

In subdivided population, 2 types of indiv: $\begin{cases} \text{local} \\ \text{in, in deme, + from local} \\ \text{out, out of deme} \end{cases}$ (2)

$$\sum_k \frac{\partial W_i}{\partial y_k} p_k = \frac{\partial W_i}{\partial y_0} p_{i0} + (n-1) \frac{\partial W_i}{\partial y_{in}} p_{in} + (N-n) \frac{\partial W_i}{\partial y_{out}} p_{out}$$

also $\sum_i W_i = \sum_i (B_i(1-\mu) + D_i) = \sum_i (B_i - D_i) = \sum_i B_i \mu$
 does not depend on the state of the population

so $(N-n) \frac{\partial W_i}{\partial y_{out}} = - \left((n-1) \frac{\partial W_i}{\partial y_{in}} + \frac{\partial W_i}{\partial y_0} \right)$

and $\sum_k \frac{\partial W_i}{\partial y_k} p_k = \frac{\partial W_i}{\partial y_0} (p_{i0} - p_{out}) + (n-1) \frac{\partial W_i}{\partial y_{in}} (p_{in} - p_{out})$

and so (1) $\omega B^* \mu \frac{\partial E[\bar{x}]}{\partial \omega} = \omega \frac{1}{N} \sum_i \left(\frac{\partial W_i}{\partial y_0} + (n-1) \frac{\partial W_i}{\partial y_{in}} \frac{(p_{in} - p_{out})}{p_{i0} - p_{out}} \right) (p_{i0} - p_{out})$

+ same behavior for all sites in expectation
 Note $p_{ij} = (1/2)v + (1/2)v^2$ so $R = \frac{p_{in} - p_{out}}{1 - p_{out}}$ pairwise relatedness

Then we decompose again the terms using total derivatives and foundites as intermediate variables

$$\frac{\partial W_i}{\partial y_k} = \sum_{l=1}^N \frac{\partial W_i}{\partial x_l} \frac{\partial x_l}{\partial y_k} \quad \text{if } l=k: -c \quad \text{if } l \neq k: \text{same deriv. } \frac{b}{n-1}$$

Conditional expected frequency
 $E[\bar{x}(t+1) | X(t)] = \frac{1}{N} \sum_{i=1}^N (B_i(1-\mu) + (1-D_i)) x_i + \frac{1}{N} \sum_{i=1}^N B_i \mu v$ (1)
 - mutants

Take expectation and let $t \rightarrow \infty$ (stationary distribution)
 $E[\bar{x}] = \frac{1}{N} \sum_i (B_i(1-\mu) - D_i) E[x_i] + \frac{1}{N} \sum_i B_i \mu v + E[\bar{x}]$

Note: with an life cycles, $\sum_i B_i$ does not depend on population composition
 In the absence of selection, $E[x_i] = v$ (mutation bias) and $D_i = 0$.

Weak selection approximation, first order
 $0 = 0 + \omega \left[\frac{1}{N} E_0 \left[\sum_i \left(\frac{\partial B_i}{\partial y_0} (1-\mu) - \frac{\partial D_i}{\partial y_0} \right) x_i \right] + \frac{1}{N} \frac{\partial E}{\partial \omega} \left[\sum_i (B_i(1-\mu) - D_i) x_i \right] + \frac{1}{N} E_0 \left[\sum_i \frac{\partial B_i}{\partial y_0} \mu v \right] + \frac{1}{N} \frac{\partial E}{\partial \omega} \left[\sum_i B_i \mu v \right] \right] + O(\omega^2)$

$0 = \omega \left[\frac{1}{N} E_0 \left[\sum_i \left(\frac{\partial B_i}{\partial y_0} (1-\mu) - \frac{\partial D_i}{\partial y_0} \right) x_i \right] + \frac{1}{N} \sum_i B_i^* \mu \frac{\partial E(x_i)}{\partial \omega} \right] + O(\omega^2)$

$\rightarrow \omega B^* \mu \frac{\partial E[\bar{x}]}{\partial \omega} = \omega \left[E_0 \left[\sum_i \left(\frac{\partial B_i}{\partial y_0} (1-\mu) - \frac{\partial D_i}{\partial y_0} \right) x_i \right] + O(\omega^2) \right]$ (1)
 $\rightarrow E[\bar{x}] = v + \omega \frac{\partial E[\bar{x}]}{\partial \omega} + O(\omega^2)$

Use total derivatives - Charles' version, intermediate variables are phenotypes: $\phi_i = \bar{x}_i$ (or ωx_i with any notation)
 Let's write $W_i = B_i(1-\mu) - D_i$ (fitness without mutation)

Then $\frac{\partial W_i}{\partial y_k} = \sum_{l=1}^N \frac{\partial W_i}{\partial \phi_l} \frac{\partial \phi_l}{\partial y_k} = \sum_l \frac{\partial W_i}{\partial \phi_l} \times x_{lk}$

(1) becomes $\omega B^* \mu \frac{\partial E[\bar{x}]}{\partial \omega} = \omega \frac{1}{N} \sum_{i,k} \frac{\partial W_i}{\partial \phi_k} \frac{\partial E[x_i x_k]}{\partial \omega} + O(\omega^2)$

expected state of pair of site written x_{ik}

(DB) Moran DB & WF differ by $1/2$

$B_i = \frac{1}{N} \sum_j f_i \frac{df_j}{dy_k} = \frac{1}{N} \sum_j f_i \frac{df_j}{dy_k} \frac{dy_k}{dx_j} = \frac{1}{N} \sum_j f_i \frac{df_j}{dx_j} \frac{dx_j}{dy_k} = \frac{1}{N} \sum_j f_i \frac{df_j}{dx_j} x_{jk}$

$\frac{\partial B_i}{\partial y_k} = \frac{1}{N} \sum_j f_i \frac{df_j}{dx_j} x_{jk}$

Moran Birth-Death $B_i = \frac{f_i}{\sum_l f_l} = \frac{f_i}{N}$

$\frac{\partial B_i}{\partial y_k} = \frac{1}{N^2} \sum_j f_j \frac{df_j}{dx_j} x_{jk}$

(DB) $\omega B^* \mu \frac{\partial E[\bar{x}]}{\partial \omega} = \omega \frac{1}{N} \sum_i \left(\frac{\partial B_i}{\partial y_0} + (n-1) \frac{\partial B_i}{\partial y_{in}} \right) (p_{i0} - p_{out})$

$\frac{\partial B_i}{\partial y_0} = \frac{1}{N^2} \sum_j f_j \frac{df_j}{dx_j} x_{j0}$

$\frac{\partial B_i}{\partial y_{in}} = \frac{1}{N^2} \sum_j f_j \frac{df_j}{dx_j} x_{jin}$

$\frac{\partial B_i}{\partial y_k} = \frac{1}{N^2} \sum_j f_j \frac{df_j}{dx_j} x_{jk} = \frac{1}{N^2} \sum_j f_j \frac{df_j}{dx_j} \frac{dx_j}{dy_k} = \frac{1}{N^2} \sum_j f_j \frac{df_j}{dx_j} x_{jk}$

$\frac{\partial B_i}{\partial y_k} = \frac{1}{N^2} \sum_j f_j \frac{df_j}{dx_j} x_{jk} = \frac{1}{N^2} \sum_j f_j \frac{df_j}{dx_j} \frac{dx_j}{dy_k} = \frac{1}{N^2} \sum_j f_j \frac{df_j}{dx_j} x_{jk}$

would be $\frac{m^2}{(N-n)^2}$ if self-replace