Lecture II: The evolution of dispersal

January 2017

Terminology

Dispersal "Any movement of individuals or propagules with potential consequences for gene flow across space" [Ronce, 2007]



Terminology

Dispersal "Any movement of individuals or propagules with potential consequences for gene flow across space" [Ronce, 2007]



Migration "Mass directional movements of large numbers of a species from one location to another."

[Begon et al., 1996]

But in population genetics, often used as a synonym of dispersal.

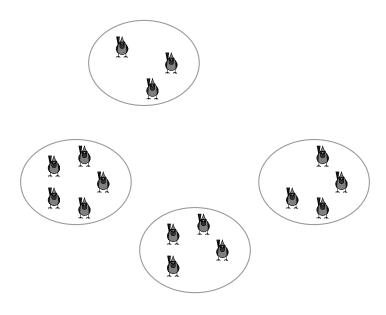


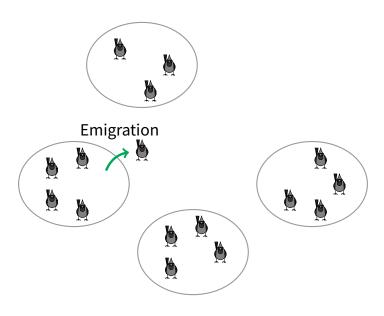
Avoid kin competition

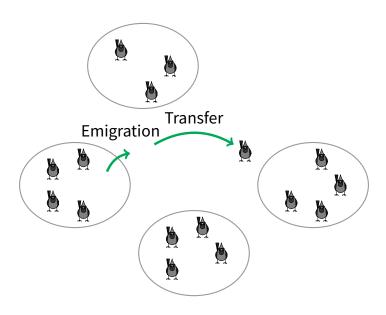
- Avoid kin competition
- Avoid inbreeding

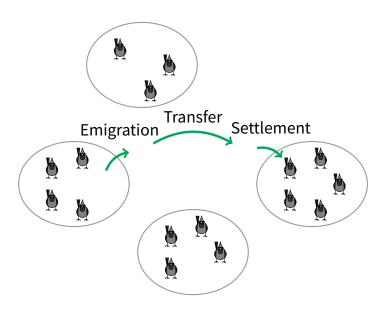
- Avoid kin competition
- ► Avoid inbreeding
- ► Explore new territories

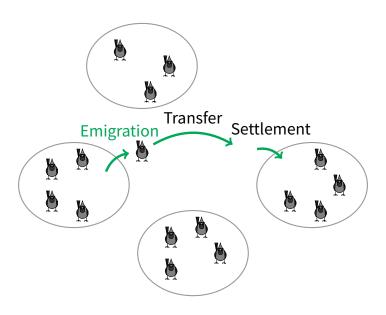
- Avoid kin competition
- Avoid inbreeding
- ► Explore new territories
- Find better conditions.











Outline

Introduction

Dispersal and kin competition Hamilton & May 1977 Island model

In spatially heterogeneous environments

Model



 $N \to \infty$

Model



N saturated sites, $N \to \infty$

Offspring production

Model

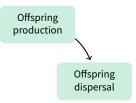


N saturated sites,

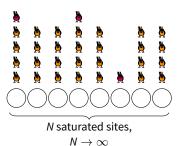
$$N \to \infty$$

Emigration probabilities: x = 0, y > 0

Cost of emigration c = 1 - p.

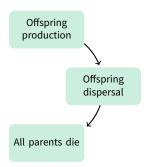


Model

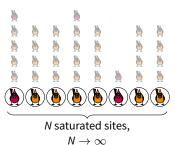


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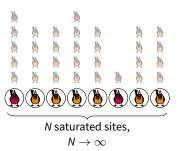
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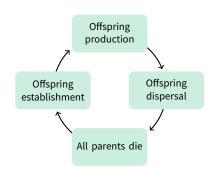
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Model



Emigration probabilities: x = 0, y > 0Cost of emigration c = 1 - p.



Invasion fitness

$$w(y,x) = \frac{1-y}{1-y+(1-c)x} + \frac{(1-c)y}{1-x+(1-c)x}$$

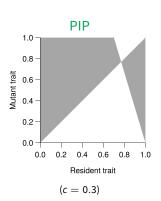
F. Débarre

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Selection gradient

$$D(x) = \left. \frac{\partial w(y,x)}{\partial y} \right|_{y=x} = \frac{(1-c)(1-x(1+c))}{(1-cx)^2}$$



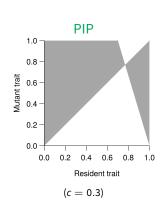
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▶ Selection gradient

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► Singular strategy

$$x^* = \frac{1}{1+c}$$



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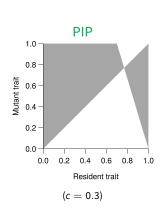
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Singular strategy

$$x^* = \frac{1}{1+c}$$

Convergence stability

$$\frac{dD(x)}{dx} = -\frac{(1-c)(1-c+(c+1)cx)}{(1-cx)^3} \le 0$$



7

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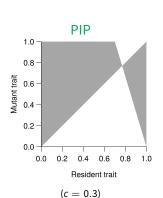
$$\frac{dD(x)}{dx} = -\frac{(1-c)(1-c+(c+1)cx)}{(1-cx)^3} \le 0$$

Uninvadability

$$\left. \frac{\partial^2 w(y,x)}{\partial y^2} \right|_{y=x=x^*} = -2(1-c)(c+1)^2 \le 0$$

F. Débarre

São Paulo, Jan 2017



7

We acknowledge that this simple model probably has few close parallels in the real world. Nevertheless it may usefully force a re-examination of some widely held ideas about migration.

[Hamilton and May, 1977]

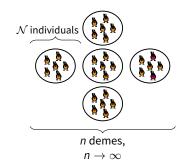
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[Hamilton and May, 1977]

Kin competition Competition between related individuals.

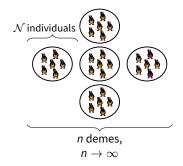






- **z**_r Emigration probability of residents
- **Z**_m Emigration probability of mutants
- c Cost of dispersal
- μ Mutation probability ($\mu
 ightarrow 0$).

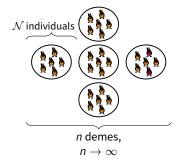




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 $q_0(z_m, z_r)$: Average frequency of mutants in demes that contain mutants.





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 $q_0(z_m, z_r)$: Average frequency of mutants in demes that contain mutants.

Invasion fitness

$$w(z_m, z_r) = \frac{1 - z_m}{1 - (q_0 z_m + (1 - q_0) z_r) + (1 - c) z_r} + \frac{(1 - c) z_m}{1 - z_r + (1 - c) z_r}$$

[Gandon and Rousset, 1999]



$$w(z_m, z_r) = \frac{1 - z_m}{1 - (q_0 z_m + (1 - q_0) z_r) + (1 - c) z_r} + \frac{(1 - c) z_m}{1 - z_r + (1 - c) z_r}$$

$$w(z_m,z_r)=\frac{1-z_m}{1-(q_0\,z_m+(1-q_0)z_r)+(1-c)\,z_r}+\frac{(1-c)\,z_m}{1-z_r+(1-c)z_r}$$

Selection gradient

$$D(z) = \frac{\partial w(z_m, z_r)}{\partial z_m}\bigg|_{z_m = z_r = z} = \frac{q - c - z(q - c^2)}{(1 - cz)^2},$$

with $q = q_0(z, z)$.



$$w(z_m, z_r) = \frac{1 - z_m}{1 - (q_0 z_m + (1 - q_0) z_r) + (1 - c) z_r} + \frac{(1 - c) z_m}{1 - z_r + (1 - c) z_r}$$

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Computing q, recursively \triangleright More on q

$$q_{t+1} = \frac{1}{\mathcal{N}} + \frac{\mathcal{N} - 1}{\mathcal{N}} \left(1 - \frac{(1-c)z}{1-cz} \right)^2 q_t$$



$$w(z_m,z_r) = \frac{1-z_m}{1-(q_0 z_m + (1-q_0)z_r) + (1-c) z_r} + \frac{(1-c) z_m}{1-z_r + (1-c)z_r}$$

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Computing q, recursively \triangleright More on q

$$q_{t+1} = \frac{1}{N} + \frac{N-1}{N} \left(1 - \frac{(1-c)z}{1-cz} \right)^2 q_t$$
$$q = \frac{1}{1 + \left((2 - \frac{(1-c)z}{1-cz}) \frac{(1-c)z}{1-cz} (N-1) \right)}$$

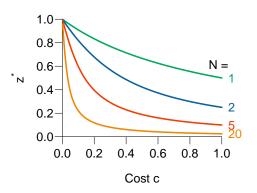
Singular strategy

$$z^* = \frac{1 + 2c \mathcal{N} - \sqrt{1 + 4c^2 (\mathcal{N} - 1) \mathcal{N}}}{2c (1 + c) \mathcal{N}}.$$



Singular strategy

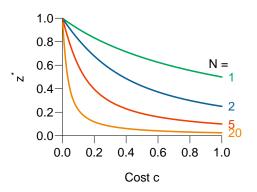
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Invadability



Invadability

$$\frac{\partial^{2} w(z_{m}, z_{r})}{\partial z_{m}^{2}} \bigg|_{z_{m}=z_{r}=z^{*}} = \frac{2}{(1-cz^{*})^{2}} \left[(1-z^{*}) \left(\frac{(q^{*})^{2}}{1-cz^{*}} + \frac{\partial q_{0}(z_{m}, z_{r})}{\partial z_{m}} \right|_{z_{m}=z_{r}=z^{*}} \right) - q^{*} \right]$$
with $q^{*} = q_{0}(z^{*}, z^{*})$



Invadability

$$\begin{split} \frac{\partial^2 w(z_m,z_r)}{\partial z_m^2}\bigg|_{z_m=z_r=z^*} &= \\ \frac{2}{(1-c\,z^*)^2}\Bigg[(1-z^*)\left(\frac{(q^*)^2}{1-c\,z^*} + \frac{\partial q_0(z_m,z_r)}{\partial z_m}\bigg|_{z_m=z_r=z^*}\right) - q^*\Bigg] \\ \text{with } q^* &= q_0(z^*,z^*) \\ \dots \end{split}$$



Invadability

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with
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. . .

▶ In this model, always *z** is always uninvadable [Ajar, 2003].

Invadability

$$\frac{\partial^{2} w(z_{m}, z_{r})}{\partial z_{m}^{2}} \bigg|_{z_{m}=z_{r}=z^{*}} = \frac{2}{(1-cz^{*})^{2}} \left[(1-z^{*}) \left(\frac{(q^{*})^{2}}{1-cz^{*}} + \frac{\partial q_{0}(z_{m}, z_{r})}{\partial z_{m}} \bigg|_{z_{m}=z_{r}=z^{*}} \right) - q^{*} \right]$$

with
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...

- ▶ In this model, always z^* is always uninvadable [Ajar, 2003].
- But with heterogeneity in deme sizes, diversification can occur [Massol et al., 2011]



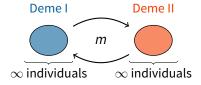
Outline

Introduction

Dispersal and kin competition

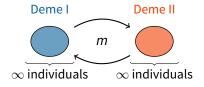
In spatially heterogeneous environments





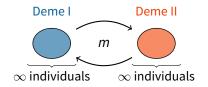
Life-cycle Selection then dispersal.



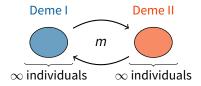


Life-cycle Selection then dispersal. Genotypes AB, Ab, aB, ab.





Life-cycle Selection then dispersal. Genotypes AB, Ab, aB, ab. ► Locus A: local adaptation Fitness:

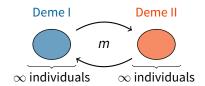


Life-cycle Selection then dispersal. Genotypes AB, Ab, aB, ab. ► Locus A: local adaptation Fitness:

$$\begin{array}{cccc} & \text{in I} & \text{in II} \\ \text{A} & 1+s & 1 \\ \text{a} & 1 & 1+s \end{array}$$

► Locus B: emigration

B
$$z$$
 b z_m .



Life-cycle Selection then dispersal. Genotypes AB, Ab, aB, ab. ► Locus A: local adaptation Fitness:

 $\begin{array}{cccc} & & & & & & & \\ & & & & & & & \\ & & & & & & \\ & & & & & & \\ & & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & \\ & & & \\ & & \\ & & \\ & & \\ & & \\ & & \\ & & \\ & & \\ & & \\ & & \\ & \\ & & \\$

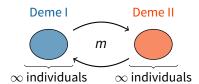
Locus B: emigration

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With AB and aB

Frequency of AB is x in deme I and y in deme II.





Life-cycle Selection then dispersal. Genotypes AB, Ab, aB, ab. Locus A: local adaptation Fitness:

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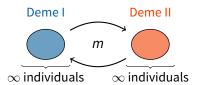
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With AB and aB

Frequency of AB is x in deme I and y in deme II.

$$x' = (1-z)\frac{(1+s)x}{(1+s)x+1-x} + z\frac{y}{y+(1+s)(1-y)}$$
$$y' = z\frac{(1+s)x}{(1+s)x+1-x} + (1-z)\frac{y}{y+(1+s)(1-y)}.$$





Life-cycle Selection then dispersal. Genotypes AB, Ab, aB, ab. Locus A: local adaptation Fitness:

- Locus B: emigration
 - B z b z_m .

With AB and aB

Frequency of AB is x in deme I and y in deme II.

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$$y' = z\frac{(1+s)x}{(1+s)x+1-x} + (1-z)\frac{y}{y+(1+s)(1-y)}.$$

$$\rightarrow$$
 Equilibrium $(\hat{x}, \hat{y}) = (\hat{x}, 1 - \hat{x})$.



Dynamics with the four genotypes

		AB	Ab	аВ	ab
Frequencies:		•	_	-	
	in deme II	<i>y</i> 1	<i>y</i> ₂	y 3	<i>y</i> ₄

Dynamics with the four genotypes

Frequencies: in deme I
$$x_1$$
 x_2 x_3 x_4 in deme II y_1 y_2 y_3 y_4

$$x'_{1} = (1-z)\frac{(1+s)x_{1}}{(1+s)(x_{1}+x_{2})+(x_{3}+x_{4})} + z\frac{y_{1}}{(y_{1}+y_{2})+(1+s)(y_{3}+y_{4})}$$

$$x'_{2} = (1-z_{m})\frac{(1+s)x_{2}}{(1+s)(x_{1}+x_{2})+(x_{3}+x_{4})} + z_{m}\frac{y_{2}}{(y_{1}+y_{2})+(1+s)(y_{3}+y_{4})}$$

$$x'_{3} = \dots$$

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$$x'_{3} = \dots$$

Invasion analysis

Local stability of the equilibrium without b,

$$(\hat{x}, 0, 1 - \hat{x}, 0, \hat{y}, 0, 1 - \hat{y}, 0)$$

► More on stability analysis



$$\begin{split} & \{ \frac{4 \ (1+s)}{4 \ (1+s)} \frac{4 \ (1+s) \ (-1+2z)}{2 \ (2+s-2z-sz+\sqrt{s^2-2s^2z+(2+s)^2z^2})^2}, -\frac{4 \ (1+s) \ (-1+2z)}{2 \ (2+s-2z-sz+\sqrt{s^2-2s^2z+(2+s)^2z^2})}, -\frac{4 \ (1+s) \ (-1+2z)}{2 \ (2+s-2z-sz+\sqrt{s^2-2s^2z+(2+s)^2z^2})}, -\frac{4 \ (1+s) \ (-1+2z)}{2 \ (2+s-2z-sz+(2+s)^2z^2)}, -\frac{4 \ (1+s) \ (2+s-2z-sz+(2+s)^2z^2)}{2 \ (2+s-2z-sz+(2+s)$$

$$\begin{split} & \{ \frac{4 \cdot (1+s) \cdot (-1+2z)}{\left\{(2+s-2z-sz+\sqrt{s^2-2s^2z+(2+s)^2z^2}\right\}^2}, -\frac{4 \cdot (1+s) \cdot (-1+2z)}{\left\{(2+s-2z-sz+\sqrt{s^2-2s^2z+(2+s)^2z^2}\right)^2}, -\frac{4 \cdot (1+s) \cdot (-1+2z)}{\left\{(2+s-2z-sz+\sqrt{s^2-2s^2z+(2+s)^2z^2}\right)^2}, -\frac{4 \cdot (1+s) \cdot (-1+2z)}{\left\{(2+s-2z-sz+\sqrt{s^2-2s^2z+(2+s)^2z^2}\right\}^2}, -\frac{1}{\left\{(2+s-2z-sz+\sqrt{s^2-2s^2z+(2+s)^2z^2}\right\}^2}, -\frac{1}{\left\{(2+s-2z-sz+\sqrt{s^2-2s^2z+(2+s)^2z^2}\right\}^2}, -\frac{1}{\left\{(2+s-2z-sz+\sqrt{s^2-2s^2z+(2+s)^2z^2}\right\}^2}, -\frac{1}{\left\{(2+s-2z-sz+\sqrt{s^2-2s^2z+(2+s)^2z^2}\right\}^2}, -\frac{1}{\left\{(2+s-2z-sz+\sqrt{s^2-2s^2z+(2+s)^2z^2}\right\}^2}, -\frac{1}{\left\{(2+s-2z-sz+(2+s)^2z^2\right\}^2}, -\frac{1}{\left\{(2+s-2z-sz+(2+s)^2z^2\right)^2}, -\frac{1}{\left\{(2+s-2z-sz+(2+s)^2z^2\right\}^2}, -\frac{1}{\left\{(2+s-2z-sz+(2+s)^2z^2\right\}^2}, -\frac{1}{\left\{(2+s-2z-sz+(2+s)^2z^2\right\}^2}, -\frac{1}{\left\{(2+s-2z-sz+(2+s)^2z^2\right)^2}, -\frac{1}{\left\{(2+s-2z-sz+(2+s)^2z^2\right\}^2}, -\frac{1}{\left\{(2+s-2z-sz+(2+s)^2z^2\right)^2}, -\frac{1}{\left\{(2+s-2z-sz+(2+s)^2z^2\right\}^2}, -\frac{1}{\left\{(2+s-2z-sz+(2+s)^2z^2\right)^2}, -\frac{1}{\left\{(2+s-2z-sz+(2+s)^2z^2\right\}^2}, -\frac{1}{\left\{(2+s-2z-sz+(2+s)^2z^2\right)^2}, -\frac{1}{\left\{(2+s-2z-sz+(2+s)^2z^2\right)^2}, -\frac{1}{\left\{(2+s-$$

 \rightarrow All eigenvalues ρ_i such that $|\rho_i| \leq 1$ when $z_m > z$

$$\begin{cases} 4 & (1+s) & (-1+2z) \\ 4 & (1+s) & (-1+2z) \end{cases} + \frac{4 & (1+s) & (-1+2z)}{\left(2+s-2z-s+\sqrt{s^2-2s^2z+(2+s)^2z^2}\right)^2}, \\ -\frac{4 & (1+s) & (-1+2z)}{\left(2+s-2z-s+\sqrt{s^2-2s^2z+(2+s)^2z^2}\right)}, \\ -\frac{4 & (1+s) & (-1+2z)}{\left(2+s-2z-s+2\sqrt{s^2-2s^2z+(2+s)^2z^2}\right)}, \\ -\frac{4 & (1+s) & (-1+2z$$

 \rightarrow All eigenvalues ρ_i such that $|\rho_i| \le 1$ when $z_m > z$ Reduced emigration probabilities are favored.



▶ Kin competition favors the evolution of emigration



- ▶ Kin competition favors the evolution of emigration
- Spatial heterogeneity only does not...

- ▶ Kin competition favors the evolution of emigration
- ➤ Spatial heterogeneity only does not... but dispersal can evolve when local conditions change with time and space.



- Kin competition favors the evolution of emigration
- Spatial heterogeneity only does not... but dispersal can evolve when local conditions change with time and space.
- Dispersal is a complicated trait to study, because it affects spatial structure (→ Lecture 4).



References

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Appendix

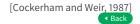
Outline

More on q

Stability analysis

New parameters:

- n Number of demes
- μ Mutation probability (infinite allele model)

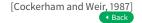


New parameters:

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$$m = \frac{1-z}{1-cz}$$
 Backward dispersal probability



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 Backward dispersal probability

Probability that two individuals came from the same deme and

• are in the same deme: $a = (1 - m)^2 + \frac{m^2}{n-1}$,

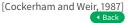


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- ▶ are in different demes: $b = \frac{1 (1 m)^2}{n 1} \frac{m^2}{(n 1)^2}$.



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Probabilities of identity by descent, with replacement:

▶ In the same deme: $q_{0,t+1} = \frac{1}{N} + \frac{N-1}{N} (1-\mu)^2 \left(a \, q_{0,t} + (1-a) \, q_{1,t} \right)$,



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- ▶ In the same deme: $q_{0,t+1} = \frac{1}{N} + \frac{N-1}{N} (1-\mu)^2 (a q_{0,t} + (1-a) q_{1,t})$,
- ▶ In different demes: $q_{1,t+1} = (1 \mu)^2 (b q_{0,t} + (1 b) q_{1,t})$,

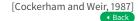
[Cockerham and Weir, 1987]

More on q(2)

$$q_{0,t+1} = \frac{1}{N} + \frac{N-1}{N} (1-\mu)^2 (a q_{0,t} + (1-a) q_{1,t}),$$

$$q_{1,t+1} = (1-\mu)^2 (b q_{0,t} + (1-b) q_{1,t}),$$

Order of limits



São Paulo, Jan 2017

F. Débarre

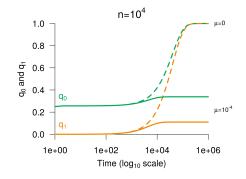
More on q(2)

$$q_{0,t+1} = \frac{1}{N} + \frac{N-1}{N} (1-\mu)^2 (a q_{0,t} + (1-a) q_{1,t}),$$

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Order of limits

When $\mu=0$, $q_{0,\infty}=q_{1,\infty}=1$.



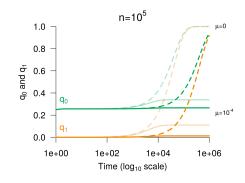
[Cockerham and Weir, 1987]

More on q(2)

$$\begin{split} q_{0,t+1} &= \frac{1}{\mathcal{N}} + \frac{\mathcal{N}-1}{\mathcal{N}} (1-\mu)^2 \left(a \, q_{0,t} + (1-a) \, q_{1,t} \right), \\ q_{1,t+1} &= \left(1 - \mu \right)^2 \left(b \, q_{0,t} + (1-b) \, q_{1,t} \right), \end{split}$$

Order of limits

- $\text{When } \mu = \textbf{0,} \\ q_{\textbf{0},\infty} = q_{\textbf{1},\infty} = \textbf{1.}$
- ▶ When $n \to \infty$, $q_{1,\infty} = 0$ and $q_{0,\infty} = \frac{1}{N} + \frac{N-1}{N} (1-\mu)^2 (a q_{0,\infty})$.



[Cockerham and Weir, 1987]

Outline

More on q

Stability analysis

Model

$$N_1(t+1) = G_1(N_1(t), N_2(t), \dots, N_k(t))$$

$$N_2(t+1) = G_2(N_1(t), N_2(t), \dots, N_k(t))$$

$$\vdots$$

$$N_k(t+1) = G_k(N_1(t), N_2(t), \dots, N_k(t))$$

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Equilibrium

$$ilde{\mathbf{N}}=(ilde{N_1},\dots, ilde{N_k})$$
, such that $G_1(ilde{N_1},\dots, ilde{N_k})= ilde{N_1}$ \vdots $G_k(ilde{N_1},\dots, ilde{N_k})= ilde{N_k}$

Write system of equations for the change over time of a small derivation from the equilibrium

Deviations from equilibrium

Define
$$n_i(t) = N_i(t) - \tilde{N}_i$$
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Write system of equations for the change over time of a small derivation from the equilibrium, and get a linear approximation of this system (Taylor series)

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$$\begin{split} n_i(t+1) &= G_i(N_1(t), \dots, N_k(t)) - \tilde{N}_i \\ &\approx 0 + \left. \frac{\partial G_i}{\partial N_1} \right|_{\mathbf{N}(t) = \tilde{\mathbf{N}}} \left(N_1(t) - \tilde{N}_1 \right) + \dots + \left. \frac{\partial G_i}{\partial N_k} \right|_{\mathbf{N}(t) = \tilde{\mathbf{N}}} \left(N_k(t) - \tilde{N}_k \right). \end{split}$$

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$$\approx 0 + \frac{\partial G_{i}}{\partial N_{1}} \bigg|_{\mathbf{N}(t) = \tilde{\mathbf{N}}} \underbrace{\left(N_{1}(t) - \tilde{N}_{1}\right)}_{n_{1}(t)} + \dots + \frac{\partial G_{i}}{\partial N_{k}} \bigg|_{\mathbf{N}(t) = \tilde{\mathbf{N}}} \underbrace{\left(N_{k}(t) - \tilde{N}_{k}\right)}_{n_{k}(t)}.$$

In matrix form:

$$\begin{pmatrix} n_1 \\ \vdots \\ n_k \end{pmatrix} (t+1) = \begin{pmatrix} \frac{\partial G_1}{\partial N_1} & \cdots & \frac{\partial G_1}{\partial N_k} \\ \vdots & \cdots & \vdots \\ \frac{\partial G_k}{\partial N_1} & \cdots & \frac{\partial G_k}{\partial N_k} \end{pmatrix} \bigg|_{\mathbf{N} = \tilde{\mathbf{N}}} \cdot \begin{pmatrix} n_1 \\ \vdots \\ n_k \end{pmatrix} (t)$$

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In matrix form:

$$\underbrace{\begin{pmatrix} n_1 \\ \vdots \\ n_k \end{pmatrix}}_{\mathbf{n}(t+1)} (t+1) = \underbrace{\begin{pmatrix} \frac{\partial G_1}{\partial N_1} & \cdots & \frac{\partial G_1}{\partial N_k} \\ \vdots & \cdots & \vdots \\ \frac{\partial G_k}{\partial N_1} & \cdots & \frac{\partial G_k}{\partial N_k} \end{pmatrix}}_{\mathbf{N} = \tilde{\mathbf{N}}} \cdot \underbrace{\begin{pmatrix} n_1 \\ \vdots \\ n_k \end{pmatrix}}_{\mathbf{n}(t)} (t)$$



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with the c_i constants determined by the initial conditions, and $\nu_{(i)}$ an eigenvector associated to the eigenvalue λ_i , i.e., $\mathbf{J} \cdot \nu_{(i)} = \lambda_i \, \nu_{(i)}$.

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Leading eigenvalue: eigenvalue with the largest modulus Modulus: for a complex number $\lambda = A + iB$,

$$|\lambda| = \sqrt{A^2 + B^2}.$$

Inspect the eigenvalues of J

