Evolutionary Dynamics Homework 6

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Problem 1: One-dimensional Fokker-Planck equation

Consider the one-dimensional Fokker-Planck equation with constant coefficients,

$$\partial_t \psi(p,t) = -m \partial_p \psi(p,t) + \frac{v}{2} \partial_p^2 \psi(p,t), \tag{1}$$

with $p \in \mathbb{R}, v > 0$.

(a) See below.

Show that for vanishing selection, m = 0,

$$\psi(p,t) = \frac{1}{\sqrt{2\pi vt}} \exp\left(-\frac{p^2}{2vt}\right) \tag{2}$$

solves the Fokker-Planck equation. To which initial condition does this solution correspond?

For vanishing selection m=0, the Fokker-Planck equation can be written as simply

$$\partial_t \psi(p,t) = \frac{v}{2} \partial_p^2 \psi(p,t) \tag{3}$$

Let's compute the left-hand side and the right-hand side to show that the solution is correct.

LHS of (3)

$$\begin{split} \partial_t \psi(p,t) &= \partial_t \left(\frac{1}{\sqrt{2\pi v t}} \exp\left(-\frac{p^2}{2v t} \right) \right) \\ &= -\frac{1}{\sqrt{2\pi v t} \cdot 2t} \cdot \exp\left(-\frac{p^2}{2v t} \right) + \frac{1}{\sqrt{2\pi v t}} \cdot \frac{p^2}{2v t^2} \cdot \exp\left(-\frac{p^2}{2v t} \right) \\ &= \frac{p^2 - v t}{2v t^2} \cdot \frac{1}{\sqrt{2\pi v t}} \cdot \exp\left(-\frac{p^2}{2v t} \right) \end{split} \tag{4}$$

RHS of (3)

$$\frac{v}{2}\partial_{p}^{2}\psi(p,t) = \frac{v}{2}\partial_{p}\left(\partial_{p}\left(\frac{1}{\sqrt{2\pi vt}}\exp\left(-\frac{p^{2}}{2vt}\right)\right)\right)$$

$$= \frac{v}{2\sqrt{2\pi vt}}\partial_{p}\left(\partial_{p}\left(\exp\left(-\frac{p^{2}}{2vt}\right)\right)\right)$$

$$= \frac{v}{2\sqrt{2\pi vt}}\partial_{p}\left(\left(-\frac{p}{vt}\right)\cdot\exp\left(-\frac{p^{2}}{2vt}\right)\right)$$

$$= \frac{v}{2\sqrt{2\pi vt}}\cdot\left(-\frac{1}{vt}\exp\left(-\frac{p^{2}}{2vt}\right) + \frac{p^{2}}{v^{2}t^{2}}\cdot\exp\left(-\frac{p^{2}}{2vt}\right)\right)$$

$$= -\frac{1}{\sqrt{2\pi vt}\cdot 2t}\cdot\exp\left(-\frac{p^{2}}{2vt}\right) + \frac{1}{\sqrt{2\pi vt}}\cdot\frac{p^{2}}{2vt^{2}}\cdot\exp\left(-\frac{p^{2}}{2vt}\right)$$

$$= \frac{p^{2}-vt}{2vt^{2}}\cdot\frac{1}{\sqrt{2\pi vt}}\cdot\exp\left(-\frac{p^{2}}{2vt}\right)$$

$$= \frac{p^{2}-vt}{2vt^{2}}\cdot\frac{1}{\sqrt{2\pi vt}}\cdot\exp\left(-\frac{p^{2}}{2vt}\right)$$
(5)

It's clear that (4) equals (5), so (2) is indeed the solution.

To investigate the initial condition, we need to examine the behavior of $\psi(p,t)$ as $t\to 0^+$. As $t\to 0^+$, we have:

- The exponential term $\exp(-p^2/(2vt))$ approaches: – 0, $\forall p \neq 0$ because $-p^2/(2vt) \rightarrow -\infty$
- 1 for p=0. The prefactor $\frac{1}{\sqrt{2\pi vt}}$ approaches ∞ for $t\to 0^+$.

This behavior is consistent with the Dirac delta function $\delta(p)$. Hence, the initial condition is

$$\psi(p,0) = \delta(p),$$

where $\delta(p)$ is the Dirac delta function defined as

$$\delta(p) = \begin{cases} 0, & \text{if } p \neq 0, \\ \infty, & \text{if } p = 0. \end{cases}$$

This means the initial distribution is concentrated at p=0, which represents the fact that the starting population is initially at a single allele frequency.

(b) Construct a solution for constant selection, $m \neq 0$, by substituting z = p - mt for p in (1). What is the mean and variance?

Let z(p,t)=p-mt, then we have $\partial_p z(p,t)=1$, and $\partial_t z(p,t)=-m$. Substitute z(p,t) into $\psi(p,t)$ to obtain $\psi(z,t)$, using the chain rule, we have

$$\frac{\partial \psi(p,t)}{\partial t} = \frac{\partial \psi(z,t)}{\partial z} \frac{\partial z}{\partial t} + \frac{\partial \psi(z,t)}{\partial t} \frac{\partial t}{\partial t}$$

$$= -m \frac{\partial \psi(z,t)}{\partial z} + \frac{\partial \psi(z,t)}{\partial t}$$
(6)

$$\frac{\partial \psi(p,t)}{\partial p} = \frac{\partial \psi(z,t)}{\partial z} \frac{\partial z}{\partial p}
= \frac{\partial \psi(z,t)}{\partial z}$$
(7)

Using (6) and (7), we can rewrite the Fokker-Planck equation:

$$\frac{\partial \psi(p,t)}{\partial t} = -m \frac{\partial \psi(p,t)}{\partial p} + \frac{v}{2} \frac{\partial^2 \psi(p,t)}{\partial p^2}
-m \frac{\partial \psi(z,t)}{\partial z} + \frac{\partial \psi(z,t)}{\partial t} = -m \frac{\partial \psi(z,t)}{\partial z} + \frac{v}{2} \frac{\partial^2 \psi(z,t)}{\partial z^2}
\frac{\partial \psi(z,t)}{\partial t} = \frac{v}{2} \frac{\partial^2 \psi(z,t)}{\partial z^2}$$
(8)

Voila! This equation (8) has exactly the same form as (3), so the solution is of the form of (2) also. We use the same form and substitute z = p - mt into it:

$$\begin{split} \psi(z,t) &= \frac{1}{\sqrt{2\pi v t}} \exp\left(-\frac{z^2}{2vt}\right) \\ \psi(p,t) &= \frac{1}{\sqrt{2\pi v t}} \exp\left(-\frac{(p-mt)^2}{2vt}\right) \sim \mathcal{N}(mt,vt) \end{split}$$

Hence, the mean is mt and the variance is vt.

Problem 2: Diffusion approximation of the Moran process

Derive a diffusion approximation for the Moran process of two species. Assume the first species has a small selective advantages.

(a) See below.

The general definition for the drift coefficient is

$$M(p) = E[X(t) - X(t-1)|X(t-1) = i]/N,$$
(9)

where p = i/N and X(t) denotes the abundance of the first allele. Evaluate this expression for the Moran process with selection. Show that this yields the result for the Wright-Fisher process from the lecture, divided by N.

From lecture 3 we know that for a *neutral* Moran process we have transition probabilities:

- $\begin{array}{l} \bullet \ \ P_{i,i+1} = p(1-p) \\ \bullet \ \ P_{i,i-1} = (1-p)p \\ \bullet \ \ P_{i,i} = p^2 + (1-p)^2 \\ \end{array}$

However, since the first species has a small selective advantage (denoted as s hereafter), we need to modify the transition probabilities. We have:

- $P_{i,i+1} = (p(1+s)/(p(1+s) + (1-p))) \cdot (1-p) = (p(1+s)/(1+sp)) \cdot (1-p)$ $P_{i,i-1} = (1-p)/(1+sp) \cdot p$ $P_{i,i} = 1 P_{i,i+1} P_{i,i-1}$

And the expectation derived from the transition probabilities is:

$$\begin{split} E[X(t)|X(t-1) &= i] = i \cdot P_{i,i} + (i+1) \cdot P_{i,i+1} + (i-1) \cdot P_{i,i-1} \\ &= i \cdot (1 - P_{i,i+1} - P_{i,i-1}) + (i+1) \cdot P_{i,i+1} + (i-1) \cdot P_{i,i-1} \\ &= i + P_{i,i+1} - P_{i,i-1} \\ &= i + \frac{p(1+s)}{1+sp} \cdot (1-p) - \frac{1-p}{1+sp} \cdot p \\ &= i + \frac{sp(1-p)}{1+sp} \end{split} \tag{10}$$

Using (10) and (9), we can calculate the drift coefficient M(p):

$$\begin{split} M(p) &= \frac{E[X(t) - X(t-1)|X(t-1) = i]}{N} \\ &= \frac{E[X(t)|X(t-1) = i] - E[X(t-1)|X(t-1) = i]}{N} \\ &= \frac{E[X(t)|X(t-1) = i] - i}{N} \\ &= \left(i + \frac{sp(1-p)}{1+sp} - i\right)/N \\ &= \frac{sp(1-p)}{N(1+sp)} \end{split} \tag{11}$$

From slide 33, we know that the drift coefficient for the Wright-Fisher process is:

$$M_{WF}(p) = \frac{sp(1-p)}{1+sp},$$

and (11) is exactly the result for the Wright-Fisher process divided by N.

(b) By a similar argument calculate the diffusion coefficient V(p).

From slides, we have

$$\begin{split} V(p) &= E\left[Var[p(t+1)]|p(t)\right] \\ &= \frac{1}{N^2} E\left[Var[X(t+1)|X(t)]\right] \end{split} \tag{12}$$

To derive Var[X(t+1)|X(t)] again, we first derive Var[X(t+1)|X(t)=i]:

$$\begin{split} Var[X(t+1)|X(t) &= i] = Var[X(t)] + Var[X(t+1) - X(t)|X(t) = i] \\ &= 0 + E[(X(t+1) - X(t))^2|X(t) = i] - E[X(t+1) - X(t)|X(t) = i]^2 \\ &= (i-1-i)^2 \cdot P_{i,i-1} + (i+1-i)^2 \cdot P_{i,i+1} + (i-i)^2 \cdot P_{i,i} - E[X(t+1) - X(t)|X(t) = i]^2 \\ &= P_{i,i-1} + P_{i,i+1} - E[X(t+1)|X(t) = i]^2 \\ &= \frac{p(1-p)}{1+sp} + \frac{(1+s)p(1-p)}{1+sp} - \left(\frac{sp(1-p)}{1+sp}\right)^2 \\ &= \frac{p(1-p)(s+2)}{1+sp} - \left(\frac{sp(1-p)}{1+sp}\right)^2 \\ &= p(1-p)\left(1 + \frac{1+s}{(1+sp)^2}\right) \end{split}$$

As (13) is not related to i at all, we can say that the expression of Var[X(t+1)|X(t)] is the same as (13). Plug (13) back to (12), we obtain the diffusion coefficient:

$$V(p) = \frac{1}{N^2} \left(p(1-p) \left(1 + \frac{1+s}{(1+sp)^2} \right) \right)$$
 (14)

(c) Use your results from (a) and(b) in the forward Kolmogorov equation to present a diffusion equation for the Moran model.

Plug in (11) and (14) into the forward Kolmogorov equation:

$$\begin{split} \frac{\partial \psi(p,t)}{\partial t} &= -\frac{\partial}{\partial p} \left[\psi(p,t) M(p) \right] + \frac{1}{2} \frac{\partial^2}{\partial p^2} \left[\psi(p,t) V(p) \right] \\ &= -\frac{\partial}{\partial p} \left[\psi(p,t) \frac{sp(1-p)}{N(1+sp)} \right] + \frac{1}{2} \frac{\partial^2}{\partial p^2} \left[\psi(p,t) \frac{1}{N^2} p(1-p) \left(1 + \frac{1+s}{(1+sp)^2} \right) \right] \end{split}$$

(d) See below.

Now assume that $s \le 1$. Approximate your results from (a) and (b) and use the general expression for the fixation probability $\rho(p_0)$ to show that the fixation probability is given by:

$$\rho(p_0 = 1/N) = \frac{1 - e^{-s}}{1 - e^{-Ns}} \tag{15}$$

The general expression for fixation probability $\rho(p_0)$ is (given in the slides) as:

$$\rho(p_0) - \frac{\int_0^{p_0} \exp\left(-\int_0^p \frac{2M(q)}{V(q)} dq\right) dp}{\int_0^1 \exp\left(-\int_0^p \frac{2M(q)}{V(q)} dq\right) dp}$$
(16)

Now we want to simplify the expression of 2M(q)/V(q). Since we have $s \le 1$, we can write:

$$\frac{2M(q)}{V(q)} = \frac{2sp(1-p)}{N(1+sp)} / \left(\frac{p(1-p)}{N^2} \left(1 + \frac{1+s}{(1+sp)^2}\right)\right)
= \frac{2Ns}{1+sp} / \frac{(1+sp)^2 + (1+s)}{(1+sp)^2}
= \frac{2Ns(1+sp)}{(1+sp)^2 + (1+s)}
\approx \frac{2Ns \cdot 1}{1^2 + 1}$$
 (p is also small)
= Ns

Given (17), we have:

$$\int_0^p \frac{2M(q)}{V(q)} = \int_0^p Nsdq = Nsp \tag{18}$$

Plug (18) back to (16), we have:

$$\begin{split} \rho(p_0 = 1/N) &= \frac{\int_0^{p_0} \exp\left(-Ns\right) dp}{\int_0^1 \exp\left(-Ns\right) dp} \\ &= \frac{1 - \exp(-Nsp_0)}{1 - \exp(-Ns)} \;, \\ &= \frac{1 - e^{-s}}{1 - e^{-Ns}} \end{split}$$

which is exactly the form of (15).

(e) Take the limit to derive a result for the fixation probability of a neutral allele, s=0. Evaluate (15) for N=10 and N=1000 for both positive and negative selection, s=2%, and s=-2%, respectively. Compare your result with ρ_1 of the exact Moran process.

Fixation probability for neutral allele

$$\lim_{s \to 0} \frac{1 - e^{-s}}{1 - e^{-Ns}} = \frac{1 - e^0}{1 - e^0} = 1$$

Approximation v.s. exact Moran process

The author preferred to write code block (although sometimes get the parentheses wrong) to do numeric calcualtions.

```
library(tidyr)
library(dplyr)
library(kableExtra)

moran_fixation_prob <- function(N, s) {
   return((1 - 1 / (1+s)) / (1 - 1 / (1+s)^N))
}

diffusion_fixation_prob <- function(N, s) {
   return((1 - exp(-s)) / (1 - exp(-N * s)))
}</pre>
```

```
rho_exact_N10_sneg <- moran_fixation_prob(10, -0.02)</pre>
rho_exact_N1000_spos <- moran_fixation_prob(1000, 0.02)</pre>
rho_exact_N1000_sneg <- moran_fixation_prob(1000, -0.02)</pre>
rho_approx_N10_spos <- diffusion_fixation_prob(10, 0.02)</pre>
rho_approx_N10_sneg <- diffusion_fixation_prob(10, -0.02)</pre>
rho approx N1000 spos <- diffusion fixation prob(1000, 0.02)
rho_approx_N1000_sneg <- diffusion_fixation_prob(1000, -0.02)</pre>
fixation_probs <- data.frame(</pre>
  Population_Size = c(10, 10, 1000, 1000),
  Selection_Coefficient = c(0.02, -0.02, 0.02, -0.02),
  Exact_Probability = c(rho_exact_N10_spos, rho_exact_N10_sneg,
                        rho_exact_N1000_spos, rho_exact_N1000_sneg),
 Diffusion_Approximation = c(rho_approx_N10_spos, rho_approx_N10_sneg,
                              rho approx N1000 spos, rho approx N1000 sneg)
)
fixation_probs %>%
  kable(format = "latex",
        col.names = c("N", "s", "Exact ()", "Diffusion Approx. ()"),
        digits = 6.
        caption = "Fixation Probabilities: Exact vs Diffusion Approximation") %>%
  kable_styling(latex_options = c("striped", "hold_position"),
                full_width = FALSE) %>%
  row_spec(0, bold = TRUE)
```

Table 1: Fixation Probabilities: Exact vs Diffusion Approximation

N	s	Exact (ρ)	Diffusion Approx. (ρ)
10	0.02	0.109144	0.109237
10	-0.02	0.091156	0.091242
1000	0.02	0.019608	0.019801
1000	-0.02	0.000000	0.000000

Problem 3: Absorption time in the diffusion approximation

In the diffusion approximation of a process with absorbing states 0 and 1 the expected fixation time, conditioned on absorption in state 1, reads:

$$\tau_1(p_0) = 2(S(1) - S(0)) \left(\int_{p_0}^1 \frac{\rho(p)(1 - \rho(p))}{e^{-A(p)}V(p)} dp + \frac{1 - \rho(p_0)}{\rho(p_0)} \int_0^{p_0} \frac{\rho(p)^2}{e^{-A(p)}V(p)} dp \right)$$

where $\rho(p)$ denotes the fixation probability,

rho_exact_N10_spos <- moran_fixation_prob(10, 0.02)</pre>

$$A(p) = \int_0^p \frac{2M(p)}{V(p)} dp, \quad \text{and} \quad S(p) = \exp(-A(p)) dp.$$

(a) Calculate the conditional expected waiting time for fixation, $\tau_1(p_0)$, of an allele of frequency p_0 in the neutral Wright-Fisher process. Approximate the result for small p_0 .

The original formula seems intimidating, so there must be some way to make it simpler. Fortunately, we have the following information:

- S(0) = 0, by definition.
- For neutral Wright-Fisher process, M(p)=0, hence A(p)=0, $\forall p$ (From slides, M(p)=p(1-p)s/(1+ps), and $s\to 0$ for neutral Wright-Fisher process).
- S(1) = 1 follows the previous point.
- V(p) = p(1-p)/N as from slide 19.
- $\rho(p) = p$ for neutral Wright-Fisher process.

With the above information, we can simplify the formula for $\tau_1(p_0)$:

$$\begin{split} \tau_1(p_0) &= 2 \left(\int_{p_0}^1 \frac{p(1-p)}{e^0 p(1-p)/N} dp + \frac{1-p_0}{p_0} \int_0^{p_0} \frac{p^2}{e^0 p(1-p)/N} dp \right) \\ &= 2 \left(\int_{p_0}^1 N dp + \frac{1-p_0}{p_0} \int_0^{p_0} \frac{Np}{1-p} dp \right) \\ &= 2 \left(N(1-p_0) + \frac{N(1-p_0)}{p_0} \left(-p_0 - \log(1-p_0) \right) \right) \\ &= 2 \left(N(1-p_0) - N(1-p_0) - \frac{N(1-p_0) \log(1-p_0)}{p_0} \right) \\ &= -\frac{2N(1-p_0) \log(1-p_0)}{p_0} \end{split}$$

For small p_0 , we can use the Taylor expansion of $\log(1-p_0)$:

$$\begin{split} \tau_1(p_0) &= -\frac{2N(1-p_0)(-p_0-p_o^2/2 - O(p_0^3))}{p_0} \\ &= 2N(1-p_0)(1+p_0/2 + O(p_0^2)) \\ &= 2N(1+p_0/2 + O(p_0^2)) - 2N(p_0 + O(p_0^2)) \\ &\approx 2N(1-p_0/2) \\ &\approx 2N \end{split} \tag{for small } p_0 \ll 1) \end{split}$$

Therefore, for small p_0 , the conditional expected waiting time for fixation is approximately 2N generations.

(b) Compute τ_0 , the conditional expected waiting time until extinction (absorption in state 0) in the neutral Wright-Fisher process. Also derive the unconditioned expected waiting time $\bar{\tau}$ until either fixation or extinction.

If the current allele of interest (say A) goes to extinction, it means the other allele a goes to fixation. If A has initial abundance p_0 , then a has initial abundance $1-p_0$.

$$\begin{split} \tau_0(p_0) &= \tau_1(1-p_0) \\ &= -\frac{2N(1-(1-p_0))\log(1-(1-p_0))}{1-p_0} \\ &= -\frac{2Np_0\log(p_0)}{1-p_0} \end{split}$$

For the unconditioned expected waiting time $\bar{\tau}(p_0)$

- The probability of fixation is p_0
- The probability of extinction is $1-p_0$

Therefore:

$$\begin{split} \bar{\tau}(p_0) &= p_0 \tau_1(p_0) + (1-p_0) \tau_0(p_0) \\ &= p_0 \left(-\frac{2N(1-p_0) \log(1-p_0)}{p_0} \right) + (1-p_0) \left(-2N \left(\frac{p_0 \log(p_0)}{1-p_0} \right) \right) \\ &= -2N(1-p_0) \log(1-p_0) - 2Np_0 \log(p_0) \\ &= -2Np_0 (1-p_0) [\log(1-p_0)/p_0 + \log(p_0)/(1-p_0)] \end{split}$$

This is the general formula for the unconditional expected waiting time until absorption in either state 0 or 1.