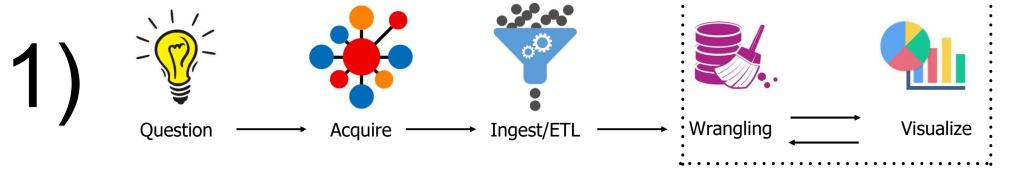


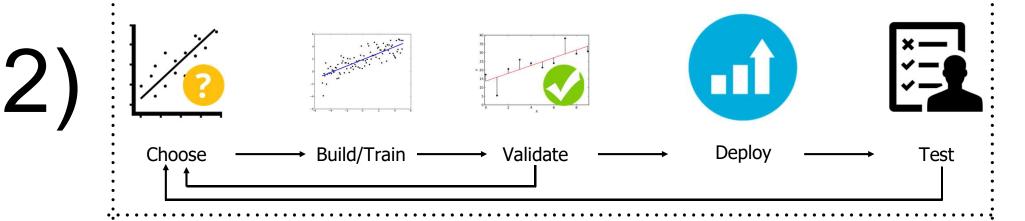
PYTHON & DATA SCIENCE WORKSHOP

https://www.facebook.com/diceanalytics/





Modelling



3)





Data Science Deconstructed

Ask a Lot of Questions **Identify All Available Datasets** Translate an ambiguous request into a Web, internal/ external databases, etc. concrete, well-defined problem **Extract Data Into Usable Format** Identify business priorities & strategy .csv, .json, .xml, etc. decisions that will influence your work 01 02 Frame the Problem Collect Raw Data **Identify Business Insights Examine Data at a High-Level** Return back to the business problem Understand every column; 06 identify errors, missing values & **Visualize Your Findings** THE corrupt records 03 **DATA SCIENCE** Commu- Keep it simple & priority-driven Clean the data nicate **PROCESS** Process Tell a Clear & Actionable Story Results the Data Throw away, replace, and/or filter Effectively communicate to corrupt /error prone / missing non-technical audiences values 05 04 **Perform** In-Depth Explore the Data **Play Around With the Data Create a Predictive Model** Analysis • Split, segment, & plot the data in Use feature vectors from step #4 different ways **Evaluate & Refine Model Identify Patterns & Extract Features** Perhaps return to step #2, 3, or 4 Use statistics to identify & test significant



variables

SKILLS REQUIRED



FRAME THE PROBLEM

- Domain Knowledge (needs)
- Product Intuition (metrics)
- Business Strategy (priorities)
- Teamwork (people & resources)



COLLECT RAW DATA

- Database Management
 - Systems: mySQL, postgreSQL, Oracle, MongoDB
- Querying Structured Databases
 SQL
- Retrieving Unstructured Info
 - Informational Retrieval / Text Mining
- Distributed Storage
 - Hadoop HDFS, Spark, Flink



PROCESS THE DATA

- Scripting Language
 - Python or R
- Data Wrangling & Cleaning
 - Python "Pandas" library
- Distributed Processing
 - Hadoop MapReduce / Spark



EXPLORE THE DATA

- Scientific Computing
 - Python: numpy, matplotlib, scipy, pandas
- Inferential Statistics
 - hypothesis testing
 - correlation vs. causation
- Experimental Design
 - A/B tests, controlled trials



PERFORM IN-DEPTH ANALYSIS

- Machine Learning
 - Supervised / Unsupervised algorithms
 - Contextual pros/cons)
- ML Tools Library
 - Python: scikit-learn
- Advanced Math
 - Linear Algebra & Multivariate Calculus



COMMUNICATE RESULTS

- Business Acumen
 - Non-technical terminology
- Data Visualization Tool(s)
 - Tableau, D3.js, Google visualize, matplotlib, ggplot, seaborn
- Data Storytelling
 - presenting & speaking
 - reporting & writing



Statistical Flow

Population

Sample

Design Study

Scope

EDA

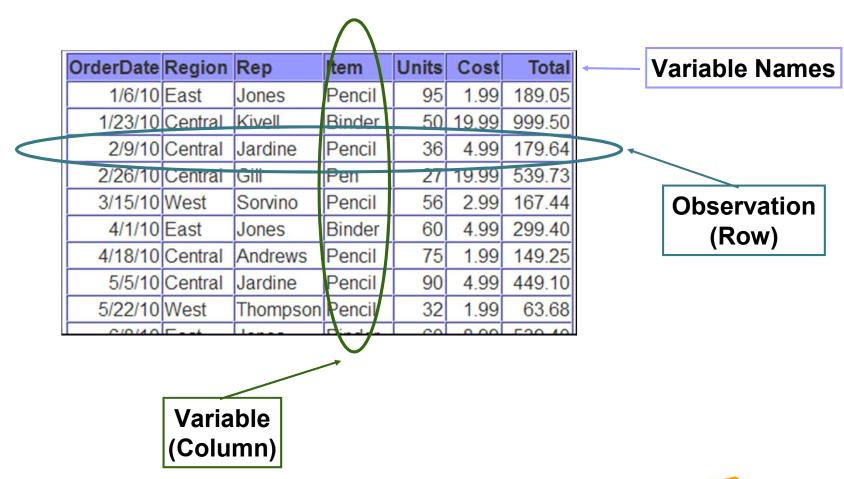
Model

Inference



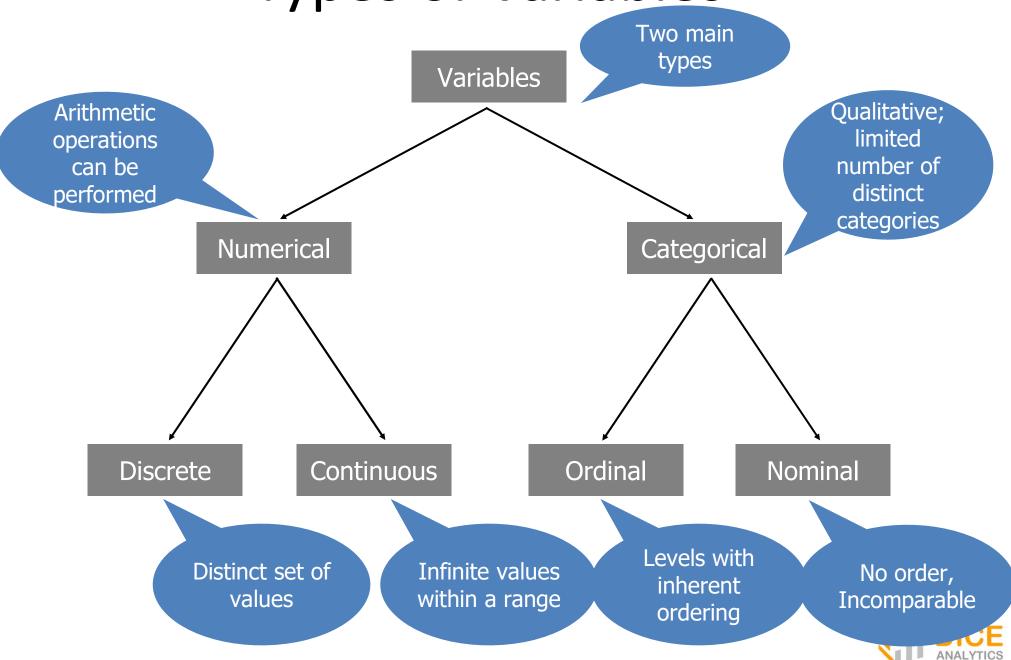
Data Organization

Data is stored in the form of a *Data Matrix*





Types of Variables



Types of Variables

http://www.statisticshowto.com/types-variables/

https://statistics.laerd.com/statistical-guides/types-of-variable.php



Types of Variables

• <u>Response Variable</u>: It is the focus of a question in a study or experiment. It is the variable we want to predict or observe. It is the dependent variable.

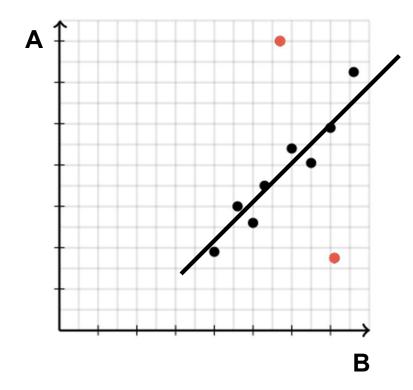
 Explanatory Variable: It is the variable on whom the response variable depends, or the variable which 'explains' the response variable. It is assumed to be independent variable.



Relationship b/w Variables

 Two variables that show connection with each other are called <u>Associated/Correlated (Dependent)</u>

- Two variables that do not show connection with each other are called <u>Independent</u>
- An observation that is away that is not close to majority of data is called <u>Outlier</u>





Sampling



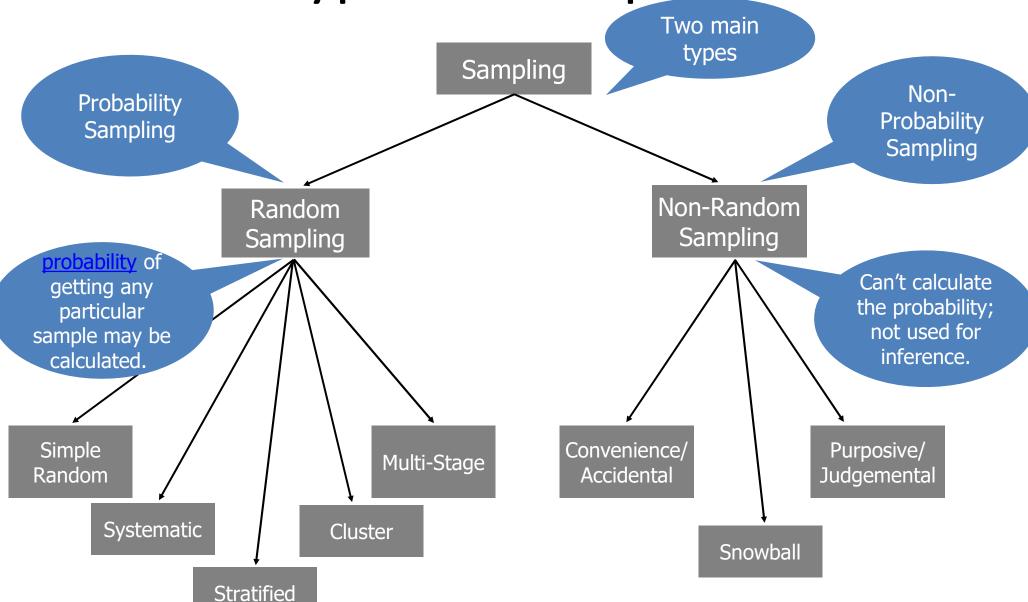
Census vs Sample

- <u>Census</u>: A census is a study of every unit, everyone or everything, in a population. It is known as a complete enumeration, which means a complete count.
- Census not mostly possible: time-consuming, expensive, population hardly still, etc.

 Sample: A sample is a subset of units in a population, selected to represent all units in a population of interest.



Types of Sampling



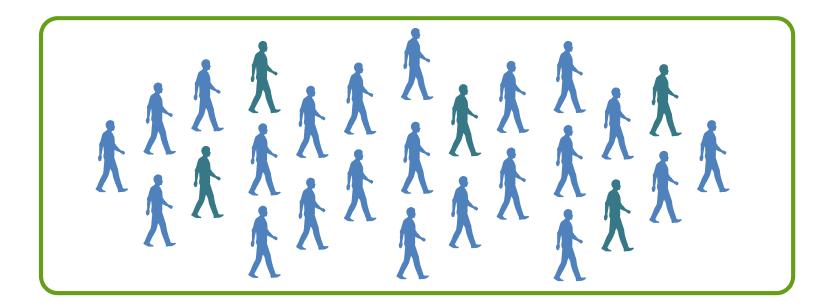


Random Sampling



Simple Random Sampling (SRS)

- Select n observations randomly from entire population
- Each observation is likely to be selected





Systematic Sampling

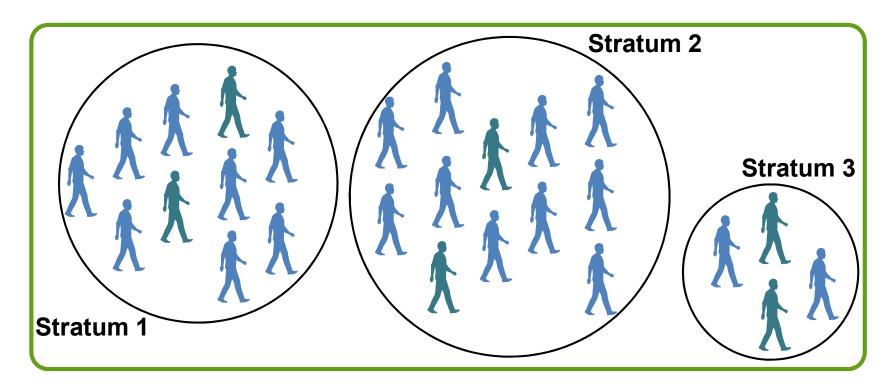
- Arrange the population according to some ordering
- Start randomly and select every kth observation

K = 4



Stratified Sampling

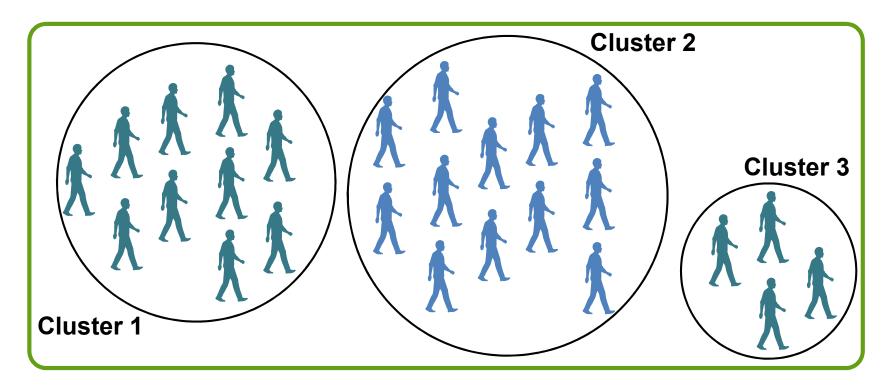
- Divide population in homogenous groups called <u>strata</u>
- Do Simple Random Sampling (SRS) from each stratum





Cluster Sampling

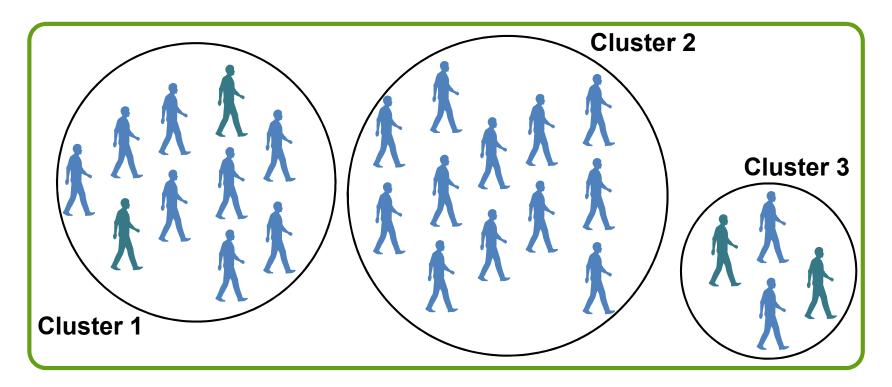
- Divide population in heterogenous groups called <u>clusters</u>
- Randomly Sample k clusters; and sample all observations within those clusters





Multi-Stage Sampling

- Divide population in heterogenous groups called <u>clusters</u>
- Randomly Sample k clusters; and do SRS within those clusters



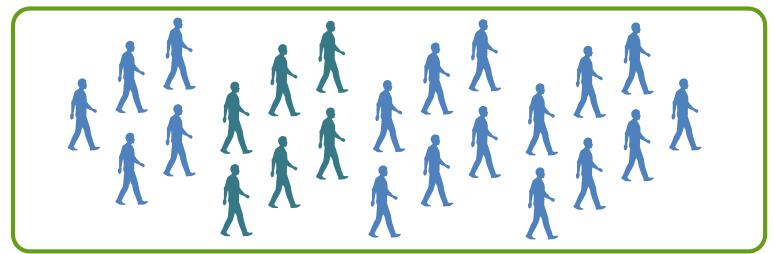


Non-Random Sampling



Convenience/Accidental Sampling

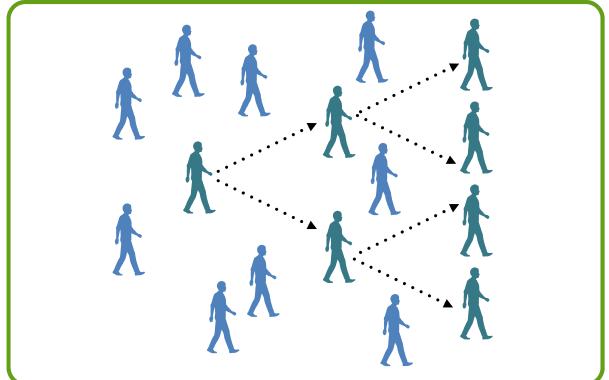
- Members of the population are chosen based on their relative ease of access.
- To sample friends, co-workers, or shoppers at a single mall, are all examples of convenience sampling.
- Such samples are biased because researchers may unconsciously approach some kinds of respondents and avoid others (Lucas 2014a), and respondents who volunteer for a study may differ in unknown but important ways from others (Wiederman 1999).





Snowball Sampling

- The first respondent refers an acquaintance. The friend also refers a friend, and so on.
- Such samples are biased because they give people with more social connections an unknown but higher chance of selection (Berg 2006), but lead to higher response rates.





Purposive/Judgmental Sampling

- The researcher chooses the sample based on who they think would be appropriate for the study.
- This is used primarily when there is a limited number of people that have expertise in the area being researched, or when the interest of the research is on a specific field or a small group.





Sampling Bias vs Selection Bias

- <u>Sampling Bias</u>: A **bias** in which a **sample** is collected in such a way that some members of the intended population are less likely to be included than others; occurs when you choose your sample which is the 1st step of a research.
- <u>Selection Bias</u>: A **bias** introduced by the **selection** of individuals, groups or data for analysis in such a way that proper randomisation is not achieved; occurs when you select which subject goes to the control group and which to the treatment group.



Sources of Sampling Bias

• <u>Convenience Sample</u>: Easily accessible people more likely to be included in the sample.

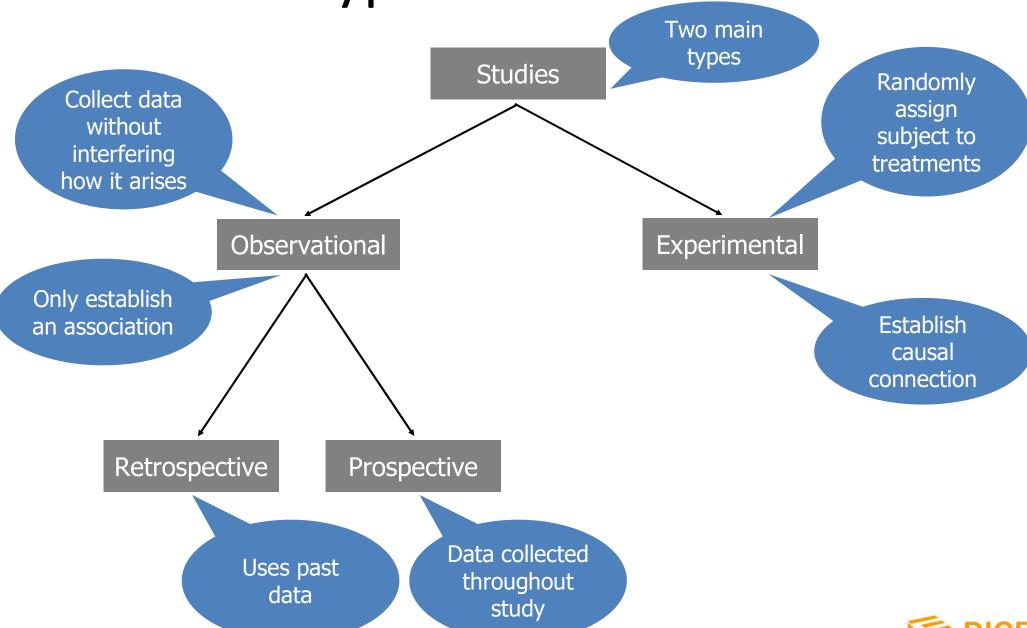
- Non-Response: If only particular type(s) of randomly sampled people respond to survey.
- Voluntary Response: Happens when sample consists of people who volunteered to respond because they are opinionated.



Study Design



Types of Studies





Correlation vs Causation

- <u>Correlation</u>: It describes the mutual relationship or connection between an independent and dependent variable.
- <u>Causation</u>: Causation, also known as cause and effect, is when an observed event or action (independent variable) appears to have caused a second event or action (dependent variable).

Correlation does not imply Causation!



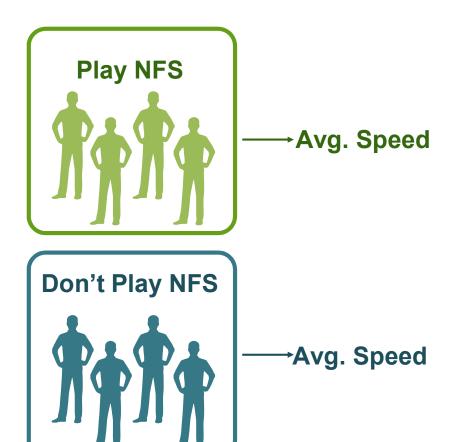
Random Assignment

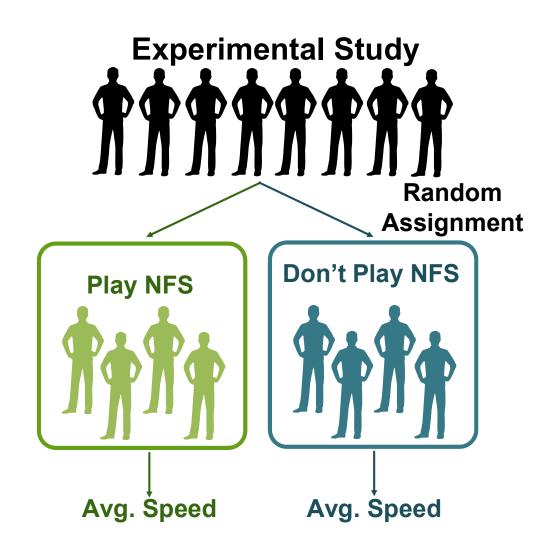
Find out relationship between <u>playing NFS</u> and actual <u>driving speed</u> of a person



Random Assignment

Observational Study



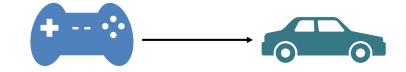




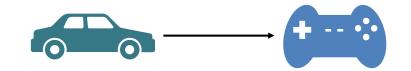
(No) Random Assignment

In Case of Observational Study:

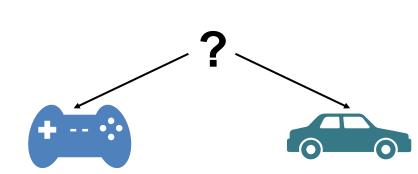
 Playing NFS causes the person to drive faster



 Driving faster causes the person to play NFS

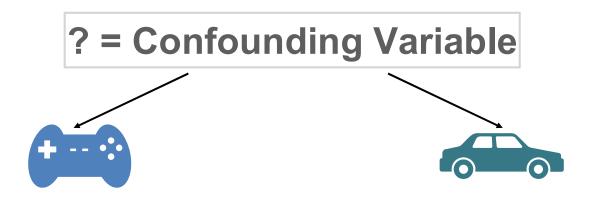


 A third variable is responsible for both these variables





Confounding Variable



Extraneous variables that affect both the explanatory and response variable, and make it look like there is a relationship (association/dependence) between them are called Confounding Variables.

Maybe because they are very rich!



Principles of Experimental Design

Control

Compare treatment of interest to a control group

Randomize

Randomly assign subjects to treatments

Replicate

Collect a sufficiently large sample; or replicate entire study

Block

Block for variables known or suspected to affect the outcome



Blocking

Find out effect of *Hi-Octane* on car's *Speed*



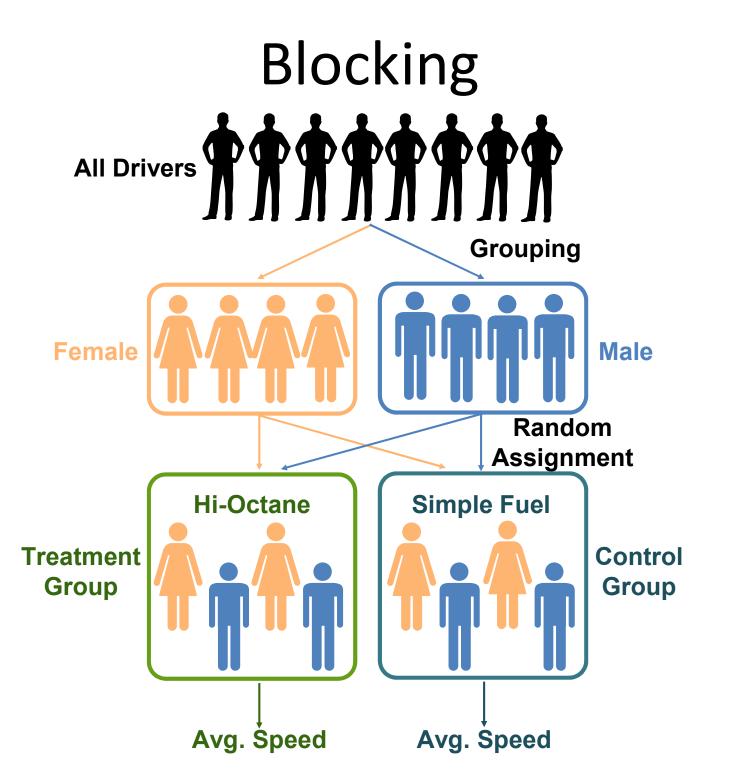
Blocking

- Control Group: Simple fuel
- Treatment Group: Hi-Octane

Male drivers may ride faster than Female drivers!

- Gender is blocking variable
- Need to 'block' Male status
- Divide sample to Male and Female groups (just like we do in Stratified Sampling)
- Randomly assign Male and Female drivers to control and treatment group, ensuring equal representation in both groups
- Now they cancel out the effect of gender, so we can say the difference in speed is solely because of Hi-Octane





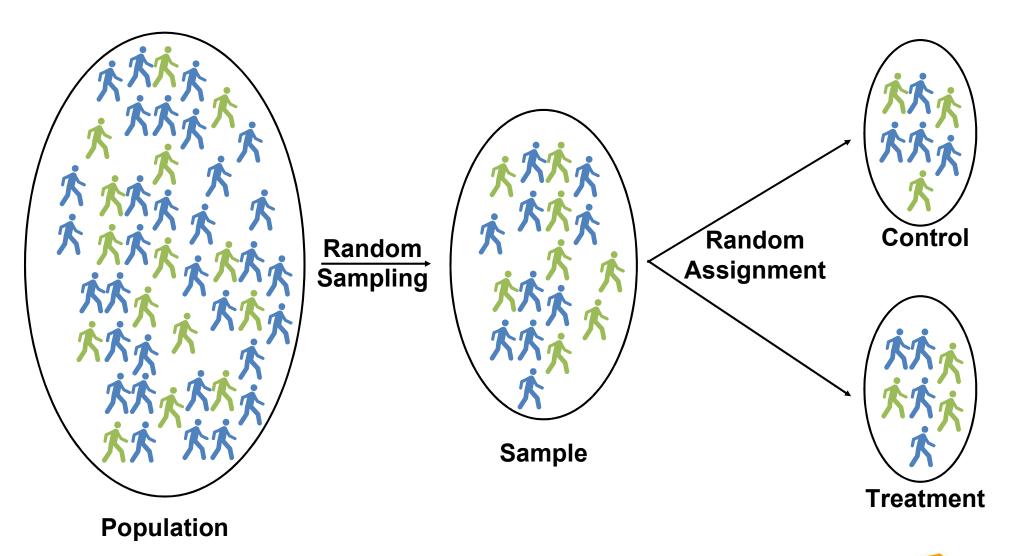


Blocking Variable vs Explanatory Variable

- <u>Explanatory Variables (factors)</u>: Conditions we can impose on experimental units.
- <u>Blocking Variables</u>: Characteristics that experimental units come with, that we would like to control for.
- Blocking is like Stratifying:
 - -> blocking during random assignment
 - -> stratifying during random sampling



Random Sampling vs Random Assignment





Scope



Most No Random Ideal Random Observational **Studies Studies** Assignment Assignment Random Causal and Not causal, Generalizability But generalizable Sampling Generalizable No Random No Causal, But Neither causal not generalizable nor generalizable Sampling Generalizability Most Bad Causation Experimental Association Studies **Studies**

