EVOLUTION IN AGE-STRUCTURED POPULATIONS
Second edition
CAMBRIDGE STUDIES
IN MATHEMATICAL BIOLOGY

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TO THE MEMORY OF MY PARENTS
CONTENTS

Preface xi
Preface to the second edition xiii

1 Models of age-structured populations 1
1.1 Introduction 1
1.2 The description of age-structured populations 2
1.2.1 Populations with discrete age-classes 2
1.2.2 Stage-structured populations 11
1.2.3 Continuous-time populations 11
1.2.4 The estimation of demographic parameters 15
1.3 Time-independent and density-independent demographic parameters 21
1.3.1 The dynamics of the population process for the female population 22
1.3.2 Populations with stable age-distributions 28
1.3.3 The dynamics of the population process for more complex types of population 41
1.4 Time-dependent and density-dependent demographic parameters 44
1.4.1 Time-dependent demographic parameters 44
1.4.2 Density-dependent demographic parameters 50
1.5 Systems of several populations 58
1.5.1 Continuous-time populations 58
1.5.2 Discrete age-class populations 61

2 The genetics of populations without selection 62
2.1 Introduction 62
2.2 Approach to genetic equilibrium 63
2.2.1 An autosomal locus: the discrete age-class case 63
2.2.2 An autosomal locus: the continuous-time case 70
2.2.3 A sex-linked locus 74
2.2.4 Two autosomal loci 75
2.3 The effects of finite population size 78
2.3.1 General considerations 78
2.3.2 Effective population number with age-structure 80
## Contents

2.3.3 The probability of fixation of a gene  
2.3.4 The frequencies of consanguineous matings in an isolated population

3 Selection: construction of a model and the properties of equilibrium populations  
3.1 Introduction  
3.2 Construction of a model of selection  
3.2.1 Genotypic parameters  
3.2.2 Models of the mating process  
3.2.3 Genotypic frequencies  
3.2.4 Non-random mating with respect to age  
3.3 Populations in genetic equilibrium  
3.3.1 Dependence of genetic equilibrium on demographic stability  
3.3.2 Equilibrium fitness measures  
3.3.3 Equilibrium in a single-locus system under selection  
3.4 Biological applications of the results  
3.4.1 Factors influencing relative fitnesses  
3.4.2 The measurement of fitness in human populations

4 Selection: dynamic aspects  
4.1 Introduction  
4.2 Approximate equations with weak selection  
4.2.1 The density-independent case  
4.2.2 The density-dependent case  
4.3 Local stability analyses  
4.3.1 Gene frequencies near 0 or 1 with density independence  
4.3.2 Gene frequencies near 0 or 1 with density dependence  
4.3.3 Gene frequencies near 0 or 1 with temporally varying environments  
4.3.4 Gene frequencies near 0 or 1 with spatially varying environments  
4.3.5 The probability of survival for a non-recessive mutant gene  
4.3.6 Local stability of a polymorphic equilibrium  
4.4 The asymptotic results of selection  
4.5 Selection on quantitative characters  
4.5.1 General considerations  
4.5.2 Prediction of the response to artificial selection  
4.5.3 Natural selection on quantitative characters  
4.6 Conclusions  
4.6.1 Density-independent populations  
4.6.2 Density-dependent populations  
4.6.3 Quantitative genetics and life-history evolution  
4.6.4 Determining optimal life-histories
The aim of this book is to summarise our current knowledge of evolutionary theory as applied to age-structured populations, i.e. populations whose members are not born into distinct generations, and where fertility and survival are functions of age. Although theoretical treatments of this subject date back to the beginnings of population genetics, its growth has been relatively slow until the last decade or so, when important contributions have been made by workers in several different areas. Population geneticists have subjected many species with age-structured populations to experiments and observations, *Drosophila* and man being the two best-known examples. A clear understanding of the implications of age-structure for such processes as selection and genetic drift is therefore necessary for a proper interpretation of experimental findings. Similarly, animal and plant breeders often wish to make predictions about the effects of artificial selection regimes imposed on age-structured populations. An understanding of the effects of natural selection on life-history phenomena, such as ageing and the timing of reproductive effort, also requires a quantitative theory of the effects of selection in relation to age. For these reasons, a survey of the present sort seems timely.

My policy has been to develop the mathematical treatment in some detail, so that the reader can see for himself the logic behind the results presented. I have deliberately kept the mathematics to an elementary level, so that anybody with a knowledge of basic calculus and matrix algebra should have no difficulty in following the derivations. A knowledge of elementary genetics is assumed. More advanced mathematical techniques are essential to a full treatment of certain topics, such as the application of bifurcation theory to the study of density-dependent populations, the treatment of temporally varying survival rates and fecundities, and the use of optimal control theory in life-history evolution. I have contented myself with citations of what I hope are the major papers on these subjects.

I have made some effort to discuss the implications of the theoretical results for empirical biologists, and many examples of the application of
the theory to real data are given or quoted. I hope that the material is
arranged in such a way that readers who are not interested in the details
of the mathematical derivations will be able to identify, and make sense
of, the major conclusions. I have tried to give a broad coverage of what I
consider to be the most important results in the field. Inevitably, much
work is only referred to or treated briefly, regardless of its intrinsic
importance, and I trust my colleagues will not feel I have unduly
emphasised my own work at their expense. By no means all the problems
in this field have been solved, and I hope I have indicated clearly where
work remains to be done.

Chapter 1 is a general survey of the basic mathematical theory of the
demography and ecology of age-structured populations, and introduces
many of the concepts and notation used in later chapters. Chapter 2 deals
with such questions as the approach to Hardy–Weinberg equilibrium and
the effects of genetic drift. It also contains a brief account of the problem
of calculating the expected frequencies of consanguineous matings in a
finite population. Chapters 3 and 4 cover the theory of gene frequency
change and equilibrium under selection. Among other things, they con-
tain discussions of the effects of demographic changes on gene frequen-
cies at polymorphic loci, the appropriateness of various fitness measures,
and the effects of selection on a quantitative character. Chapter 5 is
concerned with life-history evolution in general, with particular emphasis
on the evolution of senescence and on the reproductive effort model of
optimal life-histories.

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Preface to the second edition

Fifteen years have gone by since I wrote the first edition of this book. Some notable advances have been made in the population genetics of age-structured populations, and in the theory of life-history evolution. In addition, the literature on empirical tests of life-history theory has grown enormously. In order to include an account of these developments in the new edition, while keeping the length of the book roughly constant, I have deleted some old material as well as adding new. I have rewritten Chapter 5 completely, in an attempt to do justice to recent work on life-history evolution. I have not tried to provide a comprehensive review of this large field, but have concentrated on emphasising theoretical principles and empirical tests which seem to me to be especially illuminating.

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Models of age-structured populations

1.1 Introduction

The populations of many species of plants and animals are age-structured; if such a population is investigated at a given point in time, it is found to consist of a set of individuals who were born over a range of past times, and whose fecundities and probabilities of future survival depend on their age. The human population is an obvious example. A full ecological description of the state of such a population should therefore take into account the relative numbers of individuals of different ages (the age-structure of the population), as well as its overall size. Similarly, a description of the genetic composition of the population should contain a classification of individuals by age as well as genotype. The traditional theories of ecology and population genetics have, however, usually ignored the problem of age-structure, and have assumed that populations can be treated as homogeneous with respect to age. This is clearly legitimate in dealing with organisms with discrete generations, such as annual plants, where only individuals born in the same season are capable of interbreeding. Many species of biological importance have overlapping generations, however, and their fecundities and survival rates are rarely completely independent of age, so that their age-structure should be taken into account in any realistic model of their behaviour.

There has recently been considerable interest in extending ecological and population genetics models to include age-structure as a variable. Evolutionary theorists have also become interested in the role of natural selection in moulding the way in which fecundity and survival depend on age. The purpose of this book is to give an account of the major results which have been obtained by applying ecological and demographic models of age-structured populations to population genetics and evolutionary theory. Before going on to this topic, it is desirable to develop the basic models and concepts which are needed, and this is the subject of the present chapter.

We are primarily concerned in this chapter with the construction and analysis of mathematical models which enable us to handle changes in the
size and age-structure of biological populations. This is the subject-matter of mathematical demography, a science which dates from the work of Euler (1760), although its main development was initiated by Lotka and his associates in this century (Sharpe and Lotka, 1911; Lotka, 1925). More comprehensive treatments than the one given here may be found in the books by Keyfitz (1968), Pollard (1973), Henry (1976), Webb (1985) and Caswell (1989). It is assumed throughout this chapter that populations are genetically homogeneous, or at least constant in their genetic make-up, so that individuals can legitimately be classified solely by sex and age. It is also assumed that the populations with which we are concerned are so large that stochastic effects can be ignored, and that numbers of individuals can be adequately represented by continuous variables. The concepts and notation introduced in this chapter are essential for an understanding of later chapters, where we discuss the problems associated with describing and predicting evolutionary changes in age-structured populations.

We begin with a discussion of methods for describing the state of an age-structured population at a given time, and then show how the state at one time can be related to that at previous times. This is followed by a detailed analysis of the process of change in state of a population which lives in an unchanging environment, and whose numbers are not restricted by density-dependent factors resulting from competition for limited resources, such as food. Although most species of animals or plants only rarely, and for short periods, find themselves in such an environment, the concepts which emerge from the study of this case are fundamental to an understanding of the dynamics of age-structured populations in general. We then discuss briefly the effects of temporal fluctuations in the survival and fecundity parameters of a density-independent population, and go on to consider the problem of density-dependent factors. The chapter concludes with a brief consideration of the dynamics of systems of several, non-interbreeding populations.

1.2 The description of age-structured populations

1.2.1 Populations with discrete age-classes

Probably the simplest type of biological population which can usefully be described as age-structured is one which is characteristic of many temperate-zone species of vertebrates and perennial plants. Individuals may survive over many years, but reproduction is limited to one season of the year. In a given breeding season, individuals who were born in several different, earlier, breeding seasons may be reproductively
active, so that it is impossible to assign individuals to separate generations. The composition of the population when censused at the beginning of the breeding season in a given year \( t \) can be described in terms of the numbers of males and females falling into age-classes 1, 2, 3, etc. These correspond to individuals who were born 1, 2, 3, \ldots \) years previously. The number of females aged \( x \) at the beginning of the breeding season in year \( t \) can be written as \( n(x, t) \), and the corresponding number of males as \( n^*(x, t) \). Our problem is to relate these numbers to the state of the population in the preceding year, \( t - 1 \). There are basically two methods of doing this; one is to formulate the problem in terms of matrix algebra, and the other is to use difference equations of high order. These two methods yield equivalent results, and it is solely a matter of technical convenience as to which is to be used in approaching a particular problem. We shall therefore consider both of them in this chapter. We start by developing the matrix formulation for a number of cases, and then go on to discuss the difference equation approach.

**The Leslie matrix**

The population at the start of the breeding season in year \( t \) is composed partly of individuals who were present at the corresponding period of year \( t - 1 \), and partly of individuals who were born during the breeding season of that year. The latter individuals constitute age-class 1 for year \( t \). We can write \( P(x, t) \) and \( P^*(x, t) \) for the probabilities of survival over one year for females and males, respectively, who are present in age-class \( x \) at the start of the breeding season in year \( t \). These probabilities are usually functions of age, and will be referred to in what follows as the age-specific survival probabilities. They are also written as functions of time, to indicate that in general they are expected to change with time in response to changes in factors of the physical environment, or to changes in the numbers of individuals in the population. The probability that a female aged \( x \) in year \( t \) dies during the year can be written as \( Q(x, t) = 1 - P(x, t) \); a corresponding quantity \( Q^*(x, t) = 1 - P^*(x, t) \) can obviously be defined for males. These are the age-specific probabilities of death.

Using this notation, we have, for age-classes other than the first, the relations

\[
\begin{align*}
n(x, t) &= n(x - 1, t - 1)P(x - 1, t - 1) \quad (x > 1) \quad (1.1a) \\
n^*(x, t) &= n^*(x - 1, t - 1)P^*(x - 1, t - 1) \quad (x > 1) \quad (1.1b)
\end{align*}
\]

It is more difficult to construct a realistic model of the way in which
individuals in the first age-class are generated; in general this would require a detailed specification of the probabilities of mating events between pairs of individuals of all possible ages, and the fecundities of these matings. Fortunately, it is usually reasonable to assume that the fecundity of females of a given age is independent of the age of their mates, and that there are sufficient males available to fertilise the females. Age-specific fecundities can therefore be assigned to females; these are independent of the age-composition of the population. (This assumption is standard in demographic theory and is usually referred to as the demographic dominance of females.) Let $M(x, t)$ be the expected number of fertile eggs produced in year $t$ by a female aged $x$ at the start of the breeding season. Let the proportion of females among the offspring resulting from those eggs be $a(x, t)$; this is the primary sex-ratio for offspring of mothers aged $x$ at time $t$. The expected number of female zygotes produced by a female aged $x$ in year $t$ is thus

$$m(x, t) = M(x, t)a(x, t)$$  \hspace{2cm} (1.2)

In order to compute the number of individuals in age-class 1 at the start of the breeding season in a given year, we need to specify the probability of survival of a zygote over one year. For simplicity, we assume here that this probability, which is denoted by $P(0, t)$ for a female zygote produced in year $t$, is independent of the age of the parents. The net expected contribution from a female aged $x$ in year $t$ to the population of females aged 1 in year $t + 1$ is thus

$$f(x, t) = m(x, t)P(0, t)$$  \hspace{2cm} (1.3)

This may be called the net fecundity for the female concerned.

The assumption that the survival of an individual over the first year of life is independent of parental age is probably quite often violated in practice, especially for species with parental care, where senescence may affect the ability of parents to look after their young, or affect the quality of their eggs. For example, Perrins and Moss (1974) have shown that the rate of hatching of eggs of the great tit ($Parus$ major), in English woodland, declines with maternal age. This problem may be avoided by censusing the individuals born in year $t$ at a time when they are old enough for their future survival to be unaffected by parental age. $m(x, t)$ is then the number of female zygotes weighted by the probability of survival to this stage, and $P(0, t)$ is the probability of survival from this stage to the beginning of the breeding season in year $t + 1$.

We are now in a position to relate the state of the population at two successive times. There is generally a upper limit to the age at which
reproduction is possible or, if there is no such limit, so few individuals survive beyond a certain age that they can safely be disregarded. Let the last age-class which is taken into account have the index \( d \). Similarly, there is a lower limit to the age of reproduction, \( x = b \) say \((b \geq 1)\). We obtain from the above definitions the relation

\[
n(1, t) = \sum_{x=b}^{d} n(x, t - 1)f(x, t - 1)
\]

(1.4)

A full description of the population of immature and reproductively active females at time \( t \) is provided by the \( d \)-dimensional row vector \( n(t) \), whose components are \( n(1, t), n(2, t), \ldots, n(d, t) \). (Throughout this book a row vector consisting of \( n \) elements \((x_1, x_2, \ldots, x_n)\) is written as \( x \), and the corresponding column vector as the transposed vector \( x^\top \).) Equations (1.1a) and (1.4) can be combined compactly in the matrix equation

\[
n(t) = n(t - 1)L(t - 1)
\]

(1.5)

where \( L(t - 1) \) is the \( d \times d \) matrix defined by the expression

\[
L(t - 1) =
\begin{bmatrix}
f(1, t - 1) & P(1, t - 1) & 0 & \ldots & 0 \\
f(2, t - 1) & 0 & P(2, t - 1) & \ldots & 0 \\
\vdots & \vdots & \vdots & \ddots & \vdots \\
f(d - 1, t - 1) & 0 & 0 & \ldots & P(d - 1, t - 1) \\
f(d, t - 1) & 0 & 0 & \ldots & 0
\end{bmatrix}
\]

(1.6)

All the elements of \( L \) are zero except for the off-diagonal elements \( P(1, t - 1), P(2, t - 1), \ldots, P(d - 1, t - 1) \), and some of the elements of the first column, \( f(x, t - 1) \). A matrix such as \( L \) is referred to as a population projection matrix or Leslie matrix, after P. H. Leslie, whose papers were largely influential in promoting its widespread use in ecology and demography (Leslie, 1945, 1948). This method was also suggested independently by Bernardelli (1941) and Lewis (1942). A matrix which, like the Leslie matrix, has elements which are positive or zero is called a non-negative matrix. Such matrices have special properties, which facilitate the analysis of the population process described by equation (1.5) (Seneta, 1973). These will be discussed later, in sections 1.3.1 and 1.4.1.

In most biological applications we are interested mainly in the population of pre-reproductive and reproductive individuals, so that the vector \( n(t) \) provides an adequate description of the female population.
some special purpose it is necessary to take into account individuals who are past reproducing age, it is easy to calculate their numbers using equation (1.1a). We have

\[ n(x, t) = n(d, t + d - x) \prod_{y=1}^{x-d} P(x - y, t - y) \quad (x > d) \]

**The male population**

There are several ways to proceed if one wishes to include the male population in the analysis. One method is to define an age-specific fecundity function \( M^*(x, t) \) for males, in terms of the number of young which a male aged \( x \) is expected to father in year \( t \). If the frequency of males among these offspring is \( 1 - a^*(x, t) \), then the expected number of male progeny contributed by a male aged \( x \) in year \( t \) is

\[ m^*(x, t) = M^*(x, t)[1 - a^*(x, t)] \quad (1.7) \]

The net contribution of this male to the population of males in age-class 1 in year \( t + 1 \) can be obtained by weighting this fecundity by the probability of survival of a male zygote over the year, \( P^*(0, t) \). By analogy with equation (1.3), this is

\[ f^*(x, t) = m^*(x, t)P^*(0, t) \quad (x \geq b^*) \quad (1.8) \]

where \( b^* \) is the age at first reproduction for males. If the effective upper age-limit to male fertility is \( d^* \) (which is not necessarily equal to \( d \)), the male population at time \( t \) can be described by the \( d^* \)-dimensional row vector \( n^*(t) \), with components \( n^*(1, t), n^*(2, t), \ldots, n^*(d^*, t) \). A \( d^* \times d^* \) Leslie matrix, \( L^* \), can be constructed for the male population in the same way as for the female population in equation (1.6), giving the matrix equation

\[ n^*(t) = n^*(t - 1)L^*(t - 1) \quad (1.9) \]

This approach is not very useful from the point of view of studying the population process, since, with the assumption of female demographic dominance, the basic male fecundities \( M^*(x, t) \) must depend in general on the age-composition of the female population. This can change in time, even if environmental conditions are constant, resulting in changes in the male fecundities (which are constrained by the condition that every individual has both a male and a female parent). A more useful method is to attribute all offspring to their female parents. The net contribution of male offspring in year \( t + 1 \) from a female aged \( x \) in year \( t \) is thus

\[ f^*(x, t) = M(x, t)P^*(0, t)[1 - a(x, t)] \quad (1.10) \]
and so we obtain

\[ n^*(1, t) = \sum_{x=b}^{d} n(x, t-1) f'(x, t-1) \quad (1.11) \]

\( n(t) \) and \( n^*(t) \) can be combined into a single vector of dimension \( d + d^* \), \([n(t), n^*(t)]\), which gives a complete description of the population at time \( t \). From equations (1.1), (1.5) and (1.11), we have

\[
\begin{bmatrix} n(t), n^*(t) \end{bmatrix} = \begin{bmatrix} n(t-1), n^*(t-1) \end{bmatrix} \begin{bmatrix} L(t-1) & A(t-1) \\ 0 & B(t-1) \end{bmatrix}
\]

(1.12)

where \( L \) is the Leslie matrix defined by equation (1.6), \( 0 \) is a \( d^* \times d \) matrix of zero elements; \( A \) is a \( d \times d^* \) matrix all of whose elements are zero, apart from the first column, which consists of \( f'(1, t-1), f'(2, t-1), \ldots, f'(d, t-1) \); \( B \) is a \( d^* \times d^* \) matrix whose only non-zero elements are the off-diagonal elements \( b_{x,x+1} = P^*(x, t-1), \ x = 1, 2, \ldots, d^* - 1. \)

The approach embodied in equation (1.12) was introduced by Goodman (1969) and provides the simplest reasonably realistic method of handling an age-structured population with two sexes. For accounts of more complex approaches, the reader should consult Pollard (1973, Chapter 7), Parlett (1972) and Pollak (1990).

In species where reproduction is purely asexual, or where individuals are monoecious or hermaphroditic, as in many plants, the problems of describing a two-sex population clearly do not arise. The population process is completely characterised by a single equation similar to (1.5), except that we no longer have to discount male offspring in calculating net fecundities. The \( f(x, t-1) \) are therefore equal to \( M(x, t-1) P(0, t-1) \) in this case, and the vector \( n(t) \) refers to the whole population.

**Plants with seed dormancy**

In plant demography, however, the description of the population process is considerably complicated by the fact that seeds may persist for several years in the soil before germinating. In treating hermaphrodite plants which have seed dormancy, it is thus necessary to take into account the existence of a population of seeds of various ages, as well as the population of plants themselves. This topic is discussed in some detail by Harper and White (1974), Sarukhan and Gadgil (1974) and Harper.
Models of age-structured populations

(1977). Here we merely sketch the way in which one can describe the
essentials of the population process, by extending the matrix formulation
developed above.

It is convenient to treat both the seed and plant populations as being
censused at the beginning of the flowering season in a given year,
following the time of germination. We write \( n'(1, t) \) for the number of
seeds produced in the flowering season of year \( t - 1 \) and which survived
to year \( t \) without germinating, \( n'(2, t) \) for the number produced in year
\( t - 2 \) which survived to \( t \), etc. If \( d' + 1 \) is the maximum number of years
over which a seed can survive without germinating, the seed population is
represented by the \( d' \)-dimensional row vector \( n'(t) = [n'(1, t), n'(2, t),
\ldots, n'(d', t)] \). Seeds aged more than \( d' \) years can be ignored, since they
contribute nothing to the adult population, and hence nothing to the
future reproduction of the population. It is assumed that \( d' \geq 1 \), so that
seeds can survive more than one year; otherwise, it is unnecessary to treat
the seed population as a distinct entity. Let \( P'(x, t) \), for \( x = 1, 2, \ldots, d' \), be the probability that a seed aged \( x \) at the beginning of the
season in year \( t \) does not germinate and survives for a year. We also
define \( P'(0, t) \) as the probability that a seed produced in year \( t \) survives
to year \( t + 1 \) without germinating. Similarly, \( P"(x, t) \), for \( x = 1, 2, \ldots, d' \), is the probability that a seed aged \( x \) in year \( t \) survives to year \( t + 1 \)
and then germinates. \( P"(0, t) \) is the probability that a seed produced in
year \( t \) survives to year \( t + 1 \) and germinates. The population of plants in
year \( t \) is represented by the \( d \)-dimensional vector \( n(t) = [n(1, t), n(2, t),
\ldots, n(d, t)] \); \( n(1, t) \) is the number of plants present at the beginning of
year \( t \) which result from newly germinated seeds, \( n(2, t) \) is the number of
plants which result from seeds which germinated in year \( t - 1 \), etc. We
write \( P(x, t) \), with \( x = 1, 2, \ldots, d - 1 \), for the probability that a plant
aged \( x \) in year \( t \) survives to year \( t + 1 \) (where \( x \) is one plus the number of
years since it germinated). Similarly, \( M(x, t) \) is the expected number of
seeds that a plant aged \( x \) in year \( t \) produces in that year. It is convenient
to define two net seed fecundities, for non-germinating and germinating
seeds, respectively, by the relations

\[
f'(x, t) = M(x, t)P'(0, t) \tag{1.13a}
\]

and

\[
f"(x, t) = M(x, t)P"(0, t) \tag{1.13b}
\]

Using these definitions, we obtain the following matrix equation for the
population process
where $A$ is a $d' \times d'$ matrix whose elements are all zero, except for the off-diagonals $a_{x,x+1} = P'(x,t - 1)$; $B$ is a $d' \times d$ matrix whose only non-zero elements are those of the first column, $P''(1,t - 1)$, $P''(2,t - 1)$, etc.; $C$ is a $d \times d'$ matrix whose only non-zero elements are those of the first column, $f'(1,t - 1)$, $f'(2,t - 1)$, etc.; $D$ is a $d \times d$ matrix whose only non-zero elements are the off-diagonals $d_{x,x+1} = P(x,t - 1)$, and the elements of the first column, $f''(1,t - 1)$, $f''(2,t - 1)$, etc.

### The difference equation approach

An alternative way of representing the population processes described above is by using high-order difference equations rather than matrix algebra. As will be seen in later chapters, this method often has certain advantages for population genetics problems. In this section, the procedure for the female population, corresponding to pp. 3–6 above, will be presented. The most convenient variable, with this method, for describing the state of the population in a given year $t$ is the number of female zygotes produced in the breeding season of that year, $B_f(t)$. If $t \geq d$, none of these zygotes will have been produced by individuals who were present at the start of the breeding season in some initial year, when $t$ is taken as 0. We write $l(x,t)$ for the probability that a female survives from conception in year $t - x$ to age $x$ at the beginning of the breeding season in year $t$. From the definitions presented above, on pp. 3–4, we have

$$l(x,t) = \prod_{y=1}^{x} P(x-y,t-y) \quad (x \geq 1) \quad (1.15)$$

It is useful to define $l(0,t)$ as equal to unity. These definitions imply that $l(x,t)$ is a decreasing function of $x$; if the survival probabilities are all less than unity, as is generally the case in practice, $l(x,t)$ is a strictly decreasing function of $x$. It will be referred to here as the survival function.

It is also convenient to have an expression for the net expectation of female offspring at age $x$, attributable to a female born at time $t - x$; this is given by the relation

$$k(x,t) = l(x,t)m(x,t) \quad (1.16)$$
This quantity is often referred to as the reproductive function.

Using these definitions, it is easily seen that, for $t \geq d$, we have

$$B_f(t) = \sum_{x=b}^{d} B_f(t-x)k(x, t)$$ (1.17)

For times such that $t < d$, some of the females who were present in the initial population are still reproductively active, and their contributions to $B_f(t)$ must be taken into account. An individual aged $x - t$ initially ($x \geq t + 1$) has a probability of survival to time $t$ (when it will be aged $x$) of

$$\bar{l}(x, t) = \prod_{y=1}^{t} P(x - y, t - y)$$

The total contribution of such individuals to $B_f(t)$ is thus

$$g(t) = \sum_{x=t+1}^{d} n(x - t, 0)\bar{l}(x, t)m(x, t) \quad (t < d)$$ (1.18)

For $t \geq d$, $g(t)$ can be defined as zero, so that equations (1.17) and (1.18) can be condensed into a single formula

$$B_f(t) = g(t) + \sum_{x=1}^{t} B_f(t-x)k(x, t)$$ (1.19)

Note that $k(x, t)$ is zero for $x > d$ and $x < b$.

The number of females in age-class $x$ at time $t(t \geq x)$ is easily computed from $B_f(t-x)$ by the formula

$$n(x, t) = B_f(t-x)l(x, t)$$ (1.20a)

The definition of $B_f(t)$ as the number of female zygotes is purely a matter of convenience; any other stage in the first year of life may be chosen for this purpose, provided that $m(x, t)$ is re-defined so as to discount the fraction of zygotes which die before reaching the chosen stage. There must obviously also be a corresponding modification in $l(x, t)$. Finally, we may note that, if the primary sex-ratio is constant and independent of parental age, which is a close approximation to reality in many cases, then equation (1.19) holds with $B_f(t)$ replaced by $B(t)$, the total number of male and female zygotes. The only change is that $m(x, t)$ is replaced by $M(x, t)$ in equation (1.18). If the primary sex-ratio is $a$, then the numbers of females and males aged $x$ at time $t(t \geq x)$ are given by

$$n(x, t) = aB(t-x)l(x, t)$$ (1.20b)
\[ n^*(x, t) = (1 - a)B(t - x)l^*(x, t) \] (1.20c)

where \( l^*(x, t) \) is defined by an equation similar to (1.15), replacing \( P(x - y, t - y) \) by \( P^*(x - y, t - y) \).

1.2.2 Stage-structured populations

Plant and animal ecologists often find it easier to census individuals according to some measure of physiological or developmental state such as size, rather than by age. Models which describe the state of the population according to the numbers of individuals in different categories of this kind are known as stage-structured models (Lefcovitch, 1965). Survival probabilities and fecundities are then assumed to be dependent on stage rather than age. This enables a matrix representation of the population process, analogous to equation (1.5), to be written, in which the transition matrix generally contains more positive elements than the Leslie matrix of equation (1.6). The uses of models of this kind are discussed by Metz and Diekmann (1986), Caswell (1989), and Cochran and Ellner (1992). So far, such models have had little application to population genetics and life-history evolution. This is in part due to the difficulty of providing concise descriptions of evolutionary dynamics with these models, and in part to the fact that they fail to include any direct effect of age on survival and fecundity. They will not be considered further in this book.

1.2.3 Continuous-time populations

The models described above are exact only for populations which reproduce at discrete time-intervals, usually one year in length. There are, however, many species where there are no discrete seasons for reproduction, so that reproduction and other aspects of the population process are best thought of as taking place in continuous time. Human populations are the most familiar example, and the traditional apparatus of mathematical demography (Sharpe and Lotka, 1911; Lotka, 1925) was largely developed in an effort to understand and predict the growth of human populations. We first construct an exact description of the continuous-time model for the female population, on lines similar to the difference equation approach to the discrete age-class model described above. We then discuss how the continuous-time case can be adequately approximated by a suitable discrete-time formulation.

*Exact formulation of the continuous-time model*

Time \( t \) and age \( x \) are now taken to be continuous real variables; \( x \) occupies the closed interval \([0, d]\), where \( d \) is the upper limit to the
reproductive ages. We can define a survival function $l(x, t)$ as the probability that a female survives from conception at time $t - x$ to age $x$ at time $t$, and an age-specific death-rate $\mu(x, t)$ such that

$$\frac{\partial l(x, t)}{\partial x} + \frac{\partial l(x, t)}{\partial t} = -\mu(x, t)l(x, t) \quad (1.21)$$

With the condition $l(0, t) = 1$, this gives the relation

$$l(x, t) = \exp - \int_0^x \mu(x - y, t - y) \, dy \quad (t \geq x) \quad (1.22)$$

The probability of death in a small age-interval $[x, x + dx]$ is given by $\mu(x, t) \, dx$, and is clearly analogous to the age-specific probability of death $Q(x, t)$, defined for the discrete age-class model on p. 4. Equation (1.22) corresponds to equation (1.15).

We can similarly define the fecundity of a female aged $x$ at time $t$ as her rate of production of offspring $M(x, t)$, such that the number of offspring of both sexes which she is expected to produce in the small age-interval $[x, x + dx]$ is $M(x, t) \, dx$. More accurately, the number she is expected to produce in the interval $[x, x + x']$, given that she is alive at the end of the interval, is $\int_x^{x'} M(x + y, t + y) \, dy$. If the fraction of female zygotes among these zygotes is $a(x, t)$, her fecundity with respect to female offspring is defined by

$$m(x, t) = M(x, t)a(x, t) \quad (1.23)$$

The parallels with the discrete age-class formulae on pp. 3–4 are obvious.

We can also define the reproductive function for the continuous-time model as $k(x, t) = l(x, t)m(x, t)$, by analogy with equation (1.16).

Finally, the state of the population at any time $t$ can be described by the quantity $B_f(t)$; this is the rate of production of female zygotes by the population as a whole, such that the number of female offspring entering the population over a small time-interval $[t, t + dt]$ is $B_f(t) \, dt$. This will be referred to loosely as the birth-rate of the female population, although for genetic purposes we normally regard the population as being censused at the stage of zygote formation, before selection can act to alter its genetic make-up. Knowledge of $B_f(t)$ and the survival function $l(x, t)$ enables us to calculate the number of females in any age-interval. Let $n(x, t)$ be the density function for the number of females aged $x$ at time $t$, such that the number aged between $x$ and $x + dx$ at time $t$ is $n(x, t) \, dx$. We clearly have (for $t \geq x$)

$$n(x, t) = B_f(t - x)l(x, t) \quad (1.24a)$$
The number of females aged between \( x_1 \) and \( x_2 \) at time \( t \) is given by

\[
n(x_1, x_2, t) = \int_{x_1}^{x_2} n(x, t) \, dx
\]  

(1.24b)

We are now in a position to relate \( B_j(t) \) to the past state of the population; this can be achieved by means of an argument similar to that which leads to equations (1.17), (1.18) and (1.19). For times \( t > d \), such that no females who were present at the initial time \( t = 0 \) are still reproductively active, we have

\[
B_j(t) = \int_{b}^{d} B_j(t - x) k(x, t) \, dx
\]

(1.25)

where \( b \) is the lower limit to the reproductive ages \( (b > 0) \). For \( t \leq d \), it is necessary to take into account the contributions of females who were present at time \( t = 0 \). Let \( n(x, 0) \, dx \) be the number of females aged between \( x \) and \( x + dx \) at this initial time. The contribution of the initial set of females to \( B_j(t) \) is given by

\[
g(t) = \int_{t}^{d} n(x - t, 0) \tilde{l}(x, t, t) m(x, t) \, dx
\]

(1.26)

where

\[
\tilde{l}(x, t) = \exp \left( - \int_{0}^{t} \mu(x - y, t - y) \, dy \right).
\]

Combining this with equation (1.25) gives the continuous-time analogue of equation (1.19)

\[
B_j(t) = g(t) + \int_{0}^{t} B_j(t - x) k(x, t) \, dx
\]

(1.27)

where \( g(t) = 0 \) for \( t > d \), and \( k(x, t) = 0 \) for \( x > d \) and \( x < b \). This is an example of an integral equation.

The dynamics of the population can also be represented by a partial differential equation known as the McKendrick–Von Foerster equation (McKendrick, 1926; Von Foerster, 1959). For most biological applications, the integral equation (1.27) and the corresponding discrete age-class representations of equations (1.6) and (1.19) have proved more useful.

It is reasonable to assume that \( \mu(x, t) \) and \( m(x, t) \) are continuous functions of \( x \) and \( t \). It follows from this that \( l(x, t) \) and \( k(x, t) \) are also continuous in \( x \) and \( t \). It is not necessary for \( n(x, 0) \) to be continuous in \( x \) for equation (1.27) to be valid, but it must satisfy the standard mathematical conditions for integrability. If these conditions are met, then \( g(t) \) and
Models of age-structured populations

$B_f(t)$ are both continuous functions of $t$. (Continuity of $k(x, t)$ is, of course, not necessary for the validity of equation (1.27), and in fact equation (1.19) can be regarded as a special case of equation (1.27) if the theory of Lebesgue integration is applied. In an elementary treatment such as the present, it seems preferable, however, to treat the discrete-time and continuous-time models as distinct.) Later on we shall see how equation (1.27) can be used to study the dynamics of the size- and age-composition of populations (sections 1.3.1 and 1.4.1).

We may note that age-specific death-rates, survival functions and fecundities can be defined for males in exactly the same way as for females; as in the discrete age-class model, they will be distinguished by asterisks from their female counterparts. Furthermore, if the primary sex-ratio, $a$, can be treated as constant and independent of parental age, it is easy to show that equation (1.27) holds if $B_f(t)$ is replaced by the corresponding rate of production of male and female zygotes, $B(t)$, except that $m(x, t)$ in equation (1.26) must be replaced with $M(x, t)$. The parallel with the difference equation formulation of the discrete age-class model is, therefore, complete.

Discrete age-class approximations to continuous-time models

The discrete age-class model is obviously far more convenient for computational purposes than the continuous-time model. By dividing the continua of time $t$ and age $x$ into discrete intervals of arbitrary length, it is possible to approximate the continuous-time model to any desired degree of accuracy by a discrete age-class model, simply by making the widths of the intervals sufficiently small. For example, let the age interval $[0, d]$ be divided into $m$ equal sub-intervals of width $d/m$. Consider a set of $m$ distinct values of $x$, $\{x_i\}$ ($i = 1, 2, \ldots, m$), such that each $x_i$ falls into one of the sub-intervals or on to a boundary between sub-intervals. (The positioning of the $x_i$ with respect to the boundaries is arbitrary; in ecological applications, it is frequently the case that each $x_i$ is taken as the mid-point of the $i$th sub-interval.) Equation (1.25) is then approximated by the expression

$$\tilde{B}_f(t) = \sum_{i=1}^{m} B_f(t - x_i)k(x_i, t)(d/m)$$

If time $t$ is also divided into intervals of width $d/m$, we can approximate the continuous-time process described by $B_f(t)$ by the discrete-time process described by $\tilde{B}_f(j)$, where (for $j > m$) we have
\[ \mathcal{B}_i(j) = \sum_{i=1}^{m} \mathcal{B}_{i-1}(j-i) \tilde{k}(i, j) \]

where \( \tilde{k}(i, j) = k(x_i, t_j)(d/m) \).

Similar approximations can obviously be made to equation (1.27). The parameters of this discrete-time model are obtained directly from the continuous model. The probability of survival from age-class \( i \) to \( i + 1 \) at the \( j \)th time-interval is given by \( \mathcal{P}(i, j) = l(x_{i+1}, t_{j+1})/l(x_i, t_j) \); the probability of survival from conception in the \( i - j \)th time-interval to age-class \( i \) in time-interval \( j \) is given by \( \prod_{k=1}^{i-1} \mathcal{P}(i-k, j-k) \); the fecundity of a female in age-class \( i \) in the \( j \)th time-interval is \( m(x_i, t_j)(d/m) \), etc. With these definitions, and the appropriate change in notation, we can regard the matrix or difference equation formulations of section 1.2.1 as approximations to a continuous-time model, except that the unit of time is now the time-interval of arbitrary length \( d/m \) instead of one year. In practice, it is usually found that the continuous-time process is well approximated by the discrete model even if \( m \) is not very large. For example, in human demography it is conventional to take 5-year time-intervals as the basis for computation, thus dividing up the female reproductive and pre-reproductive life-span into nine or ten age-classes. The error resulting from this approximation seems to be negligible in comparison with other sources of error and artificialities of the model (Keyfitz, 1968, Chapter 8).

1.2.4 The estimation of demographic parameters

As we have seen above, the dynamic properties of a female population with discrete age-classes are determined by age-specific survival probabilities \( P(x, t) \) (or, equivalently, by the survival function, \( l(x, t) \), and the age-specific fecundities, \( m(x, t) \)). Similar functions can be defined for a male population and, with appropriate modifications, for a continuous-time population. We may refer to these quantities collectively as the demographic parameters for the population in question. To apply our mathematical models to a real population, it is obviously necessary to possess estimates of the demographic parameters. Such estimates are usually referred to as the vital statistics or vital rates of the population. Obtaining satisfactory estimates presents considerable technical problems, whose detailed discussion is beyond the scope of this book. Interested readers should consult one of the standard works in this field, such as Harper (1977) for plants, Deevey (1947), Southwood (1966) or Caughley (1977) for animals, Caswell (1989) for plants and animals, and Keyfitz (1968) or Henry (1976) for humans. Here, we shall simply survey rapidly the basic ideas and methods involved in obtaining vital statistics,
in order to provide some feel for the empirical basis of the models described in this chapter. For simplicity, the account is worded as if demographic parameters are completely independent of time and depend only on age and sex; it should be borne in mind that a set of vital statistics collected at any one time is frequently valid only for a limited period.

**Direct methods of obtaining vital statistics**

Two basic methods of directly collecting data, the 'horizontal' and the 'vertical' methods, may be distinguished. With the horizontal method, which is commonly used in the study of laboratory populations of animals and in the demography of historical human populations, the investigator follows the fate of a sample of individuals who were born at about the same time (a cohort), and records the times at which they die and the numbers of offspring which they produce at different ages. For example, if a laboratory strain of *Drosophila* is used as material, it is possible to follow the daily egg production of individual females throughout their life under standard culture conditions, and to record for each day the number of females surviving out of the set of newly eclosed females used to found the cohort. Separate experiments have to be conducted to determine the length of time from the egg stage to eclosion from the pupa, and the probability of survival from egg to eclosion. In this way it is possible to determine a set of values, \( l(x) \), of probabilities of survival of a female from egg to day \( x \); the fecundity \( m(x) \) is expressed as one-half of the number of eggs a female alive at day \( x \) is expected to lay over one day (this assumes a sex-ratio of one-half). The table of values of \( l(x) \) is known as the life-table, and the table of values of \( m(x) \) as the fecundity schedule. Table 1.1 shows the life-tables and fecundity schedules for adult females of the species *Drosophila pseudoobscura* grown at 25 °C, both under optimal conditions and when yeast is omitted from the culture vials. In both cases, the mean time to eclosion is approximately 13 days, but the fecundity and survival of the flies are drastically reduced by starvation.

In the vertical method for obtaining vital statistics, the investigator uses census data on the same population at two successive times, e.g. one year apart. This is the standard method used in the demography of contemporary human populations, where very many data are available from national census figures. This type of data gives one the number of individuals of a given age-class that were alive at the start of the interval, and the number who died at some time during it. The number of offspring produced during the interval by individuals who belong to a given age-class at its start can similarly be estimated. There is extensive
Table 1.1. Vital statistics for adult females of Drosophila pseudobscura

<table>
<thead>
<tr>
<th>Age ((x)) (days)</th>
<th>Optimal conditions</th>
<th>Yeast starvation</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(l(x))</td>
<td>(m(x))</td>
</tr>
<tr>
<td>13</td>
<td>0.60</td>
<td>0.0</td>
</tr>
<tr>
<td>17</td>
<td>0.60</td>
<td>23.6</td>
</tr>
<tr>
<td>21</td>
<td>0.60</td>
<td>20.2</td>
</tr>
<tr>
<td>25</td>
<td>0.60</td>
<td>25.8</td>
</tr>
<tr>
<td>29</td>
<td>0.60</td>
<td>25.1</td>
</tr>
<tr>
<td>33</td>
<td>0.59</td>
<td>16.4</td>
</tr>
<tr>
<td>37</td>
<td>0.53</td>
<td>16.6</td>
</tr>
<tr>
<td>41</td>
<td>0.46</td>
<td>17.0</td>
</tr>
</tbody>
</table>

Data of Anderson and Watanabe (1980) for females of karyotype \(AR/AR\), reared at 25 °C. One hundred newly eclosed females were used to found each cohort. Measurements of larval and pupal viability showed that the probabilities of survival to time of eclosion \(l(13)\) were 0.60 and 0.46 for the optimal and starvation conditions respectively.

Indirect methods of obtaining vital statistics

Ecologists interested in natural populations often have to resort to less direct methods for obtaining vital statistics. It is rarely possible, in dealing with a population which is living under natural conditions, to follow a cohort of individuals throughout their life, and to distinguish emigration of animals out of a study area from mortality. One method which is often used to get around these difficulties is to assume that the population is stationary in size and constant in age-structure. If this is the case, it is easy to see (equation \((1.50a)\)) that the frequency of individuals in age-class \(x\) (in a discrete age-class population) is simply \(l(x)/\sum l(x)\). If new-born individuals \((x = 0)\) are included in the sample whose age-structure is measured, we can use the fact that \(l(0)\) equals 1 to obtain the values of \(l(x)\) from the frequencies of different ages in the sample. If
Models of age-structured populations

Table 1.2. Vital statistics for U.S. females (1964)

<table>
<thead>
<tr>
<th>Age-class (x) (5-year intervals)</th>
<th>P(x)</th>
<th>l(x)</th>
<th>m(x)</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>0.9779</td>
<td>1</td>
<td>—</td>
</tr>
<tr>
<td>1</td>
<td>0.9966</td>
<td>0.9746</td>
<td>0</td>
</tr>
<tr>
<td>2</td>
<td>0.9983</td>
<td>0.9730</td>
<td>0.0011</td>
</tr>
<tr>
<td>3</td>
<td>0.9979</td>
<td>0.9710</td>
<td>0.0898</td>
</tr>
<tr>
<td>4</td>
<td>0.9968</td>
<td>0.9679</td>
<td>0.3566</td>
</tr>
<tr>
<td>5</td>
<td>0.9960</td>
<td>0.9641</td>
<td>0.4868</td>
</tr>
<tr>
<td>6</td>
<td>0.9947</td>
<td>0.9590</td>
<td>0.3453</td>
</tr>
<tr>
<td>7</td>
<td>0.9923</td>
<td>0.9516</td>
<td>0.1875</td>
</tr>
<tr>
<td>8</td>
<td>0.9887</td>
<td>0.9408</td>
<td>0.0778</td>
</tr>
<tr>
<td>9</td>
<td>0.9830</td>
<td>0.9248</td>
<td>0.0178</td>
</tr>
<tr>
<td>10</td>
<td>—</td>
<td>0.9091</td>
<td>0.0010</td>
</tr>
</tbody>
</table>

These figures are taken from U.S. census data for 1964, as analysed by Keyfitz (1968, pp. 27–33). P(0) is an estimate of the probability that a female child born at some time within a 5-year time interval will live to the end of the interval; P(1) is an estimate of the probability that a female aged between 0 and 4 years will live for 5 years, P(2) estimates the probability that a female aged between 5 and 9 years will live for 5 years, etc. Similarly, m(2) estimates the number of daughters a female aged between 5 and 9 years is expected to bear over 5 years; m(3) is the similar number for a female aged between 10 and 14 years, etc. l(x) is obtained from equation (1.15).

new-born individuals cannot, for technical reasons, be sampled adequately, it is only possible to determine the values of l(x) relative to the first age-class represented in the sample. An example of this method is provided by the work of Bulmer and Perrins (1973) on the great tit, who show that adult females (females aged 1 year and older) have an age-distribution which fits a life-table with an annual mortality of 0.52, independent of age. Males have a mortality of 0.44.

A related method, which also assumes a stationary population and constant age-structure, is to determine the distribution of age at death among a sample of dead individuals. For example, examination of the annual growth rings of the horns of mammals such as sheep enables one to determine the ages at death of a collection of skulls (Deevey, 1947; Caughley, 1966, 1977). If the population is stationary, it is easily seen that the frequency of individuals who die while in age-class x is l(x) − l(x + 1), which enables one to determine the values of l(x) relative to the value for the first age-class which can be reliably aged. A refinement of this method, which removes the necessity for assuming a stationary size
for the population as a whole, is to mark individuals at birth or some convenient age, and then to attempt to recover as many marked corpses as possible. In this way, an estimate of the distribution of age at death can be obtained for a cohort. This method has been widely applied to the study of bird populations, owing to the ease of identifying individuals by ringing (Lack, 1954, 1966).

With all these indirect methods, supplementary observations must be made to determine the fecundity schedule, and also to estimate survival probabilities for ages which cannot be included in the basic sample material. Quite apart from the errors resulting from the fact that the assumptions of the methods are frequently violated in practice, considerable sampling errors are usually attached to the vital statistics obtained for natural populations, due to the difficulty of obtaining large samples. Relatively few really reliable sets of vital statistics are available for natural populations. Enough data have been obtained, however, to give some insight into the types of life-histories adopted by different species, in terms of the way in which the vital statistics depend on age. We shall briefly consider some general features of the main types of life-history below.

Types of life-history

There are two fundamentally different types of life-history, with respect to the pattern of reproduction as a function of age. These were named semelparous and iteroparous life-histories by Cole (1954). A species is said to have a semelparous life-history when reproduction is confined to a single age-class, after a more or less prolonged juvenile stage, and is followed by death. The most familiar type of semelparous life-history is that exhibited by annual species of insects, where individuals reproduce in the breeding season following the year of their birth. There are also semelparous species in which reproduction takes place several years after conception. The pink salmon (Oncorhynchus gorbuscha), for example, breeds only at 2 years of age (Aspinwall, 1974). More extreme postponement of reproduction is exhibited by such organisms as the periodic cicada (Magicicada) of the eastern U.S.A., which reproduces and dies at 13 or 17 years (Lloyd and Dybas, 1966a, b; Bulmer, 1977). Semelparous species with long postponement of reproduction are also known in plants, e.g. the century plant Agave americana. There is thus considerable diversity among semelparous species with respect to the age of reproduction, although annual species are the commonest.

There is a wide range of life-histories among iteroparous species. They may reproduce in successive discrete breeding seasons (usually annual) as
Table 1.3. Vital statistics for female grey squirrels in North Carolina

<table>
<thead>
<tr>
<th>Age-class ($x$) (years)</th>
<th>$P(x)$</th>
<th>$l(x)$</th>
<th>$m(x)$</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>0.253</td>
<td>1</td>
<td>-</td>
</tr>
<tr>
<td>1</td>
<td>0.458</td>
<td>0.253</td>
<td>1.28</td>
</tr>
<tr>
<td>2</td>
<td>0.767</td>
<td>0.116</td>
<td>2.28</td>
</tr>
<tr>
<td>3</td>
<td>0.652</td>
<td>0.089</td>
<td>2.28</td>
</tr>
<tr>
<td>4</td>
<td>0.672</td>
<td>0.058</td>
<td>2.28</td>
</tr>
<tr>
<td>5</td>
<td>0.641</td>
<td>0.039</td>
<td>2.28</td>
</tr>
<tr>
<td>6</td>
<td>0.880</td>
<td>0.025</td>
<td>2.28</td>
</tr>
<tr>
<td>7</td>
<td>-</td>
<td>0.022</td>
<td>2.28</td>
</tr>
</tbody>
</table>

Data of Barkalow, Hamilton and Soots (1970). Annual survival was estimated by following marked individuals from the nesting stage; fecundity was determined by measurements of litter size (litters from females breeding in their first year of life have been ignored). Note that the survival probabilities for later ages are based on small samples.

in many perennial plants and vertebrates; in other species, individuals produce successive litters at short intervals, without any synchronisation into breeding seasons (but usually with a suspension of reproduction in the winter, in temperate species); females may lay eggs continuously throughout their reproductive life, as in many tropical insects. There is considerable diversity even among quite closely related species with respect to such parameters as the age of first reproduction, the human species being an example of an iteroparous species which has postponed reproduction considerably, even in comparison with its closest relatives among the great apes.

The relationship of fecundity to age is very variable among iteroparous species, and several different types of life-history can be distinguished. A pattern which is characteristic of many small birds and mammals is annual breeding, with individuals reproducing for the first time in the breeding season following the year of birth. It is often found that $M(x)$ (i.e. clutch size in birds or litter size in mammals) is more or less constant at all ages, with the possible exception of the first breeding season, where it is frequently smaller than later on. Similarly, the probability of survival over one year is generally much lower for juveniles than adults, whose annual survival rate is approximately independent of age. This type of pattern is exemplified by the great tit discussed above. A complete set of vital statistics for a natural population of the grey squirrel (Sciurus carolinensis), which illustrates this type of life-history for a mammal, is shown in Table 1.3. In the larger species of birds and mammals, the age
of first reproduction is frequently more than one year, and they generally have a considerably higher survival rate. In fact, individuals may live so long that evidence for senescence may be observed, in the form of declining survival and fecundity with advancing age (Caughley, 1966). The human data of Table 1.2 illustrate an extreme type of life-history of this sort, where a peak level of fecundity is achieved relatively quickly after the start of reproduction, followed by a slow decline. Mortality increases at an accelerated rate after a minimum around the age of first reproduction. The Drosophila data shown in Table 1.1 are similar to this in general pattern. Both the human and Drosophila vital statistics are for species which under natural conditions would have much higher mortality rates, that would tend to conceal the senescent increase in mortality with age. Another type of reproductive pattern is shown by iteroparous cold-blooded vertebrates and many perennial herbs or woody plants (Harper and White, 1974), which continue to grow in size throughout reproductive life, unlike insects or female higher vertebrates. Since egg or seed number in these species is highly correlated with size, continued growth results in a positive correlation between age and fecundity. This is seen in many fishes, amphibia and reptiles (e.g. Hodder, 1963; Tinkle and Hadley, 1973).

This is by no means an exhaustive classification of life-history types, but is intended to give the reader an idea of the great diversity of possible patterns of reproduction and survival in relation to age. Some evolutionary forces which may have been important in moulding life-history patterns will be considered in Chapter 5. Intervening chapters will deal with the population genetics theory that provides a foundation for understanding life-history evolution. We now pass on to further mathematical development of the population models, using as illustrative material some of the sets of vital statistics presented in this section.

1.3 Time-independent and density-independent demographic parameters

So far we have been concerned only with the construction of a mathematical framework for describing the population process. A solution to the population process, in the sense of specifying its size and age-composition at some arbitrary time (given its initial state), can usually be obtained only when the age-specific survival probabilities are constant over time and independent of population size. This case will be considered in some detail in this section. Its importance does not lie in the fact that natural populations frequently satisfy these conditions; clearly
they do not in general. Rather, many of the basic properties of more complex types of age-structured populations can be understood by reference to this case. We shall give most attention to the female population process described by equations (1.5), (1.19) and (1.27); the male population and the case of plants with seed dormancy are also considered briefly (section 1.3.3). Since it is assumed that the demographic parameters are constant over time, the argument $t$ will be omitted from $P(x, t)$, $l(x, t)$, $m(x, t)$ etc. We consider in turn the matrix, difference equation and integral equation approaches, before developing some general results which are valid for all three methods.

1.3.1 The dynamics of the population process for the female population

The Leslie matrix solution

Consider first of all the process described by equations (1.5) and (1.6). The elements of the Leslie matrix $L$ are now independent of time; it follows from equation (1.5) that we have

$$n(t) = n(0)L^t$$  (1.28)

If $L$ has $d$ linearly independent eigenvectors, this equation can be written in terms of the spectral expansion of $L^t$

$$n(0)L^t = n(0)\sum_{i=1}^{d} \lambda_i V_i^T U_i$$  (1.29)

where $\lambda_i$ is the $i$th eigenvalue of $L$, $U_i$ is the corresponding row eigenvector, and $V_i^T$ the corresponding column eigenvector, normalised such that $U_i V_i^T = 1$. (Readers unfamiliar with the results of matrix algebra employed here should consult a standard text on the subject; alternatively, Jacquard (1974, Appendix 2) and Caswell (1989, Appendix A) provide short summaries.)

The eigenvalues $\lambda_i (i = 1, 2, \ldots, d)$ are the roots of the characteristic equation

$$|L - \lambda I| = 0$$  (1.30)

where $I$ is the $d \times d$ unit matrix.

The eigenvectors $U_i$ and $V_i^T$ are defined by the equations

$$U_i L = \lambda_i U_i$$  (1.31a)

$$L V_i^T = \lambda_i V_i^T$$  (1.31b)

If equation (1.30) is expanded, and the definitions of equations (1.3) and
(1.16) are used, we obtain the following simple form for the characteristic equation\footnote{Many authors index the age-classes in the Leslie matrix 0, 1, 2, ... instead of 1, 2, 3, ... as here. An age-class \( x \) in such notation corresponds to age-class \( x + 1 \) in the system used here, so that equation (1.32) takes a slightly different form, with an exponent of \( \lambda^{x+1} \) instead of \( \lambda^x \). Some texts incorrectly use an exponent of \( \lambda^{-1} \).}

\[
\sum_{x=b}^d \lambda^{-x} k(x) = 1
\]  

(1.32)

(This equation can be derived in a straightforward way using the difference equation method, as we shall see on pp. 25-6.)

Equation (1.32) is sometimes called the Euler equation, owing to the derivation of a special form of it by Euler (1760). It is also frequently referred to as the Lotka equation, after A. J. Lotka (1925) who derived its continuous-time analogue and applied it to demographic problems.

**The asymptotic solution**

For large \( t \), equation (1.29) is dominated by the term contributed by the leading eigenvalue, \( \lambda_1 \), say; this is the eigenvalue with greatest modulus (if \( \lambda_1 \) is complex, the term contributed by its complex conjugate must also be taken into account). In the case when the leading eigenvalue is real, equation (1.29) approaches, with increasing \( t \), the expression

\[
n(t) \sim n(0) \lambda_1^t V_1^T U_1
\]  

(1.33)

By considering the corresponding expression for \( n(t-1) \), it is seen that \( \lambda_1 \) gives the asymptotic growth-rate of the population; it is a quantity of fundamental importance in demographic theory. It can be shown from standard matrix theory (Pollard, 1973, Chapter 4) that there is a unique, real positive root of equation (1.32) which corresponds to \( \lambda_1 \), unless the age-specific net fecundities \( f(x) \) are periodic. A periodic \( f(x) \) function is defined here as one which is non-zero only for \( x \) values which are a multiple of some fixed number other than one (e.g. reproduction occurring at ages 2, 4, 6, 8, etc.). The only periodic case likely to be encountered in biological applications is the semelparous life-history discussed above (p. 19). For iteroparous species, it is safe to assume that \( \lambda_1 \) is real and positive. These results can also be derived by the following elementary method, directly from the properties of the characteristic equation (1.32), using an approach due originally to Haldane (1927a).

Let \( \lambda_j \) be a root of equation (1.32), real or complex, and write

\[
\lambda_j = e^{\phi_j + i\omega_j} = e^{\phi_j} (\cos \omega_j + i \sin \omega_j)
\]
When \( \omega_j = m \pi \) \((m = 0, 1, 2, \ldots)\), \( \lambda_j \) is real; when \( m \) is zero or even, \( \lambda_j > 0 \); when \( m \) is odd, \( \lambda_j < 0 \). Separating real and complex parts of equation (1.32) with \( \lambda = \lambda_j \), we obtain

\[
\sum_x e^{-\phi_k x} \cos(-\omega_j x)k(x) = 1
\]
\[
\sum_x e^{-\phi_k x} \sin(-\omega_j x)k(x) = 0
\]

Consider a real and positive \( \lambda_j \), for which we simply have \( \sum_x e^{-\phi_k x} k(x) = 1 \). But \( \sum_x e^{-\phi_k x} k(x) \) is easily seen to be a strictly decreasing function of \( \phi \), when \( \phi \) is an arbitrary real variable. This implies that there is only one such real and positive \( \lambda_j \). If \( f(x) \) and hence \( k(x) \) are periodic, it is always possible to choose \( \omega = 2\pi/b \) such that the values of \( \omega x \) for the reproductive age-classes are even multiples of \( \pi \), since the only non-zero \( k(x) \) values are those for which \( x = b, 2b, 3b \) etc. For such an \( \omega \), we have

\[
\cos(-\omega x) = 1 \text{ and } \sin(-\omega x) = 0 \text{ for } x = b, 2b, 3b \text{ etc.,}
\]

so that equation (1.32) is satisfied with \( \lambda = e^{\phi_i + i\omega} \). Only if \( \omega \) is itself an even multiple of \( \pi \) is this root equal to \( \lambda_j \); this occurs if and only if \( b = 1 \), which is excluded from the category of periodic \( k(x) \). This proves that in the periodic case it is always possible to find a negative real root or complex root of equation (1.32) which is equal in modulus to the positive real root \( \lambda_j \).

If \( k(x) \) is not periodic, it is not possible to choose an \( \omega \) such that \( \cos(-\omega x) = 1 \) for all reproductive ages. Suppose that there is a complex root \( \lambda_k \) such that \( \phi_k \geq \phi_j \), i.e. the modulus of \( \lambda_k \) is greater than or equal to that of \( \lambda_j \). Then, since \( |\cos(-\omega_k x)| < 1 \) for at least some values of \( x \), we have

\[
\sum_x e^{-\phi_k x} \cos(-\omega_k x)k(x) < \sum_x e^{-\phi_k x} k(x) \leq 1
\]

since \( e^{-\phi_k x} \leq e^{-\phi_j x} \) for all \( x \). Hence \( \lambda_k \) cannot be a root of equation (1.32). \( \lambda_j \) is therefore the root of largest modulus, giving the conclusion that there is a real, positive leading eigenvalue \( \lambda_1 \) in the non-periodic case.

Further consideration of the periodic case will be deferred until p. 26. We can therefore assume that \( L \) possess a leading eigenvalue \( \lambda_1 \), which is given by the positive real root of equation (1.32). For \( t \) taken sufficiently large, equation (1.33) is the solution of the population process.

Since, by equation (1.33), each age-class asymptotically grows in size at the geometric rate \( \lambda_1 \) every time-interval, it follows that the age-structure of the population, in terms of the relative numbers of individuals in the different age-classes, is constant. The population is said to have a stable
age-distribution when this state has effectively been reached. The concept of stable age-distribution plays an important role in demographic theory, and will be discussed in detail in section 1.3.2.

Before turning to the solution of the population process by the difference equation method, we may note that equation (1.29) assumes that the eigenvectors of $L$ are linearly independent. This is usually found to be the case in practical applications (Keyfitz, 1968, p. 51); if it is not so, a biologically trivial perturbation to some of the elements of $L$ will usually suffice to make it true. It is useful to note that, even if this condition is violated, there is always a positive, real leading eigenvalue in the non-periodic case, and that the asymptotic result of equation (1.33) is valid.

**The difference equation solution**

In this section, we show how the population process can be solved by means of the difference equation formulation introduced on pp. 9–11. If there is a time-independent reproductive function $k(x)$, equation (1.19) becomes simply

$$B_f(t) = g(t) + \sum_{x=1}^{t} B_f(t - x)k(x) \tag{1.34}$$

This is an example of a renewal equation, a class of linear difference equations whose properties have been much studied by mathematicians interested in probability theory and related areas. In the present case, when $k(x) = 0$ for $x > d$ and $g(t) = 0$ for $t \geq d$, equation (1.34) can be approached by the method of generating functions, which is outlined in Appendix 1. Taking generating functions on both sides of this equation, and defining $k(0)$ as zero, we obtain

$$B_f(s) = \bar{g}(s) + B_f(s)\bar{k}(s) \tag{1.35}$$

so that

$$B_f(s) = \frac{\bar{g}(s)}{1 - \bar{k}(s)} \tag{1.36}$$

This equation is of the same form as equation (A1.3) of Appendix 1; it follows that, providing that the zeros of $1 - \bar{k}(s)$ are all distinct, we have

$$B_f(t) = \sum_{j=1}^{d} C_j s_j^{-(t+1)} \tag{1.37}$$

where the $C_j$ are given by equation (A1.6) of Appendix 1. Asymptotically we have

$$B_f(t) \sim C_1 s_1^{-(t+1)} \tag{1.38}$$
where \( s_1 \) is the zero of \( 1 - k(s) \) with smallest modulus. But \( \bar{k}(s) = \sum s^x k(x) \), so that \( s^{-1} \) can be identified with \( \lambda \) of equation (1.32). \( \lambda_1 \) is therefore equal to \( s_1^{-1} \). Evaluating the differential coefficient of \( 1 - \bar{k}(s) \) with respect to \( s \) at \( s_1 \), and using equation (A1.6) of Appendix 1 together with equation (1.38), gives the asymptotic result

\[
B_f(t) \sim \frac{d}{x=0} \sum_{x=1}^{d} \frac{x\lambda_1^{-x} k(x)}{\sum x\lambda_1^{-x} k(x)}
\] (1.39)

As with the matrix approach, this asymptotic solution is valid even if \( 1 - \bar{k}(s) \) has repeated zeros, provided that \( k(x) \) is non-periodic (see Appendix 1). The numbers of individuals in any age-class at a given time \( t \) can be obtained from equation (1.39) by using equation (1.20). Equations (1.33) and (1.39) must therefore correspond; the nature of this correspondence will be discussed below (pp. 36–7).

It is useful to note that equation (1.32), from which \( \lambda_1 \) is obtained, can be derived directly from equation (1.34). If the population has attained its stable age-distribution and is growing at rate \( \lambda_1 \), we have \( B_f(t) = C\lambda_1^t \), where \( C \) is a constant of proportionality. Substituting from this into equation (1.34) for \( t = d \), we find

\[
C\lambda_1^t = C\sum_x \lambda_1^{(t-x)} k(x)
\]

so that, cancelling \( C\lambda_1^t \) from both sides, we see that \( \lambda_1 \) must satisfy equation (1.32).

The difference equation approach to the population process has the slight advantage over the matrix method that it can be applied to situations in which reproduction continues indefinitely, without a finite upper limit \( d \) to the fertile age-classes. Techniques which are more advanced than those employed here can be used to show that, provided \( k(x) \) is non-periodic and \( \sum_{i=0}^{\infty} g(t) \) is finite, equation (1.34) has the asymptotic solution (1.39). The interested reader should consult Feller (1968, Chapter 13) for a full discussion of this result, the renewal theorem.

The periodic case

As mentioned earlier, the only periodic case of much biological interest is the semelparous form of life history. Most semelparous species effectively consist of a number of non-interbreeding populations (phases) represented in different years, the individuals born in year \( t - b \) becoming the parents of individuals born in year \( t \). The number of female births in year \( t \) is therefore connected with those in year \( t - b \) by the relation...
Parameters independent of time and density

For \( t \geq b \), the number of births corresponding to a given phase of the population increases at a geometric rate \( k(b) \), the numbers for \( t < b \) being given by the initial conditions. In certain cases, of which the most famous is the 13- or 17-year cicada, *Magicicada*, (Lloyd and Dybas, 1966a, b; Bulmer, 1977), reproduction has been synchronised so that a breeding population appears only every \( b \) years (\( b = 13 \) or 17 for *Magicicada*). Possible causes of this synchronisation are discussed by Lloyd and Dybas (1966a, b), Hoppensteadt and Keller (1976) and Bulmer (1977).

**The continuous-time model**

The continuous-time model with a time-independent reproductive function yields the continuous version of the renewal equation as a special case of equation (1.27)

\[
B_f(t) = g(t) + \int_0^t B_f(t - x)k(x) \, dx
\]

(1.41)

It is easily seen that (for \( t > d \), so that \( g(t) = 0 \)) this equation is satisfied by substituting \( B_f(t) = C e^{zt} \), where \( z \) satisfies the characteristic equation analogous to equation (1.32)

\[
\int_b^d e^{-zx}k(x) \, dx = 1
\]

(1.42)

Using the argument applied above to equation (1.32), it can be shown that equation (1.42) has a single real root, \( z_1 \) say, which exceeds the real parts of the complex roots; it is sufficient for this to be true that \( k(x) \) be continuous, non-negative, and take some positive values between \( b \) and \( d \). This condition is normally satisfied in biological applications. By analogy with the results for the difference equation approach to the discrete-time case, one might expect equation (1.41) to have a general solution of the form

\[
B_f(t) = \sum_{i=1}^{\infty} C_i e^{z_i t}
\]

(1.43)

where the \( z_i \) are the roots of equation (1.42), assumed to be distinct, and the \( C_i \) are constants given by the initial conditions. This is the form of the solution proposed by Lotka and his co-workers (Lotka, 1925). If such a solution in fact exists, then (using the fact that the real root \( z_1 \) is larger than the real parts of all the complex roots) it follows that equation (1.43) becomes asymptotically

\[
B_f(t) \sim C_1 e^{z_1 t}
\]

(1.44a)
so that the logarithm of $B_f(t)$ eventually grows at constant rate $z_1$. It follows from equations (1.24) that the same is true for the number of individuals in any age-group, so that the population asymptotically achieves a state of stable age-distribution, just as in the discrete age-class case.

There are some mathematical problems in rigorously justifying the solution represented by equation (1.43). These are discussed by Feller (1941) and Lopez (1961). It seems in practice to provide an adequate answer in most cases of biological interest and (according to the results of Lopez) is valid if $k(x)$ is continuous and positive for values of $x$ between $b$ and $d$, assuming that all the zeros of equation (1.42) are distinct. The asymptotic expression (1.44a) is known, in fact, to be valid under very wide mathematical conditions, which subsume those which are sufficient for equation (1.42) to have a single real root, larger than the real parts of the complex roots (Feller, 1966, Chapter 11). The constant $C_1$ has the value

$$C_1 = \int_0^d e^{-z_1 x} g(x) \, dx \int_b^d x e^{-z_1 x} k(x) \, dx$$

(1.44b)

which shows the analogy between equations (1.39) and (1.44a).

1.3.2 Populations with stable age-distributions

As we have seen, both discrete age-class and continuous-time populations converge with increasing time to a state in which the age-structure and rate of population growth are constant. A population which has attained this state of stable age-distribution tends to return to it if it suffers a perturbation. The age-structure and growth-rate are independent of the initial state of the population, and are determined only by the values of the demographic parameters. This is the so-called strong ergodicity property of age-structured populations. We now discuss in some detail the properties of populations in stable age-distribution.

The intrinsic rate of increase, generation time and expectation of life

As we have seen, a discrete age-class population in stable age-distribution grows at the geometric rate $\lambda_1$, given as the positive real root of equation (1.32); a continuous-time population grows at the logarithmic rate $z_1$, given by the analogous equation (1.42). In order to unify the discrete-time and continuous-time cases, it is often convenient to work with the natural logarithm of $\lambda_1$ rather than $\lambda_1$. This is designated by the symbol $r$, and can be identified with $z_1$ of the continuous-time case. It has been called the intrinsic rate of natural increase, (Lotka, 1925) or the
Malthusian parameter (Fisher, 1930). Here, \( r \) will usually be referred to simply as the *intrinsic rate of increase*; it measures the asymptotic logarithmic rate of growth in both the discrete-time and continuous-time models. In the former case, \( r \) can be determined numerically (for example, by Newton–Raphson iteration), using the fact that, by equation (1.32), we have

\[
\sum_{x=1}^{d} e^{-rx}k(x) = 1
\]

(1.45)

This equation is analogous to equation (1.42) for the continuous-time model, which can similarly be solved by a combination of numerical integration and Newton–Raphson iteration. In most cases, equation (1.45) is too complex for \( r \) to be evaluated explicitly. If \( r \) is sufficiently small that terms of order \( r^2 \) can be neglected, an approximate expression for \( r \) can be obtained by noting that a Taylor expansion of equation (1.45) gives

\[
1 = \sum e^{-rx}k(x) = \sum k(x) - r\sum xk(x) + \cdots
\]

so that

\[
r \approx \frac{\sum k(x) - 1}{\sum xk(x)}
\]

(1.46)

The quantity \( \sum k(x) \) in this expression is the net expectation of female offspring to a female zygote, and is called the *net reproduction rate*. It will be denoted by the symbol \( R \). It follows from the properties of equation (1.45) that \( R \) must exceed unity for \( r \) to be positive. If \( R = 1 \), the stable age-distribution corresponds to a population which is stationary in size \( (r = 0) \); if \( R < 1 \), then the stable age-distribution corresponds to a decreasing population \( (r < 0) \).

The quantity \( \sum xk(x) \) in the denominator of equation (1.46) is the mean age at reproduction for a female in a stationary population, and may be used as one measure of 'generation time'. The above estimate of \( r \) is clearly unreliable if \( r\sum xk(x) \) is more than a few per cent i.e. if the per 'generation' rate of population increase is high. For slowly growing populations, it provides an adequate approximation. For example, with the grey squirrel data quoted in section 1.2.4, we find that \( R = 1.119 \) and \( \sum xk(x) = 3.332 \), giving an approximate value for \( r \) of 0.036. The value obtained by Newton–Raphson iteration of equation (1.45) is 0.041, in reasonably good agreement.
Since the concept of generation time is a rather arbitrary one in the context of age-structured populations, several alternative measures have been proposed. When $r$ is high, a more appropriate measure than $\sum x k(x)$ is the quantity

$$T_0 = \frac{\sum x k(x)}{\sum k(x)}$$  \hspace{1cm} (1.47a)

This is the mean age at reproduction of a cohort of females, and is independent of any assumptions about the rate of population growth. Another definition of generation time is the value $T_1$ which satisfies the equation

$$R = e^{r T_1}$$

i.e.

$$T_1 = \frac{(\ln R)}{r}$$  \hspace{1cm} (1.47b)

Finally, another definition of generation time is the mean age of the mothers of a set of new-born individuals in a population with a stable age-distribution. Using equations (1.39) and (1.20) it is easily seen that this mean age is given by the expression

$$T = \sum x e^{-r x} k(x)$$  \hspace{1cm} (1.47c)

which appears in the denominator of equation (1.39). $T$ is in fact the modulus of the derivative of equation (1.45) with respect to $r$. As will be seen shortly, $T$ arises naturally in a number of contexts, and is therefore to be preferred to other measures of generation time. Provided $r$ is not too big, all these measures are approximately equal. Using the grey squirrel data, for example, we find $T_0 = 2.794$, $T_1 = 2.273$, and $T = 2.676$.

Generation time should not be confused with expectation of life, $E$, which is the expected value of the time of death of a new-born individual. If an individual who dies between ages $x - 1$ and $x$ is classified as dying in the $x$th interval of life, and noting that $l(0)$ is defined as equal to one, we have

$$E = \sum_{x=1}^{d} x[l(x-1) - l(x)] = \sum_{x=0}^{d} l(x)$$  \hspace{1cm} (1.48)

Since this measure takes no account of reproduction, it may differ considerably in value from the generation times. For example, in the grey squirrel case $E = 1.602$. In species with high juvenile losses, $E$ very
largely reflects the fact that most individuals die in the first age-class. This measure therefore gives a biased picture of the life-span. The bias can be corrected to some extent by defining a conditional expectation of life for individuals who have survived to age \( x \), \( E_x \), by the relation

\[
E_x = \sum_{y=x+1}^{d} (y - x)[l(y - 1) - l(y)]/l(x) \quad (x = 1, 2, \ldots)
\]

\[
= \sum_{y=x}^{d} l(y)/l(x) \quad (1.49)
\]

In the squirrel case, for example, we find \( E_1 = 2.379 \), showing a considerable increase in expectation of life over \( E \).

All these quantities can be defined for the continuous-time model simply by replacing summation over age-classes by integration over age.

*The form of the stable age-distribution*

The form of the stable age-distribution of a discrete age-class population is defined by the frequencies of individuals falling into the different age-classes. Using equations (1.39) and (1.20), it is easy to see that the frequency of individuals in age-class \( x \), \( \phi(x) \), is given by the expression

\[
\phi(x) = e^{-rx} l(x)/\sum_{x=1}^{d} e^{-rx} l(x) \quad (1.50a)
\]

Using equation (1.33) and the definition of \( U_1 \), the row eigenvector of \( L \) associated with the leading eigenvalue \( \lambda_1 \) (equation (1.31a)), it is evident that the components of \( U_1, U_1(1), U_1(2), \ldots, U_1(d) \), are proportional to \( \phi(1), \phi(2), \ldots, \phi(d) \).

With the continuous-time model, we can similarly define a probability density function \( \phi(x) \), which gives the frequency of individuals between ages \( x \) and \( x + dx \) as \( \phi(x) \, dx \). This has the form

\[
\phi(x) = e^{-rx} l(x)/\int_{0}^{d} e^{-rx} l(x) \, dx \quad (1.50b)
\]

The shape of the stable age-distribution is, from equations (1.50), obviously determined both by the intrinsic rate of increase and by the shape of the \( l(x) \) function. High fecundity, resulting in high values of \( r \), leads to a high weight being placed on early ages, so that the age-structure of the population is dominated by younger individuals. Similarly, if mortality rates are high throughout life, \( l(x) \) will fall off rapidly with age, giving a similar effect. The form of the stable age-distribution is
determined solely by the form of the \( l(x) \) function only in stationary populations, where \( r = 0 \). Provided \( r \geq 0 \), \( \phi(x) \) is always a decreasing function of \( x \).

It is interesting to note that there is no difference in the form of the stable age-distribution between two populations whose demographic parameters differ only in a component of survival which is the same for each age. Thus, if one population has the set of age-specific survival probabilities \( P(x)(x = 1, 2, \ldots) \) and the other the set \( P'(x) = KP(x) \) where \( K \) is independent of age, we have \( l'(x) = K^x P(x) \). Substituting from this into equation (1.32), it is easily seen that the value of \( \lambda_1 \) for the second population is equal to that for the first multiplied by \( K \). Using equation (1.50a), it is seen that \( \phi(x) \) is the same for both populations. This type of difference in the life-table is what one would expect if the populations differed with respect to some 'accidental' source of mortality which is indiscriminate with respect to age. Coale (1957, 1972) has shown that many human populations have changed demographically over time largely in respect to mortality factors of this sort; they have therefore remained close to a single stable age-distribution.

Changes in fecundity have quite different effects on the age-distribution; since they alter the value of \( r \), they must change \( \phi(x) \) unless compensated for by changes in mortality. This suggests that changes in age-specific survival probabilities have a much smaller effect on the age-structure of a population than do changes in fecundity, unless they are highly selective with respect to age (e.g. if juvenile survival alone is changed).

\textit{Per capita birth-rates and death-rates}

The \textit{per capita birth-rates} and \textit{death-rates} for a discrete age-class population are defined as the total numbers of new births and deaths, respectively, occurring over a single time-interval, divided by the total number of individuals alive in the time-interval in question. A population with a stable age-distribution must clearly have constant values for these quantities. If the new-born individuals are counted at the beginning of the time-interval \textit{subsequent} to that in which they were born, we obtain, using equation (1.50a), the following expression for the per capita birth-rate, \( B \), of a population in stable age-distribution

\[
B = l(1) \sum_{x=b}^{d} e^{-rx}k(x)/\sum_{x=1}^{d} e^{-rx}l(x)
\]

\[
= l(1)/\sum_{x=1}^{d} e^{-rx}l(x)
\]

(1.51a)
Similarly, the per capita death-rate for the stable population is given by

$$D = \sum_{x=1}^{d} \frac{e^{-rx}[l(x) - l(x + 1)]}{\sum_{x=1}^{d} e^{-rx}l(x)}$$  \hspace{1cm} (1.51b)$$

The geometric rate of population growth for the stable population is related to \(B - D\); since each age-class increases in number at the same geometric rate \(\lambda_1\), the total number of individuals in the population at time \(t\), \(n(t)\) is related to the number at time \(t - 1\) by the expression

$$n(t) - n(t - 1) = (\lambda_1 - 1)n(t - 1)$$

But from the definitions of \(B\) and \(D\), we must also have

$$n(t) - n(t - 1) = (B - D)n(t - 1)$$

These equations show that \(B - D\) is equal to \(\lambda_1 - 1\); this can be shown in a different way by noting that equation (1.51b) gives us

$$D \sum_{x=1}^{d} \lambda_1^{-x}l(x) = \sum_{x=1}^{d} \lambda_1^{-x}l(x) - \lambda_1 \sum_{x=1}^{d} \lambda_1^{-x}l(x) + l(1)$$

which can be combined with equation (1.51a) to yield \(B - D = \lambda_1 - 1\).

Similar quantities can be defined for continuous-time populations; by analogy with equations (1.51) we have

$$B = \int_{0}^{d} e^{-rx}k(x) \, dx / \int_{0}^{d} e^{-rx}l(x) \, dx$$  \hspace{1cm} (1.52a)$$

$$D = \int_{0}^{d} e^{-rx}\mu(x)l(x) \, dx / \int_{0}^{d} e^{-rx}l(x) \, dx$$  \hspace{1cm} (1.52b)$$

Integrating equation (1.52b) by parts and combining it with equation (1.52a) verifies that \(r = B - D\), as one would expect from the definition of \(r\) as the logarithmic rate of increase.

A population which is not in stable age-distribution will not in general have constant per capita birth-rates and death-rates. It is a well-known fact of demography that direct estimates for \(B\) and \(D\) for a population which is in the process of approaching a stable age-distribution may be poor guides to their final values, owing to changes in age-structure which alter the weights attached to ages with different survival rates and fecundities (see Figures 1.1 and 1.2).

**Reproductive value**

Fisher (1930) introduced the concept of *reproductive value* as a measure of the relative extents to which individuals of different ages, in a
population stable age-distribution, contribute to the ancestry of the future population. In a discrete age-class population, reproductive value is closely connected with the column eigenvector $V_1^T$ defined by equation (1.31b), as was first pointed out by Leslie (1948). This can be seen as follows. From equations (1.5), (1.6) and (1.31b), we obtain the following relation, which enables us to determine relative values of the components $V_1(1), V_1(2), \ldots, V_1(d)$ of $V_1^T$.
Parameters independent of time and density

(a)

(b)

Figure 1.2. The approach to stable age-distribution for a population with the demographic parameters of Table 1.3 (N. Carolina female squirrels). 
(a) and (b) are as for Figure 1.1, except that time-intervals in this case are 1 year in length.
\( \lambda V_1(x) = f(x)V_1(1) + P(x)V_1(x + 1) \quad (x = 1, 2, \ldots, d) \)  

(1.53)

(Note that \( V_1(d + 1) \) must be defined as zero.)

It is easily verified that equation (1.53) is satisfied if we write 
\[ V_1(x) = Cv(x), \]
where \( C \) is a constant of proportionality and \( v(x) \) is the reproductive value of age \( x \), defined by equation (1.54a):  

\[ v(x) = \frac{e^{rx}}{l(x)} \sum_{y=x}^{d} e^{-ry}k(y) \quad (x = 1, 2, \ldots, d) \]  

(1.54a)

Note that from equations (1.32) and (1.15) we have \( y(l) = e^{r/P(0)} \) (1.54b)

This is the reproductive value of an individual who has survived for one time-interval after conception. We can use equation (1.54a) to define the reproductive value of a newly produced female zygote (noting that \( l(0) = 1 \)) as

\[ v(0) = 1 \]  

(1.54c)

The reproductive value of a female aged \( x \) on this scale can be thought of as measuring her expected future contribution of zygotes to the population, counting both the offspring contributed at age \( x \) and those she is expected to contribute if she survives to later ages. Since the population increases in number by a factor of \( e^r \) every time-interval, the value of her contribution at a future age \( x + y \) is discounted by \( e^{-ry} \). Reproductive value is thus measured relative to a value of unity for a newly-produced female zygote.

Since only the relative values of the components of \( V_1^T \) are determined by equation (1.31b), the above choice of scale for reproductive value is arbitrary, but turns out to be convenient for genetic purposes (see Chapters 3 and 4). An alternative scale of measurement, which arises naturally in a demographic context, has been proposed by Goodman (1967). Consider the asymptotic expression for the number of individuals

\[ v(x) = \left[ e^{rx} \sum_{y=x}^{d-1} e^{-ry}k(y) \right]/l(x) \]

where age-class \( x \) is equivalent to age-class \( x + 1 \) in equation (1.54a). The reproductive value for a given age in this system is equal to \( e^{-r}P(0) \) times its value in the system used in the text.

\[ v(x) = \left[ e^{rx} \sum_{y=-x}^{d-1} e^{-ry}k(y) \right]/l(x) \]  

where \( r \) is a constant of proportionality and \( v(x) \) is the reproductive value of age \( x \), defined by equation (1.54a):  

\[ v(x) = \frac{e^{rx}}{l(x)} \sum_{y=x}^{d} e^{-ry}k(y) \quad (x = 1, 2, \ldots, d) \]  

(1.54a)

Note that from equations (1.32) and (1.15) we have \( y(l) = e^{r/P(0)} \) (1.54b)

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\[ v(0) = 1 \]  

(1.54c)

The reproductive value of a female aged \( x \) on this scale can be thought of as measuring her expected future contribution of zygotes to the population, counting both the offspring contributed at age \( x \) and those she is expected to contribute if she survives to later ages. Since the population increases in number by a factor of \( e^r \) every time-interval, the value of her contribution at a future age \( x + y \) is discounted by \( e^{-ry} \). Reproductive value is thus measured relative to a value of unity for a newly-produced female zygote.

Since only the relative values of the components of \( V_1^T \) are determined by equation (1.31b), the above choice of scale for reproductive value is arbitrary, but turns out to be convenient for genetic purposes (see Chapters 3 and 4). An alternative scale of measurement, which arises naturally in a demographic context, has been proposed by Goodman (1967). Consider the asymptotic expression for the number of individuals

\[ v(x) = \left[ e^{rx} \sum_{y=x}^{d-1} e^{-ry}k(y) \right]/l(x) \]

where \( r \) is a constant of proportionality and \( v(x) \) is the reproductive value of age \( x \), defined by equation (1.54a):  

\[ v(x) = \frac{e^{rx}}{l(x)} \sum_{y=x}^{d} e^{-ry}k(y) \quad (x = 1, 2, \ldots, d) \]  

(1.54a)

Note that from equations (1.32) and (1.15) we have

\[ v(1) = e^r/P(0) \]  

(1.54b)

This is the reproductive value of an individual who has survived for one time-interval after conception. We can use equation (1.54a) to define the reproductive value of a newly produced female zygote (noting that \( l(0) = 1 \)) as

\[ v(0) = 1 \]  

(1.54c)
in age-class \( y \) at time \( t \), given by equation (1.33)

\[
n(y, t) \sim e^{\alpha t} U_1(y) \sum_{x=1}^{d} n(x, 0) V_1(x) \quad (1.55a)
\]

where \( U_1(y) \) and \( V_1(x) \) are \( y \)th and \( x \)th components of the vectors \( U_1 \) and \( V_1 \), when normalised such that \( U_1 V_1^T = 1 \) (p. 22). We know from p. 31 that \( U_1(y) \) is proportional to \( e^{-\gamma y} l(y) \); for simplicity, let the constant of proportionality be unity. We know from the above that

\[
V_1(x) = C v(x),
\]

where \( C \) is given by the normalisation requirement. Using this, together with equations (1.54), gives

\[
1 = C \sum_{x=1}^{d} e^{-\gamma x} l(x) v(x) = C \sum_{x=1}^{d} \sum_{y=x}^{d} e^{-\gamma y} k(y)
\]

i.e.

\[
1 = C \sum_{x=b}^{d} x e^{-\gamma x} k(x)
\]

We therefore have

\[
V_1(x) = v(x)/T \quad (1.56)
\]

where \( T \) is the generation time defined by equation (1.47c). Equation (1.55a) can thus be rewritten as

\[
n(y, t) \sim e^{t(l-y)} l(y) \left( \sum_{x=1}^{d} n(x, 0) v(x)/T \right) \quad (1.55b)
\]

This may be compared with the difference equation solution of equation (1.39). It is easy to show from the definition of \( g \) in equation (1.18) that we have

\[
\sum_{x=1}^{d} n(x, 0) v(x) = \sum_{x=0}^{d-1} e^{-\gamma x} g(x)
\]

Furthermore, from equation (1.20) we have \( n(y, t) = B_f(t - y) l(y) \). Substituting these expressions into equation (1.39), we recover equation (1.55b), thus unifying the matrix and difference equation solutions.

The normalisation of reproductive value by the reciprocal of the generation time in equation (1.56) gives the quantity which Goodman (1967) has called the eventual reproductive value of age \( x \). It is a measure of the extent to which individuals aged \( x \) at time zero contribute to the ancestry of the population at some distant time \( t \), when the population has approached stable age-distribution. It is thus conceptually somewhat
Table 1.4. Reproductive value in two populations

<table>
<thead>
<tr>
<th>Age-class ($x$)</th>
<th>U.S. humans (1964) $v(x)$</th>
<th>Grey squirrels $v(x)$</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>1</td>
<td>1</td>
</tr>
<tr>
<td>1</td>
<td>1.106</td>
<td>4.119</td>
</tr>
<tr>
<td>2</td>
<td>1.201</td>
<td>6.454</td>
</tr>
<tr>
<td>3</td>
<td>1.300</td>
<td>5.670</td>
</tr>
<tr>
<td>4</td>
<td>1.312</td>
<td>5.420</td>
</tr>
<tr>
<td>5</td>
<td>1.037</td>
<td>4.866</td>
</tr>
<tr>
<td>6</td>
<td>0.586</td>
<td>4.205</td>
</tr>
<tr>
<td>7</td>
<td>0.184</td>
<td>2.280</td>
</tr>
<tr>
<td>8</td>
<td>0.093</td>
<td>—</td>
</tr>
<tr>
<td>9</td>
<td>0.018</td>
<td>—</td>
</tr>
<tr>
<td>10</td>
<td>0.001</td>
<td>—</td>
</tr>
</tbody>
</table>

Reproductive value as defined by equations (1.54) or (1.56) does not have a simple monotonic relationship with age. If $r$ is non-negative it follows from the fact that $l(x)$ is a decreasing function of $x$ that reproductive value for the juvenile age-classes is an increasing function of age. But, for the reproductive age-classes, it is evident that $\sum_{y=x}^{d} e^{-r}k(y)$ decreases with increasing $x$, so that reproductive value may decrease with age for at least some reproductive ages. With sufficiently high $r$ and no upper limit to the reproductive age-classes ($d = \infty$), examples can, however, be constructed in which reproductive value increases indefinitely with age (Hamilton, 1966). Table 1.4 shows the relationship with age of reproductive value for two of the sets of vital statistics discussed earlier in this chapter. The differences between the human and squirrel cases shown here can be understood qualitatively in terms of the differences in the ways in which their vital statistics depend on age (see Tables 1.2 and 1.3). The human population is subject to very low mortality, so that there is only a slight increase in reproductive value during pre-reproductive and early reproductive life, compared with the squirrel population. On the other hand, the fecundity of human females falls off rapidly from age-class 5 onwards, whereas squirrel fecundity is constant. Reproductive value therefore tends to decrease more sharply with advancing age for older human females than for squirrels. The significance of the behaviour of reproductive value as a function of age, in connection with evolutionary problems, will be discussed in Chapter 5.
Effects of a perturbation of the components of a Leslie matrix on the population growth-rate

The age-structure and reproductive value vectors also play an important role in the theory of the effects of perturbations in the components of the Leslie matrix on the asymptotic population growth-rate, as shown by Caswell (1978), using standard matrix perturbation theory. This represents a generalisation of earlier work by Hamilton (1966), Demetrius (1969) and Goodman (1971). The main results can be derived as follows. Let the Leslie matrix \( L \) be perturbed by the addition of a matrix of perturbations \( \delta L \), corresponding to small changes in the survival and fecundity components of \( L \). There are resultant perturbations \( \delta \lambda_1 \) and \( \delta U_1 \) in the leading eigenvalue and corresponding row eigenvector, respectively. Using the properties of eigenvalues and eigenvectors, and neglecting second-order terms in the perturbations, it is easily seen that

\[
(U_1 + \delta U_1)L + U_1\delta L \equiv \lambda_1(U_1 + \delta U_1) + \delta \lambda_1 U_1
\]

i.e.

\[
\delta U_1 L + U_1 \delta L \equiv \lambda_1 \delta U_1 + \delta \lambda_1 U_1
\]  

(1.57a)

Postmultiplying by the column eigenvector \( V_1^T \), we obtain

\[
\delta U_1 L V_1^T + U_1 \delta L V_1^T \equiv \lambda_1 \delta U_1 V_1^T + \delta \lambda_1 U_1 V_1^T
\]

Using equation (1.31b), the first terms on the left-hand and right-hand sides of this expression can be seen to cancel, yielding

\[
\delta \lambda_1 \equiv \frac{U_1 \delta L V_1^T}{(U_1 V_1^T)}
\]

(1.58a)

If only one element, \( L_{ij} \), of \( L \) is perturbed, taking the limit of equation (1.58a) as \( \delta L_{ij} \) tends to zero yields the partial derivative

\[
\frac{\partial \lambda_1}{\partial L_{ij}} = \frac{U_1(i)V_1(j)}{(U_1 V_1^T)}
\]

(1.58b)

Equations (1.58) enable one to determine the effect of small changes in the demographic parameters on the asymptotic population growth-rate or intrinsic rate of increase. This is of great utility in modelling the evolution of life-histories, as will be discussed in Chapter 5.

The rate of change of total reproductive value of a population

One final property of reproductive value is of theoretical interest. We can define the total reproductive value, \( V(t) \), of an arbitrary population (which is not necessarily in stable age-distribution) at time \( t \) by the
relation

\[ V(t) = n(t)V_1^T \]  

(1.59)

Using equation (1.5), it can be seen that this is related to the population at time \( t - 1 \) by the expression

\[ V(t) = n(t - 1)LV_1^T \]

and using equation (1.31b), this gives

\[ V(t) = \lambda_1 n(t - 1)V_1 = \lambda_1 V(t - 1) \]

(1.60)

In other words, the total reproductive value of a population increases at a geometric rate \( \lambda_1 = e^r \), regardless of whether or not it is in stable age-distribution.

The concept of reproductive value can equally well be defined for a continuous-time population, and in fact this is how Fisher (1930) originally developed it. By analogy with equation (1.54a) we can write

\[ v(x) = \frac{e^{rx}}{l(x)} \int_x^d e^{-r y} k(y) dy \]

(1.61)

Using the continuous-time solution to the population process given by equations (1.44), we can obtain the analogue of equation (1.55b)

\[ n(y, t) = e^{r(t-y)}l(y) \times \int_0^d e^{-rx} n(x, 0)v(x) dx \int_b^d x e^{-rx} k(x) dx \]

(1.62)

Similarly, if we define the total reproductive value of a population as

\[ V(t) = \int_0^d n(x, t)v(x) dx, \]

it is possible to show that

\[ \frac{dV}{dt} = rV \]

(1.63)

**Convergence to the stable age-distribution**

Obviously the properties of the stable age-distribution which we have just discussed are not of much interest unless real populations converge towards stability in a reasonably short time. The rate of convergence can be most directly studied using the difference equation solution given by equation (1.37), which can be rewritten as

\[ e^{-\gamma B}(t - 1) = C_1 + \sum_{j=2}^d C_j e^{(\zeta_j - \gamma) t} \]

(1.64)

where the \( e^{\zeta_j} \) (with \( \zeta_j = \phi_j + i\omega_j \)) are the roots of equation (1.32) other than \( \lambda_1 \). Each complex root \( \lambda_j \) has a conjugate root, \( e^{\phi_j - i\omega_j} \), whose \( C \)
Parameters independent of time and density

coefficient is the conjugate of $C_j$. Each pair of complex conjugate roots therefore contributes an oscillatory term of the form

$$e^{(\phi_j - r)t} \left[ C_j \cos(\omega_j t) + C_j' \sin(\omega_j t) \right]$$

to the right-hand side of equation (1.64). Similarly a real, negative root, $\lambda_k$ say, contributes a term of form

$$C_k e^{(\phi_k - r)t} \cos(\pi t)$$

which oscillates between positive and negative values in successive time-intervals. The amplitudes of these oscillatory terms diminish at a rate determined by the values of $\phi_j - r$, which we have already shown to be negative (pp. 23–5). The constant term in equation (1.64) comes to predominate after a time that is governed by the magnitude of the difference between $r$ and the largest of the $\phi_j$, $\phi_2$ say. During a period of time of order $1/(r - \phi_2)$, the smaller roots still contribute significantly, and generate waves in the value of $B_f(t)$, and hence in the numbers of individuals in other age-classes. Keyfitz (1968, Chapter 6; 1972) discusses the period of these waves and the rate at which they damp out, for the type of reproductive function characteristic of human populations. The complex part $\omega_2$ of the largest complex root is approximately equal to $2\pi/T_0$, where $T_0$ is the measure of generation time defined by equation (1.47a). The periodicity of the waves in a population which is converging to its stable age-distribution is thus of the order of the ‘generation time’. The modulus of the real part $\phi_2$ is such that effective convergence to the stable age-distribution takes place within approximately five generations. This high speed of convergence seems to be fairly common, on the basis of the numerical examples which have been studied in the literature. Figures 1.1 and 1.2 show two examples, using the sets of vital statistics quoted earlier.

1.3.3 The dynamics of the population process for more complex types of population

The analysis given above is valid only for the female segment of the population, or (with appropriate notational changes) for a purely hermaphrodite population. We shall now consider briefly how it can be extended to the male segment of the population, and to the case of a species of plants with a distinct seed population (see pp. 7–9).

The male population

On the assumption of time-independent demographic parameters, and the model leading to equation (1.12), it is clear that the male
population must approach an asymptotic state with a stable age-distribution at the same rate as the female population. This follows from the fact that, on the assumption of time-independent age-specific sex-ratios \( a(x) \) (which is necessary for a time-independent \( m(x) \) function), the number of male offspring produced in a time-interval is completely determined by the numbers of females of different ages. As the age-structure of the female population approaches constancy, therefore, the number of male births in a time-interval, \( B_m(t) \), must approach the same rate of increase, \( \lambda_1 \), as the number of female births, \( B_f(t) \). More formally, it is known from standard theory that the set of eigenvalues of the population projection matrix in equation (1.12) (with time-independent demographic parameters) is identical with the set of eigenvalues of the sub-matrices \( L \) and \( B \); it can easily be shown the latter are all zero, so that the leading eigenvalue of the matrix for the population of males and females is \( \lambda_1 \), giving the asymptotic rate of increase of both males and females. It follows that the proportion of males in a population which is effectively in stable age-distribution is given by an equation similar in form to equation (1.50a), with \( l(x) \) replaced by \( l^*(x) = \prod_{y=1}^{y=x} P^*(x - y) \). Furthermore, in a population in stable age-distribution the male age-specific fecundity function \( m^*(x) \), as defined by equation (1.7), must be independent of time. A male reproductive function \( k^*(x) = l^*(x) m^*(x) \) can thus be defined for the stable age-distribution; application of an argument similar to that leading to equation (1.17) for the female population gives the relation (for \( t \geq d^* \))

\[
B_m(t) = \sum_{x=b^*}^{d^*} B_m(t - x) k^*(x)
\]

(Modifications similar to those made in deriving equation (1.19) can be used to obtain an expression for \( B_m(t) \) when \( t < d^* \).) Use of the renewal theory applied to the female population process of pp. 25–6 gives us the characteristic equation

\[
\sum_{x=b^*}^{d^*} \lambda^{-x} k^*(x) = 1
\]  

(1.65)

From the above considerations, the leading root of this equation must be \( \lambda_1 \). Equations (1.32) and (1.65) thus relate the male and female demographic parameters for the stable age-distribution. The values of \( m^*(x) \) used for obtaining \( k^*(x) \) must, of course, be found either from empirical data on the male age-specific fecundity schedule for the stable age-distribution, or by detailed specification of the rules of mating with respect to male and female ages. Generation times, reproductive values
etc. can be defined using m*(x) and k*(x), just as for the female population.

Plants with seed dormancy

The population projection matrix in equation (1.14) can be handled, for the case of time-independent demographic parameters, by matrix methods similar to those used on pp. 22–5 for the female population. The characteristic equation is of a rather more complex form than equation (1.32), however, and it is simpler to derive it using the difference equation approach. Let us define the following quantities:

- \( l'(x) \) (\( x = 1, 2, \ldots, d' + 1 \)) is the probability that a seed survives to be aged \( x \), and then germinates;
- \( l(x) \) (\( x = 0, 1, \ldots, d - 1 \)) is the probability that a plant survives \( x \) years from the time of germination. We have:

\[
l'(1) = P''(0) \tag{1.66a}
\]

\[
l'(x) = P''(x - 1) \prod_{y=2}^{x} P'(x - y) \quad (x = 2, 3, \ldots, d' + 1) \tag{1.66b}
\]

\[
l(0) = 1 \tag{1.66c}
\]

\[
l(x) = \prod_{y=0}^{x-1} P(x - y) \quad (x = 1, 2, \ldots, d) \tag{1.66d}
\]

The net expected number of progeny seeds ascribable to a seed \( x \) years after it was itself formed is thus

\[
K(x) = \sum_{y=0}^{x-1} l'(x - y) l(y) M(y + 1) \quad (x = 1, 2, \ldots, d + d')
\]  

(1.67)

If we define \( B(t) \) as the total number of seeds produced by the population during the flowering season of year \( t \), we have (for \( t \geq d + d' \))

\[
B(t) = \sum_{x=1}^{d+d'} B(t - x) K(x) \tag{1.68}
\]

Applying the renewal theory results, the asymptotic rate of increase of the population is thus the leading root \( \lambda_1 \) of the equation

\[
\sum_{x=1}^{d+d'} \lambda^{-x} K(x) = 1 \tag{1.69}
\]
Provided that $K(x)$ is not periodic, $\lambda_i$ is real and positive. As the population approaches its asymptotic growth-rate, the age-structures of both the seed and adult plant populations must stabilise.

1.4 Time-dependent and density-dependent demographic parameters

The solution of the population process studied in the preceding section of this chapter is based on the assumption that the demographic parameters of the population are independent of time. This is clearly at best an approximation to the behaviour of a real population, and it obviously is important to examine the consequences of relaxing this assumption. It is useful to distinguish between two ways in which time-dependence of the demographic parameters can arise; factors of the external environment which affect survival or fecundity may vary over time (time-dependence in the strict sense), or changes in the size and age-composition of the population may cause changes in the demographic parameters (density-dependence). Of course, both of these processes may operate at the same time in a real population, but it is simplest to treat them as alternatives. We begin by discussing the case of time-dependent demographic parameters, and then go on to consider density-dependence.

1.4.1 Time-dependent demographic parameters

This situation may be further sub-divided into two. There is firstly the case of a deterministically varying environment, where we can follow a population which is exposed to some definite pattern of variation over time in its demographic parameters. Secondly, there is the case of a stochastically varying environment, where the successive sets of demographic parameters to which a population is exposed are governed by some probability law. As we shall see, it is not possible in either of these cases to obtain useful analytical solutions for the age-structure and growth-rate of the population at a given time, as was possible with time-independence. It is possible to prove, however, that the initial age-structure of the population eventually ceases to influence the age-structure and growth-rate of the population. This is the so-called weak ergodicity property of the population process, which may be contrasted with the strong ergodicity property of a population with fixed demographic parameters (p. 28); in that case, the asymptotic state is characterised by a constant, calculable age-distribution and growth-rate. For simplicity, we shall consider only the female population in the following analysis.
Deterministically varying environments

Consider first of all the case of a discrete age-class population. Its state at time $t$ is characterised by the vector of number of individuals in different age-classes, $n(t)$, and by the Leslie matrix for time $t$, $L(t)$ (cf. pp. 3–6). Using equations (1.5) and (1.6), we can express $n(t)$ in terms of the initial population vector $n(0)$, and the sequence of Leslie matrices between times 0 and $t$

$$n(t) = n(0) \prod_{u=0}^{t-1} L(u)$$

(1.70)

This result is trivial and does not provide in itself any insight into the properties of the population process, although it can be used for numerical calculations of population trajectories. Lopez (1961) used it, however, to prove that the age-structure of the population eventually becomes independent of the initial vector $n(0)$. Let the elements of the vectors $n_1(t)$ and $n_2(t)$ describe two populations with two different initial vectors, $n_1(0)$ and $n_2(0)$, but which are exposed to the same set of Leslie matrices; then, for large $t$, the following equation is satisfied

$$\frac{n_1(x, t)}{n_2(x, t)} \sim \frac{n_1(y, t)}{n_2(y, t)} \quad (x \neq y)$$

(1.71a)

If we use the fact that the demographic parameters are identical for the two populations at any given time, it is easily verified that this relation implies that the two population vectors satisfy the equation

$$n_2(t) \sim Cn_1(t)$$

(1.71b)

where $C$ is a constant whose value depends on the initial vectors $n_1(0)$ and $n_2(0)$. This turn implies that the geometric growth-rate of the population at time $t$, as measured by the relative values of (say) $n(1, t + 1)$ and $n(1, t)$, becomes dependent only on $t$, and is independent of the initial conditions.

The following conditions are sufficient for the validity of this theorem. The non-zero elements of $L$ are bounded, and greater than some fixed number $\varepsilon > 0$. There are two values of $x$, $x_1$ and $x_2$, with greatest common divisor of one, such that $f(x_1, t)$ and $f(x_2, t)$ are non-zero for all values of $t$. (This is related to the requirement of non-periodicity for the time-dependent case, discussed on p. 23.) The survival probabilities $P(x, t)$ are non-zero, as well as at least some of the elements of the initial population vector $n(0)$. The proof requires fairly lengthy matrix manipulations, which are beyond the scope of this book. The interested reader
should consult Lopez (1961), Pollard (1973, pp. 51–5) for details of the proof. Seneta (1973, pp. 69–77) has proved this theorem for the case of non-negative matrices of more general structure than the Leslie matrix; his results are used in section 2.2.

A theorem of the same sort exists for continuous-time populations, and was first stated and proved by Norton (1928), in a paper which has largely been overlooked by workers in demography. Norton assumed that the reproductive function \( k(x, t) \) in equation (1.27) is some defined function of \( t \) as well as \( x \), due to temporal variation in environmental factors affecting survival or fecundity. He assumed that there were fixed upper and lower limits, \( b \) and \( d \), to the values of \( x \) for which \( k(x, t) \) is non-zero; and that \( k(x, t) \) is an integrable function of \( x \) and \( t \). (Continuity of \( k(x, t) \) with respect to both \( x \) and \( t \) is sufficient but not necessary for the validity of Norton’s results.) Provided that there are at least some individuals of reproductive or pre-reproductive age at time \( t = 0 \), Norton showed that:

given two communities in which there is the same birth-rate and death-rate [age-specific fecundities and death-rates in our terminology], the number of births in one tends ... to a constant proportion of the number of births at that moment in the other ... Finally, if the birth-rate and death-rate are given – and provided the conditions stated are fulfilled – the eventual age-distribution is determinable; it will vary, of course, from moment to moment, but is approximately the same at the same time in all communities in which the birth- and death-rates have their given values; in particular, it will not depend on the initial age-distribution (Norton, 1928, p. 20).

As in the case of time-independent demographic parameters, the properties of the discrete age-class and continuous-time models are parallel. The general conclusion is that the age-structure and rate of growth of a population subjected to demographic parameters which vary with time tend asymptotically to values which are independent of the initial state of the population. The speed of this convergence is difficult to study analytically; some numerical studies have been made by Kim and Sykes (1976), who conclude that convergence takes place over a few generations for sets of data drawn from human populations. These results are of some importance in population genetics, as we shall in Chapters 2 and 4.
Stochastically varying environments

Cohen (1976, 1977a, b, 1979a, b) has extended these conclusions to the case of a population exposed to stochastic variation in the demographic parameters. He considers a discrete age-class model described by Leslie matrices of the same structure assumed by Lopez (1961); the sequence of Leslie matrices applied to a population is assumed to be sampled from an ergodic Markov chain (i.e. the probability law relating the state of $L$ at two successive times is such that the distribution of $L(t)$ becomes independent of $t$ for sufficiently large $t$). Under these conditions, the age-structure and growth-rate of the population exhibit stochastic ergodicity; the probability distribution of the age-structure of the population eventually become independent of the initial vector $n(0)$, and of the initial distribution of the demographic parameters. The details of the assumptions and methods involved in the proof of this result can be found in the references cited above. Cohen (1977b) gives an example of a simulation of this type of population process, which illustrates the convergence of the probability distribution of the age-structure to its limiting state.

Mean growth-rate of a population in a varying environment

An important question in relation to varying environments concerns the mean asymptotic growth rate of the population, over the set of environments encountered by the population in the deterministic case, or over the probability distribution of environments in the stochastic case. This topic has been thoroughly reviewed by Tuljapurkar (1989, 1990a, Chapters 4, 12), and only the major points relevant to the population genetic and evolutionary topics discussed in Chapters 4 and 5 will be considered here.

First, it is necessary to clarify what is meant by the mean asymptotic growth-rate of a population. The most useful definition, for both demographic and genetic purposes, is the mean logarithmic growth-rate, denoted here by $\bar{r}$. If the conditions for weak ergodicity or stochastic ergodicity are met, this quantity can be calculated from the rate of growth of any component of the population vector $n(t)$, or of any linear combination of the elements of $n(t)$ given by $n(t)e^T$, where $e$ is an arbitrary row vector. As can be seen as follows, this result is implied by the fact that the relative values of the components of $n(t)$ become independent of the initial conditions (although dependent on time). Write $n(1, t)/n(1, t - 1) = \lambda(t - 1)$, so that

$$n(1, t) = n(1, 0) \prod_{u=1}^{t} \lambda(t - u)$$  \hfill (1.72a)
Using the property of ergodicity, it follows that, for large $t$, we have

$$n(t)e^T = n(1, t)\psi(t) = n(1, 0)\psi(t)\prod_{u=1}^{t} \lambda(t-u)$$  \hspace{1cm} (1.72b)$$

where $\psi(t)$ is a bounded function, determined by $e$ and the age-structure of the population at time $t$. We can then write

$$\bar{r} = \lim_{t \to \infty} \frac{1}{t} E \left\{ \sum_{u=1}^{t} \ln \lambda(t-u) \right\} = \lim_{t \to \infty} \frac{1}{t} E \left\{ \ln n(t)e^T \right\}$$  \hspace{1cm} (1.73)$$

This relation is purely formal, however, and does not permit evaluation of $\bar{r}$ except under special conditions, or by numerical computations. One special case of interest is when fecundity and survival from age 1 onwards are independent of age, but dependent on time (Tuljapurkar, 1982; Bulmer, 1985). Let the fecundity function for all ages, including 1, be $f(t)$ at time $t$, and the survival probability be $P(t)$. Let the total number of individuals in the population at time $t$ be $n(t)$. The Leslie matrix representation of equation (1.5) is easily seen to lead to the result

$$n(t) = n(t-1)[f(t-1) + P(t-1)]$$  \hspace{1cm} (1.74)$$

so that

$$\lambda(t) = f(t) + P(t)$$  \hspace{1cm} (1.75)$$

The mean intrinsic rate of increase, $\bar{r}$, is simply the mean of $\ln \lambda(t)$, i.e. the mean of the intrinsic rates for each environment.

A related result holds for cases with arbitrary patterns of dependence of survival and fecundity on age, when environmental fluctuations take place over a time period which is long in relation to the time it takes for age-structure to equilibrate in a constant environment. Under this condition, population growth will effectively be determined by the intrinsic rates associated with each successive environment, i.e. $\bar{r}$ is approximately equal to the arithmetic mean of the intrinsic rates over all environmental states encountered by the population.

In general, there are formidable difficulties in obtaining useful analytic results on $\bar{r}$ for arbitrary patterns of environmental variation. An approximate approach for the case of small random fluctuations in the demographic parameters has been introduced by Tuljapurkar (1982, 1989, 1990a, Chapter 12). Here we will consider the results for the case when there are no correlations between the values of the demographic parameters in successive time-intervals. This is the opposite extreme to the situation with long-period fluctuations which we have just discussed.
Formulae for the case when there are correlations between successive time-intervals are given by Tuljapurkar (1982, 1989, 1990a, Chapter 12).

Let the Leslie matrix for time \( t \) be written as \( \mathbf{L}(t) = \mathbf{L} + \delta \mathbf{L}(t) \), where \( \mathbf{L} \) is the expected value of the Leslie matrix and \( \delta \mathbf{L}(t) \) is the deviation from this expectation. The expectation of \( \delta \mathbf{L}(t) \) is zero by definition; the expectations of the cross-products of its components, of the form \( E\{ \delta \mathbf{L}_{ij} \delta \mathbf{L}_{kl} \} \), give the covariances of the components in question, \( \text{Cov}(\mathbf{L}_{ij}, \mathbf{L}_{kl}) \). We have

\[
\mathbf{L}(t-1)\mathbf{L}(t-2)\ldots\mathbf{L}(0) = \mathbf{L}' + \phi(t) + \zeta(t) \tag{1.76a}
\]

where

\[
\phi(t) = \sum_{u=1}^{t} \mathbf{L}^{u-1} \delta \mathbf{L}(t-u)\mathbf{L}^{t-u} \tag{1.76b}
\]

and \( \zeta(t) \) is a function of terms involving cross-products of the form \( \delta \mathbf{L}(t)\delta \mathbf{L}(t-u) \) \((u > 1)\), plus higher-order \( \delta \mathbf{L} \) terms.

We can simplify equation (1.76a) by using the fact that

\[
\mathbf{U}_1\mathbf{L}'\mathbf{V}_1^\top = \lambda_1' \mathbf{U}_1\mathbf{V}_1^\top = \lambda_1'
\]

where \( \lambda_1 \) is the leading eigenvalue of the mean Leslie matrix \( \mathbf{L} \), and \( \mathbf{U}_1 \) and \( \mathbf{V}_1 \) are the corresponding row and column eigenvectors, normalised such that their product is one.

Using \( \mathbf{U}_1 \) as the initial vector in (1.72b), and substituting into equations (1.73) and (1.76), we obtain

\[
\tilde{r} = \ln \lambda_1 + \lim_{t \to \infty} \frac{1}{t} \mathbb{E} \left\{ \ln \left( 1 + \frac{\mathbf{U}_1 \phi(t)\mathbf{V}_1^\top + \mathbf{U}_1 \zeta(t)\mathbf{V}_1^\top}{\lambda_1'} \right) \right\} \tag{1.77}
\]

From the assumption of small \( \delta \mathbf{L} \) values, and of no covariance across time-intervals, all terms in \( \zeta \) can be neglected, and only second-order terms in \( \phi \) need to be kept in the expansion of the logarithm in equation (1.77). We also have

\[
\mathbb{E}\{(\mathbf{U}_1 \phi(t)\mathbf{V}_1^\top)^2\} = \mathbb{E}\left\{\sum_{u=1}^{t} (\mathbf{U}_1 \delta \mathbf{L}(t-u)\mathbf{V}_1^\top)^2\right\}
\]

Using this in equation (1.77) together with equations (1.31), and expanding the expectation on the right-hand side of (1.77), we get

\[
\tilde{r} \approx r - \frac{1}{2\lambda_1'^2} \sum_{ijkl} \mathbf{U}_1(i)\mathbf{V}_1(j)\mathbf{U}_1(k)\mathbf{V}_1(l) \text{Cov}(\mathbf{L}_{ij}, \mathbf{L}_{kl}) \tag{1.78a}
\]

where \( r = \ln \lambda_1 \) is the intrinsic rate of the mean Leslie matrix \( \mathbf{L} \).
From equation (1.58b), this can be written as

$$\bar{r} = r - \frac{1}{2} \sum_{ijkl} \left( \frac{\partial r}{\partial L_{ij}} \right) \left( \frac{\partial r}{\partial L_{kl}} \right) \text{Cov} (L_{ij}, L_{kl})$$

Equations (1.78) provide a relatively simple approximate expression for the mean population growth-rate in a varying environment, provided there is no autocorrelation between successive time-intervals. This expression will prove useful in the applications to population genetics and evolution in Chapters 4 and 5. Numerical results show that this expression provides an excellent approximation, even when the variability in demographic parameters is high relative to the mean (Orzack and Tuljapurkar, 1989).

The arithmetic mean of the intrinsic rates for each environment was shown above to be the appropriate measure for highly autocorrelated environments. The first two non-zero terms in the Taylor series expansion around $L$ of this quantity are

$$r - \frac{1}{2} \sum_{ijkl} \left( \frac{\partial^2 r}{\partial L_{ij} \partial L_{kl}} \right) \text{Cov} (L_{ij}, L_{kl})$$

Hence, to the order of approximation for which equations (1.78) are valid, the measures of mean growth-rate for the cases of zero and high correlation between successive environments are by no means the same (Orzack and Tuljapurkar, 1989), despite the widespread assumption in the older literature that the arithmetic mean of the intrinsic rates for each environment is the most appropriate measure of average fitness in varying environments (Schaffer, 1974b; Lacey et al., 1983).

1.4.2 **Density-dependent demographic parameters**

It is only exceptionally that organisms find themselves in an environment where there is no check to their continued increase in numbers due to the operation of factors which increase mortality or reduce fecundity, in response to increasing population density in a local area (density-dependent factors). Although the nature of such checks to population growth varies from species to species, and the full details of the mechanism of regulation of population density have probably never been worked out for even one species, there can be little doubt of their general importance in ecology (Lack, 1954; Nicholson, 1957, 1960). One of the standard texts on ecology, such as Begon, Harper and Townsend (1990) or Ricklefs (1990), may be consulted for more information about
density-dependent factors in natural populations. Our concern here is to examine how density-dependent demographic parameters can be incorporated into models of the sort outlined earlier in this chapter (pp. 3–15). It is useful to start by considering the properties of models of density-dependent populations which neglect age-structure.

Density-dependence in populations without age-structure

The classical approach to modelling a density-dependent population is to write down a differential equation which describes the rate of increase in the number of individuals, \( n \), in a local population

\[
\frac{dn}{dt} = ng(n) \tag{1.79}
\]

where \( g(n) \) is a continuous, decreasing function of \( n \). As we have done earlier, it is best to regard \( n \) as the number of females only. The simplest and most widely used form for \( g(n) \) is that of the logistic growth or Pearl–Verhulst equation

\[
g(n) = \alpha - \beta n \tag{1.80}
\]

The expression which results if equation (1.80) is substituted into (1.79) is readily integrated to give \( n \) as a function of time

\[
n(t) = \frac{\alpha n(0)}{\beta n(0) + [\alpha - \beta n(0)]e^{-\alpha t}} \tag{1.81}
\]

If the population is started with a population size close to zero, it increases initially at a logarithmic rate of approximately \( \alpha \); provided \( n \) is initially positive, the population converges smoothly towards a stationary population size of

\[
\hat{n} = \frac{\alpha}{\beta} \tag{1.82}
\]

\( \hat{n} \) is usually referred to as the carrying-capacity of the population. The logistic equation provides a reasonably good empirical fit to the form of population growth of model laboratory populations, such as population cages of \textit{Drosophila} (e.g. Buzzati-Traverso, 1955).

More complicated forms for \( g(n) \) can be studied in a similar way (May, 1973). If \( g(n) \) is a strictly decreasing function of \( n \), it is obvious from equation (1.79) that there is only one carrying-capacity or stationary population size \( \hat{n} > 0 \), which must satisfy \( g(n) = 0 \). It is not always possible to integrate equation (1.79) directly to obtain \( n \) as an explicit function of \( t \). The convergence to the carrying-capacity can be studied in such a case by examining the local stability of the equilibrium, \( \hat{n} \).
Applying Taylor's theorem to equation (1.79), for \( n \) close to \( \hat{n} \), and noting that \( g(\hat{n}) = 0 \), we obtain

\[
\frac{dn}{dt} \equiv (n - \hat{n}) \left( \frac{\partial g}{\partial n} \right) \hat{n}
\]

i.e.

\[
n(t) - \hat{n} \equiv [n(0) - \hat{n}] \exp \left[ t \left( \frac{\partial g}{\partial n} \right) \hat{n} \right]
\]  

(1.83)

Since \( g \) is a decreasing function of \( n \), \( (\partial g/\partial n)\hat{n} \) is negative, so that the exponential term decays to zero with increasing time. The equilibrium is therefore locally stable; a small perturbation away from the carrying-capacity results in a return towards equilibrium.

Equation (1.79) is clearly a highly over-simplified representation of the dynamics of a population. It is strictly valid only if the individuals have death-rates \( \mu(x, t) \) and fecundities \( m(x, t) \) which are independent of age \( x \), but depend on total population size. In order to avoid the loss of realism inherent in applying such equations to age-structured populations reproducing in continuous time, theoretical ecologists have developed similar models of density dependence for populations with completely discrete generations. These provide reasonably accurate descriptions of, for example, annual plants and many semelparous insect species of economic importance, which live only one year. Equation (1.79) can be replaced by the related first-order difference equation

\[
n(t) = n(t - 1)w[n(t - 1)]
\]

(1.84)

where \( n(t) \) is the number of adult females alive in generation \( t \), and \( w \) is the expectation of female offspring (counted at maturity) of an adult female in the preceding generation. A plausible and simple form for \( w(n) \), analogous to the logistic form for the continuous-time case, is

\[
w(n) = \alpha e^{-\beta n}
\]

(1.85)

The carrying-capacity \( \hat{n} \) now satisfies the equation

\[
w(\hat{n}) = 1
\]

(1.86)

If \( g(n) \) is a strictly decreasing function of \( n \), there is a unique solution to this equation. The dynamics of the population in the neighbourhood of the equilibrium can again be studied by a local analysis. We have

\[
n(t) - \hat{n} \equiv [n(t - 1) - \hat{n}] \left[ 1 + \hat{n} \left( \frac{\partial w}{\partial n} \right) \right]
\]

(1.87)
If \(-2 < \mathcal{N}(\partial w / \partial n) \mathcal{N} < 0\), the equilibrium is locally stable. If \(\mathcal{N}(\partial w / \partial n) \mathcal{N} < -2\), the population responds to small perturbations away from equilibrium by oscillations of progressively increasing magnitude across \(\mathcal{N}\), i.e., it is locally unstable. If \(w(n)\) is of the form given by equation (1.85), for example, we have \(\mathcal{N} = (\ln \alpha) / \beta\) and \(\mathcal{N}(\partial w / \partial n) \mathcal{N} = -\ln \alpha\), so that the equilibrium is locally stable for \(\ln \alpha < 2\), and unstable if \(\ln \alpha > 2\). Too great a sensitivity of the population growth-rate to population density thus leads to instability. But this stability analysis is valid only for the neighbourhood of an equilibrium; it is obvious that a population close to \(n = 0\) must tend to increase in number, and a population with very large \(n\) will tend to decrease, provided that \(w(n)\) is a decreasing function of \(n\). This suggests that the local instability of the equilibrium actually corresponds to some kind of bounded fluctuations in population size. This was apparently first pointed out in an ecological context by Moran (1950) and, independently, by Ricker (1954).

More detailed analyses of the dynamical behaviour of expressions of the form of equation (1.84) have considerably deepened our understanding of the complex dynamics which are possible when there is formally a locally unstable equilibrium (reviewed by May and Oster, 1976; Nisbet and Gurney, 1982; Schaffer and Kot, 1986). Consider, for example, the population model expressed by equation (1.85). If \(\mathcal{N}\) is kept fixed, so that \(\ln \alpha\) and \(\beta\) are varied proportionately, it is found that the population converges to \(\mathcal{N}\) for \(\ln \alpha < 2\). When \(\ln \alpha\) is slightly greater than 2, the population approaches a limit cycle of period 2, i.e., the population trajectory settles into a state where \(n\) switches each generation between two alternative values, one above \(\mathcal{N}\) and the other below. If \(\ln \alpha\) and \(\beta\) are further increased, this limit cycle is replaced by one with period 4 (the population approaches a state in which it switches between four successive values of \(n\)), which is in turn replaced by one of period 8, etc. The dynamical behaviour of the population thus experiences a succession of bifurcations such that, with increasing sensitivity to population density, successive stable limit cycles of period 4, 8, \ldots, \(2^m\ldots\) are generated. It can be shown that the pattern of bifurcation behaviour depends only on \(\alpha\), so that two populations with different \(\mathcal{N}\) but the same \(\alpha\) will have the same type of limit cycle (May and Oster, 1976). (The actual values of \(n\) through which the population passes are, of course, dependent on \(\beta\) as well as \(\alpha\).) When \(\ln \alpha\) exceeds 2.6924, there is an infinite number of limit cycles of differing periods; there is also an uncountable set of initial conditions for which the population trajectories are aperiodic, so that the pattern of population size with respect to time never repeats itself. This is
the so-called region of chaos. A population in this region will behave in an essentially unpredictable fashion, with periods of low population size succeeded by sudden increases in number, followed by rapid return to a low level once more.

As is obvious from the simple method described above, cyclic or chaotic behaviour depends on the species having a high reproductive capacity, so that \( \alpha \) is sufficiently large. If the rate of population growth at low density is small, then the population will settle down to a stable equilibrium level, \( \hat{n} \), as given by equation (1.86). Hassell, Lawton & May (1976) analysed data on several species, and concluded that most seem to be in the region of stability or 2-point or 4-point limit cycles, so that chaotic phenomena are perhaps infrequent in natural populations. Other workers have recently argued for an important role for cyclical or chaotic population behaviour (Schaffer and Kot, 1986; Turchin and Taylor, 1992).

Models of density-dependence with age-structure

The complex behaviour of discrete-generation populations, revealed by the studies summarised above, suggests that it is important to examine carefully the properties of age-structured populations with density-dependence. Various special cases have been studied numerically by Leslie (1948), Pennycuick, Compton & Beckingham (1968), Usher (1972), Beddington (1974), Smouse and Weiss (1975), and others. Analytical treatments have been given by Gurtin and MacCamy (1974), Oster and Takahashi (1974), Oster (1976), Rorres (1976) and Guckenheimer, Oster and Ipaktchi (1977). General surveys are given by Nisbet and Gurney (1982) and Costantino and Desharnais (1991). We will first consider some general properties of density-dependent regulation in age-structured populations, and then go on to discuss the dynamical properties of some special cases.

It is clear, both on theoretical grounds and from the results of empirical studies, that density-dependent regulation in an age-structured population does not usually involve a response of the demographic parameters to the total number of individuals present at one time. More usually, the density-dependent components of the demographic parameters respond only to the numbers of individuals in a restricted sub-group of the population. Such a sub-group has been called the critical age-group by Charlesworth (1972). For example, in many bird species density-dependence seems to occur through limited availability of food for nestlings (Lack, 1954, 1966), so that the survival of nestlings decreases with an increase in the number of nestlings in a given area. Similarly, seedling
survival in plants is affected by competition between seedlings for light and space, resulting in negative dependence of seedling survival on the numbers of seedlings (Harper and White, 1974). The critical age-group in these cases is composed simply of the nestlings or seedlings, respectively. Nicholson (1957, 1960) describes experimental populations of the sheep blowfly *Lucilia cuprina* in which fecundity of adult females was controlled by competition for a limited food supply. In this case, the critical age group consists of the entire adult female population.

The simplest type of reasonably realistic model of density-dependence in a discrete age-class model represents the age-specific survival probabilities or fecundities, for one or more age-classes, as decreasing functions of the number of individuals in the critical age-group. Let this number be $N(t)$ at time $t$. Consider, for example, the case of a species which starts reproduction $b$ years after birth, and in which juvenile survival depends on the total number of juvenile individuals. For simplicity, only the female population is considered. We can write the survival probability for a juvenile aged $x$ at time $t$ as

$$P(x, t) = P^{(l)}(x)P^{(D)}[x, N(t)] \quad (x = 0, 1, \ldots, b - 1)$$

where

$$N(t) = \sum_{x=1}^{b-1} n(x, t)$$

$P^{(l)}(x)$ is the density-independent component of juvenile survival for age-class $x$; $P^{(D)}[x, N(t)]$ is the density-dependent component, which is assumed to be a decreasing function of $N(t)$.

The population process in this case can be represented by a Leslie matrix with density-dependent survival probabilities $P(x, t)$ for ages $x = 1, 2, \ldots, b - 1$, and density-dependent net fecundities $f(x, t) = m(x)P(0, t)$. Alternatively, we can represent the process by a difference equation of the form of equations (1.17) or (1.19). We can write

$$k(x, t) = k^{(l)}(x)k^{(D)}(x, t)$$

where

$$k^{(l)}(x) = m(x)\prod_{y=0}^{b-1} P^{(l)}(y)\prod_{y'=b}^{x-1} P(y')$$

$$k^{(D)}(x, t) = m(x)\prod_{y=0}^{b-1} P^{(D)}[y, N(t + y - x)]$$
The dependence of the reproductive function on the values of $N(t)$ over a set of past times, as expressed in equation (1.88a), is typical of what happens with density-dependence in an age-structured population, and greatly complicates the analysis of its dynamics. In general, we can represent it by writing

$$k(x, t) = k(x, N_T)$$

(1.89)

where $N_T$ represents the set of values of $N(t)$ over some set $T$ of past times. As in the above example, the actual set of times included in $T$ may depend on $x$. The population process is then described in general terms (for times $t \geq d$) by the equations

$$N(t) = \sum_{x \in S} B_f(t-x)l(x, N_T)$$

(1.90a)

$$B_f(t) = \sum_{x=b}^d B_f(t-x)k(x, N_T)$$

(1.90b)

$S$ in equation (1.90a) represents the set of age-classes which comprise the critical age-group.

Appropriate modifications can be made for $t < d$ (equation (1.19)). An equivalent matrix formulation can obviously be written down, but the difference equation method of description is convenient for dealing with the problem of selection in density-dependent populations, as will be discussed in Chapters 3 and 4. The extension to the corresponding continuous-time models is straightforward and will not be given here.

**Stationary populations**

By analogy with the results for discrete-generation populations, we would expect that an age-structured population with a sufficiently low reproductive capacity at low population density should converge towards an equilibrium state with a stationary number of individuals in each age-class. In such a stationary population, equation (1.90b) gives us the relation

$$\sum_x k(x, N_T) = 1$$

(1.91a)

By definition, the number of individuals in the critical age-group is unchanging, so that each number in the set $N_T$ is constant at the equilibrium value of $N(t)$, $\hat{N}$ say. This corresponds to the carrying-capacity of the population in the simple models described above (pp. 51–2). The value of $\hat{N}$ can be determined from this equation by the following
procedure. We can imagine a set of populations held at different stationary states by some artificial means. Let the number of individuals in the critical age-group in one such imaginary population be $N$, say. If the density-dependent components of the demographic parameters are *strictly* decreasing functions of the number of individuals in the critical age-group, it follows that, in such a stationary population, the net reproduction rate is a strictly decreasing function of $N$. $\hat{N}$ is thus given as the *unique* root of the equation

$$\sum_x k(x, N) = 1$$

This equation is analogous to equation (1.45), which gives the intrinsic rate of increase in the density-independent case. As in that case, the solution usually has to be obtained by numerical iteration of the equation.

The uniqueness of $\hat{N}$, which follows from the particular model of density-dependence just described, is useful in analysing the mathematical properties of the system, especially in connection with the theory of selection with density-dependence, to be discussed in Chapters 3 and 4. It should be borne in mind that, although this class of model probably provides a fairly realistic description of the process of density regulation in many situations, other types of model with more complex properties may apply to individual cases. For example, different age-classes within the critical age-group may contribute differentially to the impact of density on the demographic parameters, or the decreasing properties of the latter as functions of $N(t)$ may hold only for a certain range of population size. Beddington (1974) discusses an example of the second sort, which results in two alternative equilibria for the same set of demographic parameters.

*Non-equilibrium behaviour*

A detailed analysis of the non-equilibrium behaviour of these models usually presents considerable mathematical difficulties, and computer studies have frequently been resorted to. Oster (1976), Charlesworth (1980), Nisbet and Gurney (1982) and Costantino and Desharnais (1991) review general aspects of the population dynamics resulting from density-dependence with age-structure, as well as studies of some individual cases. The results which they discuss indicate that age-structured populations may show cyclical or chaotic fluctuations in numbers, provided that the reproductive capacity of the species is sufficiently high. Organisms with relatively low fecundity may be expected to approach a stationary population size under density-dependence. There is still debate
as to whether these instability phenomena are important agents in contributing to cyclic population phenomena in nature (Oster and Takahashi, 1974; Hassell et al., 1976; Schaffer and Kot, 1986; Turchin and Taylor, 1992; Godfray and Grenfell, 1993).

1.5 Systems of several populations

We have so far considered only the case of a single, genetically homogeneous population. In practice, of course, a species consists of a mixture of different genotypes, and exists as a member of a community of species which interact with each other by competition, predation, parasitism, mutualism, etc. Most ecological models of interacting species have ignored the problem of age-structure, although models of predator-prey interactions, for example, have been studied by Maynard Smith and Slatkin (1973), Auslander, Oster and Huffaker (1974) and Beddington and Free (1976). It is clear from these studies that incorporating age-structure may introduce novel features into the behaviour of multi-species systems. As this book is concerned mainly with the consequences of genetic differences within a species, the problem of ecological interactions between species will be considered only briefly. An asexually reproducing population which contains a number of genetically distinct clones is formally equivalent to a system of different, non-interbreeding species. We will examine here the simplest kind of system of this sort: one in which the different sub-populations (asexual genotypes or species) have different sets of demographic parameters which remain constant over time, and are unaffected by the composition of the system. This obviously excludes the standard types of ecological interaction mentioned above, and is therefore most relevant to the process of selection within an asexually reproducing population. The models explored below were originally investigated by Pollak and Kempthorne (1970, 1971).

1.5.1 Continuous-time populations

It is convenient to examine the continuous-time model first. An asexual population with \( m \) distinct genotypes \( A_1, A_2, \ldots, A_m \) can be regarded as composed of \( m \) separate sub-populations, each with its own set of demographic parameters. The sub-population \( A_i \) is characterised by a reproductive function \( k_i(x) \), with which an intrinsic rate \( r_i \) is associated \((r_i \text{ is obtained by applying equation (1.42) to } k_i(x))\). In the case of an asexual population, there is obviously no need to distinguish between the sexes in calculating fecundities; if the system is regarded as representing a set of sexual but non-interbreeding species, we can follow our earlier practice, and treat the female part of each population only.
Let \( n_i(x, t) \, dx \) be the number of \( A_i \) individuals aged between \( x \) and \( x + dx \) at time \( t \); the total number of \( A_i \) individuals is \( n_i(t) = \int_0^t n_i(x, t) \, dx \), and the total number of individuals in the system is \( n(t) = \sum_i n_i(t) \). The theory developed on pp. 27–8 tells us that, for sufficiently large \( t \), we can write

\[
\frac{dn_i}{dt} = r_i n_i
\]  

(1.92)

The state of the system is most conveniently represented in terms of the frequencies of the different types, and the total number of individuals \( n \). The frequency of \( A_i \) at time \( t \) is defined as \( p_i(t) = n_i(t)/n(t) \); it is also useful to define the mean intrinsic rate of increase, taken over all types, as

\[
\bar{r}(t) = \sum_i p_i(t) r_i
\]  

(1.93)

Using equations (1.92) and (1.93), we find (with some rearranging) that

\[
\frac{dn}{dt} = \bar{r} n
\]  

(1.94a)

and

\[
\frac{dp_i}{dt} = p_i(r_i - \bar{r})
\]  

(1.94b)

Substituting from equations (1.94) into (1.93), we obtain an expression for the rate of change of \( \bar{r} \):

\[
\frac{d\bar{r}}{dt} = \sum_i p_i (r_i - \bar{r})^2
\]  

(1.95)

This states that the rate of change of the population growth-rate is, for large \( t \), equal to the variance among types in the intrinsic rate of increase. This is the version of Fisher’s Fundamental Theorem of natural selection appropriate for an asexual population (cf. Fisher, 1930); the derivation of a version of this theorem for sexually reproducing populations will be discussed in section 4.2. Equation (1.94b) tells us that, not surprisingly, the type with the highest intrinsic rate of increase tends to displace the others from the population; equation (1.95) implies that this process is accompanied by a simultaneous improvement in the growth-rate of the population as a whole.
The use of reproductive value

The above results are obviously valid only for $t$ taken sufficiently large that each sub-population has approached its stable age-distribution and rate of growth. Results which are valid for arbitrary $t$ can be obtained by applying the concept of reproductive value, which was introduced earlier in the chapter (p. 38). We saw that the rate of increase of the total reproductive value of a population is equal to its intrinsic rate of increase, regardless of the age-distribution. This result can be applied to each sub-population in the present case. We can define the reproductive value, $v_i(x)$, for $A_i$ individuals aged $x$ by using equation (1.61) with the demographic parameters and intrinsic rate of increase of $A_i$. The total reproductive value of the $i$th sub-population at time $t$ is defined as

$$V_i(t) = \int_0^d n_i(x, t)v_i(x) \, dx$$

The total reproductive value of the population is $V = \sum V_i$. Using equation (1.63), it follows that

$$\frac{dV_i}{dt} = r_i V_i$$

(1.96)

Weighting each individual by its reproductive value, we obtain the frequency $\tilde{p}_i$ for $A_i$, using the relation

$$\tilde{p}_i(t) = \int_0^d n_i(x, t)v_i(x) \, dx/V$$

(1.97)

We can also define $\tilde{r}$ as the corresponding mean intrinsic rate of increase,

$$\tilde{r} = \sum \tilde{p}_i r_i.$$

Using the same arguments that were applied to equations (1.92) and (1.93), we obtain

$$\frac{dV}{dt} = \tilde{r} V$$

(1.98a)

$$\frac{d\tilde{p}_i}{dt} = \tilde{p}_i (\tilde{r}_i - \tilde{r})$$

(1.98b)

Similarly

$$\frac{d\tilde{r}}{dt} = \sum \tilde{p}_i (\tilde{r}_i - \tilde{r})^2$$

(1.99)

These equations are of exactly the same form as those derived above, but are valid for all $t$. They resemble the equations which can be derived for continuous-time populations with no age-structure, i.e. in which fecundi-
ties and death-rates are independent of age (cf. Crow and Kimura, 1970, Chapter 1).

1.5.2 *Discrete age-class populations*

Somewhat similar results can be derived for the discrete age-class model. If \( n_i(t) \) is the total number of \( A_i \) individuals at time \( t \) \( (n_i(t) = \sum_{t} n_i(x, t)) \), the results derived in section 1.3.1 imply that, for large \( t \), we obtain the following expression for the change in numbers between \( t \) and \( t - 1 \).

\[
\Delta n_i(t) = n_i(t) - n_i(t - 1) = \lambda_i n_i(t - 1)
\]

where \( \lambda_i \) is the value of \( \lambda_1 \) for \( A_i \) (cf. p. 33). Similarly, we can define the total number of individuals as \( n = \sum n_i \), the frequency of \( A_i \) as \( p_i = n_i/n \), and the mean rate of increase as \( \bar{\lambda} = \sum p_i \lambda_i \), just as for the continuous model. Using the same type of argument as for the continuous-time model, we find that, for large \( t \), we have

\[
\Delta n = (\bar{\lambda} - 1)n
\]

and

\[
\Delta p_i = [p_i(\lambda_i - \bar{\lambda})]/\bar{\lambda}
\]

Similarly,

\[
\bar{\lambda} \Delta \bar{\lambda} = \sum p_i(\lambda_i - \bar{\lambda})^2
\]

These equations are analogous to equations (1.94) and (1.95). If we weight individuals by their reproductive value, we can obtain expressions analogous to equations (1.98) and (1.99). These results resemble those obtainable for an asexual population with discrete generations (cf. Nagylaki, 1992, Chapter 2).

The problem of treating the process of selection in a sexually reproducing population, where individuals can produce offspring whose genotype differs from their own, is considerably more complex than for the asexual case, and forms the subject of Chapters 3 and 4. Before considering this problem, it is desirable to investigate what happens to the genotypic composition of age-structured populations in which there is no selection, i.e. where there is a mixture of genotypes whose demographic parameters are identical. This topic is considered in the next chapter.
2

The genetics of populations without selection

2.1 Introduction

In Chapter 1, the basic concepts of mathematical demography were introduced. We shall start to apply these concepts in the present chapter to the genetics of age-structured populations. The cornerstone of population genetics theory is the Hardy–Weinberg law, which deals with a population which is not exposed to the action of any of the standard evolutionary forces: mutation, migration, selection, non-random mating, or random sampling of genes due to finite population size. In such a population, a single autosomal locus reaches an equilibrium with constant gene frequencies, and in which genotypic frequencies are predicted from the gene frequencies by the familiar Hardy–Weinberg formula (Crow and Kimura, 1970, Chapter 2). This result is conventionally derived for the case of a discrete-generation population, where gene frequencies can be shown to remain constant for all time, under the stated assumptions, and in which Hardy–Weinberg frequencies are reached after one generation. There have been a few attempts to extend this result to more general types of populations. Moran (1962, Chapter 2) gave a treatment of a continuous-time model which assumed constancy of gene frequencies for all time. Charlesworth (1974b) showed that gene frequencies become asymptotically constant, using a continuous-time model; he gave a proof for a discrete age-class model with time-independent demographic parameters in Jacquard (1974, Chapter 7). This case has also been studied by Gregorius (1976).

In the first part of this chapter, we shall be concerned with the analysis of the approach to genetic equilibrium in the absence of evolutionary factors, using both discrete age-class and continuous-time models. The cases of a single autosomal locus, a sex-linked locus and two autosomal loci will each be studied. The general conclusion reached is that constancy of gene frequencies and Hardy–Weinberg genotypic frequencies are approached only asymptotically, unless the gene frequencies of each age and sex group are equal in the initial population. We might, therefore, expect recently-founded populations to experience transient changes in
gene frequencies, which have nothing to do with the conventional forces of evolutionary change.

The other topic of this chapter is the analysis of the consequences of finite population size for evolutionary change in age-structured populations. We shall be particularly concerned with the measurement of effective population number, the parameter which summarises the impact of finite population size on gene frequency change (Wright, 1931, 1938). Numerous publications have been devoted to the definition of effective population number for age-structured populations. Somewhat heuristic treatments have been given by Kimura and Crow (1963), Nei and Imaizumi (1966), Nei (1970), Giesel (1971), and Crow and Kimura (1971). More exact approaches to the problem were introduced by Felsenstein (1971) and Hill (1972), which have been extended by Johnson (1977a, b), Choy and Weir (1978), Hill (1979), Emigh and Pollak (1979), and Nunney (1993). The account in this chapter is based on the treatments of Felsenstein and Johnson. In addition to effective population number, the probability of fixation of a neutral gene in a finite population will also be discussed. The chapter concludes with a discussion of the frequencies of consanguineous matings in an age-structured population, based on the work of Hajnal (1963).

2.2 **Approach to genetic equilibrium**

In this section we shall study the process of approach to genetic equilibrium in a number of genetic and demographic models. It is assumed that the population size is sufficiently large that the effect of random sampling of gamete frequencies can be neglected, and that other evolutionary factors such as selection, mutation and migration are also absent. All the genotypes in the population are assumed to have the same demographic parameters. It is simplest to think of these models as representing populations in which mating is at random with respect to genotype, although they can also be used to describe gene frequency change in inbreeding populations, or in populations in which there is assortative mating of a type which does not lead to genotypic differences in fecundity. We shall consider in turn the cases of a single autosomal locus, a sex-linked locus and two autosomal loci.

2.2.1 **An autosomal locus: the discrete age-class case**

*Construction of the model*

We shall first analyse the case of a population with discrete age-classes, of the type whose demographic properties were examined in
some detail in Chapter 1. Suppose that the population under consideration is segregating at an autosomal locus with \( m \) alleles \( A_1, A_2, \ldots, A_m \). There are \( m^2 \) possible genotypes if we distinguish between heterozygotes which carry the same pair of alleles, according to the maternal or paternal origins of the alleles in question. We shall adopt the convention that the ordered genotype \( A_iA_j \) has allele \( A_i \) of maternal origin, and \( A_j \) of paternal origin.

Consider first the female part of the population. If there are \( d \) age-classes, a complete description of its state at time \( t \) is provided by the set of \( m^2 \) vectors of numbers of females with all possible ordered genotypes, such that the row vector \( n_{ij}(t) = [n_{ij}(1, t), n_{ij}(2, t), \ldots, n_{ij}(d, t)] \) represents the numbers of \( A_iA_j \) females in age-classes 1, 2, \ldots, \( d \). Alternatively, we can write \( p_{ij}(x, t) \) for the frequency of \( A_iA_j \) among females of age \( x \) at time \( t \), and \( n_{ij}(t) = [p_{ij}(1, t), p_{ij}(2, t), \ldots, p_{ij}(d, t)] \) for the vector of frequencies of \( A_iA_j \) across age-classes of females. The vector of numbers of females of all genotypes, \( n(t) = \sum_{ij} n_{ij}(t) \), together with \( m^2 - 1 \) of the frequency vectors can also be used to describe the population. Similar vectors, of dimension \( d^* \), can be defined for the male part of the population. As in Chapter 1, the vectors for males will be distinguished from their female counterparts by asterisks.

An abbreviated description of the system is provided by considering numbers or frequencies of alleles rather than genotypes. Thus, we can define the vector \( n_i(t) = [n_i(1, t), n_i(2, t), \ldots, n_i(d, t)] \) of numbers of copies of allele \( A_i \) (\( i = 1, 2, \ldots, m \)) in the female population; \( p_i(t) \) is the corresponding vector of frequencies of \( A_i \) across female age-classes. We have

\[
n_i(t) = \sum_{j=1}^{m} [n_{ij}(t) + n_{ji}(t)] \quad (2.1a)
\]

\[
p_i(t) = \frac{1}{2} \sum_{j=1}^{m} [p_{ij}(t) + p_{ji}(t)] \quad (2.1b)
\]

Obviously, similar vectors \( n_i^* \) and \( p_i^* \) can be defined for the male population.

It turns out that, much as in the usual discrete-generation case (Crow and Kimura, 1970, Chapter 2), it is most convenient to study the process of genetic change in terms of gene frequencies or gene numbers rather than genotypic frequencies. We shall see below that, since each genotype is assumed to have the same demographic parameters, the gene frequencies must approach constant values which are independent of sex and age. Once these equilibrium gene frequencies have been effectively attained,
the equilibrium genotypic frequencies can be calculated knowing the rules of mating, e.g. with random mating.

Our method, therefore, will be to work directly with the vectors \( n_t \) and \( n_t^* \). The first problem is to relate the values of these vectors for some arbitrary time \( t \) to their values in the preceding time-interval, \( t - 1 \). This can be done by means of a matrix equation similar to that employed in Chapter 1 (section 1.2.1) for describing the dynamics of a genetically homogeneous population. We use the same notation as in that chapter; thus \( P(x, t) \) and \( P^*(x, t) \) are the probabilities of survival over one time-interval of a female and male, respectively, who are aged \( x \) at time \( t \). The net fecundities \( f(x, t) \), \( f^*(x, t) \) and \( f'(x, t) \), as defined by equations (1.3), (1.8) and (1.10), are also needed. These are, respectively, the expected numbers of daughters of a female aged \( x \) at time \( t \), the expected number of sons of a male aged \( x \) at time \( t \), and the expected number of sons of a female aged \( x \) at time \( t \). In addition, we define \( f''(x, t) \) as the expected number of daughters of a male aged \( x \) at time \( t \).

In the notation of Chapter 1, we have

\[
f''(x, t) = M^*(x, t)P(0, t)a^*(x, t) \tag{2.2}
\]

As discussed in Chapter 1 (pp. 6–7), the age-specific fecundities of males are in general dependent on the age-composition of the population in a highly complex way, which is determined by the precise rules of mating with respect to the ages of male and female partners. As will be seen shortly, the exact nature of this dependence is irrelevant; all that we need to know is that each genotype at the locus in question has the same demographic parameters.

We can combine \( n_t \) and \( n_t^* \) into a single \((d + d^*)\)-dimensional row vector \( \tilde{n}_t = [n_t, n_t^*] \), which describes the state of the population with respect to allele \( A_i \) (cf. section 1.2.1). Using the rules of Mendelian transmission, we can write the following matrix equation

\[
\tilde{n}_t(t) = \tilde{n}_t(t - 1)H(t - 1) \tag{2.3a}
\]

The matrix \( H(t - 1) \) in equation (2.3a) can be written in the following partitioned form

\[
H(t - 1) = \begin{bmatrix}
A(t - 1) & B(t - 1) \\
C(t - 1) & D(t - 1)
\end{bmatrix} \tag{2.3b}
\]

where \( A \) is a \( d \times d \) matrix whose only non-zero elements are those of the first column, \( a_{x1} = \frac{1}{2}f(x, t - 1) \), and the off-diagonal elements, \( a_{x,x+1} = P(x, t - 1) \); \( B \) is a \( d \times d^* \) matrix whose only non-zero elements are those
of the first column, \( b_{x1} = \frac{1}{2} f'(x, t - 1) \); \( C \) is a \( d^* \times d \) matrix whose only non-zero elements are those of the first column, \( c_{x1} = \frac{1}{2} f''(x, t - 1) \); \( D \) is a \( d^* \times d^* \) matrix whose only non-zero elements are those of the first column, \( d_{x1} = \frac{1}{2} f^*(x, t - 1) \), and the off-diagonal elements, \( d_{x,x+1} = P^*(x, t - 1) \). \( H \) is therefore a non-negative matrix. The sub-matrices of \( H \) describe the transmission of genes between different classes of individuals, so that \( H \) may be regarded as having the following schematic structure

\[
\begin{bmatrix}
\text{females to females} & \text{females to males} \\
\text{males to females} & \text{males to males}
\end{bmatrix}
\]

The same type of equation can be written for each allele in the population, and hence for the total number of alleles. The elements of the matrix are, from the assumptions which we have made, identical for every allele; only the vectors \( \tilde{n}_i(t) \) differ. We show below that this means that every allele in the system comes to increase in numbers at the same rate (which in general, of course, varies in time), so that the frequencies of all the alleles eventually become constant. To do this, we make use of the weak ergodicity property of non-negative matrices, which we discussed in section 1.4.1. As we saw there, the relative values of the components of a vector representing the state of a system described by an equation such as (2.3a) eventually become independent of the initial state of the population. Two vectors representing two systems (two alleles) which differ only in their initial vectors become strictly proportional to each other. Applied to the present case, this means that the number of copies of a given allele, \( A_i \), in a given age-class becomes proportional to the number of copies of a different allele, \( A_j \), in that age-class, so that their relative frequencies become constant, but in general depend on the initial conditions.

### Analysis of the model

We now develop a proof of the above statements, using the form of the weak ergodicity property derived by Seneta (1973, pp. 69–77). Equation (2.3a) can be used to express \( \tilde{n}_i(t) \) in terms of the initial vector \( \tilde{n}_i(0) \). We have

\[
\tilde{n}_i(t) = \tilde{n}_i(0)H^*(t)
\]

(2.4a)

where

\[
H^*(t) = H(0)H(1) \ldots H(t - 1)
\]

(2.4b)

Seneta has shown that a matrix such as \( H^* \), which is formed by taking the product of a sequence of non-negative matrices with bounded compo-
nents, tends with increasing $t$ to a state in which its rows are proportional to each other. It is assumed in Seneta’s theorem that the matrices whose product composes $H^*$ have a structure such that $H^*$ has no zero elements (i.e., it is strictly positive) when $t$ is sufficiently large. We assume for the moment that this condition is satisfied. We therefore obtain the following type of asymptotic relation between any two elements which belong to the $x$th column of $H^*(t)$

$$h^*_{ux}(t)/h^*_{vx}(t) \sim C_{uv}$$ (2.5)

$C_{uv}$ is a constant which is independent of $x$ and $t$, and depends only on $u$ and $v$ (for a given sequence of matrices composing $H^*(t)$).

Given this result, we can write the following asymptotic expression for the number of copies of $A_i$ in age-class $x$ of the female population at time $t$

$$n_i(x, t) \sim \sum_{u=1}^{d} n_i(u, 0) h^*_{ux}(t) + \sum_{u=1}^{d^*} n_i^*(u, 0) h^*_{d+u,x}(t)$$

or

$$n_i(x, t) \sim h^*_{ix}(t) C_i$$ (2.6a)

where

$$C_i = \sum_{u=1}^{d} n_i(u, 0) C_{u1} + \sum_{u=1}^{d^*} n_i^*(u, 0) C_{d+u,1}$$ (2.6b)

(The $C_{u1}$ are obtained by using equation (2.5) with $v = 1$.) A similar result can be written for the number of copies of $A_i$ in age-class $x$ of the male population

$$n^*_i(x, t) \sim h^*_{i,x}(t) C'_i$$ (2.6c)

The same type of relationship holds for the number of copies of $A_i$ in any other age-class of males or females. The same constant, $C'_i$, appears in all these expressions. $C'_i$ is determined by the initial vector $\tilde{n}_i(0)$, and by the relations between the rows of $H^*$ as expressed by equation (2.5). The matrix-derived terms, of the form $h^*_{i,x}(t)$, are of course different for different age-classes.

The same argument can obviously be applied to each allele in the population, and expressions similar to equations (2.6) can be written down by simply changing the allele subscript $i$. The equations for different alleles thus differ only with respect to the constants $C'_i$ ($i = 1, 2, \ldots, m$), given by equations of the same form as (2.6b). We have the following
asymptotic expression for the frequency of $A_t$ among females of age-class $x$:

$$p_i(x, t) \sim h^{x_t}_i(t)\frac{C_i}{h^{x_t}_i(t)\sum_{j=1}^{m} C'_j} = \frac{C_i}{\sum_{j=1}^{m} C'_j} = \hat{p}_i$$

The same asymptotic frequency, $\hat{p}_i$, clearly holds for each age-class of both the male and female parts of the population, since the $h^{x_t}_i(t)$ terms cancel top and bottom.

**Hardy-Weinberg equilibrium**

The asymptotic allele frequencies, $\hat{p}_i$, ($i = 1, 2, \ldots, m$) are therefore constant, and independent of sex and age-class. This is the desired result. It can be seen from equations (2.5)–(2.7) that any allele which is present initially in at least one reproductive or pre-reproductive age-class of the male or female population must be represented in the final population (note that all the elements of $H^*$ are positive). Furthermore, the gene frequencies will in general converge only gradually towards their asymptotic values; only if the frequencies of each allele are initially the same in each age-class of both the male and female populations will there be no change in gene frequencies over the course of time. This gradual convergence of gene frequencies towards constancy is in contrast to the standard result of population genetics for the corresponding discrete-generation model, where gene frequencies are always constant in the absence of disturbing evolutionary forces (Crow and Kimura, 1970, Chapter 2). If the population mates at random with respect to genotype, it follows from the constancy of the asymptotic allele frequencies that the genotypic frequencies for each sex and age-class are also asymptotically constant, and given by the Hardy-Weinberg formula

$$q_{ij}(x, t) = \hat{q}_{ij}(x, t) = \hat{p}_i \hat{p}_j$$

A random-mating population thus converges to a state of Hardy-Weinberg equilibrium.

**Positivity of $H^*$**

These results are based on the assumption that $H^*(t)$ is a strictly positive matrix for sufficiently large $t$ (p. 46). We now consider the conditions for the validity of this assumption. It is known that such a result holds for the product of a sequence of Leslie matrices, under conditions discussed in section 1.4.1. Now the sub-matrices $A$ and $D$ in equation (2.3b) are both Leslie matrices, describing the female and male.
Approach to genetic equilibrium

parts of the population respectively. It is reasonable to assume that they
meet these conditions if we are describing an iteroparous species (the ages
with non-zero fecundities may, of course, be different in males and
females). The other sub-matrices contribute additional positive elements.
Provided that $b_{d1}$ is always positive, the matrix $H$ in equations (2.3) thus
has the form of a Leslie matrix with some additional positive elements.
Hence, $H^*(t)$ for sufficiently large $t$ must be strictly positive.

Conclusions

These results establish that the conditions for the approach to
Hardy–Weinberg equilibrium in an age-structured population with dis-
crete age-classes are similar to those required for the age-structure of the
population to become independent of its initial state (weak ergodicity).
These conditions are likely to be satisfied for most iteroparous species.
This conclusion applies to populations exposed to environmentally-
induced fluctuations in their demographic parameters, as well as to
populations with time-independent parameters. Furthermore, since we do
not need to specify the functional form of the dependence of the
demographic parameters on time to apply this result, it holds good for
density-dependent populations, despite the fact that the value of the
matrix $H$ in equations (2.3) may depend partly on the initial vectors of
the total numbers of males and females. The rate of convergence to
Hardy–Weinberg equilibrium must be of the same order as the rate of
convergence of the population to its asymptotic state of independence of
the initial conditions, since the equations obeyed by individual alleles are
identical in form to those obeyed by the total number of alleles or the
total number of individuals. As we saw in section 1.3.2, the period of time
required for this convergence is usually a matter of a few ‘generations’.
During the process of convergence, oscillations in gene frequencies may
occur. This conclusion has been confirmed by computer calculations of
population trajectories. An example is shown in Table 2.1.

The possibility of gradual convergence of gene frequencies to con-
stancy, and of genotypic frequencies to Hardy–Weinberg proportions,
implies that populations which have recently been founded by a group of
individuals in which gene frequencies differed according to age and sex
may show significant changes in their genotypic composition in the
short-term, over and above any effects attributable to other evolutionary
factors such as selection or genetic drift. This may be of significance in
relation to studies of the genetics of human isolates, e.g. Jacquard (1974,
Chapter 16).

The convergence of the population to a state of identical, constant
Table 2.1. Convergence to equilibrium for an autosomal locus with two alleles, using demographic data for the French 1830 population. The frequency of $A_1$ was 1 in the first 5 age-classes and 0 in the others.

<table>
<thead>
<tr>
<th>Time $t$</th>
<th>Age-class</th>
<th>Age-structurea</th>
<th>Genotype frequency $\begin{pmatrix} (A_1A_1) \ (A_1A_2) \ (A_2A_2) \end{pmatrix}$</th>
<th>Frequency of $A_1$</th>
</tr>
</thead>
<tbody>
<tr>
<td>2</td>
<td>1</td>
<td>161.9</td>
<td>0.273 0.499 0.299</td>
<td>0.523</td>
</tr>
<tr>
<td></td>
<td>5</td>
<td>0.573</td>
<td>1 0 0</td>
<td>1</td>
</tr>
<tr>
<td></td>
<td>9</td>
<td>0.558</td>
<td>0 0 1</td>
<td>0</td>
</tr>
<tr>
<td>8</td>
<td>1</td>
<td>153.1</td>
<td>0.362 0.479 0.158</td>
<td>0.602</td>
</tr>
<tr>
<td></td>
<td>5</td>
<td>0.620</td>
<td>0.842 0.151 0.007</td>
<td>0.918</td>
</tr>
<tr>
<td></td>
<td>9</td>
<td>0.335</td>
<td>1 0 0</td>
<td>1</td>
</tr>
<tr>
<td>20</td>
<td>1</td>
<td>153.3</td>
<td>0.513 0.406 0.080</td>
<td>0.717</td>
</tr>
<tr>
<td></td>
<td>5</td>
<td>0.650</td>
<td>0.519 0.403 0.078</td>
<td>0.721</td>
</tr>
<tr>
<td></td>
<td>9</td>
<td>0.504</td>
<td>0.625 0.332 0.044</td>
<td>0.790</td>
</tr>
<tr>
<td>40</td>
<td>1</td>
<td>162.9</td>
<td>0.530 0.396 0.074</td>
<td>0.728</td>
</tr>
<tr>
<td></td>
<td>5</td>
<td>0.638</td>
<td>0.531 0.395 0.074</td>
<td>0.729</td>
</tr>
<tr>
<td></td>
<td>9</td>
<td>0.515</td>
<td>0.536 0.392 0.072</td>
<td>0.732</td>
</tr>
<tr>
<td>60</td>
<td>1</td>
<td>172.6</td>
<td>0.531 0.395 0.074</td>
<td>0.729</td>
</tr>
<tr>
<td></td>
<td>5</td>
<td>0.637</td>
<td>0.531 0.395 0.074</td>
<td>0.729</td>
</tr>
<tr>
<td></td>
<td>9</td>
<td>0.516</td>
<td>0.531 0.395 0.074</td>
<td>0.729</td>
</tr>
<tr>
<td>80</td>
<td>1</td>
<td>182.7</td>
<td>0.531 0.395 0.074</td>
<td>0.729</td>
</tr>
<tr>
<td></td>
<td>5</td>
<td>0.638</td>
<td>0.531 0.395 0.074</td>
<td>0.729</td>
</tr>
<tr>
<td></td>
<td>9</td>
<td>0.516</td>
<td>0.531 0.395 0.074</td>
<td>0.729</td>
</tr>
</tbody>
</table>

a The total number of individuals entering age-class 1 at time $t$ is given for age 1, and the numbers entering age-classes 5 and 9 are given as fractions of this number. (After Table 7.4 of Jacquard (1974).)

The genotypic frequencies in each age-class is dependent on the matrix $H$ having a non-periodic structure such that $H^*$ becomes strictly positive (pp. 66–9). The most biologically important case in which this condition is not fulfilled is that of a semelparous species in which the age of reproduction is at 2 or more (cf. Chapter 1, p. 26). In such a case, the different phases of the population form two or more effectively genetically isolated populations coexisting in the same locality; gene frequencies within each phase will be constant, but any initial differences will be preserved indefinitely. A probable example of such a situation has been described in the pink salmon by Aspinwall (1974).

2.2.2 An autosomal locus: the continuous-time case

It is possible to approximate the continuous-time case arbitrarily closely by using the above discrete age-class model with a large number of age-classes. It is instructive, however, to develop an exact continuous-
time model incorporating genetic variation and to use it to establish convergence of a random-mating population to a state of Hardy–Weinberg equilibrium. We shall employ the notation and concepts introduced in section 1.2.3 for the purpose of describing the dynamics of a continuous-time population. We shall assume that the primary sex-ratio, \( a \), is independent of age and time. The reproductive function for females aged \( x \) at time \( t \) is therefore equal to

\[
k(x, t) = l(x, t)M(x, t)a
\]

and the reproductive function for males is

\[
k^*(x, t) = l^*(x, t)M^*(x, t)(1 - a)
\]

These give, respectively, the expected rate of production of female offspring at time \( t \) attributable to a female zygote formed at time \( t - x \), and the expected rate of production of male offspring at time \( t \) attributable to a male zygote formed at time \( t - x \). The corresponding expected rates of production of offspring of both sexes are

\[
K(x, t) = k(x, t)/a
\]

and

\[
K^*(x, t) = k^*(x, t)/(1 - a)
\]

As in the discrete age-class model discussed above, the male fecundities are dependent on the composition of the female population in a complicated way. It is unnecessary for the present purpose to concern ourselves with the nature of this dependence; as previously, all that we need is the fact that all the demographic parameters are the same for each genotype.

It is most convenient to describe the state of the population, as far as total numbers of individuals are concerned, in terms of the rates of production of female and male zygotes by the population at time \( t \), \( B_f(t) \) and \( B_m(t) \), respectively. From the definitions given above and in Chapter 1, we have

\[
B_f(t) = g(t) + \int_0^t B_f(t - x)k(x, t)\,dx
\]

\[
B_m(t) = g^*(t) + \int_0^t B_m(t - x)k^*(x, t)\,dt
\]

where \( g(t) \) and \( g^*(t) \) are terms representing the contributions to \( B_f(t) \) and \( B_m(t) \) from individuals alive at the initial time \( t = 0 \); these terms are zero for \( t > d \) and \( t > d^* \), respectively. Using the assumption of constant sex-ratio \( a \), it is easy to see that the rate of production of zygotes of both
Populations without selection

sexes, $B(t)$, is related to $B_f(t)$ and $B_m(t)$ by the expressions

\[ B_f(t) = aB(t) \quad (2.12a) \]
\[ B_m(t) = (1 - a)B(t) \quad (2.12b) \]

Using these relations and equations (2.11), we obtain the following expressions for $B(t)$

\[ B(t) = \frac{1}{a}g(t) + a\int_0^t B(t - x)K(x, t)\,dx \quad (2.13a) \]
\[ B(t) = \frac{1}{1 - a}g^*(t) + (1 - a)\int_0^t B(t - x)K^*(x, t)\,dx \quad (2.13b) \]

Substituting from equations (2.9) and (2.10) and adding the resulting expressions, we obtain

\[ B(t) = G(t) + \frac{1}{2}\int_0^t B(t - x)[k(x, t) + k^*(x, t)]\,dx \quad (2.14) \]

where $G(t) = \frac{1}{2}[(1/a)g(t) + [1/(1 - a)]g^*(t)]$. This is a renewal equation of the same general form as that which Norton (1928) showed has a weak ergodicity property, under suitable conditions. Two such equations, which differ only in the values of the initial conditions, $G(t)$ tend asymptotically to a state in which the $B(t)$ function for one has a constant ratio to the $B(t)$ function for the other (cf. section 1.4.1). Sufficient conditions for this result to be valid are that $k(x, t)$ and $k^*(x, t)$ be continuous functions of $x$ and $t$, and that the ages of first and last reproduction for males and females, $b, b^*, d$ and $d^*$, be fixed over time. If it can be shown that each allele in the population satisfies an equation of similar form to equation (2.14), differing only in the initial conditions, then Norton's theorem can be applied to prove that allele frequencies tend to constancy as $t$ increases. We shall now proceed to demonstrate that this is indeed the case.

Since each individual receives one copy of an autosomal gene at a locus from its male parent and the other from its female parent, it is essential to distinguish between alleles of maternal and paternal origin. Consider the zygotes produced by the population at the instant $t$. Let $p_i(t)$ and $p^*_i(t)$ be the frequencies of a given allele $A_i$ among maternally-derived and paternally-derived alleles, respectively, in these zygotes. The rates of production at time $t$ of maternally-derived and paternally-derived copies of $A_i$ in new zygotes are thus given by $B(t)p_i(t)$ and $B(t)p^*_i(t)$. (Since we are dealing with the case of an autosomal locus, it is unnecessary to distinguish the sexes when calculating gene frequencies among new
zygotes.) The net frequency of $A_i$ among the zygotes produced at time $t$ is given by

$$\bar{p}_i(t) = \frac{1}{2} [p_i(t) + p_i^*(t)]$$

(2.15a)

and the net rate of production of $A_i$ genes in new zygotes is

$$2B(t)\bar{p}_i(t) = B(t)[p_i(t) + p_i^*(t)]$$

(2.15b)

Using these definitions, it is possible to derive a renewal equation for $B(t)\bar{p}_i(t)$ in a way similar to that which yielded equation (2.14).

Consider first a time $t > d, d^*$, such that all new zygotes entering the population are derived from individuals who were born after the initial time $t = 0$. Using equations (2.9), (2.10) and (2.13), we obtain

$$B(t)p_i(t) = \int_b^d B_m(t - x)\bar{p}_i(t - x)K(x, t) \, dx$$

(2.16a)

Similarly,

$$B(t)p_i^*(t) = \int_{b^*}^{d^*} B_m(t - x)\bar{p}_i(t - x)K^*(x, t) \, dx$$

(2.16b)

For $t \leq d$, equation (2.16a) must be modified by addition of a term $g_i(t)$, to take account of contributions from individuals alive initially. Similarly, for $t \leq d^*$, equation (2.16b) must be modified by the addition of a term $g_i^*(t)$. The exact form of these terms does not concern us here. If we write $G_i(t) = \frac{1}{2}[g_i(t) + g_i^*(t)]$, equations (2.16) can be combined to yield the final equation

$$B(t)\bar{p}_i(t) = G_i(t) + \left[ \int_0^t B(t - x)\bar{p}_i(t - x)[k(x, t) + k^*(x, t)] \, dx \right]$$

(2.17)

This, as anticipated above, has exactly the same form for each allele, which is that of equation (2.14) for the total number of individuals in the population. Only the nature of the initial conditions, $G_i$, differ between alleles. Hence, Norton’s theorem can be applied, and we can conclude that asymptotically the allele frequencies tend to constancy, i.e. for large $t$ we have

$$\bar{p}_i(t) = \bar{p}_i \quad (i = 1, 2, \ldots, m)$$

(2.18)
where the values of the equilibrium frequencies \( \hat{p}_i \) depend on the initial conditions \( G_i \). These frequencies are clearly the same for males and females, from what was said above; similarly, for large \( t \), such that gene frequency change among the zygotes has ceased, the gene frequencies among individuals of any age must also be constant and identical to the frequencies among the zygotes. When this state has been reached, it also follows that there is Hardy–Weinberg equilibrium for the genotypic frequencies among individuals of all ages, assuming that the population mates at random with respect to genotype. This result has been derived under conditions on the time-dependence of the demographic parameters which are almost as non-restrictive as for the discrete age-class model; the assumption of a constant primary sex-ratio has had to be made, however, in order to obtain equations to which Norton's weak ergodicity theorem can be applied. This is not an unduly restrictive condition, since empirical evidence suggests that there are only minor variations in sex-ratio with time and age of parent (cf. section 5.3.6).

### 2.2.3 A sex-linked locus

This case can be treated in a very similar way to an autosomal locus, taking into account the fact that a male (assumed to be the heterogametic sex) receives no genes from his father, whereas females receive one gene from each parent. Only the discrete age-class model will be considered here. A special case has previously been treated by Cornette (1978).

As in the autosomal case, an abbreviated description of the state of the population is given by the vectors of numbers of copies of each allele in females and males, \( n_i(t) \) and \( n^*_i(t) \) (\( i = 1, 2, \ldots, m \)). As before, \( n_i \) and \( n^*_i \) can be combined into one \((d + d^*)\)-dimensional vector \( \tilde{n}_i = [n_i, n^*_i] \) for the purposes of calculation. We can write

\[
\tilde{n}_i(t) = \tilde{n}_i(t - 1)\tilde{H}(t - 1)
\]

where

\[
\tilde{H}(t - 1) = \begin{bmatrix}
\tilde{A}(t - 1) & \tilde{B}(t - 1) \\
\tilde{C}(t - 1) & \tilde{D}(t - 1)
\end{bmatrix}
\]

(2.19b)

in which \( \tilde{A} \) is a \( d \times d \) matrix whose only non-zero elements are those of the first column, \( \tilde{a}_{x1} = \frac{1}{2}f'(x, t - 1) \), and the off-diagonal elements \( \tilde{a}_{x,r+1} = P(x, t - 1) \); \( \tilde{B} \) is a \( d \times d^* \) matrix whose only non-zero elements are those of the first column, \( \tilde{b}_{x1} = \frac{1}{2}f'(x, t - 1) \); \( \tilde{C} \) is a \( d^* \times d \) matrix whose only non-zero elements are those of the first column, \( \tilde{c}_{x1} = \frac{1}{2}f'(x, t - 1) \); and \( \tilde{D} \) is a \( d^* \times d^* \) matrix whose only non-zero elements are those of the first column, \( \tilde{d}_{x1} = \frac{1}{2}f'(x, t - 1) \).
f^*(x, t - 1); \widetilde{D} is a $d^* \times d^*$ matrix whose only non-zero elements are those of the off-diagonal, $\widetilde{a}_{x,x+1} = \widetilde{P}^*(x, t - 1)$.

$\widetilde{H}$ is of a somewhat different form from $H$, the matrix for the autosomal case. As in that case, however, it is equivalent to a Leslie matrix with some additional positive elements, provided that $\widetilde{b}_{d1}$ is positive. The same type of argument that was used for an autosomal locus can therefore be applied to this case to prove that the frequencies of $A_j$ among males and females of each age-class approach the same constant value $\hat{p}_j$, at about the same rate as the age-structure of the whole population approaches a state in which it is independent of the initial conditions. The conditions under which this is true are the same as for an autosomal locus. Since the allele frequencies in males and females become the same, the frequency of the ordered genotype $A_jA_j$ in females is given by an equation of the same form as equation (2.8). The genotypic frequencies in males are, of course, identical with the corresponding allele frequencies. These results are similar to those for the standard discrete-generation model of a sex-linked locus in a random-mating population (Crow and Kimura, 1970, Chapter 2). There may be some difference in the mode of approach to equilibrium, however. In the discrete generation case, there is a damped oscillation if male and female gene frequencies are initially different. With age-structure, such an oscillation would tend to be disguised by differences between age-classes; only if all age-classes of males in the initial population differed in gene frequencies from all age-classes of females in the same direction, would one expect oscillations as marked as in the discrete generation case (cf. Cornette, 1978).

2.2.4 Two autosomal loci

We shall now consider the case of two autosomal loci. In the discrete-generation case, it is known that the gamete frequencies for such a pair of loci eventually approach values corresponding to random combinations of alleles at the two loci, i.e. the frequency of a particular gamete in the population approaches asymptotically the product of the frequencies of the alleles which it contains. This was first proved by Robbins (1918). A similar result can be proved for the case of an age-structured population, by means of a generalisation of a method introduced by Malécot (1948) for discrete-generation populations. An account of Malécot's results is given by Crow and Kimura (1970, Chapter 2). Random mating with respect to genotype is assumed.

The problem is most conveniently treated by means of a difference equation formulation of the discrete age-class model (cf. Chapter 1, pp.
The continuous-time case follows from this straightforwardly, by increasing the number of age-classes without limit. Consider two autosomal loci \( A \) and \( B \), with alleles \( A_i (i = 1, 2, \ldots, m) \) and \( B_k (k = 1, 2, \ldots, n) \). The genetic state of the system in a given time-interval \( t \) can be characterised by the frequencies of the \( nm \) possible gametes, \( A_iB_1, A_1B_2, \ldots, A_mB_n \), among the zygotes born in time-interval \( t \). Let \( \pi_{ik}(t) \) be the frequency of gametes containing alleles \( A_i \) and \( B_k \) among the gametes of maternal origin in these zygotes. Let the corresponding frequency for paternally-derived gametes be \( \pi_{ik}'(t) \). The overall frequency of \( A_iB_k \) gametes in the zygotes is thus given by

\[
\pi_{ik}(t) = \frac{1}{2} [\pi_{ik}(t) + \pi_{ik}'(t)]
\]

As we have seen, the frequencies of alleles at autosomal loci tend to constancy with increasing time, under the conditions discussed in section 2.2.1. Let the equilibrium frequencies be \( \hat{p}_i \) for \( A_i \) and \( \hat{q}_k \) for \( B_k \). We shall assume from now on that the population has effectively reached such a state of constant gene frequencies. Since we are concerned here only with establishing that gamete frequencies approach those predicted from random combination of alleles, there is no loss in generality in making this assumption. If the alleles were combined at random into gametes, the frequency of \( A_iB_k \) gametes would be equal to \( \hat{p}_i\hat{q}_k \). The deviation from random combination can be measured by the linkage disequilibrium parameter \( D_{ik} \), defined by the equation

\[
D_{ik}(t) = \pi_{ik}(t) - \hat{p}_i\hat{q}_k
\]

We shall show that this parameter approaches zero with increasing \( t \).

We proceed in a somewhat similar manner to the analysis of the case of a single autosomal locus with continuous time (section 2.2.2); in particular, we make the assumption that the primary sex-ratio \( a \) is constant, so that the total number of zygotes born at time \( t, B(t) \), is related to the numbers of male and female zygotes, \( B_m(t) \) and \( B_f(t) \), by equations of the same form as equations (2.12). This enables us to write the following equation for \( B(t) \) (assuming \( t \geq d, d^* \), so that no zygotes are contributed by individuals alive at the initial time \( t = 0 \))

\[
B(t) = \sum_{x=b}^{d} B(t - x)k(x, t) = \sum_{x=b^*}^{d^*} B(t - x)k^*(x, t)
\]

Let the recombination fraction for loci \( A \) and \( B \) be \( c \) in females and \( c^* \) in males. For simplicity, we shall assume that these are independent of age, although it is not difficult to generalise the following argument to allow
for age-dependence of recombination fractions. Consider the set of gametes produced by females aged $x$ at a time $t > d$, $d^*$. There is a probability $1 - c$ that a gamete drawn at random from this set is non-recombinant, in which case there is a probability $\bar{n}_{ik}(t - x)$ that it is $A_iB_k$ in constitution. There is a probability $c$ that it is recombinant in origin, in which case there is a probability $\hat{p}_i\hat{q}_k$ of its being $A_iB_k$. The net frequency of $A_iB_k$ among these gametes is therefore equal to

$$(1 - c)\bar{n}_{ik}(t - x) + c\hat{p}_i\hat{q}_k$$

Summing over all ages, and using the assumption of constant sex-ratio, we thus obtain the relation

$$B(t)\pi_{ik}(t) = \sum_{x=b}^{d} B(t - x)[(1 - c)\bar{n}_{ik}(t - x) + c\hat{p}_i\hat{q}_k]k(x, t)$$

(2.22a)

A similar argument can be applied to paternally derived gametes, and yields the equation

$$B(t)\pi_{ik}^*(t) = \sum_{x=b^*}^{d^*} B(t - x)[(1 - c^*)\bar{n}_{ik}(t - x) + c^*\hat{p}_i\hat{q}_k]k^*(x, t)$$

(2.22b)

By using equation (2.21), subtracting $B(t)\hat{p}_i\hat{q}_k$ from both sides of equations (2.22), and adding the resulting equations, we obtain the following expression

$$B(t)D_{ik}(t) = \frac{1}{2}(1 - c)\sum_{x=b}^{d} B(t - x)D_{ik}(t - x)k(x, t)$$

$$+ \frac{1}{2}(1 - c^*)\sum_{x=b^*}^{d^*} B(t - x)D_{ik}(t - x)k^*(x, t)$$

(2.23)

Choose a value of $t$, $t_0$ say, which can be made arbitrarily large. Let $\bar{b}$ be the smaller of $b$ and $b^*$, and $\bar{d}$ be the larger of $d$ and $d^*$. Let $|D_{ik}(t_1)|$ be the maximum value of $|D_{ik}(t)|$ in the interval $[t_0 - \bar{d}, t_0 - \bar{b}]$. Equation (2.23) gives the inequality

$$B(t_0)|D_{ik}(t_0)| \leq \frac{1}{2}(1 - c)|D_{ik}(t_1)|\sum_{x=b}^{d} B(t_0 - x)k(x, t_0)$$

$$+ \frac{1}{2}(1 - c^*)|D_{ik}(t_1)|\sum_{x=b^*}^{d^*} B(t_0 - x)k^*(x, t_0)$$
Using equations (2.21), this reduces to
\[
|D_{ik}(t_0)| \leq [1 - \frac{1}{2}(c + c^*)]|D_{ik}(t_1)|
\] (2.24)

\[|D_{ik}(t_1)| \text{ can similarly be related to the maximum value of } |D_{ik}(t)|, \]
\[|D_{ik}(t_2)|, \text{ in the interval } [t_1 - \tilde{d}, t_1 - \tilde{d}], \text{ and so on. Continuing this back}
\]
l steps to some arbitrary time \( t_l (t_l \geq \tilde{d}) \), we have
\[
|D_{ik}(t_0)| \leq [1 - \frac{1}{2}(c + c^*)]|D_{ik}(t_l)|
\] (2.25)

If \( t_0 \) is taken sufficiently large, \( |D_{ik}(t_0)| \) can be made arbitrarily small, if \( c \) or \( c^* > 0 \), since \( |D_{ik}(t_l)| \) is bounded (its maximum possible value is \( \frac{1}{4} \)), and
\[
\frac{1}{2} \leq 1 - \frac{1}{2}(c + c^*) < 1.
\]

This proves that the frequency of \( A_iB_k \) averaged over maternal and paternal gametes, \( \bar{\pi}_{ik} \), approaches a constant value, \( \bar{\hat{p}}_i\hat{q}_k \), which corresponds to random combination of the alleles concerned. From equations (2.22) it follows that the frequencies of \( A_iB_k \) in maternal and paternal gametes, \( \pi_{ik} \) and \( \pi_{ik}^* \), also approach this limiting value. The same obviously applies to each gamete in the population. We may expect the average length of the time-interval between successive \( t_i \) to be of the same order as one of the generation time measures introduced in section 1.3.2; hence the rate of decay of the \( D \) parameters given by equation (2.25) is such that each \( D \) is reduced by a factor of approximately \( 1 - \frac{1}{2}(c + c^*) \) every 'generation'. This corresponds to the standard result for discrete-generation populations (Crow and Kimura, 1970, Chapter 2).

### 2.3 The effects of finite population size

#### 2.3.1 General considerations

In this section we shall study the process of random changes in gene frequencies due to finite population size (genetic drift), in the context of an age-structured population. We assume that other evolutionary forces, such as selection, mutation, migration and non-random mating, are absent. The only factor tending to perturb gene frequencies from the values which they would attain in an infinitely large population is the effect of random sampling of alleles between successive time-intervals. The traditional discrete-generation models of population genetics have dealt with this problem by means either of a measure of the rate of increase in the inbreeding coefficient (the probability that the two genes carried at an autosomal locus of a diploid organism are identical by descent), or by examining the rate of increase in the variance of the probability distribution of gene frequencies generated by random drift.
Crow and Kimura (1970, Chapter 7) review this topic, with reference to a number of classes of discrete-generation populations.

In many cases, the asymptotic rates of increase in both these measures are governed by similar equations, at least in to a good approximation. If this is so, the rate of genetic drift can be summarised by a single parameter, the effective population number, $N_e$. This concept was introduced by Wright (1931, 1938). For example, let the inbreeding coefficient in generation $t$ be $F(t)$. For sufficiently large $t$, it is found with discrete-generation models that we can write the approximate equation

$$F(t) \approx \frac{1}{2}N_e + \left(1 - \frac{1}{2}N_e\right)F(t - 1) \quad (2.26a)$$

Similarly, if there are two alleles $A_1$ and $A_2$ with expected frequencies $\hat{p}_1$ and $\hat{p}_2 = 1 - \hat{p}_1$, the variance in the frequency of $A_1$, $V(t)$, asymptotically satisfies the expression

$$V(t) \approx \hat{p}_1\hat{p}_2 \frac{1}{2N_e} + \left(1 - \frac{1}{2N_e}\right)V(t - 1) \quad (2.26b)$$

These relations would be satisfied for all values of $t$ in a population with the simplest type of structure: a random-mating but self-fertile hermaphrodite species with a constant adult population size $N$, and a distribution of number of offspring per adult which is binomial with a mean unity. In such an ideal population, we simply have $N_e = N$. In discrete-generation populations with a more complicated breeding structure, relations such as (2.26) may be obeyed only asymptotically, and $N_e$ may differ from the census number of breeding adults in a way which is calculable from the breeding structure of the population. One difficulty with the use of effective number as a measure of the rate of genetic drift is that the value of $N_e$ in equation (2.26a) is sometimes found to differ from that in equation (2.26b). In such cases, it is necessary to distinguish between inbreeding effective number and variance effective number (Kimura and Crow, 1963).

We shall accordingly concentrate on trying to define a measure of effective population number for the case of an age-structured population. As mentioned in the introduction to this chapter, this problem has been tackled by a number of authors. The approach used here is based on that of Felsenstein (1971), as modified by Johnson (1977b) and Emigh and Pollak (1979) for the case of a two-sex population. Some results on probabilities of fixation of genes in a finite population will also be given. The final section deals with the problem of calculating the frequencies of consanguineous matings in an isolated population, a question of some importance in human genetics.
2.3.2 Effective population number with age-structure

Construction of the model

We shall study in detail only the case of a stationary diploid, discrete age-class population with two sexes and with time-independent demographic parameters for both males and females. Furthermore, we make the somewhat artificial assumption that the numbers of males and females in each age-class are fixed over time. Survival from age-class \( x - 1 \) to \( x \) (\( x > 1 \)) is thus assumed to be random with respect to genotype, but there is interdependence between the deaths of different individuals such that constant numbers \( n(x) \) and \( n^*(x) \) of females and males are maintained in age-class \( x \). Similarly, although new-born individuals are assumed to be formed by combining gametes sampled at random from infinite pools of male and female gametes, births take place in such a way as to preserve constant numbers of males and females in the first age-class. This model is reasonably realistic for a fairly large population whose size is regulated by density-dependent factors, but is obviously unrealistic for very small populations, whose size and age-structure must themselves be subject to stochastic fluctuations. A further assumption which is needed is that mating is at random with respect to age and genotype, i.e. the age and genotype of the father of an individual are independent of the age and genotype of its mother.

We shall mostly be concerned with studying the process of genetic drift in terms of inbreeding. To do this, it is necessary to generalise the concept of inbreeding coefficient in the following way. We introduce a square symmetric matrix of order \( d + d^* \), \( F(t) \), which describes the state of the system at time \( t \). The \( xy \)th element of \( F(t) \), \( f_{xy}(t) \) (\( 1 \leq x \leq d \), \( 1 \leq y \leq d^* \)), is the probability of identity of two genes sampled at random (with replacement) from females of age-classes \( x \) and \( y \) at time \( t \) (for brevity, the term ‘identity’ will be used here as equivalent to identity by descent). Similarly, \( f_{x,d+y}(t) \) (\( 1 \leq x \leq d \), \( 1 \leq y \leq d^* \)) is the probability of identity of a gene sampled from a female aged \( x \) with one from a male aged \( y \) at time \( t \); \( f_{d+x,d+y}(t) \) (\( 1 \leq x \leq d^* \), \( 1 \leq y \leq d^* \)) is the probability of identity of two genes sampled at random from males aged \( x \) and \( y \) at time \( t \). In order to avoid clumsiness of notation, whenever it is unnecessary to specify the sex of the individuals involved, the \( ij \)th element of \( F \) is referred to as the probability of identity of a gene sampled from an individual of ‘class’ \( i \) with a gene sampled from an individual of ‘class’ \( j \) (\( 1 \leq i \leq d + d^* \), \( 1 \leq j \leq d + d^* \)). Note that \( f_{ii} \) is non-zero even for an outbred population, since two genes sampled from the same class have a chance of being the same.
Analysis of the model

The problem therefore reduces to one of determining a workable expression for $F(t)$ in terms of $F(t-1)$. In order to do this, it is useful to introduce a further matrix, $G$, which describes the flow of genes between different classes of individuals in successive time-intervals in a somewhat different way from the matrix $H$ used earlier (section 2.2.1). In general terms, the $ij$th element of $G$, $g_{ij}$, is defined as the probability that a gene of an individual of class $j$ is derived from an individual of class $i$ of the previous time-interval. Matrices of this type were introduced into population genetics by Hill (1972, 1974).

The elements of $G$ can be derived straightforwardly, using the assumptions of constancy of the demographic parameters, population size and age-structure, and the fact that each individual has one maternal and one paternal allele at an autosomal locus. Using the fecundity parameters discussed earlier in this chapter (section 2.2.1), but omitting any dependence on time, we have

$$g_{x1} = \frac{n(x)f(x)}{2\sum n(y)f(y)} \quad (x = 1, 2, \ldots, d) \quad (2.27a)$$

$$g_{d+x,1} = \frac{n^*(x)f''(x)}{2\sum n^*(y)f''(y)} \quad (x = 1, 2, \ldots, d*) \quad (2.27b)$$

These define, respectively, the probabilities that a randomly-chosen gene of a female of age-class 1 is derived from a female or male aged $x$ in the preceding time-interval. Similar probabilities can be defined for a gene of a male of age-class 1

$$g_{x,d+1} = \frac{n(x)f'(x)}{2\sum n(y)f'(y)} \quad (x = 1, 2, \ldots, d) \quad (2.27c)$$

$$g_{d+x,d+1} = \frac{n^*(x)f^*(x)}{2\sum n^*(y)f^*(y)} \quad (x = 1, 2, \ldots, d*) \quad (2.27d)$$

The other elements of $G$ may be found as follows. Clearly, a female of age-class $x$ ($x > 1$) receives her genes from a female belonging to age-class $x - 1$ in the preceding time-interval with probability 1, and similarly for males. Hence, we have

$$g_{x-1,x} = 1 \quad (x = 2, \ldots, d) \quad (2.27e)$$

$$g_{d+x-1,d+x} = 1 \quad (x = 2, \ldots, d*) \quad (2.27f)$$

All other elements of $G$ are zero.

If we assume that the primary sex-ratio is constant and independent of
parental age, these expressions can be simplified as follows (cf. Chapter 1, pp. 10-11, 25-6)

\[ g_{x1} = g_{x,d+1} = \frac{1}{2} k(x) \quad (1 \leq x \leq d) \quad (2.28a) \]

\[ g_{d+x,1} = g_{d+x,d+1} = \frac{1}{2} k^*(x) \quad (1 \leq x \leq d^*) \quad (2.28b) \]

\( G \) is similar in general structure to the \( H \) matrix used earlier in the chapter, and may similarly be regarded as being composed of four sub-matrices, each of which describes the flow of genes between a particular sub-set of the population (cf. p. 66). Unlike \( H \), however, it is constant in time, and its elements describe probabilities of transmission, not net numbers of genes transmitted. For future purposes, it is useful to note that \( G \) is the transpose of a \textit{stochastic matrix}, since the elements in each column sum to unity. From the standard results concerning stochastic matrices (e.g. Jacquard, 1974, Appendix 2), we can therefore conclude that \( G \) has a leading eigenvalue of unity, to which there corresponds a row eigenvector whose elements are all equal. An associated column eigenvector, \( q^T \), can be found by the method used for defining reproductive value (equation (1.53)). We have

\[ q(x) = \sum_{y=x}^{d} (g_{y1} + g_{y,d+1}) \quad (x = 1, 2, \ldots, d) \quad (2.29a) \]

\[ q(d+x) = \sum_{y=x}^{d^*} (g_{d+y,1} + g_{d+y,d+1}) \quad (x = 1, 2, \ldots, d^*) \quad (2.29b) \]

A measure of the generation time of the population, analogous to the measures defined in section 1.3.2, is given by

\[ \bar{T} = \frac{1}{2} \sum_{x=1}^{d+d^*} q(x) \quad (2.30) \]

If primary sex-ratio is constant and independent of parental age, we can use equations (2.28) to obtain the simplified expressions

\[ q(x) = \sum_{y=x}^{d} k(y) \quad (x = 1, 2, \ldots, d) \quad (2.31a) \]

\[ q(d+x) = \sum_{y=x}^{d^*} k^*(y) \quad (x = 1, 2, \ldots, d^*) \quad (2.31b) \]
The effects of finite population size

\[ \tilde{T} = \frac{1}{2} \sum_{x=1}^{d} xk(x) + \frac{1}{2} \sum_{x=1}^{d^*} xk^*(x) \]  

(2.31c)

In this case, \( \tilde{T} \) is equal to the mean over the female and male populations of the generation time \( T \) defined on p. 30 (note that we are assuming a constant population size).

Using these relations, we can write an asymptotic expression for \( G' \), using the same methods that yielded equation (1.55a). For large \( t \), we have

\[ G' \sim A = q^{T1/2}\tilde{T} \]  

(2.32)

where \( 1 \) is a \((d + d^*)\)-dimensional row vector, whose elements are all equal to unity. The validity of the asymptotic expression (2.32) is guaranteed if the age-specific fecundities for males and females are both non-periodic, since \( G \) then has the structure of a non-periodic Leslie matrix with additional positive elements (cf. equation (1.6)).

We are now in a position to use these definitions to obtain an expression for \( F(t) \). It is in practice somewhat more convenient to work with the probabilities \( 1 - f_{ij}(t) \) of non-identity between two genes sampled at random from individuals of class \( i \) and class \( j \) at time \( t \). (If we define \( J = 1^T1 \) as the square matrix of order \( d + d^* \) whose elements are all unity, we obtain the matrix of these probabilities of non-identity as \( J - F(t) \).) Consider firstly \( 1 - f_{11}(t) \), the probability of non-identity between two genes sampled at random from females of age-class 1 at time \( t \). In order for these genes to be non-identical, the same gene must not have been sampled twice from the same female. There is a probability of \( 1/2n(1) \) that the two genes are separate but come from the same female, in which case there is a probability \( 4g_{x1}g_{d+y,1} \) that they come from a female aged \( x \) and a male aged \( y \) at time \( t - 1 \), with probability of non-identity \( 1 - f_{x,d+y}(t - 1) \). There is a probability \( 1 - 1/n(1) \) that the two genes are sampled from two different females; if this is so, there is a probability \( g_{ii}g_{jj} \) that they are derived from individuals belonging to classes \( i \) and \( j \) at time \( t - 1 \), with probability of non-identity \( 1 - f_{ij}(t - 1) \). We thus obtain the recurrence relation

\[ 1 - f_{11}(t) = \frac{2}{n(1)} \sum_{x=1}^{d} \sum_{y=1}^{d^*} g_{x1}g_{d+y,1}[1 - f_{x,d+y}(t - 1)] \]

\[ + \left[ 1 - \frac{1}{n(1)} \right] \sum_{i,j=1}^{d+d^*} g_{ii}g_{jj}[1 - f_{ij}(t - 1)] \]  

(2.33a)
A similar expression can be derived for genes sampled from males of age-class 1:

\[ 1 - f_{d+1,d+1}(t) \]

\[ = \left[ \frac{2}{n^*(1)} \right] \sum_{x=1}^{d} \sum_{y=1}^{d^*} g_{x,d+1}g_{d+y,d+1}[1 - f_{x,d+y}(t-1)] \]

\[ + \left[ \frac{1}{n^*(1)} \right] \sum_{i,j=1}^{d+d^*} g_{i,d+1}g_{j,d+1}[1 - f_{i,j}(t-1)] \]  \hspace{1cm} (2.33b)

An expression for \( 1 - f_{xx}(t) \) (\( 1 < x \leq d \)), the probability of non-identity between two genes sampled from females of age-class \( x \) (\( x > 1 \)), can be obtained as follows. The probability that the two genes sampled are separate genes is \( 1 - 1/2n(x) \), since the chance that the same gene is sampled twice is \( 1/2n(x) \). If they are separate genes, they must have been derived from separate genes of females aged \( x-1 \) at time \( t-1 \), with probability of non-identity \( [1 - f_{x-1,x-1}(t-1)]/[1 - 1/2n(x-1)] \). We thus have

\[ 1 - f_{xx}(t) = [1 - f_{x-1,x-1}(t-1)] \]

\[ \times \frac{[1 - 1/2n(x)]}{[1 - 1/2n(x-1)]} \quad (x = 2, 3, \ldots, d) \]  \hspace{1cm} (2.33c)

Similarly,

\[ 1 - f_{d+x,d+x}(t) = [1 - f_{d+x-1,d+x-1}(t-1)] \]

\[ \times \frac{[1 - 1/2n^*(x)]}{[1 - 1/2n^*(x-1)]} \quad (x = 2, 3, \ldots, d^*) \]  \hspace{1cm} (2.33d)

For a pair of genes sampled from a male and a female both aged 1 at time \( t \), we have

\[ 1 - f_{1,d+1}(t) = \sum_{i,j=1}^{d+d^*} g_{i1}g_{j,d+1}[1 - f_{i,j}(t-1)] \]  \hspace{1cm} (2.33e)

For a female aged 1 and a male aged \( x \) (\( x > 1 \)), we have

\[ 1 - f_{1,d+x}(t) = \sum_{i=1}^{d+d^*} g_{i1}[1 - f_{i,d+x-1}(t-1)] \quad (x = 2, 3, \ldots, d^*) \]  \hspace{1cm} (2.33f)

Similarly, for a female aged \( x \) and a male aged 1, we have
The effects of finite population size

\[ 1 - f_{x,d+1}(t) = \sum_{i=1}^{d+d^*} g_{i,d+1}[1 - f_{i,x-1}(t-1)] \quad (x = 1, 2, \ldots, d) \]  

(2.33g)

For individuals in all other classes, we have

\[ 1 - f_{ij}(t) = 1 - f_{i-1,j-1}(t) \]  

(2.33h)

Equations (2.33) provide us with a set of linear recurrence relations which connect the set of probabilities of non-identity at time \( t \) with the set at time \( t - 1 \), in a form which is extremely suitable for numerical calculations. From general theory (cf. section 1.3.1), we expect that asymptotically these probabilities will change with time with a rate determined by the leading eigenvalue, \( \lambda_0 \), of the matrix corresponding to the recurrence relations. The elements of this matrix are easily deducible from equations (2.33). When this asymptotic state has effectively been attained, we can write

\[ 1 - f_{ij}(t) \sim \lambda_0[1 - f_{ij}(t - 1)] \]

or

\[ f_{ij}(t) \sim (1 - \lambda_0) + \lambda_0 f_{ij}(t - 1) \quad (i, j = 1, 2, \ldots, d + d^*) \]  

(2.34a)

Noting that the 'generation time' in the present case is defined by equations (2.30) and (2.31c) as \( \bar{T} \), and comparing equation (2.34a) with equation (2.26a), we can define effective population size by the relation

\[ 1 - \lambda_0 = 1/2\bar{T}N_e \]  

(2.34b)

\( (1 - \lambda_0) \bar{T} \) predicts the asymptotic per generation increase in the \( f_{ij} \). The problem therefore reduces to one of finding a useful approximate formula for \( \lambda_0 \).

This can be done if we assume that the population size is sufficiently large that second-order terms in \( 1/n(x) \) and \( 1/n^*(x) \) can be neglected, and that the asymptotic state represented by equation (2.34a) is approached while the \( f_{ij} \) are still small enough that second-order terms in the \( f_{ij} \) and terms such as \( f_{ij}/n(x) \) are negligible.† If these assumptions are

† The assumption about the rate of convergence to the asymptotic state has been found to be valid whenever computations of specific examples have been carried out (Felsenstein, 1971; Johnson, 1977b; Choy and Weir, 1978). Furthermore, an identical expression for \( N_e \) to that obtained here can be derived by the alternative methods of Felsenstein (1971) and Emigh and Pollak (1979). If this assumption breaks down in a particular case, the calculations which follow still provide a valid approximate method for determining the rates of change of the \( f_{ij} \) in the early generations of inbreeding.
made, we can approximate equations (2.33a and b) for low values of the
\( f_{ij} \) coefficients by the expressions

\[
 f_{11}(t) \equiv \frac{1}{2n(1)} + \sum_{i,j=1}^{d+d^*} g_{ij} g_{ji} f_i(t - 1)
\]

(2.35a)

\[
 f_{d+1,d+1}(t) \equiv \frac{1}{2n^*(1)} + \sum_{i,j=1}^{d+d^*} g_{i,d+1} g_{j,d+1} f_i(t - 1)
\]

(2.35b)

(Note that \( \sum_{x=1}^{d} g_{x1} = \sum_{y=1}^{d^*} g_{d+y,1} = \frac{1}{2} \), etc.)

Similarly, equations (2.33c and d) can be approximated for low \( f_{ij} \) by

\[
 f_{xx}(t) \equiv \frac{1}{2n(x)} - \frac{1}{2n(x - 1)} + f_{x-1,x-1}(t - 1)
\]

\((x = 2, 3, \ldots, d)\) (2.35c)

\[
 f_{d+x,d+x}(t) \equiv \frac{1}{2n^*(x)} - \frac{1}{2n^*(x - 1)} + f_{d+x-1,d+x-1}(t - 1)
\]

\((x = 2, 3, \ldots, d^*)\) (2.35d)

The remaining members of equations (2.33) can be rewritten in terms of
the \( f_{ij} \) coefficients alone; in each case a relationship of exactly the same
form is preserved merely by substituting \( f_{ij} \) for \( 1 - f_{ij} \) on both sides of the
equation. Combining these equations (2.35), we obtain the single matrix
equation

\[
 F(t) = G^T F(t - 1) G + Y
\]

(2.36)

where \( Y \) is a diagonal matrix with non-zero elements \( y_{11} = 1/2n(1) \),
\( y_{xx} = 1/2n(x) - 1/2n(x - 1) \) \((x = 2, 3, \ldots, d)\), \( y_{d+1,d+1} = 1/2n^*(1) \),
\( y_{d+x,d+x} = 1/2n^*(x) - 1/2n^*(x - 1) \) \((x = 2, 3, \ldots, d^*)\). Equation (2.36)
gives the relation

\[
 F(t) = (G^T)^t Y G \sum_{u=0}^{t-1} (G^T)^u Y G^u
\]

where \( F(0) \) is the initial value of the \( F \) matrix. (If we assume that all
individuals in the initial population are unrelated, \( F(0) \) is a diagonal
matrix with non-zero elements \( f_{xx}(0) = 1/2n(x) \) \((x = 1, 2, \ldots, d)\) and
\( f_{d+x,d+x}(0) = 1/2n^*(x) \) \((x = 1, 2, \ldots, d^*)\).)

This gives

\[
 F(t) - F(t - 1) = (G^T)^t Y G - (G^T)^{t-1} F(0) G^{t-1} + (G^T)^{t-1} Y G^{t-1}
\]

Substituting from equation (2.32), we obtain

\[
 F(t) - F(t - 1) = A^T YA
\]
The effects of finite population size

i.e.

\[ F(t) - F(t - 1) = \frac{(qYq^T)J}{4T^2} \]  
(2.37)

where \( J \) is the matrix \( 1^T1 \) (p. 83).

The effective population number

Equation (2.37) implies that each \( f_{ij} \) coefficient experiences the same increment per time-interval, equal to the scalar \( (qYq^T)/(4T^2) \). This can be compared with the expression for the change in \( f_{ij} \) given by equations (2.34). Provided that the population size is sufficiently large that equation (2.34a) is a good approximation when the \( f_{ij} \) are still small, equations (2.34) can be approximated by

\[ f_{ij}(t) - f_{ij}(t - 1) \approx \frac{1}{2N_e \hat{T}} \]  
(2.38)

A comparison of this with equation (2.37) suggests that the definition of effective population number for the age-structured case should be \( 1/N_e = (qYq^T)/2\hat{T} \). Collecting terms in \( qYq^T \), and using equations (2.29), we can write

\[ \frac{1}{N_e} = \frac{1}{4\hat{T}} \left\{ \frac{1}{n(1)} + \frac{1}{n^*(1)} + \sum_{x=2}^{d} q^2(x) \left[ \frac{1}{n(x)} - \frac{1}{n(x - 1)} \right] \right\} 
+ \sum_{x=2}^{d} q^2(d + x) \left[ \frac{1}{n^*(x)} - \frac{1}{n^*(x - 1)} \right] \]  
(2.39a)

If the primary sex-ratio is independent of parental age and equal to one-half, this expression can be simplified to the following

\[ \frac{1}{N_e} = \frac{1}{2\hat{T}B} \left\{ \frac{1}{l(1)} + \frac{1}{l^*(1)} + \sum_{x=2}^{d} q^2(x) \left[ \frac{1}{l(x)} - \frac{1}{l(x - 1)} \right] \right\} 
+ \sum_{x=2}^{d^*} q^2(d + x) \left[ \frac{1}{l^*(x)} - \frac{1}{l^*(x - 1)} \right] \]  
(2.39b)

where \( B \) is the total number of zygotes produced by the population in each time-interval (cf. Chapter 1, p. 10).

A very similar formula can be derived for the case of an hermaphrodite or monoecious species (Felsenstein, 1971; Johnson, 1977b). If the demographic parameters, etc. are defined as for the female population in the two-sex case, except that fecundities are not corrected for the sex of
offspring (cf. Chapter 1, p. 7) we obtain

\[ \frac{1}{N_e} = \frac{1}{\bar{T}B} \left\{ \frac{1}{l(1)} + \sum_{x=2}^{d} q^2(x) \left[ \frac{1}{l(x)} - \frac{1}{l(x-1)} \right] \right\} \] (2.40)

This is obviously equivalent to equation (2.39b) for the case when male and female demographic parameters are exactly equal.

Equations (2.39) and (2.40) are very useful for displaying the effects of demographic structure on population size. It is immediately obvious that when only individuals of age-class 1 are capable of reproduction, equation (2.39a) reduces to

\[ \frac{1}{N_e} = \frac{1}{4N_f} + \frac{1}{4N_m} \]

where \( N_f \) and \( N_m \) are the numbers of breeding females and breeding males, respectively \( (N_f = n(1) \) and \( N_m = n^*(1)) \). This is of course identical with the standard formula for effective number in the discrete-generation case (Crow and Kimura 1970, Chapter 7). Similarly, for a hermaphrodite species equation (2.40) gives \( N_e \) equal to the census number of breeding adults when only age-class 1 individuals are capable of reproduction. If there are no deaths at all \( (l(x) = l^*(x) = 1 \) for all \( x \), \( N_e \) from equation (2.39b) is equal to \( \bar{T}B \), the total number of individuals born into the population over a period of one generation (note that this assumes a primary sex-ratio of one-half). If there are deaths of individuals only between conception and reaching an age of unity, so that \( l(x) = l(1) \) and \( l^*(x) = l^*(1) \) for all \( x > 1 \), then \( N_e = 4\bar{T}/[1/n(1) + 1/n^*(1)] \), i.e., \( N_e \) is equal to the harmonic mean of the numbers of females and males in age-class 1 over a period of one generation. It can be seen from equations (2.39) that deaths of individuals in later age-classes reduce \( N_e \) below these values, as would be expected intuitively.

Hill’s (1972, 1979) formulae show that the terms in braces in equations (2.39) and (2.40) are equivalent to the reciprocals of terms involving variances and covariances of life-time reproductive success (Johnson, 1977b). Nunney (1993) has used Hill’s results to examine the effects of different systems of mating on effective population number.

**Examples of the calculation of effective number**

The use of these formulae for \( N_e \) can be illustrated with some of the sets of vital statistics used in Chapter 1 (Tables 1.2 and 1.3). In order to simplify the calculations, it has been assumed that the demographic parameters of males and females are equal, so that equation (2.40) can be used. All the essential features of age-structured populations are pre-
The effects of finite population size

served in this formula. In order to use the data of Chapter 1, which apply

to populations with positive intrinsic rates of increase, for the purpose of
calculating \( N_e \) from a formula which assumes a stationary population, it is

necessary to scale down the age-specific fecundities by dividing them by

the net reproduction rate for the population in question, so that the

modified age-specific fecundities correspond to those for a stationary

population (cf. Felsenstein, 1971). If this is done with the human popula-

tion data of Table 1.2, we find that

\[
N_e = \frac{BT}{1.03}
\]

where \( B \) is the number of zygotes born into the population in each
time-interval, and \( \tilde{T} \) is equal to 5.31 time-intervals (a time-interval in this
case corresponds to five years). \( N_e \) in this case is therefore almost exactly
equal to the number of zygotes which enter the population over the

course of one generation, which reflects the low mortality of the popula-

tion in question. It is also interesting to compare \( N_e \) with \( N_b \), the total

number of breeding adults. \( N_b \) can be calculated from \( B \) by the formula

\[
N_b = B \sum_{x=b}^{d} l(x)
\]

(2.41)

In the present case, taking \( b = 2 \) and \( d = 10 \), we find \( N_b = 8.63B \), so that

the effective population number is only 0.60 of the number of individuals

in the reproductive age-classes.

A similar calculation can be done for the squirrel data given in Table

1.3. In this case, the length of a time-interval is one year. We find that

\[
N_e = \frac{b\tilde{T}}{7.82}
\]

where \( \tilde{T} = 2.79 \). In this case, we have \( N_b = 0.60B \), so that \( \frac{N_e}{N_b} = 0.59 \),
a value very similar to that for the human population. The great reduction

in \( N_e \) below the number of zygotes entering the population per generation

is due to the high mortality of the squirrels, particularly in the first

year of life.

Nunney (1993) has examined the relation between \( N_e \) and \( N_b \) for a

variety of different systems of mating, and concludes that \( N_e \) is frequently

approximately equal to one-half \( N_b \).

The inbreeding coefficient of an age-structured population

Up to now we have discussed the problem solely in terms of the

matrix of probabilities of identity, \( F \). It is possible, however, to construct

a measure similar to the discrete-generation inbreeding coefficient (John-

son, 1977b). It is necessary in general to define inbreeding coefficients for
males and females separately, although they become equal if the primary sex-ratio is independent of parental age. The inbreeding coefficient for the females at time $t$, $F(t)$, is defined by analogy with the discrete-generation case as the probability of identity of the maternal and paternal genes of a female in age-class 1. We have

$$F(t) = 4 \sum_{x=1}^{d} \sum_{y=1}^{d^*} g_{x,y} f_{x,y}(t - 1)$$

(2.42a)

Similarly, for males we have the inbreeding coefficient

$$F^*(t) = 4 \sum_{x=1}^{d} \sum_{y=1}^{d^*} g_{x+y} f_{x,y}(t - 1)$$

(2.42b)

In practical applications, such as animal breeding problems, it is relatively simple to calculate these inbreeding coefficients for any desired time, by combining equations (2.42) with the results of iterating the exact recurrence relations (2.33), rather than by using the approximate asymptotic equation (2.37), or the estimate of $F(t)$ and $F^*(t)$ of $(t - 1)/(2N_eT)$ suggested by equations (2.42) and (2.38).

**Variance effective number**

In certain discrete-generation models, the effective number appropriate for calculating the rate of inbreeding differs from that for the rate of increase in variance in gene frequency (p. 79). It is therefore important to determine whether or not there is any difference in the present case. Johnson (1977a) has obtained a formula for variance effective number with age-structure which is identical with equations (2.39); the variance effective number given by Hill (1972, 1979) can also be shown to be the same as that of equations (2.39), for the sampling scheme assumed here (Johnson, 1977b). Felsenstein (1971) showed that the inbreeding and variance effective numbers are the same for the present type of model in the case of an hermaphrodite species. His argument can easily be extended to the two-sex case. We can conclude that the variance effective number is identical with the inbreeding effective number in the present case.

**Other models**

The results described above have been derived on the basis of a number of specialising assumptions, notably that the size and age-structure of the population are constant, and mating is at random with
The effects of finite population size

respect to age and genotype. The sampling scheme assumed for the formation of new zygotes implies that each member of a given age-class has an equal chance of contributing to the new-born population (i.e., a binomial distribution of family size within age-classes). Some of these assumptions can be relaxed. Felsenstein (1971) gives a treatment of variance effective number for the case of an hermaphrodite population which is growing in size but has a stable age-structure and constant population growth-rate. Hill (1972, 1979) gives a treatment of variance effective number which is capable of dealing with non-binomial family size distributions (see also Johnson (1977a, b)). Choy and Weir (1978) give recurrence relations for the \( f_j \) coefficients which can be used for computations with non-binomial family size distributions. They also give recurrence relationships for digenic identity probabilities.

2.3.3 The probability of fixation of a gene

General considerations

In this section, we shall be concerned with the problem of calculating the probability of fixation of a neutral autosomal gene with an arbitrary initial frequency in an age-structured population of finite size. In the discrete-generation case, the following argument can be used. The formula for the inbreeding coefficient in an arbitrary generation, equation (2.26a), implies that as time increases, \( F \) tends to unity, so that all gene copies at a locus are eventually descended with probability unity from one particular gene present in the initial generation. Let the frequency of given allele \( A_i \) be \( p_i(0) \) among the females of the initial generation, and \( p^*_i(0) \) among the males. With stochastic independence of the offspring distributions for different individuals, the expected frequency of \( A_i \) in one generation is equal to the actual population gene frequency of the preceding generation, in the absence of mutation, selection, etc. This implies that the expected frequency of \( A_i \) in any generation in a finite population is equal to its frequency in an infinite population with the same initial composition, i.e. to \( \frac{1}{2}p_i(0) + \frac{1}{2}p^*_i(0) \). Since the final state of the population is such that \( A_i \) is either fixed or lost, the probability of fixation of \( A_i \) is thus \( \frac{1}{2}p_i(0) + \frac{1}{2}p^*_i(0) \).

This type of argument can be extended fairly easily to the case of an age-structured population, using the finite population model developed above. The assumption of random mating with respect to age, is however, unnecessary in this context. The treatment below is based on Emigh and Pollak (1979) and Pollak (1982). (See also Emigh (1979a, b).)
The calculation of the fixation probability

The definitions of section 2.2.1 can be extended by introducing a \(d\)-dimensional vector of expected frequencies of \(A_i\) among females of the various age-classes at time \(t\), \(X^f_i(t)\); a corresponding \(d^*\)-dimensional vector, \(X^f_i(t)\), can be defined for males. As with the gene frequencies themselves, these can be combined into a single \((d + d^*)\)-dimensional vector \(\tilde{X}_i(t) = [X_i(t), X^f_i(t)]\). Using the result that, with binomial sampling, the expected frequency of a gene in a given age-class must be equal to the frequency in an infinite population with the same initial state, it is easily shown that

\[
\tilde{X}_i(t) = \tilde{X}_i(t - 1)G
\]

where \(G\) is the matrix of transmission probabilities defined on p. 81. Iterating this equation, we obtain

\[
\tilde{X}_i(t) = \tilde{X}_i(0)G^t
\]

where \(\tilde{X}_i(0)\) is the initial value of \(X(t)\), such that

\[
\tilde{X}_i(0) = [p_i(0), p^*_i(0)]
\]

Using equation (2.32), equation (2.43b) can be rewritten for large \(t\)

\[
\tilde{X}_i(t) \sim \tilde{X}_i(0)(q^T 1)/(2\tilde{T})
\]

The elements of \(\tilde{X}_i(t)\) are therefore asymptotically all equal to the same constant value, corresponding to the result derived earlier (section 2.2.1) for gene frequencies in the infinite population model with a general demographic structure. This value is the probability of fixation of \(A_i\), which we can symbolise by \(U_i\). Equation (2.45) yields the expression

\[
U_i = \left[ \sum_{x=1}^{d} p_i(x,0)q(x) + \sum_{x=1}^{d^*} p^*_i(x,0)q(d + x) \right]/2\tilde{T}
\]

A case of particular interest for evolutionary theory is when the allele in question is a new mutant, and is therefore represented initially as a single copy among the zygotes of time \(t = 0\). The probability of survival of a new mutant, \(U\), can be found from equation (2.46) as follows. If the mutant individual is female, the gene has an initial frequency of \(p_i(1,0) = 1/2B_f\), where \(B_f\) is the total number of female zygotes produced per time-interval. For \(x > 1\), we have \(p_i(x,0) = 0\). Hence, for a new mutant among the female part of the population we have

\[
U = 1/4B_f\tilde{T}
\]
The effects of finite population size

For a new mutant in the male part of the population, we find

\[ U = \frac{1}{4} B_m \bar{T} \]  

(2.47b)

where \( B_m \) is the number of male zygotes produced per time-interval. These probabilities are equal only when the primary sex-ratio is one-half.

The rate of gene substitution

These results can be used to calculate the rate of substitution of neutral mutations in evolution (Pollak, 1982). The standard method of doing this for a discrete-generation model (Kimura, 1968) assumes that there is an infinite number of possible alleles at a locus, which are selectively equivalent (neutral alleles), and that each time a mutation to such an allele occurs in a population, the allele has never previously been represented in the population. If the population has a fixed size \( N \), and the rate of occurrence of neutral mutations at the locus in question is \( u \) per generation, then \( 2Nu \) new mutations appear in the population each generation, on average. But an individual gene present in the population in any particular generation has a probability of \( 1/2N \) of being fixed by chance. Hence, the expected number of new mutations which enter the population each generation and which eventually displace all the existing alleles at the locus is

\[ K = \frac{2Nu}{2N} = u \]  

(2.48)

\( K \) is the rate of gene substitution at the locus in question. If the state of the locus is compared at times \( t_0 \) and \( t_1 \), where \( t_1 - t_0 \) is a large number of generations, the expected number of neutral allele substitutions that have occurred is equal to \( K(t_1 - t_0) \). This result plays an important role in the interpretation of data on molecular evolution (Kimura, 1983; Gillespie, 1991).

In order to extend equation (2.48) to age-structured populations, we need to introduce an appropriate definition of mutation rate. Consider an allele at a given locus in a new zygote. As the individual develops, its germ-line cells will be exposed to the risk of mutation. Let \( u(x) \) be the probability that a germ cell in a female aged \( x \) contains a mutation to a different allele. Let \( u^*(x) \) be the corresponding probability for a male germ cell. The mutation rate \( u(t) \) for time-interval \( t \) is defined as the probability that a gamete involved in a successful fertilisation in this time-interval contains a mutant allele. In general, we have (for \( t \geq d \))

\[ 2B(t)u(t) = \sum_{x=1}^{d} B(t - x)[k(x)u(x) + k^*(x)u^*(x)] \]  

(2.49a)
With a constant population size, as assumed here, the mutation rate will be independent of time, and we have

\[ u = \frac{1}{2} \sum_x \left[ k(x)u(x) + k^*(x)u^*(x) \right] \]  

(2.49b)

This definition of mutation rate can be used to determine the rate of molecular evolution as follows. New mutations entering the population at time \( t \) will be passed to age-class 1 at time \( t + 1 \). From equation (2.46), the chance of fixation of a single such mutation present in age-class 1 is the product of \( 1/2 \bar{T} \) and its initial frequency. The initial frequency of a mutation in female zygotes is \( \frac{1}{2}B_f \), and there are \( 2B_fu \) mutations per time interval. Mutations transmitted to females thus contribute a total of \( u/2\bar{T} \) mutations which are destined to be fixed. Males contribute an equal quantity. The rate of substitution, in terms of mutations fixed per time-interval, is therefore

\[ K = \frac{u}{\bar{T}} \]  

(2.50)

It follows that the rate of substitution on a time-scale of generations is the same as that given by equation (2.48).

### 2.3.4 The frequencies of consanguineous matings in an isolated population

**General considerations**

One of the classic problems of human genetics is that of finding the expected frequencies of consanguineous matings in an isolated population. This problem obviously has a close relationship with that of calculating the effective number of a population. It can be stated as follows. Consider an isolated human population of fixed size, and assume that mating is random, so that each individual in the population has the same chance of marrying a given person of the opposite sex. What is the expected frequency of occurrence of consanguineous marriages of a particular degree, e.g. first-cousin marriages? The first attempt to solve this problem was made by Dahlberg (1929, 1948), using a discrete-generation formulation. His approach can be illustrated with the case of first-cousins. Each marriage is assumed to produce two children that survive to maturity, so that the population size is constant. Assuming a sex-ratio of one-half, the number of cousins available to a man of marriageable age can be found as follows. His father had one sibling who
married and produced two children. On average, one of these is female, so that the man in question has an expectation of one marriageable female cousin on his father’s side. Similarly, he has an expectation of one on his mother’s side, giving a total expectation of two marriageable female cousins. If the number of breeding adults is $N$, there is a total of $\frac{1}{2}N$ girls of marriageable age, so that the probability of the man’s marrying a first-cousin is $2\sqrt{N} = 4/N$. If the frequency of first-cousin marriages is known, the population size $N$ can be determined. A similar calculation for uncle–niece and aunt–nephew marriages gives a frequency of $2/N$. These calculations are somewhat artificial, as they assume that there is a fixed family size of two; more realistic models, which allow for a probability distribution of family size, can readily be developed (Frota Pessoa, 1957; Jacquard, 1974, pp. 197–202).

The assumption of discrete generations, which underlies Dahlberg’s model and its later generalisations, is inadequate when dealing with data on the human population. The most serious defect of the model is that it leaves out the fact that the probability of marriage is dependent on the ages of the individuals concerned, men being on average older than their wives. The number of potential spouses of a given degree of consanguinity available to a male may therefore differ from the number available to a female. For example, uncle–niece marriages are more common than aunt–nephew marriages, simply because the probability of marriage between a woman and a much older man is higher than the probability of a marriage between a man and a much older woman. On the discrete-generation model, the frequencies of uncle–niece and aunt–nephew marriages should, however, be equal. More subtle effects can arise with marriages of a lower degree of consanguinity. For instance, first-cousin marriages can be divided into four classes, according to the sexes of the ancestors of the partners (Figure 2.1). Data on the frequencies of these classes of marriage in several populations are given in Table 2.2. It can be seen that matrilateral cross-cousin marriages are more frequent than patrilateral cross-cousin marriages, and that maternal parallel marriages

![Figure 2.1. The four kinds of first-cousin. (Squares and circles represent males and females respectively).](image-url)
Table 2.2. *Observed frequencies of the four kinds of first-cousin marriages (relative to a frequency of 100 for all first-cousin marriages)*

<table>
<thead>
<tr>
<th></th>
<th>Parallel</th>
<th>Cross</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Germany (1890–1935)</td>
<td>21</td>
<td>23</td>
</tr>
<tr>
<td>Austria (1901–1931)</td>
<td>18</td>
<td>21</td>
</tr>
<tr>
<td>Italy (1851–1957)</td>
<td>22</td>
<td>21</td>
</tr>
<tr>
<td>Japan (1948–1952)</td>
<td>22</td>
<td>18</td>
</tr>
</tbody>
</table>

After Table 26 of Hajnal (1963)

are more frequent than paternal parallel marriages. Again, the discrete-generation model predicts equal frequencies of each of these classes. The excess frequency of matrilateral cross-cousin marriages has a similar explanation to that for uncle–niece marriages. Mothers are, on average, younger at the birth of their offspring than fathers, so that a man will tend to have matrilateral cross-cousins who are younger than himself, and hence available for marriage, whereas girls who are his patrilateral cross-cousins will tend to be older than himself. The excess frequency of maternal parallel marriages is harder to explain, and requires an analysis based on the marriage model developed below. This model provides a general solution to the problem of the frequencies of consanguineous mating in an age-structured population, and was formulated by Hajnal (1963, 1976) (see also Cavalli-Sforza, Kimura and Barrai, 1966). The following account is based on that of Hajnal (1963).

*Assumptions of the marriage model*

We shall assume that the population is closed, so that no individuals enter or leave. Age and time are treated as continuous variables, as is realistic for human populations. The population is assumed to be stationary in size, and to have time-independent demographic parameters. The primary sex-ratio, $a$, is assumed to be independent of parental age (cf. section 2.2.2). Mating is at random within the population, subject to the restriction that the probability of marriage between a given couple is a known function of their ages. Specifically, we shall assume that the chance of eventual marriage between two individuals selected at birth depends only on the interval between their dates of birth.
Let the time of birth of a given boy be subtracted from that of a given girl, and write \( z \) for the resulting difference (\( z \) is measured in years, and may take negative as well as positive values). The probability that the couple will eventually marry is thus a function of \( z \). This provides a realistic model of the marriage process. Finally, it is assumed that there are no correlations between relatives with respect to their demographic parameters. This is clearly unrealistic, but is the best that can be done in the absence of adequate information on such correlations. Some of these assumptions can be relaxed to a certain extent, as will be discussed below (p. 103).

Consider consanguineous marriages of a given class \( \xi \), specified by a pedigree diagram of the type shown in Figures 2.1 and 2.2. The class \( \xi \) is specified by the sequence in which males and females occur in the lines of descent, as well as by the number of steps by which the couple in question trace their descent from a single ancestral couple.† The different categories of first-cousin marriages shown in Figure 2.1 thus correspond to different values of \( \xi \). The problem is to calculate \( Y_\xi \), the average annual number of marriages of class \( \xi \) in an isolated population. We can write

\[
Y_\xi = AN_\xi \Pi_\xi
\]

(2.51)

where \( A \) is the total number of marriages per year; \( N_\xi \) is the expected number of potential marriages of class \( \xi \) among the descendants of a couple (i.e. the average number of pairs of males and females related in the manner \( \xi \) among the descendants of a couple); \( \Pi_\xi \) is the chance that a potential marriage between such a pair actually takes place.

**The calculation of \( N_\xi \)**

We first examine the relatively straight-forward problem of calculating \( N_\xi \). Consider, for example, the case of paternal parallel cousins. Let a pair of brothers produce during their lifetimes \( n_1 \) girl and \( n_1^* \) boy babies, and \( n_2 \) girl and \( n_2^* \) boy babies respectively. The total number of potential marriages among their children is thus \( n_1 n_2^* + n_1^* n_2 \). Taking expectations, and using the above assumption of no correlations between the demographic parameters of relatives, we obtain

\[
E[n_1 n_2^*] = E[n_1^* n_2] = E[n_1]E[n_2^*]
\]

(2.52a)

Using the assumptions that the primary sex-ratio, \( a \), is independent of age...
and that the population is stationary, we have

\[ E[n^2] = \int_{b^*}^{d^*} k^*(x) \, dx = 1 \]

\[ E[n_1] = \frac{a}{1 - a} \int_{b^*}^{d^*} k^*(x) \, dx = \frac{a}{1 - a} \]

Hence

\[ E[n_1 n^2] + E[n_1 n_2] = \frac{2a}{1 - a} \tag{2.52b} \]

If a marriage produces \( j \) children, the number of possible pairs of brothers is expected to be \((1 - a)^2 j(j - 1)/2\), so that the expected number of potential paternal parallel cousin marriages among the descendents of a marriage with \( j \) children is, using equation (2.52b), \( a(1 - a)j(j - 1) \). If \( \pi_j \) is the chance that a marriage produces \( j \) live births, and if there is no correlation in demographic parameters between parents and offspring, we have

\[ N_{\xi} = a(1 - a) \sum_{j=2}^{\infty} j(j - 1)\pi_j \tag{2.53a} \]

The quantity \( \sum_{j=2}^{\infty} j(j - 1)\pi_j \) depends solely on the mean, \( \mu \), and variance, \( V \), of the distribution of numbers of births per marriage. We have

\[ \sum_{j=2}^{\infty} j(j - 1)\pi_j = \sum_{j=0}^{\infty} j(j - 1)\pi_j = V + \mu^2 - \mu \]

Substituting into equation (2.52a), we obtain

\[ N_{\xi} = a(1 - a)(V + \mu^2 - \mu) \tag{2.53b} \]

A similar calculation can be carried out for any class of consanguineous marriage. It is found that \( N_{\xi} \) is independent of \( \xi \), so that equation (2.53b) can be used for any class of marriage.

**The calculation of \( \Pi_{\xi} \)**

We now consider the calculation of \( \Pi_{\xi} \). Suppose that we select a new-born girl and a new-born boy who have relationship \( \xi \). Let \( h_{\xi}(z) \, dz \) be the probability that the girl is born between \( z \) and \( z + dz \) years after the boy, and \( \phi(z) \) be the probability of eventual marriage with a given girl born \( z \) years after the boy. We thus have

\[ \Pi_{\xi} = \int_{-\infty}^{\infty} \phi(z) h_{\xi}(z) \, dz \tag{2.54} \]
A formula for $\phi(z)$ can be found as follows. We assume that the population is sufficiently large that stochastic fluctuations in the birth-rate can be neglected. Since the population is stationary, the birth-rate $B$ is constant. Over a short time-interval $dt$, $aBdt$ girls and $(1-a)Bdt$ boys will be born. The numbers of pairs which can be formed among male births occurring between times $t$ and $t+dt$, and female births occurring between $z$ and $z+dz$ later, is thus $a(1-a)B^2dtdz$. Furthermore, in a stationary population the number of marriages occurring over a short time-interval $dt$ is equal to the number contracted over their entire life-time by the males born in any time-interval of the same length. Let this number be $A dt$, where $A$ is the rate of marriage. It follows that

$$A = a(1-a)B^2\int_{-\infty}^{\infty} \phi(z) dz$$

(2.55a)

so that

$$\int_{-\infty}^{\infty} \phi(z) dz = A/a(1-a)B^2$$

(2.55b)

Let the proportion of marriages at any time where the woman is $z$ to $z+dz$ years younger than the man be $\theta(z)dz$. $\theta(z)$ is in principle determinable from census data on the frequencies of marriages classified by the ages of the partners. Clearly, $\theta(z) = \phi(z)/\int_{-\infty}^{\infty} \phi(z) dz$. Let $C = A/B$ be the ratio of the rate of marriage to the rate of birth. Equations (2.55) thus yield

$$\phi(z) = \theta(z)C^2/a(1-a)A$$

(2.56)

The calculation of $Y_\xi$

Equation (2.56) can be combined with equations (2.51), (2.53b) and (2.54) to obtain the expression

$$Y_\xi = C^2(V + \mu^2 - \mu)\int_{-\infty}^{\infty} \theta(z)h_\xi(z) dz$$

(2.57)

$Y_\xi$ is thus independent of the absolute birth-rate and hence of the population size. The quantities $V$, $\mu$, $c$ and $\theta(z)$ can be determined from empirical data. It thus remains only to find an expression for $h_\xi(z)$ in terms of observable quantities. This can be achieved as follows. For concreteness, consider the relationship determined by the pedigree in Figure 2.2. The difference in age between the individuals of interest, $J$ and $I$, is given by $z = (t + x_1 + x_2) - (t + y + x_3) = x_1 + x_2 - (y + x_3)$. In order to calculate the probability density of $z$, $h_{\xi}(z)$, it is convenient to assume that the ages $x_1$, $x_2$ and $x_3$ of the ancestors 1, 2 and 3 of $I$ and $J$
are independently and normally distributed, as is the difference $y$ between the dates of birth of the sibs 1 and 3. Consider a set of infants born at the same instant. Let $T$ and $V_T$ be the mean and variance of the ages of their mothers, and $T^*$ and $V_{T^*}$ be the corresponding parameters for their fathers. Since the population is assumed to be stationary, we have (cf. p. 30)

$$T = \int_b^d x k(x) \, dx, \quad V_T = \int_b^d (x - T)^2 k(x) \, dx$$  \hspace{1cm} (2.58a)$$

$$T^* = \int_{b^*}^{d^*} x k^*(x) \, dx, \quad V_{T^*} = \int_{b^*}^{d^*} (x - T^*)^2 k^*(x) \, dx$$  \hspace{1cm} (2.58b)$$

The means and variances of $x_2$ and $x_3$ are $T$ and $V_T$ respectively, while the mean and variance of $x_1$ are $T^*$ and $V_{T^*}$ respectively. The mean of $y$ is equal to the mean difference between the ages of a randomly chosen pair of sibs, and is therefore zero. The variance of $y$, $V_y$, is equal to the variance of the difference in age between randomly chosen pairs of sibs. Hence, in this example the mean and variance of $z$ are equal to

$$\mu_z = T + T^* - T = T^*$$  \hspace{1cm} (2.59a)$$

$$V_z = 2V_T + V_{T^*} + V_y$$  \hspace{1cm} (2.59b)$$

More generally, let there be $\alpha$ males and $\beta$ females in the ancestry of the boy involved in the relationship $\xi$. (The original ancestral couple are not counted as ancestors for this purpose.) The mean and variance of $z$
The effects of finite population size are given by

\[ H = (\alpha_T - \alpha_f)T^* + (\beta_T - \beta_f)T \tag{2.60a} \]

\[ V_T^* = (\alpha_T + \alpha_f)V_T^* + (\beta_T + \beta_f)V_T + V_y \tag{2.60b} \]

The probability density function \( h_\xi(z) \) is thus given by

\[ h_\xi(z) = \frac{1}{\sqrt{2\pi V_\xi}} \exp \left[ -\frac{(z - \mu_\xi)^2}{2V_\xi} \right] \tag{2.61} \]

If we assume that the difference between the ages of spouses is also normally distributed with mean \( \mu_a \) and variance \( V_a \), we have

\[ \theta(z) = \frac{1}{\sqrt{2\pi V_a}} \exp \left[ -\frac{(z - \mu_a)^2}{2V_a} \right] \tag{2.62} \]

Combining this with equation (2.60), substituting into equation (2.56) and evaluating the integral, we obtain the following expression for \( Y_\xi \).

\[ Y_\xi = \frac{C^2(V + \mu^2 - \mu)}{\sqrt{2\pi S_\xi}} \exp \left[ -\frac{\Delta_\xi^2}{2S_\xi} \right] \tag{2.63} \]

where \( \Delta_\xi = \mu_\xi - \mu_a \) and \( S_\xi = V_\xi + V_a \).

Applications of the model

The application of equation (2.63) to specific populations requires estimates of the relevant means and variances. Human populations which are effectively isolated are generally examples of high-fertility, high-mortality populations, so that we need data on pre-industrial societies for these estimates. Hajnal (1963) suggests values of \( T = 25 \), \( V_T = 40 \), \( T^* = 30 \) and \( V_T^* = 70 \) for such societies. (Note that fathers have a considerably higher mean and variance than mothers.) Values of \( \mu_a = 5 \) and \( V_a = 40 \) are also suggested by Hajnal, who used data on an eighteenth-century French population to obtain \( V + \mu^2 - \mu = 20 \) and \( V_y = 55 \). \( C \) is the ratio of marriages to births, so that \( 2C \) is approximately equal to the proportion of people who survive to marriageable age. \( C \) may be taken as 0.25 for present purposes.

Armed with these figures, it is easy to calculate the expected rate of consanguineous marriages of any kind. Table 2.3 shows the values of the components of \( Y_\xi \) for the four classes of first-cousin marriage, from the above formulae, and Table 2.4 shows the relative frequencies of these marriages as a function of the mean difference in age between spouses. With \( \mu_a = 5 \), the sum of the \( Y_\xi \) is equal to 0.14; smaller values of \( \mu_a \) give a slightly bigger, and larger ones a slightly smaller, value. Thus in an
isolated population with high fertility, there would be between 1 and 1.5 first-cousin marriages per decade. This agrees well with data on the frequencies of first-cousin marriages in some nineteenth century European isolates (Hajnal, 1963). Uncle–niece and aunt–nephew marriages taken together would be expected to occur at about 0.06 of this rate, marriages between first-cousins once removed at 0.6, and second cousins at 3.0, times this rate.

The reason for this low frequency of paternal parallel first-cousin marriages is clear from inspection of Table 2.3. Although the term arising from the mean difference between the ages of spouses is the same for paternal parallel and maternal parallel marriages, both ancestors in the paternal parallel pedigree are males, whereas both are female in the maternal parallel pedigree. The variance term $S_x$ is therefore bigger in the former case; since $Y_x$ decreases with $S_x$, there is a corresponding deficit in the frequency of marriages between paternal parallel first cousins. This result may be viewed intuitively as follows. There is greater variation in the age of fathers than that of mothers, so that paternal parallel cousins are more likely to be born far apart in time, and hence are less likely to marry each other. Conversely, maternal parallel cousins are the most likely of the four classes to be born close in time; this
explains why maternal parallel cousin marriages are the most common, except when \( \mu_a \) is so big that its effect on \( \Delta_\xi \) overcomes the differences due to \( S_\xi \) (cf. Tables 2.3 and 2.4).

Comparison of Tables 2.2 and 2.4 shows that the theory seems to predict the qualitative pattern of the differences in frequencies of classes of first cousin marriages. Furthermore, Japan and Italy are known to have larger values of \( \mu_a \) than Germany and Austria, so that the differences between the top and bottom of Table 2.2 accord well with the differences between the rows of Table 2.4. The differences between the theoretical frequencies do not seem as big as the actual ones. Hajnal (1963) discusses some possible reasons for these discrepancies.

Conclusions

The results derived here show that both the absolute rates and relative frequencies of consanguineous marriages in isolated populations are highly dependent on several demographic parameters. Changes in demography, therefore, may themselves produce shifts in marriage patterns. For example, a reduction in mortality would result in an increase in \( C \), the ratio of the marriage rate to the birth rate, leading to an increase in both the absolute rate and frequency of consanguineous marriages (since \( Y_\xi \) is proportional to \( C^2 \)). The model predicts a constant annual rate of consanguineous marriages in a stationary population, independent of the population size, which seems to agree well with observations on human populations. Hajnal (1963) has also investigated the effects of population growth and migration, and Cavalli-Sforza et al. (1966) have carried out an elaborate investigation of the consequences of migration. Although some quantitative differences are introduced by these factors, the general pattern of results is not affected very much. The relative frequencies of different classes of consanguineous marriages are expected to be similar to those for a stationary population, unless the rate of population growth is very high, but the frequency of consanguineous marriages as a whole will decrease as the population size increases. The effect of migration is to lower the rate of consanguineous marriages; if male and female migration rates differ, the pattern of consanguineous marriages may be affected. Since males often have higher migration rates than females in man, one might expect migration to lower the frequency of marriages that involve a relatively large number of male ancestors (e.g. between paternal parallel cousins), and to raise the frequency of marriages involving a relatively large number of female ancestors (e.g. between maternal parallel cousins). This may help to explain the disagreements between Tables 2.2 and 2.4.
Overall, data on the frequencies of consanguineous marriages in human populations suggest that the assumption of random mating is close to the truth (Hajnal, 1963), so that people do not appear to select their spouses on the basis of relationship (except for the avoidance of incest). It is interesting to note that phenomena such as the relative infrequency of paternal parallel cousin marriages have sometimes been interpreted as evidence for such selection. Analyses of this sort can, of course, in principle be carried out for non-human populations, although data of a suitable sort are rarely available. Hajnal (1976) has analysed data on an English population of great tits, and shown that the observed frequencies of consanguineous marriages are consistent with random mating.

Finally, it is useful to note how these results can be used to estimate the size of isolates, if we have data on populations where it can be assumed that no migration occurs between random-mating units. If we know that there are $A_{\xi}$ marriages of class $\xi$ per year in the region of interest, but we expect $Y_{\xi}$ in an isolated population, the number of isolates is given by $A_{\xi}/Y_{\xi}$. If $f_{\xi}$ is the frequency of marriages of class $\xi$, then the number of marriages per isolate per year is equal to $A/(A_{\xi}/Y_{\xi}) = Y_{\xi}/f_{\xi}$. If $\bar{A}$ is the average number of marriages per year per head of population (counting individuals of all ages), and $N$ is the size of an isolate, we have $\bar{A}N = Y_{\xi}/f_{\xi}$, so that

$$N = Y_{\xi}/\bar{A}f_{\xi}$$

(2.64)

This replaces Dahlberg's (1929, 1948) formulae.
3

Selection: construction of a model and the properties of equilibrium populations

3.1 Introduction

The problem of the theory of natural selection in age-structured populations was first considered quite early in the history of population genetics, by Haldane (1927a) and Norton (1928). Until comparatively recently, there was little interest in adding to the important contributions of these authors, apart from the isolated paper of Haldane (1962). This is probably largely due to the introduction by Fisher (1930) of his well-known ‘Malthusian parameter’ method of dealing with selection in continuous-time populations, which apparently provided a simple and elegant answer to the problem, and has since been extensively employed in population genetics (Crow and Kimura, 1970). This method assumes that fixed per capita birth-rates and death-rates can be assigned to each genotype, analogous to the rates defined in section 1.3.2 for a genetically homogeneous population in stable age-distribution. The Malthusian parameter for a genotype is the difference between these two quantities, and is used in differential equations for gene frequency change in much the same way that the standard discrete-generation fitness measure of Wright and Haldane is used in difference equations. But as pointed out by several authors (Moran, 1962; Charlesworth, 1970; Pollak and Kempthorne, 1970, 1971), an individual genotype cannot have a fixed per capita birth-rate and death-rate when the genotypic composition of a population is changing under natural selection, resulting in a continual disturbance of the population’s demographic structure.

The question of the analysis of the process of selection in age-structured populations is, therefore, not satisfactorily settled by the Malthusian parameter technique, and interest in the problem revived about twenty years ago, as can be seen from the publications of Anderson and King (1970), Charlesworth (1970, 1972–4, 1976, 1990a, 1993), Pollak and Kempthorne (1970, 1971), King and Anderson (1971), Charlesworth and Giesel (1972a, b) Charlesworth and Charlesworth (1973), Lande (1982), Tuljapurkar (1982), Abugov (1986), and Orzack and Tuljapurkar (1989), among others. During this period, there has also been strong interest
among evolutionary theorists in the evolution of life-histories, which reflect the phenotypic consequences of selection in age-structured populations (reviewed by Roff, 1992; Stearns, 1992; Charnov, 1993). This topic is discussed in Chapter 5.

The present chapter is concerned with the construction of a reasonably realistic and rigorous model of the population genetics of natural selection, and with the analysis of equilibria under selection, mostly with respect to a single locus. The following chapter deals with the dynamics of selection, and with selection on quantitative characters. Both finish with a discussion of the biological implications of the main results.

The mathematical models throughout these chapters are developed mainly in terms of the difference equation approach to age-structured populations, described in section 1.2.1, since in studying the selection process this method gives a more compact notation for the highly non-linear equations which arise than the equivalent matrix formulation. It also permits an easy extension to the integral equations of the corresponding continuous-time model. Only sexually reproducing populations are considered; the relatively trivial case of selection in an asexual population has already been dealt with in section 1.5. The presentation is for the case of two distinct sexes, although hermaphrodite and monoeccious species can be described by the same equations if male and female gametes from the same individual are considered separately. Random mating with respect to genotype is assumed throughout (complete self-fertilisation has been studied by Pollak and Kempthorne, 1970). Selection is assumed throughout to be frequency independent, although considerable attention is paid to density-dependent selection.

3.2 Construction of a model of selection

3.2.1 Genotypic parameters

As in the case of selection in an asexual species (section 1.5.2), it is necessary to specify demographic parameters for each genotype. For concreteness, consider an autosomal locus with alleles $A_1, A_2, \ldots, A_n$, and let $A_tA_j$ be the genotype of an individual who received $A_i$ from its mother and $A_j$ from its father. Reciprocal effects are assumed to be absent, so that the parameters of $A_tA_j$ and $A_jA_t$ are identical. Let $P_{ij}(x, t)$ and $P_{ji}(x, t)$ be the probabilities of survival from age-class $x$ at time $t$ to age-class $x + 1$ at time $t + 1$ for $A_tA_j$ females and males respectively. As in section 1.2.1, we can derive from these the probabilities of $A_tA_j$ females and males surviving to age $x$ at time $t$ from conception at time $t - x$.
In principle, we can also define the fecundity of an $A_iA_j$ individual aged $x$ at time $t$ in terms of the number of offspring it is expected to produce in time-interval $t$. By analogy with section 1.2.1, this can be written as $M_{ij}(x, t)$ for a female and $M^*_j(x, t)$ for a male. Empirically, these fecundities can be measured simply by estimating the mean number of offspring attributable to an $A_iA_j$ female or male of age $x$ at time $t$. As in the case of a genetically homogeneous population, these parameters cannot in general be treated as independent of the age-composition of the population. The problem is even worse in the present case since, if different classes of matings have different fecundities, the net fecundity of individuals of a given genotype, age and sex may be affected by the genotypic composition of the population of the opposite sex, as well as by its age composition. This problem also arises, in a less acute form, in discrete generation models of selection (Bodmer, 1965; Kemphorne and Pollak, 1970; Prout, 1971a, b; Abegov, 1985; Nagylaki, 1987). It is usually dealt with in that case by making specialising assumptions about the way in which selection on fecundity works, and this is essentially what will be done here.

### 3.2.2 Models of the mating process

It is convenient and realistic to make the assumption used in deriving equation (1.5): the expected number of offspring produced at time $t$ by an $A_iA_j$ female aged $x$ is independent of the genotypes and ages of her mate or mates. Hence, $M_{ij}(x, t)$ is independent of the composition of the population with respect to both age and genotype, unless frequency-dependent selection is operating.

If $N_{ij}(x, t)$ is the number of $A_iA_j$ females aged $x$ at time $t$, the total number of zygotes produced at time $t$ is

$$B(t) = \sum_{ijx} N_{ij}(x, t) M_{ij}(x, t)$$  \hspace{1cm} (3.2a)

Similarly, let $N^*_{ij}(x, t)$ be the number of $A_iA_j$ males aged $x$ at time $t$. Since every individual has a mother and father, we must also have

$$B(t) = \sum_{ijx} N^*_{ij}(x, t) M^*_{ij}(x, t)$$  \hspace{1cm} (3.2b)
The summations in these equations are taken over all reproductively active individuals. Selection may well involve the ages of first and last reproduction; in order to avoid excessive use of subscripts, wherever upper and lower ages of reproduction appear \((b\) and \(d\) for females, \(b^*\) and \(d^*\) for males) these are to be taken as representing the maximum upper and minimum lower values respectively for all genotypes, unless genotypes are specifically identified by subscripts.

Since \(M_{ij}(x, t)\) is assumed to be independent of the population's composition, equations (3.2) imply that the absolute values of the \(M^*_{ij}(x, t)\) are constrained by the state of the male and female populations. The nature of this constraint is particularly simple if a second assumption is introduced: that mating of fertile individuals is random with respect to both genotype and age. Such randomness arises naturally in the following two models of the mating process.

**Sperm or pollen pool model**

In this model, breeding males at time \(t\) are assumed to shed their gametes into a common pool of effectively infinite size, from which random draws are made to fertilise the ova. Let \(\theta_{ij}(x, t)\) be the expected number of gametes contributed to the pool at time \(t\) by an \(A_iA_j\) male aged \(x\). This quantity characterises the absolute fecundity of the class of male in question in the same way that the \(M_{ij}(x, t)\) functions characterise female fecundity. The \(\theta_{ij}(x, t)\) may in principle be independent of the state of the population. The total size of the male gamete pool at time \(t\) is

\[
B^*(t) = \sum_{ijx} N^*_{ij}(x, t) \theta_{ij}(x, t)
\]

(3.3)

Only \(B(t)\) of these gametes actually fertilise ova, so that the net fecundity of an \(A_iA_j\) male aged \(x\) is

\[
M^*_{ij}(t, x) = \frac{\theta_{ij}(x, t) B(t)}{B^*(t)}
\]

(3.4)

where \(B(t)\) is determined by equation (3.2a).

This model of mating is realistic for organisms with external fertilisation, such as many species of marine animals which shed their sperm and eggs into the sea, and self-incompatible higher plants, whose pollen is dispersed by wind or insect vectors. For organisms with internal fertilisation, the following model is more appropriate.

**Mating group model**

During time-interval \(t\), females mate with males chosen by draws from a group composed of all fertile males. Independent draws are made
from the group until all the fertile females present at time $t$ have successfully mated. The quantity $\theta_{ij}(x, t)$ which characterises a male of a given class is now to be regarded as the probability that he is chosen in a draw from the group. The relative values of the $\theta_{ij}(x, t)$ are independent of the composition of the population unless there is frequency-dependent selection. The net fecundities $M^*_{ij}(x, t)$ can be calculated by a technique similar to that used for the gamete pool model, yielding an expression identical with equation (3.4). $B^*(t)$ in this case is defined purely formally, however, and has no concrete biological meaning.

This model provides a reasonably accurate description of, for example, many species of insect or small vertebrate. It does not apply to species such as man, where there is a high correlation between the ages of mates. An examination of the consequences of non-random mating with respect to age is made later in this chapter (section 3.2.4), but for the present the assumption of random mating will be retained.

### 3.2.3 Genotypic frequencies

With the single-locus case and random mating, it is easily seen that the genotypic frequencies among the zygotes produced at time $t$ are identical with those obtained by combining pairs of alleles drawn at random from a pool of maternally derived genes and a pool of paternally derived genes. As in section 2.2.1, let the frequencies of allele $A_i$ among the maternal and paternal pools be $p_i(t)$ and $p^*_i(t)$ respectively. The frequency of $A_iA_j$ among the zygotes is

$$P_{ij}(t) = p_i(t)p^*_j(t)$$

(3.5)

where the allele frequencies are given by

$$B(t)p_i(t) = \frac{1}{2}\sum_{jx} [N_{ij}(x, t) + N^*_i(x, t)]M_{ij}(x, t)$$

(3.6a)

$$B(t)p^*_i(t) = \frac{1}{2}\sum_{jx} [N^*_{ij}(x, t) + N^*_i(x, t)]M^*_{ij}(x, t)$$

(3.6b)

These equations can be simplified by assuming that the frequency of females among new zygotes has a fixed value $a$, which is independent of the ages and genotypes of the parents. The number of male and female zygotes can then be written as fixed fractions of the total number of zygotes, $aB(t)$ and $(1 - a)B(t)$ respectively. We can also write $m_{ij}(x, t)$ for the expected number of daughters of an $A_iA_j$ female, and $m^*_{ij}(x, t)$ for the expected number of sons of an $A_iA_j$ male, aged $x$ at time $t$. 
This enables us to define reproductive functions for each genotype as

\[ k_{ij}(x, t) = l_{ij}(x, t)m_{ij}(x, t) \]  
\[ k_{ij}^*(x, t) = l_{ij}^*(x, t)m_{ij}^*(x, t) \]

(cf. section 1.2.1). For \( t \geq x \), we have from equation (3.5)

\[ N_{ij}(x, t) = aB(t - x)p_i(t - x)p_j^*(t - x)l_{ij}(x, t) \]  
\[ N_{ij}^*(x, t) = (1 - a)B(t - x)p_i(t - x)p_j^*(t - x)l_{ij}^*(x, t) \]

These relationships can be used to formulate difference equations for the allele frequencies. When \( t \geq d, d^* \), all the zygotes in the population are produced by parents who were born after the initial time-interval. Equations (3.6) become, using equations (3.8) and (3.9),

\[ B(t)p_i(t) = \frac{1}{2} \sum_{j} B(t - x)[p_i(t - x)p_j^*(t - x)]k_{ij}(x, t) \]  
\[ B(t)p_j^*(t) = \frac{1}{2} \sum_{j} B(t - x)[p_i(t - x)p_j^*(t - x)]k_{ij}^*(x, t) \]

where

\[ B(t) = \sum_{ij} B(t - x)p_i(t - x)p_j^*(t - x)k_{ij}(x, t) \]  
\[ = \sum_{ij} B(t - x)p_i(t - x)p_j^*(t - x)k_{ij}^*(x, t) \]

For times earlier than this, contributions from individuals alive in the initial time-interval must be taken into account as well. We use a method similar to that of equation (1.18). Write

\[ g_i(t) = \frac{1}{2} \sum_{x = t+1}^{d} \sum_{j} [N_{ij}(x - t, 0) + N_{ij}^*(x - t, 0)]l_{ij}(x, t)M_{ij}(x, t) \]  
\[ g_i^*(t) = \frac{1}{2} \sum_{x = t+1}^{d} \sum_{j} [N_{ij}^*(x - t, 0) + N_{ij}^*(x - t, 0)]l_{ij}^*(x, t)M_{ij}^*(x, t) \]
where \( g_i(t) = 0 \) for \( t \geq d \), \( g^*(i) = 0 \) for \( t \geq d^* \); \( \tilde{l}_{ij}(x, t) \) and \( \tilde{l}^*_{ij}(x, t) \) are defined analogously to \( \tilde{l}(x, t) \) in equation (1.18). It is easily seen that the full versions of equations (3.10) for arbitrary \( t \) are

\[
B(t)p_i(t) = g_i(t) + \frac{1}{2} \sum_{x=1}^{t} B(t-x) \sum_j [p_i(t-x)p^*_j(t-x)]k_{ij}(x, t) \\
+ p_i(t-x)p^*_j(t-x)]k_{ij}(x, t) \\
+ p_i(t-x)p^*_j(t-x)]k_{ij}(x, t) \\
+ p_i(t-x)p^*_j(t-x)]k_{ij}(x, t) \\
\]

There are \( 2(n + 1) \) equations in this system, taking \( i \) over all possible values, but only \( 2n \) are independent, since \( \sum p_i = \sum p^*_i = 1 \). These equations can be simplified further if it is assumed that the life-tables for each genotype are identical for males and females, and that the age-specific fecundity functions for each genotype have the same shape with respect to age in males and females, i.e. if we have

\[
l_{ij}(x, t) = l^*_{ij}(x, t) \\
\theta_{ij}(x, t) \propto M_{ij}(x, t) \\
\]

(Note that this implies \( b = b^* \) and \( d = d^* \).)

If these relations are assumed, it is easy to see that \( p_i(t) = p^*_i(t) \) for \( t \geq d \), regardless of the initial conditions. If we shift the time-scale so that the initial time interval falls in the period when \( p_i = p^*_i \), a full description of the population’s state is given by the reduced system of \( n + 1 \) equations, \( n \) of which are independent

\[
B(t)p_i(t) = g_i(t) + \sum_{x=1}^{t} B(t-x) \sum_j [p_i(t-x)p^*_j(t-x)]k_{ij}(x, t) \\
\]

\[
B(t) = \sum_i g_i(t) + \sum_{x=1}^{t} B(t-x) \sum_{ij} [p_i(t-x)p^*_j(t-x)]k_{ij}(x, t) \\
\]
Equilibria with selection

Table 3.1. Assumptions involved in deriving the difference equations for gene frequencies

Equations (3.12) and (3.14) require that:

1. The fecundity of a mating between a given pair of individuals is determined solely by the age and genotype of the female.
2. Mating is at random with respect to age and genotype.
3. The primary sex-ratio is constant and independent of parental age and genotype.

Equations (3.14) also require that:

4. The age-specific survival probabilities of males and females of a given genotype are equal.
5. The age-specific fecundity functions of males and females of a given genotype are proportional.

The assumptions embodied in equation (3.13) are somewhat arbitrary, and are unlikely to hold for many biological situations. Equations (3.14) are useful, however, since they resemble closely the standard discrete-generation selection equations of population genetics (Crow and Kimura, 1970, p. 180), which in fact may be regarded as the special case of equations (3.14) with \( b = d = 1 \); they take the form

\[
\bar{W}(t)p_i(t) = p_i(t - 1)\sum_j p_j(t - 1)W_{ij}(t - 1)
\] (3.15a)

\[
\bar{W}(t) = \sum_\eta p_i(t - 1)p_\eta(t - 1)W_\eta(t - 1)
\] (3.15b)

where \( W_{ij}(t) \) is the fitness of \( A_iA_j \) in generation \( t \).

Equations (3.14) can therefore be used to compare the properties of age-structured populations with those of discrete-generation populations, as well as being more tractable than equations (3.12) for mathematical analysis. Equations of this form, or their matrix or continuous-time equivalents (equations (3.16)) have commonly been used in studies of the theory of selection in age-structured populations (e.g. Haldane, 1927a, 1962; Norton, 1928; Anderson and King, 1970; Charlesworth, 1970, 1972, 1976; Pollak and Kempthorne, 1971). The numerous assumptions needed to arrive at these equations are summarised in Table 3.1, and should be borne in mind whenever they are used. The earlier papers on this subject were rather casual about the nature of their assumptions concerning the mating process and mode of selection; Pollak and Kempthorne (1971) were the first to emphasise the need for clarity on these points. They in fact went further than we have done and assumed that the fecundities of each genotype were the same and were independent of age for each
fertile age-class. These assumptions are not necessary, and their general results are valid for the model leading to equations (3.14). For further details of the models described in this section, see Charlesworth (1972, 1976).

The continuous-time model

These results can easily be extended to the continuous-time case by increasing indefinitely the number of age-classes within the life-span. As in section 1.2.3, the population size at time $t$ is characterised by the population birth-rate $B(t)$, such that the number of zygotes produced in a short time-interval of length $dt$ is $B(t)dt$. The frequency of $A_i$ among maternally-derived genes of new zygotes at time $t$ is defined as $p_i(t)$, such that the number of $A_i$ genes of maternal origin among the zygotes produced between $t$ and $t + dt$ is $B(t)p_i(t)dt$; the paternal allele frequency, $p_f(t)$, is defined similarly. By re-defining $l_{ij}(x, t)$, $m_{ij}(x, t)$ and $k_{ij}(x, t)$, etc., appropriately and re-formulating the mating models in terms of arbitrarily small time-intervals, integral equations analogous to (3.12) and (3.14) can be obtained. For example, the integral analogues of equations (3.14) are

$$B(t)p_i(t) = g_i(t) + \int_0^t B(t-x)p_i(t-x)\sum_j p_j(t-x)k_{ij}(x, t)\,dx$$

(3.16a)

$$B(t) = \sum_i g_i(t) + \int_0^t B(t-x)\sum_{ij} p_i(t-x)p_j(t-x)k_{ij}(x, t)\,dx$$

(3.16b)

$$g_i(t) = \frac{1}{2} \int_0^d \sum_j [N_{ij}(x-t, 0) + N_{ji}(x-t, 0)]\tilde{l}_{ij}(x, t)M_{ij}(x, t)\,dx$$

(3.16c)

where $\tilde{l}_{ij}(x, t)$ is defined analogously to $\tilde{l}(x, t)$ in equation (1.26).

Cornette (1975) and Nagylaki (1977, pp. 79–82) should be consulted for an alternative formulation of the mating process with the continuous-time model.

The two-locus case

The discrete age-class model developed above can easily be extended to genetic systems more complex than that of a single autosomal
Equilibria with selection

locus. For example, consider a two-locus system with alleles $A_1$, $A_2$ at one locus and $B_1$, $B_2$ at the other. With the assumptions that led to equations (3.14), it is easy to derive difference equations which describe the system completely. Let $c$ be the recombination fraction, and write the frequencies of the gametes $A_1B_1$, $A_1B_2$, $A_2B_1$ and $A_2B_2$ at time $t$ as $p_1(t)$, $p_2(t)$, $p_3(t)$ and $p_4(t)$, respectively. The coefficient of linkage disequilibrium (cf. p. 76) is $D(t) = p_1(t)p_4(t) - p_2(t)p_3(t)$. For $t \geq d$ we have

$$B(t)p_i(t) = \sum_{j} B(t - x)p_i(t - x)p_j(t - x)k_{ij}(x, t)$$

$$\pm \sum_{x} cB(t - x)D(t - x)k_{14}(x, t) \quad (3.17a)$$

$$B(t) = \sum_{ijx} B(t - x)p_i(t - x)p_j(t - x)k_{ij}(x, t) \quad (3.17b)$$

where the sign attached to $c$ is positive for $i = 2, 3$ and negative for $i = 1, 4$.

The extension of these equations to times $t < d$ is trivial.

3.2.4 Non-random mating with respect to age

Before going on to explore the properties of the gene frequency equations derived in the previous sections of this chapter, it is worthwhile to consider the consequences of relaxing the assumption of randomness of mating with respect to age. The present account is based on that of Charlesworth and Charlesworth (1973). It is necessary in this case to consider all possible matings with respect to the ages of the partners. Let $\psi(x, y, t)$ be the frequency of matings at time $t$ involving females aged $x$ and males aged $y$. Let $p_i(x, t)$ be the frequency of $A_i$ among the gametes produced by females in such matings, and $p_i^*(y, t)$ be the corresponding frequency among male gametes. It is convenient to write

$$\bar{p}_i(x, y, t) = \frac{1}{2}[p_i(x, t) + p_i^*(y, t)] \quad (3.18a)$$

$$\delta_i(x, y, t) = p_i(x, t) - \bar{p}_i(x, y, t) = \bar{p}_i(x, y, t) - p_i^*(y, t) \quad (3.18b)$$

If there is random mating with respect to genotype, the genotypic frequencies among the zygotes produced by this class of matings may be written (dropping the argument $t$ for brevity)
Construction of a model of selection

\[ p_{ij}(x, y) = p_i(x)p_j(y) = \bar{p}_i^2(x, y) - \delta_i^2(x, y) \]  
\[(3.19a)\]

\[ p_{ij}(x, y) + p_{ji}(x, y) = p_i(x)p_j(y) + p_j(x)p_i(y) \]
\[= 2\bar{p}_i(x, y)\bar{p}_j(x, y) - 2\delta_i(x, y)\delta_j(x, y)\]  
\[(3.19b)\]

Taking expectations over all classes of mating, we find that the overall genotypic frequencies among the new zygotes are given by

\[ p_u(x, y) = \sum_{xy} p(x, y)[\bar{p}_i^2(x, y) - \delta_i^2(x, y)] \]
\[= \bar{p}_i^2 - \delta_i^2 + V\bar{p}_i(x, y) - V\delta_i(x, y) \]  
\[(3.20a)\]

\[ p_{ij} + p_{ji} = 2\sum_{xy} p(x, y)[\bar{p}_i(x, y)\bar{p}_j(x, y) - \delta_i(x, y)\delta_j(x, y)] \]
\[= 2\bar{p}_i\bar{p}_j - 2\delta_i\delta_j + 2 Cov[\bar{p}_i(x, y), \bar{p}_j(x, y)] \]
\[ - 2 Cov[\delta_i(x, y), \delta_j(x, y)] \]  
\[(3.20b)\]

where \( \bar{p}_i = \sum_{xy} p(x, y)\bar{p}_i(x, y) \), \( \delta_i = \sum_{xy} p(x, y)\delta_i(x, y) \), and the \( V \) and \( Cov \) functions are the variances and covariances across age-classes of the variables enclosed in square brackets.

If selection is not intense, we can ignore the second-order terms in the differences in the gene frequencies and in the \( \delta_i \) between different classes of mating (section 4.2), so that the variance and covariance terms can be dropped from equations (3.20), to a good approximation. Noting that
\[ p_i(t) = \bar{p}_i(t) + \delta_i(t) \]  
and
\[ p_i^*(t) = \bar{p}_i(t) - \delta_i(t) \]
the approximate expressions can be substituted into equations (3.9), enabling us to recover equations (3.10) and (3.12). This shows that, as far as the formal expressions for the gene frequencies are concerned, non-random mating with respect to age has only a second-order effect. But, unlike the random-mating case (section 3.2.2), it is no longer possible to define \( \theta_{ij}(x, t) \) functions which are independent of the composition of the population. This means that with non-random mating between age-classes, equations (3.10) etc. are useful only in treating equilibrium populations, and certain restricted types of non-equilibrium situations to be discussed in section 4.3.1.

With intense selection, equations (3.20) show that deviations from Hardy–Weinberg frequencies among the new zygotes can be produced even if there is random mating with respect to genotype. With just two alleles, \( A_1 \) and \( A_2 \), equations (3.20) become
The effects of differences in gene frequency between the sexes and differences between classes of mating are in opposition, the former tending to produce an excess of heterozygotes, and the latter a deficiency, over the Hardy–Weinberg frequencies. If selection acts in a similar way in males and females, it is probable that the terms in \( \delta_i^2 \) will be small, compared with the variance in gene frequencies over classes of matings, resulting in a net deficiency of heterozygotes among the new zygotes. Of course, if there is a viability advantage of the heterozygote, this deficiency of heterozygotes will diminish among older age-classes and may be replaced by an excess of heterozygotes. This may be relevant to some observations of heterozygote deficiencies in natural populations (e.g. Schaal and Levin, 1976; Charlesworth, 1991).

3.3 Populations in genetic equilibrium

In this section we examine the more important properties of age-structured populations in genetic equilibrium under selection. For convenience, the simplified selection equations (3.14) are used for deriving most of the results, although reference to the more realistic models will be made where appropriate. The treatment here is derived largely from the papers of Charlesworth (1972) and Charlesworth and Charlesworth (1973).

3.3.1 Dependence of genetic equilibrium on demographic stability

One of the most important results in the theory of selection in age-structured populations is that genetic equilibrium is generally possible only if there is demographic stability, i.e. if the population is stationary in size or growing at a constant rate, and has a constant age-structure. A corollary of this is that populations which are subject to fluctuations in growth-rate and age-structure may show fluctuations in their composition at loci undergoing selection (Anderson and King, 1970; Charlesworth and Giesel, 1972a, b). The biological implications of this result are discussed in section 3.4.1; here we will investigate its mathematical basis.

In an equilibrium population, the gene frequencies at a locus must take values independent of time, so we can write \( p_i(t) = \hat{p}_i \) in equations

\[
p_{11} = \bar{p}_1^2 - \delta_1^2 + V[\bar{p}_1(x, y)] - V[\delta_1(x, y)] \\
p_{12} + p_{21} = 2\bar{p}_1\bar{p}_2 + 2\delta_1^2 - 2V[\bar{p}_1(x, y)] + 2V[\delta_1(x, y)]
\]
(3.14), yielding the equilibrium equations for $t \geq d$

$$1 = \sum_j \hat{p}_j \sum_x \frac{B(t - x)}{B(t)} k_{ij}(x, t)$$  \hspace{1cm} (3.22a)

$$B(t) = \sum_{ijx} B(t - x) \hat{p}_i \hat{p}_j k_{ij}(x, t)$$  \hspace{1cm} (3.22b)

These may be compared with the equilibrium form of the discrete-generation equations (3.15), writing the fitnesses $W_{ij}$ as independent of time, as must usually be the case in an equilibrium population

$$1 = \sum_j \hat{p}_j \frac{W_{ij}}{\hat{W}}$$  \hspace{1cm} (3.23a)

$$\hat{W} = \sum_{ij} \hat{p}_i \hat{p}_j W_{ij}$$  \hspace{1cm} (3.23b)

Only the relative values of the $W_{ij}$ in equations (3.23) are important in determining the equilibrium gene frequencies. If we wish to define a fitness measure for the age-structured case which determines the equilibrium gene frequencies in the same way as the $W_{ij}$ in the discrete-generation case, we can therefore choose a scale in which the equivalent of $\hat{W}$ is 1, with no loss of generality; this yields the following equilibrium fitness measure for the age-structured case

$$w_{ij} = \sum_x \frac{B(t - x)}{B(t)} k_{ij}(x, t)$$  \hspace{1cm} (3.24)

(The fitness measure is given in lower case to distinguish it from the discrete-generation $W_{ij}$.)

For there to be genetic equilibrium, the relative values of these $w_{ij}$ must be independent of time, except in certain degenerate cases discussed by Charlesworth (1972). Since variation with $t$ of the function $B(t - x)/B(t)$ affects the relative contributions of different ages to the $w_{ij}$, it is difficult to imagine conditions under which a population with a varying growth-rate, and hence varying $B(t - x)/B(t)$, could be in genetic equilibrium, unless we can write for all $ij$

$$k_{ij}(x, t) = w'_{ij} k_s(x, t)$$  \hspace{1cm} (3.25)

where $k_s(x, t)$ is the reproductive function of some standard genotype, and $w'_{ij}$ is a constant independent of $x$ and $t$ but characteristic of $A_i A_j$.

It is unlikely that equation (3.25) can often be satisfied. If selection acts
at the level of age-specific survival probabilities, for example, and if genotypic differences in these are independent of age, we can write

\[ l_{ij}(x) = P_{ij}^s(x) \]

where \( l_i(x) \) is the survival function for the genotype with the highest survival probabilities, and \( P_{ij} = 1 \) is a constant characteristic of \( A_i A_j \).

Even in this simple case, equation (3.25) is violated. Similarly, if selection acts at the level of fecundity, it is likely that genotypes will differ in such parameters as age at first or last reproduction, time of peak fecundity etc., which will also be inconsistent with equation (3.25).

We may conclude, therefore, that a population will be in genetic equilibrium only if the population has a constant growth-rate and hence stable age-structure, unless selective differences among genotypes are non-specific with respect to age in such a way that equation (3.25) is satisfied. Of course, deviations from genetic equilibrium caused by changing age-structure may be so slight as not to be noticeable in practice; the conditions under which there are significant effects of changing demography are discussed in section 3.4.1.

**Proof for the time-independent case**

A formal proof that demographic stability is necessary for genetic equilibrium under selection can be given in the case when the \( k_{ij}(x, t) \) functions are independent of \( t \). We assume that the population is at genetic equilibrium but does not possess a stable age-structure, and show that this leads to a contradiction except when equation (3.25) is satisfied.

If the population is in genetic equilibrium, it can be treated from the demographic viewpoint as if it were genetically homogeneous, replacing the usual reproductive function \( k(x) \) (equation (1.16)) by the mean for the equilibrium population

\[ k(\hat{p}, x) = \sum_{ij} \hat{p}_i \hat{p}_j k_{ij}(x) \]  

(3.26)

From the standard theory described in section 1.3.1, it follows that, for a population with changing age-structure, we can write \( B(t) \) at any time as

\[ B(t) = \sum_{l=1}^{d} C_l \lambda_t^{l+1} \]  

(3.27)

where the \( \lambda_t \) are the roots of the \( d \)th degree polynomial

\[ \sum_x \lambda^{-x} k(\hat{p}, x) = 1 \]  

(3.28)
(Equation (3.27) assumes for simplicity that the $\lambda_l$ are all distinct, but the following proof can easily be extended to cover the cases of multiple roots.)

If the expression for $B(t)$ in equation (3.27) is substituted into equation (3.24), the following condition for constant $w_{ij}$ must hold for each genotype and each value of $t$

$$\sum_{l=1}^{d} C_l \lambda_l^{-x} \sum_{x} \lambda_l^{i} k_{ij}(x) = w_{ij} \sum_{l=1}^{d} C_l \lambda_l^{i+1}$$

(3.29)

By choosing an arbitrary set of values of $t$, $[t_1, t_2, \ldots, t_d]$, and substituting them into equation (3.29), we obtain a set of homogeneous linear equations in $d$ unknowns, $[y_1, y_2, \ldots, y_d]$

$$\sum_{l=1}^{d} a_{ml} y_l = 0$$

(3.30)

where

$$y_l = \sum_{x} \lambda_l^{i} k_{ij}(x) \frac{w_{ij}}{-1}, \quad a_{ml} = C_l \lambda_l^{-x}$$

But, since the choice of $t$ values is arbitrary, it must be possible for the determinant of the $a_{ml}$ to be non-zero, which implies that the $y_l$ must all be zero for equation (3.30) to hold. Each $\lambda_l$ must therefore be a root of the following equation, for every $i$ and $j$

$$\sum_{x} \lambda_l^{i} k_{ij}(x) \frac{w_{ij}}{w_{ij}} = 1$$

Since the coefficients of a polynomial equation are completely determined by their roots, this implies that the $k_{ij}(x)/w_{ij}$ are the same for each $ij$ and a given value of $x$, so that equation (3.25) is satisfied. Under any other conditions, the population cannot simultaneously be in genetic equilibrium and not have a stable age-structure.

### 3.3.2 Equilibrium fitness measures

The above results mean that there no loss in generality in writing $B(t - x)/B(t)$ in equation (3.24) as $\lambda_1^{-x}$, where $\lambda_1$ is the geometric rate of increase of the equilibrium population; in cases where the fitnesses are sensitive to changing demography, an equilibrium can only exist if the population growth-rate is constant, and in cases where the fitnesses are insensitive, the values of $B(t - x)/B(t)$ are unimportant. We can similarly write $k(x, t)$ as independent of $t$. For notational reasons which will
Equilibria with selection

become clearer when we discuss the dynamics of selection in Chapter 4, it is convenient to work with the natural logarithm of \( \lambda_1 \), which we will write as \( \hat{r} \) in order to emphasise that it is the intrinsic rate of increase for the equilibrium population (cf. equation (1.45)). The equilibrium fitness measure for equations (3.14) can therefore be written as

\[
w_{ij} = \sum_x e^{-\hat{r}x} k_{ij}(x)
\]

(3.31)

The mean fitness of the equilibrium population is, on this definition, equal to the mean Fisherian reproductive value for a new zygote (cf. equations (1.54)), which is equal to unity. The equilibrium fitness of a genotype can be thought of as its contribution to the reproductive value, at the zygote stage, of the population as a whole.

Similar conclusions concerning the interdependence of demographic stability and genetic equilibrium apply to more general situations. For example, we can use equations (3.20) for the case of non-random mating and differences in demographic parameters between the sexes. Neglecting the terms which are second-order in the gene frequency differences between the sexes, as well as the variance and covariance terms, we obtain

\[
p_{ii} \equiv \bar{p}_i^2
\]

\[
p_{ij} + p_{ji} = 2\bar{p}_i\bar{p}_j
\]

Substituting these into equations (3.9) and (3.10), and solving for equilibrium, we find that

\[
1 \equiv \sum_x \frac{B(t - x)}{B(t)} \sum_j \hat{B}_{ij}(x, t)
\]

\[
1 \equiv \sum_x \frac{B(t - x)}{B(t)} \sum_j \hat{B}_{ij}(x, t)
\]

which can be added, to yield

\[
1 \equiv \frac{1}{2} \sum_x \frac{B(t - x)}{B(t)} \sum_j \hat{B}_{ij}(x, t) [k_{ij}(x, t) + k_{ji}(x, t)]
\]

(3.32)

This is clearly analogous to equation (3.22a), and can be used to draw the same general conclusions. The equilibrium fitness measure corresponding to that of equation (3.31) is

\[
w_{ij} \equiv \frac{1}{2} \sum_x e^{-\hat{r}x} [k_{ij}(x) + k_{ji}(x)]
\]

(3.33)
This expression shows that the mean of the fitness measures for the two sexes can, with sufficiently weak selection, be used for determining equilibrium gene frequencies in the same way as the fitnesses when there are no sex differences. The same is true in the corresponding discrete-generation case (Wright, 1942; Nagylaki, 1979).

As discussed by Charlesworth (1972) and Charlesworth and Charlesworth (1973), similar results can be derived for selection models involving such additional factors as recombination and mutation (see also pp. 125–6), and expressions analogous to the discrete-generation equations for equilibrium frequencies can always be derived, with the fitness measures of equations (3.31) or (3.33) replacing the corresponding discrete-generation formulae. Finally, we may note that analogous results can be obtained with the continuous-time model; integration replaces summation and the instantaneous logarithmic growth-rate of the equilibrium population replaces $\hat{r}$.

In order to utilise these equilibrium fitness measures, it is obviously usually necessary to know the value of $\hat{r}$. Unless this can be determined in terms of more fundamental genotypic parameters, or is given by empirical data, the equilibrium fitnesses are of uncertain value except when equation (3.25) is satisfied. In the next section, we see how $\hat{r}$ can be determined in the case of a single autosomal locus under selection alone and, in the following section, in the case of selection balanced by mutation.

### 3.3.3 Equilibrium in a single-locus system under selection

#### Density-independent case

Consider first of all the system described by equations (3.14) with time-independent $k_{ij}(x, t)$ functions. A convenient way of summarising the demographic characteristics of a genotype $A_iA_j$ is by means of the intrinsic rate of increase of a homogeneous population, all of whose members have the reproductive function of $A_iA_j$, $k_{ij}(x)$. We can call this the intrinsic rate of increase of $A_iA_j$, denoted by $r_{ij}$. As in equation (1.45), $r_{ij}$ is the real root of the equation in $z$

$$w_{ij}(z) = \sum_x e^{-zx} k_{ij}(x) = 1$$ (3.34)

where $w_{ij}(z)$ generalises the fitness measure of equation (3.31) to arbitrary values of the complex or real variable $z$.

If we imagine a hypothetical population with fixed gene frequencies $p_1$, $p_2$, and
$p_2, \ldots, p_n$, its intrinsic rate of increase, $r_p$, can be defined as the real root of

$$\bar{w}(p, z) = \sum_x e^{-zx} \sum_{ij} p_i p_j k_{ij}(x) = 1$$ (3.35)

Obviously, a real population only attains the rate of growth $r_p$ if the gene frequencies are at equilibrium, but this does not prevent us from using $r_p$ as a useful means of characterising a non-equilibrium population with a given set of gene frequencies.

Using these definitions in the two-allele case, equations (3.22) become

$$1 = \hat{p}_1 w_{11}(\hat{r}) + \hat{p}_2 w_{12}(\hat{r})$$ (3.36a)

$$1 = \hat{p}_1 w_{12}(\hat{r}) + \hat{p}_2 w_{22}(\hat{r})$$ (3.36b)

Noting that $\hat{p}_1 + \hat{p}_2 = 1$, the following pair of homogeneous linear equations in $\hat{p}_1$ and $\hat{p}_2$ can be obtained from these

$$\hat{p}_1[1 - w_{11}(\hat{r})] + \hat{p}_2[1 - w_{12}(\hat{r})] = 0$$ (3.37a)

$$\hat{p}_1[1 - w_{12}(\hat{r})] + \hat{p}_2[1 - w_{22}(\hat{r})] = 0$$ (3.37b)

Solving for $\hat{p}_1$ and $\hat{p}_2$ we obtain the gene-frequency equation for a polymorphic equilibrium

$$\frac{\hat{p}_1}{\hat{p}_2} = \frac{w_{12}(\hat{r}) - w_{22}(\hat{r})}{w_{12}(\hat{r}) - w_{11}(\hat{r})}$$ (3.38)

This, as anticipated from the general results derived above, is identical in form with the standard discrete-generation equation for this case (Crow and Kimura, 1970, p. 270). For an equilibrium with $0 < \hat{p}_1 < 1$, we must have either $w_{12} > w_{11}$, $w_{22}$ or $w_{12} < w_{11}$, $w_{22}$, excluding the degenerate case $w_{11} = w_{12} = w_{22}$ which implies $r_{11} = r_{12} = r_{22} = \hat{r}$. Furthermore, equations (3.37) imply that $\hat{r}$ must lie between $r_{12}$ and the closer of $r_{11}$ and $r_{22}$, since the $w_{ij}(z)$ are strictly decreasing functions of $z$ when $z$ is real. This decreasing property also implies that $w_{12} > w_{11}$, $w_{22}$ is equivalent to $r_{12} > r_{11}$, $r_{22}$, and that $w_{12} < w_{11}$, $w_{22}$ is equivalent to $r_{12} < r_{11}$, $r_{22}$.

To determine $\hat{r}$, we note that, for equations (3.37) to be satisfied for non-zero $\hat{p}_1$ and $\hat{p}_2$, the determinant of their coefficients must vanish, i.e. $\hat{r}$ must satisfy the equation

$$[1 - w_{12}(z)]^2 = [1 - w_{11}(z)][1 - w_{22}(z)]$$ (3.39)

This equation was first derived by Norton (1928).
It can be shown as follows that there is only one real value of $z$ which satisfies both this equation and the above requirements for $\hat{r}$. Consider the case $r_{12} > r_{11}, r_{22}$. Then, from the fact that this implies $w_{12} > w_{11}, w_{22}$ at equilibrium and from standard population genetics theory, it follows that the function $\hat{w}(p, \hat{r})$ is at a global maximum with respect to $p_1$ in the closed interval $[0, 1]$, keeping $z$ fixed at $\hat{r}$ and using the constraint $\hat{p}_1 + \hat{p}_2 = 1$. Since $\hat{w}(p, z)$ is a decreasing function of $z$ when $z$ is real, this implies $\hat{r} > r_p$ for all $p_1 \neq \hat{p}_1$ in $[0, 1]$. This excludes the possibility that there is another equilibrium point which is also a maximum of $\hat{w}$ with respect to $p_1$. The existence of a minimum of $\hat{w}$ is excluded by the fact, also given by standard theory, that this requires $w_{12} < w_{11}, w_{22}$ at the equilibrium point; this is incompatible with $r_{12} > r_{11}, r_{22}$.

We conclude, therefore, that where $r_{12} > r_{11}, r_{22}$, there is a unique equilibrium point with $\hat{r}$ given by equation (3.39), and which corresponds with a global maximum in $r_p$. A similar argument can be used to establish that $r_{12} < r_{11}, r_{22}$ implies a unique equilibrium with $r_p$ at a global minimum. For practical purposes, $\hat{r}$ must usually be obtained by Newton–Raphson iteration of equation (3.39), given the $k_{ij}(x)$ functions for each genotype. An example of the calculation of $\hat{r}$ and the $w_{ij}$ for a set of genotypes of *Tribolium castaneum* is given by Moffa and Constantin (1977).

**Density-dependent case**

A very similar result can be obtained for the case when the population size is regulated by a density-dependent negative feedback mechanism, using the type of model introduced in section 1.4.2 for a genetically homogeneous population (Charlesworth, 1972). It is assumed that we can define a critical age-group consisting of a set $S$ of age-classes, such that the age-specific survival probabilities or fecundities for one or more age-classes are decreasing functions of the numbers of individuals in the critical age-group, either at the present time or at some past time, depending on the mechanism of density regulation. In a population whose numbers are changing in time, the value of $k_{ij}(x, t)$ for a given genotype is controlled by the density of the population; in general, it will depend on some set of numbers of individuals in the critical age-group over a set of earlier times $\mathcal{T}$ (equations (1.89) and (1.90)). Writing $N_T$ for this set of numbers, we can replace $k_{ij}(x, t)$ by $k_{ij}(x, N_T)$ in equations (3.14). In a stationary population the numbers in the set $N_T$ are constant in value and equal to the number of individuals in the critical age-group.
The properties of equilibrium populations can be derived if we assume that the net reproduction rate of each genotype, in a population where gene frequencies and population numbers are held fixed, is a strictly decreasing function of the number of individuals in the critical age-group. (This assumption is analogous to the decreasing property of \( w_t(z) \) as a function of \( z \), in equation (3.34).) We can define the carrying-capacity, \( N_{ij} \), of \( A_iA_j \) as the equilibrium number of individuals in the critical age-group which would be reached by a genetically homogeneous population with the demographic parameters of \( A_iA_j \). This is given as the unique root of the equation

\[
\hat{w}(N) = \sum_x k_{ij}(x, N) = 1
\]  

where \( N \) is a positive real variable corresponding to a fixed number of individuals in the critical age-group. The carrying-capacity, \( N_p \), for a mixed population with fixed gene frequencies \( p_1, p_2, \ldots, p_n \) can be defined, analogously to \( r_p \) in equation (3.35), as the root of

\[
\hat{\bar{w}}(p, N) = \sum_{ijx} p_i p_j k_{ij}(x, N) = 1
\]

Using the decreasing properties of \( w_{ij}(N) \) and \( \hat{\bar{w}}(p, N) \) as functions of \( N \), and proceeding in the same way as in the density-independent case, it is possible to establish conditions for the existence of an equilibrium in terms of the \( N_{ij} \); they are identical with the earlier conditions in the \( r_{ij} \). We also have the result that \( N_p \) takes a maximum value, \( \hat{N} \), at equilibrium when \( N_{12} > N_{11}, N_{22} \) and a minimum when \( N_{12} < N_{11}, N_{22} \); \( \hat{N} \) is given by the analogue of equation (3.39) with \( N \) substituted for \( z \). These results are identical with those for the corresponding discrete-generation model (Charlesworth, 1971).

**Multiple alleles**

With \( n \) alleles, the equilibrium is determined by the set of equations corresponding to equations (3.36)

\[
\sum_i \hat{\bar{P}}_i w_{ij}(\hat{\bar{r}}) = 1
\]  

where \( \hat{\bar{r}} \) is a real root of the determinantal equations corresponding to equation (3.39), with \( a_{ij} = 1 - w_{ij}(z) \)

\[
|A| = 0
\]

An analogous result holds for the density-dependent case, with the equilibrium carrying-capacity, \( \hat{N} \), replacing \( \hat{\bar{r}} \).
Unfortunately, it is not possible to prove the same sort of uniqueness properties for \( \hat{N} \) and \( \hat{r} \) as in the two-allele case. It can, however, be shown that there is at most one maximum in \( r_p \) or \( N_p \) with all \( n \) alleles present; this corresponds to a stable equilibrium when selection is weak (pp. 140–1). Equation (3.42b) can always be solved numerically, given the \( k_{ij}(x) \) functions.

In most other cases, it is not possible to obtain an equation such as (3.39) or (3.42) which determines \( \hat{r} \) or \( \hat{N} \) in terms of the more basic genotypic parameters. This is so even in the case of a single locus with two alleles when there are demographic differences between the sexes, since the values of the \( k^*_j(x, t) \) depend on the composition of the population, so that a detailed specification of this dependence would be necessary to obtain a complete solution for the equilibrium population. This problem can be evaded when selection is weak, using the methods described in section 4.2.1. One important case which can be treated satisfactorily, without having to assume weak selection, is when an allele at a locus is maintained at a low frequency as a result of a balance between mutation and selection (Charlesworth and Charlesworth, 1973), and this case will now be considered briefly.

**Mutation-selection balance**

In this case, we consider a ‘wild-type’ allele \( A_1 \) at a locus, which mutates to a deleterious allele \( A_2 \), kept at a low frequency by selection. The rate of mutation from \( A_1 \) to \( A_2 \) is \( u(t) \), defined by equations (2.49). The rarity of \( A_2 \) means that it is reasonable to neglect terms of order \( u^2 \) and \( up_2 \). Equations (3.12) can be modified in a simple fashion to incorporate the effect of mutations. Carrying out a similar procedure to that involved in deriving equation (3.32), we obtain the following equations for the equilibrium values of the frequencies of \( A_1 \) and \( A_2 \), averaged over male and female gametes

\[
\hat{P}_1 = \sum_x \frac{B(t - x)}{B(t)} \hat{P}_1 [\hat{P}_1 k_{11}(x) + \hat{P}_2 k_{12}(x)] - u(t) \tag{3.43a}
\]

\[
\hat{P}_2 = \sum_x \frac{B(t - x)}{B(t)} \hat{P}_2 [\hat{P}_1 k_{12}(x) + \hat{P}_2 k_{22}(x)] + u(t) \tag{3.43b}
\]

where \( k_{ij}(x) = \frac{1}{2} [k_{ij}(x) + k_{ji}(x)] \), and \( B(t) \) is now the value for the population as a whole.

For an equilibrium population, there is, as we have seen, no loss in generality in writing \( B(t - x)/B(t) \) as \( e^{-\hat{r}x} \), where \( \hat{r} \) is the rate of growth of the population as a whole. It is easily seen that, in the present case, \( \hat{r} \)
is equal to $r_{11}$ plus terms of order $u(t)$, so that equations (3.43) can be written, to a satisfactory approximation, as

$$\hat{p}_1 = \hat{p}_1[\hat{p}_{11}w_{11} + \hat{p}_{12}w_{12}] - u$$  (3.44a)

$$\hat{p}_2 = \hat{p}_2[\hat{p}_{12}w_{12} + \hat{p}_{22}w_{22}] + u$$  (3.44b)

where

$$u = \frac{1}{2}\sum_x e^{-r_{11}x}[k_{11}(x)u(x) + k_{11}^*(x)u^*(x)]$$

and

$$w_{ij} = \sum_x e^{-r_{11}x}k_{ij}(x)$$

If the relative values of the $w_{ij}$ are written as 1, 1 - $hs$ and 1 - $s$ for $w_{11}$, $w_{12}$ and $w_{22}$ respectively, equations (3.44) can be re-arranged to yield the equilibrium frequency of $A_2$ in the standard form employed in discrete-generation models (Crow and Kimura, 1970, p. 260)

$$s(1 - 2h)\hat{p}_2^2 + hs\hat{p}_2 - u \equiv 0$$  (3.45)

This equation has the familiar approximate solution for the dominant or semi-dominant case

$$\hat{p}_2 \equiv u/hs \quad (h > 0)$$  (3.46a)

and, for the recessive case,

$$\hat{p}_2 \equiv \sqrt{u/s} \quad (h \equiv 0)$$  (3.46b)

Analogous results can be derived for the density-dependent case, with the fitnesses all being evaluated at $N = N_{11}$. It may be noted that equation (3.46a) is valid for the case of non-random mating with respect to age and demographic differences between the sexes regardless of the intensity of selection, since second-order terms in $\hat{p}_2^2$ are neglected in obtaining it and the terms in $\delta_1^2$ and $V$ in equations (3.21) are also of order $\hat{p}_2^2$ in this case. A similar treatment can also be made for the case of a sex-linked locus.

### 3.4 Biological applications of the results

This section surveys the implications of the theoretical results derived in this chapter for experimental and observational studies of natural populations, with emphasis on those properties which are peculiar to age-structured as opposed to discrete-generation populations.
Biological applications of the results

Probably the most important single respect in which age-structured populations differ from those with discrete generations is the interdependence of demographic stability and genetic equilibrium. As we saw in section 3.3.1, we generally expect an age-structured population to remain precisely in genetic equilibrium only if the population is demographically stable. Changes in the age-structure of the population, induced by ecological changes, may therefore cause changes in allele frequencies at loci under selection, even if the force of selection is unchanged at the fundamental level of age-specific survival probabilities and fecundities, i.e. if the relative values of the $k_{ij}(x)$ functions for different genotypes remain unchanged for each value of $x$.

The possible magnitudes of such shifts in gene frequencies, and their relationship to the nature and strength of selection, were investigated by Charlesworth and Giesel (1972a) by means of computer calculations of population trajectories of two-allele systems, using equations (3.14). They showed that cyclical fluctuations in gene frequencies could be generated in populations exposed to regular cycles of population growth, imposed by an external ecological factor which varied in intensity in time. An example of such an effect is shown in Figure 3.1. The fluctuations in population size were produced by periodic switches of the $k_{ij}(x)$ functions such that, at a point of change over from one regime to a new one, the old $k_{ij}(x)$ values for each genotype and age were multiplied by the same factor, which was greater or less than one depending on the desired direction of change in population growth-rate. This corresponds to the effect of an environmental factor causing periodic changes in the fecundities, and which affects each genotype and age equally. $A_1A_1$ initiated and completed reproduction later than $A_2A_2$. The heterozygote $A_1A_2$ combined the favourable characteristics of the homozygotes, thereby ensuring maintenance of the polymorphism in each environment (section 4.3.1). In

![Figure 3.1. Fluctuations in gene frequency at a polymorphic locus, induced by cyclical changes in the population growth-rate. The solid line shows the successive values of log $B(t)$, and the dashed line the frequency of allele $A_1$ of the system shown in Table 3.2. Each genotype had 13 age-classes. (After Charlesworth and Giesel, 1972a, Fig. 2.).](image-url)
this example, there were three environmental states, following each other in regular succession and corresponding to the equilibrium states shown in Table 3.2, which would be attained if the population were left in a given state for a sufficiently long time.

As can be seen from Figure 3.1, the actual time-course of change of the population is such that the gene frequency and growth-rate fluctuate within the two extreme equilibrium values; when the population is switched into a new state, it starts moving towards the corresponding equilibrium, but does not have time to reach it before the environment changes again. The reason that allele $A_1$ is favoured when the population is declining but $A_2$ is favoured in the period of population increase, is that the $A_2A_2$ homozygotes reproduce relatively early. The weighting factor $e^{-qx}$ in the equilibrium fitness in equation (3.31) gives increased weight to early as opposed to late reproduction when $\hat{r}$ is positive, and vice versa when $\hat{r}$ is negative. From equation (3.38), one would therefore expect $A_1$ to be most frequent in populations with low $\hat{r}$, since the relative values of the $k_{ij}(x)$ are independent of the environmental state in this model.

This type of model is of interest in relation to observations on natural populations, where fluctuations in allele frequencies have often been reported (e.g. Dobzhansky, 1943; Dubinin and Tiniakov, 1945; Gershenson, 1945; Tamarin and Krebs, 1969; Gaines and Krebs, 1971; Gaines, McCleaghan and Rose, 1978). The observations reported by Gaines and co-workers on the transferrin locus in Microtus ochrogaster are of especial interest in this context, since they find that $Tf^E/Tf^E$ homozygotes at this diallelic locus show higher growth-rates than other genotypes in males sampled from populations in both Indiana and Kansas. Since the onset of sexual maturity is largely determined by body weight in small mammals, this suggests that males of this genotype may enter reproductive life earlier than the others. This is a species which shows population cycling, and there is evidence that the $Tf^E$ allele tends to increase in frequency during the phase of population increase.

Although it is not possible in such cases of gene frequency fluctuations to exclude other interpretations, such as selective responses to environ-
Biological applications of the results

mental factors like temperature or to population density itself (cf. King and Anderson, 1971; Charlesworth and Giesel, 1972b), nevertheless investigators should bear in mind the possibility that gene-frequency changes may be due to a purely mechanical shift in genotypic fitness as a result of changing demography, and are not necessarily caused by changes in the selection regime at the level of age-specific survival probabilities and fecundities.

The idea that genotypic fitnesses in an age-structured population may depend on the overall demographic structure of the population is not a new one. For instance, Bodmer (1968) suggested that the fitnesses of sufferers from genetic diseases with a late age of onset may change in response to changes in the mean age of reproduction of the whole population. The selection models and resulting expressions for equilibrium fitness described in the present chapter provide, however, an exact genetic foundation for analysing the effects of demography on fitnesses, as will be seen in section 3.4.1.

3.4.1 Factors influencing relative fitnesses

The way in which demographic factors can influence relative genotypic fitnesses, and thereby alter the genetic composition of equilibrium populations, can be understood in terms of the equations for equilibrium genotypic fitnesses, such as (3.31) and (3.33). For simplicity, the former is used here as a basis of discussion. Let us first consider the effect of a change in an environmental source of mortality, which affects the \( l(x) \) functions of the various genotypes in a similar way. If \( l_s(x) \) is the probability of survival to age \( x \) for a standard genotype, we can write

\[
l_{ij}(x) = l_s(x)l'_{ij}(x)
\]

where \( l'_{ij}(x) \) is a function which characterises \( A_iA_j \), and \( l_s(x) \) is common to all genotypes. Changes in the level of an external source of mortality which affects each genotype equally can be absorbed into \( l_s(x) \). For an equilibrium population in stable age-distribution, the fitness of \( A_iA_j \) is given by

\[
w_{ij} = \sum_x e^{-\rho x} l_s(x)l'_{ij}(x)m_y(x)
\] (3.47)

If the population experiences a change in the level of mortality, there may be a change in its overall age-distribution when equilibrium is re-established. This would happen if, for instance, the population was held in check by density-dependent factors affecting the survival of immature stages, and the survival probability of each adult age-class were
altered ( \( \hat{r} \) in this case would be zero for the old and new equilibrium). Any such change will normally be expressed as a shift in \( e^{-\hat{r}x} l_s(x) \) as a function of age (cf. section 1.3.2). If the age-structure were shifted in favour of younger reproductive individuals, \( e^{-\hat{r}x} l_s(x) \) must be changed in such a way as to give greater weight to the \( l'_{ij}(x)m_{ij}(x) \) for younger age-classes in equation (3.47). Genotypes which have relatively high \( k_{ij}(x) \) functions for younger ages, such as genotypes which initiate reproduction early in life or attain their peak fecundity early, would tend to have higher relative fitnesses in the new equilibrium population. The opposite would obviously be true if the age-structure of the population became weighted in favour of older individuals due to a relaxation in mortality.

A very similar analysis can be carried out for the case of a change in an environmental factor affecting age-specific fecundities, as in the computer model described above. Here, we can write \( m_s(x) \) for the fecundity function of the standard genotype, and \( m'_{ij}(x) \) for the fecundity function characterising \( A_iA_j \), so that

\[
\begin{align*}
    w_{ij} &= \sum_x e^{-\hat{r}x} l'_{ij}(x)m_s(x)m'_{ij}(x) \\
    \text{(3.48)}
\end{align*}
\]

If the population is density-independent, an increase in \( m_s(x) \) at each age results in an increase in \( \hat{r} \), with a shift in the overall age-structure towards younger individuals which is reflected in \( e^{-\hat{r}x} l_s(x) \), and conversely if \( m_s(x) \) is reduced. A change in the age-structure is therefore reflected in a change in the weighting of reproduction with respect to age.

On the other hand, changes in mortality or fecundity which do not result in any changes in the age-structure of the reproductive age-classes fail to produce alterations in the weighting of reproduction at different ages in equations (3.47) or (3.48), unless the changes are directed specifically at certain age-groups. For instance, if the same extrinsic probability of death were added to each age-class of each genotype, its effect on age-structure in a density-independent environment would be exactly cancelled out by a change in \( \hat{r} \) (cf. Chapter 1, pp. 31–2). There would be no change in \( e^{-\hat{r}x} l_s(x) \) and hence no change in the weighting of reproduction in equation (3.48). Similarly, in a population which is regulated by density-dependent mortality of immature stages, an increase in \( m_s(x) \) for each age would be compensated for by an increased mortality of the immature individuals, resulting in no change in \( \hat{r} \), no change in the age-structure of reproductively active stages, and no increased weighting of early reproduction. Only if the increase in \( m_s(x) \) were concentrated in early or late reproductive age-classes would there be
a change in weighting without a change in the age-structure of the reproductive part of the population.

These considerations suggest that demographic changes which affect the age-structure of the reproductively active part of the population can cause shifts in gene frequencies at loci under selective control, provided that the relative values of the \( k(x) \) functions for different genotypes vary sufficiently with age. Conversely, demographic changes which leave the age-structure of the reproductive age-classes unchanged are unlikely to cause gene-frequency changes, unless they are due to a source of mortality or a factor affecting fecundity which is directed specifically at certain genotypes or at a specific group of ages of the reproductive individuals. Such demographically induced changes in relative fitnesses and gene frequencies cannot occur to any significant extent if the \( k(x) \) functions are close to satisfying equation (3.25), i.e. if selection is non-specific with respect to age. As shown by Charlesworth and Giesel (1972a), selective situations in which the shapes of the \( m(x) \) functions with respect to age vary between genotypes are highly favourable for this effect. Mortality differences between genotypes are unlikely to result in shifts in relative fitnesses with demographic changes, unless there are pronounced genotypic differences in the patterns with respect to age of the survival probabilities from one age-class to the next, among the reproductively active age-classes. Such a case occurs, for example, if one genotype is afflicted with a reduced survival probability, but the onset of this is delayed until well after the beginning of reproduction.

Shifts in gene frequencies due to demographic changes might well be expected in populations which have colonised a new habitat where density-dependent restraints are lifted and there is a high population growth-rate. Although one would probably expect only a limited number of loci to be under sufficiently powerful selective control of a suitable sort for such changes to be induced, hitch-hiking effects on closely-linked loci (Maynard Smith and Haigh, 1974; Kaplan, Hudson and Langley, 1989; Stephan, Wiehe and Lenz, 1992) might cause additional changes in the composition of the gene pool.

3.4.2 The measurement of fitness in human populations

The results derived in this chapter are of obvious relevance to human geneticists interested in measuring genotypic fitnesses, for example in connection with the ‘indirect’ method of estimating mutation rates from the theoretical formulae for the balance between mutation and selection (equations (3.46)). For such purposes, it is usually assumed that the population is at equilibrium, so that the fitness measures described in
this chapter are appropriate. Since human populations generally show marked differences in vital statistics between males and females, the more general fitness measure of equation (3.33) is preferable, assuming that the necessary information on the vital statistics of the genotypes involved is available. It should be borne in mind that this formula is strictly valid only when the effects of non-random mating with respect to age, and of sex differences in gene frequencies, can be treated as second-order (section 3.2.4). As pointed out in section 4.3.1, this is necessarily the case for a rare non-recessive gene, so that studies of selection on rare dominant or semi-dominant alleles can be carried out without serious error, using equation (3.33). For rare recessives, or polymorphic loci, the use of the formula requires relatively small selection intensities.

The greatest difficulty in using the formula is that detailed information on the vital statistics of the genotypes concerned is rarely available. Cavalli-Sforza and Bodmer (1971, Chapter 6) review various indirect methods of obtaining vital statistics for genetic traits. Some examples of the applications of these methods in connection with the equilibrium fitness measure described here are given by Charlesworth and Charlesworth (1973). We shall briefly consider some of their findings.

One of the few human genetic traits for which a reasonably complete life-table has been estimated directly is the Marfan syndrome, a disease of the connective tissue and vascular system which is controlled by a rare dominant gene. Figure 3.2 shows the life-tables obtained by Murdoch et al. (1972) for U.S. men and women suffering from this disorder, together
Biological applications of the results

with normal life-tables for the same period of time. It will be seen that the disease first manifests itself as an increased probability of death early in reproductive life; the mean age of death is 32. No information on the age-specific fecundities of sufferers from the syndrome appears to be available, so that we cannot find their fitness directly; but an overestimate can be obtained by assigning them the same \( m(x) \) and \( m^*(x) \) functions as normal individuals. Using census data for the U.S. population of 1964, and the \( \hat{p} \) value corresponding to the female data for that year, we find that \( \sum e^{-\hat{p}x}k(x) \) for the Marfan syndrome is 0.91, and \( \sum e^{-\hat{p}x}k^*(x) = 0.78 \), giving an overall fitness value of 0.85. The relatively small selection coefficient of 0.15 against this trait, despite its substantial effect on the mortality of adults, is due to its delayed age of onset. Nevertheless, fitness is relatively insensitive to changing demography in this case. If the vital statistics for the 1939–41 U.S. population are used as a basis for calculation, the fitness value is almost unchanged, despite the fact that, because of lower fecundity, this population has an intrinsic rate of increase of \(-0.002\) per year compared with \(+0.016\) for 1964. This illustrates the conclusion reached above concerning the need for strong differences between the mortality patterns with respect to age of different genotypes during reproductive life if demographic structure is to affect relative fitnesses. The time of onset of the mortality effect of the Marfan syndrome coincides approximately with the onset of reproduction, and so this condition is not satisfied.

This insensitivity to demographic change is not always the case, however. Charlesworth and Charlesworth (1973) examined the fitnesses of carriers of Huntington’s chorea, another rare dominant disease. This is a disease with a late and variable age of onset. Cavalli-Sforza and Bodmer (1971) suggested that a useful approximate estimate of the effect of the disease on fitness could be obtained by assuming that sufferers cease to reproduce as soon as it is manifested. Using the data of Reed and Chandler (1958) and Reed and Neel (1959) on the distribution of age of onset in a U.S. population, it is possible to calculate the fitness of choreics relative to normal individuals, using vital statistics for different populations. The results of these calculations are shown in Table 3.3. Only the female fitnesses, \( \sum e^{-\hat{p}x}k(x) \), were calculated in this case.

It is apparent that the high-mortality, rapidly-growing Taiwan population of 1906 has the lowest selection coefficient against Huntington’s chorea. This is what one would expect from the earlier discussion, since such demographic conditions place the least weight on reproduction late in life, whereas early reproduction is not greatly affected in sufferers from this disease.
There are numerous other diseases in man which either have a simple mode of genetic determination, or for which there is evidence of the involvement of genetic factors, and which do not manifest themselves until after the start of reproductive life (Finch, 1990, Chapter 6). Examples of the former include genetically controlled cancers such as multiple polyposis of the colon (Reed and Neel, 1955) and of the latter, such traits as high blood pressure (Cavalli-Sforza and Bodmer, 1971, pp. 583–5). In all these cases, we may expect that the reduction in overall mortality rates which has accompanied advancing civilisation will have increased the selection pressure against the genes that are implicated in such conditions. Their frequencies in contemporary populations may therefore be far from reflecting the equilibrium which they would eventually reach if demographic change were to cease. The estimation of mutation rates from the formulae for mutation-selection balance (Cavalli-Sforza and Bodmer, 1971, pp. 88–92) is, accordingly, especially dangerous for genes affecting characters of this sort.

**Approximate estimators of fitness**

In most cases, even such indirect approaches to determining genotypic vital statistics are not possible, and various approximations have to be resorted to. Provided that the population is not growing too fast, the mean completed family size, discounting individuals who die before reproductive maturity, should provide a reasonably good approximation.

This is equivalent to the net reproduction rate, $R$ (section 1.3.2), which is widely used as a fitness measure in studies of natural populations (Clutton-Brock, 1988). If all that is known is that a mutant phenotype suffers an enhanced average probability of death of $Q$ per age-class, without any information on the relationship of $Q$ to age, it is possible to approximate $w_{ij}$ in equation (3.33) by

$$
\frac{1}{2} \sum e^{-rx} (1 - Q)^x [k(x) + k^*(x)] = 1 - \frac{1}{2} Q \sum x e^{-rx} [k(x) + k^*(x)]
$$
where \( k(x) \) and \( k^*(x) \) are given by the vital statistics for the normal population. The multiplier of \( Q \) on the right-hand side of this expression is the mean generation-time of the population (cf. equation (1.47c)). Similarly, if all that is known is that the mutant has a fecundity of \( 1 - K \) relative to normal, the best estimate of its fitness is \( 1 - K \).

Charlesworth and Charlesworth (1973) have discussed various other measures of genotypic fitnesses which have been proposed by other authors, and compared them with the values given by equation (3.31), for the case of Huntington's chorea. The measure discussed here has the definite advantage of being based on a clear-cut genetic model, whereas earlier measures rest on somewhat intuitive grounds. Provided that adequate data are available, there is no greater computational difficulty in using equations (3.31) or (3.33) than the less well-founded measures, with modern computing methods. Of course, the inherent sampling error and biases in the available data will usually mean that the use of an incorrect fitness measure is only one of many sources of error in estimation.
Selection: dynamic aspects

4.1 Introduction
This chapter is concerned primarily with the analysis of the process of gene frequency change under selection in age-structured populations, in particular with the question: to what extent can this process be understood in terms of fixed parameters which characterise individual genotypes, analogous to the fitnesses of genotypes in discrete-generation models? The first investigation of the problem of the dynamics of selection in the context of age-structured populations was that of Norton (1928). (According to Haldane (1927a), Norton started work on the problem as early as 1910 and had established his results by 1922.) He studied the case of a single diallelic locus with time-independent and density-independent demographic parameters for each genotype, with no sex-differences, using the continuous-time equivalents of equations (3.14). He showed that the long-term results of selection (whether one of a pair of alleles is fixed or a polymorphism is maintained) are completely predicted by the relationships between the intrinsic rates of increase of the three genotypes, in the same way that the outcome of selection can be predicted from the genotypic fitnesses in the discrete-generation case. Haldane (1927a) established some approximate results for the rates of spread of rare dominant and recessive genes in slowly growing populations, and concluded that selection proceeded in much the same way as in the discrete-generation case. Both Haldane and Norton based their work on equations which explicitly incorporated the characteristics of age-structured populations, in contrast to the approach of Fisher (1930, 1941) who used differential equations to describe gene frequency change with continuous-time models but did not derive these from more basic equations.

Later work (Haldane, 1962; Pollak and Kempthorne, 1971; Charlesworth, 1973, 1974a, 1976) has established that no single parameter can be regarded as the fitness of a genotype with arbitrary selection intensities. When selection is weak, however, it is possible to show that the intrinsic rate of increase of a genotype can be equated with its fitness, with time-independent and density-independent demographic parameters
Approximate equations with weak selection

Charlesworth, 1974a, 1976. Useful results can also be obtained for the density-dependent case with weak selection.

In this chapter, we start by examining how the basic equations for gene frequency and population size, such as equations (3.12) and (3.14), can be approximated with weak selection to yield equations which are similar in form to those for the usual discrete-generation models. We then go on to consider the case of arbitrary selection intensities, by means of the local stability properties of equilibria. Selection on rare alleles with variable environments, and the probability of survival of a favourable mutant gene, are treated. Selection on quantitative characters is also considered. Finally, the relation between evolutionary optimisation theory and the population genetics of selection with age-structure is discussed.

4.2 Approximate equations with weak selection

Most of the models in this section are developed in terms of a single autosomal locus with an arbitrary number, \( n \), of alleles, as in the preceding chapter, although the case of two loci is considered briefly. In addition, discussion is largely confined to the case of equations (3.14), with some mention of the problem of sex differences in demographic parameters. Both density-independent populations (with time-independent \( k_{ij}(x) \) functions) and density-dependent populations are considered. Before proceeding to the detailed results, it is useful to have a clear definition of what is meant by weak selection. We choose some genotype as an arbitrary standard, and write its reproductive function as \( k_s(x, t) \).

The reproductive function for another genotype \( A_iA_j \) can be written as \( k_{ij}(x, t) = k_s(x, t) + \epsilon \eta(x, t) \), where \( |\epsilon \eta(x, t)| < \epsilon \) for all \( i, j \) and \( x \), \( \epsilon \) is a positive number which measures the strength of selection. Selection is said to be weak when terms \( O(\epsilon^2) \) can be neglected, compared with terms \( O(\epsilon) \).

4.2.1 The density-independent case

Approximate equations for \( B(t) \)

The first step in obtaining expressions for changes in gene frequency and population growth-rate is to obtain an approximation for \( B(t) \). From equation (3.14b) we have the exact expression for \( t \geq d \)

\[
B(t) = \sum_x B(t - x) \sum_{ij} p_i(t - x)p_j(t - x)k_{ij}(x)
\]  

† As is conventional in mathematical usage, a function of \( \epsilon, f(\epsilon) \), is said to be of order \( \epsilon \) or \( O(\epsilon) \) if \( |f(\epsilon)/\epsilon| < \alpha \) when \( \epsilon \) is taken sufficiently small, where \( \alpha \) is a constant independent of \( \epsilon \). If \( |f(\epsilon)/\epsilon| \) approaches 0 as \( \epsilon \to 0 \), we write \( f(\epsilon) = o(\epsilon) \).
In constructing an approximate solution to this equation, we need the result that the change in gene frequency between times $t$ and $t+1$ (more technically, the first difference in gene frequency), which we can write as $\Delta p_i(t)$ for allele $A_i$, is of order $\epsilon$ for sufficiently large $t$. This result, which is what one intuitively expects, is derived in Appendix 2. As mentioned there, convergence to terms of order $\epsilon$ should normally be fairly rapid.

We now choose a time $t_0$ which is sufficiently large that convergence of the $\Delta p_i$ has taken place for each allele, and consider a time $t$ such that $t_0$ is previous to $t-d$. We can write $p_i(t-x) = p_i(t_0) + \xi_i(t-x)$, where $\xi_i$ is $O(\epsilon)$ if $t$ and $t_0$ are sufficiently close, since $\xi_i$ is simply the sum of the $\Delta p_i$ between $t_0$ and $t-x-1$. Equation (4.1) can therefore be rewritten as

$$B(t) = \sum_x B(t-x) \left[ \sum_{ij} p_i(t_0) p_j(t_0) k_{ij}(x) + 2 \sum_{ij} \xi_i(t-x) p_j(t_0) k_{ij}(x) \right] + O(\epsilon^2) \quad (4.2)$$

But $\sum_{ij} p_j(t_0) k_{ij}(x) = \sum_{ij} p_i(t_0) p_j(t_0) k_{ij}(x) + O(\epsilon)$. Substituting this into equation (4.2), and using the relation $\sum_i \xi_i = 0$, we obtain

$$B(t) = \sum_x B(t-x) \sum_{ij} p_i(t_0) p_j(t_0) [k_{ij}(x) + O(\epsilon^2)] \quad (4.3)$$

Applying to this equation the type of argument used in Appendix 2, we obtain $\Delta \ln B \sim r_0 + O(\epsilon^2)$, so that

$$B(t-x) \sim B(t) e^{-r_0 x} + O(\epsilon^2) \quad (4.4)$$

where $r_0$ is the intrinsic rate of increase associated with the mean reproductive function for time $t_0$, $\sum_{ij} p_i(t_0) p_j(t_0) k_{ij}(x)$. In other words, $r_0$ satisfies the equation

$$\sum_x e^{-z x} \sum_{ij} p_i(t_0) p_j(t_0) k_{ij}(x) = 1$$

If terms $O(\epsilon^2)$ are neglected, this gives us the approximate asymptotic result which we need. This can validly be done if selection is sufficiently weak ($\epsilon$ is sufficiently small), so that second-order terms can be neglected in the total gene frequency change which takes place over the period of time needed to stabilise the age-structure of a population with a constant reproductive function. In most cases of biological interest this should not be a long time, so that the condition is not unduly restrictive.
Intuitively, this result can be viewed in the following way. Equation (4.3) shows that the effect of selection on the mean reproductive function of the population is second order compared with the selection intensity, so that the population can be treated as though it has a fixed reproductive function and hence a fixed rate of population growth over a sufficiently short period of time.† This can be compared with the result for discrete-generation models that the rate of change of population mean fitness is second-order, with respect to the fitness differences between genotypes (Crow and Kimura, 1970, pp. 208–9; Nagylaki, 1992, Chapter 4).

**Approximate equations for** $\Delta p_i$: **type 1**

From equation (3.14a) we have the gene frequency equation for $t \geq d$

$$B(t)p_i(t) = \sum_x B(t-x)p_i(t-x)\sum_j p_j(t-x)k_{ij}(x) \quad (4.5)$$

Substituting from equation (4.4), this becomes (for $t$ sufficiently close to $t_0$)

$$p_i(t) = \sum_x e^{-r_0x}p_i(t-x)\sum_j p_j(t-x)k_{ij}(x) + O(\varepsilon^2) \quad (4.6)$$

This equation can be simplified further if we approximate $p_i(t-x)$ by $p_i(t) - x\Delta p_i(t)$. This type of approximation was first used in the context of models of age-structured populations by Haldane (1927a, 1962). It is valid under conditions discussed in Appendix 2. Substituting into equation (4.6), and dropping the arguments from $p_i(t)$ and $\Delta p_i(t)$, we obtain

$$p_i = p_i\sum_x e^{-r_0x}\sum_j p_jk_{ij}(x) - \Delta p_i\sum_x e^{-r_0x}\sum_j p_jk_{ij}(x) + O(\varepsilon^2) \quad (4.7)$$

In order to proceed further with this equation it is necessary to eliminate the unknown quantity $r_0$, which depends on the arbitrary choice of time $t_0$. This can be done by again using the argument which led from equations (4.1) to (4.3); we have

$$\sum_{ij} p_i(t_0)p_j(t_0)k_{ij}(x) = \sum_{ij} p_ip_jk_{ij}(x) + O(\varepsilon^2)$$

† Of course, as time passes, $t - t_0$ will become so large that the $O(\varepsilon^2)$ terms in equation (4.4) can no longer be neglected. This problem can be overcome by shifting to a later time $t_1$, in place of $t_0$, to which there corresponds a different intrinsic rate of increase $r_1$ (cf. p. 140). Provided that $\varepsilon$ is sufficiently small, and the demography of the population is such that the time needed to stabilise age-structure is small compared with the time needed for a change in gene frequency of $O(\varepsilon)$, there will always be some intrinsic rate of increase which can be substituted into equation (4.4).
This means that, to order $e^2$, $r_0$ can be replaced in equation (4.7) by $r_p$ as defined by equation (3.35); the subscript $p$ represents the set of gene frequencies at time $t$.

It is useful to define the following quantities:

$$T_{ij}(z) = \sum_x x e^{-zx} k_{ij}(x) \quad (4.8a)$$

$$T(z) = \sum_x x e^{-zx} \sum_{ij} p_i p_j k_{ij}(x) \quad (4.8b)$$

$$w_i(z) = \sum_j p_j w_{ij}(z) \quad (4.8c)$$

where the $w_{ij}(z)$ are given by equation (3.34).

Using these definitions, and replacing $r_0$ by $r_p$, equation (4.7) can be rearranged to give the first type of approximate expression for $\Delta p_i$

$$T(r_p) \Delta p_i = p_i [w_i(r_p) - 1] + O(e^2) \quad (4.9)$$

$T(r_p)$ in this equation may be regarded as a measure of the generation time of the population at time $t$ (cf. Chapter 1, p. 30, and provides a time-scale for the process of gene frequency change.

This equation is closely connected with the equilibrium theory developed in section 3.3.2. When $r_p = \hat{r}$, as given by equations (3.42), we have $w_i = 1$ for the alleles which are present in the population. Hence, the equilibrium solution of equation (4.9) is, if we neglect the $O(e^2)$ terms, identical with that given by the exact equations of Chapter 3. Since the $O(e^2)$ terms in equation (4.9) vanish when allele frequency change ceases and stable age-distribution is reached, it is not surprising that this equation gives an equilibrium solution identical to that given by the approach of Chapter 3 when the $O(e^2)$ terms are omitted. Furthermore, standard theory tells us that, as in the two-allele case discussed on pp. 121-3, this equilibrium corresponds to a stationary point in the function $\bar{w}(p, \hat{r})$ defined by equation (3.35), with respect to variation in gene frequency while holding $r_p$ constant at $\hat{r}$ (cf. Crow and Kimura, 1970, pp. 272-7). This implies that, as in the two-allele case, $r_p$ is at a stationary point at the equilibrium. In the neighbourhood of equilibrium, equations (4.9) can thus be satisfactorily approximated by substituting $\hat{r}$ for $r_p$.

It follows that the $w_{ij}(\hat{r})$ control the local stability of the equilibrium in the same way as the fitnesses $W_{ij}$ in the discrete-generation case. From the standard theory of the discrete-generation case, the mean fitness of the population is at a global maximum at a stable equilibrium point with all alleles present, and there is only one such equilibrium. Comparing the
discrete-generation case with the present system, using the $w^f(\hat{r})$ as
fitness parameters, we see that $\bar{w}(p, \hat{r})$ is at a global maximum at a stable
equilibrium point, neglecting the $O(\epsilon^2)$ terms; it follows that $\hat{r}$ is a
maximum of $r_p$ at such a point, and that there can be at most one stable
equilibrium with all alleles present. In the case of a diallelic locus, it
follows that the equilibrium is locally stable if $r_{12} > r_{11}, r_{22}$, and is
unstable if $r_{12} < r_{11}, r_{22}$. Study of equation (4.9) shows that these
relations also determine the global stability conditions with two alleles
and weak selection (cf. Charlesworth, 1974a).

Approximate equations for $\Delta p_i$: type 2

Although equation (4.9) provides a useful characterisation of the
dynamics of the system, it suffers from the weakness that $r_p$ is itself an
implicit function of the gene frequencies, so that it is tedious to calculate
the rate of change of gene frequencies at any given point. We now
develop a further approximation to this equation, which yields an ex-
pression for $\Delta p_i$ that depends only on gene frequency and on the
genotypic intrinsic rates of increase.

We note that we can approximate $w_i^f(r_p)$ (given by equation (3.34)),
using Taylor's theorem. We have

$$1 = w_i^f(r_{ij}) = w_i^f(r_p) + (r_{ij} - r_p)(\partial w_i^f/\partial z)_{r_p} + O(\epsilon^2)$$

so that

$$w_i^f(r_p) - 1 = (r_{ij} - r_p)T_{ij}(r_p) + O(\epsilon^2) \quad (4.10)$$

Substituting this into equation (4.9), and noting that $T_{ij}(r_p) =
T(r_p) + O(\epsilon)$, we obtain the final result

$$\Delta p_i = p_i(r_i - \tilde{r}) + O(\epsilon^2) \quad (4.11)$$

where

$$r_i = \sum_j p_j r_{ij} \quad \text{and} \quad \tilde{r} = \sum_{ij} p_i p_j r_{ij}.$$ 

An analogous procedure can be gone through for the continuous-time
model, for which we obtain

$$\frac{dp_i}{dt} = p_i(r_i - \tilde{r}) + O(\epsilon^2) \quad (4.12)$$

where the genotypic intrinsic rates of increase are now defined by the
integral equation analogue of equation (3.34).

It is clear from these results that, when selection is sufficiently weak,
the use of the genotypic intrinsic rates of increase provides a useful
Selection dynamics

approximate method of calculating rates of gene frequency change in a
density-independent environment, with the \( r_{ij} \) playing much the same role
as the \( W_{ij} \) in the discrete-generation case with weak selection (in which
case the normalising term \( \bar{W} \) on the right-hand side of equations (3.15)
can be dropped; see Crow and Kimura, 1970, p. 191). Equation (4.12)
with weak selection is also similar in form to Fisher's Malthusian para-
meter equation for gene-frequency change (Fisher, 1930, 1941). If selec-
tion is too strong, however, the \( O(\varepsilon^2) \) terms in equations (4.11) and
(4.12) may become significant, and one cannot expect the main term to
provide an adequate approximation.

Rate of change of the population growth-rate

The similarity between equation (4.11) and the more usual equa-
tions for gene-frequency change suggests that it should be possible to
derive an expression for the rate of change of the population growth-rate,
analogous to that given by Fisher's Fundamental Theorem of natural
selection. This can be done as follows. Ignore for the moment the \( O(\varepsilon^2) \)
terms in the equations. The results given above show that the rate of
growth of a population at time \( t \) is approximately \( r_p \). Using equation
(4.10) it is easy to see that \( r_p \) can be approximated to \( O(\varepsilon^2) \) by \( \bar{r} \), so that
we have (approximately)

\[
\Delta^2 \ln B(t) = \Delta \bar{r}
\]  

(4.13)

By standard theory (Crow and Kimura, 1970, pp. 205–10), we have

\[
\Delta \bar{r} = 2 \sum_i p_i (r_i - \bar{r})^2 + O(\varepsilon^3)
\]  

(4.14)

where the term on the right-hand side is the additive genetic variance in
the intrinsic rate of increase. Putting equations (4.13) and (4.14) together
gives us a version of Fisher's Fundamental Theorem: the rate of change of
the population growth-rate is approximately equal to the additive genetic
variance in the intrinsic rate of increase.

This result as it stands is purely formal, however, since terms in \( \varepsilon^2 \) were
neglected before taking the differences in equation (4.13); if the first
differences in these terms were themselves of order \( \varepsilon^2 \), they would
contribute significantly to the final result and destroy the approximation.
Retracing our steps through the successive approximations, it is possible
to show, by arguments similar to those of Appendix 2, that the first
differences in the \( O(\varepsilon^2) \) terms are themselves \( O(\varepsilon^3) \), so that equations
(4.13) and (4.14) combined give

\[
\Delta^2 \ln B(t) = 2 \sum_i p_i (r_i - \bar{r})^2 + O(\varepsilon^3)
\]  

(4.15)
Accuracy of the approximations

The utility of the results derived above depends on being able to neglect the \(O(e^2)\) and \(O(e^3)\) terms in the equations for gene frequency change and change in population growth-rate. It is obviously important to determine in practice how weak selection must be for this to be done without significant error. This question was investigated numerically by Charlesworth (1974a), who compared the results obtained from exact population trajectories with the approximate values of \(\Delta p_i\) and \(\Delta^2 \ln B\) obtained from equations (4.9), (4.11) and (4.15). He used a ten age-class system with the \(T_{ij}\) of equation (4.8a) all approximately equal to 4. A crude measure of the strength of selection in per generation terms is therefore obtainable by multiplying the differences in genotypic intrinsic rates of increase by 4. Cases of dominant, recessive and semi-dominant favourable alleles in diallelic systems, as well as heterozygote superiority, were studied. Even with selection differences of up to 40% per generation, there is fairly good agreement between the exact and approximate values for \(\Delta p_i\) and \(\Delta^2 \ln B\), after a few generations have elapsed and the age-structure of the population has stabilised. As might be expected from its relationship to the exact equilibrium equations, equation (4.9) seems to provide a better approximation than (4.11) for the neighbourhood of a polymorphic equilibrium. Away from such a point, equation (4.11) provides, if anything, a better fit. Since it does not depend on any parameters other than the gene frequencies and the \(r, y\), it is to be preferred to equation (4.9). In using these approximate equations, it should be noted that they are designed to predict the rates of change of gene frequencies and population growth-rate, given the current gene frequencies. It cannot be expected that they will necessarily provide an accurate basis for calculating a whole population trajectory, since the errors will tend to accumulate over successive time-intervals.

Sex differences in demographic parameters

The problem of handling sex differences in demographic parameters, which was avoided in the above treatment by basing it on equations (3.14), can be approached in the following way when there is random mating with respect to age (cf. Charlesworth, 1974a). Consider an arbitrary standard and homogeneous population with female reproductive function \(k_s(x)\) and growing at its intrinsic rate of increase \(r\). Each male genotype can be characterised by its age-specific fecundity function when introduced at a very low frequency into this population; from the assumption of a stable age-distribution, these fecundities will be independent of time. Denote the 'standard' reproductive function for \(A_tA_j\).
males as given by this procedure by \( k^*_y(x) \), with a corresponding intrinsic rate of increase \( r^*_y \). In a real population, with a mixture of genotypes changing in frequency under selection, the \( k^*_y(x, t) \) functions at a given time, \( t \), must differ somewhat from their standard values. If mating is at random with respect to age, and the other assumptions about the mating process which lead to equations (3.10) are fulfilled, we can write

\[
k^*_y(x, t) = k^*_y(x)(1 + \xi_t)
\]

where \( \xi_t \) is the same for each genotype, but in general changes with \( t \). Using the assumption of weak selection, terms of order \( \xi_t^2 \) can be neglected, since differences in the male reproductive functions between different populations result from differences in the composition of the female population, and these are \( O(\varepsilon) \).

Given these assumptions, the value of \( \xi_t \) at a time \( t \) close to some arbitrarily chosen fixed time \( t_0 \) is equal to \( \xi_0 + O(\varepsilon^2) \), where \( \xi_0 \) is given by

\[
(1 + \xi_0) \sum_x e^{-rx} \sum_{ij} p_i(t_0) p_j(t_0) k^*_y(x) = 1
\]

(4.16)

\( r_0 \) is defined on p. 138).

By the same type of argument that led to equation (4.11), it is possible to use equations (3.10) to obtain an approximate equation for the first difference in \( p_t = \{p_t - p_0\} \). We define a new intrinsic rate \( r^*_y \) for \( A_y A_y \) males as the root of

\[
(1 + \xi_0) \sum_x e^{-sx} k^*_y(x) = 1
\]

It turns out that \( \Delta \bar{p}_i \) is determined by the \( \bar{p}_i \) and the functions \( \frac{1}{2}(r_{ij} + r^*_y) \) in the same way that \( \Delta p_i \) in equation (4.11) is determined by the \( p_j \) and \( r_{ij} \). Furthermore, if we write \( T^*_y = \sum x e^{-rx} k^*_y(x) \), it follows from the Taylor's expansion of the equation for \( r^*_y \) that

\[
r^*_y = r_{ij} + (\xi_0 / T^*_y) + O(\xi_0^2)
\]

Neglecting terms in \( \varepsilon^2 \), \( r^*_y \) for any genotype is thus equal to \( r_{ij}^* \) plus a term common to each genotype. Hence, \( \Delta \bar{p}_i \) is determined by the fitness functions \( r_{ij} = \frac{1}{2}(r_{ij} + r^*_y) \) for each genotype, to order \( \varepsilon^2 \). We have

\[
\Delta \bar{p}_i = \bar{p}_i \left( \bar{r}_i - \sum \bar{p}_i \bar{r}_i \right) + O(\varepsilon^2)
\]

(4.17)

where

\[
\bar{r}_i = \sum_j \bar{p}_j \bar{r}_{ij}
\]
The mean of the intrinsic rates of the males and females thus provides a measure of the fitness of a genotype. This is similar to the result for weak selection in the discrete-generation case, where the means of the male and female fitness values control gene frequency change (Wright, 1942; Nagylaki, 1979b). Abugov (1986) has developed an alternative method for dealing with sex-differences in fitness with age-structure.

**The two-locus case**

The two-locus case can be approached in a similar way to the one-locus model; using equations (3.17) we have for \( t \geq d \)

\[
B(t)p_i(t) = \sum_x B(t - x)p_i(t - x)\sum_j p_j(t - x)k_{ij}(x)
\]

\[
\pm c \sum_x B(t - x)D(t - x)k_{14}(x)
\]

(4.18)

where the sign attached to \( c \) is positive for gametes 2 and 3, and negative for gametes 1 and 4.

When \( c \) is close to zero, the system can be treated as a multiple-allele systems with four alleles, so that with weak selection, equations similar to (4.9), (4.11) and (4.15) must be obeyed. When linkage is loose in relation to the intensity of selection, so that \( \epsilon \ll c \), the following approach can be used, which is similar to that of Nagylaki (1976, 1992, Chapter 8) for the discrete-generation case.

Writing \( k_s(x) \) for the reproductive function of a standard genotype, as before, equation (4.18) becomes

\[
B(t)p_i(t) = \sum_x B(t - x)p_i(t - x)k_s(x)
\]

\[
\pm c \sum_x B(t - x)D(t - x)k_s(x) + O(\epsilon)
\]

(4.19)

Using equation (2.24) we can write

\[
|D(t_0)| \leq (1 - c)|D(t_1)| + O(\epsilon)
\]

where \( t_0 \) and \( t_1 (t_1 < t_0) \) are two times separated by approximately one generation. Extending this back over a sequence of past times \( t_2, t_3, \ldots, t_l \) \( (t_{l+1} < t_l) \), as in equation (2.25), we obtain

\[
|D(t_0)| \leq (1 - c)^l|D(t_l)|
\]

\[
+ [(1 - c)^l + (1 - c)^{l-1} + \ldots + (1 - c)]O(\epsilon)
\]

\[
= (1 - c)^l|D(t_l)| + ((1 - c)[1 - (1 - c)^l]/c)O(\epsilon)
\]
Hence, if $t_0$ is taken sufficiently large, we have

$$|D(t_0)| \sim O(\varepsilon)$$

Substituting this into equation (4.19), we have for $t > t_0$

$$B(t)p_i(t) = \sum_x B(t - x)p_i(t - x)[k(x) + O(\varepsilon)]$$

Using the argument of Appendix 2, we conclude that $\Delta p_i = O(\varepsilon)$ and $\Delta^2 p_i \sim O(\varepsilon^2)$. Since $D \sim O(\varepsilon)$, $\Delta D$ for large $t$ depends on the first differences of $O(\varepsilon)$ terms. By the methods of Appendix 2, these can be shown to be $O(\varepsilon^2)$, so that $\Delta D \sim O(\varepsilon^2)$. Using the same type of argument that led to equation (4.11), we can reduce equation (4.18) to the first-order form

$$\Delta p_i = p_i(r_i - \bar{r}) \pm (cD/Ts) + O(\varepsilon^2) \tag{4.20}$$

where

$$Ts = \sum x e^{-r_i^x}k_s(x)$$

This equation is similar to that for the discrete-generation case when selection is weak compared with the recombination frequency (Nagylaki, 1976, 1992, Chapter 8); the resemblance is perfect if both sides are multiplied by $Ts$, so that changes in gamete frequencies are measured in 'per generation' terms.

This completes the analysis of the case of weak selection with time-independent and density-independent demographic parameters. We have seen how it is possible to obtain approximate equations for this case which resemble closely the familiar equations for discrete-generation selection models, and that the intrinsic rate of increase of a genotype serves as an adequate measure of its fitness in this situation. It is even possible to handle the case of sex differences in demographic parameters, but the effects of non-random mating with respect to age have not been coped with. The significance of these results is discussed in the final section of this chapter. We turn next to the case when population size is limited by density-dependent factors.

### 4.2.2 The density-dependent case

The analysis of this situation is complicated by the fact that genetic changes due to selection must generally result in changes in population size, accompanied by changes in the density-dependent sur-
vival and fecundity components. It turns out, however, that one can obtain results which are very similar to those of the previous sections. The discussion here will be confined to the single-locus model obeying equations (3.14). Time-dependence of the $k_{ij}(x, t)$ is caused by their dependence on the numbers of individuals in the critical age-group at a given time or set of times, $N_T$, as described in sections 1.4.2 and 3.3.3.

**Approximate equations for $\Delta p_i$**

Consider a population at time $t$ with gene frequencies $p_1(t), p_2(t), \ldots, p_n(t)$ among the new zygotes. Assuming random mating with respect to age and genotype, it can be characterised demographically by its mean reproductive function $\sum_{ij} p_i(t)p_j(t)k_{ij}(x, N_T)$. We assume that a genetically homogeneous population with a reproductive function equal to this, and having the same functional relationship to $N_T$, would tend to a stationary size, with a number $N_p$ of individuals in the critical age-group. It is assumed that such convergence to stationarity would take place for all reproductive functions which correspond to possible sets of gene frequencies, although the stationary population size $N_p$ is, of course, a function of gene frequency, given by equation (3.41). It is also necessary to assume that, in a population with changing gene frequencies, the number of individuals in the critical age-group becomes equal to the equilibrium number for a standard genotype, plus terms of order $\varepsilon$, and the population growth-rate becomes $O(\varepsilon)$.† This can be used to show that $\Delta p_i$ is asymptotically $O(\varepsilon)$, and that $\Delta^2 p_i$ is asymptotically $O(\varepsilon^2)$, (see Appendix 2).

The number of individuals in the critical age-group at some time $t$ sufficiently close to a fixed time $t_0$ can, from this, be approximated to $O(\varepsilon^2)$ by methods similar to those used in deriving equation (4.3). This number is given by

$$N(t) = \sum_{x \in S} B(t - x) \sum_{ij} p_i(t_0)p_j(t_0)[l_{ij}(x, N_T) + O(\varepsilon^2)]$$  \hspace{1cm} (4.21a)

and the total number of new zygotes satisfies

$$B(t) = \sum_{x} B(t - x) \sum_{ij} p_i(t_0)p_j(t_0)[k_{ij}(x, N_T) + O(\varepsilon^2)]$$  \hspace{1cm} (4.21b)

† This assumption is tantamount to the density-regulating mechanism being such that a genetically homogeneous population, exposed to continual, externally-imposed perturbations of order $\varepsilon$ to its demographic parameters, will eventually settle down to a constant size, to $O(\varepsilon)$. This size corresponds to its equilibrium size in the absence of perturbations. For a treatment of a density-dependent, discrete-generation model which does not require such an assumption, see Nagylaki (1979a).
Equations (4.21) describe, to order $\varepsilon^2$, the dynamics of a genetically homogeneous population with reproductive function $\sum_{ij} p_i(t_0) p_j(t_0) k_{ij}(x, N_T)$. From the above assumptions about the properties of this case, it follows that

$$N(t) \sim N_0 + O(\varepsilon^2) \quad (4.22a)$$

$$B(t - x) \sim B(t) + O(\varepsilon^2) \quad (4.22b)$$

where $N_0$ is the value of $N$ which satisfies $\sum_{ij} p_i(t_0) p_j(t_0) k_{ij}(x, N) = 1$.

Carrying through the same type of manipulations as were used in the density-independent case, we obtain the gene-frequency equation analogous to equation (4.9)

$$T(N_p) \Delta p_i = p_i [w_i(N_p) - 1] + O(\varepsilon^2) \quad (4.23)$$

We can make the same assumption as in Chapter 3: in a stationary population the net reproduction rate of each genotype is a strictly decreasing function of the number of individuals in the critical age-group. $N_p$ is then given as the unique root of equation (3.41), and $w_i$ and $T$ in equation (4.23) are defined in the same way as in equations (4.8) and (4.9), but substituting $k_{ij}(x, N_p)$ for $e^{-r_x} k_{ij}(x)$.

Equation (4.23) has an obvious resemblance to equation (4.9) for the density-independent case, and exactly the same properties hold for both, substituting the carrying-capacities $N_{ij}$ and $N_p$ of the density-dependent case for the $r_{ij}$ and $r_p$ of the other. There is thus at most one stable equilibrium with all alleles present, which corresponds to a maximum in $N_p$; with a system of two alleles, a stable equilibrium exists if and only if $N_{11}, N_{22} < N_{12}, N_{12}$, etc. Equation (4.23) can be approximated further on lines similar to those which yielded equation (4.11) for the density-independent case, by noting that

$$w_{ij}(N_p) = 1 + (N_p - N_{ij}) \left( \frac{\partial w_{ij}}{\partial N} \right) N_{ij} + O(\varepsilon^2)$$

Choosing an arbitrary genotype as standard, and writing $T_s$ for the value of $\sum x k(x, N_T)$ for a stationary population with the demographic parameters of the standard genotype, and $\partial w_s/\partial N$ for the corresponding value of $\partial w_{ij}/\partial N$, we obtain

$$\Delta p_i = -p_i \frac{(N_i - N) \left( \frac{\partial w_s}{\partial N} \right)}{T_s} + O(\varepsilon^2) \quad (4.24)$$
Local stability analyses

where

\[ N_t = \sum_j p_j N_{ij} \quad \bar{N} = \sum_{ij} p_i p_j N_{ij} \]

Rate of change of population density

It is also possible to derive an equation for the rate of change of \( N(t) \). Going through a similar procedure to that involved in obtaining equation (4.14), but using equation (4.24), we obtain the analogue of equation (4.15)

\[ \Delta N = -2 \left( \frac{\partial w_s}{\partial N} \right) \sum_i p_i \frac{(N_i - \bar{N})^2}{T_s} + O(e^3) \] (4.25)

Equations similar to (4.23), (4.24) and (4.25) can be derived for the discrete-generation case (cf. Charlesworth, 1971; Roughgarden, 1976; León and Charlesworth, 1978; Kimura, 1978; Nagylaki, 1979a, 1992, Chapter 4). They demonstrate that the progress of selection in a density-dependent population is controlled by the genotypic carrying-capacities, the \( N_{ij} \) (see equation (3.40)) and that selection tends to maximise the number of individuals in the critical age-group, subject to the assumption of weak selection. Numerical checks of these equations, based on computer calculations of population trajectories, show that, for moderate selection intensities, these equations provide accurate approximations for the rates of change of gene frequency and population density. For further discussions of the theory of selection in density-dependent populations, see Ginzburg (1983) and Costantino and Desharnais (1991).

4.3 Local stability analyses

The results of the preceding section have shown that, with weak selection, it is possible to develop equations which give good approximations to the changes in gene frequencies and population size. It is obviously important to extend the theoretical analysis of the dynamics of selection to situations where there is arbitrarily strong selection, so that it is not legitimate to neglect second-order terms in the selection intensity. One way of tackling this problem is to carry out local stability analyses, by neglecting second-order terms in deviations of genotypic frequencies from their equilibrium values, in the neighbourhoods of equilibria. The present section describes this approach. In addition to local analyses based on the deterministic equations, the stochastic behaviour of a mutant gene in a large population is examined by branching process
methods. Only the case of a single autosomal locus with two alleles has been studied.

4.3.1 Gene frequencies near 0 or 1 with density independence

A rare, non-recessive allele

This case may be exemplified by considering the fate of allele $A_2$ when introduced at a low frequency into a population fixed for $A_1$. Terms involving the square of the frequency of $A_2$ can be neglected in the initial stages after its introduction, while it remains rare. In principle, this enables us to deal with the case when mating is non-random with respect to age, since by inspection of equations (3.21), we can neglect terms arising from the variance in gene frequency across age-classes, and the second-order terms in the difference between the male and female gene frequencies. This enables us to use equations (3.10) even with non-random mating with respect to age. We have for $t \geq d$

$$B(t) = \sum_x B(t - x) k_{11}(x) + O(p_2)$$  (4.26)

so that asymptotically (cf. Appendix 2)

$$B(t - x) \sim B(t) e^{-r_1 x} + O(p_2)$$  (4.27)

Substituting this into equations (3.10a) and (3.10b), for the two-allele case with time-independent reproductive functions, gives (cf. p. 144)

$$\bar{p}_2(t) = \frac{1}{2} \sum_x e^{-r_1 x} \bar{p}_2(t - x) [k_{12}(x) + k^*_1(x)] + O(\bar{p}_2^2)$$  (4.28)

Note that the $k^*_1(x)$ here are the values derived using the age-specific fecundities of $A_1A_2$ males when mated to females drawn from a wholly $A_1A_1$ population in stable age-distribution. To determine their values in a given case, one would need to specify in detail the rules of mating with respect to age and genotype, or to have empirical data on the fecundities of rare $A_1A_2$ males in a largely $A_1A_1$ population.

The logarithmic rate of increase in frequency of $A_2$ is therefore given asymptotically, to order $O(\bar{p}_2)$, by the real root of

$$\frac{1}{2} \sum_x e^{-(z + r_1)x} [k_{12}(x) + k^*_1(x)] = 1$$  (4.29)

The utility of this result depends on the approach to the asymptotic values for $B(t)$ and $\bar{p}_2(t)$ being sufficiently fast that the latter is still small enough for terms of order $\bar{p}_2^2$ to be negligible after convergence has taken
Local stability analyses

As discussed in section 1.3.2, convergence to the asymptotic solutions of equations of the same form as equations (4.26) and (4.28) is normally a matter of a few generations. Provided \( \bar{p}_2 \) is initially fairly small, there should be no problem. Numerical examples indicate that in practice the result can be used for initial \( \bar{p}_2 \) values of up to 0.01 or so, even with strong selection.

When mating is at random with respect to age and genotype, and when the assumptions of equations (3.13) are satisfied (no mortality differences between males and females of the same age and genotype, and proportionality of male and female fecundities), it is possible to show that \( m_{12}^*(x) = m_{12}(x) \) plus terms of order \( p_2 \). \( k_{12}(x) \) can therefore be substituted for \( k_{12}^*(x) \) in equation (4.29), so that for large \( t \)

\[
\Delta \ln p_2 = (r_{12} - r_{11}) + O(p_2)
\]  

(4.30)

Obviously, the same analysis can be carried out for the other endpoint in gene frequency, \( p_1 = 0 \). The rate of spread of \( A_1 \) when introduced at a low frequency into an \( A_2A_2 \) population is governed by equations similar to (4.29) and (4.30), except that \( r_{22} \) is substituted for \( r_{11} \), \( p_1 \) for \( p_2 \), and the values of the \( k_{12}^*(x) \) are those for mating \( A_1A \) males to females from a stable \( A_2A_2 \) population.

A point of interest which emerges from this analysis is that, for the case of random mating with respect to age and no demographic differences between males and females (equation (4.30)), both endpoints of gene frequency are unstable if there is heterozygote superiority in the intrinsic rate of increase. In other words, each allele tends to increase in frequency when rare, so that there is a protected polymorphism in the sense of Prout (1968), when \( r_{12} > r_{11} \), \( r_{22} \). If the reverse inequality holds, both alleles tend to be eliminated when rare. This corresponds to the results of the analysis for weak selection, and to the standard results for discrete generations if the genotypic fitnesses \( W_{ij} \) are substituted for the \( r_{ij} \). It suggests that, with strong selection, there may be convergence to the equilibrium given by equations (3.38) and (3.39) when there is heterozygote advantage in the \( r_{ij} \). If \( r_{11} < r_{12} < r_{22} \), the endpoint \( p_1 = 1 \) is unstable, and \( p_2 = 1 \) is stable, which suggests that \( A_2 \) spreads towards fixation. The extent to which these conjectures are true is discussed in section 4.4.

A rare, recessive allele

If either \( r_{11} \equiv r_{12} \) or \( r_{22} \equiv r_{12} \), the above approach breaks down when \( A_2 \) or \( A_1 \), respectively, is rare, since the initial rate of spread of the recessive allele is of the same order as the square of its frequency. A
different method of approximation must therefore be used when a rare allele is recessive in its effect on the intrinsic rate of increase. We examine here equation (4.5) for a two-allele system, with $A_2$ introduced at low frequency into an $A_1A_1$ population and with $r_{12} = r_{11}$. Using the result that $\Delta p_2$ is $O(p_2^2)$, we can employ the same type of approach as for equations (4.2), (4.3) and (4.4) to obtain

$$B(t) = \sum_x B(t-x)\{[1-2p_2(t_0)]k_{11}(x) + 2p_2(t_0)k_{12}(x)\} + O(p_2^2)$$

(4.31)

where $t_0$ is a fixed time not too far from $t$. (Note that this does not assume anything about the strength of selection.) Since $r_{12} = r_{11}$, we thus have

$$B(t-x) \sim B(t)e^{-r_{11}x} + O(p_2^2)$$

Substituting this into equation (4.5) for allele $A_2$, we obtain

$$p_2(t) \sim \sum_x e^{-r_{11}x}p_2(t-x)[p_1(t-x)k_{12}(x) + p_2(t-x)k_{22}(x)]$$

$$+ O(p_2^3)$$

(4.32)

Taking first differences on both sides, we have

$$\Delta p_2 \sim \sum_x e^{-r_{11}x}\Delta p_2(t-x)k_{12}(x) + O(p_2^3)$$

so that $\Delta p_2$ approaches a constant value asymptotically, to order $p_2^3$ (cf. Appendix 2). The second-order and higher-order differences in $p_2$ can therefore be neglected, so that equations (4.32) can be expanded on similar lines to equations (4.6) and (4.7), to yield the final result

$$\Delta p_2 \sim \frac{p_2^2[w_{22}(r_{11}) - 1]}{T_{12}(r_{11})} + O(p_2^3)$$

(4.33)

If selection is weak, we can use the argument that led to equation (4.11); neglecting terms $O(e^2)$, we obtain

$$\Delta p_2 \sim p_2^2(r_{22} - r_{11})$$

(4.34)

If selection is strong, however, equations (4.33) and (4.34) yield different values for the change in gene frequency. Comparisons with the values of $\Delta p_2$ obtained by exact calculations of population trajectories show that equation (4.33) is accurate with strong selection, provided that $p_2$ is small ($<0.05$ or so), whereas equation (4.34) breaks down with strong selection, although $A_2$ can spread only when $r_{22} > r_{11}$. We therefore have the
somewhat surprising result that when selection is strong, the rate of
spread of a rare recessive gene cannot be calculated purely from its effect
on the intrinsic rate of increase, although the direction is determined by
this quantity. This contrasts with the results for the case of a gene with a
significant effect on the intrinsic rate of increase of the heterozygote.

4.3.2 Gene frequencies near 0 or 1 with density dependence

The results of the previous two sections can easily be extended to
the density-dependent case. The introduction of a rare allele will, as
before, perturb the demographic parameters of the population by an
amount of the order of the gene frequency (or square of the gene
frequency in the case of a recessive allele). If the density-regulating
mechanism is such that a genetically homogeneous population tends to a
stationary size, as is assumed throughout our treatment of density-
dependence, the analogues of equation (4.27) for this density-dependent
case with a rare, non-recessive gene are

\[ B(t - x) \sim B(t) + O(p_2) \]  \hspace{1cm} (4.35a)

\[ N(t) \sim N_{11} + O(p_2) \]  \hspace{1cm} (4.35b)

Substituting from these equations into the gene frequency equations
(3.14), we find that the asymptotic logarithmic rate of increase in the
frequency of \( A_2 \) is, to order \( p_2 \), given by the real root of the equation

\[ \sum_x e^{-\alpha x} k_{12}(x, N_{11}) = 1 \]  \hspace{1cm} (4.36)

It is easy to see that for this root to be positive we must have \( N_{12} > N_{11} \) in
the case when the \( w_{ij}(N) \) as defined in equation (3.40) are strictly
decreasing functions of \( N \). Hence, a rare gene spreads if the carrying-
capacity of the heterozygote exceeds that of the initial population.
Applying this to the other endpoint gives the result that there is a
protected polymorphism if \( N_{12} > N_{11}, N_{22} \). This is similar to the results
derived in section 4.2.2 for the case of weak selection, but is obviously
more restricted, since nothing has been proved about convergence to the
equilibrium.

A similar type of argument can be applied to the problem of the
introduction of a rare recessive gene. The analogue of equation (4.33) is

\[ \Delta p_2 \sim \frac{p_2^2[w_{22}(N_{11}) - 1]}{T_{12}(N_{11})} + O(p_2^3) \]  \hspace{1cm} (4.37)

A recessive gene, therefore, spreads if and only if the carrying-capacity of
the homozygote exceeds that of the initial population, in the case when
the \( w_{ij}(N) \) are decreasing functions of \( N \).

4.3.3 Gene frequencies near 0 or 1 with temporally varying environments
The results derived up to now have assumed that the demo-
graphic parameters of each genotype are independent of time, or that
they depend on time solely via changes in population density. There has
been much interest in the population genetics of selection in temporally
varying environments (e.g. Haldane and Jayakar, 1963; Gillespie, 1973a,
b, 1991; Karlin and Liberman, 1974), so that it is natural to enquire how
far the results can be extended to age-structured populations. The discus-
sion here is confined to the question of the conditions for maintenance of
a protected polymorphism in a temporally varying environment. The
nature of the method of analysis means, as will be seen, that the
population is density independent. As previously, the gene-frequency
equations (3.14) are used.

A rare, non-recessive allele
For \( A_2 \) introduced into an \( A_1A_1 \) population at a low frequency,
we have

\[
B(t)p_2(t) = \sum_x B(t - x)p_2(t - x)k_{12}(x, t) + O(p_2^2) \tag{4.38a}
\]

\[
B(t) = \sum_x B(t - x)k_{11}(x, t) + O(p_2) \tag{4.38b}
\]

The process represented by these equations can be expressed in terms of
the equivalent Leslie matrix formulations, using the matrices for \( A_1A_1 \)
and \( A_1A_2 \), which we can write as \( L_{11}(t) \) and \( L_{12}(t) \) for time \( t \) (cf.
equation (1.6)). Since we have assumed density independence we can use
weak ergodicity (section 1.4.1) to deduce that for sufficiently large \( t \)

\[
B(t)p_2(t) = \lambda_{12}(t - 1)B(t - 1)p_2(t - 1) + O(p_2^2) \tag{4.39a}
\]

\[
B(t) = \lambda_{11}(t - 1)B(t - 1) + O(p_2) \tag{4.39b}
\]

where \( \lambda_{11}(t - 1) \) is the geometric growth rate of a homozygous \( A_1A_1 \)
population from time \( t - 1 \) to \( t \), and \( \lambda_{12}(t - 1) \) is the rate for a population
with the demographic parameters of \( A_1A_2 \).

Combining these two equations, we obtain the analogue of equation
(4.30)

\[
\Delta \ln p_2(t) = [r_{12}(t) - r_{11}(t)] + O(p_2) \tag{4.40}
\]
where \( r_{ij}(t) = \ln \lambda_{ij}(t) \), and measures the increase in \( \ln B \) between \( t \) and \( t + 1 \), for a population with the demographic parameters of \( A_iA_j \).

From the weak ergodicity theorem, the \( r_{ij}(t) \) and \( \lambda_{ij}(t) \) are independent of the initial conditions, for sufficiently large \( t \), and are determined only by the environmental process which generates the time dependence of the demographic parameters. The difficulty with this result is that it is not generally possible to obtain an analytic expression for the values of the \( r_{ij} \) and \( \lambda_{ij} \) as functions of time. The following useful conclusions can, however, be drawn from equation (4.40).

In the first place, equation (4.40) tells us that the mean change in gene frequency between generations \( t \) and \( t + n \) for large \( t \) is given by

\[
\frac{1}{n} [\ln p_2(t + n) - \ln p_2(t)] = \frac{1}{n} \left( \sum_{j=0}^{n-1} [r_{12}(t + j) - r_{11}(t + j)] \right) + O(p_2)
\]  

(4.41)

so that, to order \( p_2^2 \), there will be an average increase in the frequency of \( A_2 \) if an \( A_1A_2 \) population has a higher arithmetic mean value of \( r_{ij}(t) \) than an \( A_1A_1 \) one. This is analogous to the result of Haldane and Jayakar (1963) for the corresponding discrete-generation case, with the \( r_{ij}(t) \) in the age-structured case playing the same role as the logarithms of the genotypic fitnesses with discrete generations. It implies that, in a deterministically varying environment, a rare gene will increase in frequency asymptotically if a population with the demographic parameters of the heterozygote has a higher mean rate of increase than the population homozygous for the original allele.

A similar conclusion can be drawn for the case of a stochastically varying environment by using the theorem of Cohen (1976, 1977a, b) that the probability distribution of age-structure, and hence rate of increase of a population at a given time \( t \), converges to a stationary distribution when exposed to Markovian temporal variation in its demographic parameters (cf. section 1.4.1). This implies that the \( r_{ij}(t) \) in equation (4.41) can be regarded as sampled from such a stationary distribution, provided that the assumptions of the theorem apply to the matrices \( L_{11}(t) \) and \( L_{12}(t) \). This enables us to apply the arguments used for the discrete-generation case by Gillespie (1973a) and Karlin and Liberman (1974), as noted by Charlesworth (1980, p. 175) and Tuljapurkar (1982). We can conclude that there is a probability close to one that \( p_a \) is bounded away from zero (i.e. the endpoint \( p_a = 1 \) is stochastically unstable) if the expectation of \( r_{12}(t) - r_{11}(t) \), over the stationary distribution generated by the environmental process, is positive.
These results can obviously be extended to the other endpoint \( p_1 = 0 \), provided that \( A_1 \) is not recessive to \( A_2 \); a protected polymorphism is maintained, therefore, if the arithmetic mean of the \( r_{ij}(t) \) for a population with the demographic characteristics of the heterozygote exceeds the means for the homozygotes.

In the notation of section 1.4.1, the value of \( \bar{r} \) for genotype \( ij \) is thus the most appropriate fitness measure in a temporally varying environment. If a homozygote for one allele has a higher \( \bar{r} \) value that the homozygote for the other allele, and the value for the heterozygote lies in-between, these results suggest that the first allele will eventually be fixed. Numerical studies are in accord with this conjecture (Orzack, 1985). Formulae for the genotypic \( \bar{r} \) values can be obtained for the special cases considered in section 1.4.1 by applying equations (1.75) and (1.78) to individual genotypes. The variances and covariances among demographic parameters induced by environmental fluctuations, as well as autocorrelations across time-intervals, can have important effects on the conditions for maintenance of genetic variability (Gillespie, 1973a, 1991; Orzack, 1985, 1993; Orzack and Tuljapurkar, 1989; Tuljapurkar, 1982, 1989, 1990a). Templeton and Levin (1979) discuss the analysis of selection with fluctuating environments for the special case of plants with seed dormancy.

A rare, recessive allele

When one allele, say \( A_2 \), is recessive to the other, the above approach breaks down for the same reasons as in the time-independent case. In the discrete-generation case, Haldane and Jayakar (1963) showed that a protected polymorphism can be maintained at a locus where one allele is completely recessive to the other, as a result of temporal fluctuations in fitnesses. This requires the arithmetic mean of the fitness of the homozygote for the recessive allele \( (A_2) \) to exceed that of \( A_1A_1 \) and \( A_1A_2 \), whereas its geometric mean must be less than the geometric mean for \( A_1A_1 \) and \( A_1A_2 \). Unfortunately, it does not seem possible to provide a general analysis similar to this for the age-structured case. The following special case proves, however, that the Haldane–Jayakar mechanism can work in an age-structured context.

Consider the situation in which there are environmental states which alternate at intervals which are very long compared with the life-span of an individual. For most of the time between the environmental changes, a population in such a situation can be described accurately by the gene frequency equations with the demographic parameters which are appropriate for the given environmental state in which it finds itself, since the
transitional states created by a change in environment occupy a negligible part of the population’s history. Hence, for a population in environmental state $i$, the rate of gene frequency change for $A_2$ when rare is approximated accurately for most of the time by equation (4.33) with the demographic parameters for environment $i$

$$\Delta p_2 \sim \frac{p_2^2[w_{22}(r_{11}^{(i)}) - 1]}{T_{12}^{(i)}(r_{11}^{(i)})} + O(p_2^3)$$

(4.42)

where $r_{11}^{(i)}$ is the intrinsic rate of increase of an $A_1A_1$ population maintained in environment $i$ indefinitely, etc. By the same type of argument as employed by Haldane and Jayakar (1963), $p_2$ tends to increase in frequency in the long term if the arithmetic mean of $w_{22}(r_{11}^{(i)}) - 1$ is positive. The same type of argument concerning the effects of the long environmental period can be applied to the other endpoint; combined with the results of section 4.3.1, we find that $A_1$ increases when rare if the arithmetic mean of $r_{12}^{(i)} - r_{22}^{(i)}$ is positive. The existence of a protected polymorphism requires both these conditions to be satisfied simultaneously.

If selection is weak, the condition for the increase of $A_2$ reduces to requiring the arithmetic mean of $r_{12}^{(i)} - r_{22}^{(i)}$ to be negative (cf. equation (4.34)). This is obviously impossible if the condition for the increase of $A_1$ is satisfied. We can conclude that a protected polymorphism cannot exist when selection is weak. (This is also true for the discrete-generation case, since the arithmetic and geometric means of the fitnesses approach each other for weak selection.) When selection is strong, however, it is possible for the condition for the spread of $A_2$ to hold when the mean of $r_{12}^{(i)} - r_{22}^{(i)}$ is positive, so that a protected polymorphism can exist under suitable forms of environmental fluctuation. An extreme example of how this is possible is as follows. Suppose there are just two environment states, in one of which $A_2A_2$ is near-lethal, so that $r_{22}$ for this state is very large and negative. Provided that $r_{22}$ for the other environment is bounded, the mean of $r_{12}^{(i)} - r_{22}^{(i)}$ will be positive if $A_1A_1$ and $A_1A_2$ have reasonable survival rates in both environments. If, however, $A_2A_2$ has a sufficiently high survival in the second environment compared with $A_1A_1$ and $A_1A_2$, the criterion for the increase of $A_2$ when rare can be satisfied, despite the near-lethality of $A_2A_2$ in the first environment.

Computer calculations of population trajectories have shown that this
approximate method predicts the conditions for maintenance of polymorphism when the environmental period is of the order of several generations, and that there is no more difficulty in generating numerical examples which satisfy the conditions than in the discrete generation case.

The conditions for the spread of a rare recessive gene, and hence for protected polymorphism, are less clear for the case when the environmental period is of the order of or smaller than the life-span, or when there is stochastic variation in the environment. In such cases, the response of the population to changing demographic parameters means that the rare gene can never settle down to a constant rate of increase in frequency. All that can be done with present theory is to calculate the asymptotic growth rates, the \( r_{ij}(t) \), for specific numerical examples.

### 4.3.4 Gene frequencies near 0 or 1 with spatially varying environments

The case of a spatially varying environment can be treated in a similar way. Two extreme models of spatial variation have been treated in the literature (Dempster, 1955). Wallace's (1968) terms 'hard' and 'soft' selection were applied to these models in the context of a discrete-generation population by Christiansen (1975). Each model assumes that the zygotes produced by a random-mating population encounter a variety of spatial patches ( niches), such that the probability that a zygote encounters a patch of type \( m \) in which its reproductive function is \( k_{ij}(m, x) \) is \( c_m \). In both cases, the individual spends the selectively relevant part of its life-history in the patch in which it is born; in the simplest situation, mating is assumed to be random between patches, although restricted migration between patches can be incorporated (Christiansen, 1975).

Under hard selection, there is no density-dependent population regulation, at least at the level of the individual patch. Under these circumstances, it is easily seen that selection acts similarly to the case of a homogeneous environment (cf. Dempster, 1955), replacing the reproductive function of the \( ij \)th genotype by its mean over all patches:

\[
\sum_m c_m k_{ij}(m, x)
\]

If the intrinsic rate of increase for the population as a whole is \( \hat{r} \), the equilibrium fitness measure equivalent to that of equation (3.31) is

\[
w_{ij} = \sum_x e^{-x} \sum_m c_m k_{ij}(m, x)
\]

The dynamics of selection can be investigated by the methods employed for the case of a homogeneous environment, using the mean reproductive function described above. In particular, genetic variation is maintained at
a single locus with two alleles if the heterozygote has the highest intrinsic rate associated with the mean reproductive function (cf. section 4.3.1).

With soft selection, there is strict density-regulation within each patch, so that the $m$th patch makes a fixed contribution $c_m$ to the pool of new zygotes at each time-interval. The single-locus system with two alleles can be analysed by the following method, which is a straightforward extension of the discrete-generation treatment introduced by Levene (1953). Let $A_2$ be a new allele invading an $A_1A_1$ population. Under soft selection, the proportion of $A_2$ alleles at time $t$ among the zygotes contributed from patches of type $m$ is approximated by

$$\sum_x B(t-x)p_2(t-x)k_{12}(m, x)/\sum_x B(t-x)k_{11}(m, x)$$

The net frequency of $A_2$, $p_2(t)$, is the sum of this over all patches, weighting each term by $c_m$. If the overall population is growing at rate $r_{11}$, given by the intrinsic rate associated with the mean reproductive function of $A_1A_1$, we get the recurrence relation

$$p_2(t) = \sum_x e^{-r_{11}x} p_2(t-x) \sum_m c_m k_{12}(m, x) / \left[ \sum_x e^{-r_{11}x} k_{11}(m, x) \right]$$

This leads to the initial increase criterion analogous to Levene's:

$$\sum_x e^{-r_{11}x} \sum_m c_m k_{12}(m, x) / \left[ \sum_x e^{-r_{11}x} k_{11}(m, x) \right] > 1$$

A symmetrical expression can be written for the case of an $A_1$ allele invading an $A_2A_2$ population:

$$\sum_x e^{-r_{22}x} \sum_m c_m k_{12}(m, x) / \left[ \sum_x e^{-r_{22}x} k_{22}(m, x) \right] > 1$$

Genetic variation will be maintained if conditions (4.44b) and (4.44c) are both satisfied. As in the discrete-generation case, the conditions for maintenance of variation for the case of soft selection differ significantly from those with hard selection. Since the denominator of equations (4.44b) and (4.44c) must be less than one for those patches in which the prevailing homozygote is at a disadvantage, it is not necessary for there to be heterozygote advantage with respect to the mean reproductive function for variation to be maintained (cf. Levene, 1953; Dempster, 1955).
4.3.5 The probability of survival for a non-recessive mutant gene

The above results are based on the assumption that the spread of a rare gene in an infinite population can be satisfactorily described from a purely deterministic viewpoint. In actual fact, a rare gene is represented by so few copies in the initial generations after its occurrence by mutation that even in an infinite population it is subject to stochastic fluctuations and a significant probability of loss. This process was first studied by Fisher (1922, 1930) and Haldane (1927b) for the case of a discrete-generation population. In this section, we shall study the problem of the probability of survival of a mutant gene in an infinite population with time-independent demographic parameters. The account is based on the treatment of Charlesworth and Williamson (1975). A somewhat more general treatment is given by Pollak (1976). The results will be used in Chapter 5 (p. 193).

We again use the model expressed in equation (4.5), which assumes random mating with respect to genotype and age, equality of age-specific survivorships of males and females of the same genotype, and proportionality of male and female age-specific fecundities. If \( A_2 \) is introduced by mutation into an \( A_1A_1 \) population, it will initially be overwhelmingly represented in \( A_1A_2 \) heterozygotes. Furthermore, if the initial \( A_1A_1 \) population is in stable age-distribution, or is held at a stationary equilibrium size by density-dependent factors, we saw on p. 151 that \( m_{12}(x) = m_{12}(x) \) plus terms of order \( p_2 \). It is therefore reasonable to assume that, neglecting terms \( O(p_2) \), the distributions of numbers of offspring for \( A_1A_2 \) males and females of a given age are identical, provided that they are completely characterised by their mean (e.g. when they are Poisson). This assumption is made from now on, and enables us to disregard the sex of individuals. A final assumption is that the primary sex-ratio is equal to one-half. This implies that the mean number of \( A_1A_2 \) offspring, of either sex, of an \( A_1A_2 \) heterozygote is equal to the mean number of its offspring that are the same sex as itself, which enables us to simplify the notation.

A new mutant gene \( A_2 \) will first manifest itself as an \( A_1A_2 \) zygote. The problem therefore reduces to that of finding the probability of survival of a line of \( A_1A_2 \) individuals descended from a single \( A_1A_2 \) zygote. Let this probability be \( U \). We can arbitrarily classify the set of progeny produced by a given zygote during the whole of its lifetime as belonging to the same 'generation', although many of them will in fact have been produced in different time-intervals. Standard branching-process theory (Feller, 1968, Chapter 12) can thus be applied to successive 'generations' descended from the initial zygote carrying \( A_2 \). This yields the probability of loss of
$A_2$ as the smallest value of $s$, in the interval $0 \leq s < 1$, which satisfies the equation

$$h_{12}(s) = s$$  \hspace{1cm} (4.45)

where $h_{12}(s)$ is the probability generating function (p.g.f.) of the distribution of the number of $A_1A_2$ offspring produced by an $A_1A_2$ individual over the whole of its life-span (cf. Appendix 1). It follows from standard theory that the mutant gene has a non-zero probability of survival if and only if the mean of this offspring distribution exceeds one, i.e. the net reproduction rate, $R_{12} = \sum k_{12}(x)$, exceeds one. This is also the condition for $r_{12} > 0$.

The functional form of $h_{12}(s)$ can be obtained as follows. Let the p.g.f. for the number of $A_1A_2$ offspring produced by an $A_1A_2$ parent aged $x$ be $h_{12}(x,s)$. Assuming that, for a parent who has survived to a given age, the offspring distributions for different earlier ages are independent, the p.g.f. for the number of $A_1A_2$ offspring produced between ages $b$ and $x$ by an individual now aged $x (x \geq b)$ is

$$H_{12}(x,s) = \prod_{y=b}^{x} h_{12}(y,s) \hspace{1cm} (4.46)$$

The probability that an $A_1A_2$ individual survives to age $x$ and dies between $x$ and $x + 1$ is $l_{12}(x) - l_{12}(x + 1)$. By partitioning the life-history of an individual into mutually exclusive events corresponding to these probabilities for all different $x$ values, it is easy to see that

$$h_{12}(s) = 1 - l_{12}(b) \sum_{x=b}^{d} [l_{12}(x) - l_{12}(x + 1)] H_{12}(x,s) \hspace{1cm} (4.47)$$

In the important case when the offspring distribution for each age-class is Poisson, so that $h_{12}(x,s) = \exp\{(s - 1)m_{12}(x)\}$, the survival probability $U$, from equation (4.45), satisfies the equation

$$U + \sum_{x} [l_{12}(x) - l_{12}(x + 1)] \exp[-\tilde{M}_{12}(x)U] = l_{12}(b) \hspace{1cm} (4.48)$$

where $\tilde{M}_{12}(x) = \sum_{y=b}^{x} m_{12}(y)$.

As shown by Charlesworth and Williamson (1975), the survival probability given by equation (4.48) is always less than that for a discrete generation model with a Poisson offspring distribution and the same mean. These authors give an example for a human population where the reduction is as high as 40% of the discrete generation value.
For $U$ close to zero and $R_{12}$ close to unity, equation (4.45) can be approximated in the standard way (e.g. Ewens, 1969, p. 80) to give

$$U \equiv \frac{2(R_{12} - 1)}{V_{12}}$$

(4.49)

where $V_{12}$ is the variance of the lifetime offspring distribution. In the case of a Poisson distribution for each age-class,

$$V_{12} \equiv \sum [t_{12}(x) - t_{12}(x + 1)]M_{12}^2(x)$$

when $R_{12} \equiv 1$.

Equation (4.49) may be thought of as giving a useful approximation to the survival probability of a mutant gene with a small effect on demographic parameters in the heterozygous state, when introduced into a stationary or near-stationary population. Using the approximation for the intrinsic rate of increase (equation (1.46)), equation (4.49) can be rewritten as

$$U \equiv \frac{2r_{12}T_{12}}{V_{12}}$$

(4.50)

where $T_{12} = \sum xk_{12}(x)$. It will be seen from these equations that small changes in the demographic parameters which affect $R_{12}$ and $r_{12}$ alone have much bigger effects on $U$ than changes which affect $V_{12}$ or $T_{12}$, provided that $U$ is small. This implies that the probability of survival of a mutant gene in a near-stationary population is largely controlled by its effect on $R_{12}$ or, equivalently, $r_{12}$.

### 4.3.6 Local stability of a polymorphic equilibrium

The purpose of this section is to extend the method of local stability analysis to the polymorphic equilibrium which exists in the two-allele case when there is heterozygote superiority or inferiority in the intrinsic rates of increase, for the time-independent case (Chapter 3, pp. 122–3). We saw on p. 141 of this chapter that heterozygote superiority in the $r_{ij}$ corresponds to global stability of the equilibrium in the case of weak selection, and heterozygote inferiority to instability. We show here that similar results hold for local stability with arbitrary selection intensities.

**Derivation of a stability criterion**

The equilibrium gene frequencies $\hat{p}_1$ and $\hat{p}_2 = 1 - \hat{p}_1$ are given by equation (3.38), and the equilibrium population growth-rate by equation (3.39). We can linearise equations (3.14), for the two-allele case, by neglecting terms in $(p_1 - \hat{p}_1)^2$. This will give an adequate approximation
Local stability analyses

for gene frequencies close to their equilibrium values. Writing \( B_1(t) = B(t)[p_1(t) - \bar{p}_1] \), we obtain

\[
B(t)\hat{p}_1 + B_1(t) = g_1(t) + \hat{p}_1 \sum_{x=1}^{t} B(t-x)k_1(x) + \sum_{x=1}^{t} B_1(t-x)K_1(x)
\]

\[ (4.51a) \]

\[
B(t)\hat{p}_2 - B_1(t) = g_2(t) + \hat{p}_2 \sum_{x=1}^{t} B(t-x)k_2(x) - \sum_{x=1}^{t} B_1(t-x)K_2(x)
\]

\[ (4.51b) \]

where

\[
k_1(x) = \hat{p}_1 k_{11}(x) + \hat{p}_2 k_{12}(x)
\]

\[
K_1(x) = 2\hat{p}_1 k_{11}(x) + (1 - 2\hat{p}_1)k_{12}(x)
\]

\[
k_2(x) = \hat{p}_1 k_{12}(x) + \hat{p}_2 k_{22}(x)
\]

\[
K_2(x) = 2\hat{p}_2 k_{22}(x) - (1 - 2\hat{p}_1)k_{12}(x)
\]

and where \( g_1(t) \) and \( g_2(t) \) are zero for \( t \geq d \).

Taking generating functions on both sides of these equations (see Appendix 1 for the properties of generating functions which are used here), we obtain

\[
\hat{p}_1 \bar{B}(1 - \bar{k}_1) + \bar{B}_1(1 - \bar{K}_1) = \bar{g}_1
\]

\[ (4.52a) \]

\[
\hat{p}_2 \bar{B}(1 - \bar{k}_2) - \bar{B}_1(1 - \bar{K}_2) = \bar{g}_2
\]

\[ (4.52b) \]

These equations are easily solved for \( \bar{B} \) and \( \bar{B}_1 \):

\[
\bar{B} = [\bar{g}_1(1 - \bar{K}_2) + \bar{g}_2(1 - \bar{K}_1)]/\Delta
\]

\[ (4.53a) \]

\[
\bar{B}_1 = [\hat{p}_2 \bar{g}_1(1 - \bar{k}_2) - \hat{p}_1 \bar{g}_2(1 - \bar{k}_1)]/\Delta
\]

\[ (4.53b) \]

where

\[
\Delta = \hat{p}_1(1 - \bar{k}_1)(1 - \bar{K}_2) + \hat{p}_2(1 - \bar{k}_2)(1 - \bar{K}_1)
\]

\[ (4.54) \]

Using the results described in Appendix 1, it follows that for sufficiently large \( t \) we have

\[
B(t) = C_1 \delta_1^{-(t+1)}
\]

\[ (4.55a) \]
where \( s_1 \) is the zero of \( \Delta \) in equation (4.54) with smallest modulus, excluding any common zeros of \( 1 - \bar{K}_1 \) and \( 1 - \bar{K}_2 \).

Similarly, for large \( t \) we have

\[
B_1(t) = C_2 s_2^{-(t+1)}
\]  

(4.55b)

where \( s_2 \) is the zero of \( \Delta \) with smallest modulus, excluding any common zeros of \( 1 - \bar{K}_1 \) and \( 1 - \bar{K}_2 \). Note that it is possible for \( s_1 \) and \( s_2 \) to be equal. Equations (4.55) assume that \( s_1 \) and \( s_2 \) are simple zeros. The modifications to these equations when \( s_1 \) or \( s_2 \) is a repeated zero of \( \Delta \) are relatively straightforward (see Appendix 1), and do not materially affect the results described below.

As with the renewal treatment of the case of a genetically homogeneous population (equation (1.34)), it is more convenient in practice to work with the natural logarithms of \( s_1^{-1} \) and \( s_2^{-1} \), which we denote by \( z_1 \) and \( z_2 \). Equations (4.55) thus become

\[
B(t) = C_1 e^{z_1(t+1)}
\]

(4.56a)

\[
B_1(t) = C_2 e^{z_2(t+1)}
\]

(4.56b)

If \( z_1 \) is a complex number, then equation (4.56a) must be replaced by

\[
B(t) = C_1 e^{z_1(t+1)} + C_1^* e^{z_1^*(t+1)}
\]

(4.57)

where \( z_1^* \) and \( C_1^* \) are complex conjugates of \( z_1 \) and \( C_1 \), respectively. A corresponding equation replaces (4.56b) if \( z_2 \) is complex.

If we can show that the real part of \( z_1 \) is equal to \( \hat{p} \), and that the real part of \( z_2 \) is less than \( \hat{p} \), the equilibrium can be said to be locally stable, since \( B(t) \) must return to its equilibrium rate of growth, and \( p_1 \) to \( \hat{p}_1 \). Conversely, if the real parts of \( z_1 \) and/or \( z_2 \) exceed \( \hat{p} \), the equilibrium must be locally unstable.

Unfortunately, it is difficult to give an analysis of the zeros of \( \Delta \) analogous to that of equation (1.32), owing to the large number of terms in equation (4.54). An indirect approach can, however, be used to establish that, when there is heterozygote superiority in the intrinsic rates of increase, the equilibrium is locally stable.

### Stability when \( r_{12} > r_{11}, r_{22} \)

We first note that \( \hat{p} \) is, from equations (3.37), the root of both \( 1 - \bar{K}_1 = 0 \) and \( 1 - \bar{K}_2 = 0 \) with largest real part (from now on, the generating functions are treated as functions of \( z = \ln(s^{-1}) \) rather than \( s \)). Assume arbitrarily that \( \hat{p}_2 > \hat{p}_1 \). Then \( K_1(x) \) is non-negative, from equations (4.51), and the zero of \( 1 - \bar{K}_1 \) with largest real part, denoted...
Local stability analyses

by $z'$, is real by the argument of pp. 23–4. The formula for $K_1(x)$ also implies that $r_{11} < \frac{z'}{r_{12}}$. Furthermore, we can write

$$K_1(x) = k_1(x) + \hat{p}_1[k_{11}(x) - k_{12}(x)]$$

This implies $z' < \hat{p}$ since it is easily seen from this formula that

$$\sum e^{-\gamma x}K_1(x) < 1.$$

If the real parts of $z_1$ and $z_2$ are greater than $\hat{p}$, we must have $z_1 = z_2$, since only the common zeros of $1 - \bar{K}_1$ and $1 - \bar{K}_2$ or $1 - \bar{k}_1$ and $1 - \bar{k}_2$ are not simultaneous roots for $\bar{B}$ and $\bar{B}_1$, and we have just established that only $1 - \bar{K}_2$ can have a zero with real part greater than $\hat{p}$. Hence, either the real parts of $z_1$ and $z_2$ are smaller than or equal to $\hat{p}$, or $z_1 = z_2$.

It can be established as follows that there can be no real zero of $\Delta$ which is greater than $\hat{p}$. Rearranging equation (4.54) gives

$$\Delta = \bar{k}_1(1 - \bar{k}_1)(1 - \bar{k}_2) + \bar{k}_2(1 - \bar{k}_2)(1 - \bar{k}_1)$$

(4.58)

It is easy to see that, for all real $\frac{z}{\bar{p}}$, $\Delta > 0$, so that $\Delta$ cannot have a real zero greater than $\hat{p}$. The only way in which the equilibrium can be unstable in this case is if there is a complex zero of $\Delta$, $z_1 = z_2$, with a real part greater than $\hat{p}$. The following argument shows that this situation cannot occur.

Substituting from equation (4.57) into equation (4.51a), we obtain

$$\bar{k}_1 e^{z(t+1)}(1 - \bar{k}_1) + \bar{k}_2 e^{z(t+1)}(1 - \bar{k}_2) + C_1 e^{z(t+1)}(1 - \bar{k}_1) + C_2 e^{z(t+1)}(1 - \bar{k}_2) = 0$$

where

$$\bar{k}_1 = \sum e^{-\gamma x}k_1(x), \quad \bar{k}_2 = \sum e^{-\gamma x}K_1(x)$$

If we write $z_1 = \alpha + i\beta$, this gives

$$e^{z(t+1)}[\bar{k}_1 C_1(1 - \bar{k}_1) + C_2(1 - \bar{k}_1)] + \bar{k}_1 C_1(1 - \bar{k}_1) + C_2(1 - \bar{k}_2) = 0$$

(4.59)

This equation must be satisfied for arbitrary $t$, which is impossible unless $z_1$ is a zero of both $1 - \bar{k}_1$ and $1 - \bar{k}_1$; this possibility has been excluded already. Hence, there cannot be a complex zero of equation (4.54) with its real part greater than $\hat{p}$.

This completes the results which are necessary to establish that the polymorphic equilibrium is locally stable if $r_{12} > r_{11}, r_{22}$. Since there is no complex zero of $\Delta$ with real part greater than $\hat{p}$, it follows that $z_1$, the largest root for $B(t)$, must equal $\hat{p}$ (see equation (4.54), and that $z_2$, the
largest root for $B_1(t)$, must have its real part less than $\hat{r}$. From the
definition of $B_1(t)$, the deviation of the gene frequency from its equilib-
rium value, $p_1 - \hat{p}_1$, must asymptotically be proportional to $e^{(z_2 - \hat{r})t}$, and
hence tend to zero. Convergence of the gene frequency to equilibrium
takes place via a series of damped oscillations. The population growth-
rate similarly converges to its equilibrium value $\hat{r}$. Computer calculations
of population trajectories show this pattern clearly (see, for example,
Figure 2 of Anderson and King, 1970). The rate of convergence is
determined by the difference between the real part of $z_2$ and $\hat{r}$. In
general, $z_2$ can be determined only by numerical evaluation of the zeros
of $\Delta$. The convergence to equilibrium via damped oscillations in this case
contrasts with the result for discrete generations with heterozygote super-
iority in fitness, where convergence to the equilibrium is monotonic (cf.
Nagylaki, 1992, Chapter 4).

**Instability when $r_{12} < r_{11}, r_{22}$**

In this case, it is easy to see from equation (4.58) that there is a
real zero of $\Delta$ which exceeds $\hat{r}$. An argument similar to the one used
above shows that there cannot be a complex zero with real part greater
than this. The population in the neighbourhood of the equilibrium thus
asymptotically moves away from the equilibrium point at a constant rate.

**4.4 The asymptotic results of selection**

Some theoretical results concerning the ultimate states reached
by populations under arbitrarily strong selection intensities have been
obtained by Norton (1928), Pollak and Kempthorne (1971) and Charles-
worth (1974a). These results all pertain to the two-allele versions of
equation (4.5), with time-independent and density-independent demo-
graphic parameters, or to the corresponding continuous-time equations.
The most complete analysis was given by Norton for the continuous-time
model. His analytical methods are too complex to describe here, but his
conclusions are summarised in Table 4.1, along with the corresponding
discrete-generation results. They agree with the results derived in section
4.2.1 for the case of weak selection, and complement the results of the
local stability analyses carried out in sections 4.3.1 and 4.3.6. The
genotypic intrinsic rates of increase, $r_{ij}$ are predictors of the ultimate
outcome of selection in the same way as the fitnesses $W_{ij}$ in the discrete-
generation case. Except when there is heterozygote superiority or in-
feriority in the $r_{ij}$, natural selection tends to fix whichever allele is
associated with the highest intrinsic rate of increase when homozygous.
### Table 4.1. The outcomes of various types of selective regimes with a diallelic locus

<table>
<thead>
<tr>
<th>Selection regime</th>
<th>Outcome of selection</th>
<th>Selection regime</th>
<th>Outcome of selection</th>
</tr>
</thead>
<tbody>
<tr>
<td>( w_{11} = w_{12} = w_{22} )</td>
<td>( p_1 ) constant (neutral equilibrium)</td>
<td>( r_{11} = r_{12} = r_{22} )</td>
<td>( p_1 ) constant, or neutral oscillations</td>
</tr>
<tr>
<td>( w_{11} &lt; w_{12} \leq w_{22} )</td>
<td>( p_1 \to 0 )</td>
<td>( r_{11} &lt; r_{12} \leq r_{22} )</td>
<td>( p_1 \to 0 )</td>
</tr>
<tr>
<td>( w_{11} \leq w_{12} &lt; w_{22} )</td>
<td>( p_1 \to 0 )</td>
<td>( r_{11} \leq r_{12} &lt; r_{22} )</td>
<td>( p_1 \to 0 )</td>
</tr>
<tr>
<td>( w_{12} &gt; w_{11}, w_{22} )</td>
<td>( p_1 \to \hat{p}_1 ) ( (0 &lt; \hat{p}_1 &lt; 1) )</td>
<td>( r_{12} &gt; r_{11}, r_{22} )</td>
<td>( p_1 \to \hat{p}_1 ), or sustained oscillations about ( \hat{p}_1 )</td>
</tr>
<tr>
<td>( w_{12} \leq w_{11}, w_{22} )</td>
<td>( p_1 \to 1 ), ( \text{or } p_1 \to 0 )</td>
<td>( r_{12} &lt; r_{11}, r_{22} )</td>
<td>( p_1 \to 1 ) ( \text{or } p_1 \to 0 )</td>
</tr>
</tbody>
</table>

* These results are due to Norton (1928), and were derived for a continuous-time model.

When there is heterozygote superiority, Norton showed that the gene frequency either converges to equilibrium or undergoes oscillations in which the gene frequency repeatedly passes through the equilibrium value. The local stability analysis given above can be extended to the continuous-time case by increasing the number of age-classes indefinitely; when combined with Norton’s result, it implies that there can be no oscillatory behaviour with heterozygote superiority; if the population converges to a neighbourhood of the equilibrium point, it must eventually reach the equilibrium itself. This result is confirmed by numerous computer calculations of particular cases. The biologically interesting possibility of a stable limit cycle in gene frequency, which was left open by Norton’s results, is therefore ruled out for the continuous-time model. No analogue of Norton’s results for discrete time has yet been obtained, except for some special cases studied by Pollak and Kempthorne (1971) and Charlesworth (1980, pp. 186–8).

### 4.5 Selection on quantitative characters

#### 4.5.1 General considerations

The results described in the preceding sections of this chapter have been concerned solely with the effects of natural selection on gene frequencies. In this section, we shall consider some aspects of the theory of the effects of selection on quantitative characters, which has been developed in recent years.
The case of selection on a single trait will be discussed first, primarily from the point of view of predicting the response to artificial selection. We will then consider selection on a multivariate character set, which is of relevance to the theory of natural selection on life-histories.

4.5.2 Prediction of the response to artificial selection

With artificial selection on a single trait, only individuals exceeding some chosen value of the character of interest to the breeder are used for breeding. In the classical, discrete-generation theory of artificial selection (Falconer, 1989, Chapter 11), it is assumed that the character under selection has a known heritability, $h^2$, defined as the proportion of the total variance in the character which is of additive genetic origin. In order to predict the rate of response to selection, we also need to know the selection differential, $S$, which is defined as the deviation from the mean of the unselected population of the mean phenotypic value of the parents chosen for breeding, in the generation under consideration. The selection response, $R$, (which should not be confused with the net reproduction rate) is defined as the deviation of the mean value of their offspring from the mean of the unselected population. If it is assumed that the parents are chosen independently of each other with respect to their phenotypic values, that linkage disequilibrium and epistatic contributions to the genetic variance are negligible, and that there are no environmental sources of resemblance between parent and offspring, we can write

$$R = h^2 S$$

(4.60)

This equation is only valid for one generation. If it can be assumed that the heritability and selection differential remain constant in time, repeated application of equation (4.60) enables us to write an equation for the cumulative selection response, $R_c$, which is defined as the deviation of the mean of the character in generation $t$ from its mean in the initial generation. We have

$$R_c = th^2 S$$

(4.61)

Provided that an estimate of $h^2$ is available from prior genetic experiments, this is a result of genuine predictive value. On theoretical and empirical grounds, we would expect $h^2$ to remain approximately constant for the first 5–10 generations of selection, so that equations (4.60 and 4.61) provide a useful means of predicting the initial rate of response to artificial selection (Falconer, 1989, Chapter 11).

Many species of economic importance, such as most domestic animals,
have populations with overlapping generations, so that it is not possible to apply equations (4.60 and 4.61) directly. It is therefore of importance to extend the above results to age-structured populations. Some approximate theory was developed by Dickerson and Hazel (1944) and Rendel and Robertson (1950). An exact matrix approach was introduced by Hill (1974, 1977) and, independently, by Elsen and Macquot (1974). Some refinements have been added by Pollak (1977). The results described below are based mainly on the papers by Hill.

Breeding values

We shall consider a population described by the discrete age-class model of section 2.3.2, in which the demographic parameters of individuals of both sexes are constant over time, and which is constant in size and age-structure. The size of the population is assumed, however, to be sufficiently large that sampling effects can be neglected. In order to develop the theory, we will make use of the concept of the additive genetic value or breeding value of an individual (Falconer, 1989, p. 118). We let $X(t) = [X(1, t), X(2, t), \ldots, X(d, t)]$ be a row vector of the mean breeding values of females aged 1, 2, \ldots, $d$ at time $t$, with respect to the character of interest. Similarly, $X^*(t)$ is the corresponding vector of mean breeding values for males aged 1, 2, \ldots, $d^*$. It is convenient to combine these into a single $(d + d^*)$-dimensional row vector, $\tilde{X}(t) = [X(t), X^*(t)]$ (cf. p. 65), which gives a full description of the state of the population at time $t$. This scheme is most easily understandable in terms of a character which is measured at a single age, $x$, such as weight at sexual maturity. A change in the mean breeding value of the individuals of a given sex who are of age $x$, from one time-interval to the next, must be exactly proportional to the corresponding change in the mean phenotypic value of the individuals in this age–sex class. Changes in the breeding values $X(x, t)$ and $X^*(x, t)$ can therefore be monitored by following the changes in mean phenotypic values of the females and males aged $x$. If, as is frequently the case, the character can be measured only in one sex, changes in the relevant breeding value can be detected through changes in the mean value of individuals of the appropriate age–sex class. Alternatively, the mean breeding value of individuals of a given class can, in principle, be determined by measurements of the mean value of the progeny produced when they are mated to a standard stock of constant composition. There is thus no theoretical difficulty in assigning breeding values to individuals on which the character of interest cannot be directly measured.
From the relationship of breeding value to the underlying allele frequencies at loci controlling the character (Falconer, 1989, p. 118), it follows that the following expression must hold in the absence of selection

\[ \tilde{X}(t) = \tilde{X}(0)G^t \] (4.62a)

where \( G \) is the matrix introduced on p. 81. For large \( t \), \( \tilde{X}(t) \) is thus given by the asymptotic expression

\[ \tilde{X}(t) \sim \tilde{X}(0)A \] (4.62b)

where \( A \) is defined on p. 83. From the form of \( A \) it follows that the mean breeding value asymptotically approaches the same, constant value for each class of individual, in the absence of selection. This reflects the approach of the underlying gene frequencies to constancy, discussed in section 2.2.1.

**Selection differentials with age-structure**

Suppose that artificial selection is imposed on the population only at time \( t = 0 \). The vector of mean breeding values \( \tilde{X}(0) \) for the unselected population is incremented by a vector \( s \), so that the vector of mean breeding values of the selected individuals is \( \tilde{X}(0) + s \). The components of \( s \) depend on the relationship of the character to age, the heritability, and the phenotypic selection differential. Assume, as above, that the character of interest can be measured only at age \( x \). If it can be measured in both sexes, the selection procedure at time \( t = 0 \) will therefore consist of selecting males and females aged \( x \) according to their phenotypes. Only these individuals or their survivors will be used for breeding at the current and future time-intervals. This selection on individuals aged \( x \) can be represented by the resulting increments in the mean values of the character, \( S \) for females and \( S^* \) for males. \( S \) and \( S^* \) thus define the selection differentials for females and males. If the heritability of the character is \( h^2 \) for both sexes, the components of \( s \) are given by the following expressions

\[ s(x) = h^2S \] (4.63a)

\[ s(d + x) = h^2S^* \] (4.63b)

\[ s(i) = 0 \quad (i \neq x, d + x) \] (4.63c)

If the character can be measured in only one sex, or if selection is practised on one sex alone, all the components of \( s \) except that for the appropriate age-sex class are zero.
The response to selection

The response to the selection imposed at \( t = 0 \), in terms of breeding values, can be found by applying the matrix \( G \) to the vector of breeding values of the selected individuals. We thus have

\[
\tilde{X}(t) = \tilde{X}(0)G^t + sG^t \tag{4.64}
\]

This assumes that the age as well as breeding value of one parent are independent of those of the other. In addition, the age-structure of the population is assumed to be unchanged by the selection regime. Comparing equations (4.62a) and (4.64), it follows that selection has resulted in an increment \( sG^t \) in the breeding value vector. Only part of this is due to the genetic effects of selection, since \( \tilde{X}(t) \) contains components which have been incremented solely because they relate to individuals who have survived from \( t = 0 \). In order to distinguish the genetic effects of selection, it is useful to define a matrix \( \tilde{G} \) which contains only the survival-related components of \( G \), i.e. in which the columns 1 and \( d + 1 \) have been deleted from \( G \). This genetic response to selection at \( t = 0 \) is thus given by the vector \( s(G^t - \tilde{G}^t) \).

If selection is now assumed to be applied at each time-interval, instead of just at \( t = 0 \), we can describe the cumulative response to selection at an arbitrary time \( t \) by a vector of the net increases from \( t = 0 \) in mean breeding values of each class of individuals, \( R_c(t) \). This vector is given by the sum of the responses due to selection in each time-interval prior to \( t \). If selection is practised with the same intensity in each time-interval, and if \( h^2 \) remains constant, we obtain the following relation

\[
R_c(t) = \sum_{u=1}^{t} s(G^u - \tilde{G}^u) \tag{4.65}
\]

For a character which can be measured at age \( x \) in both sexes, the phenotypic cumulative response at time \( t \) is given by the components \( R_c(x, t) \) and \( R_c(d + x, t) \), for females and males, respectively.

Equation (4.65) gives an exact prediction of the selection response, subject to the assumptions of the model. Provided that the matrices \( G \) and \( \tilde{G} \) and the vector \( s \) have been specified by the structure of the breeding programme, it is straightforward to use the equation to compute the expected cumulative response at time \( t \). An asymptotic formula can be given for the response at time \( t \), \( R(t) \), defined as \( R_c(t) - R_c(t-1) \). (This corresponds to the response given by the classical prediction equation (4.60).) We know that \( G^t \) approaches the matrix \( A \) of p. 83, when \( t \) becomes large. Furthermore, when \( t \geq d, d^* \), \( \tilde{G}^t \) is a matrix of zero elements, since no individual can survive so long. From equation (4.65), we have
Selection dynamics

\[ R(t) = s(G' - \tilde{G}') \]

so that

\[ R(t) \sim sA \quad (4.66) \]

The response per time-interval therefore approaches the same asymptotic value, \( R \) say, for each age-class. Using the definition of \( A \), this asymptotic response can be written as

\[ R = \sum_{i=1}^{d+d^*} s(i)q(i)/2\tilde{T} \quad (4.67a) \]

where the terms \( q(i) \) and \( \tilde{T} \) are defined on p. 82. In the case of a character which is measured at age \( x \) in both sexes, this expression can be combined with equations (4.63) to give

\[ R = h^2[Sq(x) + S^*q(d + x)]/2\tilde{T} \quad (4.67b) \]

If the character can be measured in one sex only, the \( q \) term corresponding to the other sex must be omitted from this expression.

Hill (1974, 1977) has shown that the asymptotic response is approached fairly quickly. It is therefore tempting to use \( R \) to predict the cumulative response to selection by an expression analogous to equation (4.61), i.e. by writing

\[ R_c = tR \quad (4.68) \]

As shown by Hill, this provides a good approximation to the cumulative response given by the exact equation (4.65); the difference between the exact and approximate values converges rapidly to a constant level, so that there is little cumulative error in using equations (4.67) and (4.68).

A numerical example

As an example of possible applications of these results, consider the following case of an imaginary herd of cattle, discussed by Hill (1977). The herd is maintained in such a way that all progeny are got by bulls aged two years; one-third each come from cows aged two, three and four years, respectively. The matrix \( G \) therefore has the form

\[
\begin{bmatrix}
0 & 1 & 0 & 0 & 0 & 0 \\
1/6 & 0 & 1 & 0 & 1/6 & 0 \\
1/6 & 0 & 0 & 1 & 1/6 & 0 \\
1/6 & 0 & 0 & 0 & 1/6 & 0 \\
0 & 0 & 0 & 0 & 0 & 1 \\
1/2 & 0 & 0 & 0 & 1/2 & 0
\end{bmatrix}
\]
The vector $q$ is equal to $(1, 1, \frac{2}{3}, \frac{1}{3}, 1, 1)$, so that the generation time, $\bar{T}$, is equal to 2.5 years. We assume that selection is practised for weight at one year of age, on bulls only. If weight is normally distributed with standard deviation 40 kg, and the 7% heaviest bulls are retained for breeding, then standard selection theory tells us that the selection differential on bulls is $S^* = 40 \times 1.9 = 70.6$ (cf. Falconer, 1989, Chapter 11). If the heritability of weight at one year is 0.4, we obtain the following value for $R$ from equation (4.67b)

$$R = \frac{(0.4 \times 70.6)}{(2 \times 2.5)} = 6.08 \text{ kg/year}$$

This expected response can be compared with that for a population in which males are retained for two breeding seasons and females for five, each fertile age-class being assumed to contribute an equal fraction of the progeny. If selection is again practised for weight at 1 year, but now 3.5% of males and 60% of females are selected, the selection differentials $S$ and $S^*$ are equal to 24.0 and 88.0, respectively. $\bar{T}$ is now found to be equal to 3.25 years, and we have $q(1) = q(7) = 1.0$. Equation (4.67b) gives us an expected asymptotic response of

$$R = \frac{(0.4 \times 112)}{(2 \times 3.25)} = 6.89 \text{ kg/year}$$

There is therefore little to choose between the two schemes.

### 4.5.3 Natural selection on quantitative characters

Most empirical work on the evolutionary genetics of life-history traits involves the use of quantitative genetic methods (see section 5.4). It is therefore important to have a sound theoretical framework for interpreting the effects of natural selection on quantitative characters in age-structured populations. Life-histories inherently involve a multiplicity of age-related traits, and so a multivariate description must be used. Work on this topic was initiated by Lande (1982). The basic theory has been applied to a variety of problems by Charnov (1989), Charlesworth (1984, 1990a), Kirkpatrick, Lofsvold and Bulmer (1990), Law (1991), and Kirkpatrick and Lofsvold (1992), among others.

We shall assume that the system of interest is described by a vector $z$, whose components are in general age-specific, e.g. $z$ for a given phenotype might be composed of the set of specific survival probabilities and fecundities for that phenotype (Lande, 1982), or be a growth-trajectory which determines these traits (Kirkpatrick et al., 1990; Kirkpatrick and Lofsvold, 1992). We also assume that knowledge of $z$ for a given phenotype allows the intrinsic rate of increase for that phenotype to be calculated. Following standard quantitative genetic procedures (Lande, 1982), the state of the population at a given time $t$ is determined by the
vector of means $\tilde{z}(t)$ for the new zygotes formed at time $t$, and by the corresponding matrix of additive genetic covariances $G = \{g_{kl}\}$, where $g_{kl}$ is the additive genetic covariance between the $k$th and $l$th components of $z$ ($g_{kk}$ is the additive genetic variance of $z_k$).

**Multivariate selection dynamics**

An equation for the rate of change of $\tilde{z}(t)$ can be derived as follows, on the assumption that the life-history traits are controlled by a number of loci, each of small effect, which combine additively to produce the values associated with each multi-locus genotype. It is also assumed that allele frequency changes at each locus are approximately the same as if there is no linkage disequilibrium. This is reasonable if selection on each locus is weak and linkage is loose (cf. equation (4.20); Nagylaki, 1992, Chapter 8, 1993). These two assumptions together imply that the rate of change of mean phenotype can be deduced by summing the effects contributed by the changes in allele frequencies at each locus affecting the trait in question. The following derivation is based on that of Charlesworth (1993).

Assume that the value of $z$ for a cohort of individuals determines their reproductive function $k(z, x)$, and the corresponding intrinsic rate of increase, $r(z)$. The reproductive function for a given genotype is thus the mean of $k(z, x)$ over the environmentally determined distribution of phenotypes around the mean value of $z$ for individuals with the genotype in question. Let the mean vector associated with individuals of genotype $A_iA_j$ at the $m$th locus be $z_{ijm}^m$. The assumption of additivity between loci, together with random mating of ages and genotypes, implies that

$$z(t) = 2 \sum_{ijm} p^m_i(t)p^m_j(t)z_{ijm}^m$$  \hspace{1cm} (4.69)

where $p^m_i(t)$ is the frequency of the $i$th allele at the $m$th locus. If there is additivity within as well as between loci, we can write $z_{ijm}^m = z_{im}^m + z_{jm}^m$, where $z_{im}^m$ is the effect of the $i$th allele at the $m$th locus. If there is dominance, the allelic effects depend on the state of the population at the locus in question, and hence change with time (Falconer, 1989, Chapter 7). If the number of loci affecting the trait is large, then the changes at each locus are negligible over a short period of time, and so the $z_{im}^m$ can be treated as constant even with dominance.

In order to obtain a recurrence relationship for $\tilde{z}(t)$, we introduce the variable $\tilde{z}_{ijm}^m(t) = p^m_i(t)z_{ijm}^m$. Let $\varepsilon$ be the measure of the overall strength of selection on the set of traits, analogous to the measure for a single locus introduced on p. 137. Using an argument similar to that leading to
equation (4.9), and dropping the arguments from $\hat{z}(t)$ etc., we obtain, after some manipulations (Charlesworth, 1993), the relation

$$T(r_\hat{z}) \Delta \hat{z} = 2 \sum_{im} \hat{z}_i^m \left[ w_i^m (r_\hat{z}) - 1 \right] + O(\varepsilon^2) \quad (4.70)$$

where $r_\hat{z}$ is the intrinsic rate associated with the current population mean vector $\hat{z}$, which satisfies the equation

$$\sum_x e^{-r_\hat{z}x} \sum_{ijm} p_i^m p_j^m k_{ij}^m(x) = 1 \quad (4.71)$$

$T(r_\hat{z})$ is the corresponding generation time for the population as a whole, and $w_i^m (r_\hat{z})$ is the average fitness associated with the $i$th allele at the $m$th locus in a population growing at rate $r_\hat{z}$. These are defined in the same way as the analogous quantities in equations (4.8).

The mean fitness of the population is given by the mean of the average fitnesses of each allele at each locus:

$$\bar{w}(r_\hat{z}) = \sum_{im} p_i^m w_i^m (r_\hat{z}) = 1 \quad (4.72)$$

From now on, the argument $r_\hat{z}$ will be usually be omitted from $w$, $T$, etc.

Write $\nabla \bar{w}$ for the gradient vector of $\bar{w}$ with respect to $\hat{z}$, $\nabla \bar{w} = (\partial \bar{w}/\partial \hat{z}_1, \partial \bar{w}/\partial \hat{z}_2, \ldots)$, noting that the argument $r_\hat{z}$ is held constant while carrying out the differentiations. Further manipulations (Charlesworth, 1993) lead to the result that

$$T \Delta \hat{z} = 2 \nabla \bar{w} \sum_{im} p_i^m (z_i^m - \hat{z})^T (z_i^m - \hat{z}) + O(\varepsilon^2) \quad (4.73)$$

From standard quantitative genetics theory, the assumption of additive genetic effects among loci implies that

$$G = 2 \sum_{im} p_i^m (z_i^m - \bar{z})^T (z_i^m - \bar{z}) \quad (4.74)$$

This yields

$$T \Delta \hat{z} = \nabla \bar{w} G + O(\varepsilon^2) \quad (4.75a)$$

Using implicit differentiation of equation (4.71) (cf. equations (1.58)), we have $T \nabla \bar{w} = \nabla r_\hat{z}$, where $\nabla r_\hat{z} = (\partial r_\hat{z}/\partial \hat{z}_1, \partial r_\hat{z}/\partial \hat{z}_2, \ldots)$.

Hence

$$\Delta \hat{z} = \nabla r_\hat{z} G + O(\varepsilon^2) \quad (4.75b)$$
If the $O(e^2)$ term is ignored, this is equivalent to the equation derived by Lande (1982), using normal distribution theory (which is equivalent to our assumption of small effects of each locus) and the assumption that the population is always in stable age-distribution.

If the phenotypic variances and covariances of $z$ are sufficiently small that their second and higher powers can be neglected, and the third- and higher-order derivatives of $r$ with respect to the components of $z$ are of similar magnitude to the first derivatives (Charlesworth, 1993; Abrams, Harada and Matsuda, 1993), we can write

$$\nabla r \approx \nabla r(\bar{z})$$  \hspace{1cm} (4.76)

where $\nabla r(\bar{z})$ is the vector of partial derivatives of $r$ with respect to the components of $z$ evaluated at the population mean ($[\partial r/\partial z_1]_{\bar{z}}, [\partial r/\partial z_2]_{\bar{z}}, \ldots$). Under the above conditions, $\nabla r(\bar{z})$ can be substituted for $\nabla r_z$ in equations (4.75) without much loss of accuracy. This provides a form which is usually more convenient for computational purposes. The result is exact if $r$ is a linear or quadratic function of $z$ (Charlesworth, 1993; Abrams et al. 1993).

While this analysis has assumed density-independent demography, an essentially similar analysis can be carried out for the case of density-dependent population regulation, using the gene-frequency equations derived in section 4.2.2. The intrinsic rate of increase $r$ is replaced as a fitness measure by the fitness measures of equations (4.23) or (4.24); all other expressions remain unchanged. Similarly, the case of small sex differences in demographic parameters can be analysed by the same averaging procedure employed to obtain equation (4.17). Modifications to deal with frequency-dependence are discussed by Charlesworth (1993).

**Equilibrium conditions**

At demographic and genetic equilibrium, the order $e^2$ terms which were neglected in order to derive equations (4.73) must vanish. This implies that, given the assumption of small allelic effects at each locus contributing to additive variation in $z$ (needed to obtain equation (4.70)), we have

$$0 = \nabla r_z G$$  \hspace{1cm} (4.77)

where $0$ is the vector of zero elements.

This equation has some important implications for experimental studies of life-history genetics (Charlesworth, 1984, 1990a; Charnov, 1989). It means that either all the components of $G$ are equal to zero, that $\nabla r_z = 0$, or that the determinant of $G$ is zero. If the components of $z$ are life-history variables such as fecundity or survival (or increasing functions
of these variables), the derivatives of $r$ must be positive, and so the second of these conditions cannot apply. In this case, if there is additive genetic variance in some of the $z_k$, so that the corresponding $g_{kk}$ are positive, equation (4.77) requires that there be negative additive genetic covariances between some of the $z_k$ in order for the terms on the right-hand side to sum to zero, consistent with the third of the conditions. This conclusion about equilibrium does not require that selection be frequency-independent, although the dynamical equations (4.75) assume frequency-independence (Charlesworth, 1990a, 1993).

Hence, under rather general conditions, selection theory predicts that life-history traits in equilibrium populations which exhibit additive genetic variance which is maintained by selection will be expected to show negative genetic covariances with some other additively variable traits. This has been known to specialists in animal improvement for a long time (Dickerson, 1955; Robertson, 1955). It is a consequence of the fact that selection exhausts additive genetic variance for net fitness, so that any remaining additive genetic variation in fitness-related traits must reflect the properties of genes whose beneficial effects on some traits are counter-balanced by deleterious effects on other characters (antagonistic pleiotropy: see Chapter 5, section 5.3.1). Rose (1982, 1985) has provided some explicit models of how this can be achieved. To the extent that genetic variation in life-history traits reflects non-equilibrium situations or the action of non-selective pressures, such as mutation, this conclusion is weakened (Charlesworth, 1990a; Houle, 1991). Nevertheless, it suggests that genetically variable components of the life-history may often have negative genetic covariances with other traits.

An important qualification of this conclusion follows from an analysis of the relations between genetic covariances and underlying functional constraints between life-history traits, which shows that positive genetic covariances between some pairs of traits will usually be expected at equilibrium under selection (Charlesworth, 1990a). Only in exceptional cases, such as when functional constraints exist only between isolated pairs of variables (e.g. between $z_1$ and $z_2$, $z_3$ and $z_4$, etc.), will there be only negative or zero covariances between all pairs of traits (Charnov, 1989; Charlesworth, 1990a). As discussed in Chapter 5, section 5.4.5, this complicates the use of quantitative genetic methods to infer the existence of functional relations among life-history variables.

### 4.6 Conclusions

In this section we bring together the main conclusions from the work described above, and discuss their biological implications. It is
convenient to consider density-independent and density-dependent populations separately.

4.6.1 Density-independent populations

**Fitness measures and their merits**

As should be clear from the analyses described above, the theoretical basis for predicting the rate and direction of gene frequency change is less complete than that for studying genetic equilibrium, which was described in the preceding chapter. In particular, the effects of non-random mating with respect to age have not yet been satisfactorily dealt with. When there is random mating between age-classes, and when male and female demographic parameters are the same, it is clear that the intrinsic rate of increase of a genotype, as defined by equation (3.34), is an adequate measure of its fitness for many purposes, provided that the population is in a temporally constant, density-independent environment. The ultimate fate of a new allele at a diallelic locus can be predicted from the relations between the genotypic intrinsic rates, by the results of Norton (1928), Pollak and Kempthorne (1971) and Charlesworth (1974a), which are summarised in section 4.4. The rate of spread of a rare dominant or semi-dominant allele is proportional to its effect on the intrinsic rate of increase (sections 4.3.1 and 4.3.2). Finally, with weak selection, equations for gene frequency change can be derived which are similar in form to the first-order difference equations of discrete generation population genetics, or the Malthusian parameter equations of Fisher (1930, 1941), as described in section 4.2.1.

To make equations (4.11) and (4.15) more comparable with those of discrete-generation population genetics, one can choose the generation-time of a standard genotype, \( T_s = \sum e^{-r_s x} k_s(x) \), as a biological unit of time. The equations for \( \Delta p_t \) and \( \Delta^2 \ln B \) can be multiplied by \( T_s \) and \( T_s^2 \) to obtain 'per generation' changes. (A similar operation can obviously be carried out for the corresponding continuous-time model.) The 'per generation' measure of the selective difference between two genotypes with intrinsic rates \( r_{ij} \) and \( r_{im} \) can then be taken as \( T_s (r_{ij} - r_{im}) \). By the argument used in equation (4.10), we have

\[
T_s (r_{ij} - r_{im}) = \sum_x e^{-r_s x} k_{ij}(x) - \sum_x e^{-r_s x} k_{im}(x) + O(e^2)
\]  

(4.78)

This is in turn related to equation (3.31) for equilibrium fitness, and is in fact equal to the difference \( w_{ij} - w_{im} \) as given by that equation, plus second-order terms. The equilibrium fitness measure and the intrinsic
rate of increase are thus unified by this approximation. When the intrinsic rates of increase of all the genotypes are close to zero, equation (4.78) can be further approximated to give

$$\left(\sum_x x k_s(x)\right)(r_{ij} - r_{lm}) \cong \sum_x k_{ij}(x) - \sum_x k_{lm}(x)$$

If this is used in equation (4.11) to obtain the per generation rate of change of gene frequency, we obtain the type of result that Haldane (1927a) first derived. For weak selection in a slowly growing population, we can conclude that the net reproduction rate of a genotype provides an approximate measure of its fitness, both for predicting the composition of equilibrium populations and for obtaining the rate of change of gene frequency per generation. It is widely used in studies of fitness in natural populations (Clutton-Brock, 1988).

If there are demographic differences between the sexes, so that equations (3.12) have to be used as a starting-point, the analysis of pp. 143–5 shows that, with weak selection and random mating with respect to age, the intrinsic rate of increase of a genotype is replaced as its measure of fitness by the mean of the male and female intrinsic rates, the male rate being measured under the demographic conditions of some arbitrary standard population with a stable age-structure. This fitness measure can be translated into a per generation measure by the same type of approximations used above, with the male and female intrinsic rates being multiplied by the male and female ‘generation times’, respectively. For the case of strong selection, and with possible non-random mating between age-classes, equations (4.28) and (4.29) show how the rate of spread of a non-recessive allele can be determined with demographic differences between the sexes.

The general conclusion is, therefore, that for the case of weak selection and random mating with respect to age, the intrinsic rate of increase of a genotype or, more generally, the mean of the male and female intrinsic rates, provides an adequate measure of fitness in a density-independent and constant environment. This parameter can be used in much the same way as the discrete generation fitness to predict, to a good approximation, the rate of change of frequency and ultimate composition of populations with respect to single loci. Provided that linkage is either very tight or is rather loose, in comparison with the selection intensity, the analysis of equation (4.20) shows that this is true for the two-locus model as well, with some minor qualifications.

If selection is strong, however, there is no unitary measure of fitness. As we have seen (section 4.3.1), the intrinsic rate of increase provides a
good predictor of the rate of spread of a rare non-recessive gene, but a
different fitness measure must be used for a rare recessive gene. Furthermore,
in the neighbourhood of a polymorphic equilibrium, the intrinsic
rate of increase will not predict the rate of change of gene frequency
accurately (pp. 140–3), and it also gives an inaccurate prediction of the
composition of the equilibrium population (section 3.3.3). This conclu-
sion is important for experimenters who wish to study selection by
following gene frequency changes in a continuously breeding population
such as a *Drosophila* population cage. This has been one of the classical
methods for estimating genotypic fitnesses (cf. Du Mouchel and Ander-
son, 1968). In many cases, particularly in studies of inversion polymorph-
isms, strong selection has been detected, and some evidence for apparent
changes in selection coefficients during the course of selection has been
obtained (Dobzhansky and Levene, 1951; Watanabe et al., 1970). The
fact that, in age-structured populations, strong selection may lead to
deviations from the changes in gene frequency which would be expected
on the basis of constant fitness parameters suggests the need for caution
in interpreting apparent evidence for frequency-dependent selection.

One final point concerning strong selection is worth mentioning. With
heterozygote superiority in the genotypic intrinsic rates of increase, we
have seen that the equilibrium is locally stable (section 4.3.6). The
analysis of Norton (1928) left open the possibility that there might be
sustained oscillations in gene frequency in this case, since he proved only
that there was convergence to a neighbourhood of the equilibrium, but
did not demonstrate convergence to the equilibrium itself. The existence
of such oscillations would have been of considerable interest to students
of gene frequency changes in fluctuating populations, particularly in view
of Chitty’s (1960) hypothesis that gene frequency changes may drive
cycles of population numbers. As we have seen in section 4.4, oscillations
do not appear to be possible in this case, at least as far as the continuous-
time model is concerned.

Demetrius (1975, 1989) has argued that a measure of entropy of the
life-history is an useful predictor of gene-frequency change. While this
may be true under some special circumstances (Demetrius, 1989), the
result is derived from the properties of the intrinsic rate of increase, and
therefore offers no new insights into the dynamics of selection.

*Malthusian parameters and gene frequency equations*

Fisher (1930, 1941) introduced equations for gene frequency
change in continuous-time models of the form
\[ \frac{dp_i}{dt} = p(x) (m_i - \bar{m}) \] (4.79)

where \( m_i - \bar{m} \) is the average excess in Malthusian parameters of allele \( A_i \), and \( p_i \) is the frequency of allele \( A_i \) among the population as a whole. Fisher appears to have identified the Malthusian parameter of an individual with its intrinsic rate of increase as we have defined it here (see pp. 37-8 of the 1958 edition of The Genetical Theory of Natural Selection (Fisher, 1930)). Later users of equations of this form have usually assumed that the Malthusian parameter \( m_{ij} \) of a specific genotype \( A_iA_j \) is equal to the difference between the per capita birth-rate and death-rate for that genotype, and that in principle these can be treated as independent of the genotype-composition and age-composition of the population (e.g. Crow and Kimura, 1970, pp. 190-5). (As we saw in Chapter 1, p. 33, a homogeneous population with the demographic characteristics of \( A_iA_j \) will have an asymptotic value for the difference between its per capita birth-rate and death-rate which is equal to the intrinsic rate of increase.) It is also usually assumed that the frequencies of the genotypes in the whole population can be approximated adequately by Hardy-Weinberg frequencies, so that \( m_i \) and \( \bar{m} \) in equation (4.79) can be written as

\[
m_i = \sum_j p_j m_{ij} \\
\bar{m} = \sum_i p_i m_i
\]

As mentioned in section 3.1, however, it is not in general possible to assign a fixed birth-rate and death-rate to each genotype in a population whose genetic composition and age-structure are being altered by selection. (Moran, 1962, p. 60; Charlesworth, 1970; Pollak and Kempthorne, 1971). In fact, this can strictly only be done if \( m_i(x) \) and \( d\ln l_i(x)/dx \) are independent of \( x \) for all \( x \). This is clearly impossible with the type of reproductive schedule characteristic of most organisms, where reproduction starts some time after birth, so that \( m_i(x) \) necessarily depends on \( x \).

Despite these difficulties, it turns out, as we have seen, that an equation similar in form to equation (4.79) can be used to describe gene-frequency change in an age-structured population for a continuous-time model with weak selection (equation (4.12)). The gene frequency is measured among new zygotes rather than the whole population. With sufficiently weak selection, this will not make much difference in practice. The use of this equation leads to the conclusion that the rate of change of the population growth-rate under selection is equal to the additive genetic variance in intrinsic rates of increase (the continuous-time analogue of...
equation (4.14)), which is similar to the Fundamental Theorem that Fisher derived using his Malthusian parameter equation for gene frequency change. Fisher gave no indication, however, that his results were to be restricted to the case of weak selection, together with the other assumptions discussed in Chapter 3; these assumptions are necessary for the basic difference or integral equations for gene frequency and population size to be valid. It is remarkable that equations similar to those proposed by Fisher do, in fact, provide good approximations to the rate of gene-frequency change with weak selection. Other interpretations of the Malthusian parameter approach to age-structured models have been proposed by Pollak and Kempthorne (1971), Price (1972), Price and Smith (1972), Nagylaki and Crow (1974) and Nagylaki (1992, Chapter 2). It is obviously not possible to decide which, if any, of these corresponds to what Fisher had in mind; it is more important to have some firmly-established methods for studying selection in age-structured populations. (Crow (1978) has shown how weighting of individuals by their reproductive value can overcome some of the difficulties of Fisher's approach.)

Furthermore, with temporally-varying environments the Malthusian parameters \( m_{ij} \) would be functions of the state of the environment at a given moment, rather than fixed constants; however, equations having the form of equation (4.79) are misleading when regarded as descriptive of gene-frequency change in an age-structured population in this situation. Consider for example the two-allele case with \( A_2 \) recessive to \( A_1 \). Assuming Hardy–Weinberg frequencies, and examining the criterion for \( A_2 \) to increase on the average when rare (cf. section 4.3.3), gives the condition

\[ E[m_{11}] < E[m_{22}], \]

where \( E \) denotes expectation over all environmental states. The criterion for \( A_1 \) to increase when rare is the reverse of this, so that one concludes that a polymorphism cannot be maintained by temporal changes in the environment when there is complete dominance, in contrast to the discrete-generation model of Haldane and Jayakar (1963). But we have seen that it is possible to produce examples in which the Haldane–Jayakar model works for age-structured populations, due to the fact that the rate of increase of a rare recessive gene is not controlled solely by its effect on the intrinsic rate of increase when selection is strong (section 4.3.3). Another application of equations of the form of equation (4.79) has been to regard them as approximations to discrete-generation difference equations (cf. Kimura, 1955, p. 45). It has been pointed out
that, with temporally-varying environments this form of approximation is not adequate (Jensen, 1973; Gillespie, 1973b), for rather different reasons from those just discussed for the case of age-structured populations.

4.6.2 Density-dependent populations

The results derived in section 4.2.2 show that, with weak selection, the genotypic carrying-capacities as defined by equation (3.40) control the direction of selection in the same way as the intrinsic rates of increase in the density-independent case. In the one-locus case, natural selection tends to maximise the number of individuals in the critical age-group as opposed to the maximisation of the population growth-rate in a density-independent population. It should be borne in mind that this result depends on the assumption that the net reproduction rate of each genotype is a strictly decreasing function of the number of individuals in the critical age-group, and that the maximisation principle applies to the number of individuals in the critical age-group, not necessarily to the population as a whole, or to age-classes outside the critical ones.

The idea that natural selection should, under suitable conditions, result in increasing population density goes back at least as far as Fisher (1930, Chapter 2). Its significance as a general principle was stressed especially by Nicholson (reviewed in Nicholson (1960)). Mathematical treatments for the discrete-generation case have been given by Charlesworth (1971), Roughgarden (1976), Kimura (1978), Nagylaki (1979b, 1992, Chapter 4), and Ginzburg (1983), with conclusions very similar to those described here. Asmussen (1979) has analysed density-dependent selection with cyclical or chaotic behaviour of population size.

Buzzati-Traverso (1955), Beardmore, Dobzhansky and Pavlovsky (1960) and Nicholson (1960), as well as a number of later workers (e.g. Mueller, 1991; Costantino and Desharnais, 1991), have described measurements on experimental populations in which population density or biomass has increased as a result of selection.

It should be noted that the conclusions about the control of the direction of selection by the genotypic carrying-capacities and the maximisation of population size do not require the different genotypes to respond differently to the density-dependent regulating factors. As in the discrete-generation case (Charlesworth, 1971), it is perfectly possible for genotypic differences in density-independent components of the life-cycle to result in differences in the carrying-capacities. This may be seen from an example in which density-regulation occurs through a response of the survival of immature stages to the total number of immature individuals.
Selection dynamics

at any one time, but where selection affects fecundity. Let the density-
dependent probability of survival to reproductive maturity be \( w^{(D)}(N) \), and the density-independent net expectation of offspring for an individual who has survived to adulthood be \( w^{(I)}_{ij} \) for genotype \( A_iA_j \). The genotypic carrying-capacity, \( N_{ij} \), is defined by

\[
\sum_x k_{ij}(x, N) = w^{(D)}(N)w^{(I)}_{ij} = 1
\]

From the fact that \( w^{(D)}(N) \) is a decreasing function of \( N \), it follows that genotypes with high values of \( w^{(I)}_{ij} \) have high carrying-capacities.

This result shows that the often-quoted statement of Haldane (1953), that natural selection will increase the density of a population only if genotypes differ in their ability to survive a density-dependent source of mortality, is not generally valid. Some experiments of Nicholson (1960) where he artificially adjusted the mortality rates of adult sheep blow-flies in a situation where the size of the population was limited by female fecundity, and found that population density was inversely related to mortality, provide a model which illustrates this fact.

These conclusions about the maximisation of population size depend on the assumptions of frequency-independent selection and a single-locus genetic system. With discrete generations, it is well known that mean fitness is not necessarily increased when there is frequency-dependent selection, or in multi-locus systems except when linkage is very tight or very loose (reviewed by Turner, 1970). The same applies to population size in the present case. In addition, we have considered only the case of a single-species ecological system. As shown by Levins (1975), Roughgarden (1976) and Léon and Charlesworth (1978) for discrete-generation models, evolutionary change in multi-species systems does not necessarily lead to maximisation of population size. For example, in a predator-prey system where the predator is only weakly regulated by its own numbers, evolutionary change in the predator species tends to increase its own numbers but decreases the numbers of the prey, whereas evolution of the prey species has only a trivial effect on its own numbers. Similar conclusions can easily be derived for age-structured populations by the methods described here for the single-species density-dependent case. For further discussions, see Ginzburg (1983) and Costantino and Desharnais (1991).

4.6.3 Quantitative genetics and life-history evolution

Two main theoretical techniques have been employed in the study of life-history evolution. The first is the method of evolutionary optimisation theory, which assumes that natural selection will lead to the
establishment of a life-history which maximises Darwinian fitness, subject to a set of constraints among the life-history traits in question (see e.g. Gadgil and Bossert, 1970; Schaffer, 1974a, 1983; Sibly and Calow, 1986; Roff, 1992, Stearns, 1992). In such models, fitness is usually equated to the intrinsic rate of increase, $r$, and the optimal life-history is determined by determining the maximum of $r$ with respect to variation in the set of trait values $z$, subject to an assumed set of functional constraints among the components of $z$, such that a vector function $f(z) = 0$ exists (see section 5.3 for specific examples). A genetical justification for this approach is provided by the fact that a mutant gene which perturbs the life-history can only invade the population if it confers a higher intrinsic rate of increase on its carriers (section 4.3.1), so that a population which is at a local maximum in $r$ constitutes an evolutionarily stable strategy (ESS) (Maynard Smith, 1982). Alternatively (and more realistically, given the generally polygenic control of life-history variables), a quantitative genetics model of life-history evolution can be employed (section 4.5.3). Here, the effects of selection on the population mean values of a set of life-history traits are determined by the genetic variance-covariance matrix for the set of traits, and the vector of selection gradients of $r$ with respect to the population mean values of the traits. In principle, the evolutionary equilibrium life-history can be determined from the conditions for the changes in the mean life-history variables to be zero, given knowledge of the genetic variance-covariance matrix, $G$.

The extent to which these two approaches lead to the same conclusions about the effects of natural selection on life-histories has been the subject of discussion (Clark, 1987; Charnov, 1989; Pease and Bull, 1988; Charlesworth, 1990a, 1993; Abrams et al. 1993). The results derived by Charlesworth (1990a), using a model of underlying functional constraints to generate the $G$ matrix, show that there is approximate correspondence between the results of an evolutionary optimisation model of a constrained multivariate system, and of the quantitative genetics model. The equilibrium mean values of the characters under the quantitative genetics model approximate the solution to the corresponding optimisation model, provided that the traits have small variances and covariances (section 4.5.3). This condition is likely to hold if $z$ is measured on a logarithmic scale, provided that the coefficients of variation on the original scale are small. If the fitness functions are linear or quadratic with respect to the trait vector $z$, and the constraint functions are linear, the correspondence between the quantitative genetics and optimisation models is exact. Extensions to the case of discrete variables, and to frequency-dependent selection, are discussed by Charlesworth (1990a, 1993).
### 4.6.4 Determining optimal life-histories

Under the conditions just mentioned, the problem of predicting the equilibrium attained under selection on a set of life-history traits subject to a set of constraints reduces to that of finding the maximum of the fitness measure with respect to perturbations of \( z \). If there is frequency-dependence, as is often the case when fitness is influenced by competition, the procedure can be modified by representing fitness of individuals with phenotype \( z \) as a function both of \( z \) and of the population mean \( \bar{z} \), \( w(z, \bar{z}) \), the second term representing the dependence of fitness on the composition of the population as a whole (Charlesworth, 1990a, 1993). The maximisation procedure is then carried out with respect to \( z \), holding \( \bar{z} \) constant, giving results which are equivalent to the standard ESS analysis (Charlesworth, 1990a, 1993).

In the case of a density-independent population, the optimal life-history corresponds to the maximum in the intrinsic rate of increase, \( r \). The technicalities of maximising \( r \) are simplified to some extent by the following facts, discussed in more detail by Taylor, Gourley and Lawrence (1974). If a life-history corresponding to a certain reproductive function \( k'(x) \) yields a local maximum in \( r \), \( r' \) say, the characteristic equation (1.45) gives the relation

$$
\sum e^{-r'x} k'(x) = 1
$$

(4.80a)

Any small perturbation to another reproductive function \( k''(x) \neq k'(x) \) with, by definition, \( r'' < r' \), must therefore be such that

$$
\sum e^{-r''x} k''(x) < \sum e^{-r'x} k'(x) = 1
$$

(4.80b)

The sum \( \sum e^{-r''x} k(x) \) is therefore at a local maximum when \( k(x) = k'(x) \). Conversely, if \( \sum e^{-r'x} k(x) \) is at a local maximum when \( k(x) = k'(x) \), \( r' \) is a local maximum in \( r \). This can be summed up by saying that the necessary and sufficient conditions for a life-history \( k'(x) \) to be at a local maximum in \( r \) is that the sum \( \sum e^{-r'x} k(x) \) is at a maximum when \( k(x) = k'(x) \), with \( r' \) being specified by equation (4.80a). The conditions for a global maximum in \( r \) can be expressed similarly.

Analogous principles can be developed for studying life-history evolution in a density-dependent environment (Charlesworth and León, 1976). Using the model of density-dependence developed in section 1.4.2, a given life history \( k'(x) \) is associated with a carrying-capacity \( \hat{N}' \), determined by the equation

$$
w'('\hat{N}') = \sum x k'(x, '\hat{N}') = 1
$$

(4.81a)
If \( w(N) \) is a strictly decreasing function of \( N \), it follows from the results described in section 4.3.2 that a mutant gene associated with a perturbed life history \( k''(x) \) will be eliminated if and only if

\[
w''(\hat{N}') = \sum_x k''(x, \hat{N}') < w'(\hat{N}') = 1 \quad (4.81b)
\]

This is, therefore, the necessary and sufficient condition for the life-history \( k'(x) \) to be an ESS. It is also the condition for the carrying-capacity \( \hat{N}' \) to be a maximum. There is thus no real difference between the density-dependent and density-independent cases, as far as the techniques of determining the ESS life-histories are concerned.

When applying this general procedure to a particular problem, it is important to distinguish between local and global maxima in fitness. It is perfectly possible in principle for there to be several points in the space specified by the permissible values of \( z \) which are local maxima in fitness; each of these constitutes an ESS with respect to small perturbations in \( z \). Only one of these can be a global maximum in fitness, except in the unlikely event that two or more turn out to be associated with exactly equal fitnesses. Some examples of this are discussed by Schaffer and Rosenzweig (1977). It may be difficult for a population which for historical reasons is located at an ESS to approach the global maximum, unless a mutation of large effect occurs or Wright’s shifting-balance process combining random genetic drift and selection operates (Wright, 1932). There are reasons for doubting that these two possibilities often can be realised (Charlesworth, 1990b).

A complete mathematical solution to an optimal life-history problem requires enumeration of all the life-histories corresponding to local maxima. This is usually a formidable task. Mathematical methods for determining the phenotype vector \( z \) corresponding to an optimum in general require the methods of dynamic programming or optimal control theory, which are difficult to apply to the problems of high dimensionality usually encountered in life-history theory. Accounts of these methods are provided by León (1976), Schaffer (1983), Sibly and Calow (1986), Mangel and Clark (1988), and Roff (1992, Chapter 8). Charlesworth (1990a) describes a numerical method of locating the optimum \( z \), which involves translating the set of constraints among the components of \( z \) into a \( G \) matrix and then iterating equations (4.75) to equilibrium. This is a relatively painless method of arriving at an optimum. Multiple optima can be searched for by starting with different initial \( z \) values.

In general, however, it has proved most profitable to obtain insights into the outcome of natural selection on life-histories by asking questions
about the ways in which the demographic environment can affect the
effects of selection on aspects of the life-history, without necessarily
solving explicitly for optima. Predictions about the effects of the demo-
graphic environment on the outcome of selection can then be compared
with empirical results. This approach will form much of the subject matter
of the next chapter.
5.1 Introduction

This chapter is concerned with the role of natural selection in moulding the forms of life-histories, as described by the relationships of age-specific survival probability and reproduction to age. This is an aspect of the theory of natural selection which was not investigated by Charles Darwin, although Alfred Russel Wallace briefly discussed the topic (Rose, 1991, pp. 4–5). The first detailed discussion was apparently that of August Weismann (Weismann, 1889; Rose, 1991, pp. 6–7), who was primarily concerned with the evolutionary explanation of ageing. In the absence of a well-developed science of population genetics, these early attempts were necessarily confused, and led nowhere. After the rise of population genetics, Fisher (1930, Chapter 2) and Haldane (1941, pp. 192–4) both briefly discussed aspects of selection on age-specific survival, but their remarks had little immediate impact.

The modern era was initiated by Medawar (1946, 1952), who was the first to formulate clearly the concept that the strength of selection on genes affecting survival or reproduction depends on the ages at which they exert their effects. He suggested that senescence has evolved as a result of stronger selection on effects of genes which are expressed early in life than on effects expressed later on. This idea was elaborated in relation to comparative data by Williams (1957), and made quantitative by Hamilton (1966). It is discussed in section 5.2 of this chapter. A large body of work now exists which was stimulated by these basic ideas and which provides a firm connection between population genetics models of the kinds described in the last two chapters and biological data on ageing (Rose, 1991). This is described in sections 5.3.1 and 5.4.3.

The relation of reproductive activity to age is another aspect of the life-history problem which has been intensively studied in the last twenty years. This involves such questions as the effects of selection on the age at reproductive maturity, selection for iteroparity versus semelparity, the relations between age and reproductive effort, allocation between male and female reproduction (in hermaphrodites), and offspring sex ratio (when there are separate sexes). This topic was initiated by Cole (1954),
who emphasised the selective advantage of early reproduction. The main theoretical ideas are described in sections 5.3.2–5.3.7, and their relationships to data are discussed in sections 5.4.4–5.4.6.

The viewpoint adopted here is that life-history evolution is usually caused by natural selection acting on individual variation within populations. The fact that, as we shall see, the theoretical study of life-history evolution frequently reduces to a discussion of the consequences of changes in demographic parameters for population properties, such as the intrinsic rate of increase or the carrying-capacity, does not mean that group selection (Wynne-Edwards, 1962) is being advocated. The use of these population-level variables is merely an aid to calculation (see section 4.6). While there are circumstances under which group selection can have significant evolutionary effects, there are difficulties in accepting that it can often play an important role in the evolution of characters so closely connected to fitness as life-history traits (Fisher, 1930, Chapter 2; Maynard Smith, 1976; Williams, 1966a; Charlesworth, 1990b).

In addition, while the evolutionary history of a species must impose constraints on the possible range of phenotypes attainable at any one time through mutational variation, it will be assumed that the life-history which characterises a population or species is generally close to a local equilibrium under the forces of selection and mutation, and that there is sufficient genetic variation for the population to track environmental changes which alter the position of this equilibrium. If this view is even approximately correct, it should be fruitful to compare patterns of variation within and between species with the predictions of evolutionary models, without taking evolutionary history into account. Readers can judge for themselves whether the results of this approach are indeed fruitful.

5.2 Effects on fitness of age-specific changes in life-history traits

Chapter 4 laid the foundations for the theory of the effects of natural selection on life-histories, by providing measures of Darwinian fitness in age-structured populations which predict changes in gene frequencies and the means of quantitative traits. With weak selection, and sufficiently small phenotypic variances and covariances in the life-history traits under study, the direction of evolution can be predicted to a good approximation from the relation between the appropriate measure of individual fitness and individual phenotypic value (sections 4.5.3 and 4.6). The requirement of weak selection can be relaxed as far as evolutionary equilibrium is concerned. From now on, we can thus largely avoid gene-frequency calculations, and merely use the effects on fitness of
changes in life-history traits as predictors of the direction of evolutionary change, and of evolutionary equilibria in life-history traits.

An essential first step is to determine the relationships between changes in individual life-history traits and the corresponding changes in fitness. These can be calculated from the partial derivatives of the appropriate fitness measure with respect to age-specific survival or fecundity (Hamilton, 1966; Caswell, 1978). Given these partial derivatives, the effect on fitness of any arbitrary small change in the life-history can be found. Hamilton’s expressions for the partial derivatives are thus of fundamental importance in the theory of life-history evolution.

5.2.1 Density-independent populations

We first consider the case of a density-independent population in a constant environment. In this situation, fitness can be measured by the intrinsic rate of increase, \( r \) (section 4.6.1).

Effects of changes in survival

Using equations (1.58), or applying implicit differentiation to equation (1.32), it is easily seen that the partial derivative of \( r \) with respect to survival at age \( x \) is

\[
\frac{\partial r}{\partial P(x)} = \frac{\sum_{y=x+1}^{d} e^{-r y} k(y)}{P(x) \sum_{y=b}^{d} y e^{-r y} k(y)}
\]

(5.1a)

In many respects, it is more useful to use the derivative of \( r \) with respect to the natural logarithm of \( P(x) \), since genetic effects on survival probabilities are more likely to be additive on a log scale. If this is done, we obtain

\[
\frac{\partial r}{\partial \ln P(x)} = \frac{s(x)}{T}
\]

(5.1b)

where

\[
s(x) = \sum_{y=x+1}^{d} e^{-r y} k(y)
\]

(5.1c)

and \( T \) is the familiar measure of generation time, introduced in section 1.3.2.

\( s(x) \) is a measure of the sensitivity of fitness to a change in log survival at age \( x \), expressed on a time-scale of generations rather than time-intervals. Equations such as (5.1) were first introduced by Hamilton (1966),
and have since been discussed by Demetrius (1969), Emlen (1970), Goodman (1971) and Caswell (1978). It is interesting to note that \( s(x) \) is equal to Fisher's reproductive value at age \( x + 1 \), times \( e^{-r(x+1)}l(x + 1) \), and that \( \partial r / \partial \ln P(x) \) is equal to Goodman's eventual reproductive value at age \( x + 1 \), times \( e^{-r(x+1)}l(x + 1) \).

It is evident from equations (5.1) that \( s(x) \) is a strictly decreasing function of \( x \), for ages within the reproductive period of an iteroparous species, and takes its maximum value of unity throughout the pre-reproductive period. Fitness is therefore most sensitive to age-specific changes in survival which take place early in the life-cycle; after the age of first reproduction is reached, sensitivity declines monotonically with age. As pointed out by Hamilton (1966), this has important implications for the evolution of senescence. These will be discussed in section 5.3.1.

**Effects of changes in fecundity**

A similar calculation can be carried out for the effects of small changes in age-specific fecundity. We obtain

\[
\frac{\partial r}{\partial m(x)} = \frac{e^{-r}l(x)}{T} = \frac{s'(x)}{T}
\]

where \( s'(x) \) is a measure of the sensitivity of fitness to a change in fecundity at age \( x \), expressed on a time-scale of generations. From equation (1.50a), it is evident that \( s'(x) \) is proportional to the stable age distribution of the population. In contrast to \( s(x) \), \( s'(x) \) is a strictly decreasing function of \( x \) if \( r \geq 0 \). If \( r \ll 0 \), however, \( s'(x) \) may have a non-monotonic relation with \( x \), or even be an increasing function of \( x \) (cf. Mertz, 1971a, b). This situation is probably of little biological importance, since a population cannot decline in numbers over the period of time needed for significant evolutionary change without risk of extinction. It seems safe to assume, therefore, that fitness is most sensitive to changes in fecundity early in the life-cycle. Again, equation (5.2) was first obtained by Hamilton (1966).

**5.2.2 Density-dependent populations**

In this case, the appropriate measure of fitness is the net reproduction rate, \( R \), evaluated at the equilibrium density for the population's current demographic parameters (Chapter 4, section 2.2). The sensitivity function for change in log survival at age \( x \) is

\[
s(x, \hat{N}) = \sum_{y=x+1}^{d} k(x, \hat{N})
\]
where $\hat{N}$ is the equilibrium number of individuals in the critical age-group, given by equation (1.91b). The generation time $T$ in the denominator of the equivalent of equation (5.1) is given by $\sum xk(x, \hat{N})$.

Similarly, the sensitivity function for fecundity is

$$s'(x, \hat{N}) = l(x, \hat{N})$$

(5.4)

5.2.3 The probabilities of fixation of genes with age-specific effects

The above results give insight into the effects of selection in terms of the rates of change of gene frequencies or trait means, which are of evident importance as measures of evolution over the short-term. It is, however, of interest to ask a slightly different question: what are the differences in the long-term rates of evolution between classes of genes which differ only with respect to their age of action? In order to do this, we can use the probabilities of fixation of new favorable mutations, since the fixation probability associated with a class of gene with a given effect on fitness provides an index of the rate of substitution of mutations belonging to this class (section 2.3.3; Kimura and Ohta, 1971, p. 12). Fixation probabilities also play a fundamental role in the theory of long-term response to artificial selection (Robertson, 1960; Hill, 1982).

The results of section 4.3.5 can be used to relate fixation probability $U$ to the age of action of the gene concerned, in a very similar way to that used above for the other fitness measures (Charlesworth and Williamson, 1975). As we saw, the survival probability for a new mutation introduced into a stationary, random-mating population is proportional to the amount by which it increases $R$ when heterozygous. If such a mutation also increases survival or fecundity when homozygous, it will spread to fixation if it survives in the population at all. In this case, therefore, the relative fixation probabilities of genes with different age-specific effects will be determined by the fitness sensitivities already introduced, and no new theory is required.

The situation is more complex when the initial population is increasing in numbers. In this case, the sensitivity functions must be replaced by the partial derivatives of $U$ with respect to age-specific survival or fecundity. Expressions for these are derived by Charlesworth and Williamson (1975). Their overall conclusion is that the relations with age of the derivatives of $r$ and $U$ are very similar. This suggests that both long-term and short-term rates of evolution have similar relationships with the ages of action of genes affecting survival and fecundity. An example is shown in Figure 5.1.
5.2.4 Dependence of the fitness sensitivities on the demographic environment

It is clear from the above equations that the sensitivity functions depend on the demographic state of the population, which will be affected by ecological factors such as the level of mortality due to disease or predation. Tests of evolutionary models by means of comparisons of the life-histories of different populations or species often involve comparisons between groups with different ecologies (section 5.4). It is therefore
Effects of age-specific changes on fitness

of great importance to have a clear understanding of how the demographic environment of the population affects the likely outcome of natural selection on life-histories. The relations of the sensitivities to age provide a basis for this.

Density-independent populations

The factor $e^{-ry}$ or $e^{-rx}$ in equations (5.1) and (5.2) provides a greater weighting to younger ages if $r$ is large than when it is small, implying that the sensitivities decline faster with age, the greater the rate of population growth (other things being equal). Similarly, a higher rate of mortality, giving rise to a faster decline of $l(x)$ with age, results in a faster decline in sensitivity with age (only a change in adult mortality is relevant for $s(x)$). Finally, there is a slower rate of decline of $s(x)$ with $x$ if fecundity increases with age. Figure 5.1 illustrates the effect of a higher level of mortality and higher rate of population growth in a high-fecundity human population (Taiwan in 1906), compared with a low-fecundity, low-mortality population (the U.S. population of 1939–41).

These results suggest, at first sight, that some clear predictions about differences in the effects of selection on life-histories in different demographic environments can be made. In particular, environments where populations grow fast, at least temporarily, or where mortality is high, should favour early life-history traits more strongly than environments where population size is stationary and/or mortality is low. The examples of changes in the net fitness effects of some deleterious major mutations in human populations (section 3.4.2) illustrate how demographic changes may alter the outcome of selection on genes with age-specific effects.

Some caution needs to be exercised in applying this conclusion, however, since the detailed nature of the ecological factors causing the demographic differences in question influences their effects on the contributions of different ages to net fitness (section 3.4.1; Abrams, 1993). For instance, it was shown on p. 32 that a change in log survival which is constant across ages (i.e. $P(x)$ for each age is multiplied by the same factor $K$) has no effect on the age distribution, since $r$ is changed by $\ln K$, exactly cancelling the effect of the change in survival (Coale, 1957). Such a change in survival will leave both sensitivity functions unchanged. In contrast, if adult fecundity for all ages is multiplied by the factor $K' > 1$, there will be a corresponding increase in $r$, and a faster decline in the sensitivities with age. An exactly similar effect is produced if survival at age 0 is multiplied by $K'$. In the case of $s(x)$, any change in juvenile survival will affect the sensitivities in the same way as a change in fecundity. This is true only of the adult ages in the case of $s'(x)$. 
Density-dependent populations

In the case of density-dependent populations, the effect of changes in demography induced by density-independent survival or fecundity changes depend both on the nature of the mechanism of density-dependence and on the nature of the density-independent changes. Assume first that density is regulated by a survival factor $P^{(D)}(N)$ which is applied equally to each age-class, so that survival to age $x$ can be written as $l(x, N) = P^{(D)}(N)^x I^{(0)}(x)$ (where $I^{(0)}(x)$ is the density-independent component of $l(x)$). A density-independent change in survival which is also age-independent will be exactly compensated by a change of opposite sign in $P^{(D)}(N)$, such that equation (1.916) is satisfied. There is thus no change in net survival or in the sensitivity functions. On the other hand, if there is an increase in fecundity or in survival at age 0, the compensatory decrease in $P^{(D)}(N)$ results in a faster decline in the sensitivities with age. The same is true of an increase in juvenile survival at any age in the case of $s(x)$.

An alternative mode of density-dependence is through a fecundity factor $m^{(D)}(N)$ applied to each age (density-dependent survival at age 0 has exactly the same consequences). If density-independent changes due to increases in mortality at all age-classes occur, then compensation will occur through an increase in $m^{(D)}(N)$, and the sensitivities will decline faster with age. As before, only changes in adult mortality rates are needed in the case of $s(x)$. Conversely, uniform changes in fecundities at all ages will leave the sensitivities unchanged. If density-dependence operates through juvenile survival, these conclusions still hold for $s(x)$, but apply only to the adult ages for $s'(x)$.

Conclusions

In all cases, demographic shifts which tilt the age-structure of the population in favour of younger individuals, or which result in a more rapid decline in fecundity with age, will cause an increased weighting of the sensitivity of fitness toward changes at earlier ages. Changes which have no such effects will leave the weightings unchanged (Abrams, 1993), in agreement with the findings of section 3.4.1 on the effects of demography on the fitness measure used for calculating equilibrium gene frequencies. In particular, there is no necessary relation between a high intrinsic rate of increase and a greater weighting of younger ages, as is often stated in the literature (e.g. Lewontin, 1965). This implies that comparative studies need to incorporate data on age-structure and the relation of fecundity to age, in order to be truly informative.
5.3 Theories of life-history evolution

5.3.1 The evolution of senescence

General considerations

We are now in a position to see how these results can be applied to an evolutionary interpretation of senescence and related phenomena. In demographic terms, senescence is defined as the tendency for the age-specific survival probabilities and fecundities to decline with age, for individuals of sufficiently advanced age. This process of decline at the level of life-history traits reflects the decline in performance of many different physiological functions with respect to age, and the increase in incidence of pathological factors such as cancer and cardio-vascular disease (Comfort, 1979; Finch, 1990; Rose, 1991). Examples of senescent decline in both fecundity and survival are shown in Tables 1.1 and 1.2.

The problem raised by senescence for evolutionists has been to devise a credible theory to account for the establishment and maintenance by natural selection of this apparently deleterious phenomenon. As mentioned in the introduction to this chapter, Medawar (1946, 1952) suggested that the key to solving this problem was the concept that the strength of selection on a gene, whose effects on survival or fecundity are confined to a given age or set of ages, is dependent on the age or ages in question, such that the intensity of selection is higher, the earlier the time of action of the gene. The intuitive basis for this effect is that unavoidable sources of mortality cause the size of a cohort to dwindle with advancing age, so that a gene with delayed expression will have a smaller net effect on the composition of a population than a gene which is expressed early in life.

Medawar proposed that Fisherian reproductive value should be used as an index of the intensity of selection on genes acting at a given age. This proposal was criticised by Hamilton (1966), who suggested the use of the partial derivatives of $r$ with respect to age-specific survival and fecundity as such indices, based on Fisher's (1930) Malthusian parameter formulae for gene-frequency change. As discussed above, these derivatives or the related measures described in section 5.2, can be given a concrete interpretation in terms of the population genetics of age-structured populations, whereas Medawar's suggestion was based on intuition. It seems clear that reproductive value should not be used in this context.

As Hamilton (1966) pointed out, reproductive value $v(x)$ behaves differently as a function of age from both $s(x)$ and $s'(x)$. In populations with $r \geq 0$, $v(x)$ increases with $x$ for the juvenile age-classes, whereas
s(x) is constant and s'(x) decreases (equations (5.1) and (5.2)). In addition, v(x) generally reaches its maximum value somewhat after the age of first reproduction, as may be seen in Table 1.4, whereas s(x) and s'(x) decline monotonically throughout the reproductive period. Hamilton also pointed out that, in organisms where fecundity increases monotonically with age, v(x) increases with advancing age, so that its use as an index of selection intensity would predict the reverse of senescence. The difference between the two viewpoints is, therefore, not just a technical detail.

Paths for the evolution of senescence

Having decided on the appropriate way of measuring the dependence of selection intensity on age of gene action, we can now examine ways in which this dependence may contribute to the evolution of senescence. It is useful to distinguish three possible ways in which the dependence of the strength of selection on the age of gene action may contribute to the evolution of senescence. For brevity, the discussion will concentrate on senescence in relation to survival, although very similar principles govern the evolution of senescence with respect to reproductive characteristics.

(1) Hamilton (1966) suggested that there would be a higher rate of incorporation of favourable mutations which increase survival at earlier ages within the reproductive period, compared with genes acting later. Over a long period of evolutionary time, this would have the effect of raising survival rates at early adult ages relative to survival rates later in life. An initially non-senescent life-history, with constant adult P(x) values, would thus become converted into one which exhibits senescence. The relation between age of gene action and fixation probability of a favourable mutation, discussed in section 5.2.3, provides an exact foundation for this idea.

It seems unlikely that this mechanism provides a full explanation of senescence, since so many of the physiological changes involved in senescence seem to be pathological (Finch, 1990). It is improbable that such malfunctioning was characteristic of the younger age-groups of the ancestors of present-day species, and has simply been eliminated by the accumulation of favourable gene substitutions. Rather, it seems necessary to invoke processes which involve an accumulation of genes causing harmful effects at later ages, as discussed in the following paragraphs.

(2) It is plausible to imagine the existence of genes with pleiotropic effects on survival, such that increased survival at one age is accompanied by decreased survival at another. The decline in fitness sensitivity with
increasing adult age means that a gene which increases survival early in adult life (or in the juvenile period), at the expense of a reduction in survival of comparable size later in life, is more likely to spread to fixation than a gene with the reverse pattern of effects. The accumulation of such gene substitutions would therefore tend to produce senescence, as emphasised by Williams (1957). Detailed population genetic models of this antagonistic pleiotropy path for the evolution of senescence have been described by Charlesworth (1980, pp. 208–9), Templeton (1980), and Rose (1985). Possible physiological mechanisms for generating the trade-offs between early and late survival rates required for antagonistic pleiotropy have been discussed by Williams (1957) and Kirkwood (1977, 1990). An analysis of the physiological basis of the trade-off between early fecundity versus survival and late fecundity in *Drosophila melanogaster* has been made by Rose *et al.* (1984), Service *et al.* (1985) and Service (1987).

(3) A final contributory factor to the evolution of senescence may be the maintenance of deleterious genes by mutation pressure, the *mutation accumulation* theory (Medawar, 1952; Edney and Gill, 1968). The equilibrium frequency of a deleterious allele maintained in the population by recurrent mutation from wild-type was determined in section 3.3.3 (equations (3.45) and (3.46)). As we have seen (section 4.6), the fitness measure for the mutant genotype which appears in the relevant formulae is directly related to the corresponding $r$ value, so that the same relation holds between the age of action of a mutant gene and its effect on fitness which we discussed in section 5.2 above. In the case of a dominant or semi-dominant mutation with a small effect on survival at a single age $x$, application of the above methods to equation (3.46a) gives the equilibrium frequency of heterozygotes for the mutant as

$$2u/\{[\ln P_{11}(x) - \ln P_{12}(x)]s_{11}(x)\} \quad (5.5a)$$

where $s_{11}(x)$ is the sensitivity function for the population of wild-type homozygotes.

Similarly, for a completely recessive gene, equation (3.46b) gives the frequency of mutant homozygotes as

$$u/\{[\ln P_{11}(x) - \ln P_{22}(x)]s_{11}(x)\} \quad (5.5b)$$

These equations are readily extended to situations where the effects of the mutations extend over more than one age-class, by replacing the denominator by the sum of the relevant terms over all ages involved. They show that deleterious mutant alleles with late ages of action will give rise to higher equilibrium frequencies of affected individuals than genes
with similar effects on survival early in reproductive life. If there is a sufficiently large number of loci capable of mutating to deleterious alleles with age-specific effects, there will therefore be a net decline in survival with advancing adult age. A quantitative genetics model of the consequences for senescence of age-specific mutations at a large number of loci has been analysed by Charlesworth (1990a), who showed that such senescent decline can be produced given reasonable mutation rates and a supply of age-specific mutations (see also Partridge and Barton, 1993).

An idea related to mutation accumulation was proposed by Haldane (1941, pp. 192–4), who suggested that there might be a significant selection pressure in favour of modifier genes which delay the time of onset of hereditary diseases, since this would lower the impact of the diseases on fitness. Medawar (1952) emphasised this process as a major factor in the evolution of senescence. While in principle it is clearly capable of operating, it is open to the same objection which Wright (1929a, b) raised to Fisher’s theory of the evolution of dominance (Fisher, 1928a, b). The selection pressure in favour of a modifier of a rare allele initially at equilibrium under mutation–selection balance must be of the same order as the mutation rate. For example, in the case of a recessive mutation, the equilibrium frequency of mutant homozygotes is given by equation (5.5b), and the selective advantage to a rare, unlinked modifier gene which completely abolishes its effect can be obtained by multiplying by the denominator, giving a net result of \( u \). The advantage to a gene which simply retards the age of onset of the disease is obviously considerably smaller. Such a small selective advantage seems unlikely to be capable of overcoming any deleterious pleiotropic effects of the modifier, or the effects of random genetic drift. It thus seems far more plausible to assume that mutant alleles have a wide distribution of ages of action, and that those whose effects are restricted to later parts of the life-cycle tend to accumulate at the highest frequencies.

5.3.2 The evolution of iteroparity versus semelparity

Another major question in life-history evolution has been the evolution of iteroparity versus semelparity (Cole, 1954; Charnov and Schaffer, 1973; Young, 1990). Other things being equal, it seems from the considerations of section 5.2 that selection on age-specific fecundity would favour reproduction as early as possible in the life-cycle, so that in species with seasonal environments the life-history associated with highest fitness is one in which reproduction occurs as a single burst in the first season following birth. This is, of course, the reproductive pattern characteristic of organisms such as univoltine insects, and is extremely
widespread. One limitation to this mode of reproduction is the necessity of a high level of reproduction in the single breeding season, in order to compensate for the loss of progeny from later seasons. There may well be physiological upper limits to fecundity; in organisms with parental care, there may also be a selective limitation to fecundity due to the need to have a clutch of sufficiently small size to be cared for adequately (Lack, 1954, 1966; Stearns, 1992, Chapter 7).

**Constant density-independent environments**

The strength of this limitation for a population in a constant environment can be assessed as follows, using the method of Charnov and Schaffer (1973). Consider an iteroparous life-history with an age-independent adult survival probability, \( P \), from one age-class to the next. Let the age-specific fecundity, \( m_i \), also be age-independent. Let the age of first reproduction be \( b \), and the probability of survival to this age be \( l(b) \). Then, if there is no upper age limit to survival or reproduction, the characteristic equation for the intrinsic rate of increase, \( r \), can be written as follows (cf. equation (1.32))

\[
1 = l(b)m_i(e^{-rb} + P e^{-r(b+1)} + P^2 e^{-r(b+2)} + \ldots)
\]

Summing this geometric series, we obtain

\[
1 = l(b)m_i e^{-rb}/(1 - P e^{-r}) \quad \text{(5.6a)}
\]

A semelparous life-history with the same age at maturity and same probability of survival to maturity, but with fecundity \( m_s \), has an intrinsic rate of increase, \( r_s \), given by

\[
1 = l(b)m_s e^{-rb} \quad \text{(5.6b)}
\]

For equal fitness of the two life-histories, we require \( r_s = r_i = r \), say; the condition for this is

\[
m_i/m_s = 1 - P e^{-r} \quad \text{(5.7)}
\]

For a semelparous life-history to be at an advantage, \( m_s \) must exceed the value given by this equation. The higher \( P \) and the lower \( r \), the greater the value of \( m_s \) required. If \( P = 0.5 \) and \( r = 0 \), fecundity would have to be doubled for semelparity to do as well as iteroparity; if \( P = 0.9 \) and \( r = 0 \), fecundity would have to be increased tenfold. Iteroparity is therefore favoured by high adult survival and a low rate of population growth (i.e. low fecundity or low juvenile survival); as discussed above in section 5.2.4, these are general conditions which are favourable to the evolution of late life-history traits.
This result is particularly clear when $b = 1$, so that reproduction is initiated after one season (annuality). Equations (5.6) then yield the following expressions

$$e^r = l(1)m_i + P$$  \hspace{1cm} (5.8a)  
$$e^r = l(1)m_s$$  \hspace{1cm} (5.8b)

For equality of the intrinsic rates, we therefore need

$$m_s = m_i + \frac{P}{l(1)}$$  \hspace{1cm} (5.9)

High adult survival relative to juvenile survival is thus particularly favourable to iteroparity in this case (Charnov and Schaffer, 1973). This accords well with the fact that annuality is common among insects of the temperate zone, and in plants associated with unstable, temporary habitats (Stebbins, 1950, Chapter 5). In both cases, the chance that an adult survives from year to year is presumably very low.

**Constant density-dependent environments**

These conclusions can readily be extended to density-dependent populations. The case when density-dependent regulation is exerted through an age-independent survival factor $P^{(D)}(\hat{N})$ (section 5.2.4) yields the following expressions for the equilibrium numbers of individuals in the critical age-group in iteroparous and semelparous populations respectively

$$1 = l^{(i)}(b)P^{(D)}(\hat{N})^b m_i/[1 - P^{(i)}P^{(D)}(\hat{N})]$$  \hspace{1cm} (5.10a)  
$$1 = l^{(i)}(b)P^{(D)}(\hat{N})^b m_s$$  \hspace{1cm} (5.10b)

where $P^{(i)}$ and $l^{(i)}(b)$ are the density-independent components of adult survival probability, and probability of survival to age $b$, respectively.

From the results of section 4.2.2, we can use $\hat{N}$ as a fitness measure, so that the two types of life-history confer equal fitness when $P^{(D)}(\hat{N})$ is the same for both, i.e. when

$$m_i/m_s = 1 - P^{(i)}P^{(D)}(\hat{N})$$  \hspace{1cm} (5.11)

This is obviously analogous to equation (5.7), with the density-dependent component of survival playing the same role as $e^{-r}$ in the density-independent case. Low-fecundity or low-juvenile-survival environments, with correspondingly high density-dependent survival probabilities, will therefore tend to favour iteroparity, as in the density-independent case.

If population regulation operates through a density-dependent compo-
nent of fecundity or juvenile survival, a similar line of reasoning leads to
the condition for equality of life-history type

\[ \frac{m_i^{(I)}}{m_x^{(I)}} = 1 - P^{(I)} \]  \hspace{1cm} (5.12)

where the superscript \((I)\) again indicates the density-independent com-
ponent of the relevant life-history variable. In this case, there is no effect
of juvenile survival or absolute level of fecundity on the condition. Given
the lack of a population growth or density-dependent survival term in
equation (5.12), this type of population regulation is more favorable to
iteroparity than density-independence or regulation through age-
independent survival.

**Temporally varying environments**

The ESS life-history with temporally varying environments cor-
responds to the life-history with maximum mean rate of population
growth, \( \bar{r} \), defined in section 1.4.1, since a rare non-recessive gene can
only invade the population if its \( \bar{r} \) exceeds that of the current population
(section 4.3.3). There is currently no quantitative genetic model for
age-structured populations which permits a justification for the maximisa-
tion of \( \bar{r} \) similar to that for the constant environment case discussed in
section 4.6.3. The ESS approach may be considered as providing neces-
sary but not sufficient conditions for a given life-history to be an evolu-
tionary equilibrium.

Murphy (1968) suggested that temporal variation in juvenile survival
might enable iteroparity to evolve under conditions when the demo-
graphic parameters of the average life-history would favour semelparity.
Consider, for example, the extreme case of an environment in which
there is an occasional season in which juveniles completely fail to survive.
A semelparous life-history would be lethal in this season, whereas an
iteroparous life-history would be able to produce offspring in the next
favourable season, as a result of the survival of the adults. Less extreme
variation can also promote iteroparity, as Murphy showed by computer
calculations of the spread of a gene causing iteroparity (cf. Hairston,

This principle, called *bet-hedging* by Stearns (1976), can be illustrated
with the following example, using the result of section 4.3.3 that, as far as
the spread of a rare gene is concerned, \( \bar{r} \) is equal to the average of the \( r \)
values for each environment, if reproduction starts at age 1 and there is
age-independent adult survival and fecundity.

Suppose that the environment alternates between two states, such that
the probability of survival of individuals to age 1 is \((1 + s)l(1)\) or
(1 - s)l(1), with a mean of l(1). Using equation (5.8a), $\bar{r}$ for an iteroparous population is given by

$$\bar{r}_i = \frac{1}{2} \ln \left[ \frac{l(1)(1 + s)m_i + P}{l(1)(1 - s)m_i + P} \right]$$

$$= \frac{1}{2} \ln \left[ (l(1)^2(1 - s^2)m_i^2 + 2l(1)m_iP + P^2) \right]$$

(5.13a)

Similarly, the mean for a semelparous life-history is given by

$$\bar{r}_s = \frac{1}{2} \ln \left[ l(1)^2(1 - s^2)m_s^2 \right]$$

(5.13b)

A semelparous life-history will therefore be at an advantage if $m_s$ exceeds the value given by the following expression

$$m_s^2 = m_i^2 + \frac{P[2l(1)m_i + P]}{l(1)^2(1 - s^2)}$$

(5.14)

Clearly, the higher $s$, the greater must be the fecundity of the semelparous life-history in order for it to have an advantage. The existence of environmental variation of this kind can thus mean that iteroparity is advantageous, whereas semelparity would be favoured in a constant environment.

A similar result holds if fecundity is variable. On the other hand, if adult survival alternates between $(1 + s)P$ and $(1 - s)P$, the equivalent of equation (5.14) is

$$m_s^2 = m_i^2 + \frac{P[2l(1)m_i + P(1 - s^2)]}{l(1)^2}$$

(5.15)

Fluctuations in adult survival thus tend to favour semelparity.

Bulmer (1985) has studied the effects of density-dependence on the relative advantages of iteroparity and semelparity in this type of model. When there is variation in adult survival, but density is regulated by juvenile survival, the changes in juvenile survival induced by the fluctuations in population numbers caused by the changes in adult survival tend to favour iteroparity, compared with the situation when density-dependence is absent.

Orzack and Tuljapurkar (1989) and Tuljapurkar (1989, 1990a, Chapter 15) have extended the analysis of the effects of temporal variation to more complex life-histories than this, using the theory described in sections 1.4.1 and 4.3.3. Strictly speaking, their theory cannot be applied to the contrast between semelparous and iteroparous life-histories, since semelparity violates the conditions for weak ergodicity needed for the validity of their formulae (section 1.4.1). It is possible, however, to compare two iteroparous life-histories, which both satisfy the weak ergodicity conditions, one of which has reproduction concentrated early in the
Theories of life-history evolution

reproductive period and thus approaches semelparity. This should provide insight into the effects of variable environments on the advantages of semelparity versus iteroparity.

Only the situation when there is no correlation in demographic parameters between successive time-intervals has been studied in detail. Consider the case when survival from age-class 0 to age-class 1 varies. From equation (1.3), the fecundity components of the Leslie matrix are given by \( f(x) = l(1)m(x) \). This means that we need to apply equations (1.78) to all pairs of fecundity components. Writing the covariance between \( f(x) \) and \( f(y) \) as \( \text{cov}(f(x), f(y)) = c^2 \), where \( c \) is the coefficient of variation in \( l(1) \), we have

\[
\bar{r} \approx r - \frac{c^2}{2T^2} \sum_{xy} e^{-r(x+y)} k(x) k(y)
\]

i.e.

\[
\bar{r} \approx r - \frac{c^2}{2T^2} \tag{5.16}
\]

where \( r \) and \( T \) are the intrinsic rate of increase and generation time for the mean \( L \) matrix.

Differences in \( T \) between life-histories may thus reduce or even outweigh differences in \( r \). Since \( T \) is necessarily greater for an iteroparous than for a semelparous life-history with comparable juvenile survival, this suggests that the effect of variation in early survival on the conditions for an advantage to iteroparity is quite general. Of course, this does not mean that such variation will necessarily cause iteroparity to have an advantage over semelparity in every case, but merely that the effect of a difference in \( r \) between a pair of life-histories may be weakened or overcome by a sufficiently large value of \( c \) (Orzack and Tuljapurkar, 1989).

In contrast, variance in adult survival reduces the mean intrinsic rate for iteroparous life-histories, but clearly has no effect on semelparous life-histories, again in agreement with the more specialised model considered above. As discussed by Orzack (1985), lowered sensitivity of life-history traits to environmental variation, so that \( c \) in equation (5.16) is smaller, is also favoured under these conditions.

5.3.3 The evolution of age at maturity

Equation (5.2) shows that, other things being equal, selection favours reproducing early in life, particularly when there is high mortality. Most species defer reproduction to a greater or lesser extent,
Evolution of life-histories

however, so that it is clear that things are usually not equal, and that reproduction early in life must incur a fitness penalty. Several models for such penalties have been proposed (Roff, 1992, Chapter 7; Stearns, 1992, Chapter 6; Charnov, 1993). One of the most widely discussed models assumes that there is a conflict between growth and reproduction, but that the reproductive success of mature individuals increases with their body size, so that early maturity implies lower fecundity (Schaffer, 1974a; Wiley, 1974a, b; Bell, 1976, 1980). This possibility will be considered here, as an illustration of the ways in which the evolution of the age of reproductive maturity can be modelled.

Following Schaffer (1974a), assume that growth is a multiplicative process such that size at age $x$, $G(x)$, is equal to $g(x - 1)G(x - 1)$, where $g(x)$ is the growth-rate at age $x$. The size of a zygote is $G(0)$, so that

$$G(x) = G(0) \prod_{y=1}^{x} g(x - y)$$  \hspace{1cm} (5.17)

This has the same form as $l(x)$ in equation (1.15), and it is convenient to combine growth and survival into one expression by writing

$$\bar{P}(x) = g(x)P(x)$$  \hspace{1cm} (5.18a)

$$\bar{l}(x) = \prod_{y=1}^{x} \bar{P}(x - y)$$  \hspace{1cm} (5.18b)

Finally, we can represent fecundity at age $x$ in terms of fecundity per unit body size, $\bar{m}(x)$, so that

$$m(x) = \bar{m}(x)G(x)$$  \hspace{1cm} (5.18c)

$$k(x) = \bar{l}(x)\bar{m}(x)$$  \hspace{1cm} (5.18d)

In cases when fecundity is not dependent on size, but on some other function of age such as experience in catching prey (Brown, 1987, p. 66), we can use the same type of representation, except that $G(x)$ represents growth in reproductive efficiency rather than size.

**Constant density-independent environments**

We can now apply this model to the problem of selection on the age of maturity. Consider first the case of a semelparous organism in a density-independent environment. The possible life-histories to be compared are characterised by the values of $\bar{l}(x)$ for all possible pre-reproductive ages, and the fecundity per unit weight at age $b$, $\bar{m}_s$. Both
\( \bar{l}(x) \) for \( x \leq b \) and \( \bar{m}_s \) are assumed to be independent of the value of \( b \) for the given life-history, and are determined purely by some specified model for growth, survival and reproduction. The equivalent of equation (5.6b) is thus

\[
1 = \bar{l}(b)\bar{m}_s e^{-r_s b} \tag{5.19a}
\]
i.e.

\[
r_s = \frac{\ln \bar{l}(b) + \ln m_s}{b} \tag{5.19b}
\]

For a value of \( b > 1 \) to be advantageous over \( b = 1 \), \( \ln \bar{l}(x) \) must clearly increase faster with \( x \) than \( 1/x \) declines. The age of maturity which maximises \( r_s \) must be less than that corresponding to the maximum of \( \ln \bar{l}(x) \) and hence \( \bar{l}(x) \). The higher \( \bar{m}_s \), the greater the contribution of the second term on the right-hand side of equation (5.19b), so the harder it is for late maturity to be advantageous.

A necessary condition for selection to favour \( b > 1 \) is clearly that \( \bar{l}(x) \) increases with \( x \) over at least part of its range. But \( \bar{l}(x) \) is composed of two components, one of which, \( l(x) \), decreases with \( x \), as a result of mortality. On general grounds, it is reasonable to assume that the other component, \( G(x) \), initially increases rapidly with age, and then levels off. Delayed maturity can only evolve if the decline in \( l(x) \) is sufficiently slow that it can be outweighed by the increase in \( G(x) \). The position of the maximum in \( \bar{l}(x) \) must be earlier than the point at which growth ceases (Figure 5.2). Hence, growth should never cease before the age of reproductive maturity under this class of model. High mortality or rapid deceleration of growth in the range of ages around potential ages of reproductive maturity favour early maturity; low mortality and late completion of growth favour postponement of reproduction.

Iteroparous life-histories can be analysed in a similar way. Assume for simplicity that growth and survival rates remain constant after reproductive maturity, so that \( \bar{P}(x) = \bar{P} \) for \( x \geq b \). Similarly, age-specific fecundity per unit weight is constant at \( \bar{m}_i \). The equivalent of equation (5.6a) yields the relation

\[
r_i + \frac{\ln (1 - \bar{P} e^{-r_i})}{b} = \frac{\ln \bar{l}(b) + \ln \bar{m}_i}{b} \tag{5.20}
\]

Provided that \( \bar{P} e^{-r_i} < 1 \), as required for convergence of the series equivalent to that of equation (5.6a), the left-hand side is an increasing function of \( b \) as well as \( r_i \). This implies that the conditions for an advantage to delayed reproduction are more stringent than in the case of semelparity.
Figure 5.2. The graphs of $l(x)$ (solid circles), $G(x)$ (open circles) and $\tilde{l}(x)$ (crosses) for an imaginary example. $G(x)$ and $\tilde{l}$ are in arbitrary units.

Constant density-dependent environments

These considerations can easily be extended to density-dependent environments, using the approach of section 5.3.2. The result will only be sketched here (cf. Charlesworth, 1980, pp. 235–7). As before, assume first that density-dependence is mediated through an age-independent survival/growth term $\bar{P}^{(D)}(N)$ which plays a similar role to $e^{-r}$ with no density-dependence. A high value of fecundity per unit weight implies a high equilibrium $N$, so that $\tilde{l}(x)$ increases less rapidly with age, and accentuates the advantage of early reproduction. Density-regulation of this type does not, therefore, especially favour delayed maturity.

In contrast, if fecundity per unit weight or early survival/growth are density-dependent, there is no factor corresponding to $e^{-r}$. From the equivalent of equations (5.6), it can be seen that the optimal value of $b$ in this case corresponds to the value that maximises $\tilde{l}(x)$. This type of density-regulation is thus favourable to delayed maturity.

Extensions of the models, and alternative models

Elaborations of these models, and some alternatives, are reviewed by Roff (1992, Chapter 7) and Stearns (1992, Chapter 6). A few examples which are particularly relevant to the tests of life-history theories described in section 5.4 are noted here.
Roff (1981, 1984, 1991) has proposed specific models of the relations between growth, reproduction and fecundity, applicable to lower vertebrates with indeterminate growth such as fishes, and to holometabolous insects. The fish model will be described here. It assumes that age is a continuous variable, and that growth in length follows the von Bertalanffy equation, such that body length at age \( x \) is given by

\[
L(x) = L(\infty)(1 - e^{-\kappa x})
\]  

(5.21a)

where \( L(\infty) \) is the length on completion of growth, and \( \kappa \) is the rate of growth of length. Body weight at age \( x \), \( G(x) \), is given by the cube of \( L(x) \), and fecundity per unit weight is assumed to be independent of age. If there is a density-dependent component of survival very early in life, and thereafter a constant mortality rate \( \mu \), we find that

\[
\hat{L}(x) \propto e^{-\mu x} L(\infty)^3 (1 - e^{-\kappa x})^3
\]

(5.21b)

As noted above, with this model of density-dependence the optimum \( b \) corresponds to the age which maximises \( \hat{L}(x) \). Differentiating equation (5.21b) with respect to \( x \), and setting the derivative equal to zero, yields the following expression for the optimal age at maturity

\[
b = \frac{1}{\kappa} \ln \left( \frac{3\kappa}{\mu} + 1 \right)
\]  

(5.22)

As expected from the general considerations discussed above, this model predicts that increased growth-rate and higher mortality both favour earlier maturity.

Stearns and Crandall (1981) and Stearns and Koella (1986) have developed somewhat different models, several of which assume that the survival probability of juveniles is a decreasing function of the age of maturity. As noted by Roff (1992, pp. 202-4), the mathematical predictions of these models about the optimal age at maturity are similar in general form both to each other and to those of his models.

Charnov (1990, 1993) and Charnov and Berrigan (1991a, b) have developed a theory of the evolution of age at maturity in stationary populations, by assuming that the action of selection on \( b \) can be described purely in terms of the effect of age on the chance of survival to maturity, and on the subsequent expectation of offspring (i.e. reproductive value at maturity, given by equation (1.61) with \( r = 0 \)). They further assume that the age-specific mortality rate over the range of possible ages at maturity is a constant, \( \mu \). Under the assumption of density-dependence through early survival (or adult fecundity), fitness can be measured by the density-independent component of the net reproduction rate. For a
Evolution of life-histories

possible optimum age at maturity \( x \), fitness is thus proportional to

\[
e^{-\mu x} V(x) \tag{5.23a}
\]

where \( V(x) \) is the reproductive value at age \( x \). Differentiating with respect to \( x \), and equating the derivative to zero, the condition for an age at maturity \( b \) to be optimal is

\[
\frac{\partial \ln V}{\partial x} \bigg|_{x=b} = \mu \tag{5.23b}
\]

The further assumption is made that the functional relation of \( V(x) \) to \( x \) for possible ages at maturity takes a simple form. For example, assume that it is a power function, \( V(x) \propto x^d \). Then we have

\[
\frac{\partial \ln V}{\partial x} \bigg|_{x=b} = \frac{d}{b} = \mu \tag{5.23c}
\]

This implies that the product \( b\mu \) for an optimal life-history is a dimensionless constant, \( d \), so that higher mortality favours earlier reproduction, in line with earlier conclusions.

Charnov and Berrigan (1991a), Shine and Charnov (1992) and Charnov (1993, Chapter 4) discuss how formulae relating \( V(x) \) to \( x \) for possible ages at maturity can be justified for species with indeterminate growth; Charnov and Berrigan (1991b) and Charnov (1993, Chapter 5) discuss justifications for species such as mammals with determinate growth. Their models will be considered in section 5.4.4, in relation to tests of life-history theory.

Temporally varying environments

The effects of temporal fluctuations on selection for the age of maturity of a semelparous species have been discussed by Klinkhammer and de Jong (1983), Roerdink (1988, 1989) and Tuljapurkar (1990a, Chapter 16). Their analyses indicate that variation in fecundity tends to favour delayed maturity, although the effect is not large unless the coefficient of variation is 0.5 or more. For iteroparous life-histories with variable fecundity or juvenile survival/growth, and no autocorrelation, equation (5.16) yields a similar result, since the generation time \( T \) must increase with \( b \). With a long environmental period, approximation of the mean of the left-hand side of equation (5.20) by the first term of a Taylor expansion around the parameter values for the mean life-history yields the difference between the mean intrinsic rate and the intrinsic rate of the mean life-history as
Theories of life-history evolution

\[
\bar{r}_i - r_i \equiv \frac{(1 - \tilde{P} e^{-r_i}) E \{\ln \tilde{l}(b) + \tilde{m}_i\}}{[b - (b - 1)\tilde{P} e^{-r_i}]}
\equiv \frac{(1 - \tilde{P} e^{-r_i}) c^2}{2[b - (b - 1)\tilde{P} e^{-r_i}]} \quad (5.24)
\]

If \(\tilde{P} e^{-r_i} < 1\), as is required for convergence of the series yielding equation (5.20), this implies that a higher age at maturity reduces the effect of environmental variability on the intrinsic rate of increase, and hence is more likely to be favoured in a variable environment.

A similar approximation shows that variable adult survival/growth has no major effect on the intrinsic rate of increase when there is a long environmental period. The situation is not immediately clear if there is no autocorrelation, since the numerator of the equivalent of equation (5.16) may be increased as well as the denominator (see equations (1.78)). The case when log survival probabilities at each adult age at a given time are affected by the same random deviation, with variance \(c^2\) for each age, can be examined as follows. Noting that the covariances between log survival probabilities at different ages are also equal to \(c^2\), equations (1.78b) and (5.1b) yield the relation

\[
\bar{r}_i \equiv r_i - \frac{\left[ c \sum_{x=b} s(x) \right]^2}{2T^2} \quad (5.25a)
\]

From equations (1.47c) and (5.1b), we obtain the relation

\[
T = b + \sum_{x=b} s(x) \quad (5.26)
\]

so that equation (5.25a) can be written as

\[
\bar{r}_i \equiv r_i - \frac{[c(T - b)]^2}{2T^2} \quad (5.25b)
\]

In the case considered above, when adult survival/growth and fecundity per unit weight are independent of age, the use of the properties of the geometric series and equation (5.6a) enables us to write

\[
T = b + \tilde{l}(b)\tilde{m}_i e^{-rb} \sum_{j=1} e^{-\eta j} \tilde{P}^j
\]

Hence,

\[
\bar{r}_i \equiv r_i - \frac{1}{2} \left( \frac{c}{[b(\tilde{P}^{-1} e^{r_i} - 1) + 1]} \right)^2 \quad (5.25c)
\]
If two life-histories with the same \( r_i \) and \( \tilde{P} \) values are compared, the effect of variable adult survival is to confer an advantage on the life-history with the later age of maturity, provided that \( \tilde{P} e^{-r_i} < 1 \). The magnitude of the advantage is diminished if \( r_i \) is large, or \( \tilde{P} \) is small, as might be expected on general grounds. An increased \( \tilde{P} \) value can, however, outweigh the effect of later age of maturity in the second term of equation (5.25c).

Spatially varying environments

Discussion of the consequences of spatially variable environments for the evolution of life-histories was initiated by Stearns and Koella (1986), in the context of selection for phenotypic plasticity. Following the models of section 4.3.4, it is assumed that the environment is divided into different types of spatial patch, that an individual spends the selectively relevant portion of its life in the patch in which it is born, and that the nature of the patch affects certain aspects of the life-history, such as the rate of growth in body size. It is further assumed that a single genotype can respond to the patch type, and adjust the life-history traits of its carriers (e.g. age at maturity) to match the patch type. If, for example, growth is slow in a given patch, because of scarcity of resources, it might pay to delay reproductive maturity (see the discussions above). The problem is thus to find the ESS distribution of ages at maturity which match the distribution of patch types encountered by the population.

Stearns and Koella (1986) assumed that the ESS can be found by determining the life-history for each patch type which maximises the intrinsic rate of increase within each patch. While this would be appropriate for a model of phenotypic plasticity in a temporally varying environment with a long period, where the ESS would be given by the maximal mean intrinsic rate of increase over all environments, it lacks a population genetic justification in the context of spatial variation (Houston and McNamara, 1992; Kawecki and Stearns, 1993). For the model of hard selection described in section 4.3.4, it is evident that the ESS must be such that the quantity

\[
\sum_x e^{-\hat{r}_x} k(m, x) = 1
\]

is to be maximised for each patch type \( m \), subject to the constraint that the intrinsic rate for the whole population, \( \hat{r} \), satisfies the characteristic equation

\[
\sum_x e^{-\hat{r}_x} \sum_m c_m k(m, x) = 1
\]
As discussed by Kawecki and Stearns (1993), use of these criteria can lead to predictions about the optimal adjustment of life-history traits to local conditions which differ significantly from those of Stearns and Koella (1986). Details of the nature of the predictions concerning the optimal response of age at maturity to varying growth and mortality regimes are given by Kawecki and Stearns (1993). As might be expected, earlier maturity is favoured in patches with higher mortality rates at the potential ages of maturity; later reproduction is favoured in patches where higher fecundity is attained relatively late in life (e.g. due to slower growth).

Analogous criteria can be established for the case of density-dependent populations, where the equilibrium density for the population as a whole is used in the reproductive functions for each patch, and the net reproduction rate \( R \) for each patch \( m \) is to be maximised.

While these criteria apply to a population under hard selection, somewhat different criteria will apply to the case of soft selection, where equation (4.44b) implies that the ESS is to be found by maximising the following quantity for each value of \( m \)

\[
\sum_x e^{-r_x} \frac{k(m, x)}{\left( \sum_x e^{-r_x} k(m, x) \right)} \quad (5.27c)
\]

subject to equation (5.27b), and holding the reproductive function in the denominator constant when performing the maximisation. An analogous condition can be written with density-dependence.

### 5.3.4 Reproductive effort

Much work in life-history theory has been devoted to the 'reproductive effort model', originally suggested in verbal form by Williams (1966a, b), and first studied quantitatively by Gadgil and Bossert (1970). Later investigations include those of Fagen (1972), Schaffer (1974a, b), Schaffer and Gadgil (1975), Charlesworth and Léon (1976), León (1976), Ricklefs (1977), Schaffer and Rozenzweig (1977), Michod (1979), Law (1979b) and Charlesworth (1990a). The basic idea is that, at any age \( x \) an individual can allocate a fraction \( E_x (0 \leq E_x \leq 1) \) of its available energy resources to reproduction, and the remainder to maintenance (survival) and growth. \( E_x \) is called the reproductive effort at age \( x \). Using the model developed on p. 206, we can represent the fecundity per unit weight at age \( x \) as a strictly increasing function of \( E_x \), \( \bar{m}_x(E_x) \); the product of growth and survival at age \( x \) is a strictly decreasing function of \( E_x \), \( \bar{P}_x(E_x) \).
Evolution of life-histories
(Schaffer, 1974a). (The subscript \( x \) attached to \( \bar{m} \) and \( \bar{P} \) is intended to emphasise that, in general, the nature of the functional dependence of these variables on reproductive effort will vary with age.)

The optimisation problem in this case is thus to determine the vectors of values of \( E_x \) which yield maxima in \( r \), in the density-independent case, or \( \hat{N} \), in the density-dependent case. Since fecundity per unit weight is assumed to be an increasing function of effort, it is convenient in practice to use it directly as an independent variable, \( m_x \), and to write \( \bar{P}_x \) as a function of \( \bar{m}_x \), \( \bar{P}_x(\bar{m}_x) \). When reproductive effort at age \( x \) is unity, \( \bar{m}_x \) takes a maximal value \( M_x \), such that \( \bar{P}_x(M_x) = 0 \). When reproductive effort at age \( x \) is zero, \( \bar{P}_x \) takes its maximal value \( \bar{P}_x(0) \). Various forms of functional relationships between \( P_x \) and \( m_x \) can be imagined (Figure 5.3); most of the results discussed below are based on the simplest (concave and convex) forms. Finally, it is important to note that parental care may form an important component of reproductive effort in some species; since we imagine that the survival of the young is an increasing function of the amount of parental care, this factor can be incorporated into the models by counting offspring at the age at which parental care ceases, so that \( \bar{m}_x \) includes this survival probability.

The problem of solving for the optimal life-history with this model is a formidable one, and no general analytic solutions have been found. Schaffer (1974a) and Schaffer and Rosenzweig (1977) have studied some special cases, and numerical solutions have been obtained by Gadgil and

\[ \text{Figure 5.3. Some possible functional relationships between } \bar{P}_x \text{ and } \bar{m}_x: \text{ concave (full line), convex (dashed line) and concavo-convex (dotted line). } \bar{P}_x \text{ and } \bar{m}_x \text{ are in arbitrary units.} \]
Bossert (1970), Fagen (1972) and Charlesworth (1990a). Insights into the properties of optimal life-histories can, however, be gained without knowing the full solutions, and these will be discussed below.

We first examine the conditions for \( r \) to be at a maximum with this model. Consider those age-classes for which reproductive effort is intermediate \((0 < \bar{m}_x < M_x)\). In order for \( r \) to be at a local maximum under such conditions, we need \( \partial r / \partial \bar{m}_x = 0 \) for all such age-classes. Using the derivatives of \( r \) given in equations (5.1) and (5.2), we have

\[
T \frac{\partial r}{\partial \bar{m}_x} = e^{-r x} \bar{l}(x) \left[ 1 + \frac{\partial \bar{P}}{\partial \bar{m}_x} e^{-r} \bar{v}(x + 1) \right]
\]

The quantity \( \bar{v}(x + 1) \) is Schaffer's *modified reproductive value* for age \( x + 1 \), which is analogous to Fisher's reproductive value (equations (1.54)), but takes growth as well as survival into account. We have

\[
\bar{v}(x) = \frac{e^{r x}}{\bar{l}(x)} \sum_{y=x}^{d} e^{-r y} \bar{l}(y) \bar{m}(y)
\]

A necessary condition for a maximum in \( r \) is, therefore, that for age-classes with intermediate levels of reproductive effort, we must have

\[
1 + \frac{\partial \bar{P}}{\partial \bar{m}_x} e^{-r} \bar{v}(x + 1) = 0
\]

The second-order derivatives of \( r \) for all pairs of age-classes \( x \) and \( y \) such that \( \partial r / \partial \bar{m}_x = \partial r / \partial \bar{m}_y = 0 \) are given by

\[
T \frac{\partial^2 r}{\partial \bar{m}_x^2} = \frac{\partial^2 \bar{P}}{\partial \bar{m}_x^2} e^{-r(x+1)} \bar{l}(x) \bar{v}(x + 1)
\]

\[
\frac{\partial^2 r}{\partial \bar{m}_x \partial \bar{m}_y} = 0
\]

Using the standard analytical conditions for a maximum of a function of several variables, the necessary and sufficient conditions for \( r \) to be maximised, with respect to perturbations in the age-classes with intermediate effort, are thus given by equation (5.30) and the following inequality

\[\dagger\] An alternative procedure is used by Schaffer (1974a, *et seq.*) He gives the first age-class the index 0 instead of 1, and works with \( B_x = \bar{P}(0) \bar{m}_x \) as the independent variable, instead of \( \bar{m}_x \). On this scale, \( \bar{v}_x \) is defined as \( e^{-r} \bar{P}(0) \) times the value of \( \bar{v}_x \) on the present scale (see the footnote to p. 23). If \( r \) is differentiated with respect to \( B_x \), we obtain the equivalent of equation (5.42) as \( 1 + (\partial \bar{P}/\partial B_x) \bar{v}_{x+1} = 0 \). The two notations yield identical conclusions.
Evolution of life-histories

\[ \frac{\partial^2 \tilde{P}_x}{\partial m_x^2} < 0 \quad (5.31c) \]

The necessary and sufficient conditions for \( r \) to be at a maximum with respect to perturbations at age-classes with \( \tilde{m}_x = 0 \) or \( \tilde{m}_x = M_x \) (i.e. boundary values of \( \tilde{m}_x \)) are

\[ \frac{\partial r}{\partial \tilde{m}_x} < 0 \quad (\tilde{m}_x = 0) \quad (5.32a) \]
\[ \frac{\partial r}{\partial \tilde{m}_x} > 0 \quad (\tilde{m}_x = M_x) \quad (5.32b) \]

In general, of course, perturbations to the life-history need not be confined to sets of age-classes within one of these three divisions. For small perturbations, it is easy to see that, if the above conditions are met, any change in \( r \) will be dominated by terms in which \( m_x \) is on a boundary, since the total differential of \( r, \Delta r \), is zero with respect to perturbations in the other age-classes. Conditions \( 5.30, 5.31 \) and \( 5.32 \) are thus necessary and sufficient for \( r \) to be at a local maximum.

Similar conditions can be derived for the case of a density-dependent population, using the method of p. 187. We obtain the equivalents of expressions \( 5.28, 5.30 \) and \( 5.31 \), for example, as

\[ \frac{\partial w}{\partial \tilde{m}_x} = \tilde{l}(x) \left[ 1 + \frac{\partial \tilde{P}_x}{\partial \tilde{m}_x} \tilde{p}(x + 1) \right] = 0 \quad (5.33a) \]
\[ \frac{\partial^2 w}{\partial \tilde{m}_x^2} = \frac{\partial^2 \tilde{P}_x}{\partial \tilde{m}_x^2} \tilde{l}(x) \tilde{p}(x + 1) < 0 \quad (5.33b) \]

In carrying out these differentiations, it is important to note that the demographic parameters in these equations are held constant at the values for population density equal to the carrying-capacity given by equation \( 1.91b \) (section 4.6).

Properties of optimum life-histories with the reproductive effort model: semelparity

Some useful results can be obtained merely by inspection of the above formulae. We may note first of all that a necessary condition for inequality \( 5.31c \) to be satisfied is that \( \tilde{P}_x \) be a concave function of \( \tilde{m}_x \) for at least part of its range (cf. Gadgil and Bossert, 1970; Schaffer, 1974a). If \( \tilde{P}_x \) is a convex function of \( \tilde{m}_x \), \( \frac{\partial^2 \tilde{P}_x}{\partial \tilde{m}_x^2} \) is necessarily positive throughout its range, so that there can be no age-classes with intermediate levels of reproductive effort; the life-history in such a case must be semelparous, with a number of non-reproductive age-classes terminating
in an age-class with maximal reproductive effort. This, as we have discussed earlier, is characteristic of a number of groups, and it is tempting to speculate that such semelparous life-histories are associated with convex relationships between $\bar{P}_x$ and $\bar{m}_x$.

There is, unfortunately, little direct evidence on this point. Gadgil and Bossert (1970) suggested that semelparous life-histories characteristic of Pacific salmon involve convexity; these species undertake long migrations from the ocean to rivers in order to breed, so that much energy has to be expended, and exposure to considerable risks undertaken, in order for the salmon to breed at all. This means that the curve relating $\bar{m}_x$ to $E_x$ rises very slowly at first, and then increases rapidly, whereas $\bar{P}_x$ is reduced considerably by a small increase in $E_x$. This is likely to generate a convex relationship between $\bar{P}_x$ and $\bar{m}_x$. Similarly, Schaffer and Gadgil (1975) proposed that the contrast between the semelparous agaves and the phenotypically similar but iteroparous yuccas may be traceable to greater competition for pollinators in the former, the insects preferring large inflorescences which require large stalks. Evidence for such preferences is presented by Schaffer and Schaffer (1977, 1979). Since a large expenditure of energy is probably needed to develop an inflorescence stalk, which is required for even a low level of reproduction, a convex relationship between $\bar{P}_x$ and $\bar{m}_x$ may be generated. This is particularly likely in a competitive situation, since plants with relatively small stalks may have difficulty in attracting pollinators if others with larger stalks are present, despite a significant expenditure of effort. There is evidence from a number of different plant species that larger inflorescences attract more pollinators, and enjoy greater reproductive success (Willson, 1983; Broyles and Wyatt, 1990).

Properties of optimum life-histories with the reproductive effort model: iteroparous life-histories

For studying iteroparous life-histories in which there is a number of age-classes with intermediate reproductive effort, it is useful to restrict attention to the case when the $\bar{P}_x$ have the same, concave dependence on the $\bar{m}_x$ for each age-class with intermediate reproductive effort. $\bar{P}_x(\bar{m}_x)$ can therefore be written simply as $\bar{P}(\bar{m}_x)$. The first result that we can derive in this case is that there must be an inverse relationship between $\bar{m}_x$ and $\bar{v}(x + 1)$ for all ages with intermediate reproductive effort. By the hypothesis of concavity, $|\partial \bar{P}/\partial \bar{m}_x|$ is an increasing function of $\bar{m}_x$. Equation (5.30) thus implies that, if we compare two ages $x$ and $y$, $\bar{m}_x > \bar{m}_y$ if and only if $\bar{v}(x + 1) < \bar{v}(y + 1)$. A relationship of this sort, but between Fisherian reproductive value and reproductive effort, was
Evolution of life-histories first proposed by Williams (1966b) on intuitive grounds. He suggested that what he called residual reproductive value at age \( x \), defined as the component of reproductive value attributable to future reproduction, should be inversely related to reproductive effort at age \( x \). From equation (5.29) it is evident that we have

\[
\hat{v}(x) = \hat{m}(x) + e^{-r} \hat{P}_x \hat{v}(x + 1)
\]  

(5.34)

Residual reproductive value in this formulation is thus equal to \( e^{-r} \hat{P}_x \hat{v}(x + 1) \), so that Williams’ suggestion and the use of it by Pianka and Parker (1975) are not quantitatively correct.

Another question which can be answered without knowing the full details of the optimal life-history solutions concerns the relations between reproductive effort and age. It was suggested by Williams (1966a, p. 182) and by Gadgil and Bossert (1970) that reproductive effort should increase with age, since the cost of present reproduction to future fitness depends on the number of future reproductive age-classes, which clearly decreases with increasing age. Gadgil and Bossert constructed some computer examples of optimal life-histories which always showed an increase in \( E_x \) with \( x \) for the ages with intermediate effort. Some counter-examples were, however, produced by Fagen (1972). An algebraic treatment was given by Charlesworth and León (1976), who found that reproductive effort may increase with age, but need not necessarily do so (see also Charlesworth (1990a)).

With density-independence, they found that constancy or a decrease in reproductive effort with age is favoured when \( r \) is high, survival and growth do not fall off too fast with increasing reproductive effort, and the maximal fecundity per unit weight remains low. These are circumstances in which the considerations discussed in section 5.2.4 would lead us to expect a premium on reproduction early in life. The converse conditions favour an increase in reproductive effort with age.

It is of some interest to determine the conditions under which reproductive effort remains constant for each adult age-class in an optimal life-history, particularly as several analyses have been carried out on the assumption that this is so (Schaffer, 1974a, b; Ricklefs, 1977). In such a case, we can write \( \bar{m}_x = \bar{m} \) and \( \hat{P}_x(\bar{m}_x) = \hat{P} \) for each adult age-class in the optimal life-history. If we assume that there is no upper limit to the age to which individuals can survive and reproduce (equation (5.6a)) and substitute the above values for \( m_x \) and \( \hat{P}_x(\bar{m}_x) \) into equation (5.29) we find that

\[
\hat{v}(x) = e^{rx}/\bar{l}(x) \quad (x < b)
\]  

(5.35a)
Equation (5.30) thus gives us the condition on the adult age-classes
\[
\left| \frac{\partial \bar{P}}{\partial \bar{m}} \right| = \frac{e^r - \bar{P}}{\bar{m}}
\]  
(5.36a)

Using equation (5.6a) this can be expressed as
\[
\left| \frac{\partial \bar{P}}{\partial \bar{m}} \right| = e^{-r(b-1)} \tilde{l}(b)
\]  
(5.36b)

The above results on the relations between reproductive effort and age can easily be extended to density-dependent populations, using equations (5.33). As before, the consequences of density-dependence vary according to the model assumed. With regulation through an age-independent component of survival/growth, the relevant term cancels the effect of the intrinsic rate (cf. equations (5.10)), so there is no reason for the absence of population growth to make it any easier for reproductive effort to increase with age. In contrast, regulation through density-dependent juvenile survival or adult fecundity favours an increase in reproductive effort with age.

Optimal life-histories in which reproductive effort increases so slowly with age that it appears constant for all practical purposes were found to be extremely common in numerical studies by Charlesworth (1990a). But if recurrent mutations at many loci with age-specific effects are added to the model, some of which reduce both survival/growth and fecundity at a given age as a result of their reducing the individual’s ability to harness resources, while others affect only the partitioning of resources between survival/growth and reproduction, there is a much sharper increase in reproductive effort and fecundity with age. There is a collapse in both survival and fecundity at the end of life. This is due to the fact that the decline in fitness sensitivities with age eventually leads to a situation in which mutation pressure overpowers selection. The mutational collapse occurs sooner, the higher the rate of mutation to alleles with deleterious effects on both survival and fecundity. This suggests that some degree of increase in reproductive effort with age might be expected under many ecological circumstances, although the abrupt terminal decline in fecundity predicted by this model does not seem very realistic. The properties of more general models of this kind, including deleterious mutations with effects on more than one age-class need to be investigated, since it is possible that there would be a less abrupt change in fecundity at the end of life.
Age at maturity under the reproductive effort model

It should be noted that the age of reproductive maturity $b$ is itself a component of the optimal life-history solution. The constraints imposed by the relation between survival/growth and reproductive effort affect the evolution of the age of maturity in a way distinct from that discussed in section 5.3.3, where reproduction was assumed to have no direct effect on growth or survival. In order for a life-history to be optimal under the reproductive effort model, equations (5.32a) and (5.35a) imply that

$$\frac{\partial \tilde{P}_x}{\partial \tilde{m}_x} > e^{-r \tilde{I}} (x + 1) \quad (1 \leq x < b, \tilde{m}_x = 0) \quad (5.37)$$

Joint satisfaction of this equation and equation (5.36b), for the case when reproductive effort remains constant with age, requires $\tilde{P}_x$ for age $b - 1$ to be more sensitive to increases in $\tilde{m}_x$ than $\tilde{P}_x$ for adults.

Effects of the demographic environment on reproductive effort–constant environments

It has been suggested (Gadgil and Bossert, 1970; Schaffer 1974a) that the level of reproductive effort at each age, in an optimum life-history, will depend on the level of extrinsic causes of mortality, such as predation or density-dependent factors. Thus, if adult mortality is high, the optimum life-history should have a higher level of reproductive effort at each age than if adult mortality is low. Conversely, a low level of reproductive success or juvenile survival will tend to favour a low level of reproductive effort. Analytical results concerning this question are hard to obtain because of the difficulty of solving for the optimal life-history. Schaffer’s treatment assumes that the optimal life-history is such that reproductive effort for the adults is independent of age, and that reproductive maturity occurs at age $b = 1$. Equation (5.36b) for the optimal life-history in a density-independent environment thus becomes

$$\frac{\partial \tilde{P}}{\partial \tilde{m}} = \tilde{l}(1) \quad (5.38a)$$

Assume that $\tilde{P}(\tilde{m})$ is changed by an increase in some extrinsic source of mortality to $P'\tilde{P}(\tilde{m})$, where $P'$ is less than one and independent of reproductive effort. The equivalent of equation (5.38a) for the optimal life-history under the new regime of adult mortality is

$$P'\left|\frac{\partial \tilde{P}}{\partial \tilde{m}}\right| = \tilde{l}(1) \quad (5.38b)$$

If $\tilde{P}$ is a concave function of $\tilde{m}$, satisfaction of this relation requires that
Theories of life-history evolution

The new optimal life history have a higher value of $\bar{m}$ than the old, in accordance with the idea that higher adult mortality should lead to higher reproductive effort. On the other hand, if $\tilde{l}(1)$ or $\bar{m}$ are changed to new, lower values in a similar fashion, it is easily seen that the new optimal life-history should have lower reproductive effort.

We have seen above that reproductive effort is not always independent of age in the optimal life-history, so that these results must be treated with some caution as a generalisation. Some progress in analysing the problem by a more general approach has been made by Michod (1979). Using partial differentiation, he was able to obtain expressions for the changes in the $\bar{m}_x$ values for the optimal life-history which are induced by a small change in the probability of survival at an arbitrary age $y$. More explicitly, if $\tilde{P}_y(\bar{m}_y)$ is changed to a new value $P'_y \tilde{P}_y(\bar{m}_y)$, as the result of a change in some extrinsic cause of mortality, the new solution for the optimal life-history differs from the old in a way which is calculable from Michod's formulae, provided that $P'_y$ is close to unity.

The exact consequences of a change in survival at one age-class depend on whether or not there is density dependence and, if there is, on its exact form. In density-independent populations, an increase in mortality at age $y$ will generate a new optimal life-history which has greater reproductive effort at each age up to and including $y$, but lower reproductive effort following age $y$. Thus, an increase in mortality among the juveniles will lead to a decrease in reproductive effort among the adults in this case, and a new source of mortality late in life will lead to an increase in reproductive effort at earlier age-classes. With density-dependence, it is still true that an increase in mortality at age $y$ leads to an increase in reproductive effort at all reproductive ages $x \leq y$; the consequences for ages $x > y$ depend, however, on the specific model of density dependence assumed. If density-dependence operates in such a way that $\bar{m}_x$ and $\tilde{P}_x$ are insensitive to density for $x > u$ (where $u$ is some fixed age), then there is no effect on reproductive effort at ages later than $u$. Thus, a change in mortality among the juvenile age-classes when there is density-dependent juvenile survival has no effect on the reproductive effort of the adult age-classes, in contrast to a reduction of reproductive effort in a density-independent population.

The effects of an increase in mortality in several age-classes can be complex. It is still true, however, that all reproductive ages prior to any ages at which mortality is increased will increase their reproductive effort. In a density-independent population, there will also be a reduction in reproductive effort at each age greater than the last age to be affected by a mortality change. For other ages, the nature of the changes is difficult
to determine. In a population regulated by density-dependent juvenile survival, we would expect an increase in reproductive effort at each reproductive age up to and including the last reproductive age-class at which there is an increase in mortality, and no change thereafter. Finally, the effects of an increase in mortality applied equally to each age-class also depend on the mechanism of density-dependence. In a density-independent population, such a change is exactly compensated for by a reduction in \( r \) (p. 32), so that there is no change in the characteristic equation for the optimal life-history, and hence no change in the distribution of reproductive effort (Gadgil and Bossert, 1970). This is also true in a population regulated by an age-independent, density-dependent component of survival, but in a population regulated by density-dependent juvenile survival, an increase in reproductive effort of the adults is to be expected as a result of this type of mortality change (cf. section 5.2.4).

**Effects of the demographic environment on reproductive effort—temporally varying environments**

Schaffer (1974b) initiated the study of the effects of temporally varying environments on optimal reproductive effort. The results of his approach can be sketched as follows, using the assumption that reproduction starts at age 1 and that reproductive effort is constant over age in an iteroparous life-history with no upper limit to age. Under these conditions, fitness in a density-independent population can be measured by the average of the \( r \) values for each environment (section 4.3.3). Assume, as in section 5.3.2, that the probability of survival to age 1 varies between \((1 + s)l(1)\) and \((1 - s)l(1)\). Using equation (5.13a), and assuming that the reproductive effort is equal to the optimal value for a constant environment given by equation (5.36a), we have

\[
\frac{\partial \bar{r}}{\partial \bar{m}} = \frac{l(1)s}{2} \left\{ \frac{1}{[\bar{l}(1)(1 + s)\bar{m} + \bar{P}]} - \frac{1}{[\bar{l}(1)(1 - s)\bar{m} + \bar{P}]} \right\}
\]

(5.39)

This implies that the optimal solution for a constant environment yields a negative derivative of fitness with respect to \( \bar{m} \); hence, the optimum with a varying environment must be at a lower \( \bar{m} \) value. A similar result holds when fecundity varies, but the opposite is true when adult survival varies, since there is an additional factor of \( \partial \bar{P}/\partial \bar{m} \) (which is negative) outside the bracketed term in equation (5.39).

These results can easily be extended to more general frequency distributions of survival/growth rates. Thus, fluctuations in juvenile survival/growth rates or in fecundity lead to an optimal life-history which allocates
less to reproductive effort than that for the mean life-history over the set of environments. The converse is true if adult survival varies. This suggests that long-period environmental variability in juvenile life-history traits or fecundity will generally cause the evolution of lower allocation to reproductive effort than expected for a constant environment, whereas variation in adult survival/growth has the opposite effect. This parallels the earlier results for iteroparity versus semelparity, and for age at maturity.

Similarly, if there is no autocorrelation, equation (5.16) for a general iteroparous life-history implies that larger $T$ is favoured by variable fecundity or juvenile survival/growth. Since this requires higher adult survival/growth, lower reproductive effort than in the constant environment case is favoured. Conversely, from equation (5.25c), fluctuating adult survival or growth favours reduced adult survival/growth, and hence higher reproductive effort. These conclusions may need modification when there is density-dependence in early survival, since fluctuating adult survival will then induce compensating variation in early survival, which will tend to damp out the effects of varying adult survival (Bulmer, 1985).

*Effects of the demographic environment on reproductive effort—spatially varying environments*

Kawecki and Stearns (1993) discuss the effects of varying adult or juvenile survival on optimal reproductive effort for different patch types using the hard selection model discussed above, and the reproductive effort model with constant adult survival/growth and fecundity per unit weight. They find that a phenotypic response of reduced reproductive effort is favoured in patches with higher adult survival, but variable juvenile survival should not produce any response.

### 5.3.5 Kin selection in age-structured populations

Up to now, selection has been assumed to act solely through genetic effects on the individual's own life-history traits. There are, however, instances where life-history traits seem to have been modified as a result of kin selection, i.e. in response to the fitness effects on close relatives of a change in an individual's life-history. Important examples include the evolution of sociality in insects (Charnov, 1978; Craig, 1979; Seger, 1983; Stubblefield and Charnov, 1986), and the evolution of cooperative breeding, whereby an individual defers reproduction and assists others to raise their young. This is a life-history strategy which is found in some bird and mammal species (Brown, 1987; Stacey and Koenig, 1990; Emlen *et al*., 1991).
A general theory of kin selection in age-structured populations was developed by Charlesworth and Charnov (1981), and has been extended by Rogers (1993) to include time-lags between behavioural acts and their effects on survival or fecundity. Models devised to account for special situations are described by Charnov (1978), Craig (1979), Maynard Smith (1983) and Brown (1987). The general theory for an autosomal locus will be briefly reviewed first. It is based on the conditions for invasion of a random-mating population which is initially homozygous for allele $A_1$, by a rare allele $A_2$ whose effects are assumed to be expressed in $A_1A_2$ heterozygotes. Given the possibility of asymmetric behavioural interactions between the sexes, it is necessary to distinguish between males and females; as previously (e.g. section 3.2.1), asterisks are used to denote male parameters, and the frequency of females among new zygotes is assumed to be $a$, independent of parental age or genotype. Individuals may be either donors or recipients of ‘altruistic’ acts on the part of donors, e.g. an $A_1A_2$ individual may defer reproduction in order to aid the reproduction of its parents, who are the recipients of its altruism. The question is: what differential weightings must be given to different sexes and ages of donors and recipients in determining the conditions for spread of $A_2$, given its effect on behaviour?

This question can be answered if certain simplifying assumptions are made which enable the inclusive fitness approach of Hamilton (1964) to be applied, in which the effect of altruism on the life-history trait of a recipient is weighted by its coefficient of relatedness $\rho$ with the donor. $\rho$ is here defined as the chance that the recipient contains an $A_2$ allele present in the donor, determined from the transmission probabilities for a neutral gene through the pedigree connecting donor and recipient. This is either exact or a good approximation if the effects of different behavioural interactions on the same individual are strictly additive (Cavalli-Sforza and Feldman, 1978), or if interactions occur at a low frequency or have only small effects on fitness components (Toro et al., 1982). The theory is not confined to the evolution of altruism, as the changes in life-history traits of donor and recipient are permitted to be either negative or positive.

The conditions for the spread of $A_2$ can be found as follows (for full details, see Charlesworth and Charnov, 1981). Consider first the case when interactions occur only between one pair of age and sex classes, so that only survival probabilities to the next age-classes or fecundities at the ages in question are affected. If the species is monogamous, we adopt the convention that changes in offspring number due to altruism received by a mated pair, each member of which has the same relation to the donor,
are accredited to both parents, i.e. are effectively counted twice. For interactions between individuals of the same sex, changes in survival probability to donor and recipient are each weighted by the Fisherian reproductive value of the age-class following that to which the individual in question belongs. The change to the recipient is weighted by \( \rho \). For interactions between members of the opposite sex, the change for the recipient is weighted by \( \frac{a}{1-a} \) if male, and \( \frac{1-a}{a} \) if female. \( A_2 \) will spread if the sum of the terms for the donor and recipient is positive, e.g. if the donors are females of age \( x \) and the recipients are males of age \( y \), with changes in survival \( \delta P(x) \) and \( \delta P^*(y) \) respectively, the condition is

\[
v(x + 1)\delta P(x) + \frac{\rho a}{1-a}v^*(y + 1)\delta P^*(y) > 0 \tag{5.40a}
\]

If the interactions involve fecundity rather than survival, the reproductive value weights are omitted, but otherwise the conditions are the same. If there is an interaction involving an effect on the survival probability of a donor aged \( x \) and the fecundity of a recipient aged \( y \), the reproductive value weight for the donor is retained, multiplied by a factor of \( e^{-r} \), where \( r \) is the intrinsic rate of increase of the initial \( A_1A_1 \) population. If the effects involve the fecundity of the donor and the survival probability of the recipient, the donor is unweighted and the recipient is given the relevant reproductive value weight multiplied by a factor of \( e^{-r} \). For example, if a female donor aged \( x \) experiences a change in fecundity \( \delta m(x) \) and the male recipient experiences a change in survival \( \delta P^*(y) \), the condition for spread is

\[
\delta m(x) + \frac{a\rho e^{-r}}{1-a}v^*(y + 1)\delta P^*(y) > 0 \tag{5.40b}
\]

If there are effects involving more than one age and class of relationship, these conditions can easily be extended by summing the contributions from all sources (Charlesworth and Charnov, 1981, who also describe how continuous-time populations can be modelled).

Cooperative breeding provides a specific biological situation to which these general conditions can be applied. Consider a monogamous, seasonally breeding iteroparous species in which offspring production per breeding pair per season is constant (\( m \)) in the absence of help. For simplicity, assume that the sex-ratio is one-half, and that males and females have the same demographic parameters. Let the probability of survival per season be \( P \) for breeding adults. An allele \( A_2 \) which causes some carriers to defer reproduction for one season and assist their parents to raise young appears. Let the increase in fecundity to a mated pair
which has one helper be $\delta m'$. There is a loss in fecundity $\delta m = -m$ to helpers in their first possible breeding season. Since helpers do not incur the risk of dispersal or the cost of breeding, they may also experience a change in survival $\delta P > 0$, yielding a direct fitness benefit to delayed breeding of the type we have already considered (section 5.3.4). The reproductive value weight for this term is $m/(1 - e^{-r} P)$ (equation 5.35b). The value of $\rho$ for parents and offspring is one-half, but (as noted above) the increment in fecundity must be credited to both parents if both are alive at the time they receive aid, so that $\rho$ is effectively one. If only one parent is alive, however, the value of one-half is appropriate. The probabilities of these two possible events, conditioned on at least one parent being alive, are $P^2/(1 - [1 - P]^2)$ and $2P(1 - P)/(1 - [1 - P]^2)$ respectively, yielding an overall indirect increment in fitness of $P\delta m'/[1 - (1 - P)^2]$. This assumes that the donor breeds itself if both parents die, and the nest is either deserted or occupied by non-relatives. If it remains as a non-breeder, as has usually been assumed by previous authors (Brown, 1987, p. 198), the denominator is dropped.

The condition for the spread of the gene causing helping behaviour is thus

$$\frac{m e^{-r} \delta P}{(1 - e^{-r} P)} + \frac{P \delta m'}{[1 - (1 - P)^2]} - m > 0 \quad (5.41)$$

If the survival probability $P$ is high, this equation shows that only a slight increase in survival probability to a helper is required to confer an advantage, even in the absence of a direct advantage to delayed breeding from increased survival (expressed in the first term in the inequality), provided that the increase in fecundity to the helped parents is comparable to the loss in fecundity to the helper.

If the species is only serially monogamous with a change in breeding partner each season, or is polygamous, so that the male which is assisted by the helper is not the father, then the multiplicand of $\delta m'$ in equation (5.41) is simply one-half, since the consequences of help are now credited only to the mother. The results are also easily extended to situations in which the donors defer reproduction for more than one season (Brown, 1987, p. 198). Instead of a factor of $P$ in the numerator of the multiplicand of $\delta m'$, we have a factor of $P^u$ for $u$ seasons of delayed breeding. The corresponding denominator is $1 - (1 - P)^u$.

Charnov (1978) and Craig (1979) have applied a modification of this type of model to the evolution of sociality in Hymenoptera, which have haplodiploid inheritance instead of the diploid inheritance assumed here. They imagine a species which has two breeding seasons a year, such that
the adults of the first season always fail to survive to the next year. Offspring produced in the first season of a year thus have parents who were produced in the second season of the preceding year. A female who produces a brood in the first season may also produce a brood in the second season, which is assumed to be fertilised by sperm from the male which fathered the first brood. In the initial $A_1A_1$ population, the first-season offspring found their own, independent nests. Assume that a mutant allele $A_2$ causes a female who emerges in the first season to rear the second-season offspring of her mother instead of founding her own nest, and thereby reduces her own fecundity by $m$, but increases her mother's fecundity by $\delta m'$. By the above type of argument, the condition for the spread of $A_2$ is simply $\delta m' > m$, implying that a slight disproportion in the ratio of enhancement of maternal fecundity to the reduction in offspring fecundity is sufficient to select for this form of altruism.

The same method can be applied to a mutant allele which acts in the mother, to cause a first-generation offspring to sacrifice reproduction and assist the mother (*parental manipulation*). Here, the offspring are viewed as recipients and the mother as a donor, so that change in the offspring fecundity is weighted by $\rho = \frac{1}{2}$. We now simply require $2\delta m' > m$, a much lighter condition than before (Charnov, 1978; Craig, 1979). A change in a life-history pattern may thus be brought about as a result of a genetic change expressed in an individual’s relatives rather than in the individual itself.

### 5.3.6 Sex differences and life-history evolution

Up to now, we have largely ignored the possibility that natural selection might promote different life-history patterns in males and females, or influence the sex-ratio of offspring (in species with separate sexes) or allocation to opposite sex functions (in hermaphrodite or monoecious species) as a function of age. These topics are now the subject of a large literature (reviewed by Charnov, 1982, 1993; Karlin and Lessard, 1986; Roff, 1992, Chapter 7), and will only be briefly mentioned here.

*Sex differences in timing of reproduction*

In many animal species, there is competition between males for mates which is based, in part at least, on differences in body size. Large males accordingly have the greatest reproductive success (Roff, 1992, p. 128). If such competition is sufficiently intense, there may be strong selection on males to prolong the growth period by postponing sexual
Evolution of life-histories

maturity (cf. section 5.3.3). Because of the competitive nature of selection in this case, the effects tend to be self-reinforcing, in the sense that selection for genotypes with delayed maturity is intensified once males with a modest degree of postponed reproduction have been established in the population. An evolutionary equilibrium will only be reached when the cost of further postponement of reproduction becomes large enough to balance the reproductive advantage. A model of such competitive selection on body size has been analysed by Maynard Smith and Brown (1986). Clearly, this type of selection does not operate on females, and we would therefore expect them to become sexually mature and cease growth considerably earlier than the males. This is, indeed, commonly observed (Roff, 1992, pp. 207–12).

Satellite males

An interesting outcome of selection on male reproductive success can be the maintenance of more than one type of male life-history strategy, by the frequency-dependent selection inherent in some types of sexual selection (Gross and Charnov, 1980; Gross, 1985; Gross, 1991; Roff, 1992, pp. 207–12). The commonest situation is when there are two classes of male, ‘territorial’ and ‘satellite’ (Roff, 1992, p. 209). The former are late-maturing, large, and defend territories in order to gain access to females. The latter are early-maturing, small, and obtain matings by intercepting females attracted to the territorial males or by sneaking between the male and female during fertilisation (in cases when there is external fertilisation). It is easy to see that, in principle, satellite males might be able to gain an advantage when rare, but that their reproductive success is likely to fall as they become common, owing to the lack of territorial males to attract females, and the increased competition among satellite males for access to females attracted to the remaining territorial males. If the relative net fitnesses of the two forms become equal, then the conditions for an ESS are met (Maynard Smith, 1982), and they should be maintained at that frequency. This can arise if the two forms are genotypically identical and are the products of phenotypic switching, whose frequency in the population has been adjusted by genetic modifiers to meet the ESS condition of equal fitness. Alternatively, they might represent a genetic polymorphism due (for example) to dominant and recessive alleles at a single locus. Unfortunately, the genetic basis of the differences between the forms is unknown in all cases studied so far, other than the size morphs of the swordtail Xiphophorus nigrensis, which appear to be under single-locus control (Zimmerer and Kallman, 1989).
If the ESS condition is met, and the population size is stationary, then the net reproduction rates of the two types of male must be equal. As shown by Gross and Charnov (1980), this has the consequence that the fraction of eggs fertilised by the satellite males at a given time \((h)\) should be equal to the fraction of satellite males in the population \((q)\). If \(R_s\) and \(R_t\) are the net reproduction rates of satellite and territorial males, respectively, the proportions of fertilisations accruing to the two types are \(qR_s\) and \((1 - q)R_t\). We thus have

\[
\frac{qR_s}{(1 - q)R_t} = \frac{h}{(1 - h)} \tag{5.42}
\]

But at the ESS, we have \(R_s = R_t\), and so \(q = h\).

This prediction can be tested by obtaining empirical values for both \(q\) and \(h\) (Gross and Charnov, 1980; Gross, 1991). For the case of blue-gill sunfishes, there is good agreement with the expectation of equality, but the confidence intervals are too wide to permit a rigorous test. Gross (1991) describes the results of experiments in which the frequencies of satellite males are artificially perturbed, which provide support for the hypothesis of frequency-dependent mating success.

**Sequential hermaphroditism**

This is a phenomenon whose selective causes are probably closely related to those for deferred maturity of males. It describes cases in which at least some individuals in a population start reproductive life as one sex, and change over to the other sex at some point. Sequential hermaphrodites may be *protandrous* or *protogynous*, depending on whether they start life as males or females, respectively. It is common, for example, among certain groups of fishes, such as the wrasses. Ghiselin (1969) suggested that sequential hermaphroditism has a selective advantage when there is a tendency for the fecundity of one sex to increase more rapidly with age than that of the other. Consider, for example, the case of an initially bisexual species with two age-classes, and in which the age-specific survival probabilities for males and females are equal. If the population size is stationary, then the fact that each individual has a mother and a father implies that, using the notation of section 1.2.1, we must have

\[
l(1)M(1) + l(2)M(2) = l(1)M^*(1) + l(2)M^*(2) \tag{5.43}
\]

If female fecundity is constant with age but male fecundity increases, we have \(M(1) = M(2)\) and \(M^*(1) < M^*(2)\). It follows from equation (5.43)
that $M^*(1) < M(1) < M^*(2)$. The carriers of a rare mutant gene which causes them to be female up to age-class 1, and male in age-class 2, will have an expectation of offspring of $l(1)M(1) + l(2)M^*(2)$, and so will be at a selective advantage over the rest of the population.

The main problem in the theory of the evolution of sex change is to determine what the optimal age of sex switching should be, granted that sequential hermaphroditism is advantageous. This can be tackled by an ESS approach as follows. Consider a protogynous species in which the age of switching is $x_s$. A rare mutant allele $A_2$ at an autosomal locus is introduced into the population, which changes the age of switching to $x_s$. Assuming random mating, the frequency of $A_2$ among new zygotes produced at time $t$ is equal to the mean of the frequency of $A_2$ among male and female zygotes, i.e.

$$p_2(t) = \frac{\sum_{x=b}^{x_s} B(t - x)p_2(t - x)k(x)}{2 \sum_{x=b}^{x_s} B(t - x)k(x)}$$

$$+ \frac{\sum_{x=x_s+1}^{d} B(t - x)p_2(t - x)k^*(x)}{2 \sum_{x=x_s+1}^{d} B(t - x)k^*(x)}$$

If the initial population has an intrinsic rate of increase $r$, it will be immune to invasion by $A_2$ if

$$\frac{\sum_{x=b}^{x_s} e^{-rx}k(x)}{\sum_{x=b}^{x_s} e^{-rx}k^*(x)} + \frac{\sum_{x=x_s+1}^{d} e^{-rx}k^*(x)}{\sum_{x=x_s+1}^{d} e^{-rx}k^*(x)} < 2$$

An equivalent equation can be written for the case of protandry. These equations enable predictions to be made about the timing of sex change (Warner, 1975; Warner, Robertson and Leigh, 1975; Leigh, Charnov and Warner, 1976; Charnov, 1979). Charnov (1982, Chapter 9–11) provides a comprehensive review of the theory and pertinent data. Later work is reviewed by Warner (1988), Mangel (1990) and Charnov (1993, Chapter 3). There appears to be good empirical support for the theory that sex switching is an evolved response to differences between the sexes in the relation between age and reproductive success.
Selection and the sex-ratio

Fisher (1930, pp. 158–60) was the first to propose a mechanism for the adjustment of the sex-ratio by natural selection. He pointed out that both parents make an equal genetic contribution to a zygote, with respect to autosomal loci. A male zygote will therefore have a higher net fitness (survival × fecundity) than a female zygote, in a discrete-generation population whose primary sex-ratio (sex-ratio among the zygotes, section 1.2.1) is biased towards females. (This is true even if males have a lower probability of survival than females.) Consequently, a genotype which tends to produce offspring with a higher frequency of male zygotes than average, but otherwise has the same fecundity and survival probability, will have a selective advantage. If the sex-ratio is biased in favour of males, then female-producing genotypes are favoured. The sex-ratio should therefore tend to stabilise at one-half, if there is genetic variation available at autosomal loci. This idea has been confirmed by studies of one-locus models of sex-ratio variation, e.g. Spieth (1974), Karlin and Lessard (1986). Variation at X-linked loci has similar consequences (Trivers and Hare, 1976).

As Fisher realised, a primary sex-ratio of one-half is not necessarily favoured if parents can replace offspring which die before reaching a certain age. If, for example, males tend to have a higher mortality than females, then a mother who produces a higher frequency of male zygotes than average will compensate for the greater frequency of death of these males by producing more zygotes than average, so that she will have an increased fecundity over and above any advantage due to a female-biased population sex-ratio. This results in a primary sex-ratio biased in favour of males at equilibrium. It can be shown that selection favours a sex-ratio of one-half at the age at which parents can no longer compensate for death of an offspring (Crow and Kimura, 1970, pp. 291–2).

Recurrence equations for a rare mutant allele \( A_2 \) at an autosomal locus which changes the sex-ratio among new zygotes can be derived as follows. In general, it is necessary to allow for the possibility that the sex-ratio genotype influences the age-specific survival probabilities and fecundities of the parents, since male and female offspring may incur different costs which influence the demographic parameters of their parents, e.g. male offspring might be larger than female and therefore require greater exertion on the part of the mother in species with parental care, leading to a lower survival rate of mothers who produce a high frequency of sons (Fisher, 1930; Charlesworth, 1977; Charnov, 1982, Chapter 6). For the initial \( A_1A_1 \) population, assumed to be in stable age-distribution, let the overall frequency of female zygotes be \( \bar{a} \); let the frequency of female
zygotes produced by an $A_1A_2$ mother aged $x$ be $a_x$ and the frequency of female zygotes produced by an $A_1A_2$ father aged $x$ be $a_x^*$. The corresponding values for the $A_1A_1$ population are denoted by overbars. Let the products of survival to age $x$ and number of zygotes produced at age $x$ for $A_1A_2$ females and males be $K_{12}(x)$ and $K_{12}^*(x)$ respectively. Let $p_2(t)$ and $p_2^*(t)$ be the frequencies of $A_2$ among female and male zygotes at time $t$. The following recurrence relations are obtained

$$\bar{a}B(t)p_2(t) = \frac{\bar{a}}{2} \sum_x B(t-x)p_2(t-x)K_{12}(x)a_x$$

$$+ \frac{(1 - \bar{a})}{2} \sum_x B(t-x)p_2^*(t-x)K_{12}^*(x)a_x^* \quad (5.45a)$$

$$(1 - \bar{a})B(t)p_2^*(t) = \frac{\bar{a}}{2} \sum_x B(t-x)p_2(t-x)K_{12}^*(x)(1 - a_x)$$

$$+ \frac{(1 - \bar{a})}{2} \sum_x B(t-x)p_2^*(t-x)K_{12}^*(1 - a_x^*) \quad (5.45b)$$

If the intrinsic rate of increase of the initial population is $r$, these simplify to

$$p_2(t) = \frac{1}{2} \left\{ \sum_x p_2(t-x) e^{-rx} K_{12}(x)a_x ight\}$$

$$+ \frac{(1 - \bar{a})}{\bar{a}} \sum_x p_2^*(t-x) e^{-rx} K_{12}^*(x)a_x^* \right\} \quad (5.45c)$$

$$p_2^*(t) = \frac{1}{2} \left\{ \frac{\bar{a}}{(1 - \bar{a})} \sum_x p_2(t-x) e^{-rx} K_{12}^*(x)(1 - a_x) ight\}$$

$$+ \sum_x p_2^*(t-x) e^{-rx} K_{12}^*(x)(1 - a_x^*) \right\} \quad (5.45d)$$

Using the same technique as was used for the case of the stability of an equilibrium with heterozygote advantage (section 4.3.6), the asymptotic rate of increase of $A_2$ when rare is given by the largest real root $z_1$ of

$$2 = \sum_x e^{-(r+z)x}[K_{12}(x)a_x + K_{12}^*(x)(1 - a_x^*)]$$
This provides a means for determining the ESS pattern of sex-ratio as a function of age, since a population for which the population sex-ratio pattern is such that \( z_1 \leq 0 \) for all small perturbations meets the definition of an ESS (Maynard Smith, 1982). A necessary condition for this is that the total differential of \( z_1 \) is zero at the population sex-ratio pattern, i.e. that the derivatives \( \partial z_1 / \partial a_x \) and \( \partial z_1 / \partial a_x^* \) are zero for all reproductive ages, when evaluated at the sex-ratio pattern for the population in question.

In the simplest case, there is no relation between offspring sex-ratio and parental demographic parameters, as is likely in species with no parental care. The relevant derivatives in this case are as follows

\[
\frac{\partial z_1}{\partial a_x} \propto e^{-rx} K(x) \left\{ 1 - \frac{1}{2} \sum_x e^{-rx} K^*(x) \right\} \quad (5.47a)
\]

\[
\frac{\partial z_1}{\partial a_x^*} \propto e^{-rx} K^*(x) \left\{ \frac{1}{2} \sum_x e^{-rx} K(x) - 1 \right\} \quad (5.47b)
\]

where \( K(x) \) and \( K^*(x) \) are the products of survival to age \( x \) and fecundity at age \( x \) for females and males, respectively, for the initial population. For the initial population to be in stable age-distribution, considerations analogous to those leading to equation (1.32) show that

\[
\sum_x e^{-rx} K(x) = \frac{1}{a}, \quad \sum_x e^{-rx} K^*(x) = \frac{1}{(1 - \tilde{a})}
\]

Substituting these into equations (5.47), we find the ESS conditions

\[
1 - \frac{1}{2(1 - \tilde{a})} = \frac{1}{2\tilde{a}} - 1 = 0 \quad (5.48)
\]

i.e. an overall zygotic sex-ratio of one-half is favoured, as in the corresponding discrete-generation case.

Clearly, these relations make no predictions about sex-ratio as function of parental age; as long as the overall sex-ratio is one-half, the ESS conditions can be satisfied regardless of the relations between sex-ratio and parental age.

The situation is more complicated when there are different costs of
male and female offspring, which affect the subsequent performance of
the parents. Consider the case when sex ratio is controlled only by
maternal genotype. It is reasonable to assume that offspring sex ratio at
age $x$ can only influence the demographic parameters of mothers aged $x$
or more. This effect can be encapsulated in the partial derivative of $K(y)$
with respect to $a_x$, $\partial K(y)/\partial a_x (y \geq x)$. If male offspring cost more than
female offspring in terms of maternal survival or fecundity, for example,
we would have $\partial K(y)/\partial a_x > 0$; the opposite inequality would hold if
females were more costly than males. Including these effects in the
derivatives of $z_1$, we obtain the following relation for each reproductive
age $x$

$$e^{-rx}K(x)\delta + \sum_{y>x} e^{-r_y}\left(\frac{\partial K(y)}{\partial a_x}\right)\left(\bar{a}_y \delta + \bar{a}\right) = 0 \tag{5.49a}$$

where $\delta = 2(1 - \bar{a}) - 1$ is a measure of the bias of overall zygotic sex-
ratio towards males. If males cost more than females, the second term is
positive, so that $\delta$ must be negative for an ESS to exist, i.e. the overall
zygotic sex-ratio is female-biased, as would be expected intuitively. The
opposite bias obviously occurs if females cost more than males. This is
qualitatively the same as the bias found in discrete-generation models
with different costs of males and females.

If only fecundity at age $x$ is affected by the sex-ratio of the offspring of
mothers aged $x$, as would be the case if male female and female offspring
require different amounts of resources to raise to age 1, then equation
(5.49a) becomes

$$e^{-rx}K(x)\left\{\delta + \left(\frac{\partial \ln M(x)}{\partial a_x}\right)\bar{a}_x \delta + \bar{a}\right\} = 0 \tag{5.49b}$$

where $M(x)$ is the total number of offspring produced at age $x$.

If the effect of sex-ratio on fecundity is small, neglect of second-order
terms leads to the following approximation for $\delta$

$$\delta \approx -\frac{1}{2} \left(\frac{\partial \ln M(x)}{\partial a_x}\right)_{a_x=\frac{1}{2}} \tag{5.50}$$

If $\ln M(x)$ is linear in $a_x$, then this is also the approximate sex-ratio bias at
all ages. If $\ln M(x)$ is non-linear, so that the value of the derivative
changes with $a_x$, an infinite number of solutions to (5.49b) is possible, as
in the case of no dependence of fecundity on sex ratio, with the only
constraint being that the overall sex-ratio satisfies equation (5.50).

If survival probability at age $x$ is affected by the sex-ratio of the
offspring of mothers aged $x$, as would be the case if a brood with a bias
towards males makes higher demands on the mother than a female-biased brood, we have
\[ e^{-rx} K(x) \delta + \left( \frac{\partial \ln P(x)}{\partial a_x} \right) (\bar{a}_x \delta + \bar{a}) \sum_{y > x} e^{-ry} K(y) = 0 \quad (5.49c) \]
The different functions of age on the left-hand side of this equation indicates that the ESS sex-ratio is normally constrained with respect to the relation between maternal age and sex-ratio. Insight into the possible nature of the constraints can be obtained by rewriting equation (5.49c) as follows
\[ \delta + \left( \frac{\partial \ln P(x)}{\partial a_x} \right) (\bar{a}_x \delta + \bar{a}) \sum_{u=1}^{K(x+u)} e^{-ru} \frac{K(x+u)}{K(x)} = 0 \quad (5.49d) \]
If the life-history shows senescence, the summation over \( u \) will frequently decrease with increasing \( x \). This is the case, for example, if fecundity is constant with age and survival decreases. Under this condition, the decline in the summation term must be balanced by an increase in the absolute value of its multiplicand, for equation (5.49d) to be satisfied for each \( x \). If the effects of sex-ratio on survival are small, the leading term in the multiplicand is \( (\partial \ln P(x) / \partial a_x) / 2 \).

Evolutionary stability of the sex-ratio requires \( P(x) \) to be a concave function of \( a_x \), over the range of possible ESS values, ensuring that \( z_1 \) is a maximum at the solution of equations (5.49) (cf. equation (5.31c)). It follows that, if \( P(x) \) increases with \( a_x \), implying that males cost more than females (so that \( \delta < 0 \)), the decline in the summation term with age must be balanced by an increase in \( \partial \ln P(x) / \partial a_x \), and hence there is an increase in the proportion of males with advancing maternal age. The reverse is true if males cost less than females. More complex patterns are clearly possible if the summation term does not decrease monotonically with age, as might be the case if female fecundity increases with size.

Charlesworth (1977) studied models in which there was compensation of the type likely to occur in non-contracepting human populations. Here, there is a prolonged period of natural sterility during lactation, so that a child which dies in utero or before weaning causes its mother to become fertile sooner than if the child had survived to weaning. A maternal genotype that tends to produce a higher-than-average frequency of male zygotes (which have a higher mortality than females during most of the period of life in question) is thus associated with a shorter interval between successive conceptions than average. Using the renewal model of Henry (1953, 1957) for human age-specific fecundity, Charlesworth showed that the ESS primary sex-ratio for human populations is biased in favour of males, as expected from the discrete-generation result. By the
end of the first eighteen months or so of post-natal life, death of an infant
is not likely to shorten the period of infertility significantly, and the ESS
sex-ratio at this age is biased in favour of females. If the mortality rates
for male and female infants in the first year of life in typical pre-industrial
populations are combined with typical statistics on sex-ratios at birth
(which are usually male-biased), a female-biased sex-ratio for one-year
old infants is obtained, in agreement with the ESS prediction.

It was also possible to carry out a somewhat crude analysis of the ESS
for sex-ratio as a function of parental age, as well as the overall sex-ratio.
A tendency for younger females to produce a higher frequency of males
than older females would be predicted. An intuitive interpretation of this
is that older women have a greater chance of becoming sterile or dying in
the intervals between conceptions than young women, so that they have a
lower net ability to compensate for the deaths of offspring than younger
women. Their offspring should accordingly have a sex-ratio closer to
one-half. It is interesting to note that the sex-ratio at birth in human
populations is dependent upon maternal age in the way predicted by this
theory (Novitski and Kimball, 1958; Teitelbaum, 1972). There is thus
reasonably good agreement between the expectations of sex-ratio theory
and observations on human populations. This should not be taken as
confirming the theory, since sex-ratios at birth are affected by a variety of
socio-economic factors, and alternative explanations of the association
between maternal age and sex-ratio can be put forward. If it could be
shown that the primary sex-ratio in man is indeed biased in favour of
males, the theory would be considerably strengthened, but at present this
sex-ratio is an unknown quantity.

Sex-ratio selection with variable environments

Werren and Charnov (1978) and Seger (1982) have shown that
variable environments may have significant effects on the outcome of
selection on the sex-ratio in age-structured populations. One model which
is especially relevant to insect populations is when there are two breeding
seasons, spring and autumn say. Individuals of either sex which are born
in the spring reproduce in the autumn and die. Females born in the
autumn survive over winter and reproduce and die in the spring; males
born in the autumn also reproduce in the spring, and some survive to the
autumn and reproduce. This has the consequence that males born in the
spring face competition for mates from those born in the autumn,
whereas spring-born females do not face such competition. This favours
an ESS zygotic sex-ratio which is biased towards females in the spring and
towards males in the autumn. Seger (1982) reviews data on a number of
species of solitary Hymenoptera with life-histories which fit the assumptions of this model. Statistically significant sex-ratio biases in the predicted direction are frequently observed. Charnov (1982, Chapter 6) reviews further evidence on sex-ratio shifts associated with varying environments in age-structured populations.

*Sex allocation with age-structure*

There is a large literature on the evolutionary adjustment of resources between male and female allocation in cosexual (hermaphrodite or monoecious) species, which is particularly relevant to flowering plants (Charlesworth and Morgan, 1991). Differential costs of male versus female structures in cosexuals may affect subsequent survival or growth of mature individuals, and this may have significant effects on the ESS proportion of resources allocated to different sex functions (D. Charlesworth, 1984; Charnov, 1988; Charnov and Dawson, 1989; Tuljapurkar, 1990).

5.3.7 Some general aspects of life-history theory

We shall conclude this section with a brief discussion of some general aspects of life-history theory.

*The uses of reproductive value*

Ever since Fisher introduced the concept in 1930, there has been a strong temptation for biologists to use reproductive value as a tool for analysing life-history evolution and related problems. As we have seen, it cannot be used as an index of the intensity of selection as a function of age, despite Fisher’s (1930, p. 29) remarks concerning the possible significance of the inverse relationship between reproductive value and mortality observable in human populations, and Medawar’s use of it in discussing the evolution of senescence (section 5.3.1). Reproductive value or, rather, Schaffer’s (1974a) generalisation of it, does play an important role in determining the conditions for an optimal life-history under the reproductive effort model (section 5.3.4). Furthermore, as discussed by Schaffer and by Taylor *et al.* (1974), Caswell (1980), Charlesworth (1980, pp. 265–7), Schaffer (1981), Goodman (1982) and Yodzis (1982), there is a certain sense in which reproductive value at each age is maximised in optimal life-histories. This may be seen as follows. As we saw on p. 186, the condition for an optimal life-history in a density-independent population is that the sum $\sum e^{-rx} k(x)$ be at a local maximum with respect to perturbations in $k(x)$, $r$ being held constant at the value which makes the original sum unity. If we consider a certain age $y$, variations in the
life-history at age $y$ and later ages affect only the component $\sum_{x=y} e^{-rx} k(x)$ of the above sum, regardless of the particular constraints assumed in the optimisation model. Hence, if a given life-history maximises $\sum e^{-rx} k(x)$, the sum $\sum_{x=y} e^{-rx} k(x)$ must be maximised with respect to perturbations at ages $y, y+1, \ldots$, the life-history at earlier ages being held constant. This is true for each value of $y$. Conversely, maximisation of each sum $\sum_{x=y} e^{-rx} k(x)$ implies that the life-history is optimal. The maximisation of reproductive value follows from this, since we have

$$v(y) = \frac{e^{ry}}{l(y)} \sum_{x=y} e^{-rx} k(x)$$

and $l(y)$ is unaffected by perturbations at ages $y$ and beyond. Similar remarks apply to density-dependent populations. There is, of course, nothing of particular biological significance in this maximisation of reproductive value, since any product of $\sum_{x=y} e^{-rx} k(x)$ and a factor that is independent of events at age $y$ and beyond would have the same properties.

The restriction that $r$ be held constant while carrying out perturbations to the life-history can be relaxed for the case of perturbations to age-classes which do not have boundary values for the life-history variables under consideration (e.g. age-classes with intermediate reproductive effort). The total differential of $r$, $dr$, must be zero at a local maximum in $r$ with respect to such perturbations, so that the restriction of constant $r$ when finding maxima in reproductive value, etc. is unnecessary in this case. With the reproductive effort model, described in section 5.3.4, $v(y)$ (or, more generally, $b(y)$, if growth as well as survival is taken into account) is maximised with respect to perturbations at ages $x < y$ as well as $x \geq y$. (The restriction of constant $r$ must be applied if these ages have reproductive efforts of 0 or 1.) This is because $P(x)$ or $\tilde{P}(x)$ depend only on $m(x)$ or $\tilde{m}(x)$ in this model, so that perturbations to reproductive value at age $y$ leave later ages unaffected. The result is, of course, not true if the effects of changes in reproductive effort at ages $x < y$ influence survival or growth at age $y$ and beyond. There is therefore a much stronger sense in which selection can be said to maximise reproductive value in reproductive effort models, than in more general optimisation models.

Another role for reproductive value has been suggested by Slobodkin (1974). He proposed that the life-history of a prey species which is subjected to an increased rate of predation at a certain age would tend to evolve in such a way that reproductive value is withdrawn maximally from
that age. This possibility has been investigated analytically by Michod (1979) and Law (1979b), using the methods described on p. 221. They were able to confirm Slobodkin's suggestion for certain types of population, although its general validity remains to be established.

Finally, MacArthur and Wilson (1967, Chapter 4) suggested that a colonising species would be selected to send out individuals to found new colonies at an age when these individuals have maximal reproductive value, on the grounds that their probability of successfully founding a new population would be highest at this age. This was investigated by Williamson and Charlesworth (1976), using branching process theory. They found that there is often no particular relationship between reproductive value at a given age and the probability of survival of a population founded by an individual of that age.

Maximisation of population parameters

A number of life-history studies have been published which make use of the maximisation of population parameters. For example, Holgate (1967) and Mountford (1971, 1973) treated certain problems by finding the maximum probability of survival of a population. Cohen (1966, 1967, 1968, 1970) maximised the geometric mean of the growth-rate of a population in a temporally varying environment. As has already been discussed (section 5.1), it is necessary to demonstrate that natural selection can actually have the effect of maximising the population parameter in question, unless one is willing to invoke group selection. The use of survival probability can be justified in selectionist terms by regarding the calculations in question as dealing with the probability of survival of a mutant gene rather than a whole population, as mentioned in section 5.2.3. Although the calculations often turn out to be identical, the conceptual distinction between the two procedures is important. Similarly, under suitable conditions, Cohen's models can be translated into considerations of the rate of spread of a rare gene, along the lines indicated in section 4.3.3.

5.4 Tests of theories of life-history evolution

In the previous section of this chapter, we have discussed a variety of theoretical ideas about the ways in which natural selection can mould life-histories. In the past 15 years, a great deal of effort has been expended on attempts to test these ideas against data from both experiments and species comparisons. This literature has recently been the subject of several books (Finch, 1990; Rose, 1991; Roff, 1992; Stearns, 1992; Charnov, 1993). There is no point in trying to emulate these here.
Instead, a number of case-studies of specific topics will be described, which are intended to illustrate some of the main biological points involved.

5.4.1 *Some principles of life-history evolution*

In order to introduce the empirical studies, some of the main principles which have emerged from the theoretical work will first be summarised.

*(i) Age-specificity and effects on fitness of changes in life-history variables*

As we saw in section 5.2.1, the sensitivity of fitness to changes in age-specific survival probabilities or fecundities decreases with age (apart from changes in survival during the pre-reproductive period in the case of survival, or changes in fecundity in the unimportant case of a rapid and sustained decline in population size). This implies that, other things being equal, there is always a selective premium on genetic improvements in life-history traits early in life, relative to comparable changes later on. This provides an inbuilt selection pressure towards the evolution of senescence, early reproductive maturity, and semelparous rather than iteroparous reproduction. The fact that organisms often postpone reproduction or are iteroparous means that there must be costs attached to early or intense reproduction, as modelled in sections 5.3.2–5.3.4. Determining the nature of these costs has been a major goal of empirical research.

*(ii) Effects of a constant demographic environment on the evolution of the life-history*

The demographic environment (sources of mortality such as predation and disease, over which the individual has little control; the rate of population growth-rate; the mode of density-dependent population regulation) greatly influences the extent to which the sensitivity of fitness to changes in life-history traits alters with age. As discussed in section 5.2.4, the decline with age in the sensitivity functions depends on the age-structure of the population. Ecological circumstances which tilt the age-structure towards younger individuals, such as high adult mortality in populations regulated by adult fecundity or by survival very early in life, lead to a faster rate of decline in the sensitivities, compared with populations in which a high proportion of old individuals is maintained. As described in sections 5.2.4 and 5.3.2–5.3.4, populations of the first type should tend to evolve more rapid senescence, should reproduce
earlier and be more likely to be semelparous, and should invest in higher reproductive effort if iteroparous, than populations of the second type. Conversely, under the reproductive effort model of section 5.3.4, higher juvenile mortality favours lower reproductive effort. In addition, populations with high adult survival, a high degree of sensitivity of survival to reproductive effort and a high maximum fecundity per unit weight are likely to experience an increase in reproductive effort with age, especially if they are density-regulated through fecundity or juvenile survival.

(iii) Effects of variation in the demographic environment
The expected effects of temporally varying demographic parameters on the evolution of the life-history are more complex. As described in section 5.3.2, temporal variation in juvenile survival allows iteroparity to evolve under lighter conditions than when the environment is constant. The reverse is true when adult survival varies. Similarly, delayed reproduction is favoured by temporally varying juvenile survival or growth-rate when fecundity is size-dependent, but the effect of varying adult rates depends on the autocorrelation between time-intervals (section 5.3.3). Large variances in survival or fecundity may be necessary to overcome differences in intrinsic rate for the mean life-histories, however, so that such effects of temporal variation are far from inevitable. Reduced reproductive effort is favoured by varying juvenile survival or fecundity; the opposite is true when adult survival varies (section 5.3.4). Facultative adjustments of the life-history to local conditions are favoured when there is spatial variation in the demographic environment. The expected nature of these adjustments is broadly in line with those predicted for populations evolving in a constant environment, except that variable juvenile survival has no effect on the optimal level of reproductive effort (sections 5.3.3 and 5.3.4).

(iv) Consequences of kin selection or parental manipulation
Under suitable conditions the direct effects of selection on the life-history traits of individuals of the kind just described may be outweighed by the fitness consequences of these traits for close kin, or individuals may be manipulated into adopting life-histories which benefit their kin. In particular, deferred reproduction may be favoured if non-breeders help their parents raise young (section 5.3.5).

(v) Sex effects on life-history evolution
Males and females may exhibit different age-specific patterns of fitness sensitivities (section 5.3.6). This can result in relatively minor
adjustments to the life-histories of the two sexes, such as postponement of reproductive maturity in males relative to females when there is strong size-based sexual selection. More dramatic patterns, such as age-dependent sex switching, may evolve when male and female reproductive success have sufficiently different patterns of dependence on age or size. Finally, if sex-ratio (in species with separate sexes) or allocation of reproductive resources to different sex functions (in cosexual species) can be adjusted either genetically or facultatively, age-specific patterns of sex allocation or offspring sex-ratio may be favoured. Evidence relating to sex effects has already been discussed briefly, and we will not consider this topic further.

5.4.2 The nature of tests of theories of life-history evolution

The theoretical principles of life-history evolution just described follow logically from the rules of genetics and demography. It is obviously unnecessary to test some of the most basic principles, since we know that these rules are empirically valid. It is not a matter for controversy or empirical investigation whether the sensitivity of fitness to changes in life-history traits declines with age, or how the rate of decline is affected by the demographic environment. However, even the most basic principles will have no consequences for what happens in evolution if ancillary requirements are not satisfied, notably that genetic variability of the appropriate type must be available if selection is to be effective. Thus, although a selection pressure in favour of senescence is an inevitable outcome of genetics and demography when there is a distinction between somatic and reproductive structures, senescence cannot evolve if there is no age-specificity of genetic effects on life-history traits (section 5.3.1). As in other areas of evolutionary biology, the collection of data on the physiological and genetic phenomena assumed in different types of model forms an extremely important component of empirical work on life-history evolution.

The other component concerns the question of whether or not life-history patterns observed in nature agree with theoretical predictions. Conversely, one can ask which type of evolutionary model is most likely to explain an observed pattern, e.g. why is the characteristic age at reproductive maturity of a species often negatively correlated with its species-specific mortality rate, when related species are compared (section 5.4.5)? The difficulty here is that, as is often the case in evolutionary biology, similar patterns can be produced by different causes, e.g. both the size-related fecundity and reproductive effort models predict earlier
reproduction when mortality is high (sections 5.3.3 and 5.3.4). Thus, causes cannot be securely inferred from effects.

Ideally, a combination of research on the nature of life-history variation within species and analyses of the products of evolution under different demographic environments should provide a resolution of this problem. As we shall see, although a great deal of valuable data on both aspects of life-history evolution has been amassed, this ideal has been rarely, if ever, been approached. On the one hand, we have evidence in many systems for the physiological and genetic phenomena postulated in many of the models of life-history evolution; on the other, patterns of life-history differences between populations or species which match the predictions of the models have been repeatedly described. There can be little doubt that life-histories are indeed being moulded by selective forces of the kind described above; what we are usually unsure of is the exact nature of these forces in individual cases.

5.4.3 The evolution of senescence

This topic has recently been the subject of two full-length books (Finch, 1990; Rose, 1991), and so will only be briefly discussed here.

Genetic evidence

There is ample evidence for age-specific genetic effects on life-history traits, which are required for senescence to evolve in response to the decline in fitness sensitivity with age (section 5.3.1). This evidence comes from studies of a variety of species, from Caenorhabditis elegans through Drosophila melanogaster to humans, and both major gene and quantitative genetic effects have been described (Finch, 1990, Chapter 6; Harrison, 1990; Rose, 1991, Chapter 4).

What is less clear from the genetic studies is which of the two most plausible scenarios for the evolution of senescence, mutation accumulation and antagonistic pleiotropy, is likely to be the more important factor in the evolution of senescence, or indeed whether both of them may have contributed significantly. The evidence concerning mutation accumulation is as follows. Consider first a mutant allele at a locus that is expressed only at a given age $x$. From equations (5.5) of section 5.3.1, it is easily seen that the population's equilibrium reduction in log survival probability at age $x$ (below that for the wild-type homozygote) is proportional to the mutation rate to the allele in question, $u$, divided by the fitness sensitivity for age $x$. If the effects of different loci are approximately
additive, a large decline with age in survival probability requires that the sum of the \( u \) values over all loci be large, except for very late in life when the fitness sensitivities approach zero. Given the fact that mutation rates per locus are typically of the order of \( 10^{-5} \) or less, this requires the existence of a substantial number of loci capable of mutating to deleterious alleles with highly age-specific effects. Similar results apply to fecundity.

Research into the total rate of mutation to deleterious alleles in several species suggests that in *Drosophila* and flowering plants there is probably an average of at least one new deleterious mutation per diploid individual per generation (Crow and Simmons, 1983; Kondrashov, 1988; Charlesworth *et al.*, 1990; Houle *et al.*, 1992). The rate is likely to be substantially higher in mammals, with their much greater genome size (Kondrashov, 1988). Thus, a critical requirement of the mutation accumulation theory is met: there is a wealth of mutational input into the population of deleterious alleles at a large number of loci.

The question of whether their effects are sufficiently often age-specific to account for senescent decline is less easy to answer. In humans, research in medical genetics has uncovered numerous cases of low frequency, highly deleterious alleles maintained by mutation, whose effects are age-dependent (Finch, 1990, Chapter 6). The case of Huntington's chorea, discussed in section 3.4.2, is a classic example of a mutant gene whose expression is deferred until late in life. Major genes of this kind clearly meet the requirements of the mutation accumulation theory, but their rarity means that their net contribution to the net decline in survival or fecundity with age is likely to be small. Genetic factors with more minor effects have been implicated in many disorders associated with ageing in humans, such as cancer and heart disease, although the mechanisms by which the genes concerned are maintained in the population have mostly not been elucidated (Finch, 1990, Chapter 6). Although mutation accumulation seems an attractive candidate as a contributory factor to human senescence, its proportionate contribution remains to be determined.

In *Drosophila*, Mueller (1987) maintained a laboratory population of *D. melanogaster* for over 120 generations under conditions in which early reproductive performance was favoured but where there was no selection for reproduction late in life. No increase in early performance was observed, but fecundity late in life declined, suggesting that deleterious mutations affecting fecundity late in life had accumulated in the absence of counter-selection. On the other hand, mutations which have been accumulated for 44 generations on the second chromosome of *D. melano-
Tests of theories of life-history evolution

Gaster in the almost complete absence of selection show high positive genetic correlations between life-history components measured early and late in life (Houle et al., in preparation). This indicates that there is little scope for age-specific mutations of the type postulated in the mutation accumulation theory. Similarly, Rose and Charlesworth (1981) found no statistically significant increase with age in additive genetic variance in female fecundity, despite the fact that the mutation accumulation theory predicts that additive variance due to deleterious mutations should increase with age in parallel with the decline in mean, since both reflect higher frequencies of deleterious alleles with late ages of expression (Charlesworth, 1990a). More recently, Hughes and Charlesworth (1994) and Hughes (in preparation) have found significant increases with age in the additive genetic variances of mortality rate and male mating success. Other evidence is reviewed by Rose (1991, Chapters 3–4), and Partridge and Barton (1993). Overall, there is only limited support for the mutation accumulation theory from studies of Drosophila. This may simply reflect the difficulty of obtaining statistically significant evidence.

There is now extensive genetic evidence, primarily from Drosophila, that early reproduction has adverse effects both on survivorship and on late reproduction (Rose, 1991, Chapters 3–4; Partridge and Barton, 1993). This comes from quantitative genetic studies in which early female fecundity is found to be negatively genetically correlated with longevity and with fecundity later in life (Rose and Charlesworth, 1981; Service and Rose, 1985), and from experiments in which laboratory populations derived from natural populations are maintained under conditions where individuals reproduce unusually late in life (Rose, 1991, Chapter 3). Increased longevity and late female fecundity are frequently observed in such experiments. Increased late fecundity is commonly, but not always (Partridge and Fowler, 1992), accompanied by a decrease in early fecundity. The decline in early fecundity could in theory be due to the accumulation of mutations with effects specific to early life, but the absence of evidence for such age-specific mutations in accumulation experiments and the fact that the rate of decline is too fast for the observed rate of mutational decline for early female fecundity, even if all mutations were age-specific, rule this possibility out (Charlesworth, 1984; Houle et al., in preparation).

Overall, these experiments demonstrate very clearly that ample additive genetic variability with respect to life-history traits is available in natural populations of Drosophila, of a kind which permits experimentally induced changes in the demographic environment to produce the theoretically expected shifts in the timing of senescence. Negative genetic
correlations between early and late fecundity, and between early fecundity and adult survival rate, seem to play a major role in producing these shifts.

Comparative evidence

The two major candidates for evolutionary causes of senescence both make similar predictions concerning the effects of the demographic environment on the intensity of selection for senescence (see section 5.2.4), and so comparisons of species or populations cannot be expected to shed light on which is the more important factor. Comparative studies can, however, provide useful evidence on whether or not observed differences in patterns of ageing among taxa have evolved in the ways predicted by theory. This question was first discussed by Williams (1957), whose paper laid the foundations for the interpretation of the comparative biology of ageing in the light of evolutionary theory. More recent discussions are given by Comfort (1979), Finch (1990, Chapters 1–5), Rose (1991, Chapter 5), Austad and Fischer (1991), and Promislow (1991).

There is one caveat which should be borne in mind when considering the comparative evidence. Much of the data are expressed in terms of estimates of the expectation of life (mean longevity) or the maximum life-span characteristic of the species. The latter statistic is sample-size dependent, and hence rather unreliable. Both measures are affected by the initial mortality rate of the group of individuals being followed, as well as the rate of increase in mortality with age (Finch, 1990; Promislow, 1991). Few comparative studies have attempted to separate out these two determinants of life-span, only the second of which reflects senescence (see Finch, Pike and Whitten (1990) and Promislow (1991) for exceptions). While it seems reasonable to suppose that the great differences in life-span between mice and humans, for example, must reflect differences in the rate of ageing, this is less clear for closer taxonomic comparisons. There is also very little information on senescent decline in reproductive capacity. Use of data on captive populations, where extrinsic sources of mortality are minimised, and senescence is much easier to detect, should help deal with this problem (Austad and Fischer, 1991).

The first point brought out by Williams (1957) was that senescence is expected to be universally present in organisms in which there is a clear distinction between parent and offspring, even if reproduction is asexual. No such distinction can be made in multicellular organisms which reproduce by vegetative means which involve splitting of the whole organism into nearly equal parts, or in unicellular organisms. Groups such as
ciliates provide an exception to the latter. Their cells contain a vegetative
macronucleus responsible for synthesis of most proteins and a micro-
nucleus which participates in sexual reproduction, following which the
macronucleus is reconstituted from the micronucleus. Consistent with
theory, ciliates display a form of senescence if individuals are maintained
by asexual reproduction, associated with genetic deterioration of the
macronucleus (Bell, 1988).

The rules seem to be well obeyed by multicellular organisms, where
senescence seems to be observed whenever detailed studies have been
made and the demography of the species makes it feasible to detect
senescence (Williams, 1957; Comfort, 1979; Rose, 1991). Finch (1990,
pp. 208–24) points to some cases of long-lived species in which there is no
apparent evidence for senescence. These are difficult to evaluate, as
long-term studies in captivity have not been done in these cases. Bell
(1984) compared species of freshwater invertebrates which reproduced by
fission with sexual species, and showed that the former lacked detectable
senescence with respect to mortality whereas the latter often exhibited
senescence, in accord with theory. Martinez and Levinton (1992) have
demonstrated senescence in two species of asexually reproducing animals
in which there is a clear-cut distinction between parent and offspring.

Second, demographic environments with high external sources of mort-
ality such as predation should select for faster rates of senescence than
populations with low intensities of such factors (see section 5.2.4 for some
qualifications on this). Austad (1993) has shown that an island population
of the Virginia opossum exhibits higher survival, slower senescence and
reduced reproductive rate than mainland opposums, from which the
island population presumably originated. While it remains to be conclus-
ively established that these differences are genetic, they are in the
direction expected if predation pressure is relaxed on the island compared
with the mainland. As pointed out by Williams (1957), the prediction also
corresponds well to major differences between groups. Large body size in
mammals is associated with longer maximum life-span (Comfort, 1979;
Prothero and Jürgens, 1987), and larger species tend to have lower
mortality rates in the wild (Caughley, 1966; Harvey and Read, 1988;
Promislow, 1991), presumably because of factors such as lower predation
rates. Promislow (1991) failed, however, to find the expected relation
between mortality rate of reproductively mature individuals and the rate
of increase in mortality with age in data on natural populations of
mammals. This may reflect the difficulty in obtaining reliable estimates
from such data. Significant associations between the rate of senescence
and other life-history traits which are expected to be selected for by high
mortality (such as early development and high reproductive rate – section 5.4.1) were, however, detected by Promislow. Flying, gliding and tree-dwelling species of mammals tend to have much longer life-spans than terrestrial species of comparable size (Austad and Fischer, 1991, 1992). It is reasonable to suppose that their modes of life protect against predation. The same applies to birds when compared with mammals (Williams, 1957; Prothero and Jürgens, 1987; Calder, 1990). It is in groups which have high levels of unavoidable mortality, such as small insects, that rapid senescence is expected and observed.

Third, species where fecundity increases with age, such as many cold-blooded vertebrates (section 1.2.4), should show relatively low rates of senescence with respect to survival, since increasing fecundity tends to reduce the differential in favour of younger age-classes that is seen in the fitness sensitivities. This is in agreement with the observation that species of large reptiles, amphibia and fish may have very high maximum life-spans (Williams, 1957; Comfort, 1979).

Fourth, the theory suggests that senescence with respect to survival should start immediately after the age of first reproduction, since the sensitivity \( s(x) \) is constant for all pre-reproductive ages, and then starts to decline. Owing to the inadequacies of the data on animal life-tables, and the confounding effects of differences in ecology between juvenile and adult stages in many species, it is difficult to be certain how generally this prediction is verified. Human life-tables do show a minimum in mortality rates at or near the age of puberty, with a continuous increase with age after that point (see Williams, 1957; Hamilton, 1966, and Table 1.2); the rise in mortality after puberty fits the theory well. This raises the question of why juvenile mortality rates should be so high in many species. On the present theory, there should be no selection for changes in the distribution of mortality within the pre-reproductive period, except insofar as any age-specific gene effects overlap the beginning of reproduction; this might lead to some tendency for higher survival rates early in the pre-reproductive period. It is easy to invoke developmental and ecological reasons for high juvenile mortality, such as the small size of juveniles and consequent vulnerability to predation and accidental death. In vertebrates, the existence of a period of immunological tolerance in very young individuals necessarily implies a relatively high rate of death due to infectious disease.

Finally, there should be no opposition to the spread of genes which reduce survival after the age of termination of reproduction, in species where there is an abrupt termination. This does not, of course, imply that the end of reproduction should necessarily be followed by immediate
death, merely that selection against senescent decline is ineffective after reproduction has ceased. Possible advantages of the extended post-menopausal lives of human females, in terms of better ability of non-reproducing women to care for children or grand-children, have been discussed by Williams (1957), Hamilton (1966) and Rogers (1993). This presupposes the post-menopausal survival of a significant fraction of women under primitive conditions, for which there is evidence from studies of hunter-gatherer societies (Howell, 1976).

5.4.4 Iteroparity versus semelparity

Three circumstances which tend to favour iteroparity over semelparity have been discussed in sections 5.3.2 and 5.3.4: (i) high adult survival (if an iteroparous life-history is achieved), and a relatively low ratio of semelparous to iteroparous fecundities; (ii) a fluctuating environment in which fecundity or juvenile survival rather than adult survival varies; and (iii) a concave rather than convex relation between reproductive effort and survival or growth. Conditions (i) and (iii) are closely connected, since condition (iii) implies that only a modest gain in fecundity would result from a shift of all resources into reproductive effort in the case of a concave curve, compared with the gain for a convex curve (Fig. 5.3).

Phenotypic evidence

Experimental manipulations of some semelparous species have demonstrated that the survival of individuals can be greatly prolonged by preventing or delaying reproduction, e.g. removal of the optic gland of octopuses prevents reproduction and death (Wodinsky, 1977), and castration of Pacific salmon greatly prolongs their life (Robertson, 1961). Other examples are reviewed by Finch (1990, Chapter 2). These examples suggest that individuals of these species have been programmed to re-allocate so many resources to reproduction that death ensues.

Genetic evidence

There is evidence in some cases for within-species variation in semelparity versus iteroparity, e.g. in rice (Sano, Morishima and Oka, 1982). This suggests that at least some species have the capacity to respond to a pressure of selection to change their life-history in this radical way. An unusual example of genetic variation for semelparity versus iteroparity has been studied by Grosberg (1988) in the ascidian
Botryllus. This is a colonial marine animal, where the zooids produced by asexual budding of an individual originating by sexual reproduction remain physiologically connected. Zooids reproduce sexually and shelter the developing young within their body. In a population studied by Grosberg, semelparous and iteroparous morphs could be clearly distinguished and the polymorphism appears to be genetic, although the details of the genetics are unknown. Grosberg showed that the semelparous morph reproduces earlier, and has higher reproductive output per clutch, than the iteroparous morph. There is also evidence for temporal variation in the survival to maturity of both morphs. The difference in reproductive output between the two morphs is qualitatively in accord with conditions (i) and (iii). The means by which the polymorphism is maintained is obscure, but might be related to the observed temporal variation in survival (Hairston et al., 1970; Bulmer, 1985).

Comparative evidence
This has been thoroughly reviewed by Roff (1992, pp. 246–50), who concludes that semelparous species do indeed show higher levels of investment in reproduction compared with iteroparous relatives, in comparisons across taxa in groups as different as grasses (Wilson and Thompson, 1989) and gastropods (Calow, 1978). Again, this is qualitatively consistent with conditions (i) and (iii). Young (1990) reached a similar conclusion from a review of a large number of studies of plants. He also attempted to test condition (i) quantitatively, using data from his long-term demographic study of two closely related species of Lobelia on Mount Kenya, one of which is semelparous and the other iteroparous, and which differ in habitat. The demographic parameters of the iteroparous species in the habitat to which it is adapted seem to fit the requirement for iteroparity to be stable to invasion by semelparity (Young, 1990).

There is little direct evidence on the shape of the curve relating survival to reproductive effort which would allow any firm conclusion concerning the importance of condition (iii). Some indirect evidence that semelparous species may have life-styles which constrain the trade-off curve to be convex was discussed in section 5.3.4.

Roff (1991, 1992, pp. 287–90) has reviewed the available evidence on the possible role of environmental variability in influencing the evolution of iteroparity versus semelparity (condition (ii)), and concludes that there is no firm evidence for a correlation between iteroparity and temporal variability.
5.4.5 Age at maturity

Again, we have discussed several conditions under which delayed reproduction can be favoured, despite the decline with age in sensitivity of fitness to changes in fecundity (sections 5.3.3–5.3.5). These are: (i) an increase in reproductive success with increasing body size, with a trade-off between growth and reproduction; (ii) temporal or spatial variation in fecundity or juvenile survival; (iii) a trade-off between reproductive effort and subsequent survival or growth; and (iv) benefits to close kin from deferral of reproduction.

Phenotypic evidence for trade-offs

There is a mass of data supporting the existence of a phenotypic correlation between size and reproductive success, in a variety of different groups of plants and animals, including mammals (Roff, 1992, pp. 126–8). Such a trade-off is required for condition (i) to be effective. There is also evidence for negative phenotypic correlations between reproduction and growth, particularly in plants and fishes (Roff, 1992, pp. 154–5). This type of trade-off is required under both conditions (i) and (iii).

For reproductive effort to be effective in setting the age at maturity (condition (iii)), it is necessary that the sensitivity of survival or growth to increased fecundity be greater for younger individuals (equation (5.37)). Such a pattern can be achieved in a number of different ways. In Drosophila, a minimum body weight must be obtained for pupation to be successful (Bakker, 1959); hence the speed of development cannot be advanced beyond a point which prevents this weight from being attained at the time of pupation. Early reproduction may place a greater strain on younger individuals, leading to lower survival and growth rates for a given level of realised reproductive success than for older individuals (Lack, 1954, Chapter 6), but there appears to be only a small amount of direct evidence for this (Promislow and Harvey, 1990). In birds and mammals, increased experience in parental care and food gathering may mean that a higher fecundity can be achieved by older individuals for a given level of survival or growth. There is good evidence for an increase with age in foraging skills in a number of bird species (Brown, 1987, p. 66), and for an increase in breeding success with previous experience (Clutton-Brock, 1988, Chapters 12, 13, 16, 17). In humans and some other mammals, infant mortality is higher for very young mothers than for somewhat older mothers (Stearns, 1992, p. 130; Promislow and Harvey, 1990, p. 428), which suggests that their realised reproductive success for a given level of reproductive effort may be lower.
Genetic evidence

Genetic correlations between body size and female fecundity have been demonstrated in *Drosophila* (Robertson, 1957). There is also evidence for a genetic correlation between body size and male mating success (Wilkinson, 1987). Such correlations are clearly required for the validity of model (i).

There is a wealth of quantitative genetic data from a diverse range of species demonstrating the existence of heritable variation in age at maturity (references in Roff and Mousseau, 1987; Mousseau and Roff, 1987). For example, the capacity of *Drosophila* populations to respond to a pressure of artificial selection on age at maturity has been repeatedly demonstrated (e.g. Clarke, Maynard Smith and Sondhi, 1961). There is also genetic evidence for trade-offs between growth and reproduction, mostly from fishes (Reznick, 1983) and plants (Law, 1979a; Geber, 1990), as required by conditions (i) and (iii).

In fish populations exposed to intensive commercial fishing, the high mortality involved has been shown to lead to the evolution of earlier maturity (Law and Grey, 1989). Edley and Law (1988) applied size-specific artificial selection to laboratory populations of *Daphnia magna*, mimicking predation pressures. Populations where only large individuals were removed showed increased rates of growth in body size early in life, accompanied by delayed maturity, compared with populations in which only small individuals were removed. Most of the response appeared to be due to changes in the populations where large individuals were removed. The increase in early fecundity in these populations (at ages earlier than those where the increase in mortality was applied) is in accord with the theoretical expectations under the reproductive effort model (p. 221).

A particularly informative experiment on the consequences of altering the demographic environment has been done by Reznick, Bryga and Endler (1990) on the guppy, *Poecilia reticulata*, which links evidence on between-population differences to the effects of experimentally-induced changes in demography. Previous work had shown that populations of guppies subject to different predation regimes exhibit genetic differences in their life-histories which correspond with theoretical expectation (Reznick, 1982). Fish from sites where the main predator is a species which prefers adult guppies reproduce earlier than fish from sites where the main predator prefers juveniles. This is consistent with the hypothesis that the predator's preference for adult guppies causes higher mortality around the age at maturity, favouring earlier reproduction than when juveniles are preferred, assuming that density-dependent juvenile mortal-
ity compensates for mortality due to increased predation on juveniles. Either conditions (i) or (iii) above could be involved in this response. Guppies were taken from the first type of environment and released into a site where the predator preferred juveniles. After eleven years (30–60 generations), the laboratory-reared grandchildren of fish taken from the original and transplanted populations showed differences which indicated that the transplanted stock had indeed undergone evolutionary change of the expected kind, i.e. it had evolved later reproduction.

**Comparative evidence**

A general prediction from models of type (i) concerns the relation between age at maturity and age at cessation of growth. As shown in section 5.3.3, these models for the evolution of the age of reproductive maturity predict that it should never follow the age of cessation of growth, a point which was first stressed by Gadgil and Bossert (1970). The prediction seems to be fairly well borne out by the evidence; although there are many species in which growth is prolonged past sexual maturity, there are only a few examples where growth ceases much before maturity. These will be discussed below. As pointed out by Gadgil and Bossert, this type of model would lead one to expect the times of sexual maturity and cessation of growth to coincide most closely in species where growth is rapid and then stops abruptly, as in holometabolous insects, since the maximum of $\hat{l}(x)$ will then be close to the point at which $G(x)$ levels off (Figure 5.2). In species where slow growth continues for a long time, on the other hand, we would expect sexual maturity to be considerably in advance of the cessation of growth, as in cold-blooded vertebrates.

A strategy which has been widely employed to test models based on condition (i) is to compare their predictions with data on closely related species or different populations of the same species which have been exposed to different demographic conditions, and whose ages at maturity can be assumed to have evolved in response to these conditions. For example, if the model underlying equation (5.22) is correct, the age at maturity for a given population can be determined knowing only the growth coefficient $\kappa$ and the mortality rate $\mu$ for the population in question, both of which can be estimated from data which are independent of the age at maturity. Agreement between predicted and observed values of age at maturity is taken as confirmation of the model.

The results of attempts to apply this approach to a number of different models have been reviewed by Roff (1992, pp. 183–207). As he points out, there is often impressive agreement between observed and predicted values, but this may not mean very much, as alternative models may
generate rather similar predictions. As has been repeatedly stressed in this chapter, high mortality favours early reproduction under almost any life-history model. Comparative studies which conform with this prediction therefore validate the general principles involved, but do not necessarily shed much light on the details of the nature of the costs of early maturity which must be involved.

There are now comparative studies of age at maturity in a variety of taxonomic groups which demonstrate the expected relation between mortality and age at maturity, e.g. in fish (Charnov and Berrigan, 1991a, b), snakes and lizards (Shine and Charnov, 1992), birds (Saether, 1988), and mammals (Promislow and Harvey, 1990, 1991). The most recent studies on mammals adjust for the effects of both body size and phylogeny on the association between mortality and age at maturity, leaving little doubt that we are seeing evolutionary responses to differences in demographic conditions (Promislow and Harvey, 1991; Harvey and Keymer, 1991).

Charnov and Berrigan (1991a) and Charnov (1993, Chapter 5) have attempted to carry the analysis of the relation between mortality and age at maturity in mammals further by deriving their phenomenological equation for predicting the ESS age at maturity (equation 5.23c) from a model of growth. This model assumes that the growth of juvenile mammals follows a well-known empirically-derived differential equation, \( \frac{dW}{dx} = Ax^{0.75} \), where \( A \) is a constant of proportionality. This leads to the relation between weight \( W \) and age for a possible age at maturity \( x \)

\[
W(x)^{0.25} = \frac{0.25A}{(1 - \delta^{0.25})} 
\]  

(5.51)

where \( \delta \) is the ratio of weight at maturity to weight at weaning.

It is also assumed that growth ceases abruptly at maturity and that fecundity is constant through adult life and is proportional to the rate of increase in weight at maturity. The biological justification for the last assumption is that resources previously used for growth are now devoted to reproduction (Kozlowski and Wiegert, 1986, 1987; Kozlowski, 1992). Using these assumptions in conjunction with the growth equation, it easily shown that equation (5.23c) becomes

\[
\mu = AW^{0.75} 
\]  

(5.52)

Combining equations (5.51) and (5.52), we obtain the relation

\[
b\mu = 3(1 - \delta^{0.25}) 
\]  

(5.53)

The result of plotting estimated values of \( b\mu \) against \( \delta \) for 23 species of mammals gives good agreement with the curve predicted by equation
Tests of theories of life-history evolution

(5.53) (Charnov and Berrigan, 1991a, Charnov, 1993, Chapter 5), suggesting that the model is well-founded. The fact that equation (5.52) predicts the observed allometric relation between mortality and adult body size (Charnov, 1993, Chapter 5) seems at first sight very satisfactory, but is in reality a source of concern. This equation is a consequence of the assumption that body size evolves in response to the mortality rate; in fact, many other ecological factors must affect the selective significance of body size (LaBarbera, 1986, Roff, 1992, pp. 120–2), rendering this assumption suspect. As with the other tests of quantitative agreement between observed and predicted age at maturity, the possibility that somewhat different causes may produce the same effect should be borne in mind in this case.

Cooperative breeding

As discussed in section 5.3.5, there is evidence from a number of species of birds and mammals that deferred reproduction is sometimes associated with helping close kin raise their young (condition (iv) above). Equations (5.40) and (5.41) show that such deferral of reproduction is most likely to be favoured by kin selection when there is a close relationship between donors and recipients, and when the recipients receive a large benefit from the aid. The highest relatedness value is achieved with monogamy, when both parents receive aid from their offspring. This requires high survival of breeding individuals from year to year (equation (5.41)).

A large number of empirical studies on cooperative breeding have now been carried out on birds, and a smaller number on mammals (Brown, 1987; Stacey and Koenig, 1990; Emlen et al., 1991). The general conclusion from these studies is that there is indeed good evidence in most cases for close genetic relationships between helpers and helped, with parent and offspring being the most common relationship (Brown, 1987, Chapter 12; Stacey and Koenig, 1990; Emlen et al., 1991). In cases where there is the opportunity to assist a variety of different individuals, donors seem to aid close relatives preferentially (Curry and Grant, 1990; Emlen, 1990; Emlen et al., 1991; Emlen and Wrege, 1994). There is also evidence in many cases that the reproductive success of recipients is indeed enhanced by aid (Brown, 1987, Chapter 11; Smith, 1990; Emlen and Wrege, 1994), although it is not always possible to exclude the possibility that the apparent effect of aid is due to confounding environmental effects, such as aid accruing preferentially to breeders in high-quality territories (Woolfenden and Fitzpatrick, 1990). Cooperatively breeding species also seem to have high survival rates (Brown, 1987, pp. 41–3; Smith, 1990), as
Evolution of life-histories is required for a substantial effect of kin selection. In some cases, the effect of donors on the reproductive success of closely-related recipients seems to be sufficiently large that kin selection alone could have been effective in promoting delayed breeding (Brown, 1987, Chapter 13).

These facts do not, however, constitute proof that deferral of reproduction evolved by kin selection in most cases. As we have seen, high adult survival is also favourable to deferral of reproduction under individual selection, especially if early breeding entails a greater risk of death (e.g. because of the greater hazards associated with dispersal for inexperienced individuals). Furthermore, as already noted there are considerable uncertainties associated with the estimates of the benefits of aid from non-breeders. In addition, recent behavioural and genetic studies have called into question the extent to which apparently monogamous species are truly monogamous, so that relatedness values may be lower than the estimates from purely behavioural studies (Quinn et al., 1987; Birkhead and Møller, 1992). There is presently no consensus on the extent to which kin selection as opposed to individual selection has promoted the evolution of deferred breeding, although it seems probable that the giving of aid by non-breeders is a product of kin selection (Brown, 1987, pp. 41–3; Smith, 1990, p. 605; Emlen et al., 1991). The key life-history question is whether deferred breeding in these cases would have evolved if there were no donation of aid by non-breeders, and this has proved hard to answer conclusively. Even if, as seems probable, kin selection has in fact facilitated the evolution of deferred breeding (Emlen and Wrege, 1994), it should be remembered that the age at maturity in most species with helping at the nest is usually only delayed by a year or two (Brown, 1987; Stacey and Koenig, 1990). This is relatively modest delay, compared with many cases in which individual selection has caused delays of several years, as discussed next.

An example of postponed reproduction

A life-history which, at first sight, presents several problems for the theories discussed above is exhibited by island-breeding sea-birds which feed offshore, particularly tropical species. As discussed by many authors (Lack, 1954, Chapter 6, 1966, Chapter 16; Wynne-Edwards, 1955, 1962, Chapter 5; Ashmole, 1963; Goodman, 1974), these birds show a combination of traits associated with apparently low reproductive effort, including postponement of reproduction for several years, a clutch size of one, prolonged incubation and nestling periods, frequent abandonment of the nest by parents, and long intervals between successive clutches. The extreme example is provided by the royal and wandering
albatrosses, *Diomedea epomophora* and *D. exulans*, which reach sexual maturity at about 10 years of age, have an incubation period of around 80 days, and a nestling period of about 250 days; since it takes over a year to rear a chick successfully, breeding occurs only once every two years (Lack, 1966; Goodman, 1974). These features have been interpreted by Wynne-Edwards (1955, 1962) as evidence that the life-history has evolved by group selection in favour of a low reproductive potential, in order to avoid exhaustion of the food supply and consequent extinction of the population. As mentioned on p. 190, there are strong theoretical objections to group selection, and we shall now consider some alternative interpretations of this form of life-history, which have been proposed by Ashmole (1963), Lack (1966) and Goodman (1974).

In the first place, the small clutch size of these species is understandable in terms of the difficulty they have in feeding their young; since the birds are large in size and restricted to breeding on islands, many adults will be competing for food within foraging range, so that strong density-dependent mortality of nestlings would be expected. Evidence for such mortality is described by Ashmole (1963), Lack (1966) and Nelson (1969). Experimental addition of an extra chick to nests of the Laysan albatross, *D. immutabilis*, increases mortality due to starvation, to such an extent that there is a considerable net decrease in the number of chicks raised per pair (Rice and Kenyon, 1962). There is thus little difficulty in accounting for the small clutch size. The prolonged nestling period can be interpreted in terms of slow growth of the young, evolved in response to the limited food supply (Ashmole, 1963; Lack, 1966; but see Shea and Ricklefs, 1985). The long incubation period of the eggs is more difficult to explain. As suggested by Lack, it may be impossible to prolong the nestling period except by a general slowing-down of growth, which also retards the development of the egg. The main question of interest here is the reason for the prolonged period of sexual immaturity, despite the fact that growth is essentially completed before the birds can fly (Ashmole, 1963). There is good evidence for a very low annual mortality of full-grown birds (5–10%), presumably due to the virtual absence of predation (Goodman, 1974). The intensity of selection in favour of early breeding is therefore low. Because of the early completion of growth, any explanation based on the ideas discussed in section 5.3.4 has to be couched in terms of an increase in potential reproductive efficiency with age. As suggested by Ashmole (1963), it is probable that older birds are more efficient than younger ones in foraging for food, and that the more time the birds have to acquire skill in hunting the greater their chance of raising a chick successfully. In view of the high mortality of chicks by
starvation, there could well be a threshold level of skill needed for reproductive success, below which no chick can be reared.

Evidence for increased reproductive efficiency with age has already been discussed, and it is an attractive explanation for postponed reproduction. For example, in the Atlantic fulmar both sexes show increased breeding success for ten years after first breeding (Ollason and Dunet, 1988). This is a seabird in which males and females have a modal age of first reproduction of eight and twelve years respectively, with some females postponing breeding until 19, and there is a high (97%) probability of annual survival. There is also evidence that early breeding reduces the survival of females, consistent with the model, e.g. fecundity in the second and third years has a correlation coefficient of $-0.28$ with lifespan.

The hypotheses discussed above can be examined quantitatively by means of the reproductive effort model (cf. Goodman, 1974). We can use equations (5.36) for the adult age-classes, since the data show that adult survival and fecundity are approximately constant. We may take $\tilde{P}$ to be equal to the annual adult survival probability, $P$, as there is no growth of adults. Since parental care is an important component of reproductive effort, $\bar{m}$ is taken to be one-half the average number of chicks reared per nest. The populations are probably regulated by density-dependent nestling mortality, so that $\bar{m}$ is density-dependent and $r$ is zero. Equations (5.36) thus give us the optimal life-history equation

$$\frac{\partial P}{\partial \bar{m}} = \frac{(1 - P)}{\bar{m}} = l(b)$$

Taking $1 - P$ as equal to 0.05, an increase in $\bar{m}$ of, say, 10% of its value for the optimal life-history need only lead to a reduction in adult survival of about 0.5% in order for this relation to be satisfied. The powerful effect of high adult survival in promoting low reproductive effort is emphasised by this fact.

We must also consider the effects of increases in fecundity on survival among the juvenile age-classes, since an optimal life-history must be stable to perturbations such that reproduction occurs earlier in life. In the present case, condition (5.37) becomes

$$\frac{\partial P_x}{\partial \bar{m}_x} > l(x + 1) \quad (1 \leq x < b, \bar{m}_x = 0)$$

Since $l(x) \leq 1$, and decreases with age, a sufficient condition for this condition to be satisfied is that the slope of survival plotted against $\bar{m}_x$ be greater than one at the origin, for all the juvenile age-classes. The
postulated increase in breeding skill with age can be incorporated into this model in the form of a decrease with age in the sensitivity of $P_x$ to higher $\bar{m}_x$; the sensitivity is measured by the value of $|\partial P_x/\partial \bar{m}_x|$ for a fixed $\bar{m}_x$. This is because a greater degree of reproductive effort will be needed to attain a given value of $\bar{m}_x$, the lower the hunting ability of the parents. Since the survival probability of full-grown birds is high, only a modest increase in breeding skill with age is necessary to favour postponed reproduction, as may be seen by comparing equations (5.54) and (5.55).

**Variation in survival or fecundity**

There is little evidence on whether temporal variation in fecundity or juvenile survival tends to promote the evolution of delayed reproduction (section 5.3.3). There is some evidence for facultative adjustments of the life-history to local environmental conditions of the kinds expected theoretically. For example, in *Drosophila*, poor nutrition leads to slower growth and later age at maturity (Gebhardt and Stearns, 1988). Similar results have been observed in guppies (Reznick, 1990). Whether this pattern is an evolved response, or is simply an unavoidable response to harsh conditions, is an open question (cf. Roff, 1992, pp. 229–32).

### 5.4.6 Reproductive effort

The reproductive effort model of section 5.3.4, which postulates a trade-off between present reproduction and future survival or growth, has probably been the subject of more investigation and debate than any other part of life-history theory. Here, we will be concerned with the question of the extent to which there is empirical evidence that the characteristic level of age-specific reproduction of a species is often controlled by this type of constraint.

**Phenotypic evidence**

There are two categories of purely phenotypic evidence concerning trade-offs between reproductive effort and survival or growth: (i) the observation of a negative correlation between reproduction and survival or growth in an unmanipulated population; and (ii) the detection of a reduction in growth or survival in individuals whose reproductive effort has been artificially increased (or the converse if reproductive effort is decreased). The results of both kinds of studies have recently been reviewed by Bell and Koufopanou (1985), Partridge and Sibly (1991), Reznick (1992), Roff (1992, Chapter 8) and Stearns (1992, Chapter 4).

It is clear that, while studies of type (i) often reveal the expected
negative correlations, they frequently do not: non-significant negative correlations or even significant positive phenotypic correlations have been reported in a variety of species (see Appendix 1 of Stearns, 1992). There is a plausible explanation for this, which is consistent with the existence of an underlying trade-off. If there is variation in the quality of the environment to which individuals have been exposed, individuals raised in good environments may have high fecundity and high survival/growth, compared with individuals raised in poor conditions (Charlesworth, 1984; van Noordwijk and de Jong, 1986), obscuring any trade-off which might be observed with individuals raised under constant conditions. Differences between individuals subject to natural phenotypic variation in reproductive effort of known cause are not so subject to this difficulty. For example, the phenomenon of mast-seeding in plants, where a season of intense seed-set in a perennial is followed by a one or more seasons of reduced seed set and reduced growth (Harper and White, 1974), is indicative of an adverse effect of reproductive effort. Similarly, in red deer, nursing mothers have higher mortality and lower fecundity in the next season than non-reproductive females of comparable age (Clutton-Brock, Guinness and Albon, 1983). Mating has been shown to incur a risk of reduced survival to both males and females in many species of arthropods and vertebrates (Stearns, 1992, Appendix 2). For example, the mating calls of some species of male neo-tropical frogs attract predatory bats, so that they suffer a higher risk of predation (Tuttle and Ryan, 1981).

Manipulation experiments are also not open to the problem of artefactual positive correlations, and can provide valuable information on the physiological basis for trade-offs. For example, Maynard Smith (1958) showed that sterilisation of *Drosophila subobscura* females by radiation or by the maternally-acting mutant gene *grandchildless* caused increased adult survival. In the blue tit, Nur (1988) has shown that an artificial increase in clutch size leads to lower survival and subsequent reproduction of females. The review in Appendix 2 of Stearns (1992) shows a general, but not universal, tendency for effects of this kind.

**Genetic evidence**

Again, there are two categories of evidence: (i) evidence for negative additive genetic covariances or correlations between reproductive effort and survival or growth; and (ii) responses of the life-history to natural or artificial changes in the demographic environment of a population which agree with the predictions of the reproductive effort model. Ideally, both types of evidence should be obtained for the same
system, but this has only been done a few times (see below for some examples).

There are, however, difficulties with both approaches. Genetic covariances and correlations have high sampling errors, and satisfactory estimates can only be obtained in breeding experiments where large numbers of individuals are measured (Falconer, 1989, Chapter 19). If accurate estimates can be obtained, negative genetic correlations may not be observed even if there are trade-offs at the genetic level, for two reasons. First, as pointed out in section 4.5.3, some positive genetic correlations between pairs of constrained variables in a multivariate system are inevitable, even in a population at equilibrium under selection alone. Large negative genetic correlations may only be observed between pairs of variables which both exhibit a trade-off between each other and are virtually independent of all other variables (see Table 1 of Charlesworth, 1990a). Second, by analogy with the above argument concerning artefactual positive environmental correlations, there may be a substantial component of genetic variation in (for example) survival and fecundity which reflects the overall ability of the organism to harness resources for both components of fitness, rather than the partitioning of resources between them. Variation arising from deleterious alleles maintained in the population by mutation may be an important contributor to this, as there is evidence that such mutations frequently affect multiple life-history traits adversely (Charlesworth, 1990a; Houle, 1991; Houle et al., in preparation). In addition, if the population has recently been exposed to a novel environment, there may be genotypic variation for adaptation to this environment, involving similar effects on multiple fitness components (Service and Rose, 1985).

For these reasons, measurements of genetic correlations should be carried out on populations which have been allowed to adapt to the experimental environment for several generations; inbred stocks should also be avoided, since these are more likely to reveal the effects of recessive or partially recessive deleterious mutations (Charlesworth, 1990a; Houle, 1991). Service and Rose (1985) have demonstrated how a large negative genetic correlation (between early fecundity and a correlate of survival in Drosophila melanogaster) can be significantly reduced if the population is raised in a novel environment. Similarly, Rose (1984) showed that inbreeding can change this genetic correlation from negative to positive.

Estimates of genetic correlations from artificial selection experiments may be problematical for additional reasons, unless they are well-replicated and the selection lines were maintained with large population sizes.
This is because genetic drift in small populations can cause substantial genetic correlations due to linkage disequilibrium between loci controlling the traits, which may obscure genetic correlations present in the base population (Falconer, 1977, 1989, Chapter 19). In addition, the fixation of deleterious alleles by drift may cause a correlated decline in multiple life-history traits.

Experiments in which suitable precautions have not been taken when measuring genetic correlations by either method may account for some of the positive genetic correlations between fecundity and survival reported in the literature (see Appendix 1 of Stearns, 1992). The numerous failures to detect negative genetic correlations between reproductive effort and survival/growth have led some authors to question the utility of the genetic approach (Bell and Koufopanou, 1985; Pease and Bull, 1988; Partridge and Sibly, 1991). However, neither type of phenotypic evidence can provide a conclusive proof that natural selection is constrained by any trade-offs which they may suggest. For this reason, genetic data are required to test the evolutionary significance of such trade-offs (Charlesworth, 1990a; Reznick, 1992). As will be shown below, there are examples of properly designed genetic experiments which are encouraging for the view that there is an evolutionarily significant cost of reproductive effort.

(1) As has already been mentioned, there is evidence from breeding experiments for a large negative genetic correlation between early fecundity and survival in an old-established laboratory population of *D. melanogaster* (Rose and Charlesworth, 1981; Service and Rose, 1985). In agreement with this finding, and with the results of the manipulation experiments of Maynard Smith (1958) on *D. subobscura*, maintenance of stocks derived from this population by breeding from large numbers of old flies has led to a pronounced decline in early female fecundity, accompanied by increased longevity (section 5.4.3; Rose, 1991, Chapter 3).

(2) Law, Bradshaw and Putwain (1977) and Law (1979a) have carried out genetic studies on natural populations of the annual meadow grass, *Poa annua*, which parallel these *Drosophila* results, except that the alterations in the demographic environment to which responses were observed were natural rather than experimental.

Despite the species' name, the plants studied by these authors had a maximum life-span of about 18 months. Reproduction is initiated approximately 15 weeks after germination in winter, and continues through spring and summer of the first year. If the plant survives to the following year, a second period of reproduction in spring and summer precedes
death. These plants are largely self-fertilising, so that reproduction is effectively clonal. Life-history data were collected on plants raised from seed in standard conditions, and derived from 29 different natural populations. These could be divided into two broad categories of roughly equal size, 'opportunist' and 'pasture'. The opportunist populations were subject to a good deal of density-independent mortality due to disturbance, and the pasture populations were relatively stable communities, subject to regulation by density-dependent seedling mortality. Differences between the two classes of population, as well as between populations within each class, were detected in a number of different life-history variables. Fecundity, as measured by number of inflorescences per plant, was much higher in the second than the first year of life for the pasture populations. Opportunist populations mostly had higher fecundity in the first year than the pasture populations, and a more even distribution of reproduction between the two years. They also reproduced significantly earlier within the first year than did the pasture populations, and the plants were smaller. The opportunist populations had higher mortality than the pasture populations; this was concentrated in the period following reproduction in the first year, which suggests that it is due to the adverse effect of reproductive effort on subsequent survival. There is no doubt that the differences observed in this study are genetic, since the plants were raised under the same conditions. The differences are consistent with theoretical expectations for the differences between populations subject to density-dependent juvenile mortality, and populations subject to high age-independent and density-independent mortality. The observation that fecundity increases with age in the density-dependent pasture populations is particularly interesting, in view of the predictions of the reproductive effort model discussed on pp. 218–19.

Law (1979a) also carried out measurements of the mean values with respect to various life-history traits, for families produced by individual plants taken from both types of population. These can be used to infer genetic correlations. There was evidence for negative genetic correlations between early reproduction and subsequent survival, growth and reproduction, in agreement with the differences between populations just described.

(3) Reznick (1983) demonstrated genetically-determined trade-offs between reproduction and growth in laboratory comparisons of guppies taken from different locations in the wild. The results of Reznick (1982), discussed in connection with age at maturity in section 5.4.5, indicated that female fecundity early in life was significantly reduced in populations where there was little predation pressure on adults, compared with
Evolution of life-histories

populations where the dominant predator preferred adults. This difference was reproduced when the demographic environment was altered by transplanting fish from the first to the second type of environment, and allowing the population to evolve for many generations (Reznick et al., 1990). Again, the differences were shown to be genetic rather than environmental. While there is no direct evidence for a trade-off between reproduction and survival/growth in these data, the response of fecundity and offspring weight to changed demography is as predicted for reproductive effort (section 5.4.1), although their measure of reproductive effort did not show a statistically significant change.

Comparative evidence

One of the main conclusions of section 5.3.4 was that reproductive effort might often be expected to increase with age, under conditions summarised in section 5.4.1. Roff (1992, pp. 264–9) has discussed the evidence from studies of a variety of species as to whether this prediction is met. There are considerable difficulties in measuring reproductive effort, and use has frequently to be made of indices such as the ratio of weight of gonads to somatic parts, which are not always valid measures. Any conclusions must therefore be treated cautiously. Roff (1991) showed that in many fish species the sufficient condition for reproductive effort to increase with age of Charlesworth and León (1976) is met: the product of growth and survival for minimal reproductive effort, $\tilde{P}(0)$, exceeds one. In cod and plaice, there is good evidence that the proportion of available energy devoted to reproduction does increase with age (Roff, 1992, p. 264). A survey of 30 species of animal (including fourteen fish species) and one plant (a palm) using both energy budget and data on relative clutch mass showed that an increase in reproductive effort with age is observed in 27 cases (Roff, 1992, Table 8.4). An additional case in a bird species, the collared flycatcher, has recently been described (Pärt, Gustafsson and Moreno, 1992). In the majority of these cases, however, there is inadequate information to test the theoretical conditions quantitatively.

The expectation that the level of reproductive effort will be affected by the demographic environment (section 5.4.1) has been examined by species comparisons in both birds (Saether, 1988) and mammals (Promislow and Harvey, 1990, 1991). In both cases, the effects of body size were removed statistically, and phylogenetic corrections were applied in the case of the mammalian data. The major finding is that measures of reproductive output per unit time, such as litter or clutch size and gestation length (in mammals), are significantly correlated with adult
mortality rates, in the direction predicted by theory. Less satisfactorily, high juvenile mortality is also positively correlated with higher reproductive output in the mammalian data, in contradiction to theoretical expectation. As discussed by Promislow and Harvey (1990, 1991), this may be at least partly artefactual, since juvenile and adult mortality rates are highly correlated across species in their studies, so that there is little information on the effects of their relative values. In addition, juvenile mortality is often estimated from information on the age-structure and reproductive rates of older individuals.

A final difficulty is that the data used in these analyses were collected under field conditions. If populations are regulated by density-dependent fecundity or mortality, then a positive correlation between fecundity and mortality across species would be expected purely on the basis that the net reproduction rate must equal one (Sutherland, Grafen and Harvey, 1986). The observed patterns could thus be due solely to demographic rather than evolutionary causes. While it seems unlikely that the correlations for variables such as age at maturity and gestation length, which seem to be species-specific when studied in captive populations (Promislow and Harvey, 1991), could be explained in this way, it might apply to litter or clutch size. Measures of life-history traits for captive populations of the species used in these studies will help clarify this question.

5.4.7  *r*-selection and *K*-selection

Correlations of the type we have discussed between the demographic environment of a species and its life-history have been recognised in general terms for a long time, and led to the concept of *r*-selection versus *K*-selection (MacArthur and Wilson, 1967). *r*-selection is said to operate on species which are not usually subject to density-dependent regulation of population size, and which undergo frequent episodes of rapid population growth. It leads to the evolution of early maturity, high reproductive effort, and short life-span. *K*-selection acts on populations held in check by density-dependent factors, and leads to the evolution of delayed maturity, low reproductive effort, and long life-span.

While the distinction between these modes of selection has had considerable value, it should be remembered that the original formulations of *r*-selection and *K*-selection were in terms of models that largely neglected the factor of age-structure, and were based on the logistic growth equation described by equation (1.79) (MacArthur, 1962; MacArthur and Wilson, 1967). It is not very surprising that some important aspects of the impact of demographic factors on the outcome of selection are missing from the *r*-selection versus *K*-selection contrast. For instance, we have
seen in several contexts that the effects of density regulation by means of a mortality factor applied equally to each age-class are similar to the effects of a high rate of population growth (section 5.4.1). If we were to compare two populations which differed only in that one was held in check by this form of density dependence, and the other was density independent, no differences in the outcome of selection would be expected. On the other hand, population regulation by means of density-dependent juvenile survival or fecundity reduces the intensity of selection for early breeding, compared with a density-independent population. As in the cases discussed in section 3.4.1, differences in age-structure are the important factors. We have also seen that temporal fluctuations in survival rates and fecundities may have consequences for the form of the life-history which are not included within the compass of $r$-selection and $K$-selection (section 5.4.1). It would therefore seem wisest to investigate the effects of demographic factors on the outcome of life-history evolution by means of models which take account of details of the age-specificity of these factors, rather than by appealing to the blanket concept of $r$-selection and $K$-selection.

Similar views were expressed in the older literature by Wilbur, Tinkle and Collins (1974) and Mertz (1975). Recent surveys of life-history evolution have concurred with these criticisms (Mueller, 1991; Promislow and Harvey, 1990, 1991; Roff, 1992, pp. 44–6; Stearns, 1992, pp. 206–7). The approach of comparing the predicted and observed responses of individual life-history traits to specific changes in the demographic environment is widely accepted as being more fruitful.

5.5 Conclusions
The theories and facts reviewed in this chapter seem to show that we have a valid broad picture of the forces which control the evolution of age-specific patterns of survival and reproduction. There can be little doubt that senescence has evolved in response to the decline with age in the sensitivity of fitness to changes in survival and reproduction. Similarly, the empirical evidence is in good agreement with the notion that the evolution of the age at maturity, and of semelparity versus iteroparity, is controlled by a balance between the inherent advantage of early reproduction, and a cost of reproduction in terms of survival or growth. There is less strong, but still quite persuasive, evidence that the level of reproductive effort per breeding period is also modulated by a balance between immediate reproductive advantage and a cost to future survival and reproduction. It seems clear that differences in demographic environment, especially those due to differences in extrinsic sources of mortality,
have been responsible for observed evolutionary shifts in patterns of genetically controlled survival and reproduction. Similarly, there has been considerable success in interpreting patterns of age-specific allocation to different sex functions in sequential hermaphrodites.

Despite the proliferation of theoretical and empirical studies of life-history evolution in recent years, there are major gaps which need to be filled, as in any other thriving field. There is still a shortage of adequate experimental and comparative information on the response of the rate of ageing to different demographic environments, as opposed to studies of longevity. There is also uncertainty concerning whether or not differences in reproductive patterns between species are caused by density-dependent responses to different demographic environments rather than by evolutionary change. More comparative studies of captive populations, especially of within-species differences between populations which have been exposed in the wild to different demographic conditions, and more experimental studies of genetic responses to artificially induced changes in demography, would be helpful. There is very little evidence on whether or not variable environments have the expected consequences for life-history evolution, and suitable systems for investigating this remain to be explored. On the theoretical level, there is currently no treatment of the quantitative genetics of life-history evolution in variable environments, and the incorporation of the effects of mutational variation on the form of the life-history is just starting to be explored. These problems are a challenge for the next generation of research workers.
APPENDIX 1

Generating functions and their properties

The purpose of this appendix is to summarise the properties of generating functions used in this book. For a more detailed treatment, Feller (1968, Chapter 11) should be consulted.

Consider a finite or infinite sequence of real numbers \(a(0), a(1), a(2), \ldots\). The generating function of this sequence, \(\bar{a}(s)\), is defined by the function

\[
\bar{a}(s) = \sum_{j=0}^{\infty} s^j a(j)
\]

(A1.1)

where \(s\) is a real or complex variable; \(\bar{a}(s)\) is defined only for values of \(s\) for which the sum converges. For a finite sequence of \(n + 1\) numbers, the upper limit of the summation is taken as \(n\). If the sequence \(a\) defines a probability distribution, then \(\bar{a}(s)\) is called a probability generating function, for obvious reasons. It is easily seen that the mean of the distribution is given by \((d\bar{a}(s)/ds)_{s=1}\); more generally, \((d^r\bar{a}(s)/ds^r)_{s=1}\) gives the \(r\)th factorial moment of the distribution.

One very useful property of generating functions is exploited in the manipulation of the convolution, \(c\), of two sequences \(a\) and \(b\). The convolution is itself a sequence, whose \(j\)th element \(c(j)\) is defined by

\[
c(j) = \sum_{k=0}^{j} a(k)b(j - k)
\]

(A1.2a)

It is easily seen that the generating function of \(c\) is given by the following simple expression

\[
\bar{c} = \bar{a}\bar{b}
\]

(A1.2b)

Another useful property of generating functions arises from the fact that they can often be expanded into partial fractions. Suppose we have a generating function \(\bar{a}\) of the form

\[
\bar{a}(s) = \bar{b}(s)/\bar{d}(s)
\]

(A1.3)

where \(\bar{b}\) and \(\bar{d}\) are both polynomials in \(s\) without common zeros. Assume
for simplicity that \( \tilde{b} \) is of lower degree than \( \tilde{d} \), and that \( \tilde{d} \) has \( m \) distinct zeros, so that we can write

\[
\tilde{d}(s) = (s - s_1)(s - s_2) \ldots (s - s_m)
\]  

(A1.4)

It is known from algebra that in this case \( \tilde{a} \) can be expanded into partial fractions

\[
\tilde{a}(s) = \sum_{j=1}^{m} \frac{C_j}{(s_j - s)}
\]  

(A1.5)

where the \( C_j \) are constants. The value of \( C_1 \) can be found by multiplying both sides of equation (A1.5) by \( s_1 - s \), and letting \( s \) approach \( s_1 \). The product \((s_1 - s)\tilde{a}(s)\) approaches \( C_1 \) as \( s \to s_1 \); from equations (A1.3) and (A1.4), we have

\[
(s_1 - s)\tilde{a}(s) = \frac{-\tilde{b}(s)}{(s - s_1)(s - s_2) \ldots (s - s_m)}
\]

As \( s \to s_1 \), the numerator approaches \( \tilde{b}(s_1) \) and the denominator tends to \((s_1 - s_2)(s_1 - s_3) \ldots (s_1 - s_m)\), which is equal to \((d\tilde{d}/ds)_{s=s_1}\). A similar procedure applied to each zero in turn yields the general expression

\[
C_j = \frac{-\tilde{b}(s_j)}{(d\tilde{d}/ds)_{s=s_j}}
\]  

(A1.6)

We also have

\[
\frac{1}{(s_j - s)} = \frac{1}{s_j(1 - s/s_j)} = \frac{1}{s_j} \left[ 1 + \frac{s}{s_j} + \left(\frac{s}{s_j}\right)^2 + \left(\frac{s}{s_j}\right)^3 + \ldots \right]
\]

provided that \(|s| < |s_j|\). Substituting from this into equation (A1.5) and equating coefficients of \( s \) on both sides, we obtain the coefficient of \( s^k \)

\[
a(k) = \sum_{j=1}^{m} \frac{C_j}{s_j^{k+1}}
\]  

(A1.7)

If \( s_1 \) is a zero of \( \tilde{d} \) which is smaller in modulus than any other, then as \( k \to \infty \), we obtain the asymptotic expression

\[
a(k) \sim \frac{C_1}{s_1^{k+1}}
\]  

(A1.8)

Relatively trivial modifications are required to deal with cases when \( \tilde{b} \) is of the same or greater degree than \( \tilde{d} \), or when \( \tilde{d} \) has repeated zeros. The asymptotic expression (A1.8) still holds in these cases, provided that \( s_1 \) is a unique zero; a related expression can be obtained even when \( s_1 \) is a repeated zero (Feller, 1968, p. 285).
APPENDIX 2

Asymptotic values of $\Delta p_i$ and $\Delta^2 p_i$

Equation (4.1) gives the number of births at time $t$. It can be written as

$$B(t) = \sum_{x} B(t - x)[k_{s}(x) + O(\varepsilon)]$$

(A2.1)

where the term $O(\varepsilon)$ depends on the gene frequencies at time $t - x$ and on the genotypic differences in the reproductive functions. It follows that it must be possible to choose two positive numbers of order $\varepsilon$, $\varepsilon_1$ and $\varepsilon_2$, such that $k_{s}(x) - \varepsilon_1 < k_{s}(x) + O(\varepsilon) < k_{s}(x) + \varepsilon_2$, for all values of $t$. It is easily seen that

$$\sum_{x} B_1(t - x)[k_{s}(x) - \varepsilon_1] < B(t) < \sum_{x} B_2(t - x)[k_{s}(x) + \varepsilon_2]$$

(A2.2)

where $B_1(t)$ and $B_2(t)$ are the solutions to the renewal equations associated with the reproductive functions $k_{s}(x) - \varepsilon_1$ and $k_{s}(x) + \varepsilon_2$ (the initial conditions for $B_1$ and $B_2$ are assumed to be the same as for $B$, $B(0)$, $B(1), \ldots, B(d-1)$).

Using the standard solution to the renewal equation, given by (1.39), for sufficiently large $t$ we have

$$B_1(t) \sim C_1 e^{(r - \eta_1)t} < B(t) < B_2(t) \sim C_2 e^{(r + \eta_2)t}$$

(A2.3)

where $r - \eta_1$ and $r + \eta_2$ are the intrinsic rates of increases associated with the reproductive functions $k_{s}(x) - \varepsilon_1$ and $k_{s}(x) + \varepsilon_2$. Both $\eta_1$ and $\eta_2$ are $O(\varepsilon)$, as is $C_1 - C_2$. Equation (A2.3) implies that

$$\Delta \ln B \sim r_{s} + O(\varepsilon)$$

(A2.4)

Exactly the same argument can be applied to equation (4.5). We have

$$B(t)p_i(t) = \sum_{x} B(t - x)p_i(t - x)[k_{s}(x) + O(\varepsilon)]$$

(A2.5)

As in the previous case, positive numbers of order $\varepsilon$, $\varepsilon_3$ and $\varepsilon_4$, can be
assigned, such that (for sufficiently large $t$) we have
\[ C_3 e^{(r_3-\eta_3)t} < B(t) p_i(t) < C_4 e^{(r_3-\eta_4)t} \] (A2.6)
where $\eta_3$ and $\eta_4$ are defined analogously to $\eta_1$ and $\eta_2$, and are $O(\varepsilon)$. We therefore obtain the asymptotic expression
\[ \Delta \ln Bp_i \sim r_i + O(\varepsilon) \] (A2.7)
Subtracting equation (A2.4) from equation (A2.7), we find that
\[ \Delta \ln p_i \sim O(\varepsilon) \] (A2.8a)
from which we obtain
\[ \Delta p_i \sim O(\varepsilon) \] (A2.8b)

Having established that $\Delta p_i$ is asymptotically of order $\varepsilon$, we can now consider the magnitude of $\Delta^2 p_i$, the second difference in $p_i$. The $O(\varepsilon)$ term in equation (A2.8b) for a given time $t$ is a complicated function, $\phi_i$, say, of the sequences of $O(\varepsilon)$ terms in equations (A2.1) and (A2.5) at successive times $t$, $t-1$, $\ldots$. Similarly, $\Delta p_i$ for the preceding time $t-1$ is a function, $\phi_i^{t-1}$, of the $O(\varepsilon)$ terms for $t-1$, $t-2$, $\ldots$. At time $t-1$ we have $\Delta^2 p_i = \phi_i - \phi_i^{t-1}$, which is of the same order as the first differences in the $O(\varepsilon)$ terms in equations (A2.1) and (A2.5). These terms involve only the products of gene frequencies and constant $O(\varepsilon)$ terms given by the genotypic differences in reproductive functions. Without loss of generality, we can assume that $t$ is taken sufficiently large that the gene frequency terms are changing by $O(\varepsilon)$, so that the first differences in the $O(\varepsilon)$ terms are $O(\varepsilon^2)$. This establishes that $\Delta^2 p_i$ is asymptotically $O(\varepsilon^2)$.

This result can be used to establish the approximation for $p_i(t)$ in terms of $p_i(t-x)$ and $\Delta p_i(t)$, used on p. 139. We have
\[ p_i(t) = p_i(t-x) + \sum_{y=1}^{x} \Delta p_i(t-y) \] (A2.9)
Furthermore, for $y \geq 1$ we have (writing $\Delta^2 p_i(t) = \Delta p_i(t+1) - \Delta p_i(t)$)
\[ \Delta p(t) = \Delta p(t-y) + \sum_{u=1}^{y} \Delta^2 p(t-u) \] (A2.10)
But $\Delta^2 p_i$ is asymptotically $O(\varepsilon^2)$, so that, combining equations (A2.9) and (A2.10), we obtain
\[ p_i(t) \sim p_i(t-x) + x\Delta p_i(t) + O(\varepsilon^2) \] (A2.11)

These asymptotic results are based on the properties of the renewal equation. The period of time needed for convergence to the asymptotic
solution of this equation is of the order of a few ‘generations’ in biological applications, as discussed in section 1.3.2. The above results should, therefore, be valid after a relatively short time has elapsed.

The case of a density-dependent population can be analysed in a similar way, using the assumption of section 4.2.2 that the population growth rate tends to $O(\varepsilon)$ and the number of individuals in the critical age-group, $N$, tends to the equilibrium number for a standard genotype, plus $O(\varepsilon)$ terms. By the argument leading to equations (A2.4) and (A2.8), this implies that, to order $O(\varepsilon)$, $\Delta \ln B(t)$ and $\Delta \ln B(t)p_i(t)$ also tend to zero. This implies that equation (A2.8) applies to this case also.

Using this result, equations (4.21) and (4.22) of the text can be immediately derived, by the same argument that led to equation (4.4). These imply that the changes in the $O(\varepsilon)$ terms in equation (A2.8b) are again $O(\varepsilon^2)$, since they can be partitioned into a component which reflects the effect of changes in $N$ on the reproductive function of the standard genotype, and components reflecting the products of changes in genotype frequencies and deviations of the corresponding genotypic reproductive functions from the reproductive function of the standard genotype. By equations (4.22), the first term must be $O(\varepsilon^2)$; from (A2.8b), the other terms must also be $O(\varepsilon^2)$. 

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References


References


References


References


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References


References

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References


References


References


References


<table>
<thead>
<tr>
<th>Author</th>
<th>Pages</th>
</tr>
</thead>
<tbody>
<tr>
<td>Abrams, P. A.</td>
<td>176, 185, 193, 195, 196</td>
</tr>
<tr>
<td>Abugov, R.</td>
<td>105, 107, 145, 224</td>
</tr>
<tr>
<td>Albon, S.</td>
<td>260</td>
</tr>
<tr>
<td>Antonovics, J.</td>
<td>50</td>
</tr>
<tr>
<td>Ashmole, N. P.</td>
<td>256, 257</td>
</tr>
<tr>
<td>Aspinwall, N.</td>
<td>19, 70</td>
</tr>
<tr>
<td>Assimacopulos, S.</td>
<td>244, 245, 261</td>
</tr>
<tr>
<td>Auslander, D.</td>
<td>58</td>
</tr>
<tr>
<td>Austad, S. N.</td>
<td>246, 247, 248</td>
</tr>
<tr>
<td>Bakker, K.</td>
<td>251</td>
</tr>
<tr>
<td>Barkalow, F. S.</td>
<td>20</td>
</tr>
<tr>
<td>Barrai, I.</td>
<td>96, 103</td>
</tr>
<tr>
<td>Barton, N. H.</td>
<td>245</td>
</tr>
<tr>
<td>Beardmore, J. A.</td>
<td>183</td>
</tr>
<tr>
<td>Beckingham, L.</td>
<td>54</td>
</tr>
<tr>
<td>Beddington, J. R.</td>
<td>54, 57, 58</td>
</tr>
<tr>
<td>Bell, G.</td>
<td>206, 247, 259, 262</td>
</tr>
<tr>
<td>Bernardelli, H.</td>
<td>5</td>
</tr>
<tr>
<td>Berrigan, D.</td>
<td>209, 210, 254, 255</td>
</tr>
<tr>
<td>Birkenhead, T. R.</td>
<td>256</td>
</tr>
<tr>
<td>Bodmer, W. F.</td>
<td>107, 129, 132, 133, 134</td>
</tr>
<tr>
<td>Bossert, W. H.</td>
<td>185, 213, 215, 216, 217, 218, 220, 222, 253</td>
</tr>
<tr>
<td>Bradshaw, A. D.</td>
<td>262</td>
</tr>
<tr>
<td>Brown, J. L.</td>
<td>206, 223, 224, 226, 251, 255, 256</td>
</tr>
<tr>
<td>Brown, R. J.</td>
<td>228</td>
</tr>
<tr>
<td>Broyles, S. B.</td>
<td>217</td>
</tr>
<tr>
<td>Bryga, H.</td>
<td>252</td>
</tr>
<tr>
<td>Bull, J. I.</td>
<td>185, 262</td>
</tr>
<tr>
<td>Bulmer, M. G.</td>
<td>18, 19, 27, 48, 173, 204, 223, 250</td>
</tr>
<tr>
<td>Buzzati-Traverso, A. A.</td>
<td>51, 183</td>
</tr>
<tr>
<td>Calder, W. A.</td>
<td>248</td>
</tr>
<tr>
<td>Calow, P.</td>
<td>185, 187, 250</td>
</tr>
<tr>
<td>Canada, D.</td>
<td>245, 261</td>
</tr>
<tr>
<td>Caswell, H.</td>
<td>2, 11, 15, 22, 39, 191, 192, 237</td>
</tr>
<tr>
<td>Caughley, G.</td>
<td>15, 18, 21, 247</td>
</tr>
<tr>
<td>Cavalli-Sforza, L. L.</td>
<td>96, 103, 132, 133, 134, 224</td>
</tr>
<tr>
<td>Chandler, J. H.</td>
<td>133</td>
</tr>
<tr>
<td>Chitty, D.</td>
<td>180</td>
</tr>
<tr>
<td>Choy, S. C.</td>
<td>63, 85, 91</td>
</tr>
<tr>
<td>Christiansen, F. B.</td>
<td>158</td>
</tr>
<tr>
<td>Clark, A. G.</td>
<td>185</td>
</tr>
<tr>
<td>Clarke, C. W.</td>
<td>187</td>
</tr>
<tr>
<td>Clarke, J. M.</td>
<td>252</td>
</tr>
<tr>
<td>Clutton-Brock, T. H.</td>
<td>134, 179, 251, 260</td>
</tr>
<tr>
<td>Coale, A. J.</td>
<td>32, 195</td>
</tr>
<tr>
<td>Cochran, M. E.</td>
<td>11</td>
</tr>
<tr>
<td>Cohen, D.</td>
<td>239</td>
</tr>
<tr>
<td>Cohen, J. E.</td>
<td>47, 155</td>
</tr>
<tr>
<td>Cole, L. C.</td>
<td>19, 189, 200</td>
</tr>
<tr>
<td>Collins, J. P.</td>
<td>266</td>
</tr>
<tr>
<td>Comfort, A.</td>
<td>197, 246, 247, 248</td>
</tr>
<tr>
<td>Compton, R. M.</td>
<td>54</td>
</tr>
<tr>
<td>Cooke, F.</td>
<td>256</td>
</tr>
<tr>
<td>Cornette, J. L.</td>
<td>75, 113</td>
</tr>
<tr>
<td>Costantino, R. F.</td>
<td>54, 57, 123, 149, 183, 184</td>
</tr>
<tr>
<td>Coyle, A. M.</td>
<td>199</td>
</tr>
<tr>
<td>Craig, R.</td>
<td>223, 224, 226, 227</td>
</tr>
<tr>
<td>Crandall, R. E.</td>
<td>209</td>
</tr>
<tr>
<td>Curry, R. L.</td>
<td>255</td>
</tr>
<tr>
<td>Crow, J. F.</td>
<td>61, 62, 63, 68, 75, 78, 79, 88, 105, 112, 126, 139, 142, 181, 182, 231, 244</td>
</tr>
<tr>
<td>Dahlberg, G.</td>
<td>94, 95, 104</td>
</tr>
<tr>
<td>Darwin, C. R.</td>
<td>189</td>
</tr>
<tr>
<td>Dawson, T. E.</td>
<td>237</td>
</tr>
<tr>
<td>De Jong, G.</td>
<td>260</td>
</tr>
<tr>
<td>De Jong, T. J.</td>
<td>210</td>
</tr>
</tbody>
</table>
Deevey, E. S., 15, 18
Demetrius, L., 39, 180, 192
Dempster, E. R., 158, 159
Desharnais, R. A., 54, 57, 149, 183, 184
Dickerson, G. E., 169, 177
Diekmann, O., 11
Dobzhansky, T., 128, 180, 183
Dorey, M. L., 199
Dubinin, N. P., 128
Du Mouchel, W. H., 180
Dunet, G. M., 258
Dybas, H. S., 19, 27

Edley, M. T., 252
Edney, E. B., 199
Endler, J. A., 252
Ehler, S., 11
Elsen, J. M., 169
Emigh, T. E., 63, 79, 85, 91
Emlen, J. M., 192
Emlen, S. T., 223, 255, 256
Euler, L., 2, 23
Ewens, W. J., 162

Fagen, R. M., 213, 215, 218
Falconer, D. S., 168, 169, 170, 173, 174, 261, 262
Feldman, M. W., 224
Feller, W., 26, 28, 160, 268, 269
Felsenstein, J., 63, 79, 85, 87, 89, 90, 91
Finch, C. E., 134, 197, 198, 239, 243, 244, 246, 247, 249
Fischer, K. E., 246, 248
Fitzpatrick, J. W., 255
Fowler, K., 245
Free, C. A., 58
Frota-Pessoa, O., 95

Gadgil, M., 7, 185, 213, 214, 216, 217, 218, 220, 222, 253
Gaines, M. S., 128
Geber, M. A., 252
Geberhardt, M. D., 259
Gershenson, S., 128
Ghiselin, M. T., 229
Giesel, J. T., 154, 155, 156, 131
Gill, R. W., 199
Gillespie, H. J., 63, 154, 155, 156, 183
Ginzburg, L. R., 149, 183, 184
Godfrey, H. C. J., 58
Goodman, D., 237, 256, 257, 258
Goodman, L. A., 7, 36, 37, 39, 192
Gourley, R. S., 186, 237
Grafen, A., 265
Grant, P. R., 255
Gregorius, H.-R., 62

Grey, D. R., 252
Grenfell, B. T., 58
Grosberg, R. K., 249, 250
Gross, M. R., 228, 229
Guckenheimer, J., 54
Guinness, F. E., 260
Gurney, W. S. C., 53, 54, 57
Gurtin, M. E., 54
Gustafsson, L., 264

Hadley, N. F., 21
Haigh, J., 131
Hairson, N. G., 203, 250
Hajnal, J., 63, 96, 97, 101, 102, 103, 104
Haldane, J. B. S., 23, 105, 112, 136, 139, 154, 155, 156, 157, 160, 179, 182, 184, 189, 200
Halpern, B. L., 132
Hamilton, R. B., 20
Hamilton, W. D., 38, 39, 189, 191, 192, 197, 198, 224, 248, 249
Harada, Y., 176, 185
Hare, H., 231
Harper, J. L., 7, 15, 21, 50, 55, 260
Harrison, D. E., 243
Harvey, P. H., 247, 251, 254, 264, 265, 266
Hassell, M. P., 54, 58
Hazel, L. N., 169
Heckel, D. G., 50
Henry, L., 2, 15, 235
Hill, W. G., 63, 81, 88, 90, 91, 169, 172, 193
Hodder, V. M., 21
Hoffmaster, D. K., 244, 245, 261
Holgate, P., 239
Hoppensteadt, F. C., 27
Houle, D., 177, 244, 245, 261
Houston, A. D., 212
Howell, N., 249
Hudson, R. R., 131
Huffaker, C., 58
Hughes, K. A., 245, 261
Hutchinson, E. W., 199

Ihara, J. T., 245, 261
Imaizumi, Y., 63
Ipakitchi, A., 54

Jacquard, A., 22, 62, 69, 70, 82, 95
Jayakar, S. D., 154, 155, 156, 157, 182
Jensen, D., 183
Johnson, D. L., 63, 79, 85, 87, 88, 89, 90, 91
Jürgens, K. D., 247, 248

Kallman, K. D., 228
Kaplan, N. L., 131
Karlin, S., 154, 155, 227, 231
Kawecki, T. J., 212, 213
Keller, J. B., 27
Kempthorne, O., 58, 105, 106, 107, 112, 136, 166, 167, 178, 181, 182
Kenyon, K. W., 257
Keyfitz, N., 2, 15, 17, 18, 25, 41
Keymer, A. E., 254
Kim, Y. J., 46
Kimball, A. W., 136
Kimura, M., 61, 62, 63, 68, 75, 78, 79, 88, 93, 96, 103, 105, 112, 126, 139, 142, 149, 181, 182, 183, 193, 231
King, C. E., 105, 112, 116, 129, 166
Kirkpatrick, M., 54
Kirkwood, T. B. L., 199
Klinkhammer, P. G. L., 210
Koella, J., 209, 212, 213
Koenig, W. D., 223, 255, 256
Kondrashov, A. S., 244
Kot, M., 53, 54, 58
Koufopoulos, V., 259, 262
Kozlowski, J., 254
Krebs, C. J., 128
Kuzma, J. W., 132
LaBarbera, M., 255
Lacey, E. P., 50
Lack, D., 19, 50, 54, 201, 256, 257
Lande, R., 105, 173, 176
Langley, C. H., 131
Law, R., 213, 239, 252, 262, 263
Lawrence, C. E., 186, 237
Lawton, J. H., 54, 58
Lefcovitch, L. P., 11
Leigh, E. G., 230
Lenz, M. W., 131
León, J. A., 149, 184, 186, 187, 213, 218, 264
Leslie, P. H., 5, 34, 54
Lessard, S., 227, 231
Levene, H., 159, 180
Levin, D. A., 116, 156
Levins, R., 184
Levinton, J. S., 247
Lewis, E. G., 5
Lewontin, R. C., 196
Liberman, U., 154, 155
Lloyd, M., 19, 27
Lofsvold, D., 173
Lopez, A., 28, 45, 46, 47
Lotka, A. J., 2, 11, 23, 27, 28
MacArthur, R. H., 239, 265
MacCamy, R. C., 54
MacKinley, M. D., 199
McCleaghlan, L. R., 128
McKendrick, A. G., 13
McKusick, V. A., 132
McNamara, J. M., 212
Macquot, J. C., 169
Malécot, G., 75
Mangel, M., 187, 230
Martinez, D. E., 247
Matsuda, H., 176, 185
May, R. M., 51, 53, 54, 58
Maynard Smith, J., 58, 131, 185, 190, 224, 228, 233, 252, 260
Medawar, P. B., 189, 197, 199, 200, 237
Mertz, D. B., 266
Metz, J. A. J., 11
Michod, R. E., 213, 221, 224, 239
Moffa, A. M., 123
Møller, A. P., 256
Moran, P. A. P., 53, 62, 105, 181
Moreno, J., 264
Morgan, M. T., 237
Morishima, H., 249
Moss, D., 4
Mountford, M. D., 239
Mousseau, T. A., 252
Mueller, L. D., 183, 244, 266
Murdoch, J. L., 132
Murphy, G. I., 203
Nagylaki, T., 61, 107, 113, 121, 139, 145, 146, 147, 149, 166, 174, 182, 183
Neel, J. V., 133, 134
Nei, M., 63
Nelson, J. P., 257
Nicholson, A. J., 50, 55, 183, 184
Nisbet, R. M., 53, 54, 57
Norton, H. T. J., 46, 72, 105, 112, 122, 136, 166, 167, 178, 180
Novitski, E., 236
Nunney, L., 63, 88, 89
Nur, N., 260
Ohta, T., 193
Oka, H.-I., 249
Ollason, J. C., 258
Orzack, S. H., 50, 105, 156, 204, 205
Oster, G. F., 53, 54, 57, 58
Parker, W. S., 218
Pärta, T., 264
Parlett, B., 7
Partridge, L., 245, 259, 262
Pavlovsky, O., 180, 183
Pease, C. M., 185, 262
Pennycuick, C. J., 54
Perrins, C. M., 4, 18
Pianka, E. R., 218
Pike, M. C., 246
Pollak, R. A., 7
Pollard, J. H., 2, 7, 23, 46
Price, G. R., 182, 199
Author Index

Promislow, D. E. L., 246, 247, 248, 251, 254, 264, 265, 266
Prothero, J. W., 247, 248
Prout, T., 107, 151
Putwain, P. D., 262

Quinn, J. S., 256
Quinn, T. W., 256

Ratnieks, F. L. W., 223, 255, 256
Read, A. F., 247
Real, L., 50
Reed, T. E., 133, 134
Reeve, H. K., 223, 255, 256
Rendel, J. M., 169
Renzwick, D. N., 252, 259, 262, 263, 264
Rice, D. W., 257
Ricker, W. E., 53
Ricklefs, R. E., 50, 213, 218, 257
Robbins, R. B., 169, 177, 193
Robertson, D. R., 230
Robertson, F. W., 252
Robertson, O. H., 249
Roerdink, J. B. T. M., 210
Roff, D. A., 106, 185, 187, 206, 208, 209, 227, 228, 239, 250, 251, 252, 253, 255, 259, 264, 266
Rogers, A. R., 224, 249
Rorres, C., 224
Rose, M. R., 177, 189, 197, 199, 239, 243, 245, 246, 247, 261, 262
Rose, R. K., 128
Rosenzweig, M. L., 187, 213, 214
Roughgarden, J., 149, 183, 184
Ryan, M. J., 260

Saether, B. E., 254
Sano, Y., 249
Sarukhan, J., 7
Schaal, B. A., 116
Schaffer, M. V., 217
Seger, J., 223, 236, 237
Seneta, E., 5, 46, 66
Service, P. M., 245, 261, 262
Sharpe, R. F., 2, 11
Shea, R. E., 257
Shellman-Reeve, J., 223, 255, 256
Sherman, P. W., 223, 255, 256
Shine, R., 210, 254
Sibly, R. M., 185, 187, 259, 262
Simmons, M. J., 244
Sladkin, M., 58
Slobodkin, L., 238, 239
Smith, C. A. B., 182
Smith, J. N. M., 255, 256

Smouse, P. E., 54
Sondhi, K. C., 252
Soots, R. F., 20
Southwood, T. R. E., 15
Spieth, P. T., 231
Stacey, P. B., 223, 255, 256
Stearns, S. C., 106, 185, 201, 203, 206, 208, 209, 212, 213, 239, 251, 259, 260, 262, 266
Stebbins, G. L., 202
Stephan, W., 131
Stubblefield, J. W., 223
Sutherland, W. J., 265
Sykes, Z. M., 46

Takahashi, Y., 54, 58
Tamarin, R. H., 128
Taylor, A. D., 54, 58
Taylor, H. M., 186, 237
Teitelbaum, M. S., 236
Templeton, A. R., 156, 199
Thompson, K., 250
Tinaikov, G. G., 128
Tinkle, D. W., 21, 203, 250, 266
Toro, M., 224
Townsend, C. R., 50
Trivers, R. L., 231
Tuljapurkar, S., 47, 48, 49, 50, 105, 155, 156, 204, 205, 210, 237
Tanner, J. R. G., 184
Turchin, P., 54, 58
Tuttle, M. D., 260

Usher, M. B., 54

Von Foerster, H., 13

Walker, B. A., 132
Wallace, A. R., 189
Wallace, B., 158
Warner, R. R., 230
Watanabe, T. K., 17, 180
Webb, G. F., 2
Weir, B. S., 63, 85, 91
Weismann, A., 189
Weiss, K. M., 54
Werren, J. H., 236
White, B. N., 256
White, J., 7, 21, 55, 260
Whitten, M., 246
Wiegert, R. G., 254
Wiel, T. H. E., 131
Wilbur, H. M., 203, 250, 266
Wiley, R. H., 206
Wilkinson, G. S., 252
Williams, G. C., 189, 190, 199, 213, 218, 246, 247, 248, 249
Williamson, J. A., 160, 161, 193, 239
<table>
<thead>
<tr>
<th>Author</th>
<th>Pages</th>
</tr>
</thead>
<tbody>
<tr>
<td>Willson, M. F.</td>
<td>217</td>
</tr>
<tr>
<td>Wilson, A. M.</td>
<td>250</td>
</tr>
<tr>
<td>Wilson, E. O.</td>
<td>239, 265</td>
</tr>
<tr>
<td>Wodinsky, J.</td>
<td>249</td>
</tr>
<tr>
<td>Woolfenden, G. E.</td>
<td>255</td>
</tr>
<tr>
<td>Wrege, P. H.</td>
<td>223, 255, 256</td>
</tr>
<tr>
<td>Wright, S.</td>
<td>63, 79, 105, 121, 145, 187, 200</td>
</tr>
<tr>
<td>Wyatt, R.</td>
<td>217</td>
</tr>
<tr>
<td>Wynne-Edwards, V. C.</td>
<td>190, 256, 257</td>
</tr>
<tr>
<td>Wodinsky, J.</td>
<td>249</td>
</tr>
<tr>
<td>Yodzis, P.</td>
<td>237</td>
</tr>
<tr>
<td>Young, T. P.</td>
<td>200, 250</td>
</tr>
</tbody>
</table>
SUBJECT INDEX

a, 4
a*, 6
additive genetic value, see breeding, value
additive genetic variance, see variance, additive genetic
Agave americana, see century plant
agaves, 217
age
at reproductive maturity, 5, 19–21, 108, 118, 128, 201, 248, 265
evolution of, 189–90, 205–13, 220, 240–2, 251–9, 265–6
of cessation of growth, 207, 253–4
of death, 18–19, 133–4
of gene action, see age-specific effects of genes
of onset of genetic diseases, 129, 133–4, 200
age-classes, 2–5, 17–8, 34–5, 70
critical, see critical age-group
frequency of, see age-distribution
age-distribution, 6, 17–19, 31, 39, 44, 46–7, 60, 70, 75, 130–1, 181, 196, 233, 266
change in, 34–5, 70, 116–7, 127–34, 181, 196
convergence to stable, 34–5, 40–1, 70
age-specific changes in life history traits, 190–95
age-specific survival probability, see survival rate, age-specific
age-structure, see age-distribution
ageing, see senescence
albatross
Laysan, 257
royal, 256–7
wandering, 256–7
allele frequencies, 62, 64, 70, 74, 91, 109, 113–7, 270–2
asymptotic, 62, 67–70, 72–6
approximations for, 138–49, 150–9
sex differences in, 115, 120, 125, 132, 150
alleles, multiple, 64, 124–5
allocation, age-specific
to male versus female reproduction, 189, 227, 237, 242, 267
allometric relations, 255
altruism, 223–4, 227
amphibia, 21, 248
animal breeding, 90, 168–73, 177
animals
domestic, 168
marine, 108, 250
antagonistic pleiotropy, 177
and the evolution of senescence, 198–9, 243, 245–6
arthropods, 260
ascidians, 250
assortative mating, 63
asymptotic
age-distribution, see age-distribution, stable
population growth rate, 23–8, 39, 42–4, 60, 138–9, 158, 181
mean, over time, 47–50, 155–7, 203–5, 210–12
results of selection, 166–7, 171–3
aunt–nephew marriages, 95, 102
Austria, 96, 103
autocorrelation, 50, 156, 241
average excess, 181
b, 5
b*, 6
B, 10
B_t, 9
B_m, 42
bats, 260
behavioural interactions, 224–7
bet-hedging, 203–5
Subject Index

bifurcation, 53
binomial distribution, 79, 91–2
biomass, 183
birds, 19, 54, 223, 248, 251, 254–9, 264
birth-rate, 12, 103
  per capita, 32–3, 105, 181
blood pressure, high, 134
blowfly, sheep, 55, 184
body size, 11, 21, 128, 173, 206–10, 247,
  251–2, 254–5
  polymorphism in, 228–9
  sexual selection on, 227–9, 242, 252
Botryllus, 250
branching process, 160–1, 239
breeding
  adults, number of, 88–9
  season, 2–3, 9, 20, 173, 200–1, 225–7,
  236–7, 260
  value, 169–72
bulls, 172–3
Caenorhabditis elegans, 243
cancer, 134, 197, 244
carrying-capacity, 51–6, 190, 216
  genotypic, 124, 148–9, 153–4, 183–4
  maximisation of, 186–7
census data, 16–17
century plant, 19
chaos, 54, 57–8, 183
characteristic equation, 22–3, 42–3
cicada, periodic, 19, 27
ciliates, 247
cclone, 58, 263
clutch size, 20, 201, 256–7, 260, 264
cod, 264
coefficient of relatedness, 224–7, 256
coefficient of variation, 185, 205
cohort, 16–17, 19, 197
colonising population, 69, 131, 239
comparisons
  between populations or species, 190,
  194–6, 242–3, 246–8, 250, 252–5,
  264–5, 267
compensation, reproductive, 231, 235–6
competition, 55, 58
  for mates, 227–9
  for pollinators, 217
concave function, 216–7, 220, 235, 249
conceptions, interval between, 235–6
consanguineous matings, 63, 94–104
constraint, functional, 177, 185–7, 261
continuous-time populations, 11–14, 17,
  27–9, 31, 33, 40, 46, 51–2, 58–60, 62,
  70–4, 96–104, 112–13, 121, 136,
  141–2, 166–7, 180–3, 225
approximation by discrete age-classes, 14–5, 76
convergence
  to constant population size, 51–3, 147–8
to genetic equilibrium, 151, 153, 166–7,
  180
to stable age-distribution, 34–5, 40–1,
  138, 151, 166
convex function, 216–7, 249
convolution, 268
cooperative breeding, 223, 225–7, 255–6
correlation, 49, 97, 109, 156, 205, 258
  genetic, 245–6, 252–3, 260–3
  phenotypic, 251, 259–62
cosexual species, 237, 242
cousins, see first cousins
covariance, 49, 88, 115, 120, 158, 205
genetic, 174, 176–7, 185, 260–1
  phenotypic, 176, 185, 190
cows, 172
critical age-group, 54–7, 123–4, 147–9, 183,
  193
cycle
  of allele frequencies, 127–9, 167, 180
  of population size, 53–4, 57–8, 127–9,
  180
d, 5
d*, 6
D, 76
Daphnia magna, 252
dehth-rate, 3, 12, 14, 50, 89, 133–5, 195,
  209, 231, 235, 253–5, 257–8, 260, 265
  per capita, 32–3, 105, 181
declining populations, 192, 240
deer, red, 260
deleterious mutations, 125–6, 131–4, 195,
  199–200, 219, 244–5, 261–2
demographic dominance of females, 4, 6,
  107
demographic environment, 127–34, 188,
demographic parameters, 15–21, 39
density-dependence of, see density
  dependence
estimation of, 15–19, 132–5
genotypic, 64–5, 106–13, 118, 127–35,
  148, 153–7
human, 15–18, 34, 38, 132–5
of natural populations, 15–19, 35, 38,
  134–5, 179, 246–50, 264–5
sex differences in, 88, 111–12, 143–5,
  150–1, 176, 227–37
spatially varying, 158–9, 212–13, 223
time-dependent, 44–50, 69, 154–8,
  182–3, 210–2
time- and density-independent, 22–44,
  59–62, 69–70, 80–104, 118–23,
  136–47, 150–3, 160–82, 191–3, 195,
  201–3, 206–10, 213–22
demographic stability, see age-distribution,
  stable
demography, 2, 11, 16, 23, 25, 33, 62
density dependence, 2, 44, 50–8, 69, 80, 131, 265, 267
of age-independent survival (mortality), 130, 202–3, 208, 219, 222, 263, 266
of fecundity, 55, 184, 196, 203, 208, 219, 240–1, 266
with selection
and discrete generations, 149, 183–4
derivatives
determinant, 122, 124, 176
diallelic locus, see two-allele systems
differential coefficient, 26, 269,
differential equation, 51–2, 105, 254
partial, 13
differentiation, implicit, 175, 191
digenic identity probabilities, 91
*Diomedea, see albatross
discrete time model, see discrete age-class model
disease, 194, 248
cardio-vascular, 197, 244
genetic, 132–4, 200, 244
dominance, evolution of, 200
dominant alleles, 126, 132–3, 136, 143, 178, 199
donors of altruism, 224–7, 255–6
drift, see genetic drift
*Drosophila*, 16, 21, 51, 180, 244–5, 251–2, 259
*Drosophila melanogaster*, 199, 243–5, 261–2
*Drosophila pseudoobscura*, 16–7
*Drosophila subobscura*, 260, 262
dynamic programming, 187

ESS (evolutionarily stable strategy), 185–7, 212, 228–30, 233–7, 254
effective population number, 63, 79–80, 85–91, 94
inbreeding, 79, 90
variance, 79, 90–1
eigenvalues, 22, 42,
leading, 22–4, 39, 42–3, 49, 82, 85
eigenvectors, 22–3, 34, 37, 39, 49, 82
linear independence of, 22, 25
energy, 213, 217
entropy, 180
environment, 3, 6, 128–9, 190, 261
constant, 2, 6, 201–3, 206–10, 220–2, 240–1
quality of, 259–60
varying, see spatially varying environments, temporally varying environments
epistasis, 168
equilibrium
evolutionary, 185, 190–1, 228
for multivariate traits under selection, 176–7
gene frequencies, see allele frequencies, equilibrium
genotype frequencies, 65, 199–200
Hardy–Weinberg, see Hardy–Weinberg frequencies
size of populations, 51–7, 147, 124–5
stability of, for allele frequencies, see stability of equilibria, local, for allele frequencies
stability of, for population size, see stability of equilibria, local, for population size
unstable, 141, 165–6
ergodicity
stochastic, 47–8
strong, 28, 44
weak, 44–7, 66–7, 69, 72, 155, 204
Euler equation, 23
European populations, 102
evolutionary history, 190, 264
expectation of life, 30–1, 245–6, 248
experience, increase with age, 206, 251, 257–9
external fertilization, 228
extinction, 192
extrinsic mortality, 220, 240, 246–8, 266

\( f \), 4
\( f^* \), 5
\( F \), 80
facultative adjustment of life-history, see plasticity, phenotypic
family size, 95, 134
fecundity
age-independent, 201–5, 254, 258

of males, 6, 42, 65, 107–12, 143, 160, 229–30

genetypic, see demographic parameters, genotypic
per unit size, 206–23, 241–2
relation with age, 20–1, 195, 198, 229–30, 235, 248, 263
schedule, 16, 19, 42
variable, 204–5, 210–13, 222–3, 259

female population, 3–6, 9–13, 16–18, 22–7, 29–41, 44–50, 51–8, 64
finite population size, 62–3, 78–104
first cousins
cross, matrilateral, 95–6, 102
cross, patrilateral, 95–6, 102
marriage between, 94–104
parallel, maternal, 95–6, 102, 103
parallel, paternal, 95–8, 102, 103–4
fish, 21, 209, 229, 248, 251–4, 264
fishing, commercial, 252
fitness, 105, 176–7, 185–7, 190
approximate estimators of, 134–5, 178–9
effects of demography on, 127–34
for equilibrium population, see equilibrium, fitness measures
inclusive, 224
mean of population, see mean fitness of a population
over temporally varying environments, 50, 155–8, 182–3
sensitivities, see sensitivity of fitness
fixation probability
of a favourable mutation, 193, 198
of a neutral mutation, 91–4
flycatcher, collared, 264
French population, 70, 101
frequency-dependent selection, 109, 176–7, 180, 184–6, 228–9
frogs, 260
fulmar, Atlantic, 258
fundamental theorem of natural selection, 59, 142, 182

G, 81, 174
gamete frequencies, 75–8, 114, 145
gastropods, 250
gene frequencies, see allele frequencies
generating functions, 25, 163–5, 268–9
probability, 161, 268
generation time, 28–30, 37, 41, 78, 82–3, 85, 88–9, 140, 173, 175, 178–9, 191–3, 205
genetic drift, 69, 78–94, 187, 262
genotype frequencies, 59, 68–70, 74–5, 110–16, 181, 272
genotype, ordered, 64, 75
Germany, 96, 103
germination, 8–9, 43, 262
gestation length, 264–5
gradient vector, 175–6, 185
grandchildless mutation, 260
grases, 250
great apes, 20
group selection, 190, 239, 257
growth

variable, 212–13
of populations, see population growth
guppy, 252–3, 259, 263–4

h², 168
H, 65
haplodiploidy, 226
Hardy–Weinberg frequencies, 62, 68–71, 74, 115–16, 181
rate of convergence to, 69–70
heritability, 168, 170, 173
hermaphrodites, 7, 41, 79, 87–8, 90–1, 227
sequential, 229–30, 267
heterozygote frequency, 116, 199
heterozygote inferiority, 162, 166
heterozygote superiority (advantage), 127, 143, 151, 153, 155–6, 159, 162, 164–7, 180
hitch-hiking, 131
homogeneous linear equations, 122
homozygote frequency, 199–200
horizontal method of estimating
demographic parameters, 16
human populations, 11, 16–18, 20–1, 32, 34, 38, 41, 45, 46, 70, 89, 94–104, 161, 194–5, 237, 243–4, 246, 248–9, 251
measurement of fitness in, 131–5
sex ratio of, 235–6
human reproductive value, 38, 237
Hunter–gatherer societies, 249
Huntington’s chorea, 133–5, 244
Hymenoptera, 226, 236–7
ideal population, 79
identity by descent
probability of, 80–7, 90
inbred stocks, 261
inbreeding coefficient, 78–91
incest, avoidance of, 104
incubation of eggs, 257
Subject Index

indeterminate growth, 209
Indiana, 128
indirect estimation of mutation rates, 131–2, 134
indirect estimation of vital statistics, 17–19
infant mortality, 236, 251
inflorescences, 217
insects, 19–21, 52, 109, 202, 236, 248
holometabolous, 209, 253
univoltine, 19, 200
instability
of allele frequency equilibrium, 141, 164, 166–7
of population size equilibrium, 53–4, 57–8
integral equation, 13, 22, 27–8, 182
intrinsic rate of increase, 28–33, 39–40, 120–3, 133–4, 138, 140–1, 143, 162, 190, 196, 201, 230, 232, 241
maximisation of, 186–7, 212–13, 215–16
mean
of a phenotype, 173, 185–7, 212–13
of a population, 59–60, 141–2
over time, 48–50, 182, 205–4, 212
partial derivatives of, 176, 191–2
invertebrates, 247
isolates, human, 69, 94–104
Italy, 96, 103
iteroparity, 19–21, 23, 69, 192, 217–19, 225
evolution of, 189, 200–5, 217, 240–1, 249–50, 266
Japan, 96, 103
juvenile
mortality, 30, 241, 257, 265
predation on, 252–3
stages, see pre-reproductive ages
survival, 32
k, 9
k*, 42
K-selection, 265–6
Kansas, 128
kin selection, 223–7, 241, 251, 255–6
l, 9
l*, 11
L, 5
laboratory populations, 16, 51, 180, 183–4, 244–5, 262–3
Lebesgue integration, 14
Leslie matrix, 3–7, 11, 22–5, 39, 45–50, 55–6, 68–9, 83, 154, 205
mean, over time, 49, 205
life-histories
iteroparous, see iteroparity
optimal, see optimal, life-histories
semelparous, see semelparity
life-history evolution, 11, 19–21, 39, 106, 173, 185–267
life-history traits
genetic variation in, 242–6, 252–3, 261–4
life-span, see expectation of life
life-tables, 16–19, 248
of human populations, 16–18, 132–3, 248
limit cycle, 53–4, 57–8, 167, 180
linkage, see recombination
linkage disequilibrium, 76–8, 114, 145–6, 168, 174, 262
litter size, see clutch size
lizards, 254
Lobelia, 250
logistic growth, 51–2, 265
longevity, see expectation of life
Lotka equation, 23
Lucilia cuprina, see blowfly, sheep
m, 4
m*, 6
M, 4
M*, 6
McKendrick–Von Foerster equation, 13
macronucleus, 247
Magicicada, see cicada, periodic
male population, 3, 6–7, 22, 41–3, 64
Malthusian parameter, 29, 105, 142, 180–3, 197
mammals, 18, 20, 128, 210, 223, 244, 247–8, 251, 254–5, 264–5
manipulation of life-history traits, 249, 260, 262
Marfan syndrome, 132–3
Markov process, 47, 155
marriage
age at, 95–103
consanguineous, see consanguineous
matings
models, 95–104
mast-seeding, 260
mating
group, 108–9
process, models of, 4, 7, 65, 107–9, 112–13
matrix
equations, 2, 5, 7–9, 22, 26, 37, 39, 45, 65, 82, 86, 106, 170–1, 175
Leslie, see Leslie matrix
non-negative, 5, 45, 66, 75, 83
population projection, 5, 43
spectral expansion of, 22–3
stochastic, 82
strictly positive, 67–9
maximisation
of intrinsic rate of increase, see intrinsic
rate of increase, maximisation of
of mean fitness, 123, 140–1, 185–7
Subject Index

of population parameters, 190, 239
of population size, 124, 148–9, 183–4, 186–7
of reproductive value, 238–9
maximum life-span, 246–8, 262
meadow grass, annual, 262–3
mean fitness of a population, 140–1, 175
Mendelian transmission, 65, 224
menopause, 248–9
mice, 246
micronucleus, 247
Microtus ochrogaster, 128
migration, 62–3, 103–4
modifier genes, 200
molecular evolution, 94
monocious populations, 7, 87, 106, 227
monogamy, 224–6, 256
serial, 226
mortality, see death-rate
Mount Kenya, 250
multicellular organisms, 246–7
multi-locus system, 174, 184
multi-species system, 58, 184
multiple alleles, 124–5, 145
multiple polyposis of the colon, 134
multivariate traits, selection on, 168, 174–7, 185–7, 261
mutation, 62–3, 93, 121, 177, 185, 187, 260, 267
deleterious, see deleterious mutations
favourable, 198
neutral, 63, 91–4
rate of, 93–4, 125, 131, 134
per genome, 244–5
mutation-accumulation theory of senescence, 199–200, 243–5
mutation–selection balance, 121, 125–6, 198–9
mutational collapse, 219
mutualism, 58

\[ n, 3 \]
\[ n^*, 6 \]
\[ N_c, 79 \]
natural populations, 17–21, 51, 54, 126–9, 134–5, 179, 245, 247, 262–4
demographic parameters of, see demographic parameters, of natural populations
net reproduction rate, 29, 57, 89, 134, 148, 179, 183, 193, 213, 229, 265
Newton–Raphson iteration, 29, 123
non-random mating, with respect to age, 109, 114–16, 120, 132, 146, 150, 178
with respect to genotype, 62
non-recessive alleles, see dominant alleles, semi-dominant alleles
non-senescent life-histories, 198
normal distribution, 101, 173, 176
North Carolina, 20, 35
Norton’s theorem, 46, 72–3
octopuses, 249
offspring number, distribution of, 79, 91, 160–2
Onchorhynchus gorbuscha, see salmon, pink
opposum, Virginia, 247
optimal age at maturity, 207–10, 213, 228
control theory, 187
life-histories, 185–8, 213–22, 237–9, 258
optimisation theory, evolutionary, 137, 185–8
oscillation in allele frequencies, 69–70, 75, 166–7, 180
in population frequencies, 53–5, 57–8, 180, 183

\[ P, 3 \]
\[ P^*, 3 \]
palm, 264
parasitism, 58
partial fractions, 268–9
Pearl–Verhulst equation, 51
pedigrees, 99–100, 224
period, environmental, 156–8
periodic fecundities, 23–4, 26–7, 44, 70
perturbations to life-history, 39, 185–7, 215–16, 221–2, 238, 258
phenotypic value, 168–9, 173–7
mean, of population, 168–9, 174–6, 185–7
plaice, 264
plants
annual, 1, 19, 52, 202
demography of, 7, 250
flowering, 244, 250–2
perennial, 2, 20–1, 260
self-fertilising, 263
self-incompatible, 108
with seed dormancy, 7–9, 22, 43–4
plasticity
phenotypic, 212–13, 223, 228, 241, 259
pleiotropy theory of ageing, see antagonistic pleiotropy
Poa annua, see meadow grass, annual
Poecilia reticulata, see guppy

Poisson distribution of offspring number, 160–1, 194
pollen pool, 108
pollinators, 217
polygamy, 226
polymorphism, 122–3, 132, 136, 143, 162–7, 180, 228–9, 250
inversion, 180
protected, see protected polymorphism polynomial, 118–9
population
cage, 51, 180
captive, 246, 265, 267
density, see population size
high-fecundity, high mortality, 101, 133, 195
low-fecundity, low-mortality, 133, 195
process, solution of, 21–8, 40–1, 44
slowly growing, 179
population growth
asymptotic, see asymptotic, population growth rate
density-dependent, see density dependence
effect of selection on, 137, 142–3, 181–3, 203
geometric, 24, 27, 33, 40, 45, 119
population size, 21, 51–8, 123–5
constant, see stationary populations
effect of selection on, 148–9, 183–4
effective, see effective population number
postponement of reproduction, see reproduction, deferral of
post-reproductive ages, 5–6, 248–9
predation, 58, 184, 194, 220, 238–9, 247–8, 252–3, 263–4
pre-industrial societies, 101, 236, 249
pre-reproductive ages, 5, 15, 38, 46, 68, 192, 240
density-dependent mortality of, see density dependence, of juvenile survival (mortality)
primary sex-ratio, see sex-ratio
probability distribution of age-structure, 47, 155
probability of death, see death rate
probability of fixation, see fixation probability
probability of survival
of a mutant gene, 160–2, 193–4, 239
of a population, 239
protandry, 229–30
protected polymorphism, 151, 153–9
protogyny, 229–30
q, 82
Q, 3
Q*, 3
quantitative characters, selection on, 167–77, 184–5, 190, 200
r, 28
r-selection, 265–6
R, 29
random mating
with respect to age, 80, 91, 108–9, 112, 114–15, 147, 151, 160, 174, 179, 193, 224, 230
with respect to genotype, 63, 68–9, 71, 75, 79–80, 91, 108–9, 112, 114, 147, 151, 160, 174, 193, 224, 230
with respect to relationship, 96, 104
random sampling of alleles, see genetic drift rate of evolution, 193, 198
recessive alleles, 126, 132, 136, 143, 151–3, 180, 182, 199, 261
recipients of altruism, 224–7, 255
recombination, 121, 145, 174, 179, 184
fraction, 76–7, 114, 145–6
relatives, behavioural interactions between, 223–7, 255–6
renewal equation, 25–7, 42–3, 72–3, 270–1
model of human fecundity, 235
theorem, 26
reproduction
asexual, 7, 58–61, 106, 246–7, 250
sexual, 106, 247, 250
termination of, 248–9
reproductive activity
in relation to age, 189, 251–3
in relation to age, 189, 217–19, 241, 258–9, 264
reproductive function, 10, 12, 25, 41, 58, 117–18, 137–9, 143–5, 147, 158–9, 186, 270–2
reproductive maturity, age of, see age at reproductive maturity
reproductive value, 33–40, 60–1, 82, 182, 237–9
eventual, 36–8, 192
Fisher’s, 33–40, 120, 192, 197–8, 210, 215, 217, 225–6, 238–9
Goodman’s, see reproductive value, eventual
residual, 218
Schaffer’s modified, 215–18
total, 39–40, 60
reptiles, 21, 248
rice, 249
response to artificial selection, see selection, response

s, 191
s', 192
salmon
    Pacific, 217, 249
    pink, 19, 70
satellite males, 228–9
Sciurus carolinensis, see squirrel, grey
seed dormancy, see plants, with seed dormancy
seedlings, 55
selection, 62–3, 69, 91
    artificial, 168–73, 193
        on life-history traits, 245, 261–2, 267
        coefficient, 133–4, 180
        differential, 168, 170, 173
        group, see group selection
        hard, 158
        intensity of, 136–7, 139, 149, 174, 198, 237
        response, 168, 171–3, 193
        sexual, 227–9, 242
        soft, 158–9
        strong, 115, 149, 151–3, 157, 179–80, 182
        weak, 115, 120–1, 125, 136–49, 151, 157, 166, 174, 179, 181–2, 190
self-fertile populations, 79, 263
semelparity, 19, 23, 26–7, 52, 70
    evolution of, 200–5, 216–7, 240–1, 249–50, 266
semi-dominant alleles, 126, 132, 143, 178, 199
senescence, 4, 20–1, 235
    evolution of, 189–90, 192, 197–200, 237, 240, 243–9, 266
Seneta’s theorem, 46, 66–7
sensitivity of fitness, 191–6, 197–9, 219, 237, 240–2, 248
    to change in age-specific fecundity, 192–3, 240–1, 251, 266
    to change in age-specific survival, 191–2, 198–9, 240–1, 266
sequential hermaphroditism, see hermaphrodites, sequential
    in variable environments, 236–7
sex change, evolution of, 229–30, 242, 267
sex differences
    in fitness, 120–1, 132–3, 143–5, 176, 179, 228
    in timing of reproduction, 227–8
sex-linkage, 62, 74–5, 231
sex-ratio
    and parental age, see parental, age, and offspring sex-ratio
    at birth, 236
primary, 4, 42, 71, 77, 81, 88, 90, 93, 96, 109, 112, 160, 189, 224–5, 231–7
    selection on, 231–6
    in variable environments, 236–7
sheep, 18
shifting-balance process, 187
snakes, 254
sociality
    evolution of, 223, 226–7
    somatic versus reproductive structures, 242, 246–7, 264
spatially varying environments, 158–9, 212–13, 223, 241, 251
sperm pool, 108
squirrel, grey, 20, 29–31, 38, 89
stability of equilibria
    global, for allele frequencies, 140–1, 148, 166–7, 180
    local, for allele frequencies, 140, 149–59, 162–6, 180
    local, for population size, 51–4, 58
stable age-distribution, see age-distribution, stable
stage-structured populations, 11
stationary populations, 17–18, 29, 32, 51, 79–80, 89–90, 96, 99–100, 103, 123–5, 147–9, 160–1, 169, 193, 209, 229
sterilisation, 249, 260
stochastic instability, 155
substitution rate
    of favorable mutations, 193
    of neutral mutations, 93–4
sunfish, bluegill, 229
survival function, 9, 12, 14, 132, 206, 209
survival rate
    adult, 199, 201–6, 222–3, 225–6, 241, 249, 256–9, 262–5
    density-dependent, see density dependence
genotypic, see demographic parameters, genotypic
juvenile, 196, 201–5, 209–10, 214, 241, 249
    variable, 203–5, 210–13, 222–3, 241, 250, 259
swordtail, 228
synchronised breeding, 27
systems of several populations, 58–61

T, 30
Taiwan population, 133–4, 194–5
Taylor’s theorem, 29, 50, 52
temperate zone, 2, 20, 202
temperature, 129
temporally varying environments
deterministic, 44–6, 155–8
effects on genotypic fitnesses, 127–31, 133–4
stochastic, 44, 46–50, 155–6, 158
territorial males, 228–9
territories, 228, 255
time-intervals, 11, 14–15, 32, 34–5, 77–8, 81, 94, 110, 171, 191
correlations in demographic parameters between, 48–50, 156, 205, 241
tit
blue, 260
great, 4, 18, 104
total differential, 216, 233, 238
trade-offs, 199, 250–2, 260–4
transferrin locus, 128
*Tribolium castaneum,* 123
two-locus systems, 62, 75–8, 113–14, 145–6, 179
two-sex populations, 7, 106, 169
US population, 17–18, 34, 38, 132–4, 194–5
uncle–niece marriages, 95–6, 102
unicellular organisms, 246–7

\(v,\) 36

variance
additive genetic, 168, 177
in intrinsic rate of increase, 142, 181

in life-history traits, 245
in age at marriage, 101
in allele frequencies, 79, 115–6, 120
in demographic parameters, 156, 241
in intrinsic rate of increase, 59–60
in parental age, 101
in reproductive success, 88, 162
phenotypic, 176, 185, 190

variance–covariance matrix, additive genetic, 174–7, 185
variation, within populations, 190, 228–9, 243–5, 250, 252–3, 261–4

varying environments, *see* spatially varying environments; temporally varying environments
vascular system, 132
vegetative reproduction, 246
vertebrates, 2, 20, 260
cold-blooded, 21, 209, 248, 253
vertical method of estimating demographic parameters, 16–17
vital statistics, *see* demographic parameters
Von Bertalanffy equation, 209
Von Foerster equation, *see* McKendrick–Von Foerster equation

\(w_q,\) 117, 120
wrasse, 228

*Xiphophorus nigrensis, see* swordtail

yucca, 217
zooid, 250
Preface

English. I have not always taken their advice, and the remaining defects are, of course, entirely my responsibility.

Finally, I must express my gratitude to Chris Cannings for suggesting that I write this book and for his encouragement during its preparation.

Preface to the second edition

Fifteen years have gone by since I wrote the first edition of this book. Some notable advances have been made in the population genetics of age-structured populations, and in the theory of life-history evolution. In addition, the literature on empirical tests of life-history theory has grown enormously. In order to include an account of these developments in the new edition, while keeping the length of the book roughly constant, I have deleted some old material as well as adding new. I have rewritten Chapter 5 completely, in an attempt to do justice to recent work on life-history evolution. I have not tried to provide a comprehensive review of this large field, but have concentrated on emphasising theoretical principles and empirical tests which seem to me to be especially illuminating.

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