

# LD with Mutator state Switching

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We have 2 distinct states (mutator and non mutator) where rates of switching are  $r_{ij}$  and growth rate is identical  $\lambda$ . Say state 2 is the mutator state.

Equilibrium fraction of state 2:

$$\hat{f}_2 = \frac{r_{12}}{r_{12} + r_{21}}; \hat{f}_1 = 1 - \hat{f}_2$$

The absolute number of bacteria:

$$N(t) = N_1(t) + N_2(t) = N_0 e^{\lambda t}$$

Where  $\lambda$  is growth rate, assumed to be identical for both states.

Assume we sample  $N$  individuals, and make each the origin of a distinct colony. The time dependent fraction of state 2 in a colony is:

$$f_i(t) = \frac{N_i(t)}{N(t)}$$

The rate equation for it:

$$\frac{df_2(t)}{dt} = r_{12}(1 - f_2(t)) - r_{21}f_2(t)$$

Solutions depend on the IC of state of originator:

$$f_2(t) = \begin{cases} \hat{f}_2 + (1 - \hat{f}_2)e^{-(r_{12}+r_{21})t} & p = \hat{f}_2 \\ \hat{f}_2(1 - e^{-(r_{12}+r_{21})t}) & p = 1 - \hat{f}_2 \end{cases}$$

$$\langle f_2(t) \rangle = \hat{f}_2 \left( \hat{f}_2 + (1 - \hat{f}_2)e^{-(r_{12}+r_{21})t} \right) + (1 - \hat{f}_2) \hat{f}_2(1 - e^{-(r_{12}+r_{21})t}) = \hat{f}_2$$

As can be expected. As in LD, average number of mutation events  $m$ :

$$\frac{d\langle m(t) \rangle}{dt} = \mu N_2(t) = \mu f_2(t) e^{\lambda t}$$

Integrating:

$$\langle m(t) \rangle = \mu \times \begin{cases} \frac{\hat{f}_2}{\lambda} (e^{\lambda t} - 1) + \frac{1}{r_{12} + r_{21} - \lambda} (1 - \hat{f}_2) (1 - e^{-(r_{12} + r_{21} - \lambda)t}) & p = \hat{f}_2 \\ \hat{f}_2 \left( \frac{1}{\lambda} (e^{\lambda t} - 1) - \frac{1}{r_{12} + r_{21} - \lambda} (1 - e^{-(r_{12} + r_{21} - \lambda)t}) \right) & p = 1 - \hat{f}_2 \end{cases}$$

And we can approximate:

$$\langle m(t) \rangle = \mu \times \begin{cases} \hat{f}_2 \frac{e^{\lambda t}}{\lambda} + (1 - \hat{f}_2) \frac{1 - e^{-(r_{12} + r_{21} - \lambda)t}}{r_{12} + r_{21} - \lambda} & p = \hat{f}_2 \\ \hat{f}_2 \left( \frac{e^{\lambda t}}{\lambda} - \frac{1 - e^{-(r_{12} + r_{21} - \lambda)t}}{r_{12} + r_{21} - \lambda} \right) & p = 1 - \hat{f}_2 \end{cases}$$

Now average again on IC to get the true average across colonies:

$$\begin{aligned} \langle m(t) \rangle &= \mu \times \left( \hat{f}_2^2 \frac{e^{\lambda t}}{\lambda} + (\hat{f}_2 - \hat{f}_2^2) \frac{1 - e^{-(r_{12} + r_{21} - \lambda)t}}{r_{12} + r_{21} - \lambda} + (\hat{f}_2 - \hat{f}_2^2) \left( \frac{e^{\lambda t}}{\lambda} - \frac{1 - e^{-(r_{12} + r_{21} - \lambda)t}}{r_{12} + r_{21} - \lambda} \right) \right) \\ &= \frac{\mu}{\lambda} \hat{f}_2 e^{\lambda t} \end{aligned}$$

Now lets say we run an experiment to time  $t_f$ . The expected number of survivors at time  $t_f$  originating from mutations that happened at time  $t$  is:

$$\langle S(t) \rangle = \langle m(t) \rangle e^{\lambda(t_f - t)} = \frac{\mu}{\lambda} \hat{f}_2 e^{\lambda t_f}$$

So it seems the addition to the mean relative to the standard LD process is a trivial factor of  $\hat{f}_2$ , and the average number of survivors originating from time  $t$  does not depend on  $t$ , exactly as in the original setting.

Now we calculate the variance of the number of mutations.