

Model 3, negative assortative mating (incl. 3.20 of EK)

18.1

Assumptions 2-4 again the same. The first is now

- * female AA mate only male aa

- * female aa mate only male AA

- * female Aa mate male AA and aa with equal preference

Remark: one could make other assumptions and still call it neg. ass. mating, e.g., only exclude mating of the same genotype.

Mating table:

		Male		
		AA	Aa	aa
		Un	Vn	Wn
Female	AA	Un	0	Un
	Aa	Vn	$V_n \cdot \frac{U_n}{U_n + W_n}$	0
	aa	Wn	Wn	0

row sums must be U_n, V_n, W_n

← sum is V_n , elements are proportional to U_n, W_n (relative freq.)

Corresponding freq. of AA, Aa, aa:

$$\begin{cases} - & - & 0,1,0 \\ \frac{1}{2}, \frac{1}{2}, 0 & - & 0, \frac{1}{2}, \frac{1}{2} \\ 0,1,0 & - & - \end{cases}$$

For the next generation:

$$\begin{cases} U_{n+1} = \frac{1}{2} \frac{U_n V_n}{U_n + W_n} \\ V_{n+1} = U_n + W_n + \frac{1}{2} V_n \left(\frac{U_n}{U_n + W_n} + \frac{W_n}{U_n + W_n} \right) = U_n + \frac{1}{2} V_n + W_n \\ W_{n+1} = \frac{1}{2} \frac{V_n W_n}{U_n + W_n} \end{cases} \quad \text{non-linear system}$$

Note: $V_{n+1} = \underbrace{U_n + W_n}_{1-V_n} + \frac{1}{2} V_n = 1 - \frac{1}{2} V_n \Rightarrow V_{n+1} + \frac{1}{2} V_n = 1$, can be solved

Homogeneous solution is $V_{n,h} = C \cdot \left(-\frac{1}{2}\right)^n$ (solves $V_{n+1} + \frac{1}{2} V_n = 0$),

a particular solution is $V_{n,p} = \frac{2}{3}$ ($\Rightarrow V_{n+1} + \frac{1}{2} V_n = \frac{2}{3} + \frac{1}{2} \cdot \frac{2}{3} = 1$) \Rightarrow

$$V_n = \frac{2}{3} + C \cdot \left(-\frac{1}{2}\right)^n. \quad n=0 \Rightarrow V_0 = \frac{2}{3} + C \Rightarrow C = V_0 - \frac{2}{3} \quad \text{and} \quad V_n = \frac{2}{3} + (V_0 - \frac{2}{3}) \left(-\frac{1}{2}\right)^n \quad (*)$$

Observe $V_n \rightarrow \frac{2}{3}$, $n \rightarrow \infty$, independent of V_0 (and U_0, W_0)!

One could now use $U_{n+1} = \frac{1}{2} \frac{U_n V_n}{U_n + W_n} = \frac{1}{2} U_n \frac{V_n}{1-V_n}$ and plug in $(*)$ for V_n ,

to get a linear but not constant coefficient equation for U_n , not easy to solve.

Instead, we use $W_n = 1 - V_n - U_n$ and analyze the 2D system for U_n and V_n :

$$\begin{cases} U_{n+1} = \frac{1}{2} \frac{U_n V_n}{1 - V_n} \\ V_{n+1} = 1 - \frac{1}{2} V_n \end{cases}$$

Steady states: $\begin{cases} \bar{U} = \frac{1}{2} \frac{\bar{U} \bar{V}}{1 - \bar{V}} \\ \bar{V} = 1 - \frac{1}{2} \bar{V} \end{cases} \Rightarrow \bar{V} = \frac{2}{3} \Rightarrow \bar{U} = \frac{1}{2} \underbrace{\frac{\bar{U} \cdot \frac{2}{3}}{1 - \bar{U}}}_{=\bar{U}}$ always satisfied

$\Rightarrow (\bar{U}, \frac{2}{3})$ are steady states for all \bar{U} . For the 3D system $\bar{W} = 1 - \bar{U} - \bar{V} = \frac{1}{3} - \bar{U}$ so $(\bar{U}, \frac{2}{3}, \frac{1}{3} - \bar{U})$ are steady states.

Stability:

$$J(u, v) = \begin{pmatrix} \frac{1}{2} \frac{v}{1-v} & \frac{u}{2(1-v)^2} \\ 0 & -\frac{1}{2} \end{pmatrix} \Rightarrow J(\bar{U}, \frac{2}{3}) = \begin{pmatrix} 1 & \frac{9\bar{U}}{2} \\ 0 & -\frac{1}{2} \end{pmatrix} \Rightarrow \underbrace{\lambda_1 = 1}_{\text{critical}}, \underbrace{\lambda_2 = -\frac{1}{2}}_{\text{OK for stability}} \quad |\lambda_2| < 1$$

Observe:

$$\frac{U_{n+1}}{W_{n+1}} = \frac{\frac{1}{2} \frac{U_n V_n}{U_n + W_n}}{\frac{1}{2} \frac{V_n W_n}{U_n + W_n}} = \frac{U_n}{W_n} \Rightarrow \frac{U_n}{W_n} = \text{constant} = \frac{U_0}{W_0} = \alpha$$

For steady state $\frac{\bar{U}}{\bar{W}} = \frac{\bar{U}}{\frac{1}{3} - \bar{U}} = \alpha \Rightarrow \frac{\bar{U}}{\alpha} = \frac{1}{3} - \bar{U} \Rightarrow \bar{U} = \frac{1}{3(1+\frac{1}{\alpha})} = \frac{1}{3(1+\frac{W_0}{U_0})} = \frac{U_0}{3(U_0 + W_0)}$

and $\bar{W} = \frac{1}{3} - \bar{U} = \frac{W_0}{3(U_0 + W_0)} \Rightarrow$

$$(\bar{U}, \bar{V}, \bar{W}) = \left(\frac{U_0}{3(U_0 + W_0)}, \frac{2}{3}, \frac{W_0}{3(U_0 + W_0)} \right), \text{ steady states for all } U_0, W_0$$

Small changes in U_0, W_0 give small (but non-zero) changes in \bar{U} and \bar{W}
 \Rightarrow steady states \sim neutral.

(Compare discrete SIR model which also has $\lambda_1 = 1$)

Tests	$n=0$	1	2	3	∞	$n=0$	1	2	∞
U_n	$\frac{1}{3}$	$\frac{1}{2}$	$\frac{5}{24}$	$\frac{7}{48}$	$\dots \rightarrow \frac{1}{6}$	0.2	0.1	0.15	$\dots \rightarrow \frac{3}{15}$
V_n	$\frac{1}{3}$	$\frac{5}{6}$	$\frac{7}{12}$	$\frac{15}{24}$	$\dots \rightarrow \frac{2}{3}$	0.5	0.75	0.625	$\dots \rightarrow \frac{2}{3}$
W_n	$\frac{1}{3}$	$\frac{1}{2}$	$\frac{5}{24}$	$\frac{7}{48}$	$\dots \rightarrow \frac{1}{6}$	0.3	0.15	0.225	$\dots \rightarrow \frac{3}{15}$

Note: p_n and q_n are not constant in this model

$$p_0 = 0.2 + \frac{0.5}{2} = 0.45, p_1 = 0.1 + \frac{0.75}{2} = 0.475, \dots p_n \rightarrow \frac{2}{15} + \frac{1}{3} = \frac{7}{15} \approx 0.467$$

Age structures of populations and Leslie matrices

18.3

Suppose there are m age classes in a population with $P_{1,n}, P_{2,n}, \dots, P_{m,n}$ individuals in the classes at time n (n could be year for humans, $m \approx 100$, or each class could be a 10-year interval).

Let α_j = number of births per unit time from each individual in class j , $\alpha_j \geq 0$
 σ_j = fraction of class j that survives to class $j+1$, $0 < \sigma_j \leq 1$ assumed.

At time $n+1$ we have

$$\text{newborns } P_{1,n+1} = \alpha_1 P_{1,n} + \alpha_2 P_{2,n} + \dots + \alpha_m P_{m,n}$$

$$\text{survivors } P_{j,n+1} = \sigma_{j-1} P_{j-1,n}, \quad j = 2, \dots, m$$

$$\text{Write } \bar{P}_n = \begin{pmatrix} P_{1,n} \\ \vdots \\ P_{m,n} \end{pmatrix} \text{ and } A = \begin{pmatrix} \alpha_1 \alpha_2 \dots \alpha_{m-1} \alpha_m \\ \sigma_1 0 \dots 0 0 \\ 0 \sigma_2 \dots 0 0 \\ \vdots \vdots \vdots \vdots \\ 0 0 \dots \sigma_{m-1} 0 \end{pmatrix} \quad (m \times m \text{ matrix})$$

\Rightarrow we have the linear discrete system

$$\bar{P}_{n+1} = A \bar{P}_n \quad A \text{ is called a } \underline{\text{Leslie matrix}}$$

If A has eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_m$ with eigenvectors $\bar{v}_1, \bar{v}_2, \dots, \bar{v}_m$,
the solutions are

$$\bar{P}_n = c_1 \lambda_1^n \bar{v}_1 + c_2 \lambda_2^n \bar{v}_2 + \dots + c_m \lambda_m^n \bar{v}_m$$

(it is also possible to write down \bar{P}_n in the case that A has $< m$ eigenvectors)

Suppose that λ_1 has the largest absolute value, $|\lambda_1| \geq |\lambda_j|, j=2, \dots, m$. Then

$$\frac{1}{\lambda_1^n} \bar{P}_n = c_1 \bar{v}_1 + c_2 \underbrace{\left(\frac{\lambda_2}{\lambda_1}\right)^n}_{\rightarrow 0, n \rightarrow \infty} \bar{v}_2 + \dots + c_m \underbrace{\left(\frac{\lambda_m}{\lambda_1}\right)^n}_{\rightarrow 0, n \rightarrow \infty} \bar{v}_m \rightarrow c_1 \bar{v}_1, \quad n \rightarrow \infty \quad \Rightarrow$$

$$\bar{P}_n \approx c_1 \lambda_1^n \bar{v}_1 \text{ if } n \text{ large}$$

\bar{v}_1 is called the stable age distribution, its components show the number of individuals in each age class and can be used to predict future age distribution in a population. It should have all its components ≥ 0 and λ_1 real positive to be realistic.

Perron-Frobenius theory for matrices A with all elements $a_{ij} > 0$ or $a_{ij} \geq 0$ can be applied (positive/non-negative matrices).

If $\alpha_m > 0$ (may be not the case for humans) and if, e.g., two consecutive λ_j, λ_{j+1} are > 0 , then A is an "irreducible" and "primitive" non-negative matrix and a theorem of Frobenius says:

A has a real eigenvalue $\lambda_1 > 0$ with $\lambda_1 > |\lambda_j|$ for $j=2, \dots, m$. The eigenvector \bar{v}_1 of λ_1 has all its components > 0 (and no other eigenvector \bar{v}_j has all components ≥ 0).

This theory fits our model, it is also the theory behind Google's PageRank and many results about Markov chains. If $\alpha_m = 0$ theorems guarantee slightly weaker results.

Example with $m=4$

$$A = \begin{pmatrix} 0 & 0.5 & 0.7 & 0.1 \\ 0.9 & 0 & 0 & 0 \\ 0 & 0.9 & 0 & 0 \\ 0 & 0 & 0.8 & 0 \end{pmatrix}$$

Survival is 90% in first 2 steps, then 80%.
Births: most (0.7) in class 3, 0.5 in class 2,
few (0.1) in class 4, none in class 1.

One finds

$$\lambda_1 \approx 1.03, \bar{v}_1 \approx \underbrace{\begin{pmatrix} 0.31 \\ 0.27 \\ 0.24 \\ 0.18 \end{pmatrix}}_{\substack{\text{real positive} \\ \text{all elements } > 0 \\ \text{scaled to have sum 1}}} , \quad \lambda_2 \approx -0.13, \bar{v}_2 \approx \underbrace{\begin{pmatrix} 0.00 \\ 0.02 \\ -0.15 \\ 0.99 \end{pmatrix}}_{\substack{\text{mixed signs}}}, |\lambda_2| < \lambda_1$$

$$\lambda_{3,4} \approx -0.45 \pm 0.54i, \bar{v}_{3,4} \text{ complex}, |\lambda_{3,4}| \approx 0.7 < \lambda_1$$

Solutions to $\bar{P}_{n+1} = A\bar{P}_n$ are

$$\bar{P}_n = c_1 (1.03)^n \bar{v}_1 + c_2 (-0.13)^n \bar{v}_2 + \underbrace{c_3 \lambda_3^n \bar{v}_3 + c_4 \lambda_4^n \bar{v}_4}_{\substack{\rightarrow 0 \text{ fast} \\ \text{can be written on real form} \\ \rightarrow 0 \text{ quite fast}}}$$

For large n , $\bar{P}_n \approx c_1 (1.03)^n \begin{pmatrix} 0.31 \\ 0.27 \\ 0.24 \\ 0.18 \end{pmatrix}$, means that population grows

by 3% per unit time, and there will be approx. 31%, 27%, 24%, and 18%, respectively, of the population in the different age classes.