# Infinite-patch metapopulation models: branching, convergence and chaos

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\*Buckley, F.M. and Pollett, P.K. (2010) Limit theorems for discrete-time metapopulation models. Probability Surveys 7, 53-83.

















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*Extinction*: occupied patches remain occupied independently with probability *s*.

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*Notation*: Bin(m, p) is a binomial random variable with m trials and success probability p.

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#### **CE Model**



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#### **CE Model - Evanescence**



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## **CE Model** c'(0) < (1-s)/s



#### **CE Model** c'(0) > (1-s)/s







**Prelude** If c(0) = 0 and c has a continuous second derivative near 0, then, for fixed n,

$$Bin(N-n, c(n/N)) \xrightarrow{D} Poi(mn), \text{ as } N \to \infty,$$

where m = c'(0).

#### **Infinite-patch SPOM**

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**Claim** The process  $(n_t, t = 0, 1, ...)$  is a *branching process* (Galton-Watson process) whose offspring distribution has pgf  $G(z) = (1 - s + sz)e^{-ms(1-z)}$ .

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(We think of the census times as marking the 'generations', the 'particles' being the occupied patches, and the 'offspring' being the occupied patches that they notionally replace in the succeeding generation.)

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So, for example,  $\mathsf{E}(n_t|n_0) = n_0\mu^t \ (t \ge 1)$ .

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**Theorem** Extinction occurs with probability 1 if and only if  $m \le (1 - s)/s$ ; otherwise total extinction occurs with probability  $\eta^{n_0}$ , where  $\eta$  is the unique fixed point of *G* on the interval (0, 1).

# **CE Model** c'(0) < (1-s)/s $(\eta = 1)$



# **CE Model** c'(0) > (1-s)/s $(\eta^{n_0} = 0.0020837)$



Assume the following structure:

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where  $m(n) \ge 0$ .

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We may take N to be simply  $n_0$  or, more generally, following Klebaner\*, we may interpret N as being a 'threshold' with the property that  $n_0/N \rightarrow x_0$  as  $N \rightarrow \infty$ .

\*Klebaner (1993) Population-dependent branching processes with a threshold. Stochastic Process. Appl. 46, 115–127.

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For example,  $\mu(x)$  might be of the form

- $\mu(x) = rx(a x) \ (0 \le x \le a)$  (logistic growth);
- $\mu(x) = x e^{r(1-x)}$  ( $x \ge 0$ ) (Ricker dynamics);
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We can establish a *law of large numbers* for  $X_t^{(N)} = n_t/N$ , the number of occupied patches at census *t* measured *relative to* the threshold.

**Theorem** For the infinite-patch CE model with parameters *s* and  $\mu(x)$ , let  $X_t^{(N)} = n_t/N$  be the number of occupied patches at census *t* relative to the threshold *N*.

Suppose that  $\mu$  is continuous with bounded first derivative.

If  $X_0^{(N)} \xrightarrow{2} x_0$  as  $N \to \infty$ , then  $X_t^{(N)} \xrightarrow{2} x_t$  for all  $t \ge 1$ , where  $(x_t)$  is determined by  $x_{t+1} = f(x_t)$   $(t \ge 0)$ , where  $f(x) = s(x + \mu(x))$ .

#### **Infinite-patch SPOM - Ricker dynamics**



Bifurcation diagram for the infinite-patch deterministic CE model with Ricker growth dynamics:  $x_{n+1} = 0.3 x_n (1 + e^{r(1-x_n)})$  (*r* ranges from 0 to 7.2).

#### **Infinite-patch SPOM - Ricker dynamics**



Simulation (open circles) of the infinite-patch CE model with Ricker growth dynamics, together with the corresponding limiting deterministic trajectories (solid circles). Here s = 0.3, N = 200 and (a) r = 0.84, (b) r = 1 (c) r = 4, (d) r = 5.