# 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 stochasticity, demographic stochasticity, 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)
  - o 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

Data types	Den	nographic vital rate
Number of offspring per female	<b>→</b>	Fecundity
Nest/den success or failure	Breedi	ng success probability
Number of young returning next year	Recr	uitment probability
Radio telemetry	Compined	. , , , , , , , , , , , , , , , , , , ,
Individual mark-recapture/resight	Survivai	probability (seasonal or annual)
Annual plot census (plants)	Transiti	on probability (annual)
Growth rate		1 7 7





### 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

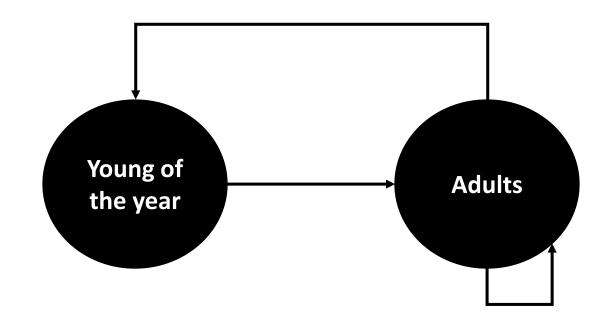
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- 3. Use state variables and state-specific vital rate estimates to create a conceptual model and matrix model





# Conceptual model

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



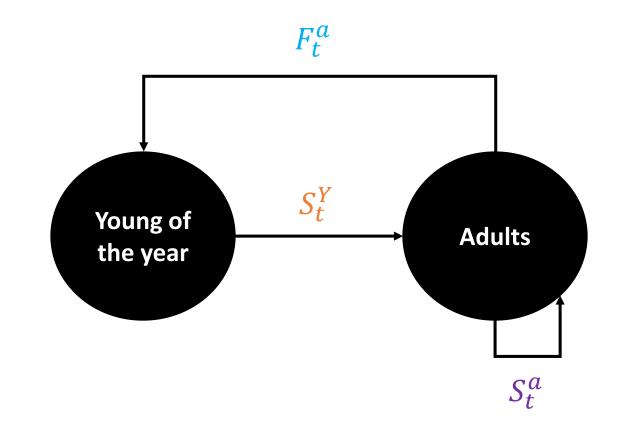






# Conceptual model

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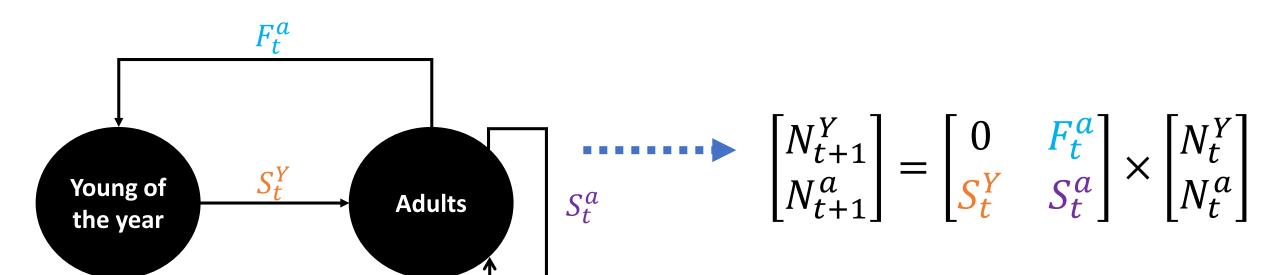






### Create matrix model

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

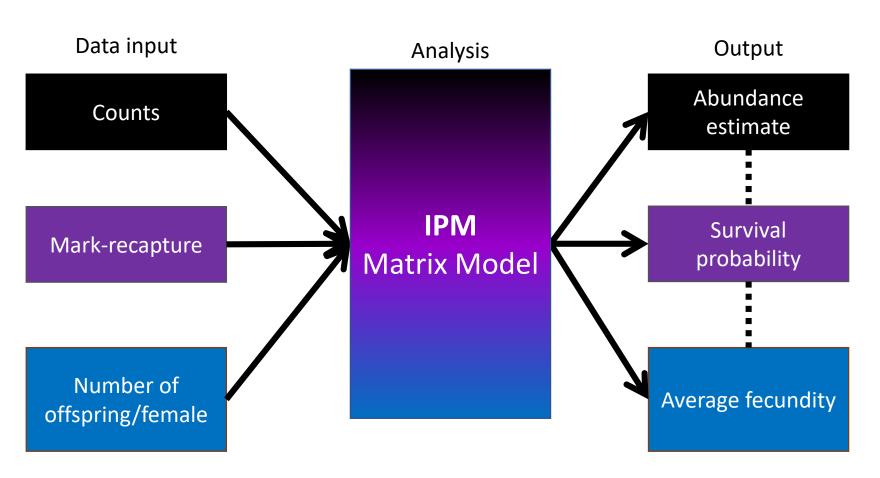






### Integrated population models

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



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





### Matrix projections

$$\begin{bmatrix} N_{t+1}^{Y} \\ N_{t+1}^{a} \end{bmatrix} \neq \begin{bmatrix} 0 & F_{t}^{a} \\ S_{t}^{Y} & S_{t}^{a} \end{bmatrix} \times \begin{bmatrix} N_{t}^{Y} \\ N_{t}^{a} \end{bmatrix}$$

$$* N_{t} \quad \longleftrightarrow \quad N_{t+1} \neq \lambda * N_{t}$$





### Matrix projection outputs

- Abundance over time
- Population growth rate
  - $\circ$  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
  - O 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
  - o Poisson distribution summed over all the individuals in the population

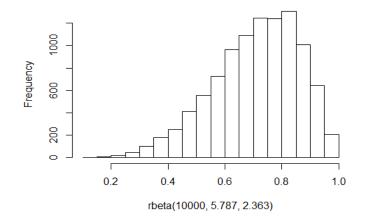


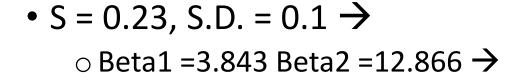
### Survival rate distribution

#### Histogram of rbeta(10000, 5.787, 2.363)

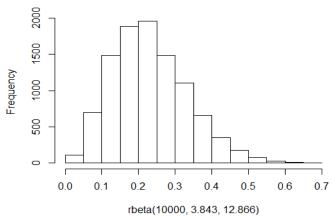
• S = 0.71, S.D. = 0.15 
$$\rightarrow$$

• Beta1 = 5.787 Beta2 = 2.363  $\rightarrow$ 





#### Histogram of rbeta(10000, 3.843, 12.866)

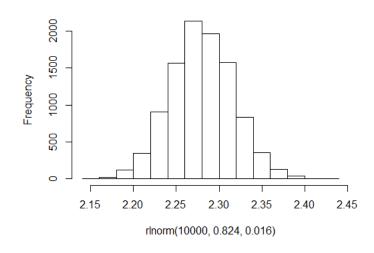




## Fecundity distribution

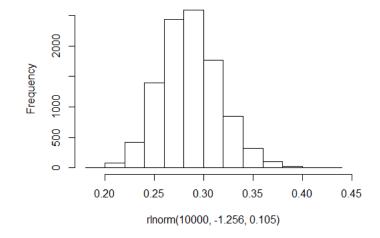
• F = 2.3, S.D. = 
$$0.3 \rightarrow$$
  
• s1= 0.824, s2 = 0.016  $\rightarrow$ 

#### Histogram of rlnorm(10000, 0.824, 0.016)



#### Histogram of rlnorm(10000, -1.256, 0.105)

• F = 0.3, S.D. = 0.1 
$$\rightarrow$$
  
• s1= -1.256, s2 = 0.105  $\rightarrow$ 









### Demographic stochasticity

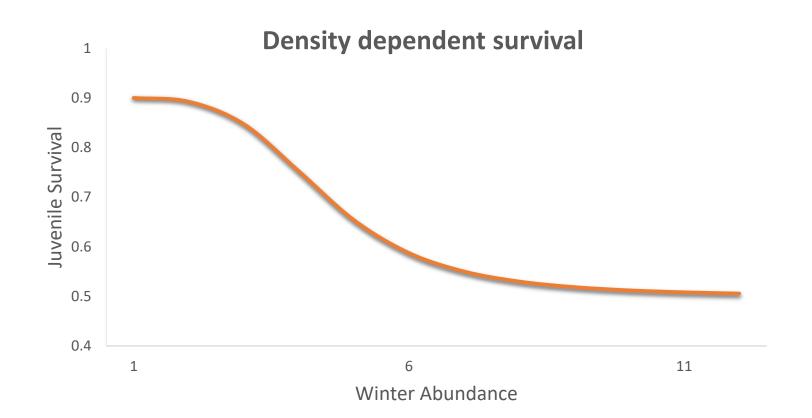
- Animals live or die as whole animals not as fractions
  - Without demographic stochasticity mean survival probability = 0.8
    - 6 individuals \* 0.8 = 4.8 individuals at the next time step
  - With demographic stochasticity
    - Model survival using a Binomial distribution
      - Computer picks 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] > ncrit) 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]/ncrit)





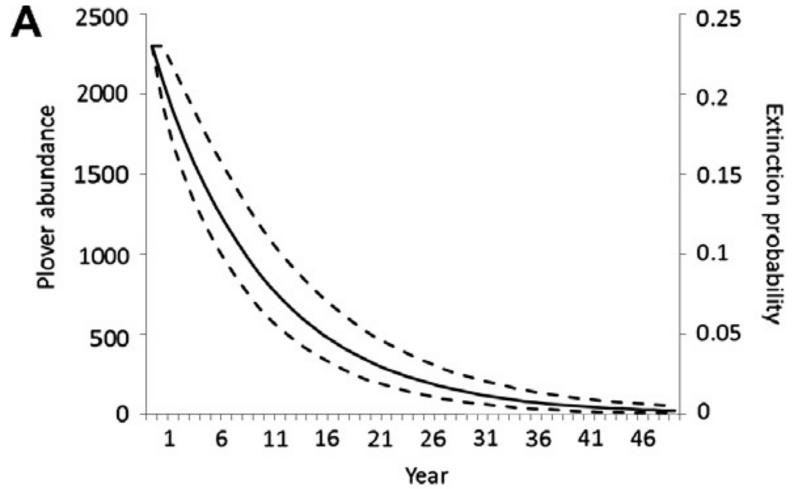
### Parametric uncertainty

- 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





# Projection without sampling variance







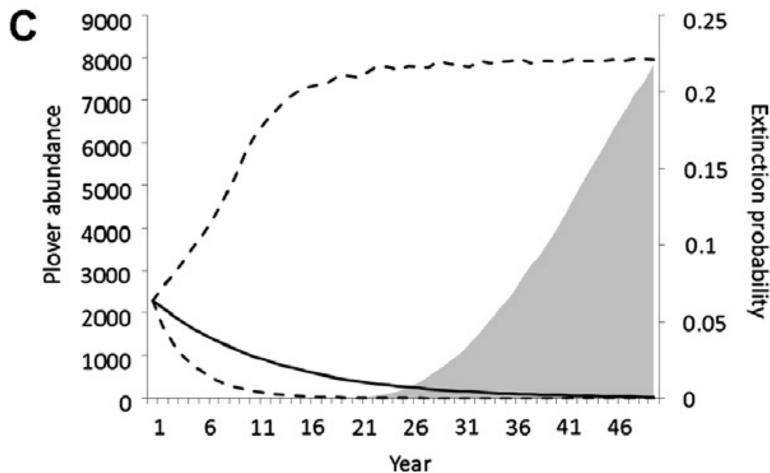
### Great, so what do we do about it?

- Modeled by selecting a random value for the mean and variance from a distribution that represents potential parameter values
- The sampled mean and variance are used to generate a new distribution for the parameter
- The vital rate or population growth rate is then randomly selected from the new distribution





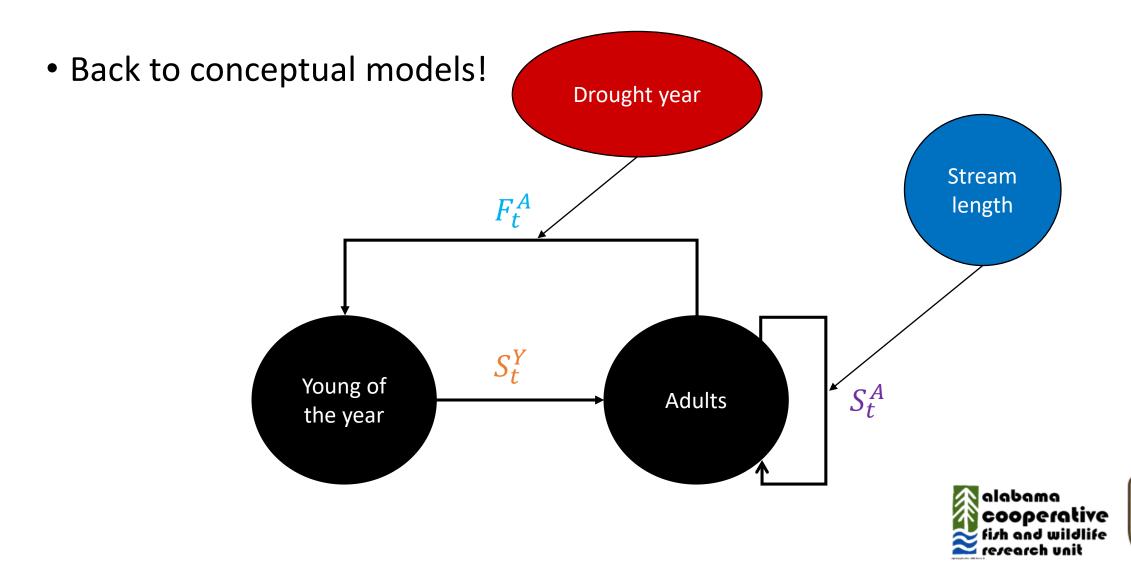
# Projection with sampling variance





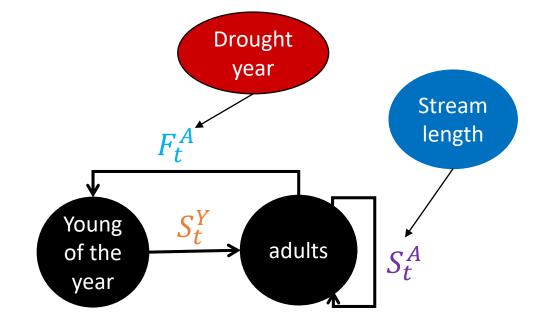


# Modeling environmental effects



### Incorporating environmental covariates

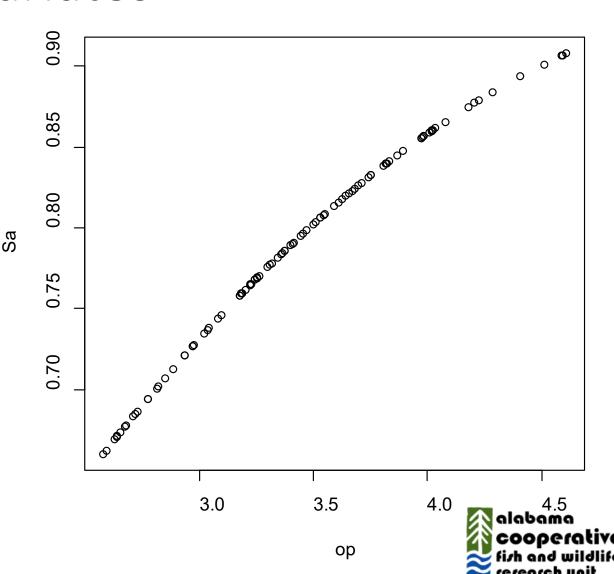
- Conditionally linked events
  - Use IF→THEN statements to link a demographic parameter to some other randomized event
    - E.g., if a Bernoulli trial for drought returns a 1, then mean fecundity is 1.1 offspring per female, but if it returns a 0 then mean fecundity is 2.3 offspring per female





### Environmental covariates

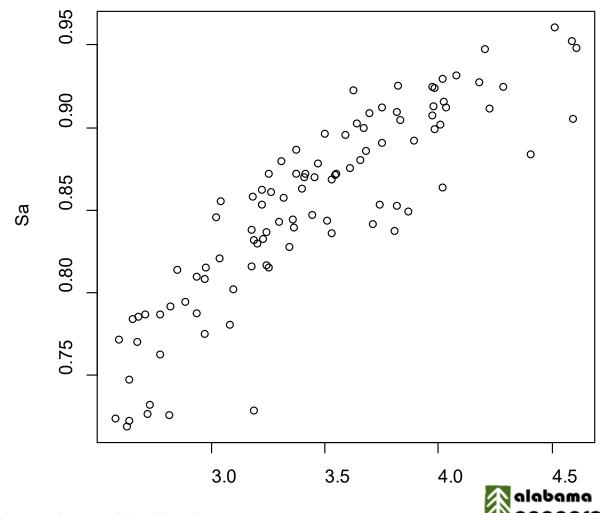
- Adult survival (Sa) can be a function of some other environmental parameter/variable
  - ("op" for other parameter,e.g., stream length)





### Environmental covariates

 Adult survival (Sa) can be a function of "op" but it has variability



op

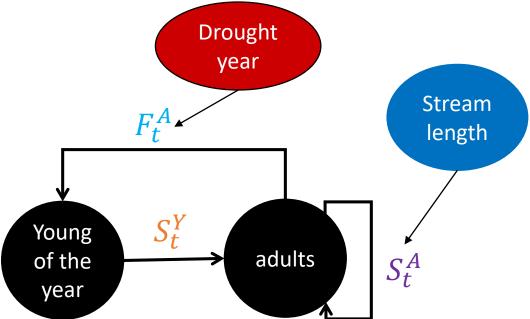


### Inputting scenarios

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:
  - $\circ P(extinction) \sim b_1(Initial\ N) + b_2(drought\ freq) + b_3(MaxPop) \dots$ 
    - This is a binomial GLM

• Determine which 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
- $P(Qe100) = -5.602 + (18.42 \times MDR) (5.363e 6x NAI) (1.797e 6x MaxPop)$







### 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?

