

## Lecture 4 - Density-independent stochastic growth

### Announcement:

Hand back Q1

Bring laptops next class

### Today's concepts:

Stochasticity (Environmental vs. Demographic)

Expectation vs. Variance (uncertainty)

Geometric vs. Arithmetic expectation

Review: Exponential (geometric) as first-order approximation

Show examples of population time series

Discrete vs. continuous Show videos

All previous equations have been deterministic

... if we start with same  $N(0)$  and  $r$  we get exactly the same answer.

Add stochasticity: "Adding stochastic shell to deterministic core of our model"

**Stochastic** (from the Greek for *aim* or *guess*) = random with respect to considered variables

A stochastic process is one whose subsequent state is determined by a random element.

$$N_T = N_0 e^{(r \pm \text{noise})T} = N_0 (\lambda \pm \text{noise})^T$$

E.g., 4 years of stochastic growth

$$N_0 = 1$$

$$\lambda_1 = 2$$

$$\lambda_2 = 1$$

$$\lambda_3 = 3$$

$$\lambda_4 = 2$$

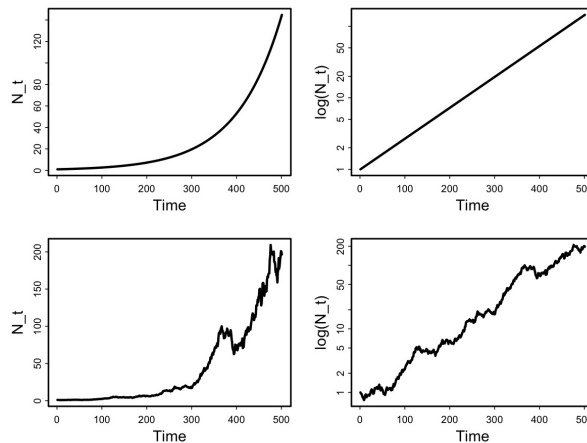
$$N_1 = N_0 \lambda_1 = 1 \cdot 2 = 2$$

$$N_2 = N_1 \lambda_2 = 2 \cdot 1 = 2$$

$$N_3 = N_2 \lambda_3 = 2 \cdot 3 = 6$$

$$N_4 = N_0 \lambda_1 \lambda_2 \lambda_3 \lambda_4 = 12$$

Show code in R...



If we wanted to determine:

Expectation (mean) of  $N_T$  ( $\bar{N}_T$ ) and variance of  $N_T$  ( $\sigma_{N_T}^2$ ) over all time points:

$$\bar{N}_T = E[N_T] = \frac{1}{T} \sum_t^T N_t = \frac{1 + 2 + 2 + 6 + 12}{5} = 4.6$$

$$\begin{aligned} \sigma_{N_T}^2 &= Var[N_T] = \frac{1}{T} \sum_t^T (N_t - \bar{N}_T)^2 \\ &= \frac{(1 - 4.6)^2 + (2 - 4.6)^2 + (2 - 4.6)^2 + (6 - 4.6)^2 + (12 - 4.6)^2}{5} = 83.2/5 = 16.64 \end{aligned}$$

Note that  $\sigma$  = standard deviation

and that  $CV$  (coefficient of variation) =  $\sigma^2/\bar{N}$

## Environmental Stochasticity - temporal variation in population's per capita growth rate

Importance of distinguishing between geometric vs. arithmetic mean...

Example: Dynamics of two populations with same initial size:

$$N(3) = N_0 \lambda_1 \lambda_2 \lambda_3$$

$$\text{Population A: } N_A(3) = N_0 \cdot 2 \cdot 1 \cdot 3 = 6 \cdot N_0$$

$$\text{Population B: } N_B(3) = N_0 \cdot 2 \cdot 2 \cdot 2 = 8 \cdot N_0$$

$$\bar{\lambda} = \frac{1}{T} \sum_t^T \lambda_t$$

On natural (arithmetic) scale,  $\bar{\lambda}_A = \bar{\lambda}_B = 2$ .

*So why does population B grow more?*

Appropriate measure is the geometric mean (remember: popn growth is a multiplicative process):

(No standard symbol for geometric mean)

$$\text{geometric mean } \lambda = \left( \prod_t^T \lambda_t \right)^{\frac{1}{T}} = \sqrt[T]{\prod_t^T \lambda_t}$$

$$\text{Geometric mean of } \lambda_A = \sqrt[3]{\lambda_1 \cdot \lambda_2 \cdot \lambda_3} = \sqrt[3]{2 \cdot 1 \cdot 3} = 1.817...$$

$$\text{So in fact } N_A(3) = N_0 \cdot 1.817 \cdot 1.817 \cdot 1.817 = 6 \cdot N_0$$

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### Side note: Contrasting arithmetic vs. geometric mean

Two numbers (expressed relative to  $x$ ):  $\overleftarrow{a} \quad \overrightarrow{b}$  What is  $x$ ?

$$a - x = x - b$$

$$a + b = 2x$$

$$\frac{a+b}{2} = x$$

$$\frac{a}{x} = \frac{x}{b}$$

$$a \cdot b = x \cdot x$$

$$a \cdot b = x^2$$

$$\sqrt{a \cdot b} = x$$

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Will show that: *Natural log of geometric mean  $\lambda$  = arithmetic mean of natural log of the  $\lambda$ 's*

$$\text{Since... } \ln(b^a) = a \ln(b) \Rightarrow \ln \left( \left( \prod_{t=1}^T \lambda_t \right)^{1/T} \right) = \frac{1}{T} \cdot \ln(\lambda_1 \cdot \lambda_2 \dots \lambda_T)$$

$$\begin{aligned} \text{Since... } \ln(ab) = \ln(a) + \ln(b) \Rightarrow &= \frac{1}{T} (\ln(\lambda_1) + \ln(\lambda_2) + \dots + \ln(\lambda_T)) \\ &= \frac{1}{T} \cdot T \cdot \overline{\ln(\lambda)} \\ &= \overline{\ln(\lambda)} \end{aligned}$$

Thus, could also calculate as:

$$\text{Geometric mean } \lambda = e^{\overline{\ln(\lambda)}}$$

An alternative approximation (provided in Case, pg. 35):

$$\begin{aligned} \text{Geometric mean } \lambda &= e^{\overline{\ln(\lambda)}} \approx e^{\ln(\bar{\lambda}) - \frac{\sigma_{\lambda}^2}{2\bar{\lambda}^2}} \\ &\quad (\text{Note effects of } \sigma_{\lambda} \text{ and of } \bar{\lambda}.) \end{aligned}$$

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**Class R exercise** - random vector draws - compare geometric and arithmetic means

- Effect of  $\sigma = 0$ , and increasing  $\sigma$  values:

Geometric mean will always be less than arithmetic mean

The more variation, the lower the geometric mean

- Effect of sample size,  $n$ :

Little effect. very low  $n$  might have more variation in depression amount

- Effect of  $\bar{\lambda}$ :

Raises and lowers, but note that at  $\bar{\lambda} \approx 1$ , geometric mean  $\lambda < 1$  (declining population)!!

**What does this mean for the dynamics of a given focal population?**

| <i>Population change</i> | Deterministic $\lambda$ | Deterministic $\ln(\lambda)$ | Environmental noise $\ln(\lambda \pm \text{noise})$            |
|--------------------------|-------------------------|------------------------------|--|
| No change                | 1                       | 0                            | NA   |
| Growth                   | $> 1$                   | $> 0$                        | $\frac{\sigma_\lambda^2}{2\lambda^2} < \ln(\bar{\lambda}) > 0$ |
| Decline                  | $< 1$                   | $< 0$                        | $\ln(\bar{\lambda}) - \frac{\sigma_\lambda^2}{2\lambda^2} < 0$ |

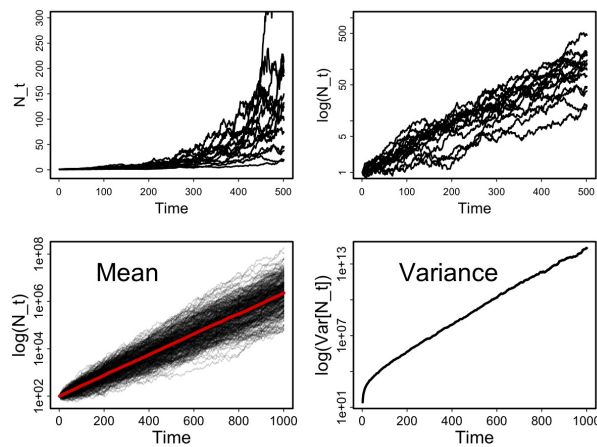
**What are the expected mean and variance of  $N_T$  for an “average” (typical) population?**

That is, we will now ask a different question...

⇒ What is the expected population size  $N_T$  over an ensemble of replicate populations?

(The average of  $n$  replicate populations.)

Run in R...



Will show that

$$\bar{N}_T = N_0 \bar{\lambda}^T = N_0 e^{\bar{r}T}$$

And that...

when there is *environmental variation* only:

$$\sigma_{\ln(N_T)}^2 = T \cdot \sigma_{\ln(\lambda_t)}^2$$

when there is only *demographic variation*:

$$\sigma_{N_T}^2 = \begin{cases} 2N_0 \bar{b}T & \text{if } \bar{b} = \bar{d} \\ \frac{\bar{b} + \bar{d}}{\bar{b} - \bar{d}} N_0 e^{\bar{r}T} (e^{\bar{r}T} - 1) & \text{if } \bar{b} \neq \bar{d} \end{cases}$$

**How to get  $E[N_T]$ ?**

Let  $\lambda$  be a random variable from a normal distribution.

Any given  $\lambda$  from this distribution is denoted by  $\lambda_t$ .

$$\begin{aligned}
 N_T &= N_0 \prod_{t=1}^T \lambda_t = N_0 \cdot (\lambda_1 \lambda_2 \dots \lambda_T) \\
 E[N_T] &= N_0 \cdot E\left[\prod_{t=1}^T \lambda_t\right] \quad (\text{since } N_0 \text{ is a constant}) \\
 &= N_0 \cdot \prod_{t=1}^T E[\lambda] \\
 &= N_0 \cdot E[\lambda]^T \\
 &= N_0 \cdot \bar{\lambda}^T
 \end{aligned}$$

**How to get variance?**

Transform by taking the log...

$$\ln(N_T) = \ln(N_0) + \ln\left(\prod_{t=1}^T \lambda_t\right) = \ln(N_0) + \boxed{\sum_{t=1}^T \ln(\lambda_t)}$$

Now working on the arithmetic scale!

Allows us to apply Central Limit Theorem (pg. 41 of Case):

*The sum of independent, identically distributed random variables  $x_i$  tends to asymptote to the normal density distribution, no matter the underlying distribution of  $x_i$*

Under a sufficiently large number of independent random variable draws:

$$\begin{aligned}
 \boxed{\sum_{i=1}^n x_i} &= \mathcal{N}(n \cdot E[x], n \cdot \text{Var}[x]) \\
 &= \mathcal{N}(n\bar{x}, n\sigma_x^2)
 \end{aligned}$$

*The mean of the sum equals the sum of all individual means  
The variance of the sum equals the sum of all individual variances*

Inserting  $\ln(\lambda_t)$  for  $x_i$ , we thus have

$$\begin{aligned}
 \sum_{t=1}^T \ln(\lambda_t) &= \mathcal{N}(T \cdot E[\ln(\lambda_t)], T \cdot \text{Var}[\ln(\lambda_t)]) \\
 &= \mathcal{N}(T \cdot \overline{\ln(\lambda_t)}, T \cdot \sigma_{\ln(\lambda_t)}^2)
 \end{aligned}$$

Therefore, we have

$$E[\ln(N_T)] = \ln(N_0) + T \cdot \overline{\ln(\lambda_t)}$$

and

$$\sigma_{\ln(N_T)}^2 = T \cdot \sigma_{\ln(\lambda_t)}^2$$

To translate expectation back to arithmetic scale:

Even though the expected value of  $N_T$  is  $E[N_T] = N_0 \cdot \bar{\lambda}^T$ ,

$$N_T \sim \log\mathcal{N} \text{ with median} = e^{\ln(N_0) + T \cdot \overline{\ln(\lambda)}}$$

Take-home message:

**When modeling  $N_{t+1} = \lambda N_t$  with environmental noise in growth rate, error must be  $\log\mathcal{N}$ !**

In Problem Set #2 you should therefore use  $N_{t+1} = N_t(\lambda \pm e^\epsilon)$ , where  $\epsilon \sim \mathcal{N}(0, \sigma_\epsilon^2)$ .

## Summary

- The solution to  $Var[N_T]$  is sensitive to assumed model-formulation!
- Dependent on how stochasticity is assumed to affect  $\lambda$ .
- If you assume  $N_T = N_0(\lambda \pm \epsilon)^T$  with  $\epsilon \sim \mathcal{N}(0, \sigma_\epsilon^2)$  you'll get a different prediction!

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**Demographic stochasticity** - Between-individual variation in per capita growth rate  
Analogy of flipping a coin (not perfect, but will do).

$H$  = birth and  $T$  = death

**Q:** How many  $H$  and  $T$  in 1000 flips? How many in 100? How many in 4?  
For fair coin,  $P(H) = P(T) = 0.5$  and  $P(H) + P(T) = 1 \Rightarrow$  Binomial distribution.

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## Class exercise in R

Number of extinctions as function of starting population size using binomial.  
Going to assume:

$$P(birth) = \frac{b}{b+d}$$
$$P(death) = \frac{d}{b+d}$$

where  $b$  and  $d$  are per capita birth and death rates, such that  $r = b - d$ .

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For true population (no longer binomial):

$$P(birth) = \frac{b}{b+d+o}$$
$$P(death) = \frac{d}{b+d+o}$$
$$P(other) = 1 - [P(b) + P(d)]$$

Won't go through derivations, but expectation of  $N(t)$  is still:

$$\overline{N}_t = N_0 e^{\bar{r}t}$$

But for variance:

$$\sigma_{N_T}^2 = \begin{cases} 2N_0 \bar{b} T & \text{if } \bar{b} = \bar{d} \\ \frac{\bar{b} + \bar{d}}{\bar{b} - \bar{d}} N_0 e^{\bar{r}T} (e^{\bar{r}T} - 1) & \text{if } \bar{b} \neq \bar{d} \end{cases}$$

Side note...Probability of extinction:

$$P(ext) = \left(\frac{d}{b}\right)^{N_0}$$

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Of course, environmental and demographic stochasticity are not mutually exclusive!  
Temporal variation among individuals will also causes temporal variation in  $\lambda$ .  
See Case for example combining the two.

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