

## SSA 200

### Strategic Use of Data

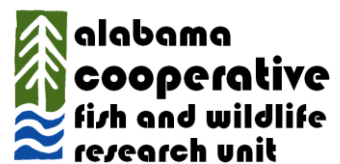
Lecture slides, activities, and additional supplementary materials are available online at:  
**[ssa200.auburn.edu](http://ssa200.auburn.edu)**

What is a model?

The purpose of modeling

- Statistical analysis of data
- Use statistical analysis to predict the future
- Explaining variation
- Using data analysis to understand ecological processes
- Predict patterns in the future
- Evaluate competing hypothesis about how the system works

*Updated Dec 2019*



## Statistical distributions

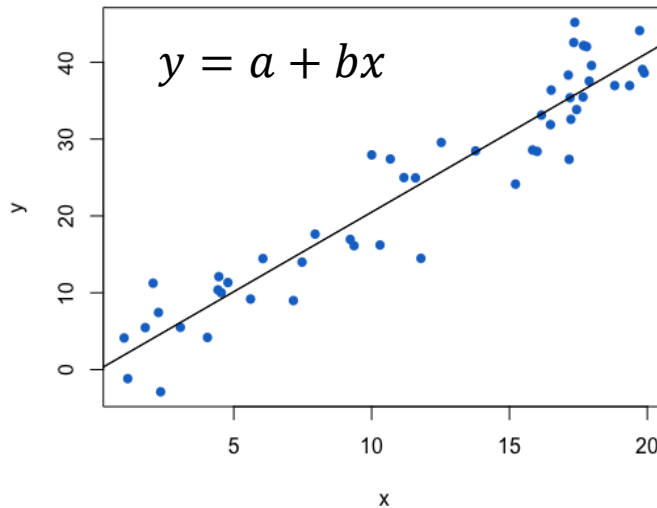
Name	Continuous or Discrete	Bounds	Common applications	Shape	Notes
Normal	Continuous	$-\infty, \infty$	Linear regression		
Binomial	Discrete	0 or 1	Occupancy Survival		
Multinomial	Discrete	0, $\infty$	State transitions		
Poisson	Discrete	0, $\infty$	Count data		
Negative Binomial	Discrete	0, $\infty$	Counts with many zeros		
Log-normal	Continuous	0, $\infty$	Population-level productivity (projections)		
Beta	Continuous	0, 1	Population rates (projections)		
Uniform	Continuous	User-defined	Variety of applications (projections)		

## Linear regression and AIC

**General** linear model – response variable ( $y$ ) has a Normal distribution

**Generalized** linear model – response variable ( $y$ ) has some other distribution

- **Logistic** regression – Binomial distribution
- **Poisson** regression – Poisson distribution



Parameter	Estimate	SE	$t$	p-value
Intercept	-0.22	1.29	-0.169	0.866
$b$	2.07	0.0978	21.16	< 0.0001

Model	AIC	$\Delta$ AIC	Np	$w_i$
Int + Covariate 1 + Covariate 3	345.8	0	3	0.82
Int + Covariate 1 + Covariate 2 + Covariate 3	349.1	3.1	4	0.18
Int	359.8	14.0	2	0
Int + Covariate 2	361.1	15.3	1	0

## Types of uncertainty

**Partial controllability** – We are unable to control the exact management actions taken in a system.

Examples:

- Setting management goals – we may intend to fully restore a habitat, but may not be able to implement the exact management goals due to other logistical constraints

**Observational uncertainty** – We are unable to perfectly observe the state of natural systems.

Examples:

- Count data – in almost all cases, we cannot count every individual present at a specific location, but instead assume there is some probability of detecting individuals

**Environmental variation** – Stochastic environmental fluctuations mean that conditions typically vary randomly from year to year.

Examples:

- Predicting effects of temperature – we may estimate a relationship between temperature and survival probability that we can use to predict survival under future temperature conditions, but temperature will likely vary in a stochastic way from year to year.

**Ecological uncertainty** – We have an imperfect understanding of how ecological systems work.

Examples:

- Metapopulation dynamics – we think a set of populations function as a metapopulation, but have not conducted studies to explicitly estimate immigration among sites, and therefore we are unsure to what extent immigration plays a role in measured population growth rate at each site.

## Some key terms

**Response/dependent variable** – in a statistical model, the variable that you are interested in better understanding or predicting (the “y” variable)

**Predictor/independent variable** – in a statistical model, the variable(s) that explain some of the observed variation in the response variable (the “x” variables)

**Covariate** – an environmental or ecological quantity that usually represents a stressor or species need and is included in a model as a predictor variable

**Parameter** – statistical quantities that are estimated to explain the relationship between predictor and response variables. Can also be used to refer to demographic vital rates of interest

**Collinearity** – occurs when two predictor variables in the same model are correlated with each other

**Overfitting** – occurs when too many predictor variables are included in the model, resulting in a model that is not very useful for prediction

**AIC** – stands for Akaike’s Information Criterion – a metric used to rank models based on how well they fit the data with a penalty for the number of covariates in the model (to avoid overfitting)

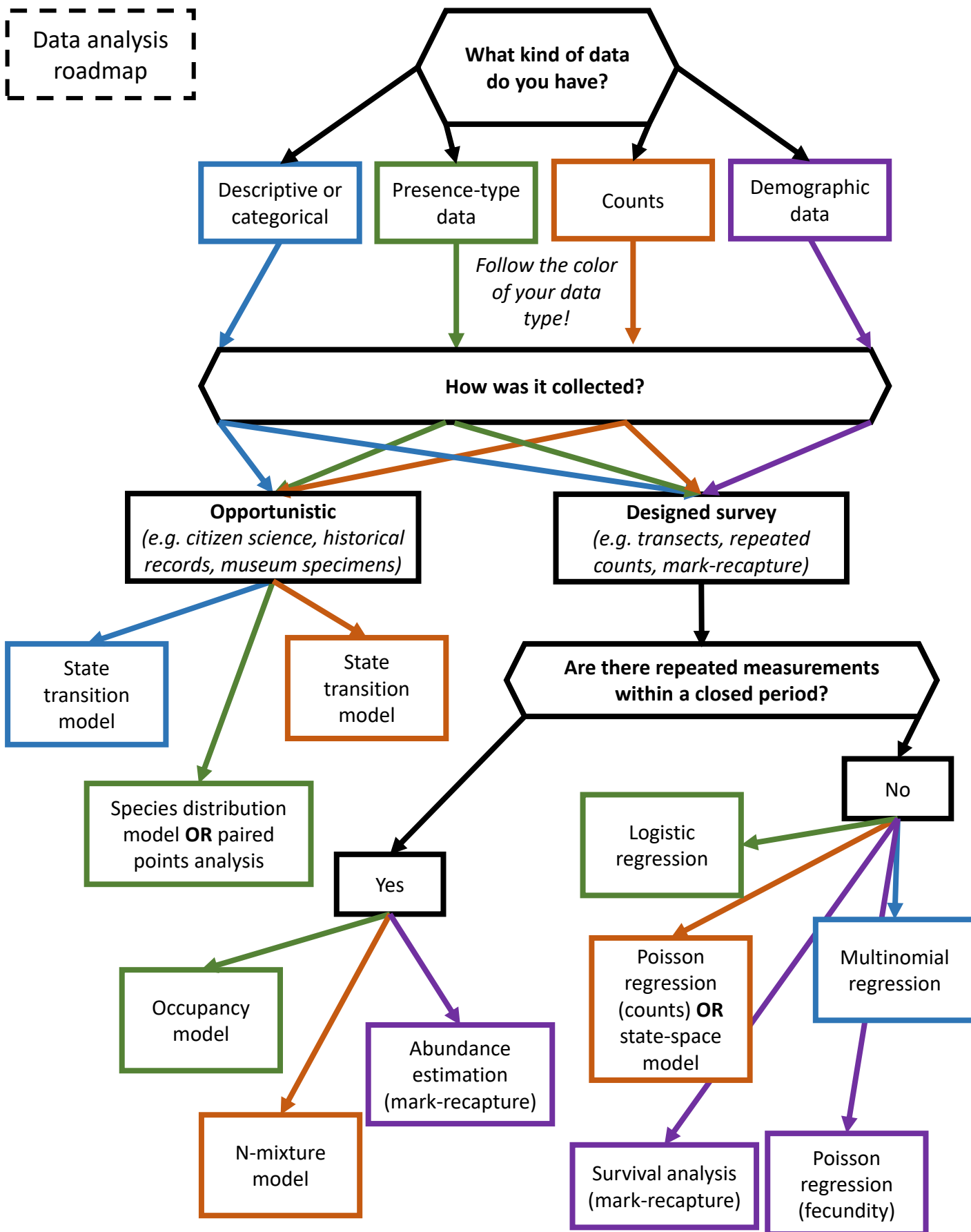
**Intercept** – the theoretical value of the response variable if all predictors were equal to zero

**Null model** – the “intercept-only” model that does not include any covariates

**Global model** – the most complex model in the model set that includes all covariates

**Population closure** – an important concept for occupancy and abundance estimation, a population is considered “closed” when there are no births, deaths, immigration, or emigration

Data analysis  
roadmap



*This roadmap is to serve as a general guide and is not an exhaustive list of all analysis options.  
Also, **always check the specific assumptions of your planned modeling approach!***