

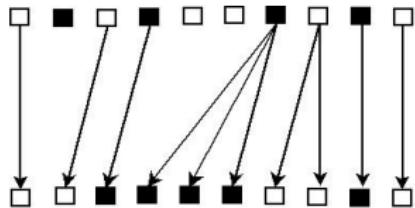
# Quasi equilibrium methods in population genetics

Ola Hössjer  
Dept. of Mathematics  
Stockholm University

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# Wright Fisher Model

- ▶ Population of  $2N$  gene (marker) copies
- ▶ Allele 1 and 2
- ▶ Nonoverlapping generations
- ▶ Constant population size
- ▶  $P_t$  = frequency of Allele 1, Generation  $t$ ,



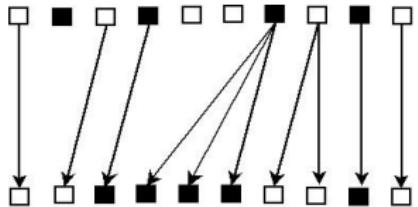
$$2N = 10,$$

$$P_t = 0.4,$$

$$P_{t+1} = 0.5.$$

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$$\begin{aligned} 2N &= 10, \\ P_t &= 0.4, \\ P_{t+1} &= 0.5. \end{aligned}$$

Alleles drawn randomly from parental generation:

$$P_{t+1}|P_t \sim \text{Bin}(2N, P_t)/(2N)$$

## Variance effective population size $N_{eV}$

Write

$$P_{t+1} = P_t + \varepsilon_{t+1}$$

where  $\varepsilon_{t+1}$  is genetic drift, with  $E(\varepsilon_{t+1}|P_t) = 0$  and

$$\text{Var}(\varepsilon_{t+1}|P_t) = \frac{P_t(1 - P_t)}{2N_{eV}} \xrightarrow{\text{WF}} N_{eV} = N.$$

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Rule of thumb:

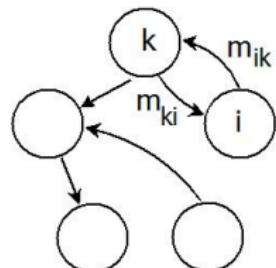
More inbreeding  $\implies$  More genetic drift  $\implies$  Smaller  $N_{eV}$

In general  $N_{eV} \neq N$  due to

- ▶ Diploid population
- ▶ Spatial structure
- ▶ Varying reproductivity
- ▶ Time varying population size
- ▶ Overlapping generations

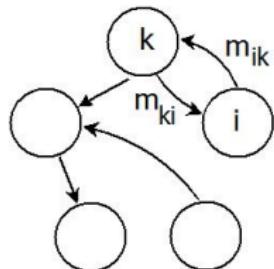
# Structured population

- ▶  $s$  subpopulations
- ▶  $m_{ki}$  migration rate from Subpopulation  $k$  to  $i$
- ▶  $N$  total population size ( $= 2N$  gene copies)



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$$\begin{aligned}\mathbf{M} &= (m_{ki})_{k,i=1}^s = \text{migration matrix} \\ Na_i &= \text{constant subpopulation sizes}\end{aligned}$$

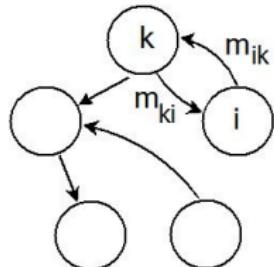
where  $\sum_{i=1}^s a_i = 1$  and

$$\mathbf{a} = (a_1, \dots, a_s) = \mathbf{aM},$$

is a left eigenvector of  $\mathbf{M}$  with eigenvalue 1.

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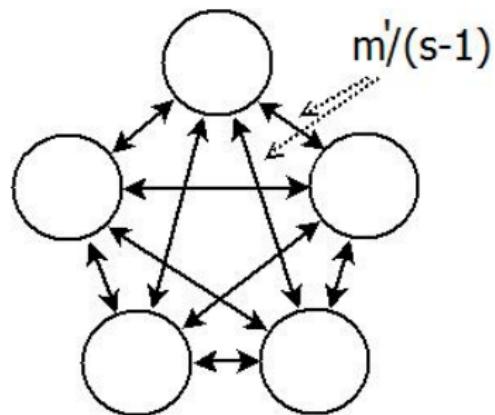
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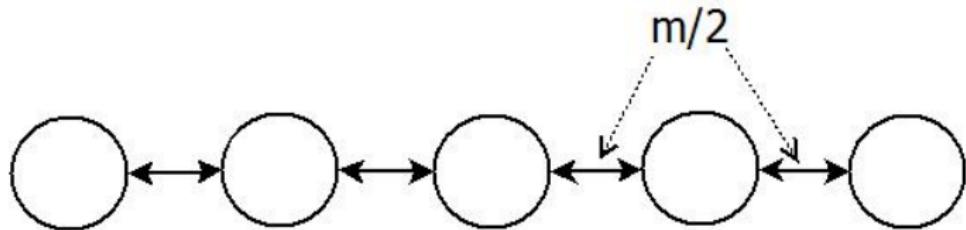
$$\text{Overall migration rate } m' = \sum_{i=1}^s a_i(m_{i\cdot} - m_{ii}) = 1 - \sum_{i=1}^s a_i m_{ii}.$$

## Island model



$$m_{ki} = \begin{cases} 1 - m', & i = k, \\ m'/(s-1), & i \neq k. \end{cases}$$

## Linear stepping stone model

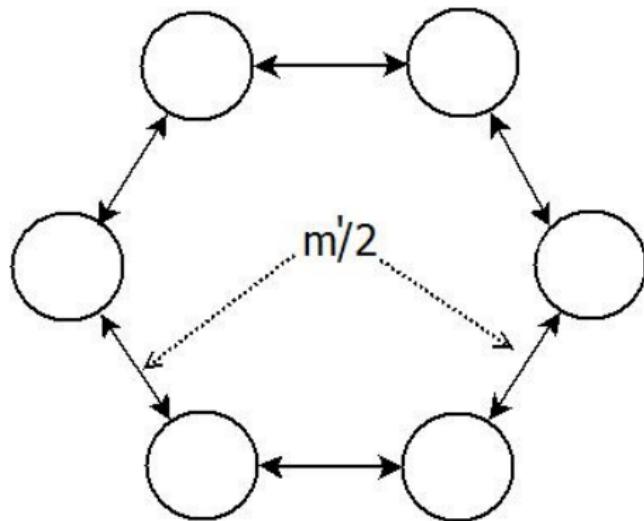


$$m_{ki} = \begin{cases} 1 - m|\mathcal{N}_k|/2, & i = k, \\ m/2, & |i - k| = 1, \\ 0, & \text{otherwise,} \end{cases}$$

where  $\mathcal{N}_k$  is the neighbourhood of  $k$ . Hence

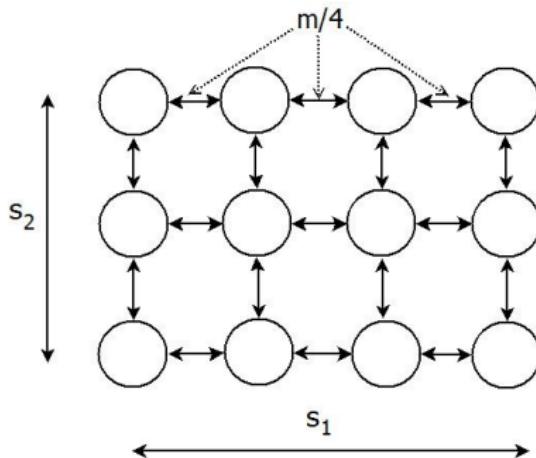
$$m' = \frac{2}{s} \cdot \frac{m}{2} + \frac{s-2}{s} \cdot m = \frac{m(s-1)}{s}.$$

## Circular stepping stone



$$m_{ki} = \begin{cases} 1 - m', & i = k, \\ m'/2, & (i - k \bmod s) = 1 \text{ or } s - 1, \\ 0, & \text{otherwise.} \end{cases}$$

## Rectangular stepping stone ( $s = s_1 s_2$ )

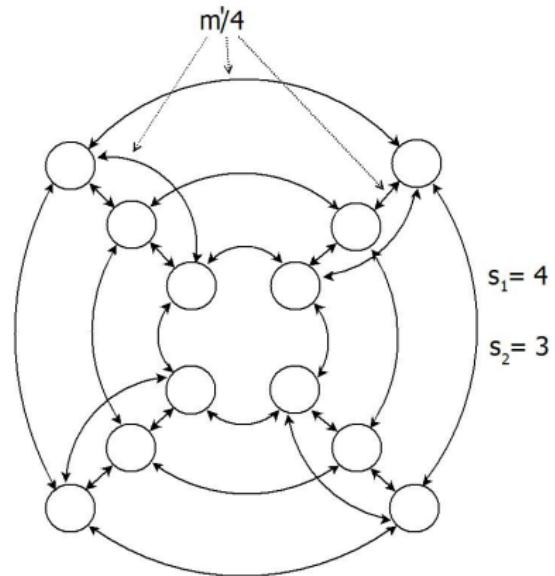


$$m_{(k_1, k_2), (i_1, i_2)} = \begin{cases} 1 - |\mathcal{N}_{(k_1, k_2)}| m/4, & (i_1, i_2) = (k_1, k_2), \\ m/4, & |i_1 - k_1| + |i_2 - k_2| = 1, \\ 0, & \text{otherwise.} \end{cases}$$

where  $\mathcal{N}_{(k_1, k_2)}$  is the neighbourhood of  $(k_1, k_2)$ . This yields

$$m' = m(1 - 0.5(s_1^{-1} + s_2^{-1})).$$

# Torus stepping stone



$$m_{(k_1, k_2), (i_1, i_2)} = \begin{cases} 1 - m', & (i_1, i_2) = (k_1, k_2), \\ m'/4, & (i_1 - k_1 \bmod s_1) = 1 \text{ or } s_1 - 1, i_2 = k_2, \\ m'/4, & (i_2 - k_2 \bmod s_2) = 1 \text{ or } s_2 - 1, i_1 = k_1, \\ 0, & \text{otherwise.} \end{cases}$$

## Backward migration matrix $B$

Let

$$b_{ik} = P(\text{parent of Subpop } i \text{ gene from Subpop } k) \approx \frac{a_k m_{ki}}{a_i}.$$

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Since

$$\sum_{k=1}^s b_{ik} = \frac{1}{a_i} \sum_{k=1}^s a_k m_{ki} = \frac{a_i}{a_i} = 1,$$

the backward matrix

$$\mathbf{B} = (b_{ik})_{i,k=1}^s$$

is transition matrix of Markov chain with equilibrium distr

$$\gamma = (\gamma_1, \dots, \gamma_s).$$

## Example of $\mathbf{M}$ and $\mathbf{B}$

$$\mathbf{M} = \begin{pmatrix} 0.8 & 1.2 \\ 0.1 & 0.4 \end{pmatrix}$$

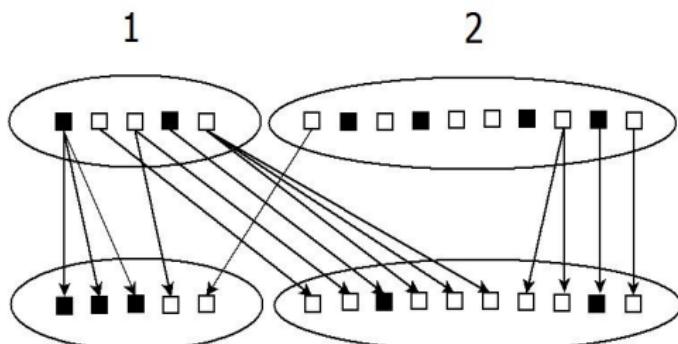
↓

$$\mathbf{a} = (1/3, 2/3)$$

$$\mathbf{B} = \begin{pmatrix} 0.8 & 0.2 \\ 0.6 & 0.4 \end{pmatrix}$$

↓

$$\gamma = (3/4, 1/4).$$



It can be shown that

$$m_{k\cdot} = \sum_{i=1}^s m_{ki} = 1, \quad k = 1, \dots, s \implies \gamma = \mathbf{a}.$$

## Allele frequencies, fixation index

Let

$P_{ti}$  = frequency of Allele 1 in Subpop  $i$

and

$$P_t = \sum_{i=1}^s a_i P_{ti} = \text{frequency of Allele 1 in whole pop}$$

in Generation  $t$ .

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in Generation  $t$ . Then

$$F_{ST} = \text{fixation index} = \frac{\sum_{i=1}^s a_i (P_{ti} - P_t)^2}{P_t(1 - P_t)}$$

quantifies **spatial diversity of subpopulations**.

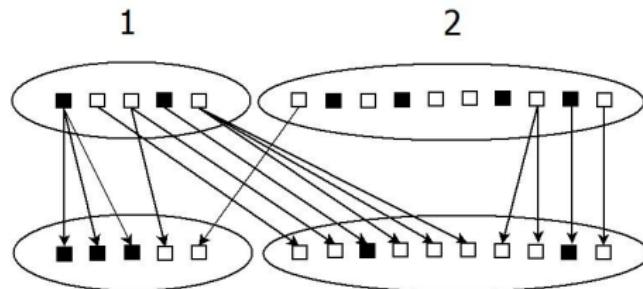
More migration  $\implies$  Smaller  $F_{ST}$

More genetic drift  $\implies$  Larger  $F_{ST}$

## Allele frequency evolution

Let

$$P_{ti} = \text{freq of Allele 1, Subpopulation } i, \text{ Generation } t$$



$$\begin{aligned} P_{t1} &= 0.4 & P_{t2} &= 0.4, \\ P_{t+1,1} &= 0.6 & P_{t+1,2} &= 0.2. \end{aligned}$$

We get the recursion

$$\begin{aligned} P_{t+1,i} &\approx P(\text{Parent of random gene of Subpop } i, \text{ Gen } t+1, \text{ is Allele 1}) \\ &\approx \sum_{k=1}^s b_{ik} P_{tk}. \end{aligned}$$

## Allele frequency evolution, vector form

Putting

$$\mathbf{P}_t = (P_{t1}, \dots, P_{ts})^T,$$

we get the recursion

$$\mathbf{P}_{t+1} = \mathbf{B}\mathbf{P}_t + \boldsymbol{\varepsilon}_{t+1}, \quad (1)$$

with vector valued genetic drift term satisfying

$$E(\boldsymbol{\varepsilon}_{t+1} | \mathbf{P}_t) = \mathbf{0} \text{ and } \text{Var}(\boldsymbol{\varepsilon}_{t+1} | \mathbf{P}_t) = \boldsymbol{\Sigma}(\mathbf{P}_t).$$

- ▶ (1) is vector valued heteroscedastic AR process
- ▶ Nonstationarity since  $\mathbf{B}$  has largest eigenvalue 1
- ▶  $\boldsymbol{\Sigma}(\cdot)$  depends on reproduction scheme

# Reproduction scheme 1: Fertilization precedes migration

**Gametes** from  $N_{ek} \leq Na_k$  breeders

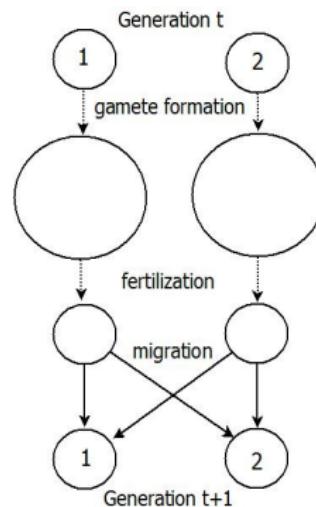
$$\tilde{P}_{tk} | P_{tk} \sim \text{Hyp}(2Na_k, 2N_{ek}, P_{tk}) / (2N_{ek}),$$

followed by **fertilization**

$$P_{tki}^* | \tilde{P}_{tk} \sim \text{Bin}(2Na_k m_{ki}, \tilde{P}_{tk}) / (2Na_k m_{ki})$$

and **migration**

$$P_{t+1,i} = \sum_{k=1}^s b_{ik} P_{tki}^*.$$



## Reproduction scheme 2: Migration precedes fertilization

**Gametes** from  $N_{ek} \leq Na_k$  breeders

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followed by **migration**

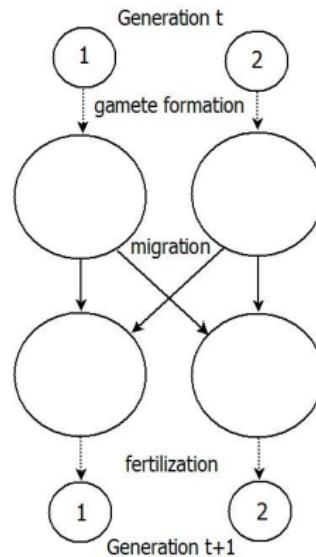
$$\check{P}_{ti} = \sum_{k=1}^s B_{ik} \tilde{P}_{tk},$$

where

$$(B_{i1}, \dots, B_{is}) \sim \text{Dir}(\alpha(b_{i1}, \dots, b_{is})),$$

and **fertilization**

$$P_{t+1,i} | \check{P}_{ti} \sim \text{Bin}(2Na_i, \check{P}_{ti}) / (2Na_i).$$



# Cointegration idea

Decompose allele frequency vector as

$$\begin{aligned}\mathbf{P}_t &= P_t^\gamma (1, \dots, 1)^T + \mathbf{P}_t^0 \\ &= \text{overall frequency} + \text{spatial frequency fluctuations}\end{aligned}$$

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This gives recursion

$$\begin{cases} P_{t+1}^\gamma &= P_t^\gamma + \varepsilon_{t+1}^\gamma, & (\text{genetic drift part}) \\ \mathbf{P}_{t+1}^0 &= \mathbf{B}^0 \mathbf{P}_t^0 + \varepsilon_{t+1}^0, & (\text{spatial fluctuation recursion part}), \end{cases}$$

where

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and by Perron-Frobenius' Theorem and Jordan decomposition

$$\mathbf{B} = \mathbf{V} \begin{pmatrix} 1 & \dots & & \\ 0 & d_2 & \dots & \\ \vdots & \ddots & \ddots & \dots \\ 0 & \dots & 0 & d_s \end{pmatrix} \mathbf{V}^{-1} \text{ and } \mathbf{B}^0 = \mathbf{V} \begin{pmatrix} 0 & \dots & & \\ 0 & d_2 & \dots & \\ \vdots & \ddots & \ddots & \dots \\ 0 & \dots & 0 & d_s \end{pmatrix} \mathbf{V}^{-1}$$

## Quasi equilibrium

Dividing  $\mathbf{P}_t$  by  $\sqrt{P_t^\gamma(1 - P_t^\gamma)}$  makes:

- ▶ Spatial fluctuation part quasi stationary.
- ▶ Genetic drift part still non-stationary.

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The **drift covariance matrix**

$$\boldsymbol{\Sigma} = \frac{\text{Cov}(\mathbf{P}_{t+1}|P_t^\gamma)}{P_t^\gamma(1 - P_t^\gamma)} = \frac{E(\boldsymbol{\Sigma}(\mathbf{P}_t)|P_t^\gamma)}{P_t^\gamma(1 - P_t^\gamma)}$$

and **spatial fluctuation covariance matrix**

$$\boldsymbol{\Lambda} = \frac{\text{Cov}(\mathbf{P}_t^0|P_t^\gamma)}{P_t^\gamma(1 - P_t^\gamma)}$$

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$$\Lambda = \frac{\text{Cov}(\mathbf{P}_t^0|P_t^\gamma)}{P_t^\gamma(1 - P_t^\gamma)}$$

can be used to find

$$N_{eV} \approx \frac{1 - (\mathbf{a} - \gamma)\Lambda(\mathbf{a} - \gamma)^T}{2(\mathbf{a}(\mathbf{B} - \mathbf{I})\Lambda(\mathbf{B} - \mathbf{I})^T\mathbf{a}^T + \mathbf{a}\Sigma\mathbf{a}^T)} \quad \gamma = \frac{1}{2\mathbf{a}\Sigma\mathbf{a}^T}$$

and

$$F_{ST} \approx \frac{\sum_{i=1}^s a_i ((\mathbf{I} - \mathbf{1}\mathbf{a})\Lambda(\mathbf{I} - \mathbf{1}\mathbf{a})^T)_{ii}}{1 - (\mathbf{a} - \gamma)\Lambda(\mathbf{a} - \gamma)^T} \quad \gamma = \sum_{i=1}^s a_i \Lambda_{ii}.$$

## Algorithm

Linear system of equations

$$\begin{aligned}\text{vech}(\boldsymbol{\Sigma}) &= \mathbf{f} - \mathbf{F}\text{vech}(\boldsymbol{\Lambda}), \\ \text{vech}(\boldsymbol{\Lambda}) &= \mathbf{G}\text{vech}(\boldsymbol{\Sigma}),\end{aligned}$$

in the  $s(s + 1)$  unknown parameters

$$\begin{cases} \text{vech}(\boldsymbol{\Sigma}) = \{\Sigma_{ik}; i \geq k\} \\ \text{vech}(\boldsymbol{\Lambda}) = \{\Lambda_{ik}; i \geq k\} \end{cases}$$

with  $\mathbf{F}$  and  $\mathbf{G}$  square matrices of order  $s(s + 1)/2$  and  $\mathbf{f}$  a column vector of length  $s(s + 1)/2$ .

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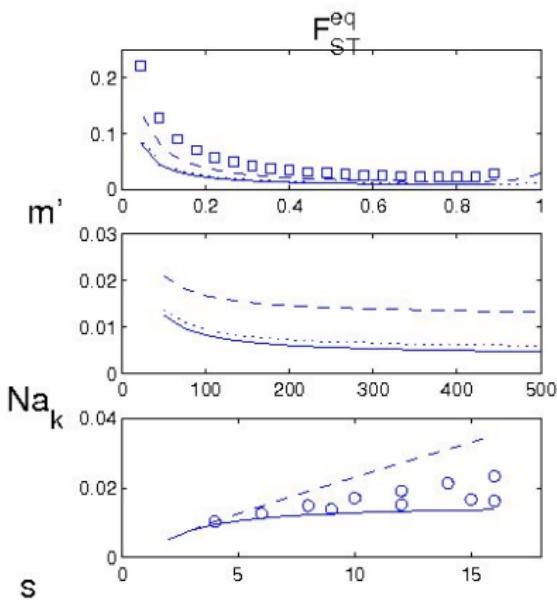
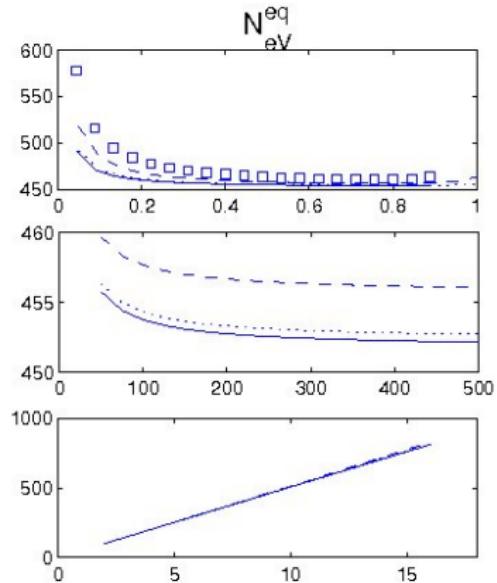
Input parameters are:

- ▶ Migration matrix  $\mathbf{M} \rightarrow \mathbf{B} \rightarrow \mathbf{G}$
- ▶ Reproduction scenario  $\rightarrow \boldsymbol{\Sigma}(\cdot) \rightarrow \mathbf{f}, \mathbf{F}$

# Fertilization before migration

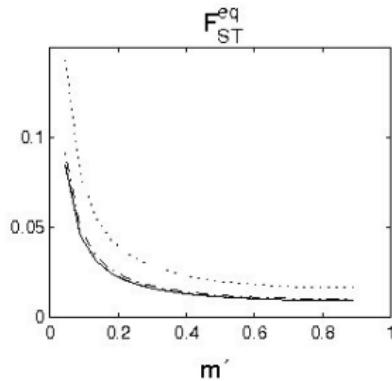
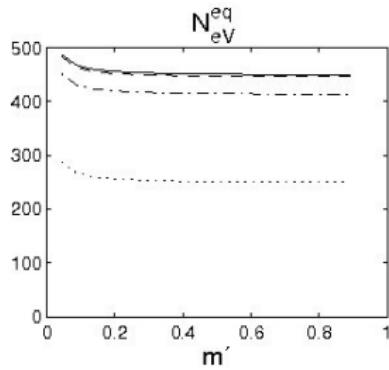
Migration model	Symbol
Island	Solid
Torus	Dotted and circles
Circular	Dashed
Linear stepping stone	Squares

Fixed parameters
$N = 450$
$s = 9$
$m' = 0.4$
$a_k = 1/9$
$N_{ek} = N/9 = 50$

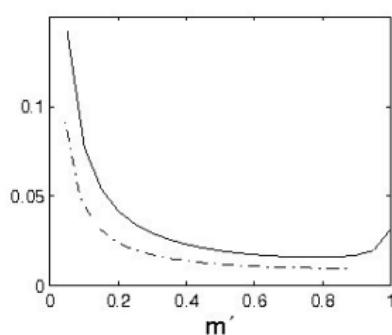
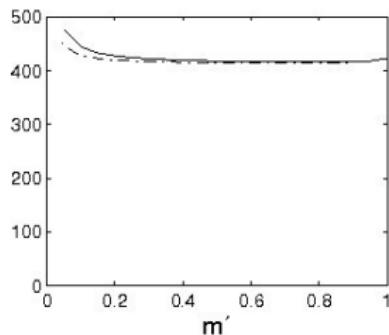


# Migration before fertilization

$$s = 9, N = 450, a_k = 1/9, N_{ek} = Na_k$$

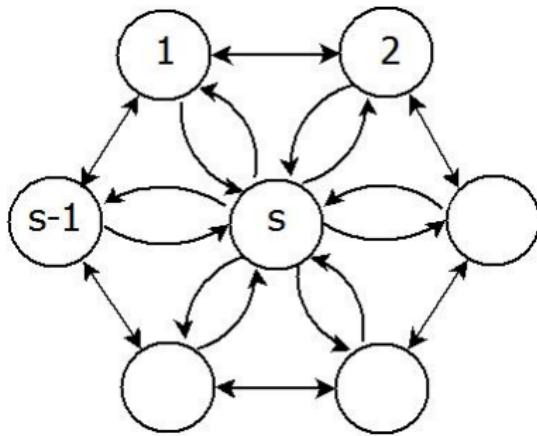


Model	$\alpha$	Symbol
Island	$\infty$	Solid
	100	Dashed
	10	Dash-dotted
	1	Dotted



Model	$\alpha$	Symbol
Island	10	Dash-dotted
Circular stepstone	10	Solid

## Demographic reservoir (source)

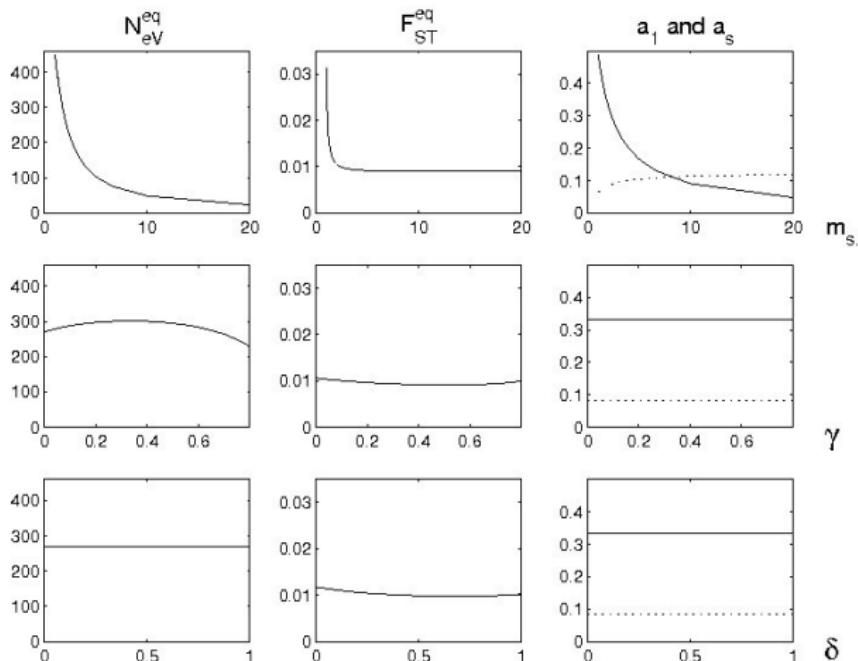


$$m_{1\cdot} = \dots = m_{s-1,\cdot} < 1 < m_{s\cdot}$$

and

$$m_{ki} = \begin{cases} m_s \cdot \beta, & k = s, i < s, \\ m_s \cdot (1 - (s-1)\beta), & k = i = s, \\ m_1 \cdot \gamma, & k < s, i = s, \\ m_1 \cdot \delta / 2, & k < s, i - k = \pm 1 \bmod s - 1, \\ m_1 \cdot (1 - \gamma - \delta), & k = i < s, \\ 0, & \text{otherwise.} \end{cases}$$

# Demographic reservoir, fertilization precedes migration



Top :

$$m_{1\cdot} = \dots = m_{s-1\cdot} = 1/m_{s\cdot}$$

Right :

$a_1$  : dotted,  
 $a_s$  : solid

---

## Fixed parameters

$$\begin{aligned}s &= 9 \\N &= 450 \\N_{ek} &= Na_k \\&\delta = 0.2 \\&\gamma = 0 \\m_{1\cdot} &= 0.5 \\m_{s\cdot} &= 2\end{aligned}$$

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# Extensions

- ▶ Multipel markers (Nei's  $G_{ST}$  instead of  $F_{ST}$ )
- ▶ Spatially invariant migration
  - ▶ Decompose  $\mathbf{B}$  by Fourier analysis (instead of Jordan)
  - ▶ Much faster algorithm
- ▶ Spatial correlations

$$\text{Corr}(P_{ti}, P_{tk}) = \frac{\Lambda_{ik}}{\sqrt{\Lambda_{ii}\Lambda_{kk}}}$$

at quasi equilibrium.

- ▶ Varying (sub)population sizes
- ▶ Diploid populations
- ▶ Overlapping generations

## References

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