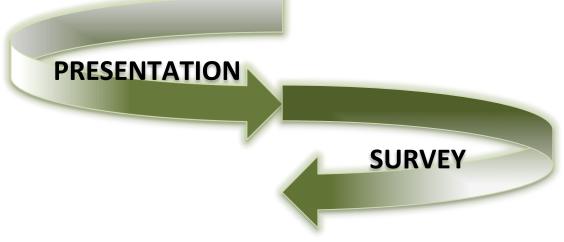
The Key Concepts and Steps in Data Science

Engin A. Sungur Statistics Discipline

University of Minnesota, Morris

OUTLINE



- INTRODUCTIONS & BACKGROUND INFORMATION
- STEPS/STAGES OF DATA SCIENCE/STATISTICS
 - QUESTION/PROBLEM
 - O DATA COLLECTION
 - DATA MANIPULATIONS
 - O EXPLORATORY DATA ANALYSIS
 - COMFIRMATORY DATA ANALYSIS
 - COMMUNICATING THE FINDINGS
 - FORMULATING NEW QUESTIONS/PROBLEMS
- GENERAL REMARKS



LEARNING OBJECTIVES

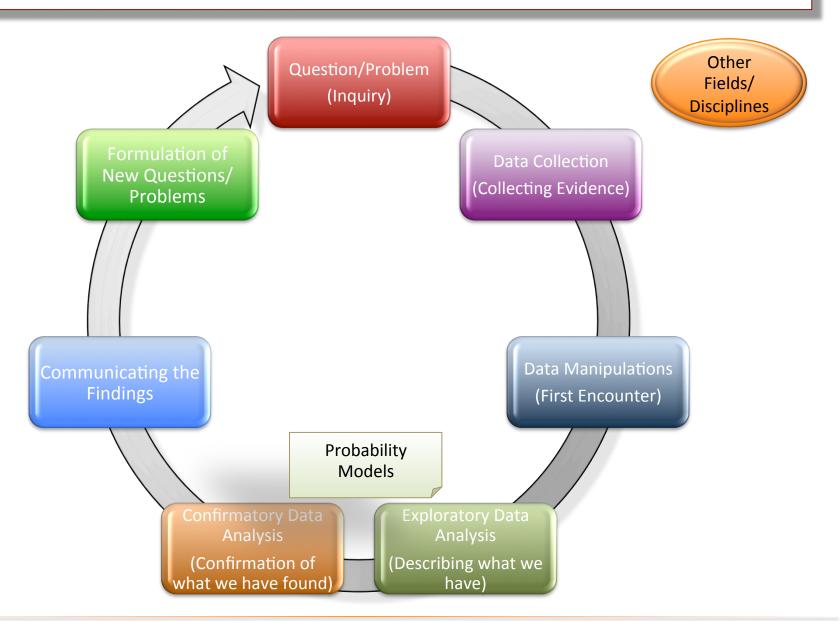


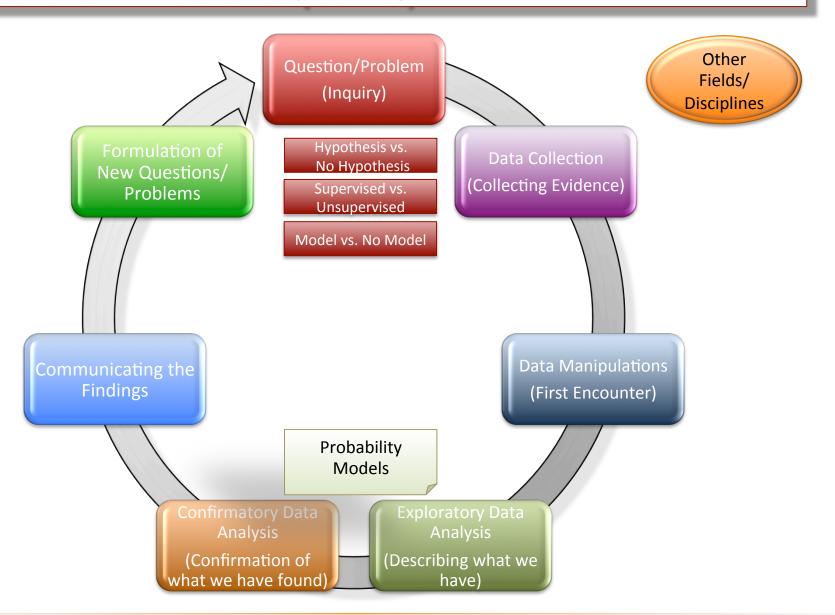


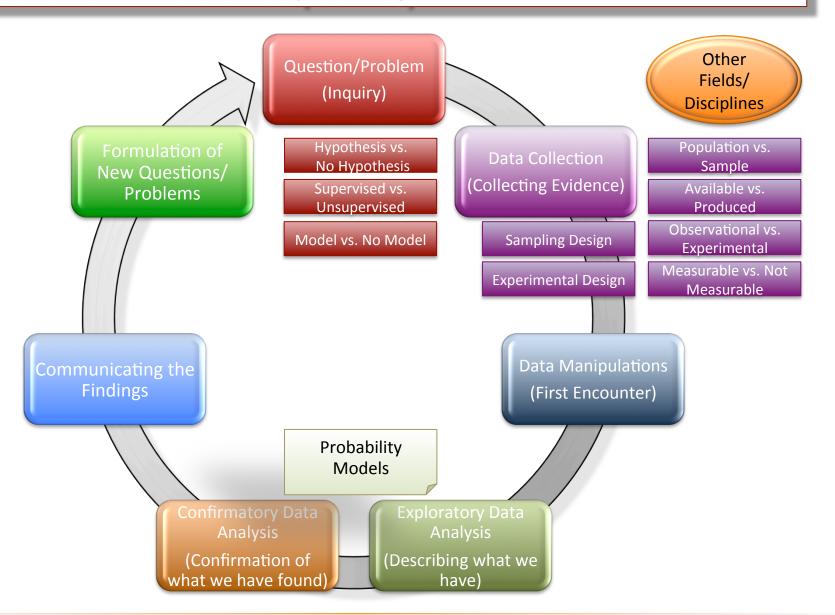
- IDENTIFY THE SEVEN STAGES OF THE DATA SCIENCE (you)
- LEARN COMMON CONCEPTS IN EACH (you)
- GET FAMILIAR WITH SOME SATATISTICAL TECHNIQUES/ METHODS/TOOLS THAT ARE AVAILABLE (you)

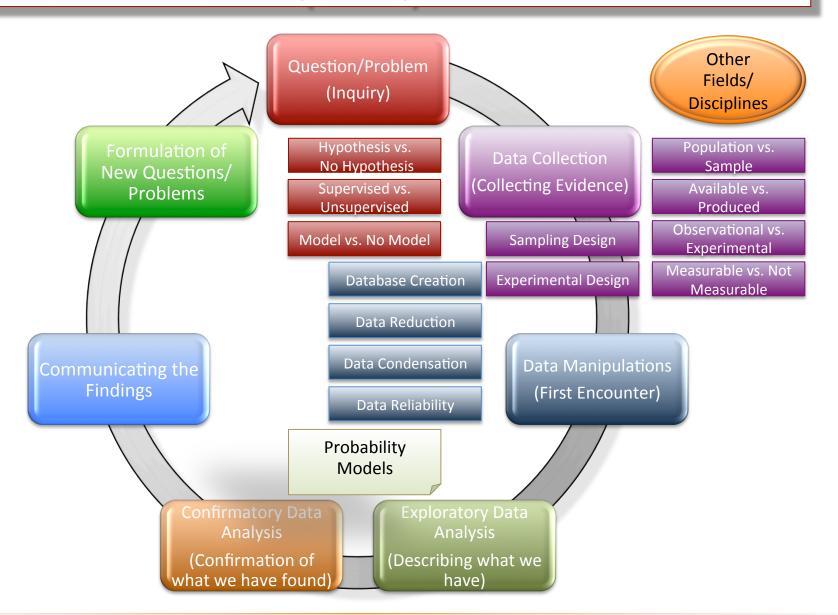
- LEARN ABOUT THE STATISTICAL
 ASPECT OF THE RESEARCH
 PROJECTS (me)
 - FORMALIZE THE TYPE OF STATISTICAL NEED OF YOUR PROJECTS (me)
- GET READY FOR YOUR QUESTIONS (me)

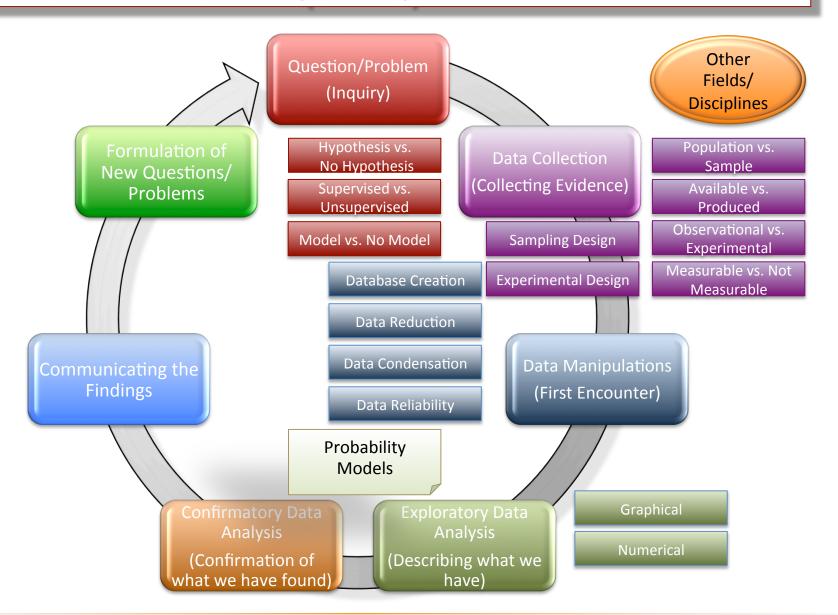
STAGES OF DATA SCIENCE

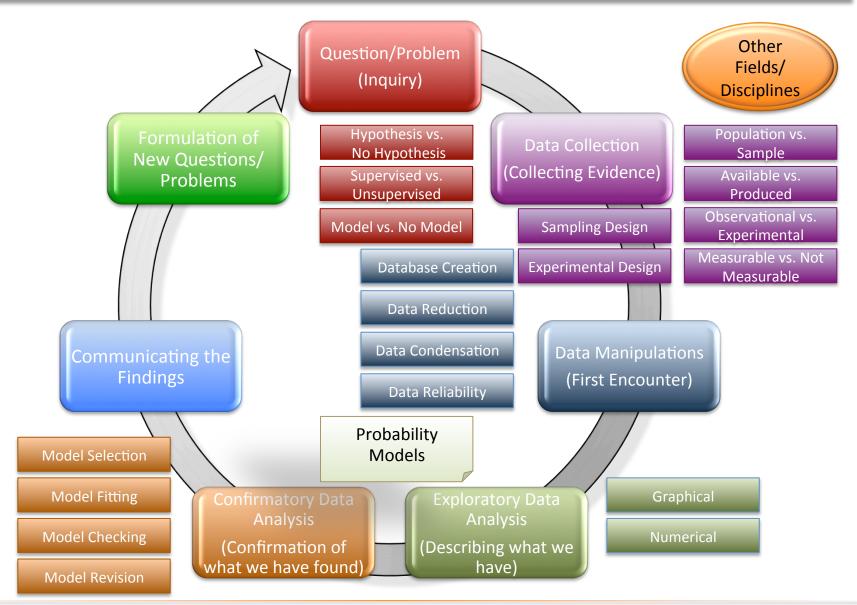


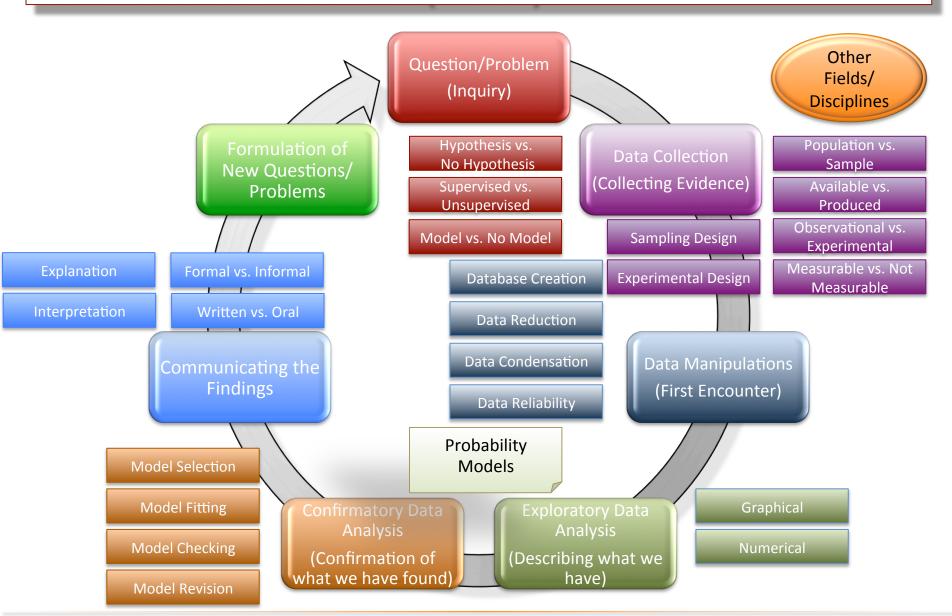


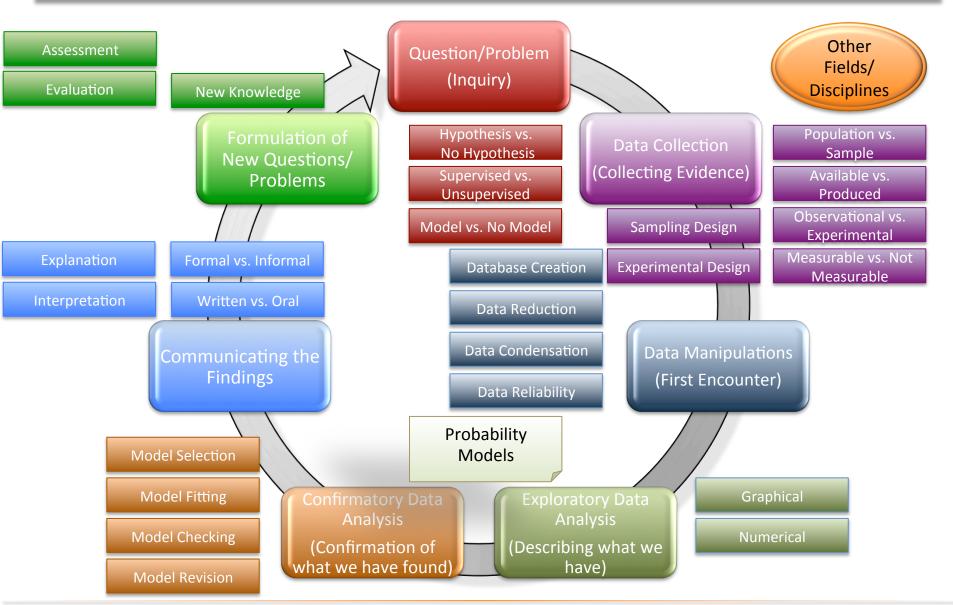




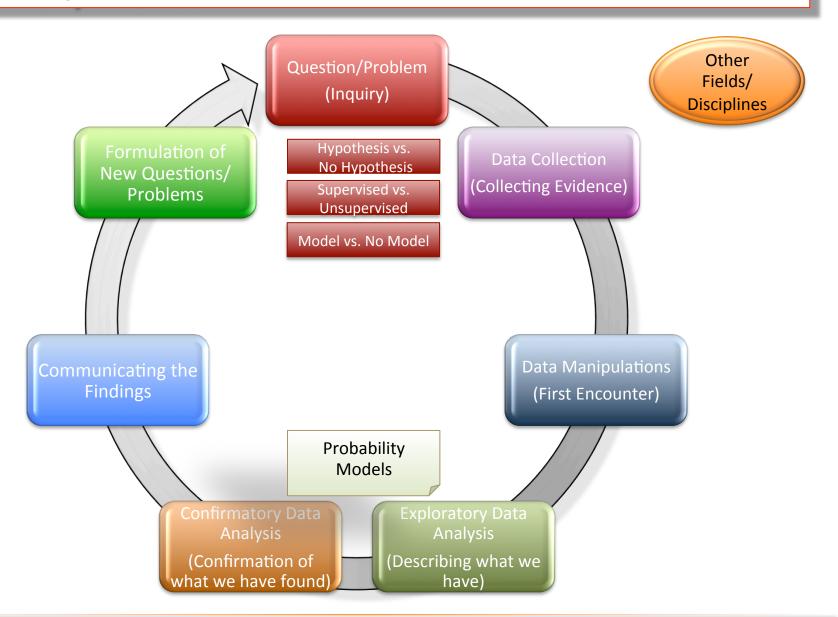








QUESTION/PROBLEM



QUESTION/PROBLEM

EXPLORATORY/DESCRIPTIVE MULTIVARIATE ANALYSIS

DATA MINING

CONFIRMATORY/INFERENTIAL MULTIVARIATE ANALYSIS

DATA CRAFTING

DATA EXPLORATION PATTERN RECOGNITION

STATISTICAL INFERENCE

LOOKING FOR PATTERNS

EXPLORING RELATIONSHIPS

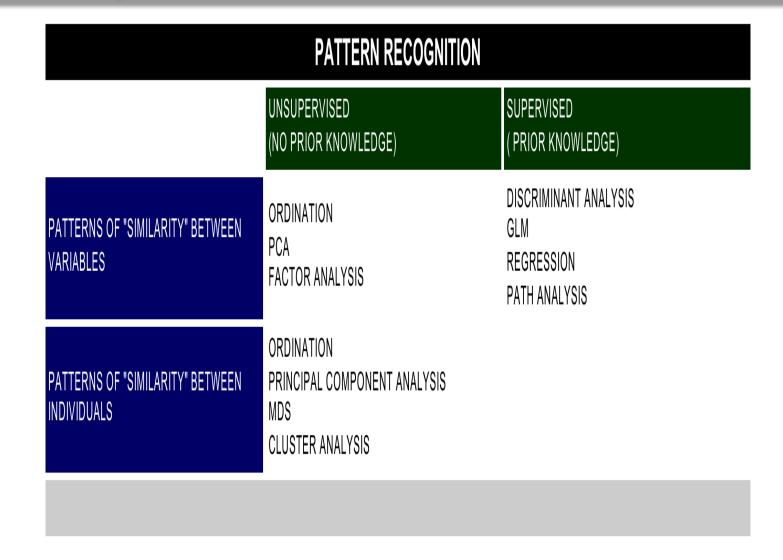
TEST HYPOTHESES

FIT & TEST THE MODELS

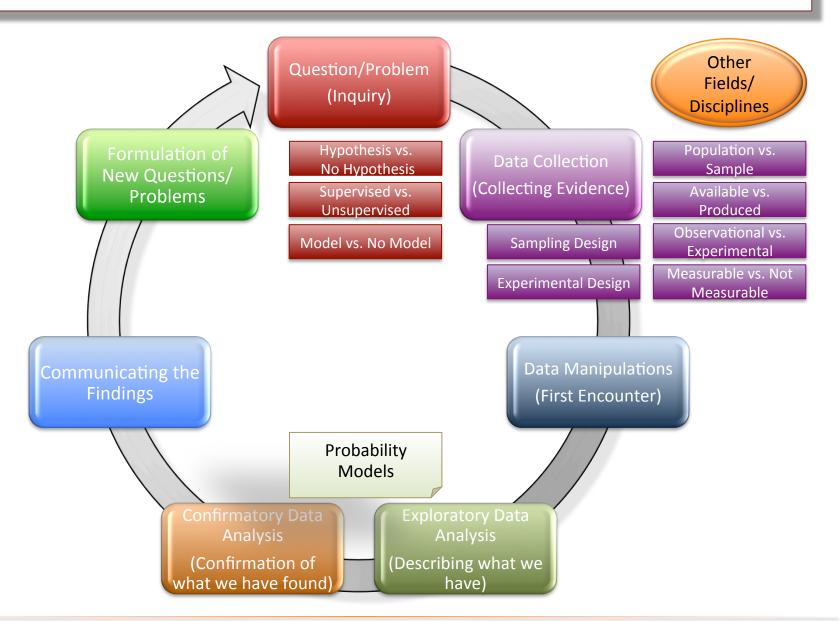
FORM HYPOTHESES

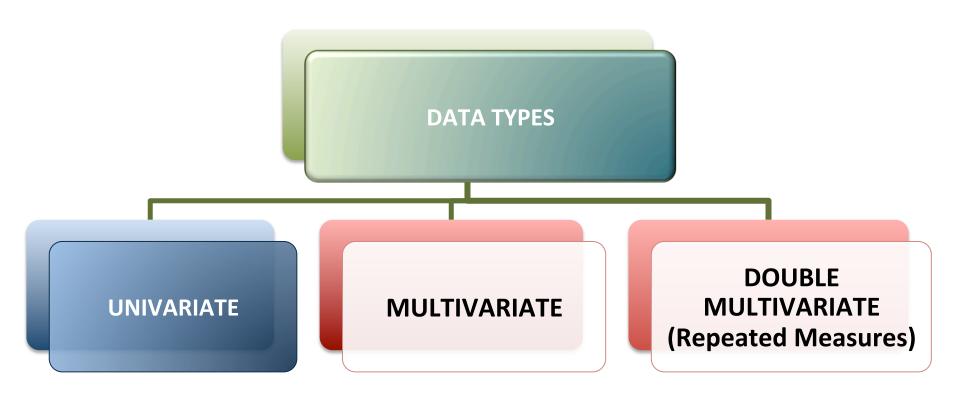
SELECT MODELS

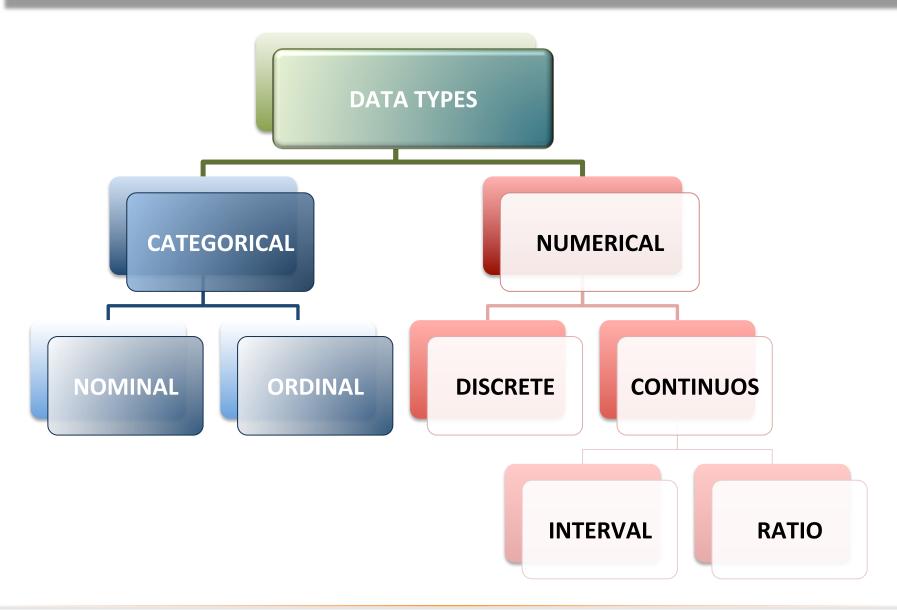
QUESTION/PROBLEM

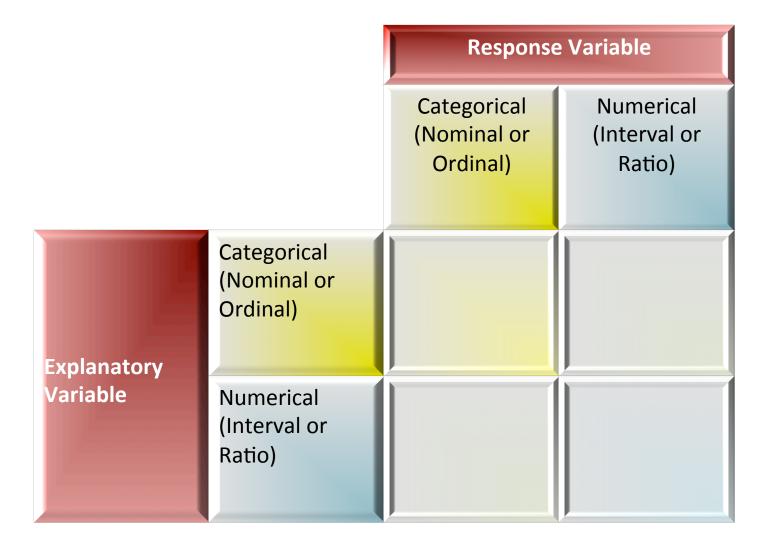


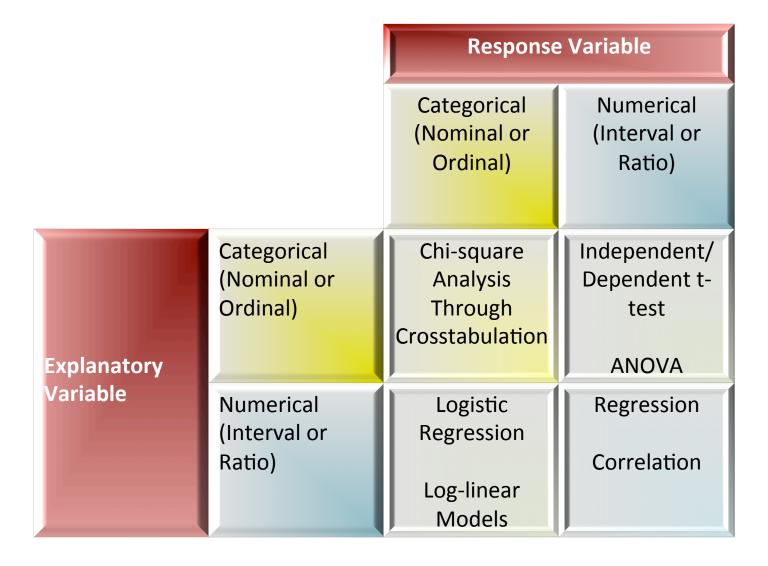
DATA COLLECTION







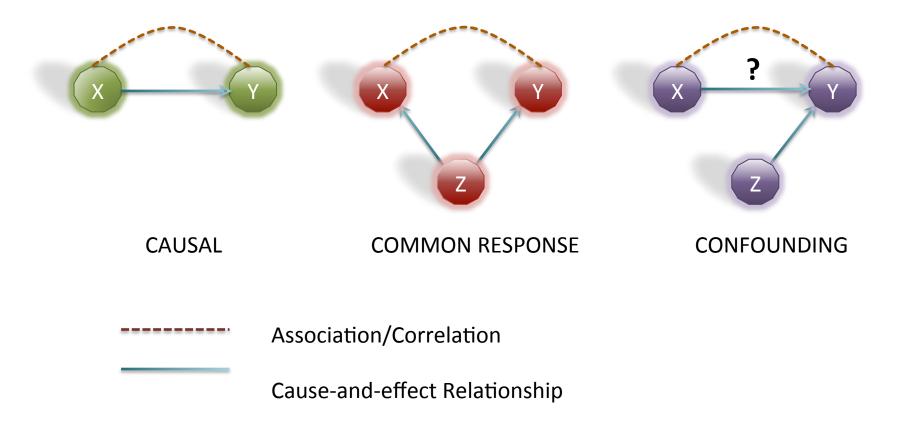




DATA COLLECTION: TYPES OF RELATIONSHIPS

Association/Correlation does not imply Causation Dependence does not imply Causation

(but it sure is a hint Lynd & Stevenson (2007), Tufte (2006), von Eye & DeShon (2011)).



DATA COLLECTION: DESIGN OF EXPERIMENTS

Causal relationships can only be set through experiments.

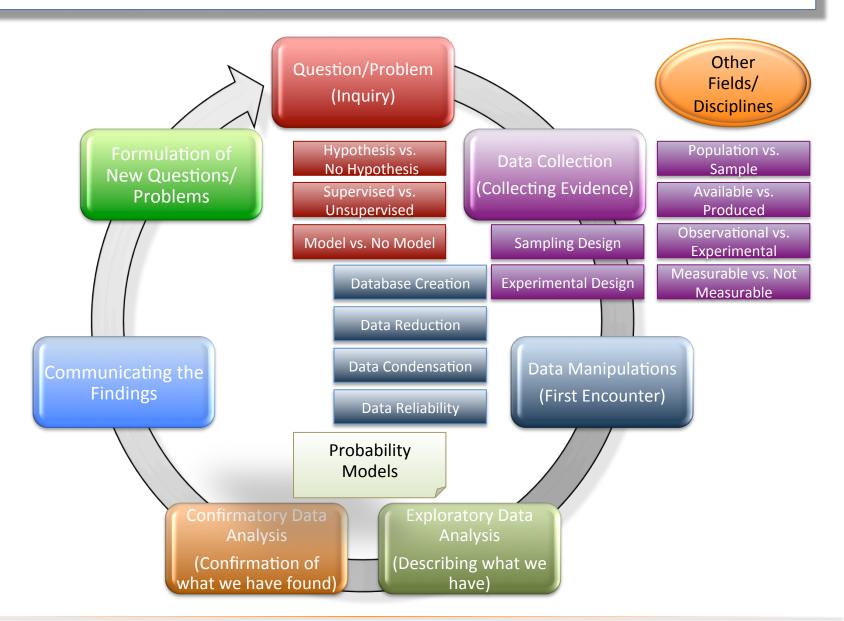
PRINCIPALS OF DESIGN OF EXPERIMENTS

CONTROL

RANDOMIZE

REPLICATE

DATA MANIPULATIONS



DATA MANIPULATIONS: DATA RELIABILITY

Data reliability is a state that exists when data is sufficiently complete and error free to be convincing for its purpose and context.

- > COMPLETE: Includes all of the data elements (variables/fields) needed for the analysis
- > ACCURATE:
 - ➤ **CONSISTENT:** The data was obtained and used in a manner that is clear and well-defined enough to yield similar results in similar analysis
 - > CORRECT: The data set reflects the data entered at the source and/or properly represents the intended results.
- > UNALTERED: The data reflects source and has not been tampered with.

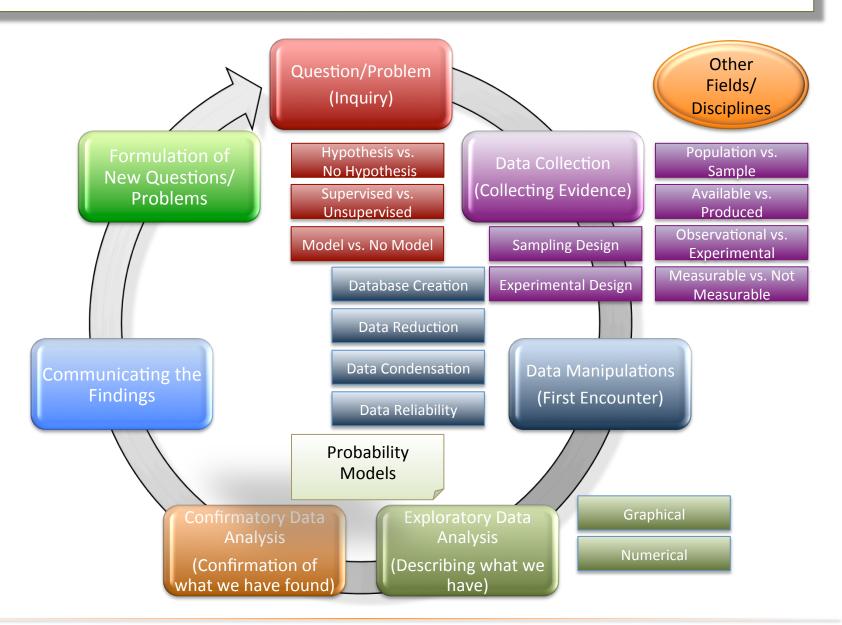
DATA MANIPULATIONS: DATABASE

Database is an organized collection of data

- Easy to use (data entry and data manipulations)
- Dynamic
- Interactive
- Open to collaboration
- Integrated
- Piece of paper
- Word processor (Microsoft Word)
- Microsoft Excel
- Microsoft Access
- Statistical software package (R, StatCrunch, SPSS, SAS etc.)
- Any program that uses SQL (Structured Query Language)
- Google Docs
- Google Fusion Tables

(UMM Data Services Center: http://mnstats.morris.umn.edu/UMMDataServicesCenter.html)

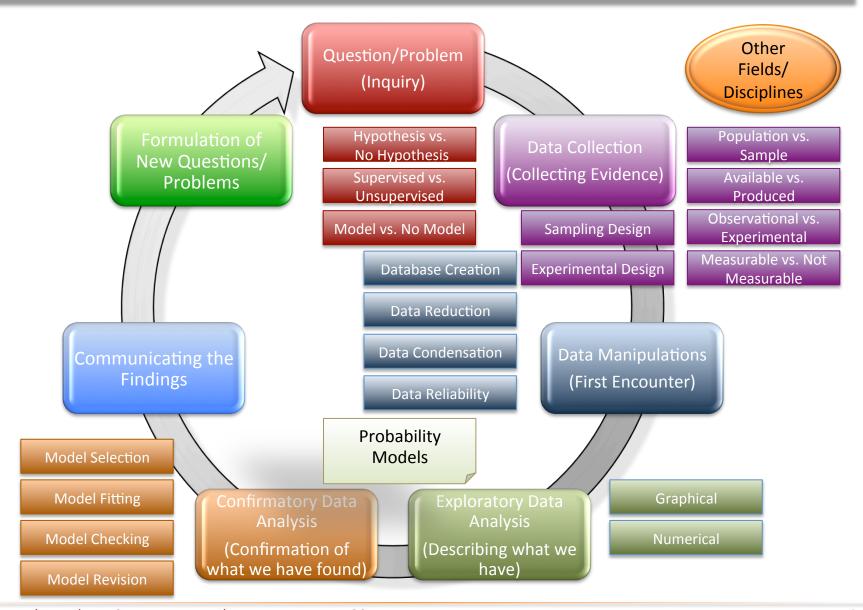
EXPLORATORY DATA ANALYSIS



EXPLORATORY DATA ANALYSIS

- Dynamic
- □<u>Interactive</u>
- ☐ Database integrated graphical displays

Correct selection of numerical and graphical summary techniques and methods



$$Y_1, Y_2, ..., Y_n$$
 are i.i.d. from $N(\mu, \sigma^2)$

$$Y_1, Y_2, \dots, Y_n \Rightarrow DATA$$

i.i.d. ⇒ independent and identically distributed

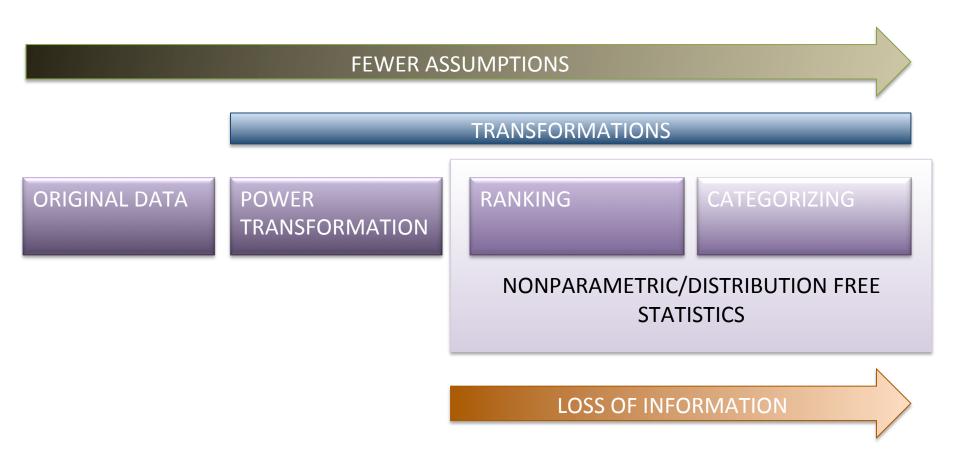
 \Rightarrow Simple random sample from the same population $N(\mu,\sigma^2) \Rightarrow$ Normal Distribution

$$Y_1, Y_2, \dots, Y_n$$
 are i.i.d. from $N\left(\beta_0 + \sum_{i=1}^p \beta_i \mu_{X_i}, \sigma^2\right)$

$$\beta_0 + \sum_{i=1}^p \beta_i \mu_{X_i} \Rightarrow \text{Linear Model N}(...,\sigma^2) \Rightarrow \text{Constant Variance}$$

CONFIRMATORY DATA ANALYSIS: TRANSFORMATIONS

WHAT TO DO WHEN THE MODEL ASSUMPTIONS ARE VIOLATED?

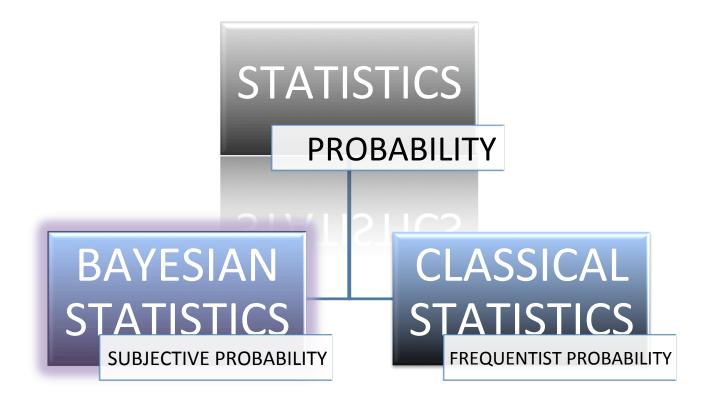


CONFIRMATORY DATA ANALYSIS: NONPARAMETRIC STATISTICS

- RANK-BASED METHODS
- PERMUTATION TESTS R.A. FISHER (1935)
- BOOTSRAP METHODS

 TAKE A SAMPLE OF SAME SIZE FROM THE SAMPLE WITH

 REPLACEMENT
- CURVE SMOOTHING
 NO LINEAR OR NONLINEAR MODEL



How would you describe in plain English the characteristics that distinguish Bayesian from Frequentist reasoning?

(http://stats.stackexchange.com)

Here is how I would explain the basic difference to my grandma:

I have misplaced my phone somewhere in the home. I can use the phone locator on the base of the instrument to locate the phone and when I press the phone locator the phone starts beeping.

Problem: Which area of my home should I search?

Frequentist Reasoning:

I can hear the phone beeping. I also have a **mental model** which helps me identify the area from which the sound is coming from. Therefore, upon hearing the beep, I infer the area of my home I must search to locate the phone.

Bayesian Reasoning:

I can hear the phone beeping. Now, apart from a **mental model** which helps me identify the area from which the sound is coming from, I also know the locations where I have misplaced the phone in the past. So, I combine my inferences using the beeps and my **prior information** about the locations I have misplaced the phone in the past to identify an area I must search to locate the phone.

Tongue firmly in cheek:

A *Bayesian* defines a "probability" in **exactly the same way that most non-statisticians do** - namely an indication of the plausibility of a proposition or a situation. If you ask him a question, he will give you **a direct answer** assigning probabilities describing the plausibilities of the possible outcomes for the particular situation (and state his prior assumptions).

A *Frequentist* is someone that believes probabilities represent long run frequencies with which events occur; if needs be, he will **invent a fictitious population** from which your particular situation could be considered a random sample so that he can meaningfully talk about long run frequencies. If you ask him a question about a particular situation, he will **not give a direct answer**, but instead make a statement about this (possibly imaginary) population. Many non-frequentist statisticians will be easily **confused** by the answer and interpret it as Bayesian probability about the particular situation.

P-VALUE? https://www.youtube.com/watch?v=ez4DgdurRPg

Very crudely I would say that:

Frequentist: Sampling is infinite and decision rules can be sharp. Data are a **repeatable random sample** - there is a frequency. **Underlying parameters are fixed i.e. they remain constant during this repeatable sampling process**.

Bayesian: Unknown quantities are treated probabilistically and the **state of the world can always be updated.** Data are observed from the realised sample. Parameters are unknown and described probabilistically. **It is the data which are fixed**.

CONFIRMATORY DATA ANALYSIS: SOME TECHNIQUES

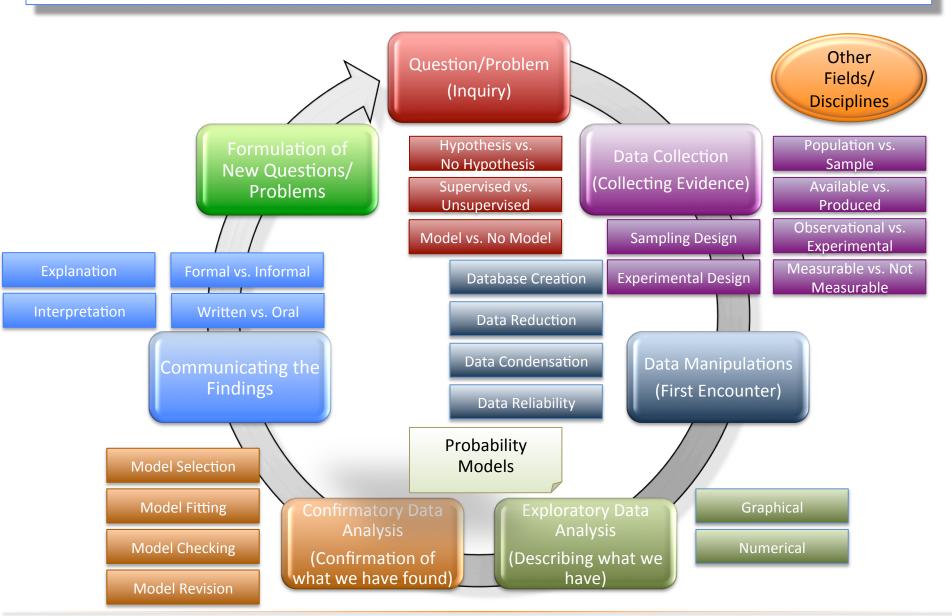
♦ MULTIVARIATE TECHNIQUES

http://mnstats.morris.umn.edu/multivariatestatistics/overview.html

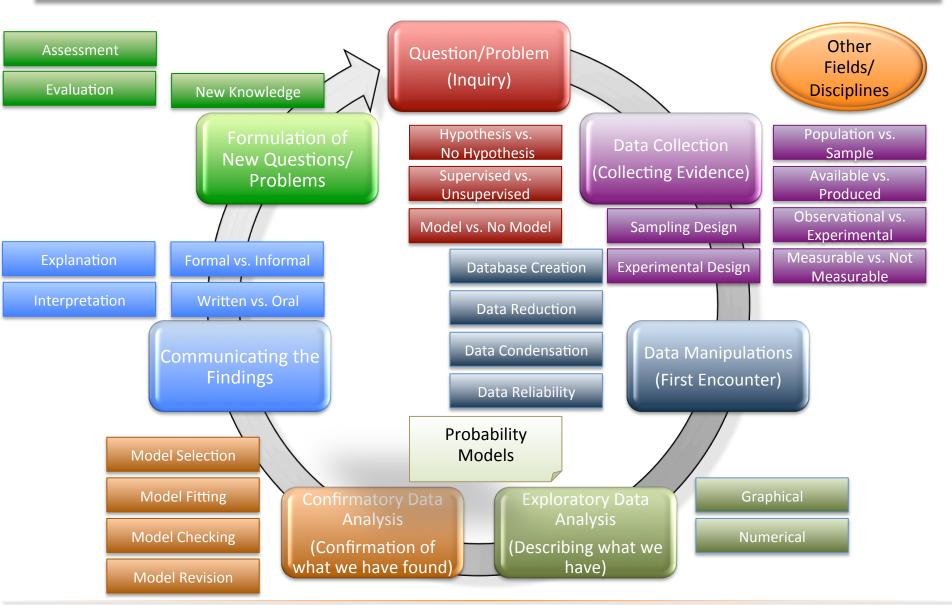
◆ NONPARAMETRIC/DISTRIBUTION FREE TECHNIQUES

http://mnstats.morris.umn.edu/introstat/nonparametric/learningtools.html

COMMUNICATING THE FINDINGS



FORMULATING NEW QUESTIONS/PROBLEMS



CONCLUDING REMARKS

SEE ME FOR A HELP

QUESTIONS?

• IF NOT, I HAVE SOME FOR YOU. PLEASE TAKE THE TEST BEFORE YOU LEAVE