

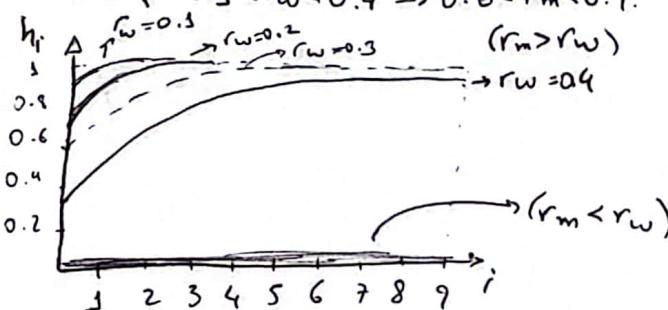
① Suppose a cell colony is described by the (const. pop.) Moran process. Suppose that if r_w is the division rate of the wild type cells and r_m is the division rate of mutants. Then the prob. of mutant fixation (mutants taking over) starting from i mutant cells out of n cells is given by:

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

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

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

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

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

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

$$\left\{ \begin{array}{l} m_{03} = 0.1 + 0.4(1 + m_{03}) + 0.3(1 + m_{13}) + 0.2(1 + m_{23}) \\ m_{13} = 0.1 + 0.7(1 + m_{13}) + 0.2(1 + m_{23}) \\ m_{23} = 0.1 + 0.9(1 + m_{23}) \end{array} \right.$$

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

$$m_{13} = 10$$

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

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

⇒ The mean time to reach state 3 starting at 0 is 10.

④ Suppose we have a continuous time birth process with immigration. That is, in time Δt , $\text{Prob}(j \rightarrow j+1) = (L_j + a)\Delta t$, $\text{Prob}(j \rightarrow j) = 1 - (L_j + a)\Delta t$, 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 $\frac{d\varphi_j(t)}{dt}$ as

$$\varphi_{j-1}(L_{j-1} + a) - \varphi_j(L_j + a) \Rightarrow j \geq 1. \text{ For } j=0, \frac{d\varphi_0(t)}{dt} = \underset{k \neq 0}{\cancel{\varphi_1(t)}} - \varphi_0(L_0 + a), \text{ and we have that } \varphi_0(0) = 1, \varphi_k(0) = 0 \forall k \neq 0$$

$$\text{b) } \frac{d}{dt} \varphi_1(t) = \varphi_0(L_0 + a) - \varphi_1(L_1 + a) = \varphi_0(a) - \varphi_1(L + a) \Rightarrow \varphi_1(t) = e^{-at} \varphi_0(t)$$

$$= ae^{-at} - \varphi_1(L + a) \rightsquigarrow \text{non-homogeneous linear dif. eq.}$$

MATH112B - HW6 (Raphael F. Levy - U10356477)

(4) b) cont.: $\frac{d}{dt} \varphi_1(t) = ae^{-at} - \varphi_1(L+a) \Rightarrow \frac{d}{dt} \varphi_1(t) + \varphi_1(L+a) = ae^{-at}$. 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: $e^{(L+a)t} \frac{d\varphi_1(t)}{dt} + e^{(L+a)t} (L+a)\varphi_1(t) = ae^{(L+a)t} e^{-at} \Rightarrow \frac{d}{dt} (\varphi_1(t)e^{(L+a)t}) = ae^{Lt}$

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

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

↳ 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: $\frac{d\langle X \rangle}{dt} = \sum_{j=0}^{\infty} j \frac{d\varphi_j(t)}{dt} = \sum_{j=0}^{\infty} j [\varphi_{j-1}(L(j-1)+a) - \varphi_j(Lj+a)]$

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

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

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 $\langle X \rangle(t) =$

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

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