

# Exponential Growth Part III: Randomness and Stochasticity

EFB 370: Population Ecology

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# Goals (concepts / buzzwords)

- I. Random variables:

- mean / variance / standard deviation / distributions / expected value
  - Binomial and Normal distributions

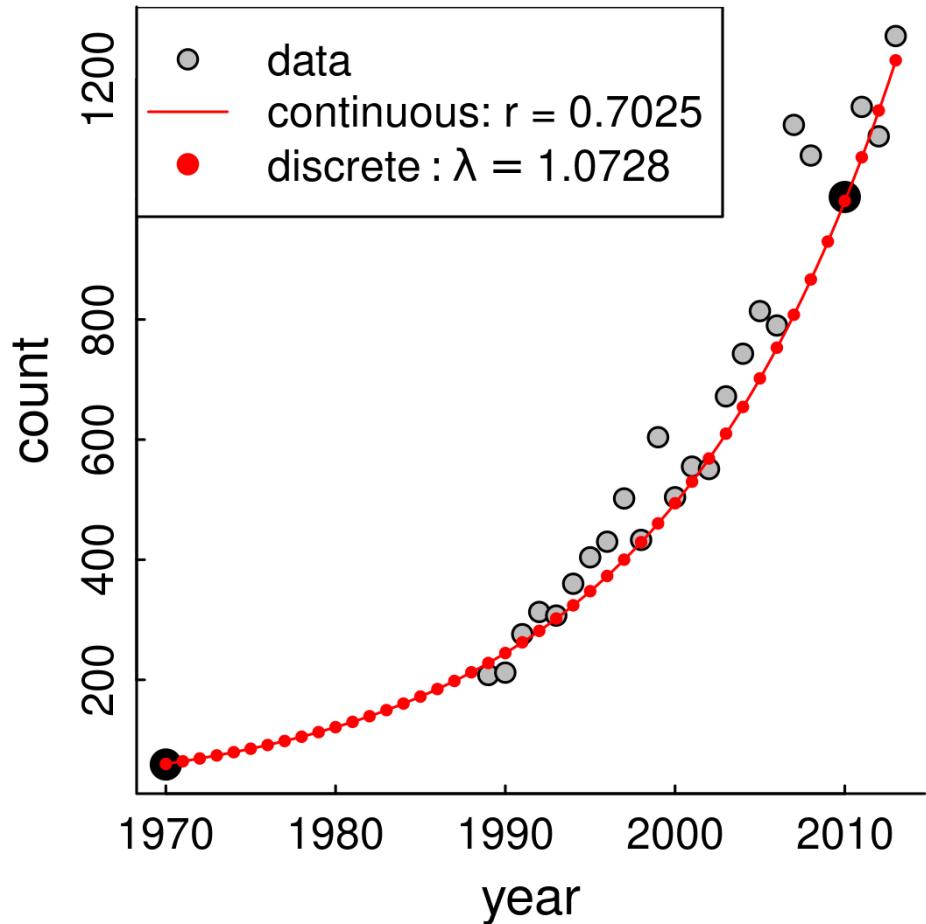
- II **stochastic** = randomness in time

- **demographic stochasticity** = randomness in individual processes
  - **environmental stochasticity** = randomness affecting populations

## Suggested reading:

- Gotelli: Chapter 1 (second half)

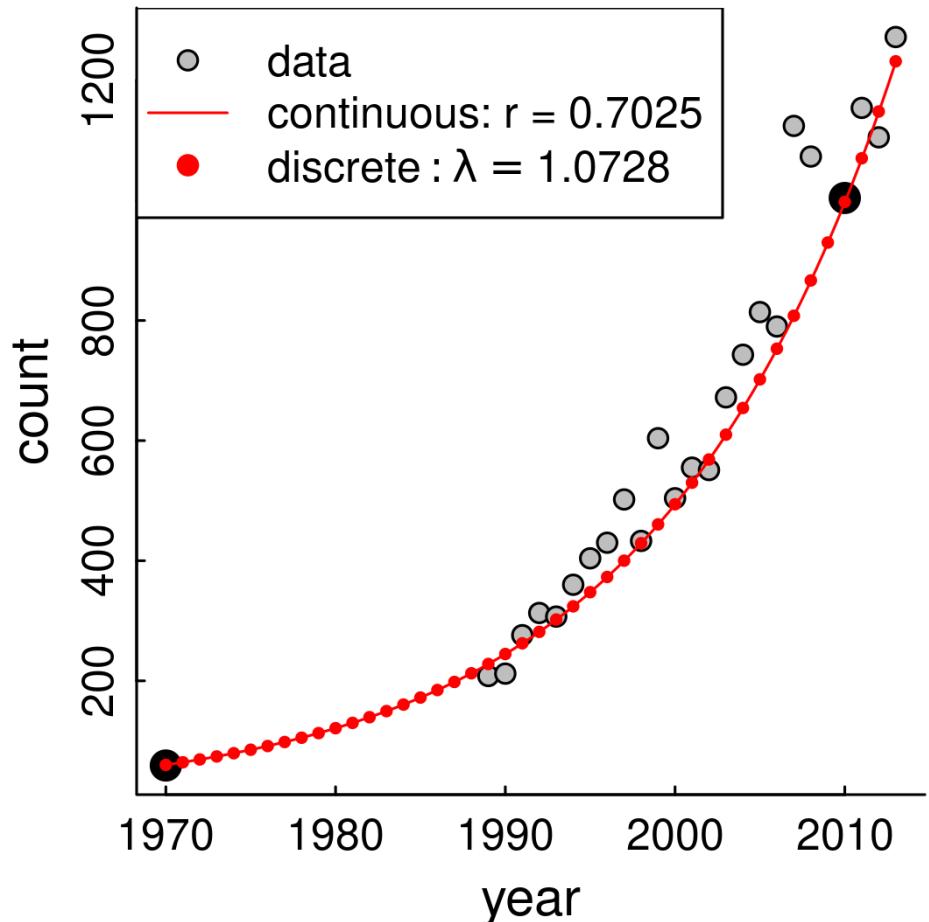
If the exponential model is so tidy ... why isn't it *perfect*?



Because ... *randomness!*(?)

*What are some potential sources of "randomness" in the sea otter population process?*

# If the exponential model is so tidy ... why isn't it *perfect*?



Because ... *randomness!*(?)

*What are some potential sources of "randomness" in the sea otter population process?*

- Environment good / bad affecting all animals ...
- Randomness in birth-death affecting individual animals ...
- Unexpected immigration / emigration ...
- Observation error ...

# Brief Intro to Random Variables

Easiest problem in boring arithmetic:

$$1 + 1 = 2$$

Easiest problem in random arithmetic:



+



= ?

# One coin flip is Random

Random variables have *possible events* associated with *probabilities*.

event	numeric	probability
	0	$1/2$
	1	$1/2$

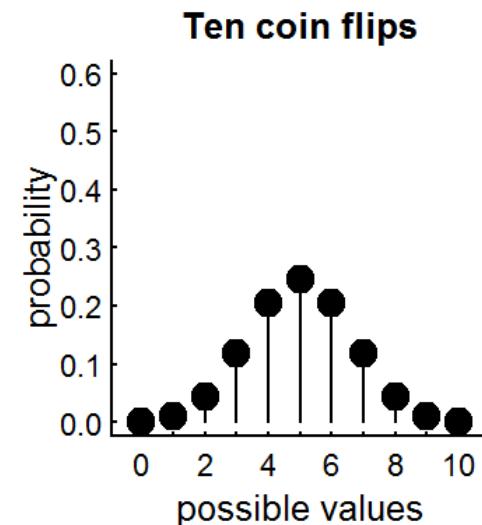
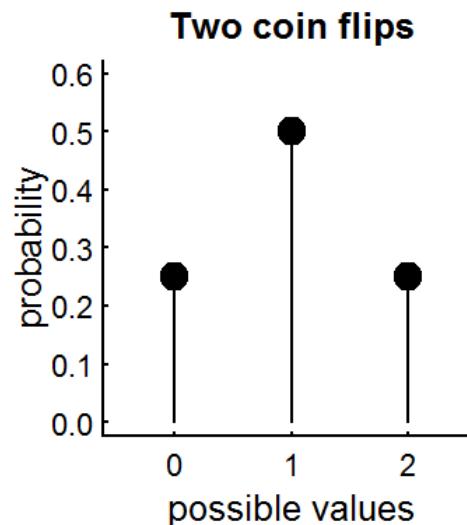
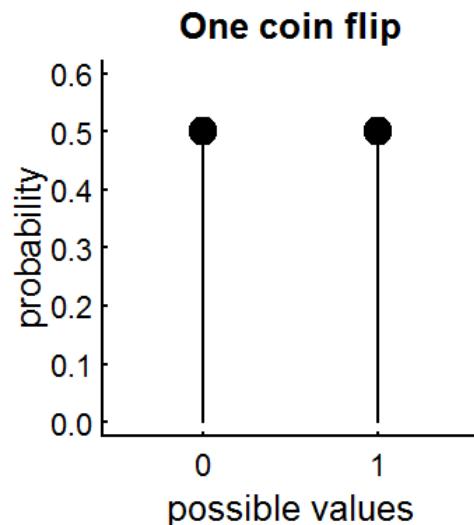
# Two coin flips is also Random

Random variables have *possible events* associated with *probabilities*.

event	numeric	probability
	0	$1/4$
 OR 	1	$1/2$
	2	$1/4$

# Random variables...

... are described by **Probability Distributions**. Probability Distributions have **names** and **parameters** that describe the distribution.



<b>One flip</b>	=	$X \sim \text{Bernoulli}(p = 1/2)$ .
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<b>Two flips</b>	=	$X_1 + X_2 = Y \sim \text{Binomial}(p = 1/2, n = 2)$ .
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<b>n flips</b>	=	$\sum_{i=1}^n X_i = Y \sim \text{Binomial}(p = 1/2, n = n)$ .
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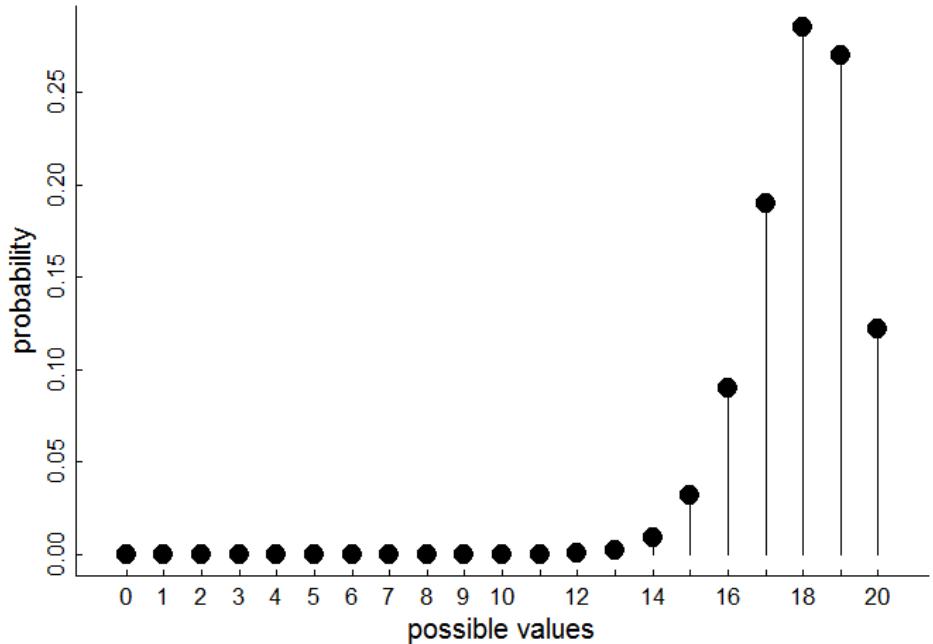
# A bit about the binomial distribution

$X \sim \text{Binomial}(n, p)$  describes the sum of  $n$  random events, each of which has probability  $p$ .

## Pop ecology example:

If you have 20 sea otters ( $X_t = 20$ ), each with a 90% probability of survival / year, how many next year?

**Answer:** A random variable with distribution:  
 $X_{t+1} \sim \text{Binomial}(n = 20, p = 0.9)$



# Key questions about a random variable:

$$X_{t+1} \sim \text{Binomial}(n = X_t, p = p_{\text{surv}})$$

**Q1. How many would you *expect* to survive?**

$$E(X) = \mu_X = np = 0.9 \times 20 = 18$$

This value is the **Expectation** of the distribution:  $E(X)$  or  $\mu(X)$ . It would be the *mean* value if you could repeat the experiment an infinite amount of times.

**Q2. How much *variability* is there in this process?**

$$SD(X) = \sigma_x = \sqrt{n p(1 - p)} = 1.34$$

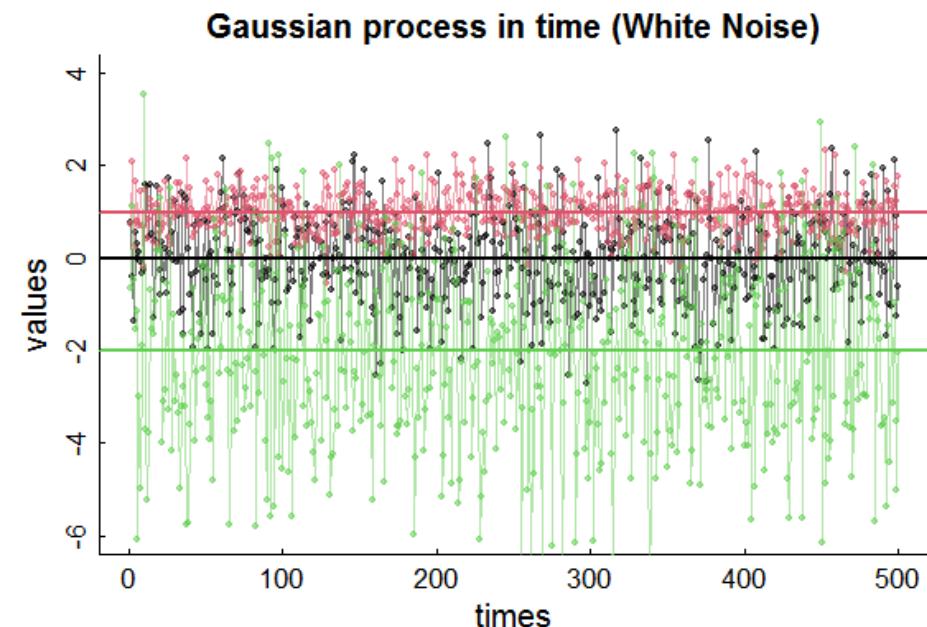
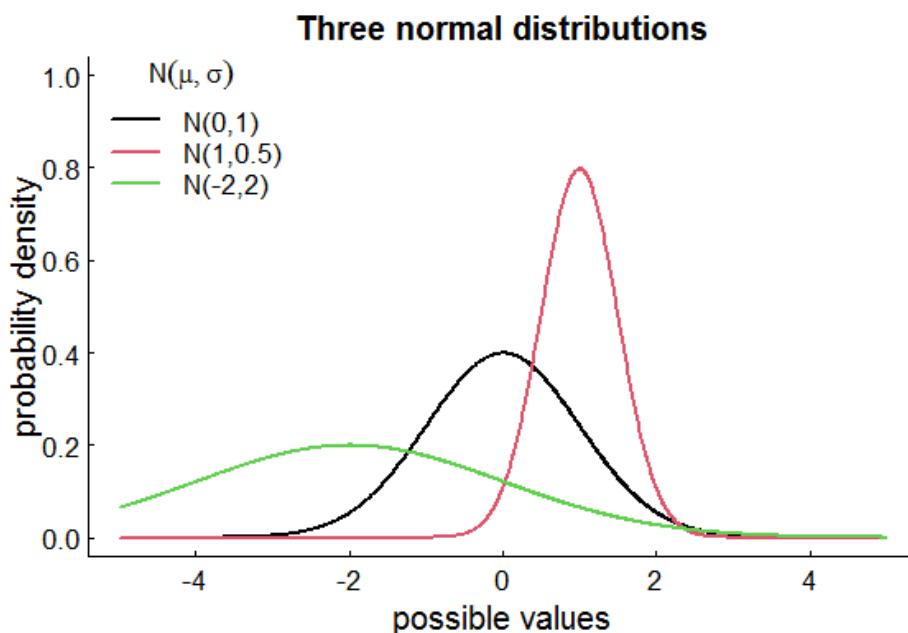
The **standard deviation** of a random variable:  $SD(X)$  or  $\sigma_x$  quantifies how concentrated the distribution is around the *mean*. *Approximately* 95% of the probability is within 2 standard deviations.

# Continuous random variables

There are a bunch. The most famous is the **Normal** or **Gaussian** distribution:

$$X \sim \mathcal{N}(\mu, \sigma)$$

It has two parameters:  $\mu$  and  $\sigma$  which are - unsurprisingly - the *mean* (expectation) and *standard deviation* (spread) of the variable.



# Once upon a time in Germany ....

GN4480100S8

Deutsche Bundesbank

Weltkarte Frankl

Frankfurt am Main  
1. September 1999



ZEHN DEUTSCHE MARK

# Some probability distributions

We use these to **model** random processes:

name	notation	possible values	models	mean	standard deviation
Normal	$\mathcal{N}(\mu, \sigma)$	$(-\infty, \infty)$	bell-shaped	$\mu$	$\sigma$
Exponential	$Exp(\lambda)$	$[0, \infty)$	random events	$\frac{1}{\lambda}$	$\frac{1}{\lambda}$
Poisson	$Poisson(\lambda)$	$[0, \infty)$	positive count data (e.g. births)	$\lambda$	$\sqrt{\lambda}$
Bernoulli	$Bernoulli(p)$	$[0, 1)$	binary outcomes (e.g. deaths)	$p$	$\sqrt{p(1 - p)}$
Binomial	$Binomial(n, p)$	$[0, n)$	many binary outcomes	$np$	$\sqrt{np(1 - p)}$

# Demographic Stochasticity

- **Stochasticity** means: **Randomness in time**.
- **Demography** is the **Science of Population Dynamics**. Often it refers specifically to **births** and **deaths** (and **movements** ... but we're still looking at closed population).
- Individually, *all* demographic processes are stochastic. An individual has some *probability* of dying at any moment. An individual has some *probability* of reproducing at a given time.

Q: How important is *individual* randomness for a *population* process?

More specific Q: **What is the probability of extinction?**

# Human Experiment with Life and Death

- 15 students
- Flip a survival coin.
  - If you die (**TAILS**) sit down, if you live (**HEADS**) stay standing
- Flip a reproduction coin.
  - If you reproduce (**HEADS**) call on another student to stand



# What do we predict from this experiment?

Starting with  $N_t$ : expected number of survivors (S):

$$E(S) = p_s N_t$$

Expected number of new individuals (babies - B):

$$E(B) = p_b E(S)$$

New population equals survivals + new babies:

$$E(N_{t+1}) = E(S) + E(B) = p_s N_t + p_b p_s N_t = p_s (1 + p_b) N_t$$

So (in our coin flip example)

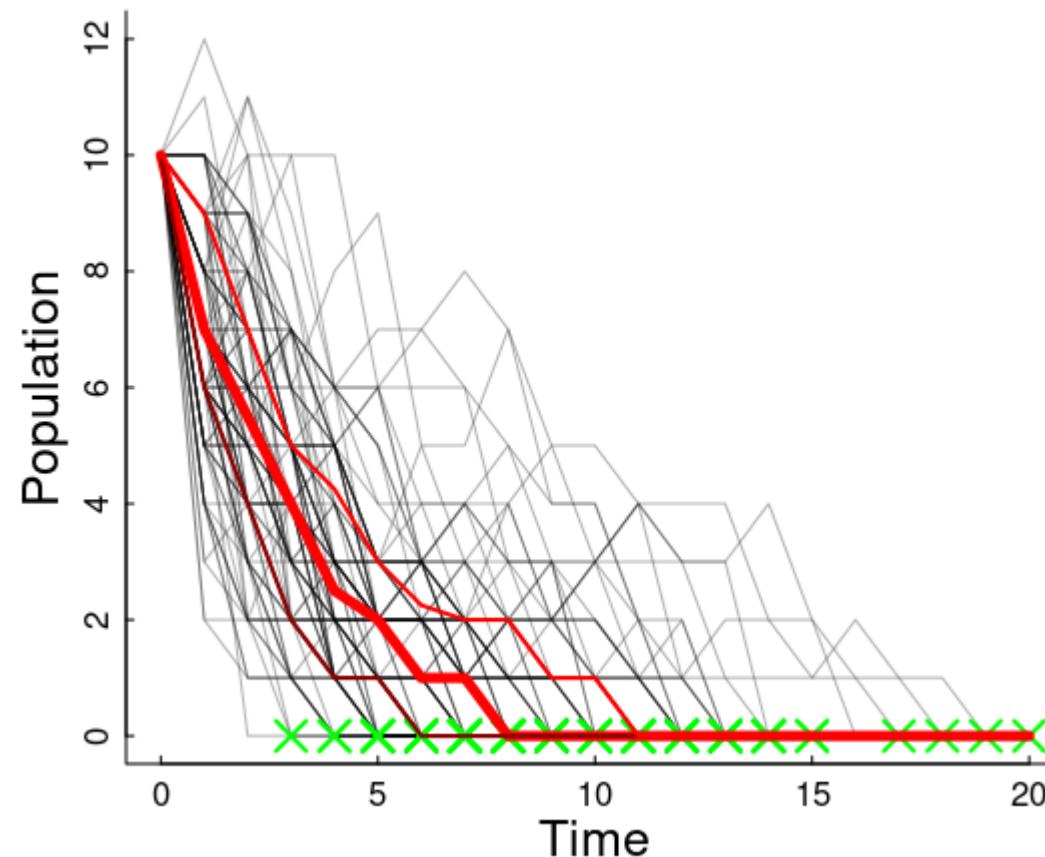
$$\hat{\lambda} = p_s (1 + p_b) = 0.75$$

.

**What does that mean for population growth!?**

# Cranking this experiment very many times.

<https://egurarie.shinyapps.io/StochasticGrowth/>



# Some predictions for a similar continuous model

Assume birth rate  $b$  and death rate  $d$ , and growth rate  $r = b - d$ .

The mean of the process is also exponential growth

$$E(N_t) = N_0 e^{rt}$$

If birth rate = death rate:

$$SD(N_t) \approx \sqrt{2N_0 bt}$$

Note, increases as square root of time. There's a somewhat more complex formula for births  $\neq$  deaths.

More importantly:

$$P(\text{extinction}) = \begin{cases} \text{if } b > d; & (d/b)^{N_0} \\ \text{if } d > b; & 1 \end{cases}$$

Note that even when birth rate = death rate,  $P(\text{extinction}) = 1$ , i.e. eventual extinction is certain. This is very similar to the eventual probability of fixation for genetic drift.

Also, probability of extinction is lower for smaller  $N_0$ .

# Environmental Stochasticity...

Some random aspect of the environment that influences  $r$  (whether via **births** or **deaths** or **both**) for the **entire population**.:

$$R \sim Dist(\mu_r, \sigma_r)$$

- $\mu_r$  is the *mean growth rate*
- $\sigma_r$  is the *variability on the growth rate*

Can you think of some examples, e.g. for sea otters?

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# Human Experiment

The professor IS the environment!

1.  $N_0 = 16$
2. If the environment flips **Heads**, the population doubles.
3. If the environment flips **Tails**, the population halves.



# Environmental Stochasticity: Analysis

This model can be written:

$$N_{t+1} = N_t 2^R$$

Where  $R$  is a random variable:

value	prob.
-1	1/2
+1	1/2

$$E(R) = \mu_r = 0$$

So population should not grow, on average.

$$E(N(r)) = N_0 e^{\mu_r t} = N_0$$

And:

$$SD(R) = \sigma_R = 1$$

- Are you doomed to extinction?
- Does this remind you of another process from earlier in class?

Experiment here: <https://egurarie.shinyapps.io/StochasticGrowth>

# A stochastic population is a random variable!

$$N(t) \sim \text{Dist}(\mu_N(t), \sigma_N(t))$$

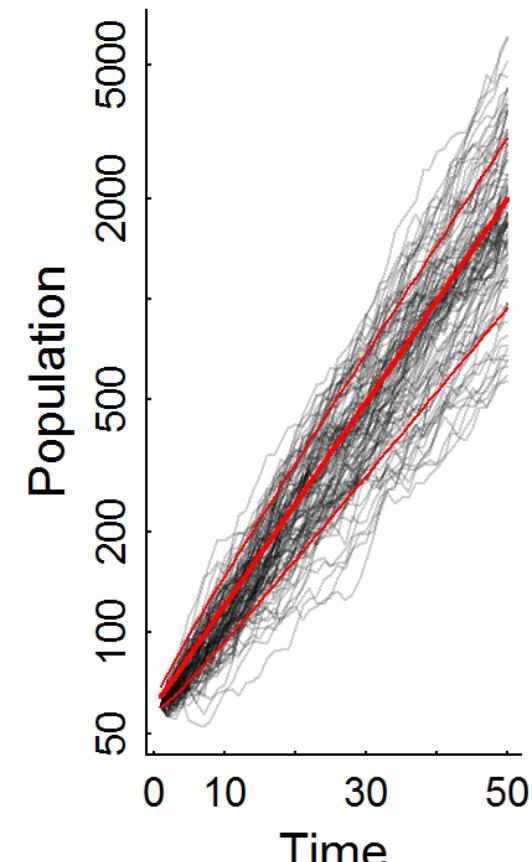
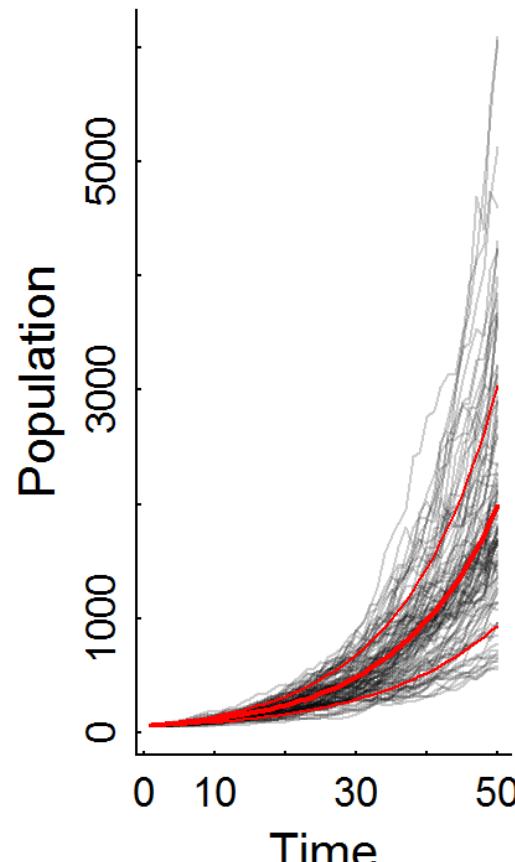
Sea otter example, with  $\mu_r = 0.07$  and  $\sigma_r = 0.07$ :

The mean / expectation is the same as for the exponential growth function, but with the mean growth rate substituted:

$$\mu_N(t) = N_0 e^{\mu_r t}$$

But the standard deviation of the function increases with time!

$$\sigma_N^2(t) \approx N_0 e^{\mu_r t} \sqrt{e^{\sigma_r^2 t} - 1}$$



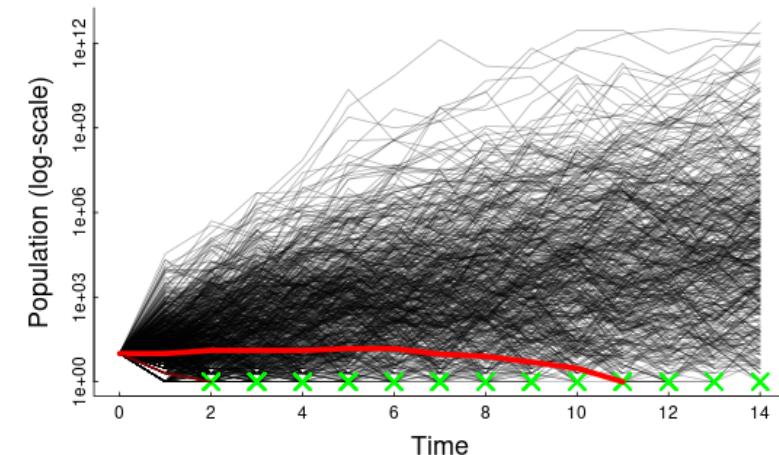
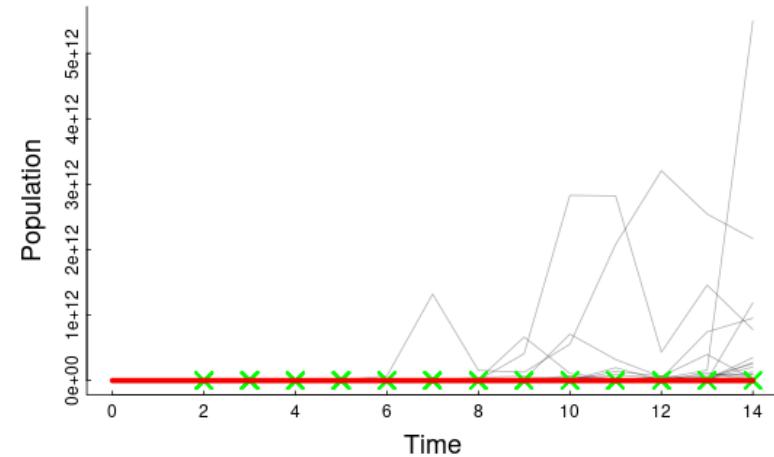
# If things are TOO random that spells trouble!

According to theory, if:

$$\sigma_r > 2\mu_r$$

extinction is nearly certain - even if  $\mu_r$  is positive and on average there is growth.

Experiment here:  
<https://egurarie.shinyapps.io/StochasticGrowth>



# Key Takeaways

1. Nature is a combination of **very complex** and **truly random**. So we have to get comfortable with randomness
2. **Demographic stochasticity** refers to **individual randomness**, and particularly effects **small populations**.
3. **Environmental stochasticity** affects **everyone similarly**, big or **small** populations, and too much randomness bodes ILL.

