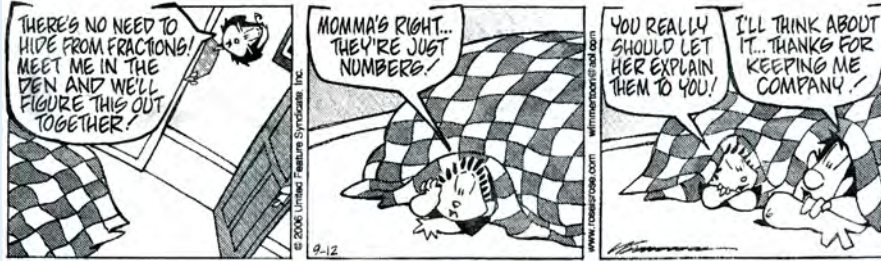


## Quantitative Modeling

# Structured Populations

**ROSE IS ROSE** | Pat Brady and Don Wimmer



S.R. Borrett

Bio534

## Announcements & Assignments

### Labs

- For each task (1) describe what you did, (2) the results (include graphs), and (3) describe what it means.
- Make sure to address all questions posed in the initial assignment
- put code in an appendix
- Lab 4 due on Friday

### Project Topic Selection

- 1 page description - see assignment
- proposed due date **Oct 19**.

### Exam

- ~10 multipart questions (short answer, sketch, essay)
- cover material up to generalized control functions.

## Meeting Objectives

- Distinguish **structured** population models from unstructured population models
- Discuss **age** and more **general class** structured populations
- Contrast **r** and **K selected** life history characteristics
- Use **life table** technique to estimate structured population growth rates
- Use Matrix population models to **forecast** future population size and **stable age distribution**.

## Why Structured Populations?

### Unstructured Population Models

- Assume all individuals are genetically and ecologically equivalent (Identical Individuals II)
- Overlapping generations with births and deaths at any time (continuous models)

### Structured Models

#### - Individual Distribution Models

- Represent differences among individuals by placing them into different classes (e.g. age, sex, size)
- Still assuming II within a class

#### - Individual Configuration Models (IBM, ABM)

- Each individual is modeled computationally
- No II assumption

## individual distribution models

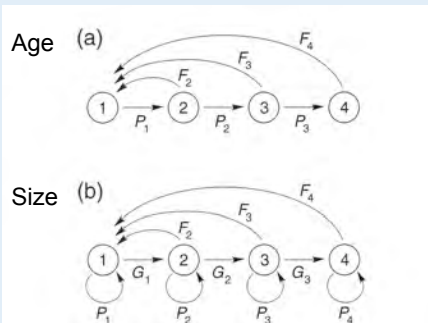
### matrix population models

For more information see

Caswell, H. 2001. Matrix Population Models: Construction, Analysis, and Interpretation. Sinauer Associates, Inc. Publishers. Sunderland, MA.

## Age Structured Model

- Partition population into classes based on a trait that influences the process rates (e.g. age)
- Choose a projection interval ( $\Delta t$ )
  - Width of classes may equal the projection interval so that all individuals in a class move on the next time step, but this may not be true



$$\mathbf{A}_a = \begin{pmatrix} 0 & F_2 & F_3 & F_4 \\ P_1 & 0 & 0 & 0 \\ 0 & P_2 & 0 & 0 \\ 0 & 0 & P_3 & 0 \end{pmatrix}$$

$$\mathbf{A}_b = \begin{pmatrix} P_1 & F_2 & F_3 & F_4 \\ G_1 & P_2 & 0 & 0 \\ 0 & G_2 & P_3 & 0 \\ 0 & 0 & G_3 & P_4 \end{pmatrix}$$

Caswell 2001

Figure 4.1: (a) A life cycle graph for an age-classified life cycle, in which the width of the age classes equals the projection interval. (b) A life cycle graph for the standard size-classified model. Nodes represent size classes, and an individual can grow no more than a single size class in the interval  $(t, t+1)$ .

## Leslie Matrix

When we partition the population into  $a = 0, \dots, A$  age classes, then we can calculate the future population density of any age class as

$$n_a(t + 1) = p_{a-1}n_{a-1}(t) \text{ for } a > 0$$

Where

$p_x$  = the probability that an  $x$ -year-old survives to be age  $x + 1$

If we assume that per individual birth rates are only a function of age, then

$f_a$  = the average number of newborns next year, per age- $a$  female this year

$$n_0(t + 1) = \sum_{a=0}^A f_a n_a(t)$$

We can summarize the whole population dynamics as

$$\begin{bmatrix} n_0(t + 1) \\ n_1(t + 1) \\ \vdots \\ n_A(t + 1) \end{bmatrix} = \begin{bmatrix} f_0 & f_1 & f_2 & \dots & f_A \\ p_0 & 0 & 0 & \dots & 0 \\ 0 & p_1 & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & & p_{A-1} & 0 \end{bmatrix} \begin{bmatrix} n_0(t) \\ n_1(t) \\ \vdots \\ n_A(t) \end{bmatrix} \rightarrow \mathbf{n}(t + 1) = \mathbf{L}\mathbf{n}(t)$$

Where **L** is the Leslie matrix

Eilner & Gukenhimer 2006

## Assumptions, Caveats, and Warnings

This is a **linear, time-invariant** model

- It is also a discrete-time, exponential or geometric model

Caswell's (2001, p 30) distinction between

- **Forecasting** - predict what *will* happen
- **Projection** - what *would* happen, given certain hypotheses

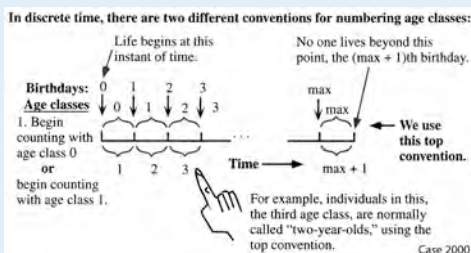
Counting **everyone** or just **females**

- Just females is common
- Assumes males are not limiting reproduction

When to census: **prebreeding** vs. **postbreeding**

- Pulse birth populations
- Post:  $f_a = p_a \times m_{a+1}$
- Pre:  $f_a = m_a \times p_0$

Numbering schemes



## Generalization: Stage Classes

Populations can be partitioned using any relevant trait

- Age (Leslie Matrix)
- Developmental Stage (L. Lefkovich 1965)
- Movement between patches (space)
- ...or any combination!

Thus, the “individual distribution” structured population model is quite general

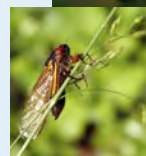
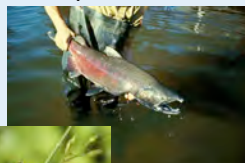
This is also a useful approach when modeling semelparous species

### semelparous

organism that reproduces just once in its lifetime  
(e.g., salmon, cicadas)

### iteroparous

organism that reproduces many times



## Solution to the Projection Equation

Let  $\mathbf{A}$  represent the general projection matrix, where  $a_{ij}$  is the probability that an individual in the type- $j$  (compartment  $j$ ) will move to type- $i$  during  $\Delta t$

Then our model is

$$\mathbf{n}(t + 1) = \mathbf{A}\mathbf{n}(t)$$

As we saw with the scalar discrete time model,

we can find the general solution as ...

$$\mathbf{n}(1) = \mathbf{A}\mathbf{n}(0)$$

$$\mathbf{n}(2) = \mathbf{A}\mathbf{n}(1) = \mathbf{A}(\mathbf{A}\mathbf{n}(0)) = \mathbf{A}^2\mathbf{n}(0)$$

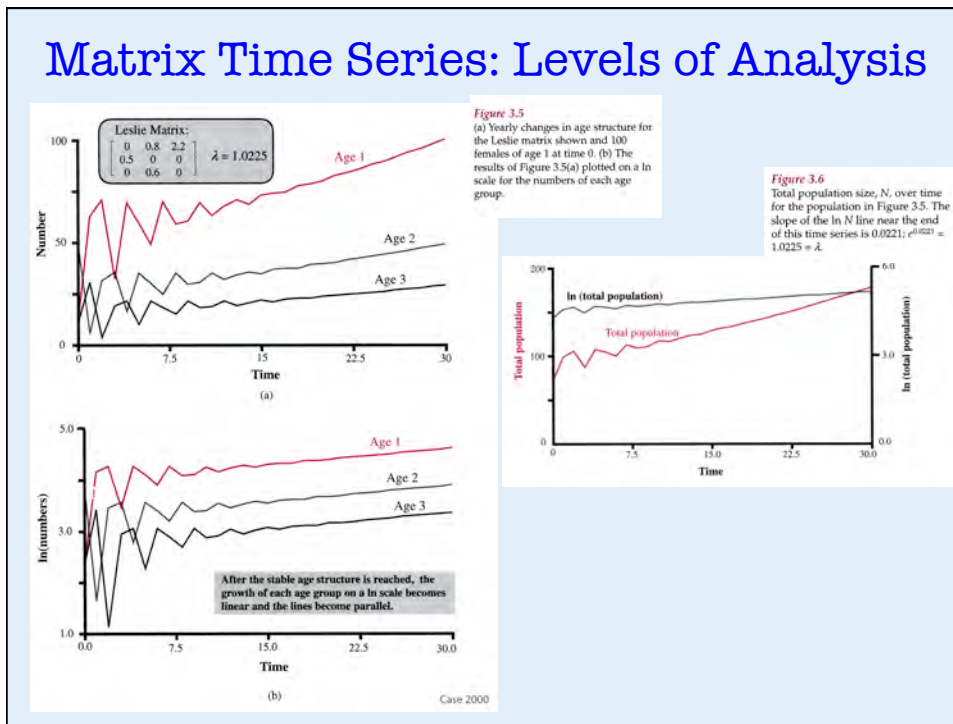
$$\mathbf{n}(3) = \mathbf{A}\mathbf{n}(2) = \mathbf{A}(\mathbf{A}^2\mathbf{n}(0)) = \mathbf{A}^3\mathbf{n}(0)$$

⋮

$$\mathbf{n}(t) = \mathbf{A}^t\mathbf{n}(0)$$

Thus, given  $\mathbf{A}$  and an initial population distribution  $\mathbf{n}(0)$  we can project the future population at time  $t$

# Matrix Time Series: Levels of Analysis



## Analysis

We can analyze the projection matrix  $A$  to learn the following

1. Asymptotic rate of population growth → dominant eigenvalue
2. Stable stage distribution → right dominant eigenvector
3. Reproductive value of a stage → left dominant eigenvector

The dynamics of  $A^t$  are determined by the characteristic polynomial

$$\pi(\lambda) = \det(\lambda I - A)$$

Where

$I$  is the matrix identity matrix

$\lambda_i$  are the roots of the characteristic polynomial where  $\pi(\lambda) = 0$

The eigenvalues  $\lambda_i$  must satisfy  $AW = \Lambda W$  and  $VA = V\Lambda$

$$A^t = W \times \Lambda^t \times W^{-1},$$

$$A^t = W \times \begin{bmatrix} \lambda_1^t & 0 & \dots & 0 \\ 0 & \lambda_2^t & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \lambda_n^t \end{bmatrix} \times W^{-1}$$

The columns of  $W$  are the corresponding right eigenvectors

The columns of  $V$  are the corresponding left eigenvectors

# Example: USA Population

## Example 4.3 An age-classified population

Keyfitz and Fieger (1971) give an age-classified matrix, with 5-year age classes and a projection interval of 5 years, for the United States population in 1966. The entries are

$i$	$F_i$	$P_i$
1	0	0.99670
2	0.00102	0.99837
3	0.08515	0.99780
4	0.30574	0.99672
5	0.40002	0.99607
6	0.28061	0.99472
7	0.15260	0.99240
8	0.06420	0.98867
9	0.01483	0.98274
10	0.00089	

The eigenvalues of this matrix (in decreasing order of absolute magnitude), their magnitudes, and the angle  $\theta$  (as a fraction of  $\pi$ ) defined in the complex plane by each are

$\lambda_i$	$ \lambda_i $	$\theta/\pi$
1.0498	1.0498	0.0000
$0.3112 + 0.7442i$	0.8067	0.3739
$0.3112 - 0.7442i$	0.8067	-0.3739
$-0.3939 + 0.3658i$	0.5375	-0.7618
$-0.3939 - 0.3658i$	0.5375	0.7618
$0.0115 + 0.5221i$	0.5223	0.4930
$0.0115 - 0.5221i$	0.5223	-0.4930
$-0.4112 + 0.1204i$	0.4284	-0.9093
$-0.4112 - 0.1204i$	0.4284	0.9093
-0.0852	0.0852	1.0000

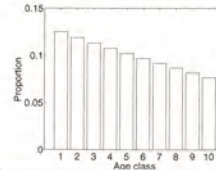
Figure 4.6 shows this spectrum plotted in the complex plane. (The computation of eigenvalues and eigenvectors is discussed in Section 4.8.) The largest eigenvalue is real and positive. The second largest is a complex conjugate pair at angles of about  $\theta = \pm 0.37\pi$  in the complex plane. This pair would generate oscillations with a period  $2\pi/\theta \approx 5.4$  projection intervals. Since the projection interval is 5 years, the oscillation produced by this pair of eigenvalues has a period of 27 years. Caswell 2001

## Stable Stage Distribution (right ev)

### Example 4.4 An age-classified model

The eigenvector  $w_1$  corresponding to  $\lambda_1 = 1.0498$  for the United States population in 1966 (Example 4.3), scaled so that  $w_1$  sums to 1, is

$$w_1 = \begin{pmatrix} 0.1252 \\ 0.1189 \\ 0.1131 \\ 0.1075 \\ 0.1020 \\ 0.0968 \\ 0.0917 \\ 0.0867 \\ 0.0817 \\ 0.0765 \end{pmatrix}$$



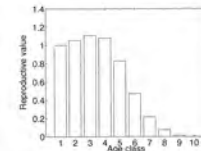
The pattern shown by  $w_1$  — decreasing proportions in successive age classes — is typical for age-classified populations. Caswell 2001

## Reproductive Value (left ev)

### Example 4.7 An age-classified population

The reproductive value vector  $v_1$  for the United States population (Example 4.3) is

$$v_1 = \begin{pmatrix} 1.0000 \\ 1.0532 \\ 1.1064 \\ 1.0787 \\ 0.8293 \\ 0.4724 \\ 0.2165 \\ 0.0752 \\ 0.0149 \\ 0.0008 \end{pmatrix}$$



Note the increase in  $v_1$  from birth to a maximum in age class 3 (ages 10 – 15 years), and its subsequent decline. This pattern is typical for age-classified populations. Caswell 2001

# Analysis Assumptions

These matrix analysis assume that **A** is

**Positive** (non-negative)

all elements of **A** are  $\geq 0$

**Irreducible**

If  $(I + A)^{s-1}$  is positive, where  $s$  is size of **A**

**Primitive**

Iff  $A^{(s^2-2s+2)}$  is positive

See Caswell 2001 Chapter 4 for more detail and consequences of violating these assumptions.

## Generalizing the Matrix Model

- Stochastic Matrix Models
  - Density-Dependent Matrix Models
- Both relax the assumption that  $A$  is a constant coefficient matrix
- Continuous Size Distributions
- Integral projection matrix

## Additional Demographic Considerations:

*How do we estimate growth rates?*

demography:

the statistical study of the size and structure of populations and changes within them

Life Table Analysis

## Demographic Parameters

- **Cohort:** group of individuals born at the same time
- $x$  = age class of a population
- $n_x$ : number of individuals of age ( $x$ )
- $l_x$ : probability at birth of surviving to age  $x$
- $d_x$ : number of individuals who have died between age classes ( $n_x - n_{x+1}$ )
- $q_x$ : age specific mortality ( $d_x/n_x$ )
- $b_x$ : age specific birth rate
- $s_x$ : survival rate ( $1 - q_x$ )

## Life Table Analysis – Life History

Table 3.1 Standard life-table calculations.<sup>a</sup>

$x$	$S(x)$	$b(x)$	$l(x) = S(x)/S(0)$	$g(x) = l(x+1)/l(x)$	$l(x)b(x)$	$l(x)b(x)x$	Initial estimate $e^{-rx}l(x)b(x)$	Corrected estimate $e^{-rx}l(x)b(x)$
0	500	0	1.0	0.80	0.0	0.0	0.000	0.000
1	400	2	0.8	0.50	1.6	1.6	0.780	0.736
2	200	3	0.4	0.25	1.2	2.4	0.285	0.254
3	50	1	0.1	0.00	0.1	0.3	0.012	0.010
4	0	0	0.0		0.0	0.0	0.000	0.000

Data

$R_0 = \sum l(x)b(x)$	= 2.9 offspring	$\Sigma = 4.3$	$\Sigma = 1.077$	$\Sigma = 1.000$
-----------------------	-----------------	----------------	------------------	------------------

### Survivorship

$l(x)$  = prob. ind. survives to age  $x$  from  $x=0$

### Survival Probability

$g(x)$  = prob. ind. survives to age  $x+1$  from  $x$

**Net Reproductive Rate =  $R_0$**

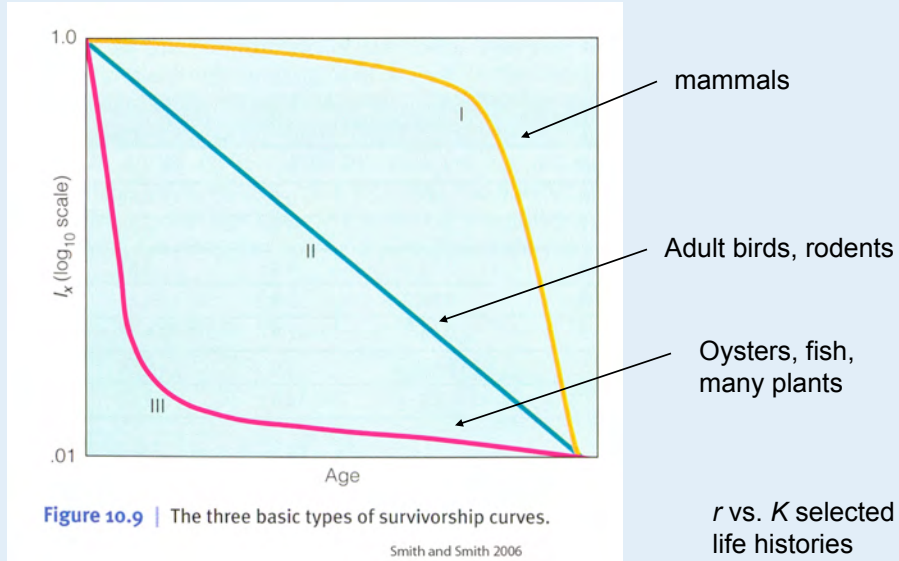
**Generation Time =  $G$**

$G = \frac{\sum l(x)b(x)x}{\sum l(x)b(x)}$	= 1.483 years
$r$ (estimated) = $\ln(R_0)/G$	= 0.718 individuals/(individual • year)
Correction added to estimated $r$	= 0.058
$r$ (Euler)	= 0.776 individuals/(individual • year)

<sup>a</sup> The  $x$ ,  $S(x)$ , and  $b(x)$  columns are supplied. All others are calculated from these.

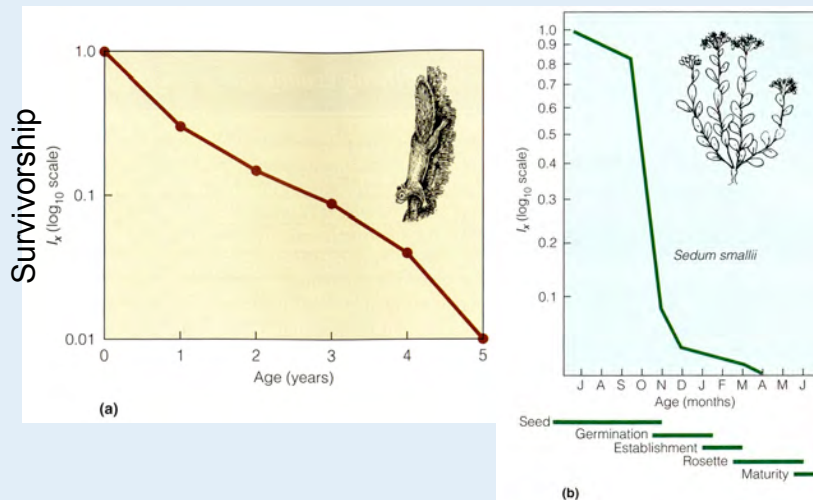
Demography

## Idealized Survivorship Curves



Demography

## Example Survivorship Curves



**Figure 10.7** | Survivorship curve for (a) gray squirrel based on Table 10.1 and (b) *Sedum smallii* based on Table 10.3.

Smith and Smith 2006

## Age Specific Survival and Fertilities for Leslie Matrix

**Table 3.2** Calculation of age-specific survival probabilities and fertilities for the Leslie matrix. Data from Table 3.1. Notice that the first row of the table is blank for  $P_i$  and  $F_i$ , because we begin counting age classes at 1, not 0.

$x$	$i$	$l(x)$	$b(x)$	$P_i = l(i)/l(i-1)$	$F_i = b(i)P_i$
0		1.0	0		
1	1	0.8	2	0.80	1.60
2	2	0.4	3	0.50	1.50
3	3	0.1	1	0.25	0.25
4	4	0	0	0.00	0.00

The resulting Leslie matrix is:

$$A = \begin{bmatrix} 1.6 & 1.5 & 0.25 & 0 \\ 0.8 & 0 & 0 & 0 \\ 0 & 0.5 & 0 & 0 \\ 0 & 0 & 0.25 & 0 \end{bmatrix}$$

**Table 3.1** Standard life-table calculations.<sup>a</sup>

$x$	$S(x)$	$b(x)$	$l(x) = S(x)/S(0)$	$g(x) = l(x+1)/l(x)$	$l(x)b(x)$	$l(x)b(x)x$	Initial estimate $e^{-rx}l(x)b(x)$	Corrected estimate $e^{-rx}l(x)b(x)$
0	500	0	1.0	0.80	0.0	0.0	0.000	0.000
1	400	2	0.8	0.50	1.6	1.6	0.780	0.736
2	200	3	0.4	0.25	1.2	2.4	0.285	0.254
3	50	1	0.1	0.00	0.1	0.3	0.012	0.010
4	0	0	0.0		0.0	0.0	0.000	0.000

Gotelli

### Demography

## Dynamic vs. Time-Specific Life Table

### Dynamic life table

Follows a cohort through time.

### Time-specific Life

Sample a population at a given time to obtain a distribution of age classes.

Assumes

- (1) Sample each age class in proportion to its numbers
- (2) constant age specific demographic rates.

Demography

### Life Table Analysis

$n_x$  = Number of individuals  $l_x$  = Survivorship  $d_x$  = Number dead

$q_x$  = Age-specific death rate

Table 10.1 | Gray Squirrel Life Table

$x$	$n_x$	$l_x$	$d_x$	$q_x$
0	530	1.0	371	0.7
1	159	0.3	79	0.5
2	80	0.15	32	0.4
3	48	0.09	27	0.55
4	21	0.04	16	0.75
5	5	0.01	5	1.0

Smith and Smith 2006

Demography

### Birth Rate & Fecundity Tables

Table 10.4 | Gray Squirrel Fecundity Table

$x$	$l_x$	$b_x$	$l_x b_x$
0	1.0	0.0	0.00
1	0.3	2.0	0.60
2	0.15	3.0	0.45
3	0.09	3.0	0.27
4	0.04	2.0	0.08
5	0.01	0.0	0.00
$\Sigma$		10.0	1.40

Mean number of females born in each age group

Survivorship

Birth rate (per female)

Gross reproductive rate

Net Reproductive Rate  
If greater than 1 the females are replacing themselves

## Projecting Population Growth

Table 10.6 | Population Projection Table, Squirrel Population

Age	Year (t)										
	0	1	2	3	4	5	6	7	8	9	10
0	20	27	34.1	40.71	48.21	58.37	70.31	84.8	101.86	122.88	148.06
1	10	6	8.1	10.23	12.05	14.46	17.51	21.0	25.44	30.56	36.86
2	0	5	3.0	4.05	5.1	6.03	7.23	8.7	10.50	12.72	15.28
3	0	0	3.0	1.8	2.43	3.06	3.62	4.4	5.22	6.30	7.63
4	0	0	0	1.35	0.81	1.09	1.38	1.6	1.94	2.35	2.83
5	0	0	0	0	0.33	0.20	0.27	0.35	0.40	0.49	0.59
Total $N(t)$	30	38	48.2	58.14	68.93	83.21	100.32	120.85	145.36	175.30	211.25
Lambda	$\lambda$	1.27	1.27	1.21	1.19	1.21	1.20	1.20	1.20	1.20	1.20

See description on p. 211,  
Smith and Smith 2006

$$\lambda = \frac{N(t+1)}{N(t)} \quad \text{Finite multiplication rate} \quad \begin{aligned} \lambda &= e^r \\ r &= \ln \lambda \end{aligned}$$

$$N(t) = N(0)\lambda^t$$

## Summary

Discussed a way of building **demographic information** and **life history** into our models

Replaced a single state-variable with **multiple state variables**

- describe the dynamics of subpopulations

Analyzed matrix models do determine the

- **geometric growth rate**
- **stable population distribution**, and
- **reproductive value**

Suggested that matrix population models could be used in more general ways, including to represent **meta-populations** or **spatial structure**

## Coming Up

Exam

Spatial Models

Two state-variable models

- Competition
- Consumer-resource dynamics
- Ecosystem vs. population point of view

Three state-variable models

Additional Notes

## Activity

In teams work on Problem 3.2

{ (Tyler, Nikolai, Matt), (Andria, Liz), (Lisa, Ali) }

3.2. Here is a set of hypothetical life-table data for a population of snails:

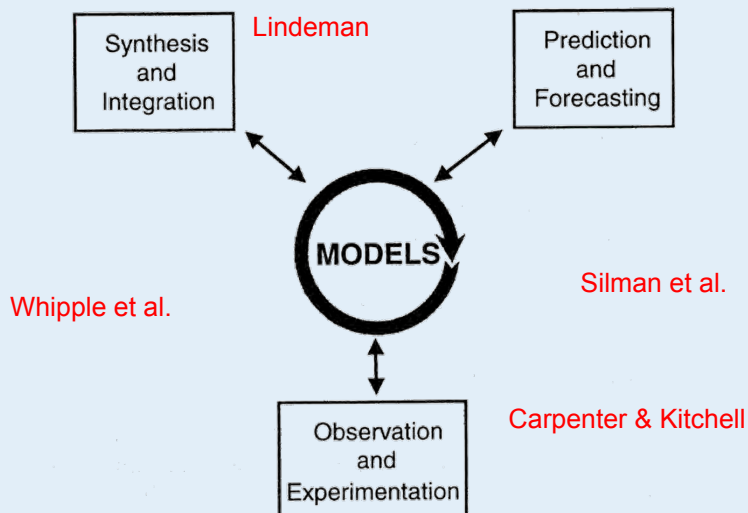
Age in years ( $x$ )	$S(x)$	$b(x)$
0	500	0.0
1	400	2.5
2	40	3.0
3	0	0.0

- Complete the life-table analysis by calculating  $l(x)$ ,  $g(x)$ ,  $R_0$ ,  $G$ , and the estimate of  $r$ . Calculate the exact value of  $r$  with the Euler equation.
- Determine the stable age and reproductive value distributions for this life table.

Use R or Excel for life table construction

Can use Biodem package to acquire matrix multiplication or use loop

## Model Use



## Analysis Assumptions

These matrix analysis assume that **A** is

**Positive** (non-negative)

all elements of **A** are  $\geq 0$

**Irreducible**

If  $(\mathbf{I} + \mathbf{A})^{s-1}$  is positive, where  $s$  is size of **A**

**Primitive**

Iff  $\mathbf{A}^{(s^2-2s+2)}$  is positive

See Caswell 2001 Chapter 4 for more detail and consequences of violating these assumptions.

These analyses assume that **A** is a primitive, irreducible, and non-negative ma

**A** is primitive iff  $\mathbf{A}^{s^2-2s+2}$  is positive