

# Demographic Matrix Population Models

**SSA 200**

# Applications to SSAs

- Models are designed to output useful metrics on future resiliency and redundancy
  - Future abundance, extinction probability, population growth rate
- Models allow us to predict future condition of the populations and characterize uncertainties

# Lecture outline

- Simple model construction
  - Review of data types/analysis
  - From conceptual to quantitative
- Incorporate environmental covariates and density dependence
- Model environmental and parametric uncertainty

# Constructing the model

1. Choose state variables (age, stage)
  - Dependent on data and species
2. Use demographic data to estimate vital rates for each state
  - Fecundity, survival probability, recruitment probability
3. Use state-specific vital rate estimates to create matrix model

# State variables

- Age classes (Leslie Matrix)
  - Equal time intervals & all individuals advance at next time
    - Short lived species with age-specific data
- Stage classes (Lefkovitch Matrix)
  - Unequal time intervals
  - Population divided by developmental stage or size
    - Difficult to age individuals but can get length, height, etc.
    - Juvenile, subadult
    - Seeds, dormancy, small plants, large plants

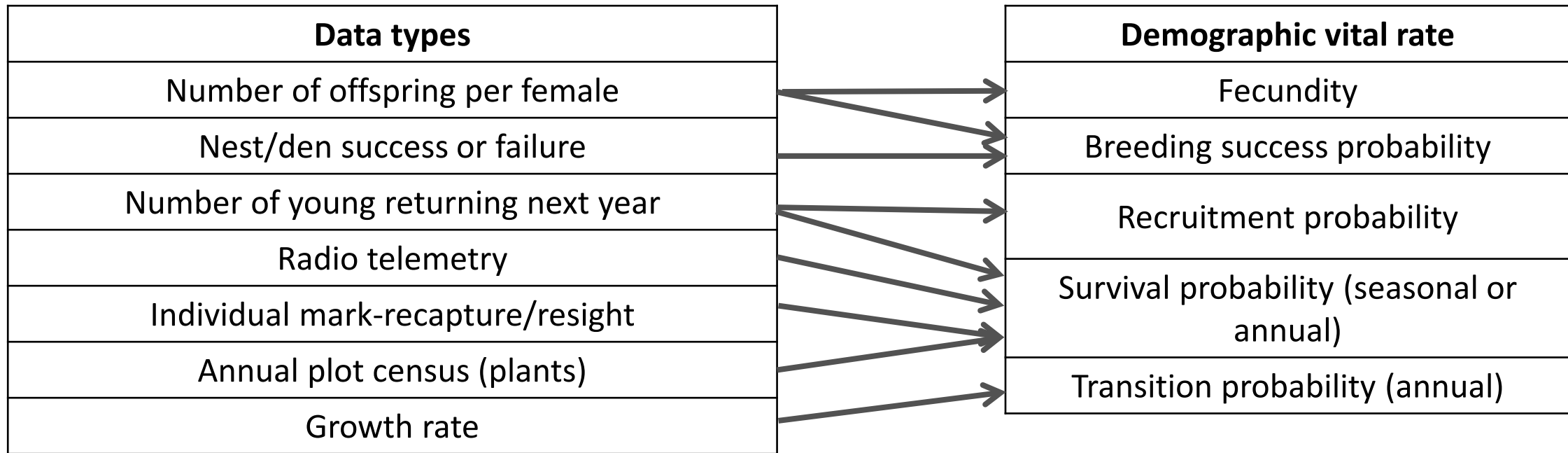


# Constructing the model

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# Demographic data

- Depends on ecology/life history of species



# Estimate vital rates

- Fecundity
  - Number of offspring/female -> Poisson GLM
  - Successful breeding (yes/no) -> Binomial GLM
- Annual survival/mortality
  - Radio telemetry -> known-fate models
  - Individual CMR -> Cormack-Jolly-Seber (CJS) models
  - Proportion of marked plants alive next year
- OR – use values reported in literature/expert opinion

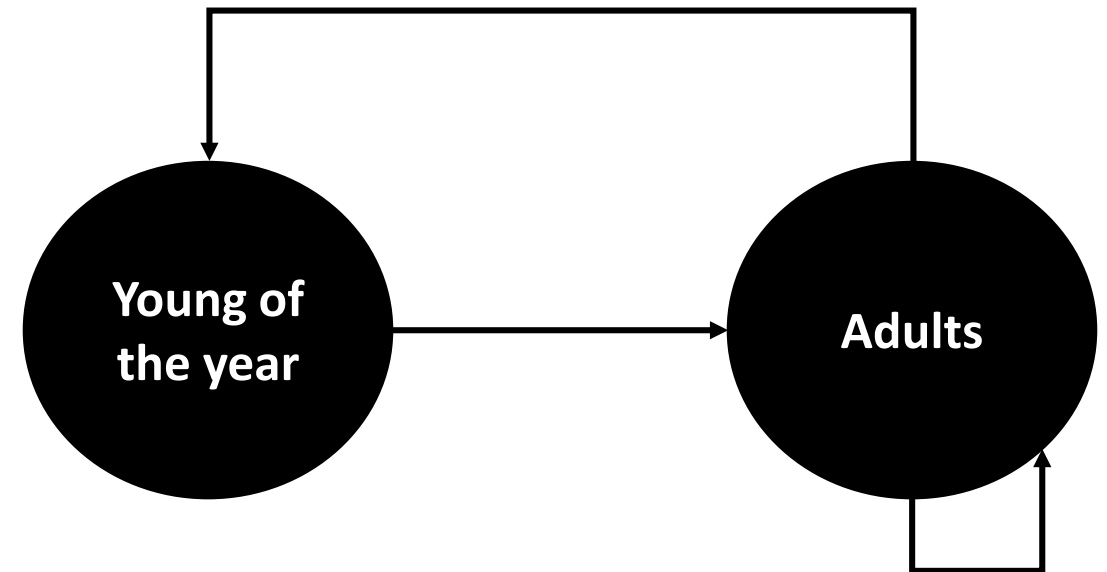


# Constructing the model

1. Choose state variables (age, size, stage)
  - Dependent on data and species
2. Use demographic data to estimate vital rates for each state
  - Fecundity, survival probability, recruitment probability
3. Use state variables and state-specific vital rate estimates to create a conceptual model and matrix model

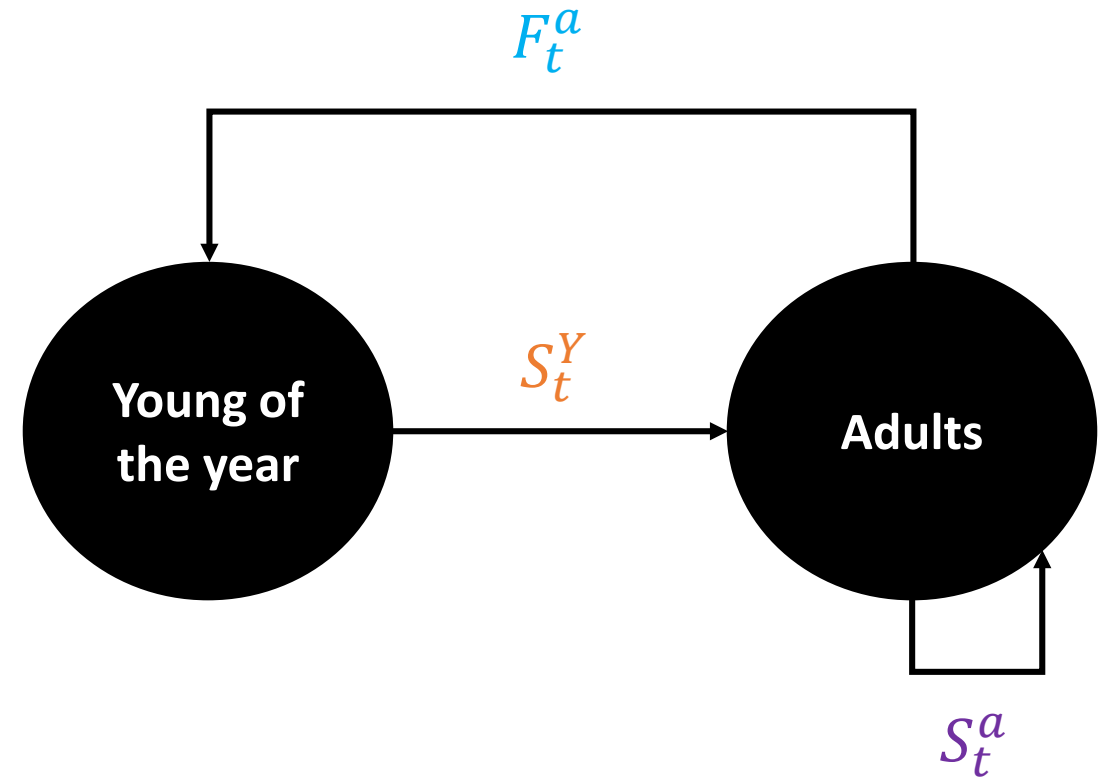
# Conceptual model

- Species with two stage classes
  - Survival
    - Adults – CMR data
    - Young of year – in literature
  - Fecundity
    - Number of offspring/ adult female



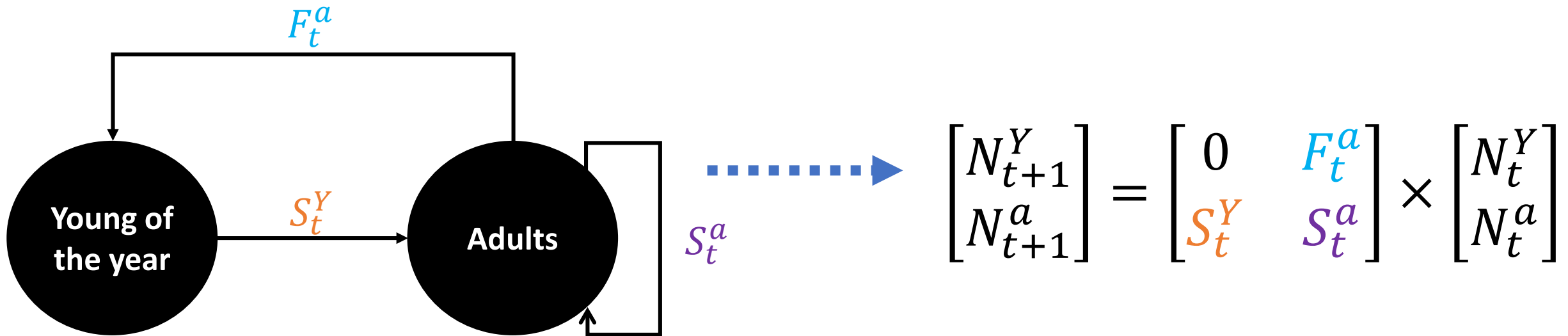
# Conceptual model

- Species with two stage classes
  - Survival
    - Adults – CMR data
    - Young of year – estimate in literature
  - Fecundity
    - Number of offspring/ adult female



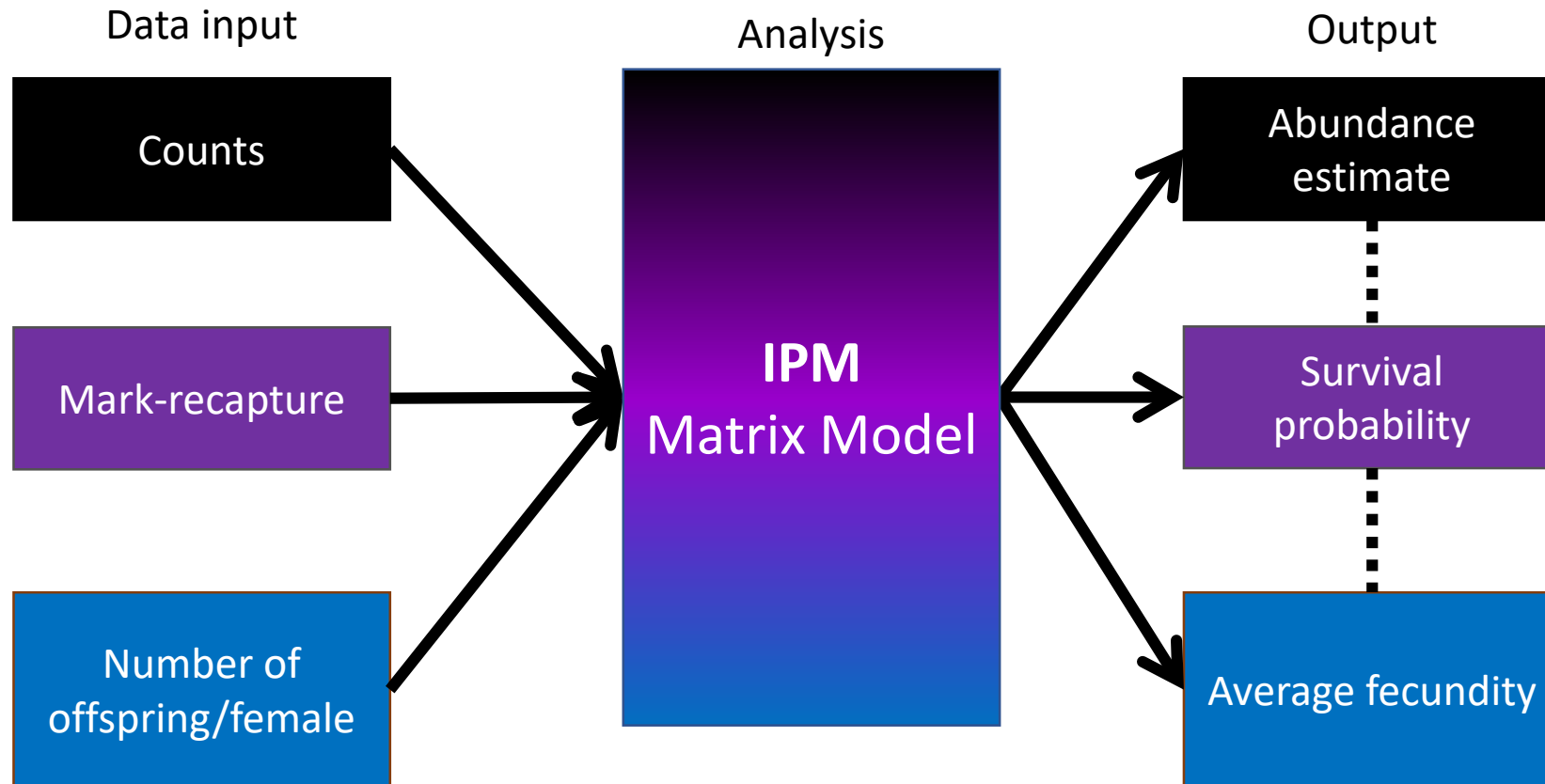
# Create matrix model

- Matrix models are used to present, analyze, and project population dynamics



# Integrated population models

$$\begin{bmatrix} N_{t+1}^Y \\ N_{t+1}^a \end{bmatrix} = \begin{bmatrix} 0 & F_t^a \\ S_t^Y & S_t^a \end{bmatrix} \times \begin{bmatrix} N_t^Y \\ N_t^a \end{bmatrix}$$



- Combine demographic data with counts
- Use the model to make projections

# Matrix projections

$$\begin{bmatrix} N_{t+1}^Y \\ N_{t+1}^a \end{bmatrix} = \begin{bmatrix} 0 & F_t^a \\ S_t^Y & S_t^a \end{bmatrix} \times \begin{bmatrix} N_t^Y \\ N_t^a \end{bmatrix}$$

$$N_{t+1} = A * N_t \quad \longleftrightarrow \quad N_{t+1} = \lambda * N_t$$

# Matrix projection outputs

- Abundance over time
- Population growth rate
  - Lambda ( $\lambda$ )
    - $\lambda = 1.0$  stationary
    - $\lambda = 1.10$  increasing 10% per year
    - $\lambda = 0.90$  decreasing 10% per year
- Extinction and/or quasi-extinction risk
- Sensitivity and elasticity

## Population Resiliency

# Sensitivity and elasticity

$$\begin{bmatrix} N_{t+1}^Y \\ N_{t+1}^A \end{bmatrix} = \begin{bmatrix} a_{1,1} & a_{1,2} \\ a_{2,1} & a_{2,2} \end{bmatrix} \times \begin{bmatrix} N_t^Y \\ N_t^A \end{bmatrix}$$

$$Sensitivity = \frac{\delta \lambda}{\delta a_{2,2}}$$

Sensitivity is the rate of change in population growth ( $\lambda$ ) with respect to a change in any element of the matrix.

$$Elasticity = \frac{a_{2,2}}{\lambda} \frac{\delta \lambda}{\delta a_{2,2}}$$

Elasticity analysis estimates the effect of a **proportional** change in the demographic rates on population growth ( $\lambda$ ).



# How to estimate sensitivity and elasticity

- Program R – Package ‘PopBio’

Population matrix  $\rightarrow \begin{bmatrix} 0 & F_t^A \\ S_t^Y & S_t^A \end{bmatrix} \rightarrow \begin{bmatrix} 0 & 1.2 \\ 0.3 & 0.8 \end{bmatrix}$

Elasticity matrix  $\rightarrow \begin{bmatrix} 0 & 0.222 \\ 0.222 & 0.554 \end{bmatrix}$

# Simple future condition assessments

- Using sensitivity and/or elasticity output
  - Results indicate population growth is most sensitive to **adult survival**
    - Conceptual modeling and lit review suggest that adult survival is negatively affected by drought frequency
    - Drought frequency will increase over next 50 years
  - What can we expect given this information?
    - Adult survival will likely decrease
    - Population growth will likely decrease
    - If climate predictions are accurate, future resiliency will decrease

# Forms of uncertainty

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

# Incorporating uncertainty

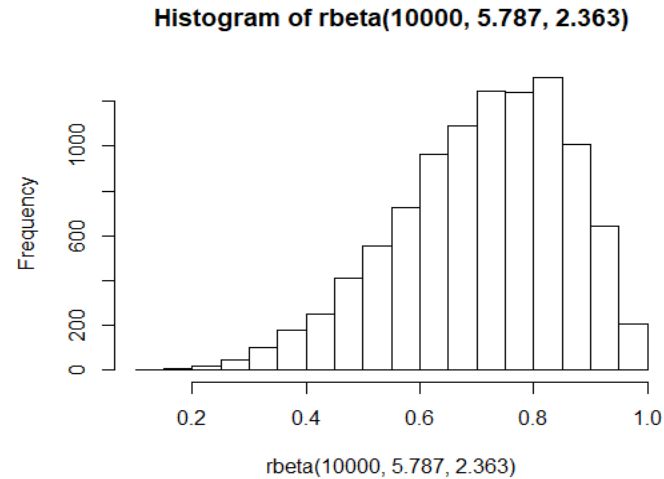
- Use statistical distributions and functional relationships
  - Environmental variation/stochasticity
  - Demographic stochasticity
  - Ecological/structural uncertainty
    - Density dependence
  - Parametric uncertainty

# Environmental stochasticity

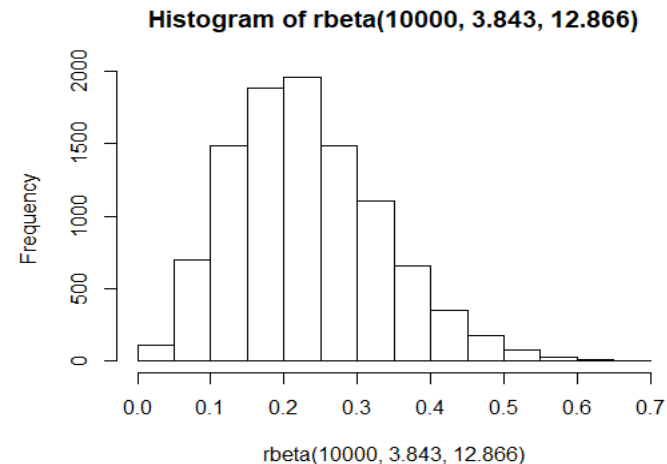
- Survival parameters typically drawn from a beta distribution
  - Continuous but restricted between 0 and 1
  - Very flexible
- Fecundity parameters have two typical methods
  - Log-normal, bounded by 0 and infinity
  - Poisson distribution summed over all the individuals in the population

# Survival rate distribution

- $S = 0.71$ , S.D. = 0.15 →
  - Beta1 = 5.787 Beta2 = 2.363 →

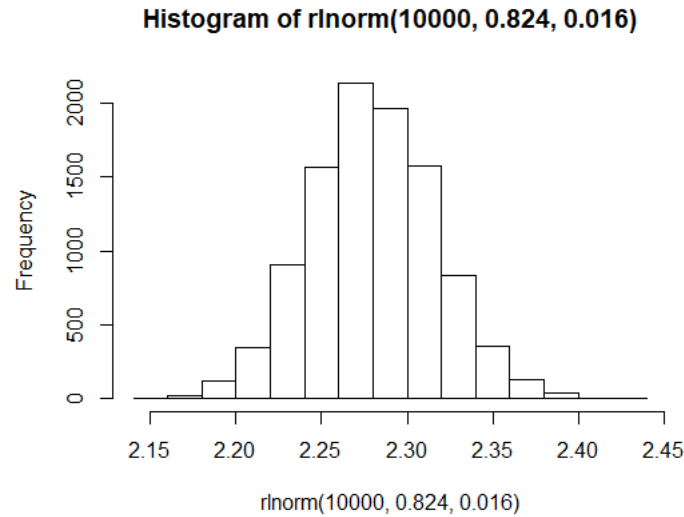


- $S = 0.23$ , S.D. = 0.1 →
  - Beta1 = 3.843 Beta2 = 12.866 →

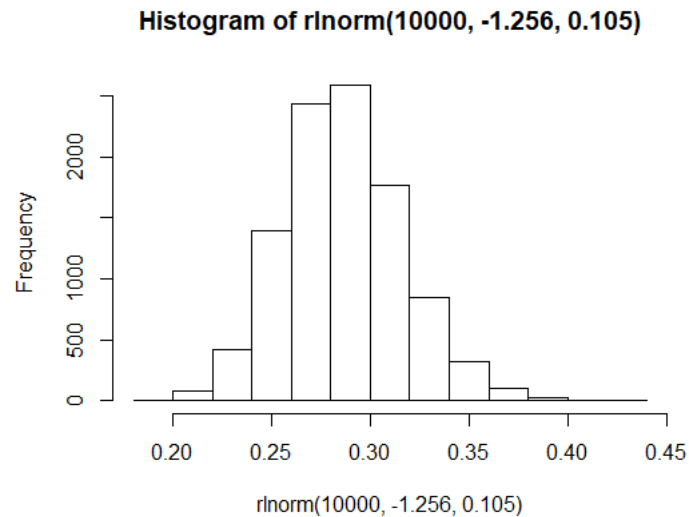


# Fecundity distribution

- $F = 2.3$ , S.D. = 0.3 →
  - $s1 = 0.824$ ,  $s2 = 0.016$  →



- $F = 0.3$ , S.D. = 0.1 →
  - $s1 = -1.256$ ,  $s2 = 0.105$  →



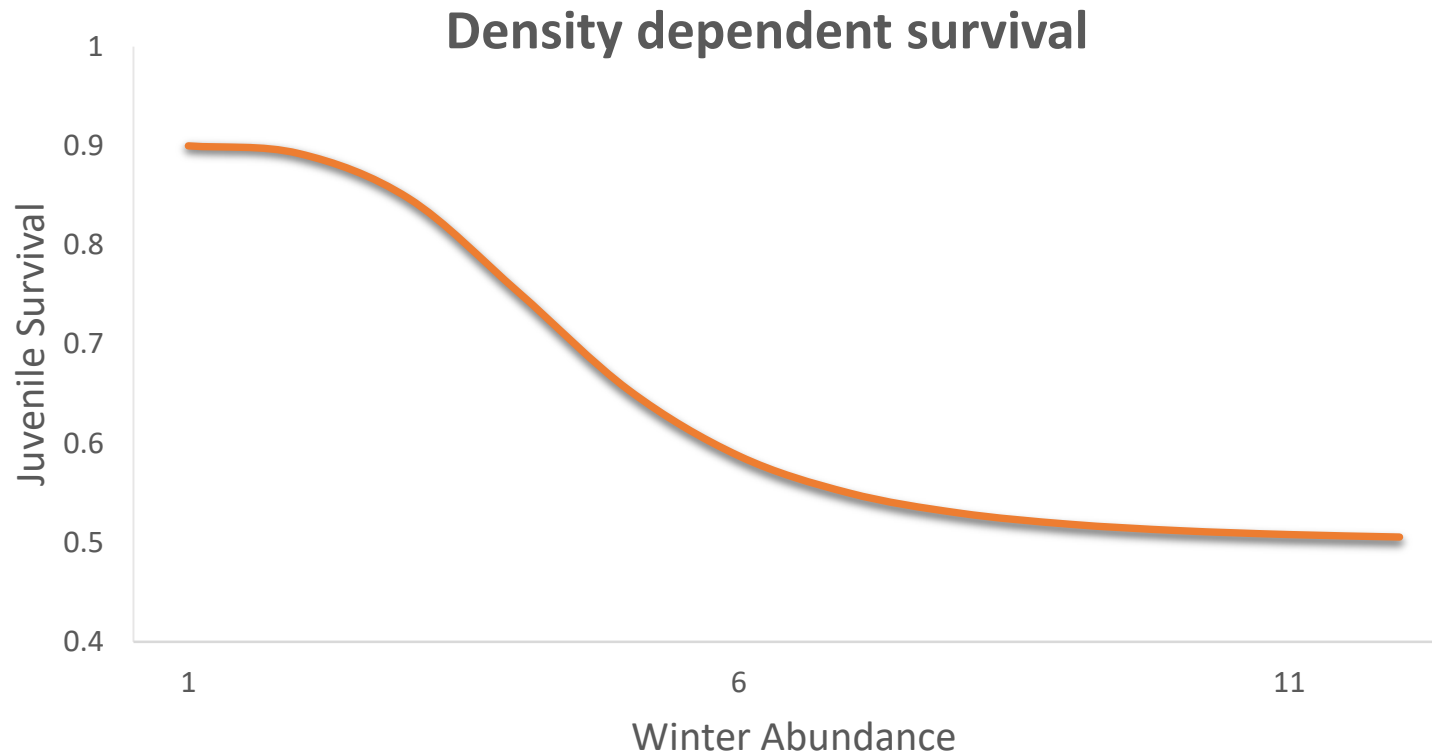
# Demographic stochasticity

- Animals live or die as whole animals - not as fractions
  - Without demographic stochasticity mean survival probability = 0.8
    - $6 \text{ individuals} * 0.8 = 4.8 \text{ individuals at the next time step}$
  - With demographic stochasticity
    - Model survival using a Binomial distribution
      - Computer picks 6 random numbers: 0.32, 0.89, 0.81, 0.11, 0.94, 0.70
      - 3 out of the 6 individuals die because their random picks were less than the mean of 0.8



# Ecological or structural uncertainty

- Density dependence
  - Model parameters are a function of population density



# Modeling density dependence

## Threshold Density dependence

- If the population exceeds a ceiling threshold, fecundity is equal to zero
  - if  $(n[i,j] > n_{crit}) F[i,j] = 0$
- As the population approaches some ceiling threshold, fecundity gets smaller and smaller
  - $F[i,j] = F[i,j] * (1 - n[i,j] / n_{crit})$

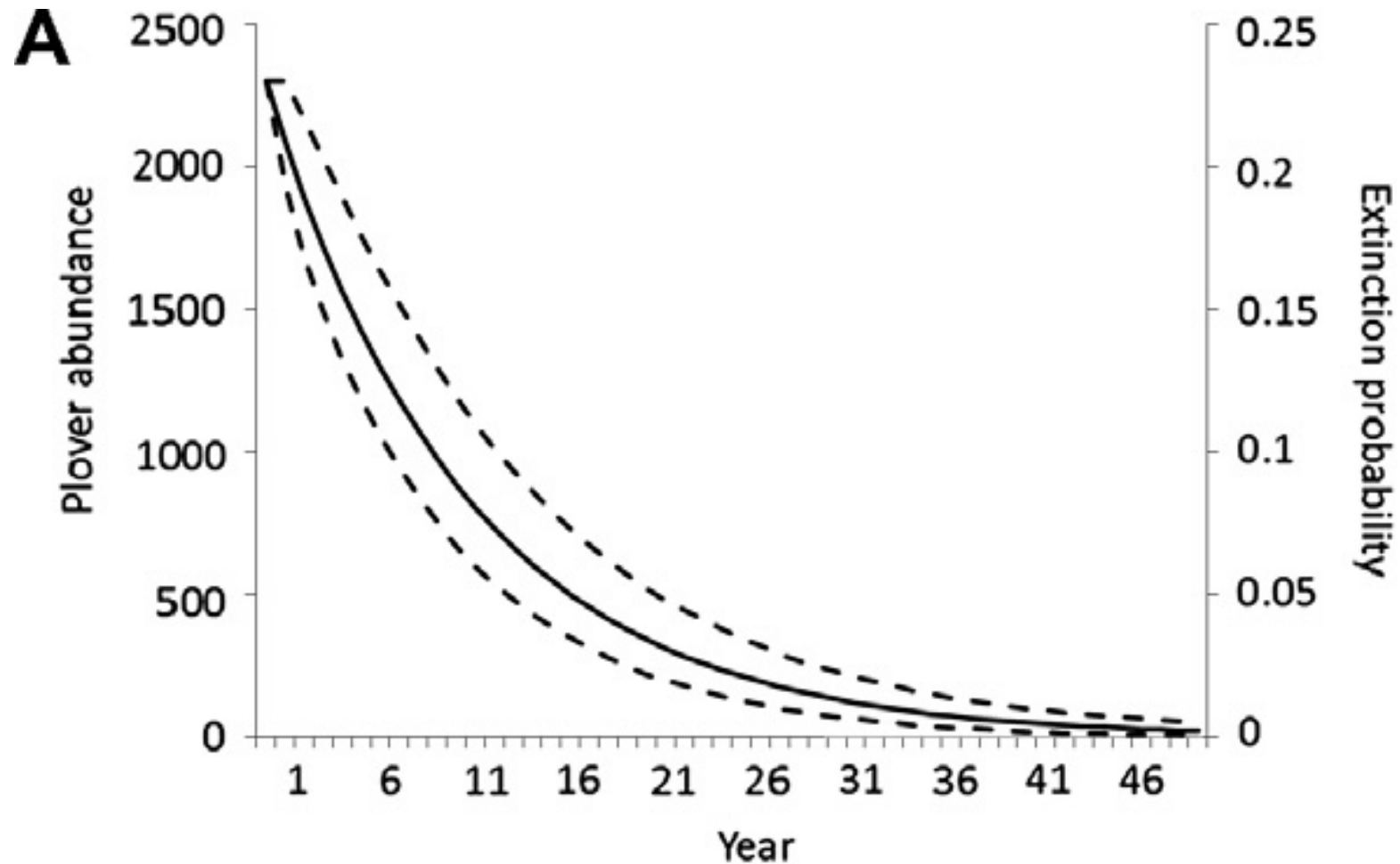
# Parametric uncertainty

- Parameters values are not precisely known
  - Variance or standard deviation estimates for parameters estimated over years conflate environmental variation with sampling variance
    - Sampling variance is the result of only using a specific number of individuals or locations to study a phenomenon
    - Happens with every wildlife study because we can't study every individual in every location

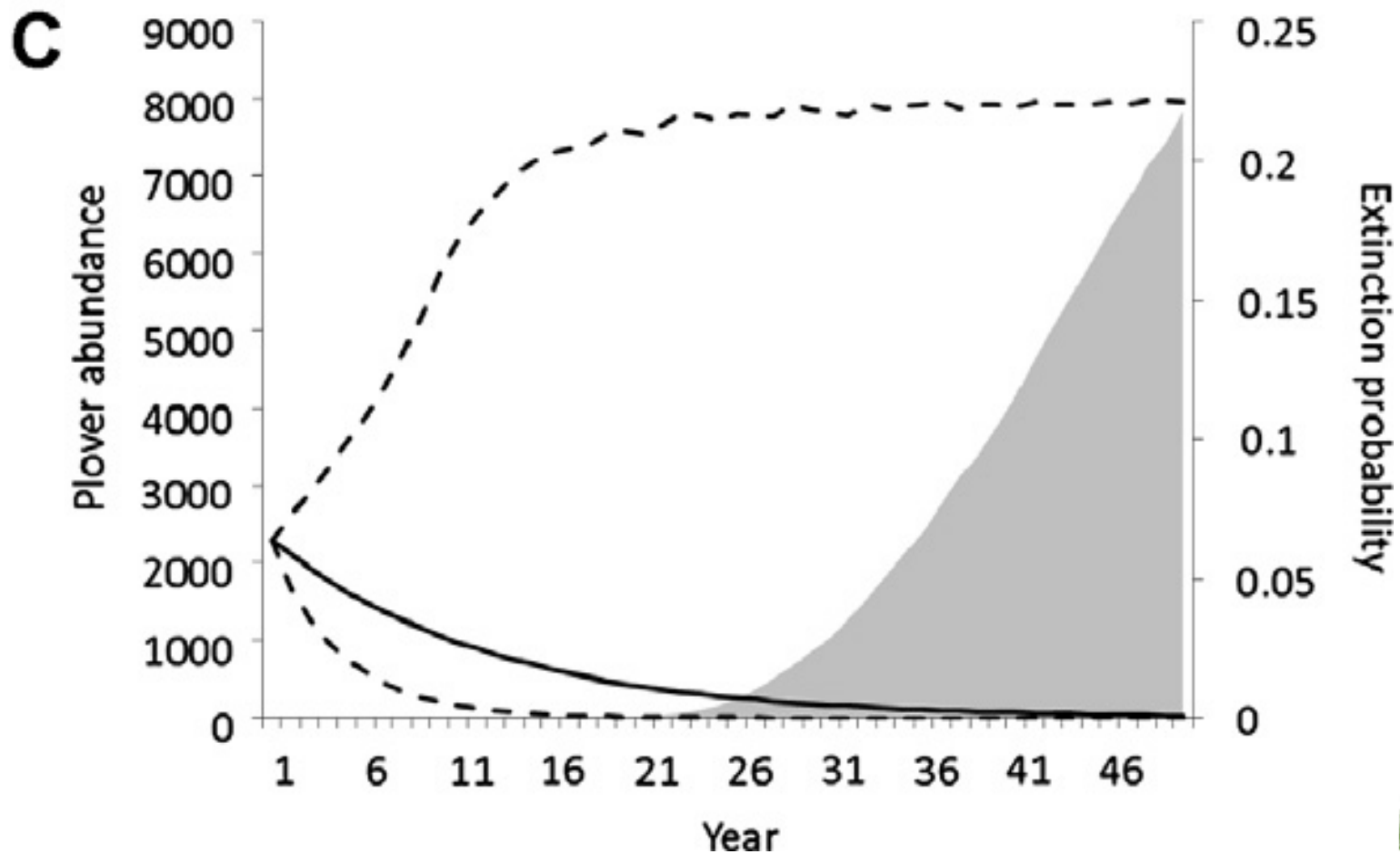
# Great, so what do we do about it?

- Most models will discard sampling variance
- Not always a good idea!
  - Pretending that we know system parameters with precision
    - Less variation in the model predictions
    - Could affect assessment of management choices

# Projection without sampling variance

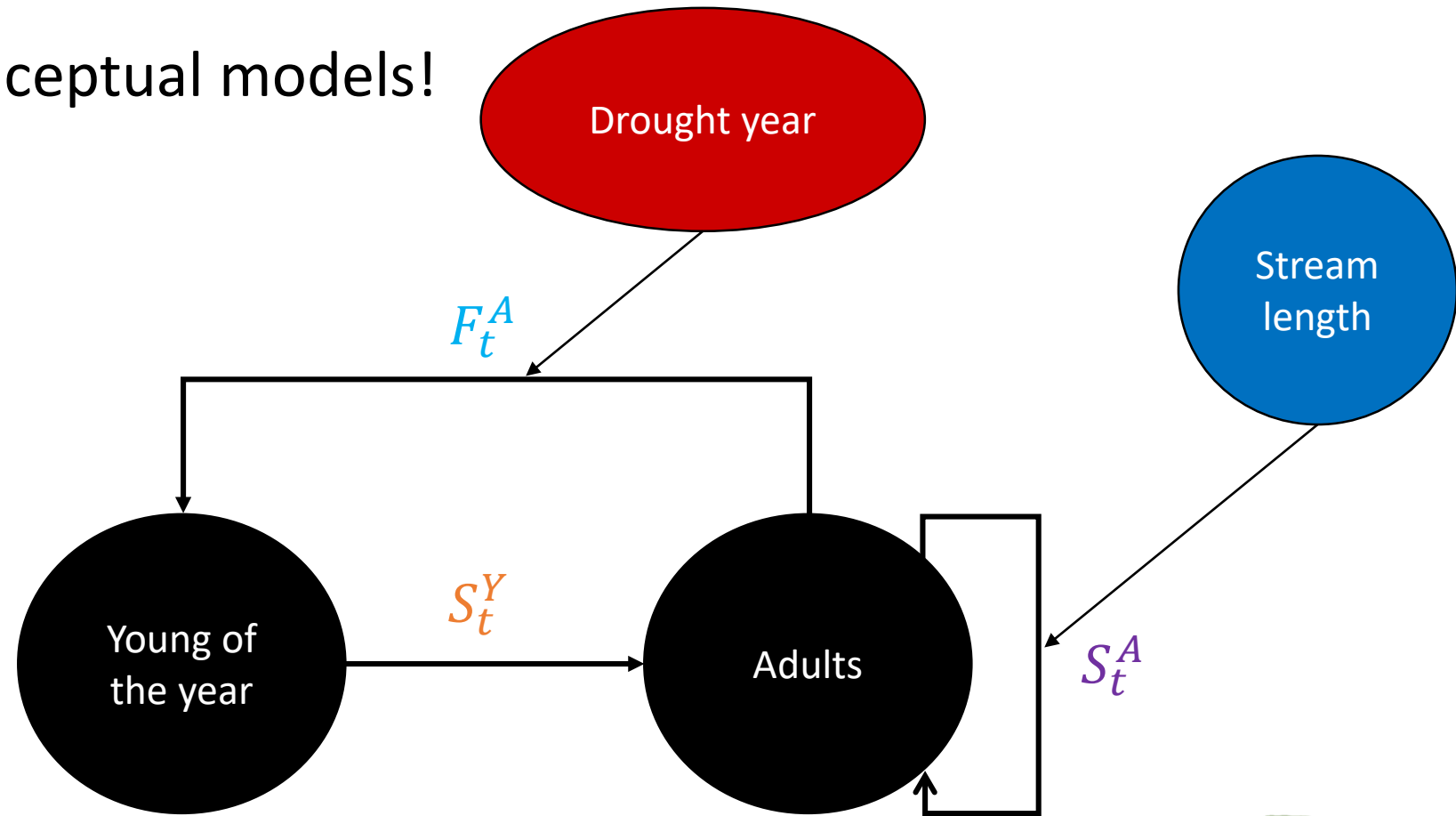


# Projection with sampling variance



# Modeling environmental effects

- Back to conceptual models!



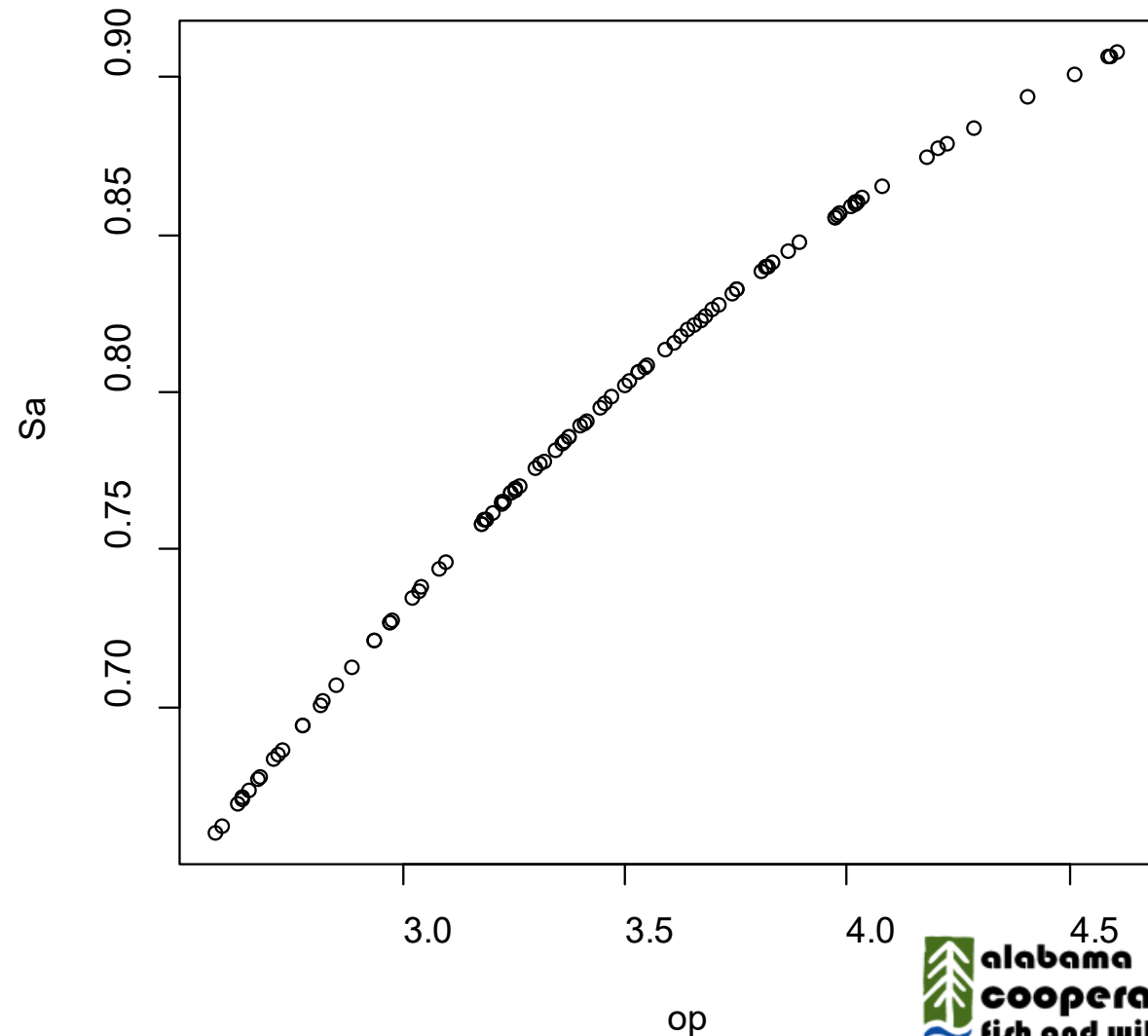
# Incorporating environmental covariates

- Conditionally linked events
  - Use **IF→THEN** statements to link a demographic parameter to some other randomized event
    - E.g., if a Bernoulli trial for drought returns a 1, then mean fecundity is 1.1 offspring per female, but if it returns a 0 then mean fecundity is 2.3 offspring per female



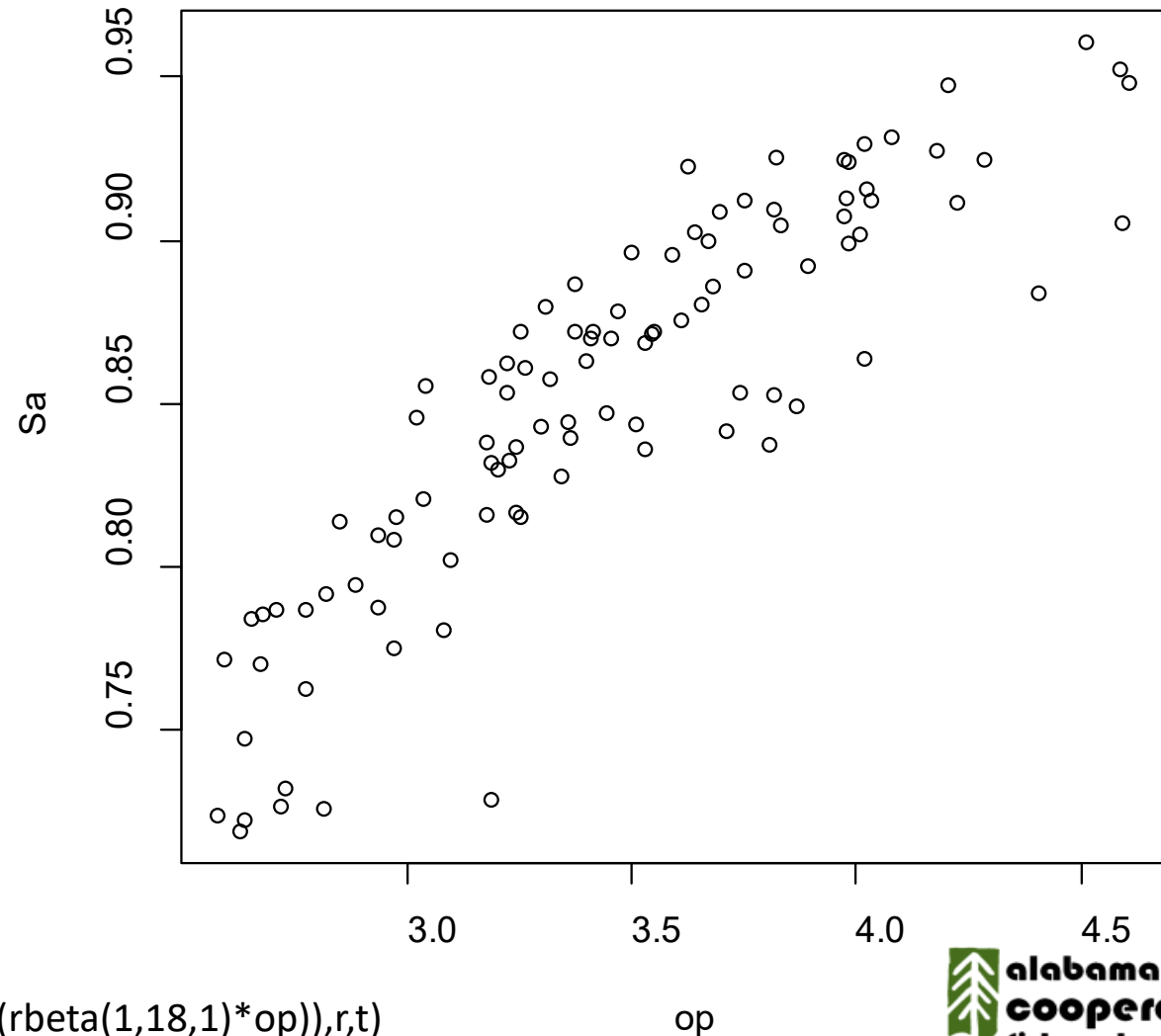
# Environmental covariates

- Adult survival ( $S_a$ ) can be a function of some other environmental parameter/variable
  - (“op” for other parameter, e.g., stream length)



# Environmental covariates

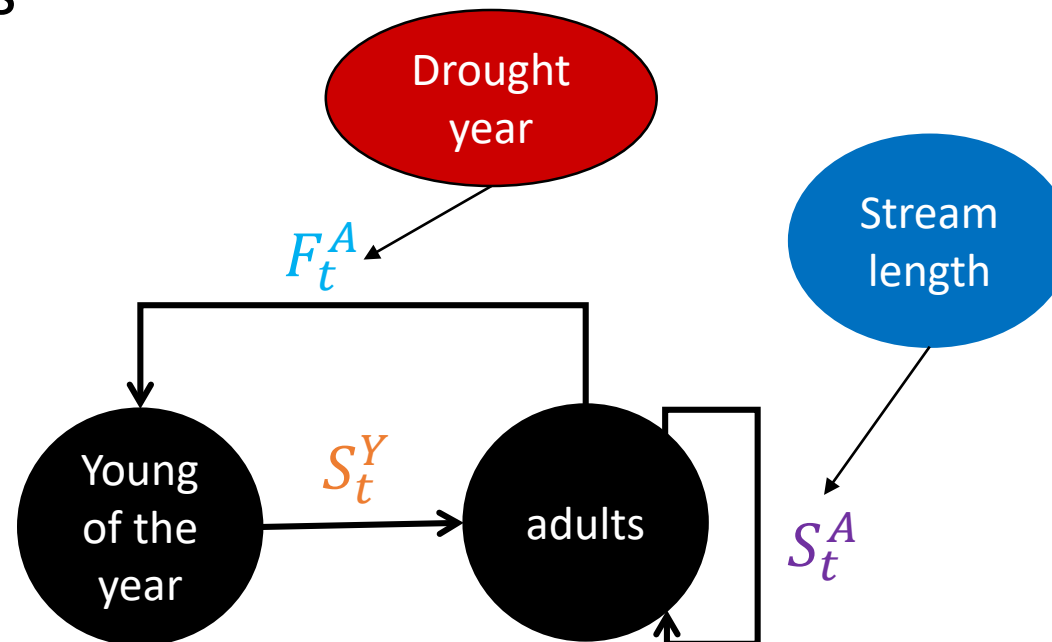
- Adult survival ( $S_a$ ) can be a function of “op” but it has variability



$S_a = \text{matrix}(\text{plogis}(-\text{rnorm}(1, 1.4, .1) + (\text{rbeta}(1, 18, 1) * \text{op})), r, t)$

# Inputting scenarios

- Use the conceptual model and sensitivity analysis to guide scenarios
  - i.e., what ecological factors affect the most sensitive parameters?
- Design scenarios to explore the expected range of future variation in important covariates



# Using this structure to build GLMs

- Generate lots of output values (abundance, P(extinction), etc.) with lots of corresponding input values
- Use a multi-variate GLM to assess the importance of each variable of interest:
  - $P(\text{extinction}) \sim b_1(\text{Initial } N) + b_2(\text{drought freq}) + b_3(\text{MaxPop}) \dots$
  - This is a binomial GLM
- What factors most effect the output metric of interest?

# Sonoran desert tortoise example



- MDR = mean drought rate
- NAI = Initial Number of adults
- MaxPop = habitat based maximum population size
  - You could input different values of MDR, NAI or MaxPop to predict the corresponding  $P(Qe100)$ , i.e., input alternative future scenarios
- $P(Qe100) = -5.602 + (18.42 \times MDR) - (5.363e^{-6} \times NAI) - (1.797e^{-6} \times MaxPop)$

# Review

- What metrics are available for assessing population resiliency using a demographic matrix model?
- What sources of uncertainty may make it difficult to predict population dynamics?
- How can we incorporate uncertainty in our population projections?

# Questions?

