

# Stochastic Population Dynamics: Persistence, Extinction, and Quasi-stationarity

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# Focus and Structure of Tutorial

Mathematically, focus on **discrete-time** stochastic models.

Go to **MS 34, Monday 8:30–10:00am**, for **continuous-time models**

Dang Nguyen Hai (8:30am) on Stochastic Differential Equations

Alex Hening (8:55am) and Edouard Strickler (9:20am) on  
Piecewise Deterministic Markov Processes

Mads Hansen (9:45am) on quasi-stationarity for continuous-time  
Markov chains

**Today: Three parts** ( $\sim 35$  minutes each  $+5$  minute breaks)

- I. Environmental Stochasticity for single species
- II. Environmental stochasticity for interacting species
- III. Demographic stochasticity and quasi-stationarity



Introduction  
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ES: Single species  
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ES: Communities  
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Demographic stochasticity  
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References





# COMMENTARII ACADEMIAE SCIENTIARUM IMPERIALIS PETROPOLITANAE.

TOMVS V.

AD ANNOS *clb*l*ccc* xxx. et *clb*l*ccc* xxxi.



PETROPOLI,  
TYPIS ACADEMIAE.

## SPECIMEN THEORIAE NOVAE DE MENSURA SORTIS. AVCTORE *Daniele Bernoulli.*

1.

... *in tempore quo Casus ... considerari*

*"these terms should then be multiplied together. Then of this product a root must be extracted the degree of which is given by the number of all possible cases"* Stearns [2000]

If  $\mathbb{P}[f(0) = f_i] = \frac{1}{k}$ , then

$$\exp(r) = \sqrt[k]{f_1 f_2 \dots f_k}$$

Geometric mean



*ON POPULATION GROWTH IN A RANDOMLY  
VARYING ENVIRONMENT*

BY R. C. LEWONTIN AND D. COHEN\*

DEPARTMENT OF BIOLOGY, UNIVERSITY OF CHICAGO; AND DEPARTMENT OF BOTANY,  
HEBREW UNIVERSITY, JERUSALEM

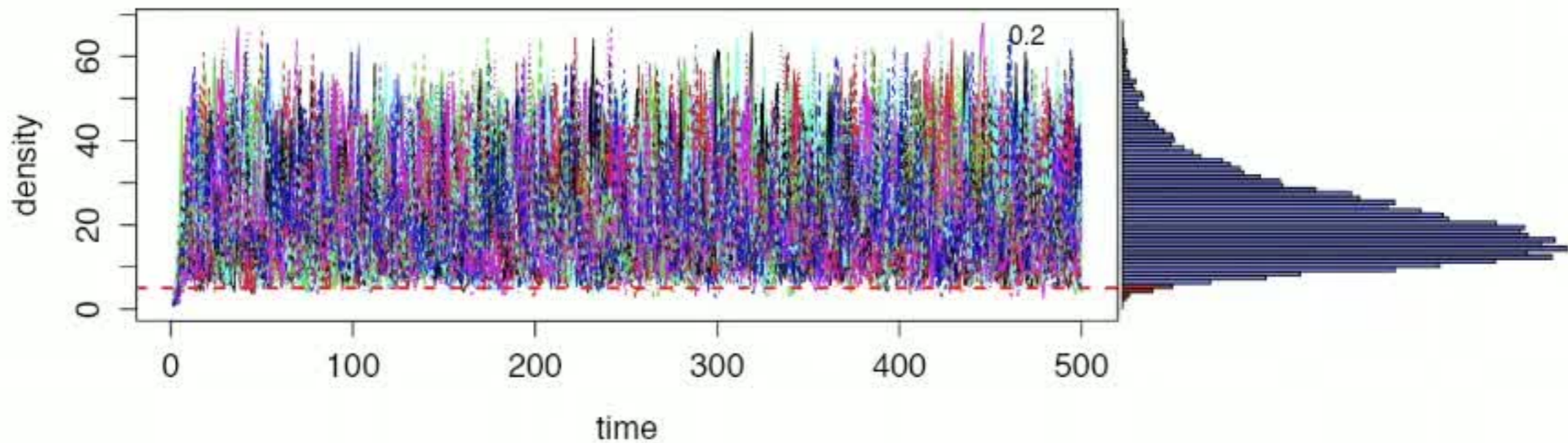
*Communicated February 10, 1969*

*Abstract.*—If a population is growing in a randomly varying environment, such that the finite rate of increase per generation is a random variable with no serial autocorrelation, the logarithm of population size at any time  $t$  is normally distributed. Even though the expectation of population size may grow infinitely large with time, the probability of extinction may approach unity, owing to the difference between the geometric and arithmetic mean growth rates.



“Anyone who believes exponential growth can go on forever in a finite world is either a madman or an economist” – Kenneth Boulding, economist and President Kennedy’s Environmental Advisor

$$X(t+1) = f(X(t), \xi(t))X(t) \quad \xi(1), \xi(2), \dots \text{i.i.d.} \quad (\star)$$



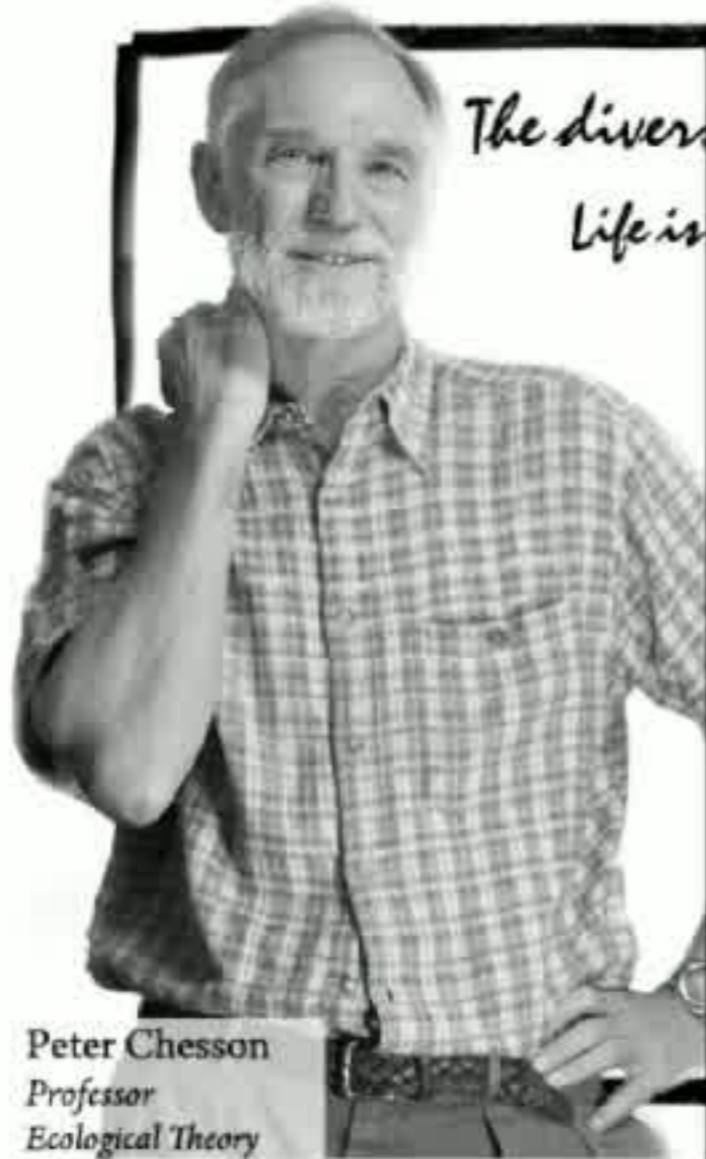
**Stochastic persistence in probability** [Chesson, 1982, Ellner, 1984]: For all  $\varepsilon > 0$ , there is a  $\delta > 0$  such that

$$\limsup_{t \rightarrow \infty} \mathbb{P}[X(t) \leq \delta | X(0) = x] \leq \varepsilon$$

whenever  $X_0 = x > 0$ .

*arbitrarily unlikely to be below arbitrarily small densities far into the future*





“This criterion requires that the probability of observing a population below any given density, should converge to zero with density, uniformly in time. Consequently it places restrictions on the expected frequency of fluctuations to low population levels. Given that fluctuations in the environment will continually perturb population densities, it is to be expected that any nominated population density, no matter how small, will eventually be seen. Indeed this is the usual case in stochastic population models and is not an unreasonable postulate about the real world. Thus a reasonable persistence criterion cannot hope to do better than place restrictions on the frequencies with which such events occur.”

# Density-dependent models

$$X(t+1) = f(X(t), \xi(t))X(t) \quad \xi(1), \xi(2), \dots \text{i.i.d.} \quad (\star)$$

When do we get persistence? If  $X(0) \approx 0$  but positive, then

$$X(t) \approx \prod_{s=0}^{t-1} f(0, \xi(s))X(0) \Rightarrow r = \mathbb{E}[\log f(0, \xi(1))]$$

**Theorem** [Ellner, 1984, Gyllenberg et al., 1994, Benaïm and Schreiber, 2019] If  $r > 0$ , then  $(\star)$  is stochastically persistent almost surely and in probability. If  $r < 0$ , then for all  $\varepsilon > 0$  there is a  $\delta > 0$  s.t.

$$\mathbb{P}\left[\lim_{t \rightarrow \infty} \frac{1}{t} \log X(t) = r \mid X(0) = x\right] \geq 1 - \varepsilon \text{ for } x \in (0, \delta)$$

If  $x = 0$  is **accessible** (see Benaïm and Schreiber [2019]) &  $r < 0$ , then  $\lim_{t \rightarrow \infty} \frac{1}{t} \log X(t) = -r$  with probability one for  $X(0) > 0$ .

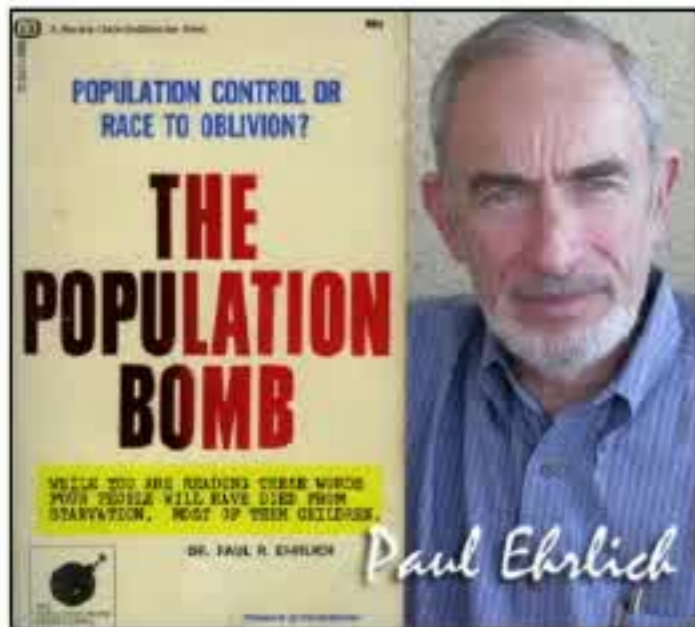
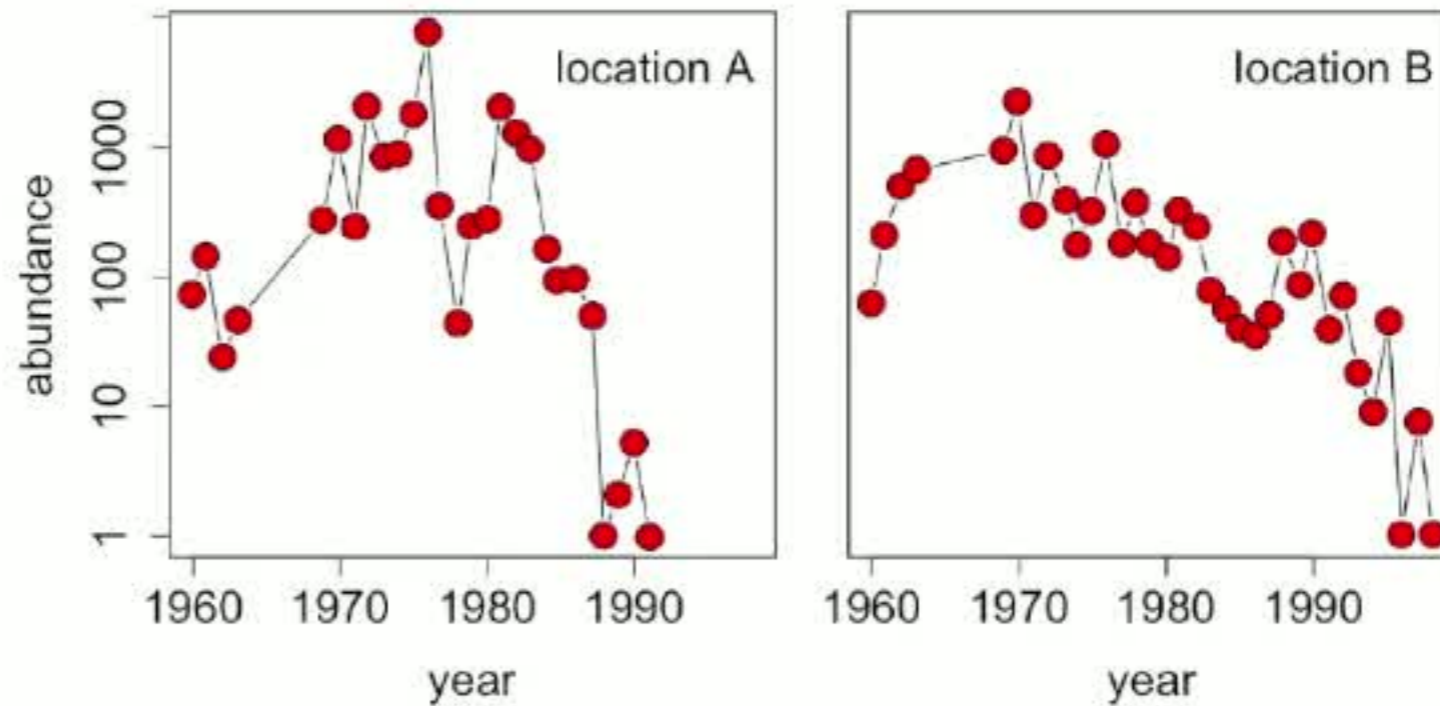


# Climate induced extinction?



Bay checkerspot

2 checkerspot populations went extinct in 1990s



Paul Ehrlich



*Plantago erecta*



*Castilleja exserta*



*Castilleja densiflora*



# Climate induced extinction?



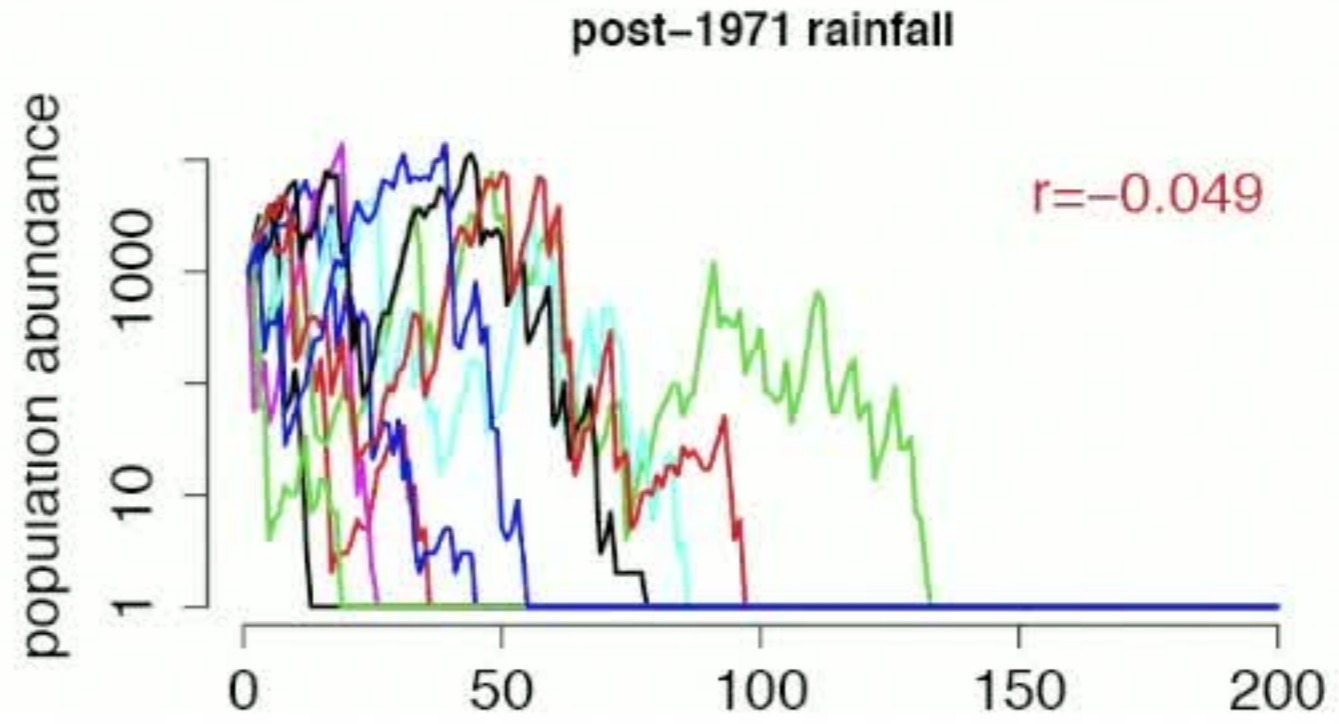
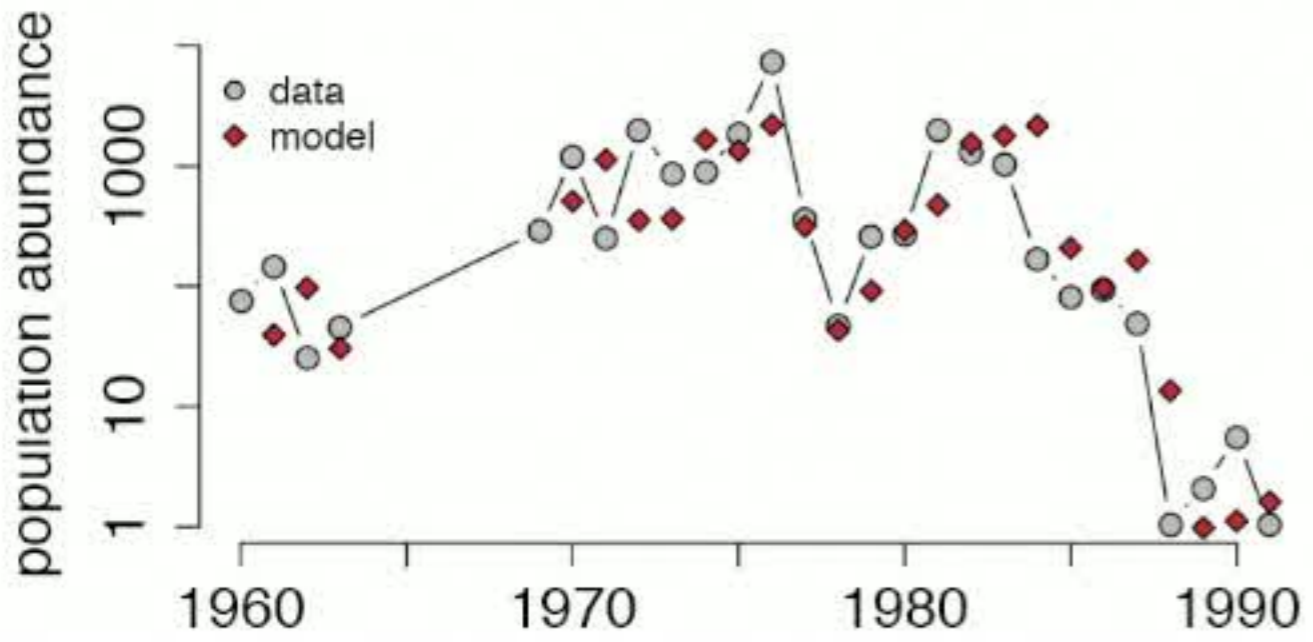
Bay checkerspot

2 checkerspot populations went extinct in 1990s

A **simplified** version of McLaughlin et al. [2002]:

$$X(t + 1) = X(t) \exp(a_0 + a_1 X(t) + a_2 \xi(t)^{\theta_2})$$

w/  $\xi(t)$  = precipitation. Here  $r = \mathbb{E}[a_0 + a_1 \xi(t)^{\theta_2}]$





Two approaches to showing uniqueness of  $\mu_+$ :

**Monotonicity** [Ellner, 1984, Chueshov, 2002]:  $x \mapsto xf(x, \xi)$  increasing

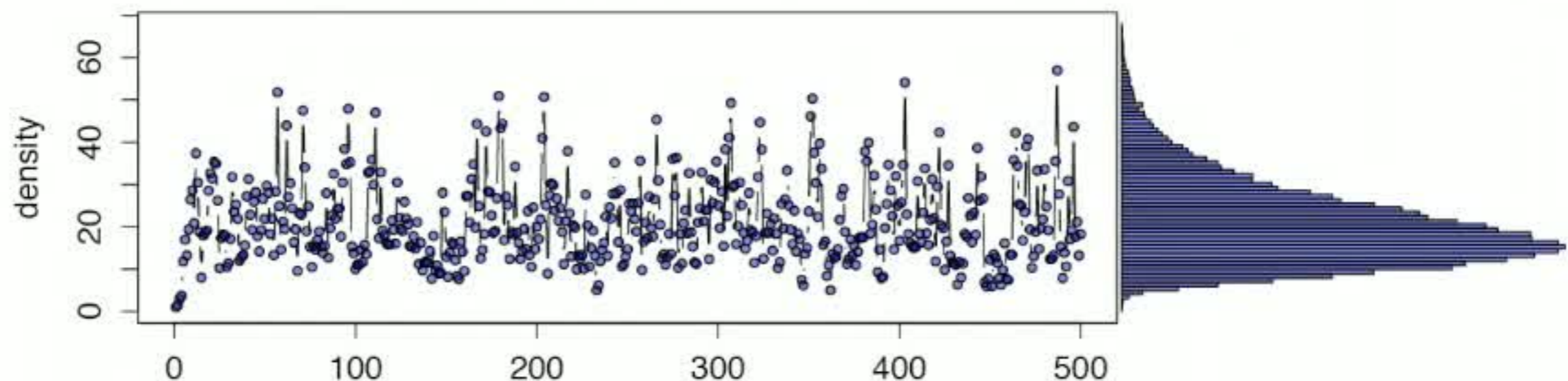
**Irreducibility** [Meyn and Tweedie, 2009, Schreiber et al., 2011]:

$\mathbb{P}[X(t) \in A | X(0) = x] \geq \nu(A)$  for all  $x > 0$ , some  $t \geq 1$ , and some probability measure  $\nu$  on  $(0, \infty)$

Both of approaches extend to higher dimensions.

For PDMPs, there exist Lie bracket conditions on the vector fields

See talks by Alex Hening (8:55am) and Eduoard Stricker (9:20am)  
in Monday session MS34





Introduction

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ES: Single species

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ES: Communities

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Demographic stochasticity

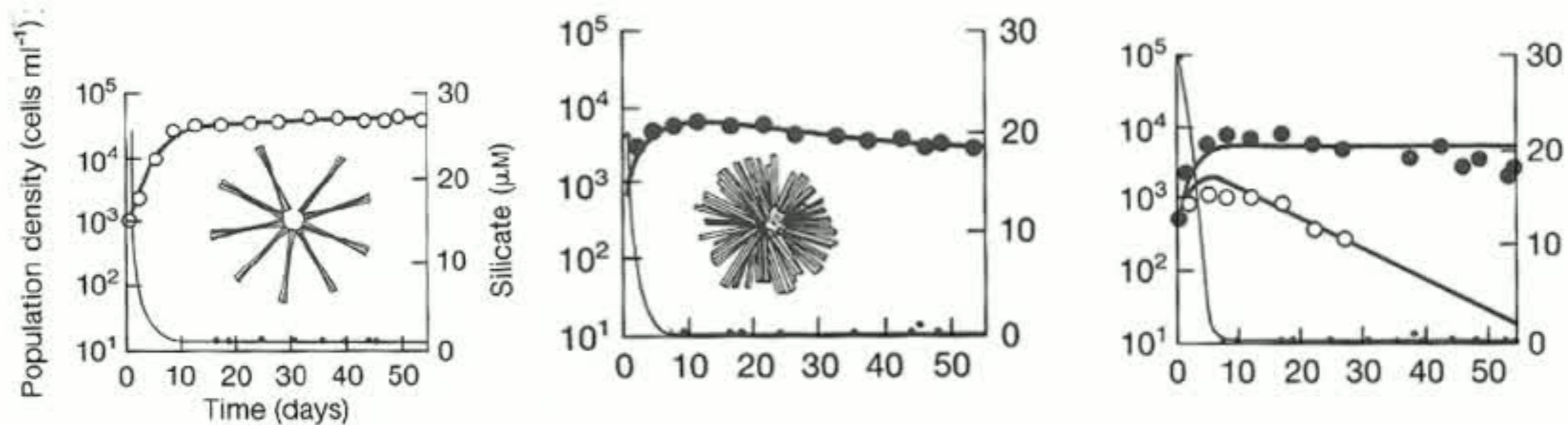
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References





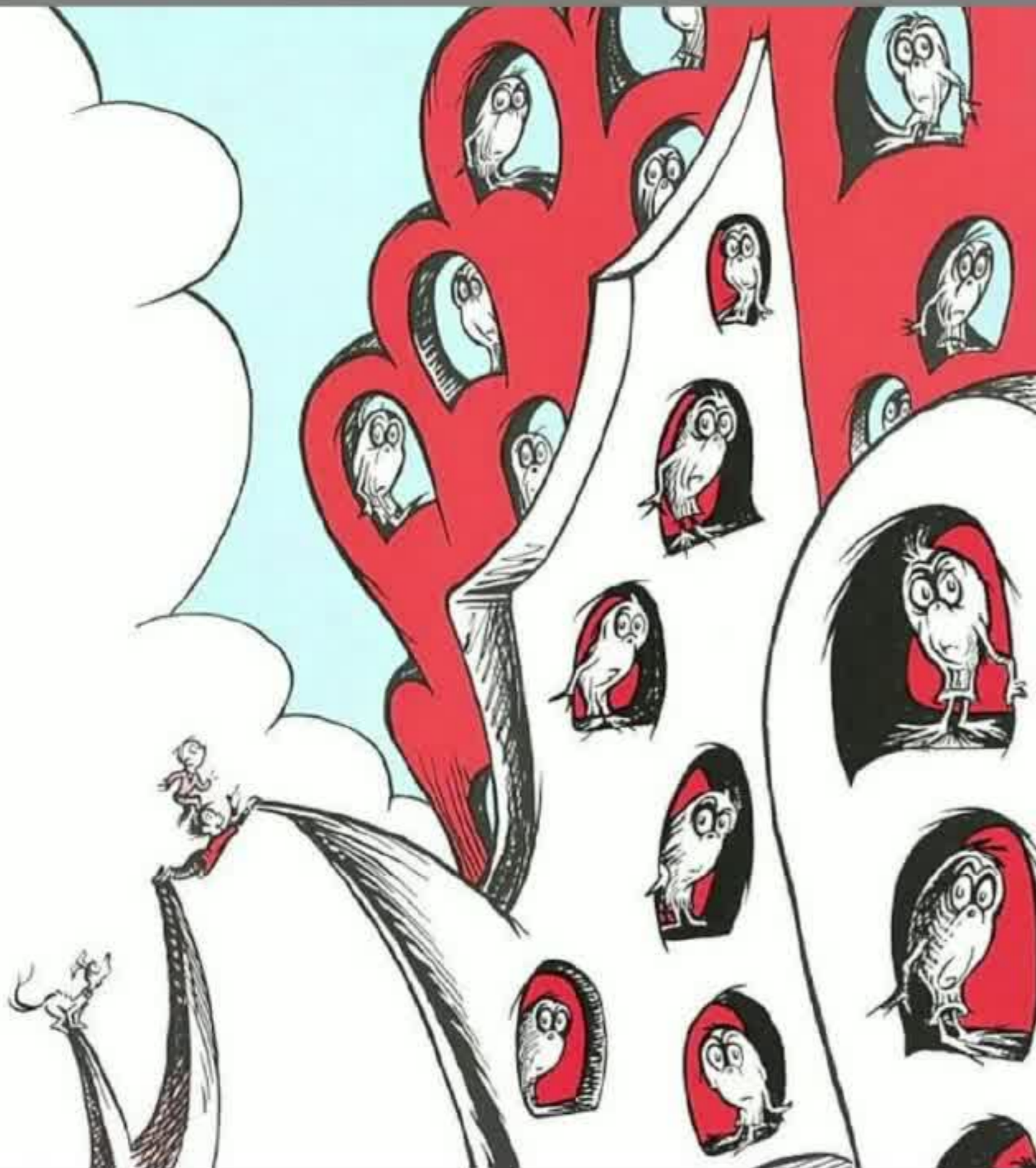
The  $R^*$  rule [Volterra, 1926, Tilman, 1977]: The competitor that suppresses a single limiting resource to the lowest equilibrium value excludes all other competitors.



Competitive exclusion principle [McGehee and Armstrong, 1977]: At most  $k$  species can coexist at a stable equilibrium on  $k$  limiting resources.

Paradox of the plankton [Hutchinson, 1961]: “The diversity ... was explicable primarily by a permanent failure to achieve equilibrium as the relevant external factors changed.”





“And NUH is the letter I use to spell Nutches, Who live in small caves, known as Niches, for hutches. These Nutches have troubles, the biggest of which is the fact there are many more Nutches than Niches. Each Nutch in a Nich knows that some other Nutch Would like to move into his Nich very much. So each Nutch in a Nich has to watch that small Nich or Nutches who haven't got Niches will snitch.”

Dr. Seuss

*On Beyond the Zebra*



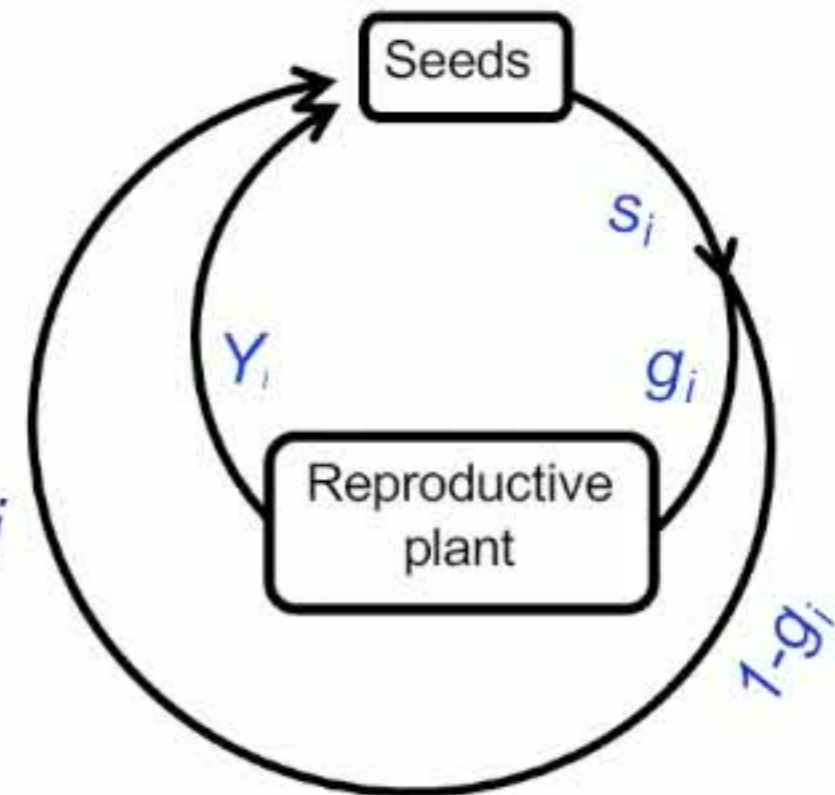
# Annual plant model

$X_i(t)$  seed density of species  $i$

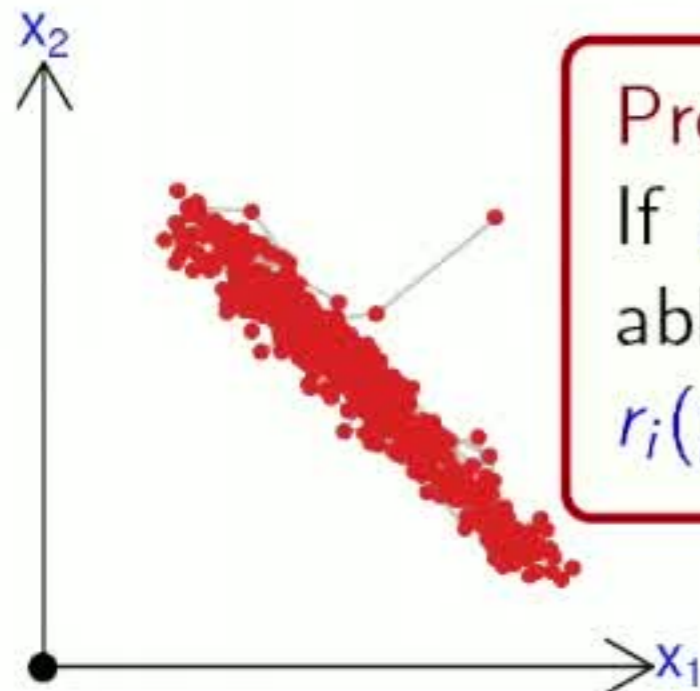
$s_1 = s_2$  seed survival

$g_i(t)$  germination probability of species  $i$

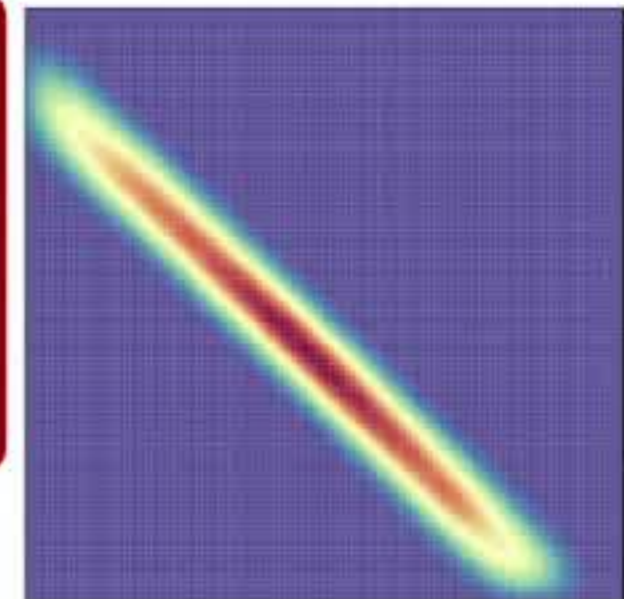
$Y_1 = Y_2$  yield of a plant



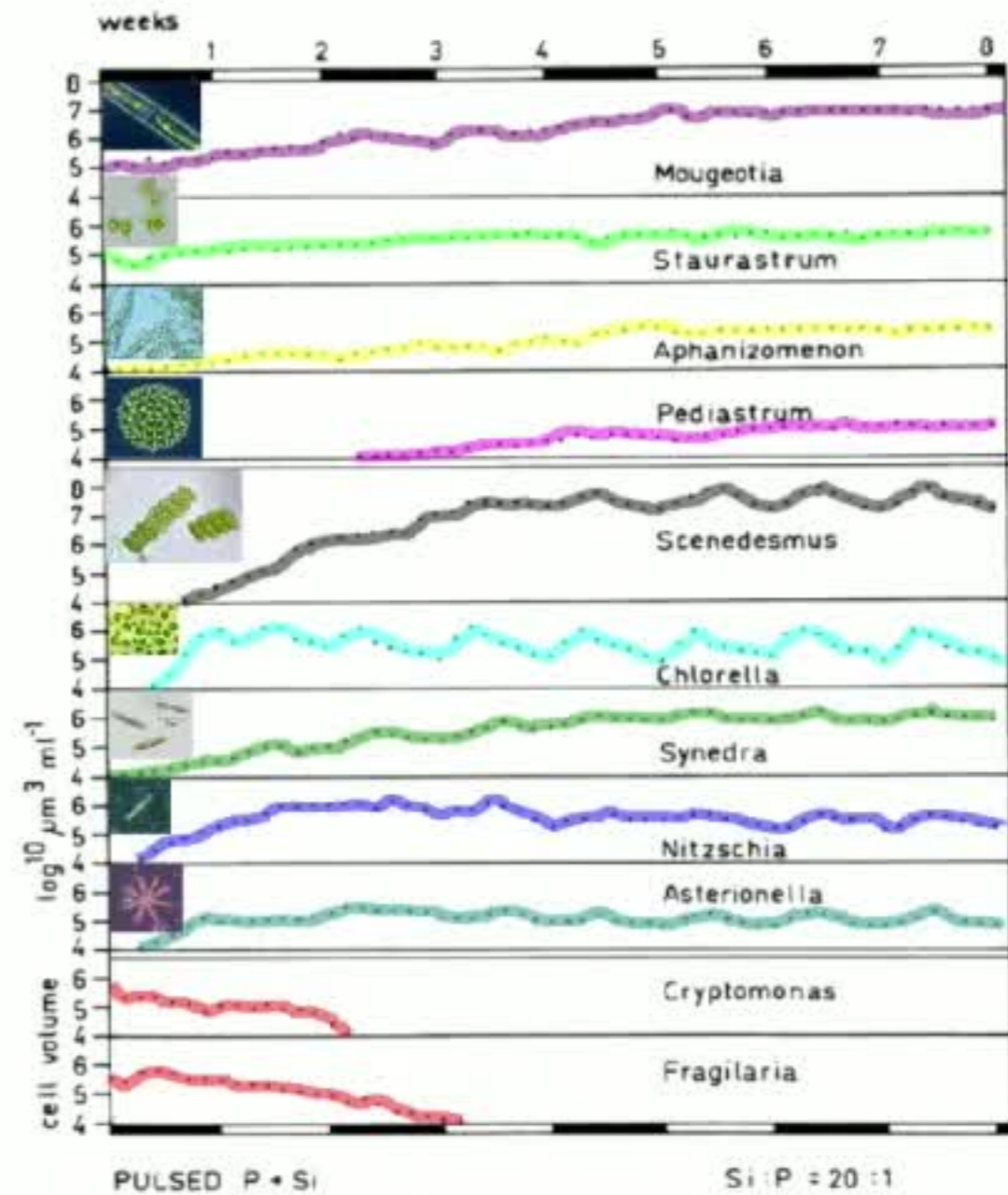
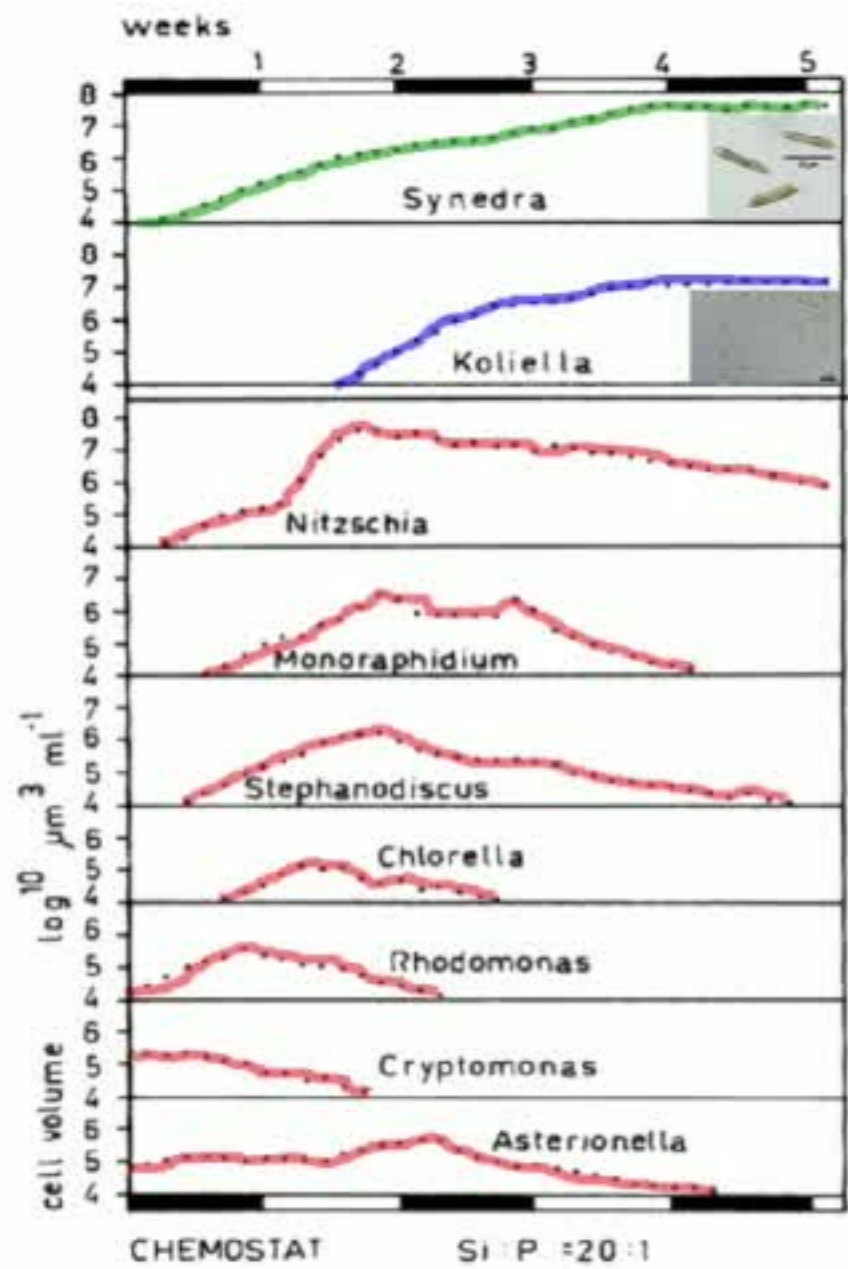
$$X_i(t+1) = X_i(t) \left( \frac{Y_i g_i(t) s_i}{1 + \sum_{j=1}^2 g_j(t) s_j X_j(t)} + (1 - g_i(t)) s_i \right)$$



**Proposition** [Chesson, 1988]:  
If  $g_1(t), g_2(t)$  are exchangeable and  $\text{Var}[g_1(t)] > 0$ , then  $r_i(\mu_j) > 0$  for  $j \neq i$ .







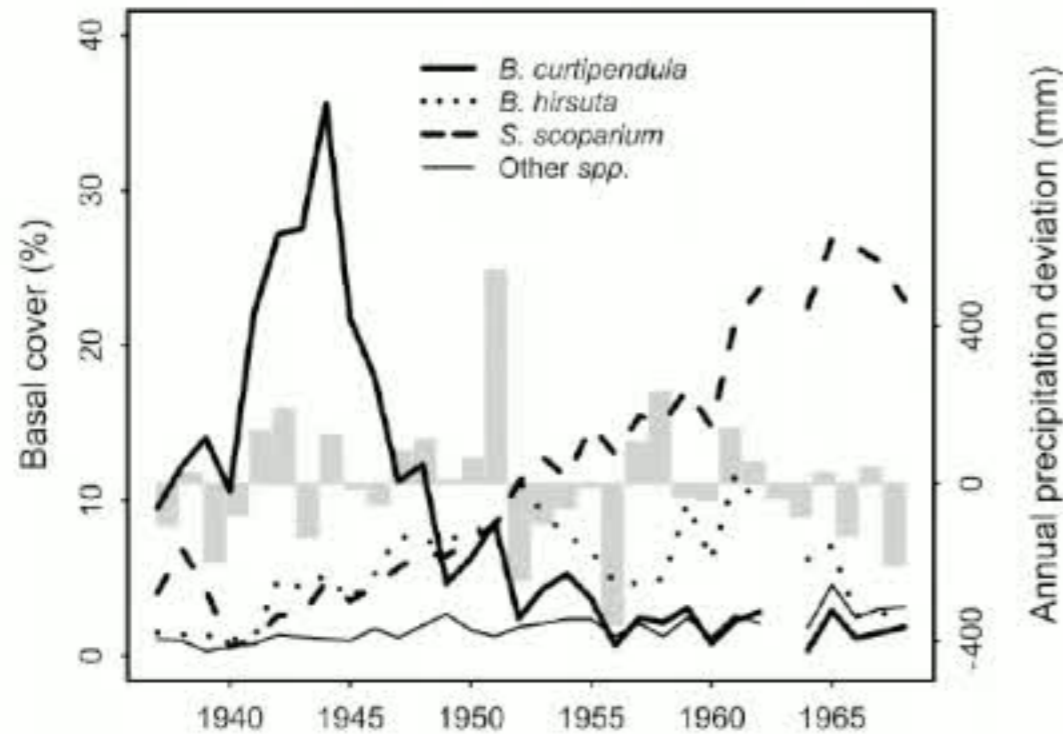
Constant inflow of Si and P

Fluctuating inflow of Si and P

Sommer [1984]



# Storage in Kansas prairies [Adler et al., 2006]



Fit a stochastic model using a hierarchical Bayesian approach

Computed  $r_i$  for stochastic and deterministic environments

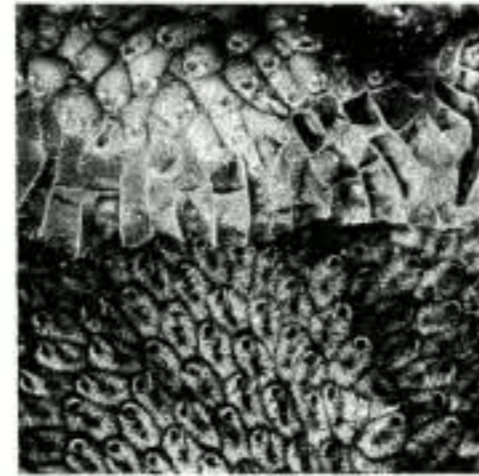


# The childhood game



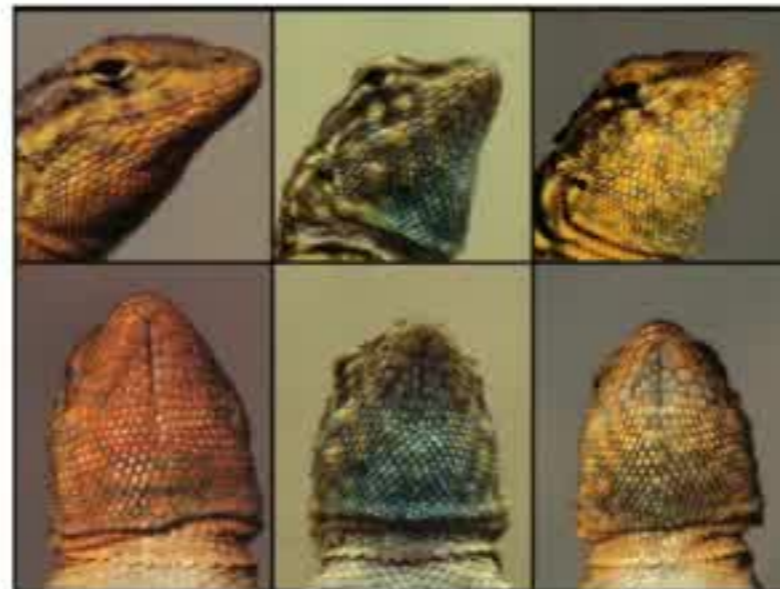
# Cryptic coral reef communities

Buss & Jackson Am. Nat. 1979



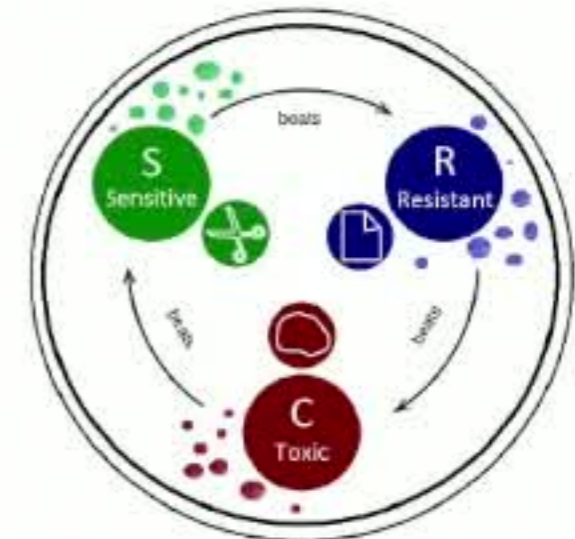
# Side-blotched lizards

Sinervo & Lively Nature 1996



# Chemical warfare in *E. coli*

Kerr et al. Nature 2003





Lest **biologists** suspect your **model** untrue, Keep probability in view.

*Populations are far from the continuous matter, or flows, or fields of classical mathematical physics. They are essentially discrete and built up by individuals, who may show great variation in behavior. Jagers [2010]*

*Demographic stochasticity describes the random fluctuations in the population size that occur because the birth and death of each individual is a discrete and probabilistic event. –Brett Melbourne (2012)*

Demographic stochasticity is typically modeled with Markov chains  $X(t)$  on a countable state space  $\mathcal{S}$

Dynamics determined by the transition matrix

$$P_{xy} = \mathbb{P}[X(t+1) = y | X(t) = x]$$



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ES: Communities

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Demographic stochasticity

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References





# Branching processes

$X(t) \in \{0, 1, 2, 3, \dots\} = \mathbf{Z}_+$  # of individuals in population

offspring distribution:  $p_k$  probability individual has  $k$  offspring

$$X(t+1) = Y_1(t+1) + Y_2(t+1) + \dots + Y_{X(t)}(t+1) \quad (\star)$$

where  $Y_i(t+1)$  are independent with  $\mathbb{P}[Y_i(t) = k] = p_k$

**Limit Theorem** Athreya and Ney [2004] Assume mean number of offspring  $R_0 = \sum_{k \geq 0} kp_k < +\infty$  and  $p_0 > 0$ . If  $R_0 \leq 1$ , then

$$\mathbb{P}[X(t) = 0 \text{ for some } t \geq 1 | X(0) = 1] = 1$$

If  $R_0 > 1$ , then

$$s^* = \mathbb{P}[X(t) = 0 \text{ for some } t \geq 1 | X(0) = 1] < 1$$

and is the minimal solution to  $s = \sum_k p_k s^k$  for  $s \in [0, 1]$ .





“Anyone who believes exponential growth can go on forever in a finite world is either a madman or an economist” – Kenneth Boulding, economist and President Kennedy's Environmental Advisor



Any population allowing individual variation in reproduction, ultimately dies out—unless it grows beyond all limits, an impossibility in a bounded world.

Jagers [2010]



# Example: Host-Parasitoid Dynamics



# Example: Host-Parasitoid Dynamics



**Parasitoid:** An organism that, during its development, lives in or on the body of a single host individual, eventually killing that individual.



# Example: Host-Parasitoid Dynamics





# Example: Host-Parasitoid Dynamics

## 1995: Herren

SHARE



### Dr. Hans Rudolf Herren

#### SWITZERLAND

**DR. HANS RUDOLF HERREN**, recipient of the 1995 World Food Prize, was only 31 years old when he took a new job in Africa and landed right in the middle of an unprecedented crisis: an insect, the cassava mealybug, was devastating the continent's staple crops, and widespread hunger was emerging as a real possibility. Within ten years, Dr. Herren had almost single-handedly developed a chemical-free biological control for the mealybug, eliminated the threat to cassava production, averted disastrous famine, and saved upward of 20 million lives.





# Resampling from the Past [Aldous et al., 1988]

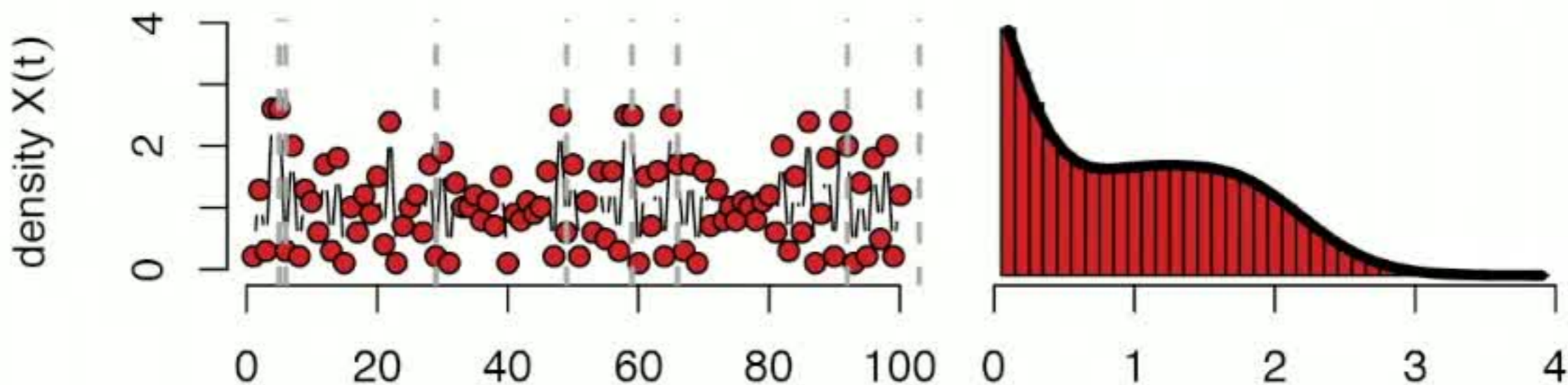
To approximate QSD, define  $Y(t)$  on  $\mathcal{S}_+$  by:

- ▶ Choose  $\tilde{Y}$  according to  $\mathbb{P}[\tilde{Y} = y | Y(t) = x]$
- ▶ If  $\tilde{Y} \in \mathcal{S}_+$ , set  $Y(t+1) = \tilde{Y}$
- ▶ If  $\tilde{Y} \in \mathcal{E}$ , then update  $Y(t+1)$  according to

$$\mathbb{P}[Y(t+1) = y] = \frac{\#\{0 \leq s \leq t : Y(s) = y\}}{t+1}$$

Aldous et al. [1988] proved that

$$\lim_{t \rightarrow \infty} \frac{\#\{0 \leq s \leq t-1 : Y(s) = x\}}{t} = \pi(x) \text{ with probability one}$$





# California Annuals in Serpentine Soils ( $k = 2$ )

$\lambda_i$  maximal per-capita seed production of species  $i$

$\alpha_{ij}$  competitive effect of species  $j$  on species  $i$

$$x_1(t + 1) = x_1(t) \frac{\lambda_1}{1 + \alpha_{11}x_1(t) + \alpha_{12}x_2(t)}$$

$$x_2(t + 1) = x_2(t) \frac{\lambda_2}{1 + \alpha_{22}x_2(t) + \alpha_{21}x_1(t)}$$

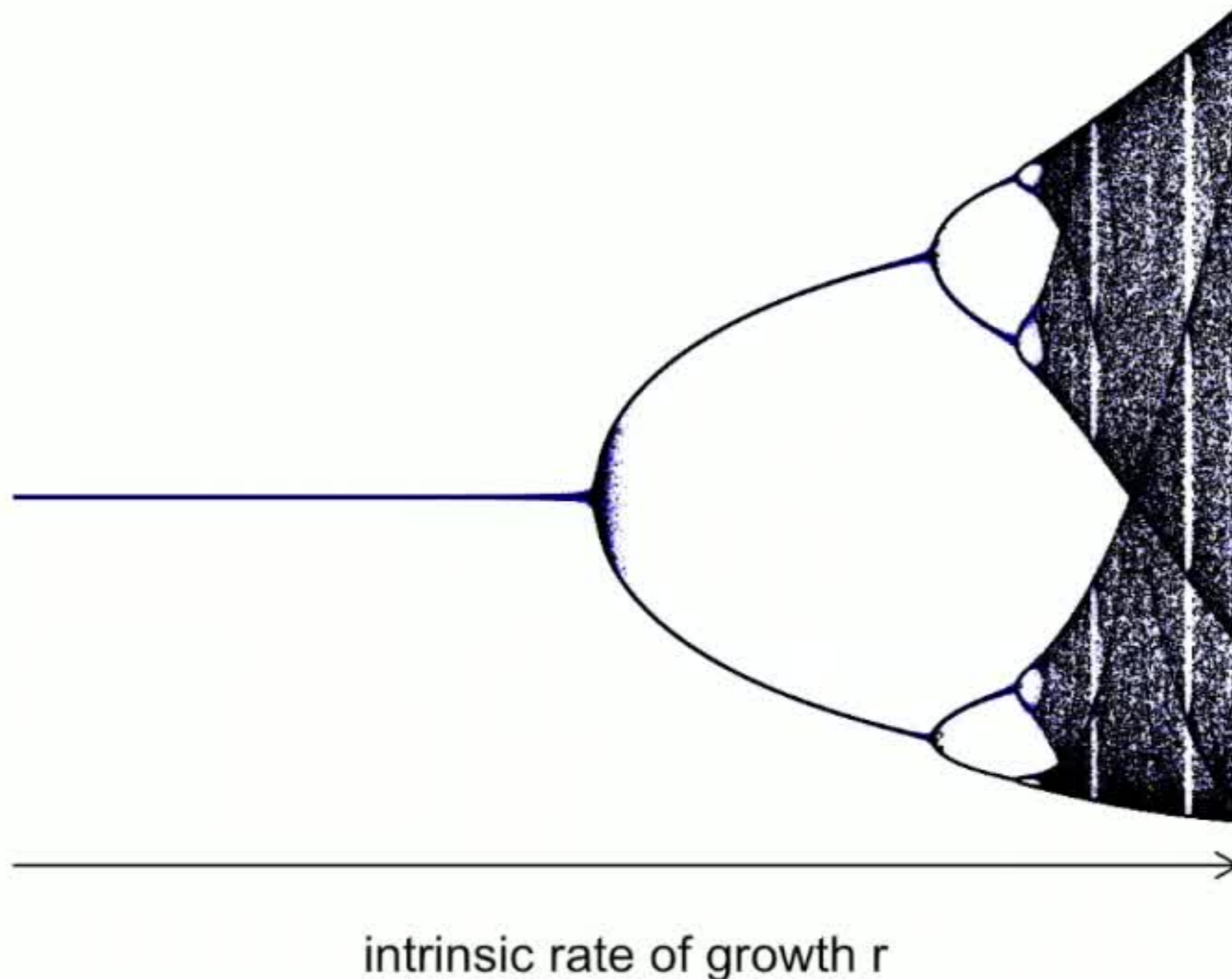
Schreiber et al. [2018]





# Revenge of the Ricker ( $k = 1$ )

$x_{t+1} = x_t \exp(r(1 - x_t))$ . Kozlovski [2003]  $\Rightarrow$  an open and dense set of positive  $r$  values satisfying Theorem's assumptions





# Example: Host-Parasitoid Dynamics



**Parasitoid:** An organism that, during its development, lives in or on the body of a single host individual, eventually killing that individual.