

# Integrodifference equations in spatial ecology

## Lecture 10: Dispersal success approximation

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# Approximations

- simplify analysis
- simplify computations
- accommodate data-poor situations
- prioritize data collection

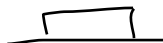
think:

$$N_{t+1} = \bar{F}(N_t) \quad \text{vs}$$

$$N_{t+1}(x) = \int K(x,y) F(N_t(y)) dy$$

Recall:

$$N_{t+1} = S F(N_t)$$



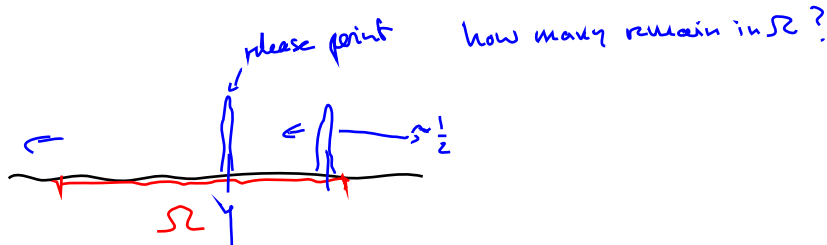
S: prob. of staying in the patch

$K(x,y)$

empirically:  $\forall y$ : release many organisms at  $y$

follow each of them.

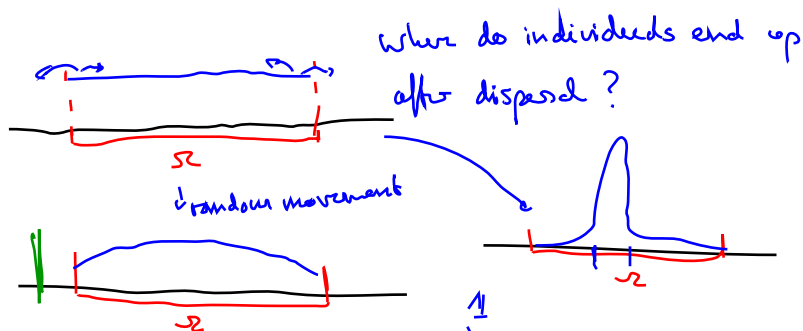
# Point-release experiment



Dispersal success function  $S(y) = \int_{\Omega} K(x, y) dx$  staying in  $\Omega$ .

Average dispersal success:  $\bar{S} = \frac{1}{|\Omega|} \int_{\Omega} S(y) dy$

# Area release experiment



Redistribution function  $U(x) = \int_{\Omega} K(x, y) dy$

NOT a probability, but  $\bar{U} = \bar{S} \leq 1$ .

$$\bar{U} = \frac{1}{|\Omega|} \int_{\Omega} U(x) dx = \frac{1}{|\Omega|} \iint |\kappa(x, y)| dx dy = \bar{S} \leq 1$$

# Example: Laplace

$$K(x-y) = \frac{a}{2} e^{-a|x-y|}$$

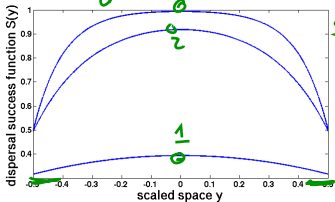
$$S(y) = U(y) = 1 - e^{-\hat{L}/2} \cosh(ay),$$

and the average dispersal success is

$$\bar{S} = 1 - \frac{1 - e^{-\hat{L}}}{\hat{L}}$$

$$\frac{1}{2} \left( 1 - e^{-ax} e^{-\frac{aL}{2}} - e^{ax} e^{-\frac{aL}{2}} + 1 \right)$$

$$1 - e^{-\frac{aL}{2}} \cosh(ax)$$



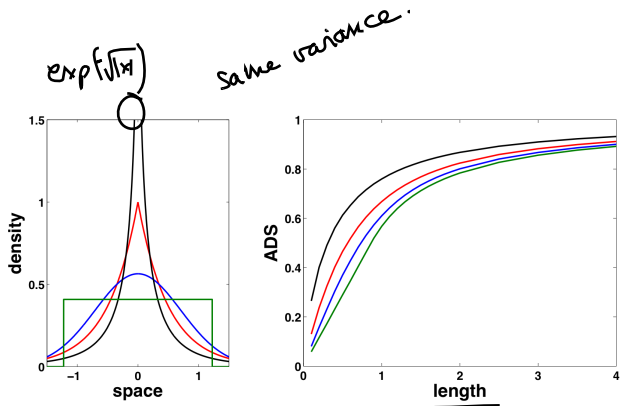
$$[-\frac{1}{2}, \frac{1}{2}] \quad a_1, a_2, a_3$$

$$S(x) = \int_{-L/2}^{L/2} \frac{a}{2} e^{-a|x-y|} dy = \int_{-L/2}^x \frac{a}{2} e^{-a(x-y)} dy + \int_x^{L/2} \frac{a}{2} e^{-a(y-x)} dy$$

$$= \frac{a}{2} e^{-ax} \left( \frac{1}{a} e^{ay} \right) \Big|_{-L/2}^x + \frac{a}{2} e^{ax} \left( -\frac{1}{a} e^{-ay} \right) \Big|_x^{L/2}$$

$$= \frac{1}{2} \left( e^{-ax} (e^{ax} - e^{-aL/2}) + e^{ax} (-e^{-L/2} + e^{-ax}) \right)$$

# Comparison of kernels



Top hat  
Gaussian  
Laplace

# Dispersal success approximation of $N^*$ (van Kirk & Lewis 1995/1997)

steady state

$$\rightarrow N^*(x) = \int_{\Omega} K(x, y) F(N^*(y)) dy,$$

$$\bar{N}^* = \frac{1}{|\Omega|} \int_{\Omega} N^*(x) dx$$

average

$$F(N^*(x)) = F(\bar{N}^*) + F'(\bar{N}^*) (N^*(x) - \bar{N}^*)$$

plug in

$$N^*(x) = \int_{\Omega} K(x, y) [F(\bar{N}^*) + F'(\bar{N}^*) (N^*(y) - \bar{N}^*)] dy$$

$$= F(\bar{N}^*) \underbrace{\int_{\Omega} K(x, y) dy}_{u(x)} + \int_{\Omega} \underbrace{K(x, y) F'(\bar{N}^*) (N^*(y) - \bar{N}^*)}_{\text{small}} dy$$

$$N^*(x) \approx F(\bar{N}^*) u(x)$$

integrate

$$\bar{N}^* = F(\bar{N}^*) \bar{S} + \frac{1}{|\Omega|} \int_{\Omega} \underbrace{S(y) F'(\bar{N}^*) (N^*(y) - \bar{N}^*)}_{\text{small \& steady}} dy$$

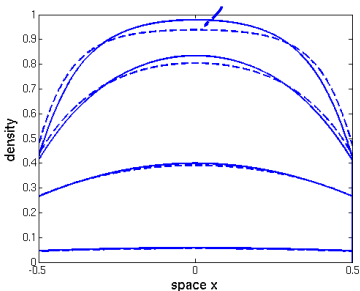


Result =  $\bar{N}^* = \bar{S} F(\bar{N}^*)$   
 $N^*(x) \approx F(\bar{N}^*) u(x)$

Small

# Illustration

Steady state  $N^*(x)$  vs  $F(N^*) \cdot U(x)$  with Laplace kernel



$$\underline{F(N^*)S(x)}$$

Dispersal Success approximation.



# Eigenvalue approximation

$$\lambda \phi(x) = \int_{\Omega} K(x, y) \phi(y) dy, \quad \bar{\phi} = \frac{1}{|\Omega|} \int_{\Omega} \phi(x) dx = \underline{1}.$$

$$\begin{aligned} \lambda \phi(x) &= \int K(x, y) [1 + b - 1] dy \\ &= \int K(x, y) dy + \int K(x, y) (b(y) - 1) dy \\ &= u(x) + \text{"small"} \end{aligned}$$

$$\lambda \cdot 1 = \bar{5}$$

$$\Rightarrow \lambda_{ap} = \bar{5}, \quad \phi_{ap} = \frac{u(x)}{\bar{5}}$$

# A conservative estimate

Lemma:

Assume that  $K$  is symmetric. Then  $\lambda_{\text{ap}} \leq \bar{\lambda}$ .

$\Rightarrow$  if pop. can persist according to  $\bar{\lambda}$ , then it can persist according to  $\lambda$

Proof:

$$\lambda = \max_{\substack{\|u\|=1 \\ \int_{\Omega} u^2}} \iint_{\Omega} u(x) K(x,y) u(y) dy dx \Rightarrow \frac{1}{|\Omega|} \iint_{\Omega} K(x,y) dx dy = \bar{\lambda}$$

choose  $u(x) = \frac{1}{\sqrt{|\Omega|}}$

# Improving the approximation: weighted average

$$\lambda \phi = \int K(x, y) \phi(y) dy$$

$$\lambda \int \phi \cdot 1 \cdot dx = \iint K(x, y) \phi(y) dy dx = \int S(y) \phi(y) dy$$

Notation:  $\langle f, g \rangle = \int_{\Omega} f(x)g(x) dx$   $\langle S, \phi \rangle$

$$\bar{S} = \frac{1}{|\Omega|} \iint K(x, y) dy dx$$

$\int S(y) \cdot 1 dy$   
 $\langle S, 1 \rangle$

$$|\Omega| = \int 1 \cdot 1 dx = \langle 1, 1 \rangle$$

Eigenvalue:  $\lambda = \frac{\langle S, \phi \rangle}{\langle 1, \phi \rangle}$

we have  $\phi(x) \approx U(x)$

Average:  $\bar{S} = \frac{\langle S, 1 \rangle}{\langle 1, 1 \rangle}$

Weighted Average:  $\lambda_{ap}^{(2)} = \frac{\langle S, U \rangle}{\langle 1, U \rangle} = \frac{\int_{\Omega} S(x)U(x) dx}{\int_{\Omega} U(x) dx}$

Reimer of 2016: improve  $\bar{S}$



average  $\bar{S} = \frac{1}{|\Omega|} \int S(y) dy$  with uniform weight

Ideal: weight =  $N^*(x)$ , use  $N^*(x) \sim U(x)$   $\bar{S} := \frac{\int S(y) U(y) dy}{\int U(y) dy}$

# Understanding the approximation

Use

$$\text{cov}(S, U) = \frac{1}{|\Omega|} \int_{\Omega} (S(y) - \bar{S})(U(y) - \bar{S}) dy = \frac{1}{|\Omega|} \int_{\Omega} S(y)U(y) dy - \bar{S}^2$$

and re-write

$$\lambda_{\text{ap}}^{(2)} = \bar{S} \left( 1 + \frac{1}{\bar{S}^2} \text{cov}(S, U) \right)$$

$$\lambda_{\text{ap}} = \bar{S} \leq \lambda$$

if  $K$  is symmetric

if  $K$  is symmetric, then  $\text{cov}(S, U) = \text{var}(S) > 0$

$$\Rightarrow \lambda_{\text{ap}}^{(2)} > \lambda_{\text{ap}} \rightarrow \text{right direction.}$$

Jonas's challenge: show that  $\lambda_{\text{ap}}^{(2)} \leq \lambda$  if  $K$  is symmetric!

# The power method I

Goal: Given Matrix  $A$ , with pos. dom. eigenvalue  $\lambda$ , Find  $\lambda$  numerically

Power method: define  $\phi_n$  by  $\phi_0$  has norm 1.

$$\phi_{n+1} = \frac{A \phi_n}{\|A \phi_n\|} \quad \text{has norm 1} \quad \text{then} \quad \|A \phi_n\| \rightarrow \lambda \quad \text{as } n \rightarrow \infty$$
$$\phi_n \rightarrow \phi$$

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Apply these ideas to  $\int K(x,y) \phi(y) dy$ .

Start with  $\phi_0(x) = \frac{1}{|\Omega|}$  const. then  $\|\phi_0\|_{L^2} = 1$

$$\phi_1 = \frac{1}{\%} \int K(x,y) \frac{1}{|\Omega|} dy = \frac{1}{\%} \frac{1}{|\Omega|} U(x) = \frac{1}{\% |\Omega|} U(x)$$

$\% = \int \frac{1}{|\Omega|} U(x) dx = \bar{U}$        $\lambda_{op}, \phi_{op}$  are first step of power method.

## The power method II

$$\phi_1 = \frac{1}{\int |\Omega|} U(x) dx$$

$$\phi_2 = \frac{1}{\int \cdot} \int K(x,y) \frac{1}{\int |\Omega|} U(y) dy \quad \rightarrow \text{need to know } K(x,y)$$

$$\therefore = \iint K(x,y) \frac{1}{\int |\Omega|} U(y) dy dx = \int S(y) U(y) dy \frac{1}{\int |\Omega|} = \lambda_{\text{of}}^{(2)}$$

$$\lambda \phi = \int K(x,y) \phi(y) dy$$

$\uparrow$   
 $R(y)$

Eigensvalue .. in (2004)

$$\lambda \phi(x) = \int K(x,y) \underbrace{F'(N^*(y))}_{R(y)} \phi(y) dy$$

$\downarrow$   
 $N^*(y) \approx \bar{N}^*$  How good is this?

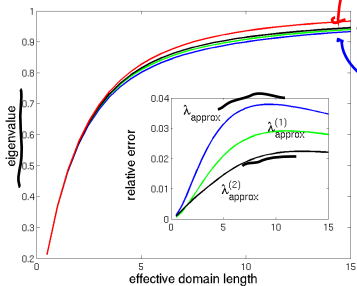
# Symmetric example

The Laplace kernel

$$1 - \frac{1 - e^{-\hat{t}}}{\hat{t}}$$

$$\lambda_{\text{ap}} = \bar{S}$$

$$\lambda_{\text{ap}}^{(2)} = \bar{S} + \text{var}(S)$$

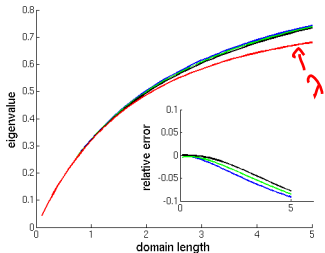


# Asymmetric example: the eigenvalue

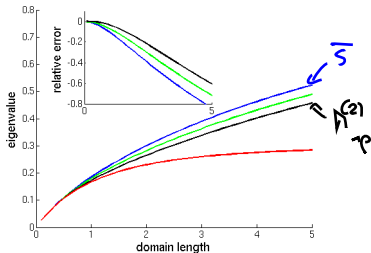


The asymmetric Laplace kernel

$$\lambda_p^{(2)} = \bar{s} \left( 1 + \frac{\text{cov}(s, u)}{\bar{s}^2} \right)$$



weak asymmetry



stronger asymmetry



# Asymmetric example: the steady state

