

Extinction in metapopulations with environmental stochasticity driven by catastrophes

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A metapopulation model

look left



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Some basic notation...

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- $\mathbf{X} = [X \ Y]^T$. The state of the metapopulation:
 - *X* is the number of suitable patches;
 - *Y* is the number of occupied patches.
- r, c, e, γ, p . Rates and other parameters.

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- Each occupied patch produces migrants at rate *c*, which may colonise empty, suitable patches

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- Each local population goes extinct at rate *e*

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on $S = \{(x, y) \mid x, y \in \mathbb{N}, 0 \le y \le x \le N\}.$

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$$(x, y) \rightarrow (x - (i + j), y - j)$$
 at rate
 $\gamma \begin{pmatrix} x - y \\ i \end{pmatrix} \begin{pmatrix} y \\ j \end{pmatrix} p^{i+j} (1 - p)^{x-i-j}.$

Finite State-space Processes

When N is finite, we can hope to evaluate measures of interest directly.

- Extinction times (a.k.a. first passage or 'hitting' times) are almost surely finite!
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If these are easy to calculate (or approximate), we can use them to analyse the system.



 $Q_C \boldsymbol{\tau} = -1$

(Generally, take the minimal, non-negative solution.)

Extinction Times



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$Q_C \mathbf{m} = -\lambda \mathbf{m}$

(λ is the eigenvalue with max. real part.)

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Quasi-stationary distributions



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Direct computation of hitting times, etc., becomes infeasible as N gets large: $\#S = \frac{1}{2}(N+1)(N+2).$ Direct computation of hitting times, etc., becomes infeasible as N gets large: $\#S = \frac{1}{2}(N+1)(N+2).$

• To make progress, we need good approximations: *e.g.* stochastic differential equations for the limit as $N \to \infty$?

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$$\mathbf{a}(\mathbf{U}) = \begin{bmatrix} \partial u / \partial t \\ \partial v / \partial t \end{bmatrix} = \begin{bmatrix} r(1-u) \\ cv(u-v) - ev \end{bmatrix},$$

with initial conditions $U(s, 0) = \lim_{N \to \infty} X(s, 0)/N$. (Kurtz, 1970)

Catastrophes in the Limit

Treating catastrophes as a separate component...

- The arrival rate of catastrophes is unaffected by scaling.
- As $N \to \infty$, if T_1 is a catastrophe time,

$$\frac{\mathbf{X}(s,T_1)}{N} \xrightarrow{P} (1-p)\mathbf{U}(s,T_1-).$$

A Stochastic Integral Equation

The limiting, scaled process:

$$d\mathbf{U}(s,t) = \mathbf{a}(\mathbf{U}(s,t))dt + \int_{\mathbf{M}} \mathbf{c}(\mathbf{U}(s,t),\mathbf{m}) \mathcal{P}[d\mathbf{m},dt;\gamma]$$

- $c(\mathbf{U}, d\mathbf{m})$ describes effect of catastrophes
- Poisson random measure \mathcal{P} describes arrival of catastrophes and their magnitudes, m.
- Generalised Itô fomula gives first passage times. (Gihman & Skorohod, 1972)

First passage times, $\tau_G(\mathbf{U}_0)$, into a closed set $S \setminus G$ (i.e. out of G), starting from \mathbf{U}_0 , are a twice continuously differentiable solution, $g(\mathbf{U})$, to

 $(Lg)(\mathbf{U}) = -1, \mathbf{U} \in G$ $g(\mathbf{U}) = 0, \mathbf{U} \notin G,$

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• In the present case, $(Lh)(\mathbf{U})$ is given by $(Lh)(\mathbf{U}) = \nabla h(\mathbf{U}) \cdot \mathbf{a}(\mathbf{U}) - \gamma h(\mathbf{U}) + \gamma h((1-p)\mathbf{U}).$

Slightly different conditions:

- $g(\mathbf{U})$ should be *continuous* along all trajectories $\mathbf{U}(s,t)$, and piecewise smooth along other smooth paths.
- $g(\mathbf{U})$ should be *bounded* for all \mathbf{U} .

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Solve in 'steps': G_n is the region from which at least n catastrophes are needed to leave G.

The solution has the form

$$e^{-\gamma t}g(\mathbf{U}(s,t)) = -\int_0^t \gamma e^{-\gamma t}g((1-p)\mathbf{U}(s,t))dt$$
$$-\gamma^{-1}\left[1-e^{-\gamma t}\right] + C_1(s),$$

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$$C_1(s) = \int_0^\infty \gamma e^{-\gamma r} g\big((1-p)\mathbf{U}(s,r)\big) dr + \gamma^{-1}.$$

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Hence (along trajectories that remain within *G*)

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- if G = G₁, g(U) = γ⁻¹ for all U on trajectories remaining in G;
- if $\mathbf{U}_{\infty} = \lim_{t \to \infty} \mathbf{U}(s, t)$ is in G, then $g(s, \infty -) = \gamma^{-1} + g((1 - p)\mathbf{U}_{\infty}).$

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- $g(\mathbf{U}(s,t)) = \gamma^{-1}$, for all trajectories not leaving G_1 in finite time,
- solutions for trajectories heading out of *G* using the deterministic hitting time and a truncated exponential law, and
- a system of DEs $[\partial u/\partial t, \partial v/\partial t, \partial g/\partial t]$ gives first passage times for trajectories starting on higher steps.



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The general case is a little more difficult.

• Recall that

$$g(\mathbf{U}(s,t)) = \frac{1}{\gamma} + e^{\gamma t} \int_{t}^{\infty} \gamma e^{-\gamma r} g((1-p)\mathbf{U}(s,r)) dr.$$

The general case is a little more difficult.

• Define a mapping $K: H \rightarrow H$,

$$K(f(s,t)) := \frac{1}{\gamma} + e^{\gamma t} \int_t^\infty \gamma e^{-\gamma r} f\left((1-p)\mathbf{U}(s,r)\right) dr,$$

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with *H* being the set of bounded functions $f: G \to \mathbb{R}_+$ under the condition

$$f(\mathbf{U}(s,t)) \ge \frac{1}{\gamma} + e^{\gamma t} \int_{t}^{\infty} \gamma e^{-\gamma r} f\left((1-p)\mathbf{U}(s,r)\right) dr.$$

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- $f \ge K(f), f \in H$, so we might hope that the iterative application of K would lead to a fixed point, but...
- $H \neq \emptyset$ is equivalent to the existence of a solution, h, to

 $(Lh)(\mathbf{U}) \le -1, \mathbf{U} \in G$ $h(\mathbf{U}) \ge 0, \mathbf{U} \notin G,$

 \equiv to a condition from Gihman & Skorohod for the existence of a solution $\tau_G \leq h$.

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 - (i) take $G' \supset G$ so that the first passage out of G' only depends on u;
 - (ii) find $h(u) = \tau_{G'}(u)$ (using H & T);
 - (iii) then h(u) satisfies the inequality condition for all v such that $(u, v) \in S$.



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