# Diffusion approximation for a spatially realistic structured metapopulation model

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AUSTRALIAN RESEARCH COUNCIL Centre of Excellence for Mathematics and Statistics of Complex Systems



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- - Different patch sizes

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# **Analytical Models**

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• No spatial structure - homogenous mixing

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- Day and Possingham (1995)
  - Variability in patch size and position
  - Discrete-time Markov chain
- But...
  - does not account for local population dynamics
  - computationally intensive ( $2^k \times 2^k$  for a k-patch system)

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- Surprise surprise... a continuous-time Markov chain!

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- Patch abundance local population dynamics?

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- $e_i$  *i*-th unit vector

### **The model - CTMC**

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 $Q = (q(i,j), i, j \in S),$ 

so that q(i, j) represents the rate of transition from state i to state j, for  $j \neq i$ , and q(i, i) = -q(i), where

$$q(i) := \sum_{j \neq i} q(i, j) \ (< \infty)$$

represents the total rate out of state *i*.

### **The model - transition rates**

#### **Birth**

$$q(n, n + e_i) = b \frac{n_i}{N_i} (N_i - n_i), \ \forall i \in K$$

#### Migration

$$q(n, n - e_i + e_j) = \gamma_{ij} \frac{n_i}{N_j} (N_j - n_j), \ \forall i \neq j, i, j \in K$$

#### Death

$$q(n, n - e_i) = \mu n_i, \ \forall i \in K$$

where  $K = \{1, ..., k\}$ .

*Definition*: A one-parameter family of Markov chains  $\{P_{\nu}, \nu > 0\}$  with state space  $S_{\nu} \subset \mathbb{Z}^{D}$  is called density dependent if there exists a set  $E \subseteq \mathbb{R}^{D}$  and a continuous function  $f : E \times \mathbb{Z}^{D} \to \mathbb{R}$ , such that

$$q_{\nu}(k,k+l) = \nu f\left(\frac{k}{\nu},l\right), \qquad l \neq 0.$$

[Kurtz (1970)]

# **Functional law of large numbers**

*Theorem*: Suppose that f(x, l) is bounded for each l and that F, where  $F(x) = \sum_{l} lf(x, l)$ , is Lipschitz continuous on E. Then, if

$$\lim_{\nu \to \infty} X_{\nu}(0) = x_0,$$

we have, for fixed  $\tau > 0$  and for all  $\epsilon > 0$ , that

$$\lim_{\nu \to \infty} \Pr\left( \sup_{t \le \tau} |X_{\nu}(t) - X(t, x_0)| > \epsilon \right) = 0,$$

where  $X(\cdot, x)$  is the unique trajectory satisfying

$$X(0,x) = x, \quad X(t,x) \in E, \ 0 \le t \le \tau, \quad \frac{\partial}{\partial t} X(t,x) = F(X(t,x)).$$

[Kurtz (1970)]

### **Functional central limit theorem**

 $\sqrt{\nu} \left( X_{\nu}(t) - X(t, x_0) \right) \rightarrow$ Gaussian Diffusion

$$\sqrt{\nu} \left( X_{\nu}(t) - x^* \right) \to N(0, \Sigma_t)$$

#### Long-term

 $\mathsf{E}(X_{\nu}) \approx x^*$  $\mathsf{Var}(X_{\nu}) \approx \frac{1}{\nu} \Sigma$  where  $\Sigma = \lim_{t \to \infty} \Sigma_t$ .

#### [Kurtz (1971)]

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$$f(x, x + e_i) = b \frac{x_i}{\rho_i} (\rho_i - x_i), \ \forall i \in K$$

$$f(x, x - e_i + e_j) = \gamma_{ij} \frac{x_i}{\rho_j} \left( \rho_j - x_j \right), \ \forall i \neq j, i, j \in K$$
$$f(x, x - e_i) = \mu x_i, \ \forall i \in K.$$

### **Deterministic approximation**

The functional law of large numbers gives us

$$\frac{dx}{dt} = F(x)$$

Therefore we have a system of k differential equations with the *i*-th given by

$$\frac{dx_i}{dt} = \left(b - \mu - \sum_{j \neq i} \gamma_{ij}\right) x_i + \sum_{j \neq i} \gamma_{ij} x_j + \frac{x_i}{\rho_i} \left[\sum_{j \neq i} \frac{\gamma_{ij}}{\rho_j} x_j (\rho_i - \rho_j) - bx_i\right]$$

$$\frac{dx_1}{dt} = \left(b - \mu - \gamma - \frac{b}{\rho}x_1\right)x_1 + \gamma x_2$$
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Fixed points and stability

Trivial fixed point: (0,0)Stable if  $b - \mu < 0$ , saddle if  $0 < b - \mu < 2\gamma$  and unstable if  $b - \mu > 2\gamma$ .

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SL fixed point:  $\left(\frac{1}{2b}(b-\mu), \frac{1}{2b}(b-\mu)\right)$ Unstable if  $b-\mu < -2\gamma$ , saddle if  $-2\gamma < b-\mu < 0$ and stable if  $b-\mu > 0$ .

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Another pair: Real and saddles if  $|b - \mu| > 2\gamma$ .





### **Different patch sizes?**



### **General stability analysis**

The nonlinear system can be written in the linearised form

$$\frac{d\underline{\mathbf{x}}}{dt} = A\underline{\mathbf{x}} + h(\underline{\mathbf{x}})$$

where

$$A = \Gamma + (b - \mu)I$$

in which  $\Gamma$  is a q-matrix with diagonal entries given by  $-\sum_{j\neq i} \gamma_{ij}$  and off-diagonal entries  $\gamma_{ij}$ , and  $h(\underline{\mathbf{x}})$  consists of higher order terms such that  $||h(\underline{\mathbf{x}})|| = o(||\underline{\mathbf{x}}||)$ , as  $||\underline{\mathbf{x}}|| \to 0$ .

### **General stability analysis**

#### Determine stability by considering the eigenvalues $\sigma$ of A

$$A\underline{\mathbf{X}} = \sigma\underline{\mathbf{X}}$$

so we have

 $(b-\mu)\underline{\mathbf{X}} + \Gamma\underline{\mathbf{X}} = \sigma\underline{\mathbf{X}}$ 

and therefore

$$\Gamma \underline{\mathbf{X}} = [\sigma - (b - \mu)] \,\underline{\mathbf{X}}$$

so finally we have

$$\sigma_i = \lambda_i + b - \mu$$

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where  $\lambda_i$  is the *i*-th eigenvalue of  $\Gamma$ . Therefore, SL fixed point is always stable if

 $b > \mu$ .

If the patches are close to homogenous in size, we can approximate the equilibrium mean population density by using the logistic model [Verhulst (1838)]

$$\frac{dy}{dt} = by(1-y) - \mu y$$

with equilibrium  $y^* = \frac{1}{b}(b - \mu)$ , which is stable if  $b > \mu$ , where  $y = \sum_{i=1}^{k} x_i$ .

The equilibrium density at each patch will then be given by

$$x_i^* = \frac{1}{kb}(b-\mu), \quad i = \{1, 2, \dots, k\}$$

• Full analysis of model - fixed points and stability

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- Listen to Michael's talk and then have some pizza!

### **Acknowledgements**

Phil Pollett and Hugh Possingham

Ben Cairns and David Sirl

The University of Queensland

#### and



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