

A convenient form of complexity
in ecology: multiple timescales

Chris Cosner

gcc@math.miami.edu

Research partially supported
by NSF grants

DMS-9303708

DMS-1118623

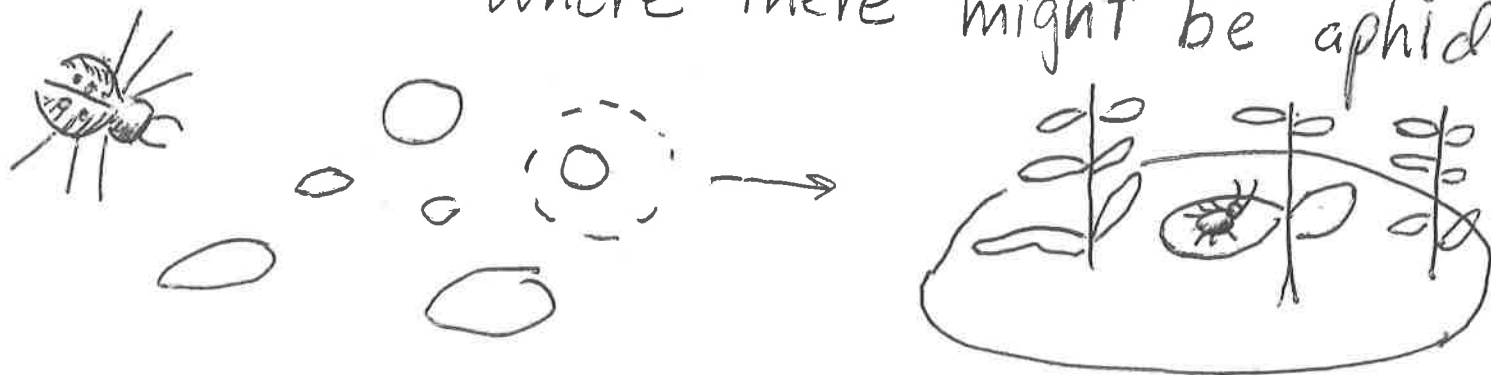
DMS-1514752

Research in collaboration with R.S. Cantrell
or R.S. Cantrell, M.A. Lewis, and Y. Lou

Case 1: Aphids and Ladybugs on Patches

- Aphids:
- Many generation per summer
 - Aphids mostly stay on one patch of plants, move short distances randomly

- Ladybugs:
- One generation per summer
 - Ladybugs move distances up to a kilometer, aggregate where there might be aphids



Motivating Question:

In a classical diffusive model

$$\frac{\partial v}{\partial t} = D \Delta v + R \left(1 - \frac{v}{K}\right) v \quad \text{on } \Omega$$

$$v = 0 \quad \text{on } \partial\Omega,$$

population density increases with patch size.

Observation: • As aphids recolonized a cleared area where patches of plants had regrown, this pattern held true in the first year.

• In later years it did not.

⇒ Could recolonization by Ladybugs cause that?

A model:

Patches Ω_n in \mathbb{R}^2 , $n=1, \dots, N$

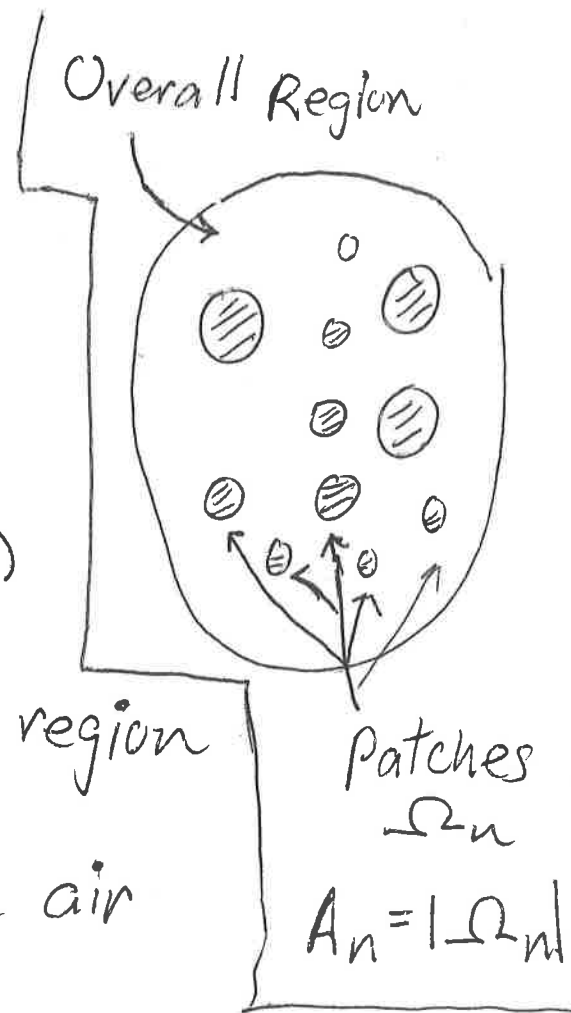
v_n = aphid density in patch n

(assume $v_n(x,0) \geq 0, \neq 0$, on Ω_n)

P = total Ladybug population in region

P_n = # of Ladybugs on Ω_n

$P_A = P - \sum_{n=1}^N P_n$ = # of Ladybugs in air



$$\frac{\partial v_n}{\partial t} = D \Delta v_n + R \left(1 - \frac{v_n}{K}\right) v_n - c \frac{P_n v_n}{A_n} \text{ in } \Omega_n$$

$$v_n = 0 \text{ on } \partial \Omega_n$$

$$\frac{dP_n}{dt} = I_n P_A - E_n P_n = I_n \left(P - \sum_{m=1}^N P_m\right) - E_n P_n$$

immigration ← emigration

days →

minutes →

Fast timescale τ : Assume $P_n \rightarrow$ equilibrium P_n^*

$$0 = I_n \left(P - \sum_{m=1}^N P_m \right) - E_n P_n \quad n=1, \dots, N$$

(sum over n , solve for $\sum_{n=1}^N P_n$, substitute back.)

$$P_n^* = \frac{(I_n / E_n) P}{1 + \sum_{m=1}^N (I_m / E_m)}$$

(Note: we could allow $P = P(t)$, but did not.)

Options for I_n and E_n include $I_n = i A_n^p$, $E_n = e$

where $A_n = |\Omega_n|$, $V_n = \int_{\Omega_n} v_n(x, t) dx$, $(i \text{ and } e \text{ constants, also } p, q, r, \text{ all } > 0)$
 $\frac{A_n^p \bar{v}_n(t)^r}{A_n^p \bar{v}_n(t)^r}$

(This captures predator aggregation (concentration) on patches with higher prey densities)

Slow timescale

$$\frac{\partial v_n}{\partial t} = D \Delta v_n + R \left(1 - \frac{v_n}{K}\right) v_n - c \frac{P_n}{A_n} v_n \text{ on } \Omega_n,$$

← predator density

$$v_n = 0 \text{ on } \partial \Omega_n, \quad n = 1, \dots, N$$

with

$$A_n = |\Omega_n|$$

$$V_n = \int_{\Omega_n} v_n dx$$

$$P_n = P_n^* = \frac{(I_n/E_n)P}{1 + \sum_{m=1}^N (I_m/E_m)} = \frac{(i/e) A_n^{p+q} V_n^r P}{1 + (i/e) \sum_{m=1}^N A_m^{p+q} V_m^r}$$

so

$$\frac{\partial v_n}{\partial t} = D \Delta v_n + R \left(1 - \frac{v_n}{K}\right) v_n - c \frac{(i/e) A_n^{p+q} (P/A_n) V_n^r v_n}{1 + (i/e) \sum_{m=1}^N A_m^{p+q} V_m^r}$$

(cooperative system: if $v_m \uparrow$ then $V_m \uparrow$

so if $v_m \uparrow$ for $m \neq n$ then $\partial v_n / \partial t \uparrow$.) (V_m constant in x)

Application - Effects of patch size on aphid population

At equilibrium $v_n = v_n^*$ satisfying

$$0 = D \Delta v_n^* + R \left(1 - \frac{v_n^*}{K}\right) v_n^* - C v_n^* \text{ in } \Omega_n, v_n^* = 0 \text{ on } \partial \Omega_n$$

C = constant, depending on V_1, \dots, V_N and A_1, \dots, A_N

but $0 = D \Delta \theta + (a - b \theta) \theta$ in Ω , a, b constant

$$\theta = 0 \text{ on } \partial \Omega$$

is well understood - so analysis is still sometimes possible.

If we assume large scale aggregation of Ladybugs - so $P = P_0 \left(\sum_{n=1}^N A_n \right) = P_0$ (total patch area)

then for some parameter ranges we get

(Patch 1 largest)



Case 2: Evolution of dispersal

Time / space scales: (medium-large mammals)

- i) Random walk, step by step
(seconds - minutes / meters)
- 2) Population level dispersal (diffusion, advection)
(hours - days / kilometers)
- 3) Population interactions and dynamics
(weeks - years / species ranges
(up to hundreds of kilometers))
- 4) Evolution of dispersal
(thousands of years / species ranges)

Random walks \rightarrow Population level dispersal



$x - \Delta x$ x $x + \Delta x$

At each time step Δt ,
move one space step Δx

Cases:

(0) $P(x|x-\Delta x) = P(x+\Delta x, x) = p_0(x) \rightarrow \frac{\partial p}{\partial t} = \frac{\partial}{\partial x} (D_0(x) \frac{\partial p}{\partial x})$

(1) $\begin{cases} P(x|x-\Delta x) = p_1(x-\Delta x) \text{ (departure point)} \\ P(x+\Delta x, x) = p_1(x) \end{cases} \rightarrow \frac{\partial p}{\partial t} = \frac{\partial^2}{\partial x^2} (D_1(x) p)$

(2) $\begin{cases} P(x|x-\Delta x) = p_2(x) \text{ (arrival point)} \\ P(x+\Delta x|x) = p_2(x+\Delta x) \end{cases} \rightarrow \frac{\partial p}{\partial t} = \frac{\partial}{\partial x} (D_2^2(x) \frac{\partial}{\partial x} (\frac{p}{D_2(x)}))$

Diffusion limit
 ($k=0, 1, 2$)

$\frac{(\Delta x)^2}{\Delta t} = C, D_k(x) = C p_k(x)$

Let $\Delta x, \Delta t \rightarrow 0$

$p(x, t) =$ continuous PDF
 for limit

PDE form for p :

Random walk \rightarrow Population dispersal

vs.

Population dispersal \rightarrow equilibrium distribution
(closed environment)
(no-flux b.c.)

Cases again:

$t \rightarrow \infty$

$$(0) \quad \frac{\partial p}{\partial t} = \frac{\partial}{\partial x} \left(D_0(x) \frac{\partial p}{\partial x} \right)$$

$$p(x, t) \rightarrow C$$

$$(1) \quad \frac{\partial p}{\partial t} = \frac{\partial^2}{\partial x^2} (D_1(x)p)$$

$$p(x, t) \rightarrow C / D_1(x)$$

$$(2) \quad \frac{\partial p}{\partial t} = \frac{\partial}{\partial x} \left(D_2^2(x) \frac{\partial}{\partial x} \left(\frac{p}{D_2(x)} \right) \right) \quad p(x, t) \rightarrow C D_2(x)$$

The form of random walk (scale $\Delta x, \Delta t$)
determines the possible patterns arising from dispersal
(scale x, t)

Population level models (movement)
(timescale t)

$u(x, t, T) =$ population density

$t \sim$ movement
 $T \sim$ population dynamics

"Classical" models: $t = T$

$$\frac{\partial u}{\partial t} = \nabla \cdot [\mu(x) \nabla u - u \vec{P}(x)] + f(x, u)$$

but timescales suggest we look at

$$\frac{\partial u}{\partial t} = \nabla \cdot [\mu(x) \nabla u - u \vec{P}(x)], \text{ let } t \rightarrow \infty; \text{ then}$$

we expect $u \rightarrow \phi(x) U(T)$

$$\nabla \cdot [\mu \nabla \phi - \phi \vec{P}] = 0 \quad \star$$

Then look for $U(t)$

$$\frac{\partial U}{\partial T} = F(U) = ??$$

\star The assumptions on movement in the $(\Delta x, \Delta t)$ scale constrain possible forms of ϕ

Population level models (dynamics timescale T)

logistic case: $L =$ dispersal operator

$\phi =$ principal eigenfunction of L

$$u(x, t, T) \rightarrow \phi(x) \bar{U}(T) \quad f(x, u) = (m(x) - u)u \quad (\text{so } L\phi = 0)$$

$$\phi(x) \frac{d\bar{U}}{dt} = (L\phi)\bar{U} + [m(x) - \phi(x)\bar{U}(T)] \phi(x) \bar{U}(T)$$

$$= 0$$

Later normalize
($\int_{\Omega} \phi dx = 1$)

Integrate x , \div by $\int_{\Omega} \phi(x) dx$

$$\frac{d\bar{U}}{dt} = \left(\frac{\int_{\Omega} m(x) \phi(x) dx}{\int_{\Omega} \phi(x) dx} \right) \bar{U}(T) - \left(\frac{\int_{\Omega} \phi(x)^2 dx}{\int_{\Omega} \phi(x) dx} \right) \bar{U}(T)^2$$

$$\text{Let } \bar{u} = \frac{\bar{U}(T)}{|\Omega|}, \quad \bar{\phi} = \frac{\int_{\Omega} \phi(x) dx}{|\Omega|}, \quad \bar{m} = \frac{\int_{\Omega} m(x) dx}{|\Omega|}$$

$$\text{Then } \frac{\int_{\Omega} m\phi dx}{\int_{\Omega} \phi dx} = \frac{\bar{m}\bar{\phi}}{\bar{\phi}} = \bar{m} + \frac{\text{cov}(m, \phi)}{\bar{\phi}} \quad \left(\begin{array}{l} \text{"scale transition"} \\ \text{Chesson} \end{array} \right)$$

Population Interactions (scale T)

Suppose that u and v are competing populations that on a single time scale would be described by

$$\frac{\partial u}{\partial t} = L_1 u + [a(x) - b(x)u - c(x)v]u$$

$$\frac{\partial v}{\partial t} = L_2 v + [d(x) - e(x)u - f(x)v]v.$$

Using the rescaled system where $u = \bar{U}(\tau)\phi_1(x)$, $v = \bar{V}(\tau)\phi_2(x)$

with ϕ_i the normalized eigenfunction of $L_i = 0$, $i=1,2$

and \bar{U} = total population for u , \bar{V} = total population for v

gives

$$\frac{d\bar{U}}{d\tau} = \left[\left(\int_{\Omega} a\phi_1 \right) - \left(\int_{\Omega} b\phi_1^2 \right) \bar{U} - \left(\int_C c\phi_1\phi_2 \right) \bar{V} \right] \bar{U}$$

$$\frac{d\bar{V}}{d\tau} = \left[\left(\int_{\Omega} d\phi_2 \right) - \left(\int_{\Omega} e\phi_1\phi_2 \right) \bar{U} - \left(\int_{\Omega} f\phi_2^2 \right) \bar{V} \right] \bar{V}.$$

This allows comparison of dispersal strategies if

the species have similar ecology, so that $a=d$, $b=c=e=f$

Evolution (long timescale)

Approach (motivated by adaptive dynamics)

- Evolution occurs as an initially small population of mutants with better adaptation invades and replaces a resident population
- This occurs rarely relative to the timescale T of ecological interactions
- The traits that will evolve are those that allow a population using them to resist invasion by populations with other traits (Evolutionarily Steady Strategies or ESS)

Evolution (pairwise invasion analysis)

(Assume resident and invader differ only in their dispersal pattern)

$$\text{For } \frac{dU}{dT} = \left[\left(\int_{\Omega} \phi_1^m \right) - \left(\int_{\Omega} \phi_1^z \right) U - \left(\int_{\Omega} \phi_1 \phi_2 \right) V \right] U \quad \left(\int_{\Omega} \phi_j dx = 1 \right)$$

$$\frac{dV}{dT} = \left[\left(\int_{\Omega} \phi_2^m \right) - \left(\int_{\Omega} \phi_1 \phi_2 \right) U - \left(\int_{\Omega} \phi_2^z \right) V \right] V$$

the semitrivial equilibrium $(U^*, 0)$ has $U^* = \frac{\int_{\Omega} \phi_1^m}{\int_{\Omega} \phi_1^z}$
(as $T \rightarrow \infty$)

so the competing population V can invade $(U^*, 0)$ if

$$\sigma = \left(\int_{\Omega} \phi_2^m \right) - \left[\left(\int_{\Omega} \phi_1 \phi_2 \right) \left(\int_{\Omega} \phi_1^m \right) / \int_{\Omega} \phi_1^z \right] > 0.$$

Observation: This approach cannot distinguish between strategies that produce the same distributions (ϕ_i 's), for example L and DL for constant D. It can distinguish diffusion from diffusion + advection

Evolution (continued)

An evolutionarily steady class of strategies

Suppose that $m(x) > 0$, and $\phi_1(x) = \frac{m(x)}{\bar{m}}$ (motivated by the ideal free distribution)

(This will be true for $L_1 = \nabla \cdot \mu(x) \nabla(u/m)$, which includes

$L_1 = \nabla \cdot [\nabla u - u \frac{\nabla m}{m}]$ for $\mu = m$ or $L_1 = \Delta(u/m)$ for $\mu = 1$)

The criterion for stability of $(u^*, 0)$ gives

$$\int_{\Omega} \phi_2 m - \frac{\left(\int_{\Omega} \frac{m}{\bar{m}} \phi_2 \right) \left(\int_{\Omega} \frac{m^2}{\bar{m}} \right)}{\int_{\Omega} \frac{m^2}{\bar{m}^2}} = 0 \quad (\text{neutral stability})$$

The criterion for stability of $(0, v^*)$ is given by

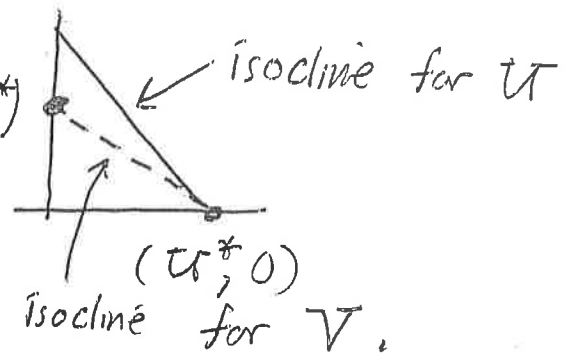
$$\sigma = \frac{1}{\bar{m}} \left[\int_{\Omega} m^2 - \frac{\left(\int_{\Omega} m \phi_2 \right)^2}{\int_{\Omega} \phi_2^2} \right]. \quad \text{By Hölder's inequality,}$$

$\sigma > 0$ if $m^2 \neq k m \phi_2$,
that is, $m \neq k \phi_2$, for
some constant k . (unstable!)

Evolution (continued)

It turns out that if $\phi_1 = \frac{m(x)}{\bar{m}}$, solving for a possible positive equilibrium in the fast dispersal system gives $V = 0$.

Phase portrait :



It follows that U excludes V .

Thus, since ϕ_2 is arbitrary, strategies that yield $\phi_1 = \frac{m(x)}{\bar{m}}$ are evolutionarily steady relative to those that do not. *

(* Related results for patch models, nonlocal models, reaction-diffusion-advection exist but take much more math.)

References

Case 1: R.S. Cantrell and C., SIAM Review 38 (1996), 256-286.

Case 2: R.S. Cantrell, C., M. Lewis, Y. Lou, JMB, epubs
DOI: 10.1007/s00285-18-1302-02