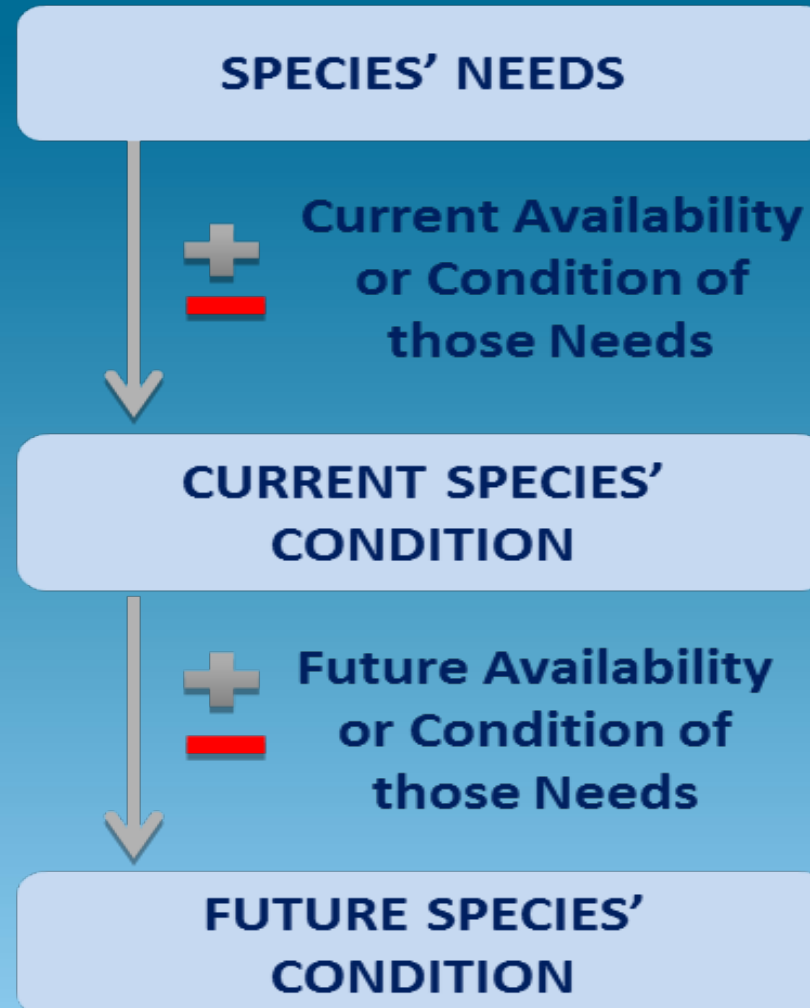


# Species Status Assessments

**SSA 200**



# Species Status Assessment



# Objectives of this class

- Become familiar with common terminology and approaches used in population modeling including constraints, weaknesses, and underlying assumptions.
- Identify an appropriate analytical approach for assessing the status of a species given the available data.
- Communicate/understand relevant analytics and interpret population modeling results pertinent to an SSA.
- Communicate population modeling results to decision makers as part of SSA results.

*“Models can do cool things for biologists and decision makers, and we want the biologists to think they’re cool—not voodoo.”*

# What this class is not:

- A class intended to teach technical skills in analysis or modeling
- A computer coding class to learn how to use R or some statistical program
- Etc...

# Day 1

- Types of data typically available and used in SSAs and associated analyses
  - Descriptive data
  - Presence and presence/absence data
  - Count data over time and space
  - Demographic data (mark recapture, nest monitoring)

# Day 1

- Analysis of existing data can give insight into:
  - Species ecological needs
    - Relating data on the species to environmental variables
  - Species current status
    - Current redundancy
    - Current resiliency

# Day 2

- Projection models for each data type
  - Randomized categorical projections
  - Single-state and multistate occupancy modeling
  - Linear projection models, Poisson process projection models
  - Matrix population models, matrix state transition models

# Day 2

- Projection models give insight into:
  - Future redundancy
  - Future resiliency
- Using the results of the current status and needs analysis to predict future redundancy and resiliency



# Questions?

# The purpose of modeling

- Statistical analysis of data
  - Assess and describe patterns
  - Estimate system or population parameters
  - Relate population parameters to ecological or environmental variables
  - Estimate and characterize variability in data
- Use statistical analysis to predict the future
  - Organize and structure your uncertainty/variability

# George Box's Pipe



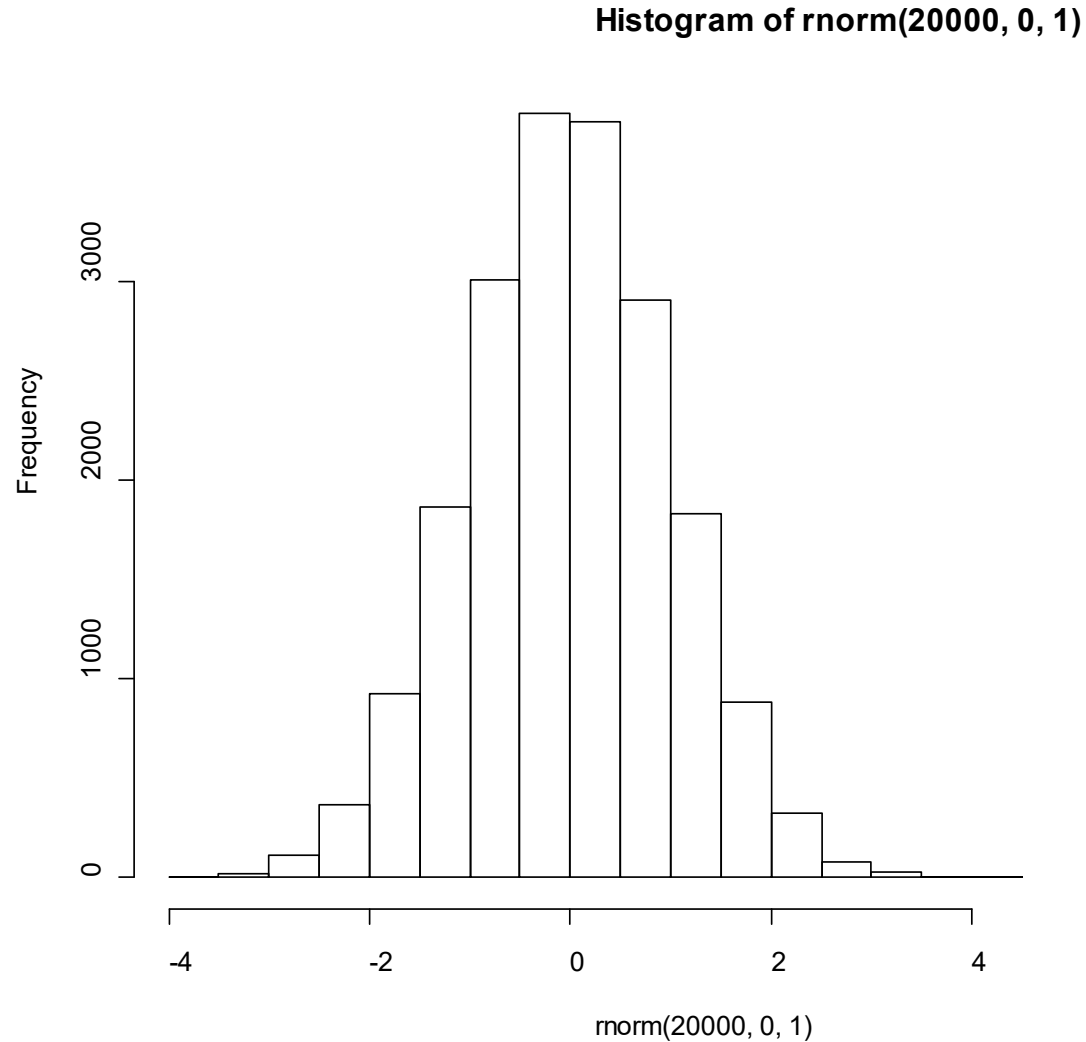
# Forms of uncertainty

- Partial controllability
- Observational uncertainty
- Environmental variation
- Ecological uncertainty
- Demographic stochasticity

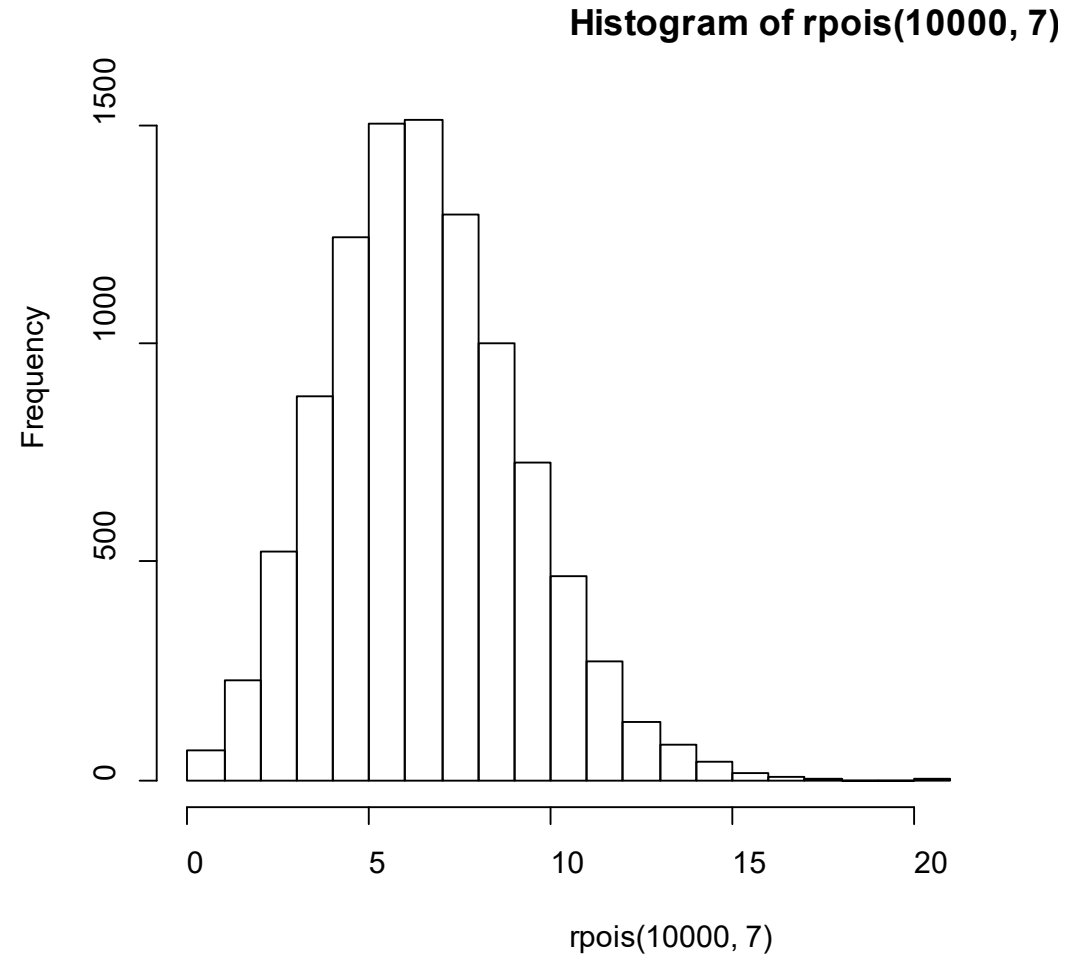
# Statistical distributions characterize uncertainties

- We use statistical distributions to generate parameter values with some variation
- The distributions have some predefined properties that reflect our knowledge or data of how a parameter behaves
  - e.g., Mean and SD or variance

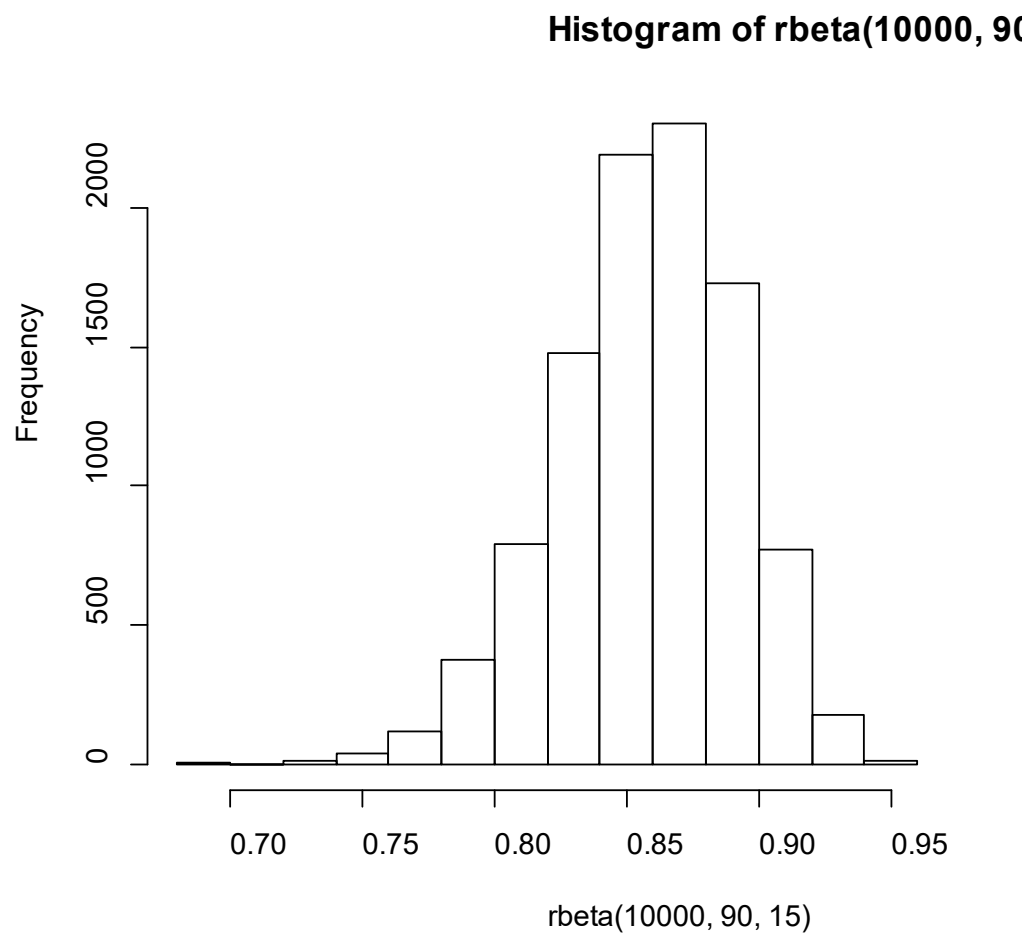
# The normal distribution



# The Poisson distribution



# The beta distribution



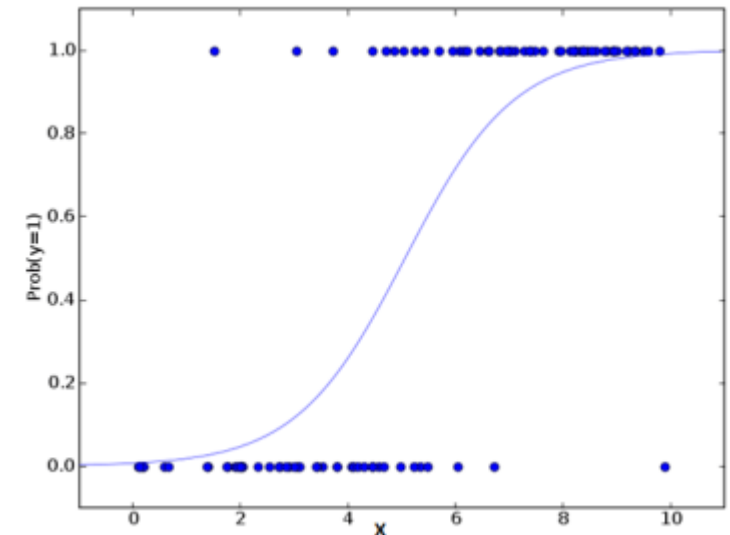
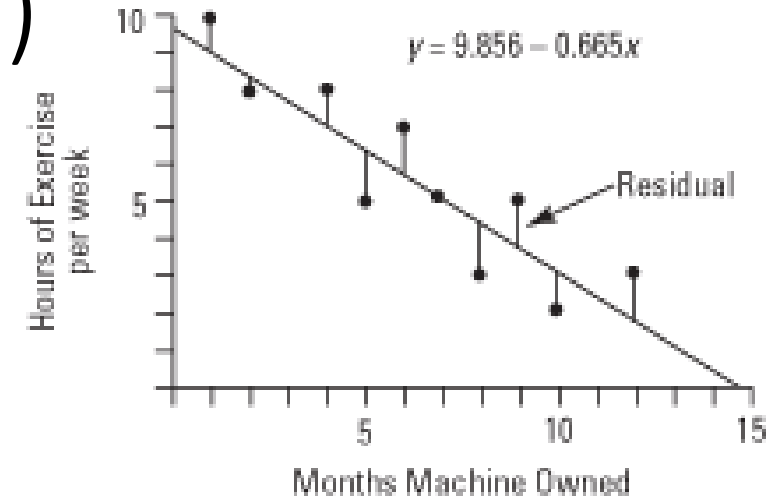


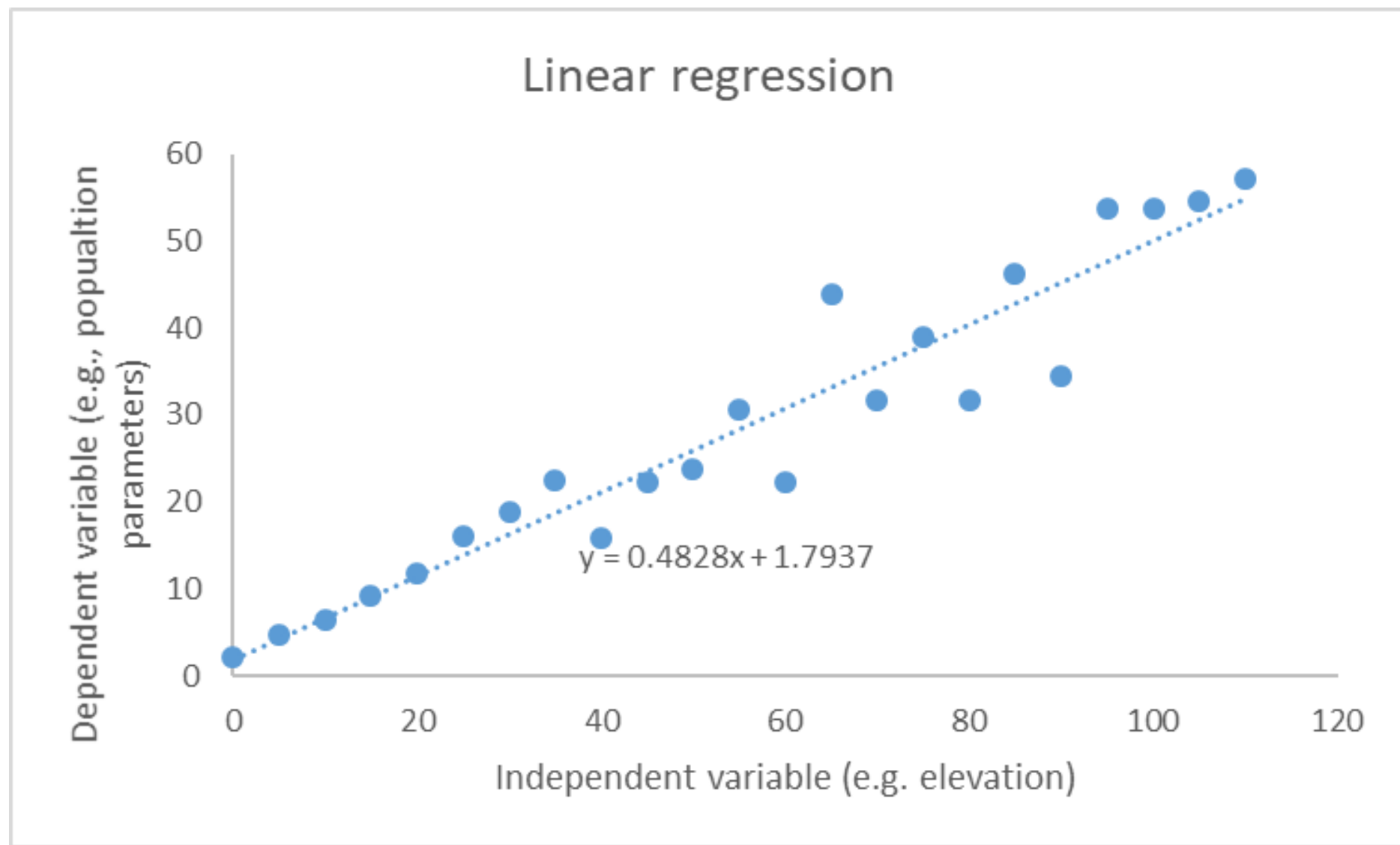
# Application and use in the class and beyond

- The analysis tools and projection tools will refer to these statistical distributions
  - E.g., count data → Poisson distribution
- Simulation models will use these distributions to mimic variability in a virtual setting

# Generalized Linear Models (GLM)

- Regression modeling is the central premise for relating parameters to environmental variables
- Least squares regression – normal distribution
- GLMs use other underlying distributions
  - Poisson (Count data)
  - Binomial (Logistic regression)
  - Negative binomial (count data with lots of zero counts)





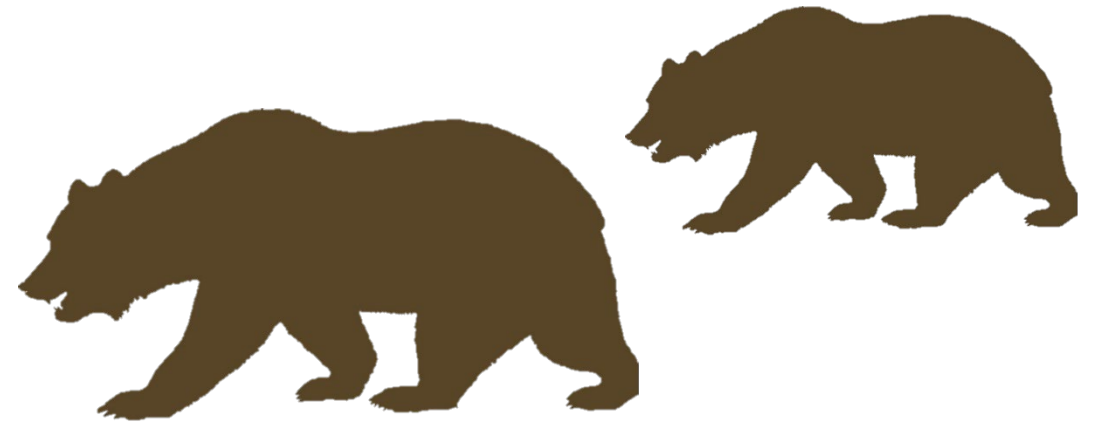
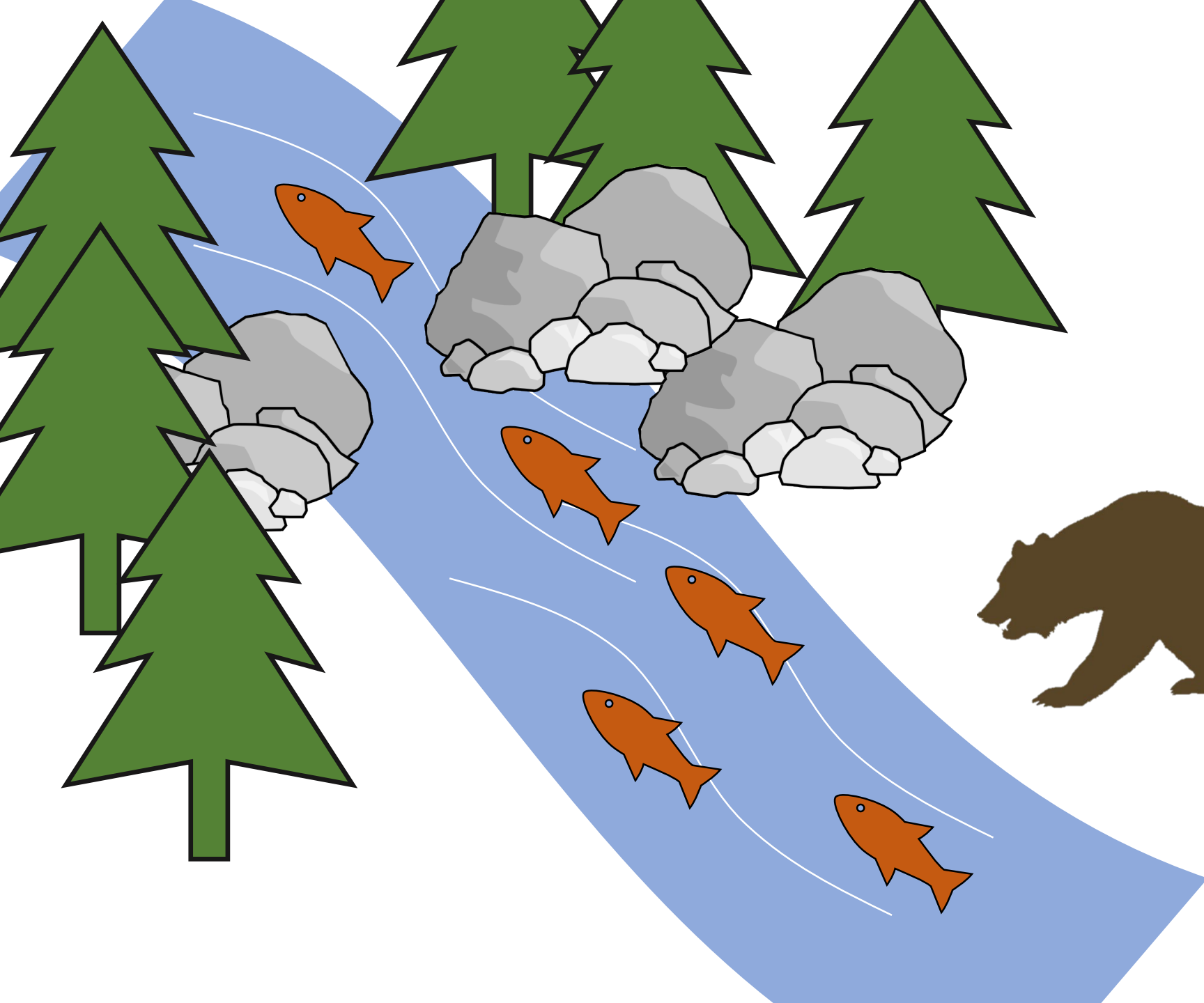
Multiple regression:  
relating parameters to multiple variables

$$y = b_1X + a$$

*The parameter  $y$  is a function of  $X$ .*

$$y = b_1X + b_2Z + b_3K + a$$

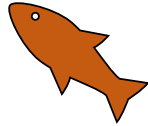
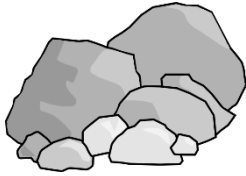
*The parameter  $y$  is a function of  $X$ ,  $Z$ , and  $K$ .*



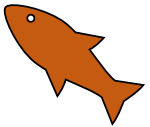
# What variables best predict bear occurrence?



$$y = b_1W + a$$



$$y = b_1W + b_2T + b_3R + b_4S + a$$



$$y = b_2T + b_4S + a$$

*None of the above*

$$y = a$$

***How do we know which one is better??***

# Comparing models using AIC

- AIC is a metric used to rank competing models that are fit with the same data
- Find the model that fits the best with the fewest number of parameters (parsimony)

Model	Model fit	Number of parameters	AIC rank
water	Not great	2	3
water + trees + rocks + salmon	Better	5	2
trees + salmon	Best	3	1
<i>none of the above</i>	Worst	1	4

# Questions?

