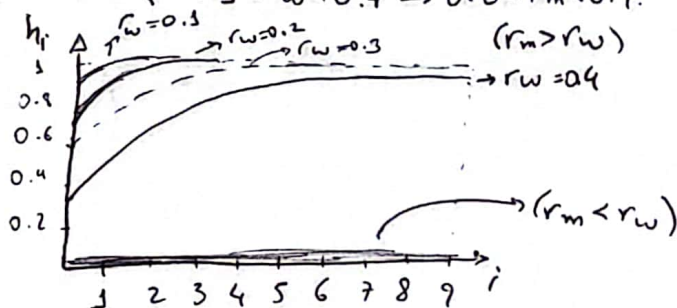


① Suppose a cell colony is described by the (const. pop.) Moran process. Suppose that if  $r_w$  is the division rate of the wild type cells and  $r_m$  is the division rate of mutants. Then the prob. of mutant fixation (mutants taking over) starting from  $i$  mutant cells out of  $N$  cells is given by:

$$h_i = \frac{1 - \left(\frac{r_w}{r_m}\right)^i}{1 - \left(\frac{r_w}{r_m}\right)^N}$$

a) Assume that  $r_m > r_w$ . Calculate the prob. of mut. fix.  $h_i$  for  $0 < i \leq 10$  for very large pop sizes (take  $\lim_{N \rightarrow \infty} h_i$ ). Plot this as a function of  $i$ .  
b) The same with  $r_m < r_w$ .

$\Rightarrow$  a) When  $r_m > r_w$ ,  $\frac{r_w}{r_m} < 1$ . Because of this,  $\lim_{N \rightarrow \infty} \left(\frac{r_w}{r_m}\right)^N = 0$ , so as  $N$  goes to  $\infty$ , the denominator goes to 1, so our fixation probability becomes  $h_i = 1 - \left(\frac{r_w}{r_m}\right)^i$ . As  $r_m + r_w = 1$ , the plot below was done for  $0.1 < r_w < 0.4 \Rightarrow 0.6 < r_m < 0.9$ .



b) When  $r_m < r_w$ ,  $\frac{r_w}{r_m} > 1$ , so  $\lim_{N \rightarrow \infty} \left(\frac{r_w}{r_m}\right)^N = \infty$ . Our denominator becomes  $\lim_{N \rightarrow \infty} (1 - \left(\frac{r_w}{r_m}\right)^N) = -\infty$ .

Because the numerator remains finite,  $\lim_{N \rightarrow \infty} h_i = \lim_{N \rightarrow \infty} \frac{1 - \left(\frac{r_w}{r_m}\right)^i}{1 - \left(\frac{r_w}{r_m}\right)^N} = 0$ . Therefore the fixation prob. goes to 0 for any  $i$  when mutant cells divide less than wildtype cells.

② Consider a Moran process with two types: the wild type with fitness  $r_w$  and a mutant with fitness  $r_m$ . Suppose the mutant can experience a back mutation, where, as a result of a mutant cell division, a wild type daughter cell is created. We assume that this mutation happens with prob.  $\beta$  every time a mutant divides, and no other mutations happen in the system. Consider the state space  $j \in \{0, 1, \dots, N\}$ , where  $j$  is the number of mutants. a) Write down the probs.  $\text{Prob}(j \rightarrow j+1)$  and  $\text{Prob}(j \rightarrow j-1)$  in a single step. b) What are the absorbing states of this system?

$\Rightarrow$  a) To go from  $j \rightarrow j+1$ , we have to consider the prob. that a wild-type dies, which is  $\frac{N-j}{N}$ , the prob. that a mutant divides, which is  $\frac{r_m j}{r_m j + r_w(N-j)}$ , but also consider that there is the probability for the mutant cell to generate a mutant daughter, which is  $(1-\beta)$ . Therefore,  $\text{Prob}(j \rightarrow j+1) = \left(\frac{r_m j}{r_m j + r_w(N-j)}\right) (1-\beta) \left(\frac{N-j}{N}\right)$ .

For  $j \rightarrow j-1$ , there is the prob. that a mutant cell is chosen to die, and that a mutant cell dies and a wild-type is chosen to divide. The full prob. is given by:  $\text{Prob}(j \rightarrow j-1) = \left(\frac{r_m j}{r_m j + r_w(N-j)}\right) \beta \left(\frac{j}{N}\right) + \left(\frac{r_w(N-j)}{r_m j + r_w(N-j)}\right) \left(\frac{j}{N}\right)$ .

b) In the Moran process with back mutation ( $\beta > 0$ ), the only absorbing state will be at  $j=0$ , as no new mutant cells will be created.

③ Suppose a Markov Chain has states 0, 1, 2, 3. Find the mean time to reach state 3 starting from state 0 for the Markov Chain whose transition matrix is given by

$$P = \begin{matrix} & \begin{matrix} 0 & 1 & 2 & 3 \end{matrix} \\ \begin{matrix} 0 \\ 1 \\ 2 \\ 3 \end{matrix} & \begin{pmatrix} 0.4 & 0.3 & 0.2 & 0.1 \\ 0 & 0.7 & 0.2 & 0.1 \\ 0 & 0 & 0.9 & 0.1 \\ 0 & 0 & 0 & 1 \end{pmatrix} \end{matrix}$$

⇒ Let  $m_{ij}$  be the expected time to get to  $j$  starting from  $i$ :

$$\begin{cases} m_{03} = 0.1 + 0.4(1 + m_{03}) + 0.3(1 + m_{13}) + 0.2(1 + m_{23}) \\ m_{13} = 0.1 + 0.7(1 + m_{13}) + 0.2(1 + m_{23}) \\ m_{23} = 0.1 + 0.9(1 + m_{23}) \end{cases}$$

$$0.7(1 + m_{13}) + 0.2(1 + m_{23}) \Rightarrow m_{13} = 0.1 + 0.7 + 0.7m_{13} + 0.2 + 0.2(10) \Rightarrow 0.3m_{13} = 1 + 2 = 3$$

$$m_{13} = 10$$

$$0.9(1 + m_{23}) \Rightarrow m_{23} = 0.1 + 0.9 + 0.9m_{23} \Rightarrow 0.1m_{23} = 1 \Rightarrow m_{23} = 10$$

$$\rightarrow m_{03} = 0.1 + 0.4 + 0.4m_{03} + 0.3 + 0.3(10) + 0.2 + 0.2(10) \Rightarrow 0.6m_{03} = 1 + 3 + 2 = 6 \rightarrow m_{03} = 10$$

⇒ The mean time to reach state starting at 0 is 10.

④ Suppose we have a continuous time birth process with immigration. That is, in time  $\Delta t$ ,  $\text{Prob}(j \rightarrow j+1) =$

$(L_j + a)\Delta t$ ,  $\text{Prob}(j \rightarrow j) = 1 - (L_j + a)\Delta t$ , and all other processes have prob. zero. Denote by  $\varphi_j(t)$  the prob. to have  $j$  individuals at time  $t$ . Assume that, initially, there are zero individuals. a) Write down the master equation for  $\frac{d\varphi_j(t)}{dt}$ , including the equation for  $j=0$  and the initial conditions for all  $\varphi_j(t)$ . b) Find  $\varphi_j(t)$  for  $j=0, 1, 2, \dots$  c) Derive the ODE for the mean number of individuals,  $\langle X \rangle = \sum_{j=0}^{\infty} \varphi_j(t)j$ , and the initial condition. d) Solve this ODE to find  $\langle X \rangle$  as a function of time.

⇒ a) Since we have migration but we don't have death ( $\text{Prob}(j \rightarrow j-1) = 0$ ), we can write  $\frac{d}{dt} \varphi_j(t)$  as

$$\varphi_{j-1}(L_{j-1} + a) - \varphi_j(L_j + a) \Rightarrow j \geq 1. \text{ For } j=0, \frac{d}{dt} \varphi_0(t) = -a\varphi_0(t), \text{ and we have that } \varphi_0(0) = 1, \varphi_k(0) = 0 \quad \forall k \neq 0$$

$$\begin{aligned} \text{b) } \frac{d}{dt} \varphi_1(t) &= \varphi_0(L_0 + a) - \varphi_1(L_1 + a) = \varphi_0(a) - \varphi_1(L_1 + a) \\ &= ae^{-at} - \varphi_1(L_1 + a) \rightarrow \text{non-homogeneous linear dif. eq.} \end{aligned}$$



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(4) b) cont.:  $\frac{d}{dt} \varphi_1(t) = a e^{-at} - \varphi_1(L+a) \Rightarrow \frac{d}{dt} \varphi_1(t) + \varphi_1(L+a) = a e^{-at}$ . To solve this we use the integrating factor given by  $\mu(t) = e^{\int (L+a) dt} = e^{(L+a)t}$ . Now, multiplying both sides of the equation by the factor:  $e^{(L+a)t} \frac{d}{dt} \varphi_1(t) + e^{(L+a)t} (L+a) \varphi_1(t) = a e^{(L+a)t} e^{-at} \Rightarrow \frac{d}{dt} (\varphi_1(t) e^{(L+a)t}) = \int a e^{Lt} dt$

$$\Rightarrow \varphi_1(t) e^{(L+a)t} = \frac{a}{L} e^{Lt} + C \Rightarrow \varphi_1(t) = \frac{a}{L} e^{Lt} e^{-(L+a)t} + C e^{-(L+a)t} = \frac{a}{L} e^{-at} + C e^{-(L+a)t}$$

Given that  $\varphi_1(0) = 0 \Rightarrow \varphi_1(0) = \frac{a}{L} e^0 + C e^0 = \frac{a}{L} + C = 0 \rightarrow C = -\frac{a}{L} \Rightarrow \varphi_1(t) = \frac{a}{L} e^{-at} - \frac{a}{L} e^{-(L+a)t} = \frac{a}{L} (e^{-at} - e^{-(L+a)t})$

(This will get complicated to calculate recursively, so let's use the mean number of individuals instead.)

c) The mean number of individuals is given by  $\langle X \rangle = \sum_{j=0}^{\infty} \varphi_j(t) \cdot j$ , so taking the time derivative we get to:  $\frac{d\langle X \rangle}{dt} = \sum_{j=0}^{\infty} j \frac{d\varphi_j(t)}{dt} = \sum_{j=0}^{\infty} j [\varphi_{j-1}(L(j-1)+a) - \varphi_j(Lj+a)]$

$$= \sum_{j=1}^{\infty} j (L(j-1)+a) \varphi_{j-1} - \sum_{j=0}^{\infty} j (Lj+a) \varphi_j = \sum_{j=1}^{\infty} j L(j-1) \varphi_{j-1} + \sum_{j=1}^{\infty} j a \varphi_{j-1} - \sum_{j=0}^{\infty} L j^2 \varphi_j - \sum_{j=0}^{\infty} j a \varphi_j$$

$$= \sum_{j=0}^{\infty} L j^2 \varphi_j - \sum_{j=0}^{\infty} L j \varphi_j + a \sum_{j=0}^{\infty} \varphi_j - a \sum_{j=0}^{\infty} j \varphi_j$$

Using that  $\sum_{j=0}^{\infty} \varphi_j(t) = 1$  and  $\sum_{j=0}^{\infty} j \varphi_j(t) = \langle X \rangle$ , we

can rewrite the above as  $\frac{d\langle X \rangle}{dt} = L\langle X \rangle + a - a\langle X \rangle = (L-a)\langle X \rangle + a$

d) To solve the ODE  $\frac{d\langle X \rangle}{dt} = (L-a)\langle X \rangle + a$ , we have the general solution  $\langle X \rangle(t) =$

$$C e^{(L-a)t} + \frac{a}{L-a} \rightarrow \langle X \rangle(0) = 0 \Rightarrow C + \frac{a}{L-a} = 0 \rightarrow C = -\frac{a}{L-a}$$

$$\Rightarrow \langle X \rangle(t) = -\frac{a}{L-a} e^{(L-a)t} + \frac{a}{L-a} = \frac{a}{L-a} (1 - e^{(L-a)t})$$