

Reading the Secrets of Biological Fluctuations

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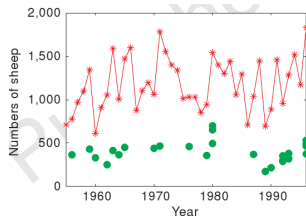
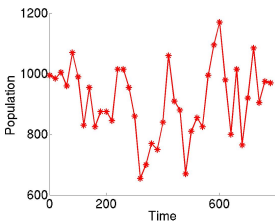
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- 1 Noisy Biology
- 2 Fluctuation Regimes
- 3 Model Choice
- 4 Macroscopic Phenomena
- 5 Fluctuation Dominance

Why study fluctuations?

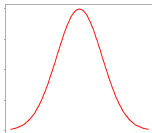
- Biology is noisy and we want to understand it.
- Stochasticity can drive phenomena we would miss in deterministic models.
- Fluctuations hold the key to deeper biological understanding?



Grenfell et al. (1998) *Nature*

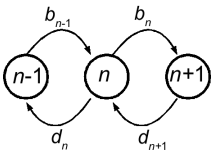
Variables at the Macroscopic and Individual Levels

- **Deterministic models** describe **macroscopic** behavior
- Individual based model are described by transition rates between states – a *Markov process*
- **Macroscopic variable** ϕ is independent of details of system (intensive), i.e. *population density*
- **Individual-based variable** n depends on system size (extensive), i.e. *population number*
- In a given area Ω with a macroscopic density ϕ , we'd expect to find $\langle n \rangle = \phi\Omega$ *on average*, which is more accurate with larger Ω .



Theory of Fluctuations

Markov process



$$\Rightarrow n = \Omega\phi + \Omega^{1/2}\xi \Rightarrow$$

Linear Noise Approximation



Fundamental Equations

$$\frac{d\phi}{dt} = \alpha_{1,0}(\phi) + \alpha_{1,0}''(\phi)\sigma^2 \tag{1}$$

$$\frac{d\sigma^2}{dt} = 2\alpha_{1,0}'(\phi)\sigma^2 + \alpha_{2,0}(\phi) \tag{2}$$

$$\alpha_{1,0}(\phi) = b(\phi) - d(\phi), \quad \alpha_{2,0} = b(\phi) + d(\phi)$$

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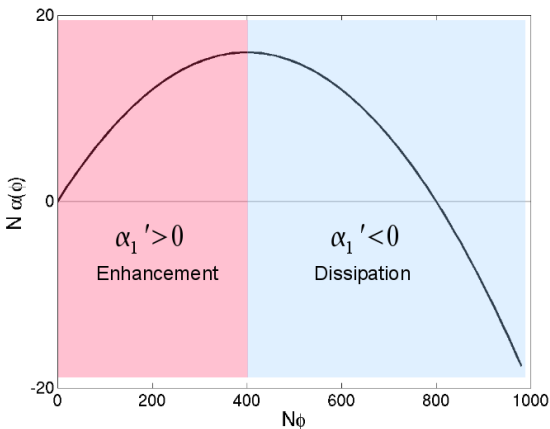
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Distinct Fluctuation Regimes

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$$\frac{dn}{dt} = \underbrace{c \frac{n}{N} \left(1 - \frac{n}{N}\right)}_{b_n} - \underbrace{e \frac{n}{N}}_{d_n}$$

$$\frac{d^2n}{dt^2} = 2\alpha'_{1,0}(\phi)\sigma^2 + \alpha_{2,0}(\phi)$$

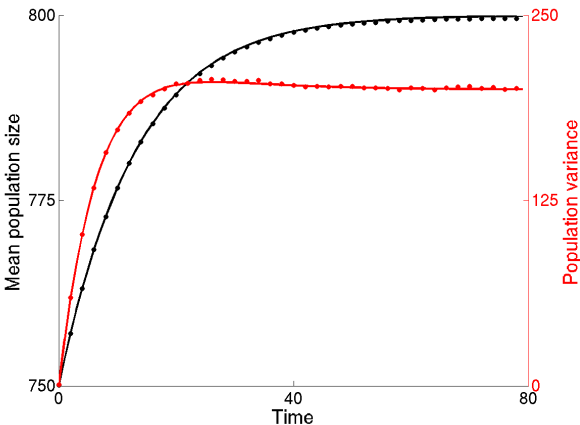


Near Equilibrium: Fluctuation Dissipation Regime

In the dissipation regime, fluctuations exponentially relax to the equilibrium level

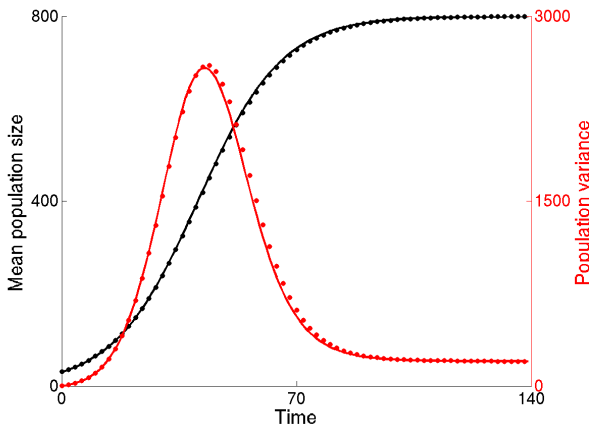
$$\hat{\sigma}^2 = \frac{b(n)+d(n)}{2[d'(n)-b'(n)]}$$

- $N = 1000$, $e = 0.2$,
 $c = 1$
- $\hat{n} = N \left[1 - \frac{e}{c}\right] = 800$
- $\hat{\sigma}^2 = N \frac{e}{c} = 200$
- Dots are simulation averages, lines are theoretical prediction



Fluctuation Enhancement

With an initial condition starting deep in the enhancement regime, fluctuations grow exponentially. At $N = 400$, dissipation takes over and fluctuations return to the same equilibrium as before.

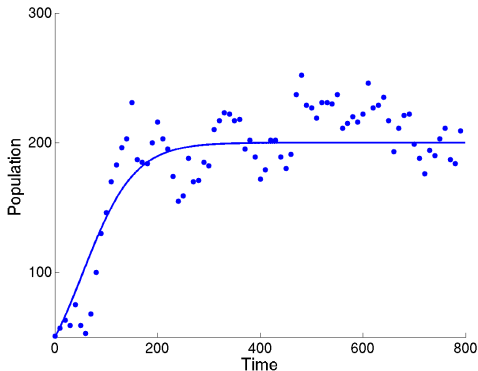


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Which model best describes this data?

$$\frac{dn}{dt} = \underbrace{c \frac{n}{N} \left(1 - \frac{n}{N}\right)}_{b_n} - \underbrace{e \frac{n}{N}}_{d_n}$$

$$\frac{dn}{dt} = \underbrace{rn}_{b_n} - \underbrace{\frac{rn^2}{K}}_{d_n}$$

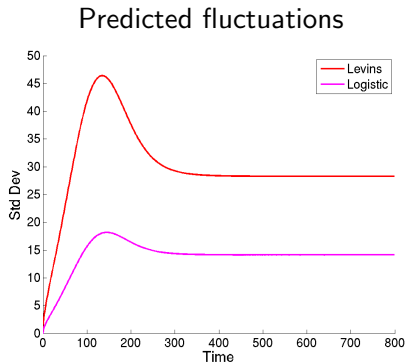


... and why does it matter?

Using the Information Hidden in the Fluctuations

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- 1 Independently parameterize **birth & death** rates, see which is density dependent
- 2 Works with **single realization** at equilibrium
- 3 With replicates: The dynamic equations can determine **functions** $b(n)$ and $d(n)$
- 4 Uses **more information** to inform model choice
- 5 Can **discount weights** of points from high-variance regions when model-fitting



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Stochastic Corrections: Deflation and Inflation

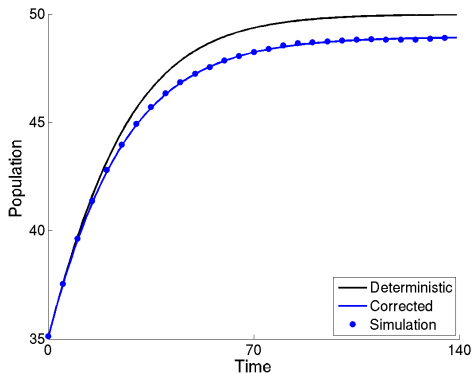
$\alpha''_{1,0}(\phi) < 0 \implies$ Fluctuations suppress the average relative to the deterministic approximation.

Our theory accurately predicts the extent of this effect.

Recall $\alpha_{2,0} = b_n + d_n$ controls the magnitude of this effect.

Ecological and evolutionary consequences for when variability is favorable?

$$\frac{d\phi}{dt} = \alpha_{1,0}(\phi) + \alpha''_{1,0}(\phi)\sigma^2$$

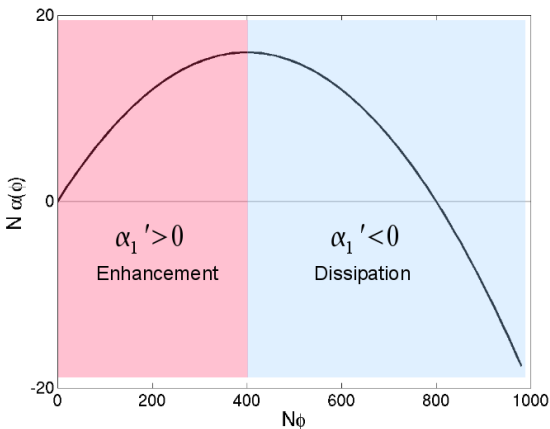


Fluctuation Phenomena: Deflation

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$$\frac{dn}{dt} = \underbrace{c \frac{n}{N} \left(1 - \frac{n}{N}\right)}_{b_n} - \underbrace{e \frac{n}{N}}_{d_n}$$

$$\frac{d\phi}{dt} = \alpha_{1,0}(\phi) + \alpha''_{1,0}(\phi)\sigma^2$$



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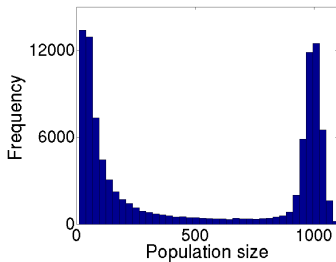
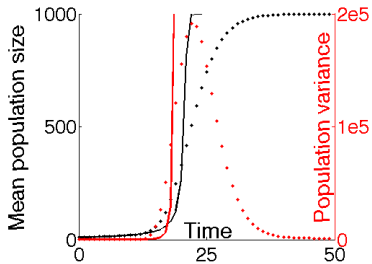
Fluctuation Dominance

Far from equilibrium, enhancement can expand the fluctuations until they reach the macroscopic scale.

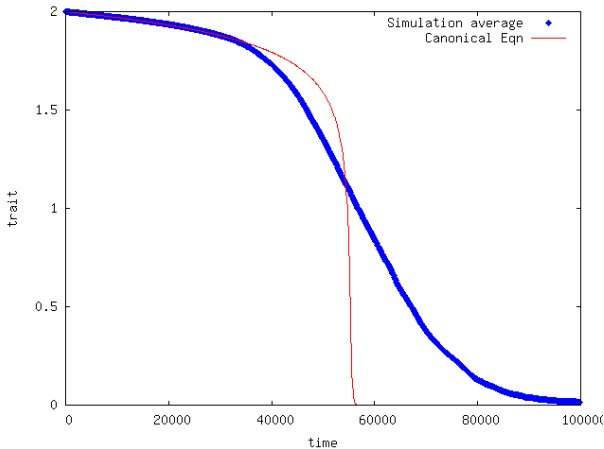
- Variance equation fails dramatically
- Mean trajectory need not follow the deterministic trajectory
- Bimodal distribution of trajectories can emerge
- **Conjecture:** occurs when neighborhood exists for which $\alpha_{1,0} \approx 0$ and $\alpha'_{1,0} \approx 0$



Breakdown of the approximation



Breakdown of the Canonical Equation of Adaptive Dynamics



Further Topics

This approach can be applied to a variety of stochastic processes in biology...

- **The multivariate theory:** multiple species or age structured populations. Predicts covariances as well.
- **Macroevolutionary theory:** inferring speciation and extinction rates from phylogenetic trees
- **Adaptive dynamics:** quantifying uncertainty in the canonical equation, correcting for fluctuations.

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