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 inbreeding

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- 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



Model



Offspring production



Emigration probabilities: x = 0, y > 0

Cost of emigration c = 1 - p.





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Cost of emigration c = 1 - p.

Invasion fitness

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

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Selection gradient

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



(c = 0.3)

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Convergence stability

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



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Convergence stability

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

Uninvadability

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



(c = 0.3)

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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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 ($\mu \rightarrow$ 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]

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

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Selection gradient

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

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 \frown 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 • More on q

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

São Paulo, Jan 2017

Singular strategy

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

Singular strategy

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Singular strategy

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×

Invadability

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$$\frac{\partial^2 w(z_m, z_r)}{\partial z_m^2} \bigg|_{z_m = z_r = z^*} = \frac{2}{(1 - c z^*)^2} \left[(1 - z^*) \left(\frac{(q^*)^2}{1 - c z^*} + \frac{\partial q_0(z_m, z_r)}{\partial z_m} \bigg|_{z_m = z_r = z^*} \right) - q^* \right]$$

with $q^* = q_0(z^*, z^*)$

Invadability

$$\frac{\partial^2 w(z_m, z_r)}{\partial z_m^2} \bigg|_{z_m = z_r = z^*} = \frac{2}{(1 - c \, z^*)^2} \left[(1 - z^*) \left(\frac{(q^*)^2}{1 - c \, z^*} + \frac{\partial q_0(z_m, z_r)}{\partial z_m} \bigg|_{z_m = z_r = z^*} \right) - q^* \right]$$

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. . .

Invadability

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

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

Invadability

$$\frac{\partial^2 w(z_m, z_r)}{\partial z_m^2} \bigg|_{z_m = z_r = z^*} = \frac{2}{(1 - c z^*)^2} \left[(1 - z^*) \left(\frac{(q^*)^2}{1 - c z^*} + \frac{\partial q_0(z_m, z_r)}{\partial z_m} \bigg|_{z_m = z_r = z^*} \right) - q^* \right]$$

with $q^* = q_0(z^*, z^*)$

- ▶ 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: in I in II A 1+s 1 a 1 1+s



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

 in I
 in II
 A
 1 + s

 Locus B: emigration B

b z_m .



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

With AB and aB

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

Locus A: local adaptation Fitness: in II in I 1 + s1 Α 1 1 + sа Locus B: emigration В Ζ b Z_m.



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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: in I in II 1 + sΑ 1 1 1 + sа Locus B: emigration В Ζ b Z_m .

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

Dynamics with the four genotypes

		AB	Ab	aВ	ab
Frequencies:	in deme I	<i>x</i> ₁	<i>x</i> ₂	<i>X</i> 3	<i>X</i> 4
	in deme II	<i>Y</i> 1	<i>Y</i> 2	Уз	<i>Y</i> 4

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$$\begin{aligned} x_1' &= (1-z)\frac{(1+s)x_1}{(1+s)(x_1+x_2)+(x_3+x_4)} + z\frac{y_1}{(y_1+y_2)+(1+s)(y_3+y_4)} \\ x_2' &= (1-z_m)\frac{(1+s)x_2}{(1+s)(x_1+x_2)+(x_3+x_4)} + z_m\frac{y_2}{(y_1+y_2)+(1+s)(y_3+y_4)} \\ x_3' &= \dots \end{aligned}$$

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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

$$\begin{split} & \text{ev} = \text{Eigenvalue}(3)a(\ //\ \text{FullSign}(1f) \\ & \frac{4(1+s)}{\left(2+s-2z-sz+\sqrt{s^2-2s^2}z+(2+s)^2z^2\right)^2}, \quad -\frac{4(1+s)(-1+2z)}{\left(2+s-2z-sz+\sqrt{s^2-2s^2}z+(2+s)^2z^2\right)^2}, \quad -\frac{4(1+s)(-1+2z)}{\left(2+s-2z-sz+\sqrt{s^2-2s^2}z+(2+s)^2z^2\right)^2}, \\ & 1_* \left[4+2\sqrt{s^2-2s^2}z+(2+s)^2z^2+s^2(-1+z)(-1+zm)+4z(-1+zm)-s\left(4-4z+\sqrt{s^2-2s^2}z+(2+s)^2z^2\right)(-1+zm)-2z\left(2+\sqrt{s^2-2s^2}z+(2+s)^2z^2\right)zm-\sqrt{2}, \\ & \left[\left(s^2(-1+z)^2+2\left(1+2z^2+\sqrt{s^2-2s^2}z+(2+s)^2z^2-z^2(2+z)^2z^2+(2+s)^2z^2\right)(-1+zm)-2z(2+\sqrt{s^2-2s^2}z+(2+s)^2z^2-z^2(2+\sqrt{s^2-2s^2}z+(2+s)^2z^2))\right)\right] \\ & - \left[2+s-2z-sz+\sqrt{4sz+(s(-1+z)+2z)^2}\right]^2, \quad \left[4+2\sqrt{s^2-2s^2}z+(2+s)^2z^2\right] \\ & \left[2+s-2z-sz+\sqrt{4sz+(s(-1+z)+2z)^2}\right]^2, \quad \left[4+2\sqrt{s^2-2s^2}z+(2+s)^2z^2\right] \\ & z+2\sqrt{s^2-2s^2}z+(2+s)^2z^2\right] \\ & - \left[2+\sqrt{s^2-2s^2}z+(2+s)^2z^2\right] \\ & - \left[2+\sqrt{s^2-2s^2}z+(2+s)^2z$$

$$\begin{split} & e^{-z} \in \mathsf{figenvalue}(\mathsf{Iac})/// \mathsf{fullSimplify} \\ & \{ \frac{4(1+5)}{\left(2+5-2z-5z+\sqrt{4^2-2s^2}z+(2+5)^2z^2\right)^2}, -\frac{4(1+5)(-1+2z)}{\left(2+5-2z-5z+\sqrt{8^2-2s^2}z+(2+5)^2z^2\right)^2}, -\frac{4(1+5)(-1+2z)}{\left(2+5-2z-5z+\sqrt{8^2-2s^2}z+(2+5)^2z^2\right)^2}, -\frac{4(1+5)(-1+2z)}{\left(2+5-2z-5z+\sqrt{8^2-2s^2}z+(2+5)^2z^2\right)^2} \right) \\ & \mathsf{1}_1 \left(4+2\sqrt{8^2-2s^2}z+(2+5)^2z^2 + 8^2(-1+z)(-1+zm) + 4z(-1+zm) - 5\left(4-4z+\sqrt{8^2-2s^2}z+(2+5)^2z^2 + (2+5)^2z^2 \right) (-1+zm) - 2z\left(2+\sqrt{8^2-2s^2}z+(2+5)^2z^2 + (2+5)^2z^2 + (2+5)^2z^2$$

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

$$\begin{split} & e^{-z} \in \mathsf{figenvalue}(\mathsf{Iac}) / t^{\mathsf{VIIISimplify}} \\ & \{ \frac{4 (1+5)}{(2+5-2z-5z+\sqrt{4^2-2s^2}z+(2+5)^2z^2)^2} \}^2 + \frac{4 (1+5) (-1+2z)}{(2+5-2z-5z+\sqrt{8^2-2s^2}z+(2+5)^2z^2)^2} \}^2 + \frac{4 (1+5) (-1+2z)}{(2+5-2z-5z+\sqrt{8^2-2s^2}z+(2+5)^2z^2)^2} \}^2 \\ & 1_{\mathsf{I}} \left(4+2\sqrt{8^2-2s^2}z+(2+5)^2z^2 + 8^2 (-1+z) (-1+zm) + 4z (-1+zm) - 5 \left(4-4z+\sqrt{8^2-2s^2}z+(2+5)^2z^2 \right) (-1+zm) - 2z (2+\sqrt{8^2-2s^2}z+(2+5)^2z^2) (-1+zm) + 4z (-1+zm) - 2z (2+\sqrt{8^2-2s^2}z+(2+5)^2z^2) (-1+zm) + 2z 2z (2+\sqrt{8^2-2s^2}z+(2+5)^2z^2) (-1+zm)$$

 \rightarrow All eigenvalues ρ_i such that $|\rho_i| \leq 1$ when $z_m > z$ Reduced emigration probabilities are favored. ▶ Kin competition favors the evolution of emigration

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- Spatial heterogeneity only does not...

A few take-home messages

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

A few take-home messages

- ▶ 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)



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

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



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

Probability that two individuals came from the same deme and

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



New parameters:

- n Number of demes
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$$m = \frac{1-z}{1-cz}$$
 Backward dispersal probability

Probability that two individuals came from the same deme and

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

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

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- μ Mutation probability (infinite allele model)

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 Backward dispersal probability

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

Probabilities of identity by descent, with replacement:

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

New parameters:

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

$$m = \frac{1-z}{1-cz}$$
 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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Probabilities of identity by descent, with replacement:

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

• In different demes:
$$q_{1,t+1} = (1-\mu)^2 (b q_{0,t} + (1-b) q_{1,t})$$
,



More on q (2)

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

Order of limits



More on q (2)

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

Order of limits

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



[Cockerham and Weir, 1987]

More on q (2)

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

Order of limits



Time (log₁₀ scale)

[Cockerham and Weir, 1987]
Outline

More on q

Stability analysis

Model

$$N_{1}(t+1) = G_{1}(N_{1}(t), N_{2}(t), \dots, N_{k}(t))$$

$$N_{2}(t+1) = G_{2}(N_{1}(t), N_{2}(t), \dots, N_{k}(t))$$

$$\vdots$$

$$N_{k}(t+1) = G_{k}(N_{1}(t), N_{2}(t), \dots, N_{k}(t))$$

Model

$$N_{1}(t+1) = G_{1}(N_{1}(t), N_{2}(t), \dots, N_{k}(t))$$

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$$\vdots$$

$$N_{k}(t+1) = G_{k}(N_{1}(t), N_{2}(t), \dots, N_{k}(t))$$

Equilibrium

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

$$G_1(\tilde{N}_1,\ldots,\tilde{N}_k) = \tilde{N}_1$$
$$\vdots$$
$$G_k(\tilde{N}_1,\ldots,\tilde{N}_k) = \tilde{N}_k$$

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

Deviations from equilibrium

Define $n_i(t) = N_i(t) - \tilde{N}_i$.

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

Deviations from equilibrium

Define $n_i(t) = N_i(t) - \tilde{N}_i$.

$$n_i(t+1) = G_i(N_1(t),\ldots,N_k(t)) - \tilde{N}_i$$

2 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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In matrix form:

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

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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., $\mathbf{J} \cdot \boldsymbol{\nu}_{(i)} = \lambda_i \boldsymbol{\nu}_{(i)}$.

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[?]

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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





▲ Back