Ydata: Introduction to Data Science



Lecture 02: Cause and Effect

Overview

Brief demonstration on Jupyter Notebooks

The history of Data Science continued...

Experiments and observational studies

- Some motivating examples
- Association vs. causation
- Next class: John Snow and the Broad Street Pump

Assignment 0

Please try assignment 0 to test that you have a working environment to run Jupyter Notebooks

- This assignment does not need to be turned in
- Requires <u>installing Anaconda and the datascience package</u>, or using <u>Google</u> colab
- Instructions are on Canvas and assignment 0 is on the class calendar page

Let's take a quick look at using Jupyter Notebooks...

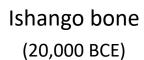
Questions

Continuation of the history of Data Science...

The history of Data Science

(a very incomplete list)

Data



Cuneiform tablets (4,000 BCE)

Quipus in South America (1100-1500)

Demographics

(1600's)

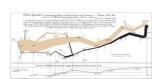
Golden age of data visualization

(1850-1900)

Big data

(now)





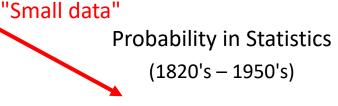


Probability

Key Take Away

Probability models dominated data analysis prior to using computational methods

Initial development (1600's)



Math Stats dominates (1900-1960's)

Computers

Abacus (2400 BCE)



Antikythera mechanism (100 BCE)



Analytical Engine (1800's)



Hollerith Tabulating Machine (1890)



Mainframes, PCs, Internet, etc. (1950-present)



"Big data"

Brief history of Data Science: the rise of Data Science

Probability models initially dominated data analysis but the rise of powerful computers and plentiful data has given rise to new approaches to analyzing data.

- John Tukey (1962) looks for a broadening of data analysis beyond mathematics
- Breiman (2001) describes a mathematical modeling culture and algorithmic culture
- The term "Data Science" starts being used in the 2000's to describe computational approaches to analyzing data
 - Donoho (2017). 50 Years of Data Science.

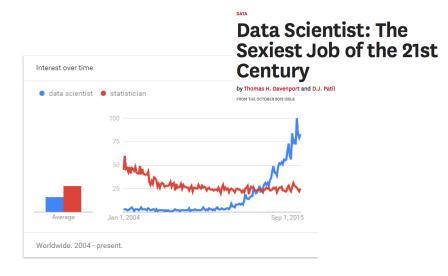
THE FUTURE OF DATA ANALYSIS1

By John W. Tukey

Princeton University and Bell Telephone Laboratories

Statistical Modeling: The Two Cultures

Leo Breiman



Big Data

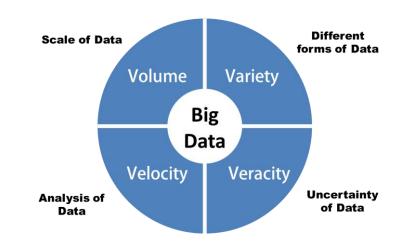
New insights:

- Lots of new data from Internet, sensors etc., can be mined to transform our understanding in a range of fields
 - E.g., health, cosmology, social sciences, etc.

New approaches and tools:

- Hypothesis test pick up on very small (meaningless) effects with very large samples
- Data manipulation, programming, computational infrastructure are needed to extract insights





New ways to choose the best methods

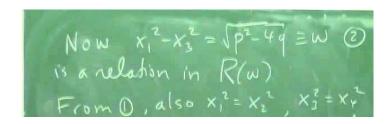
Statistics focuses on mathematical models (probability distributions) to analyze data

 Best methods are the ones that have mathematical guarantees (proofs)

Data Science empirically evaluates data analysis methods

• Best methods are the one that gives the most insight *in practice*

The proof is in the math



The proof is in the pudding



<u>Data Science vs. Statistician video</u>

So what is Data Science?

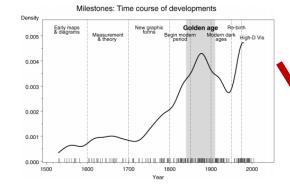
Data Science is a broadening of data analyses beyond what traditional Statistical mathematical/inferential analyses to use more computation

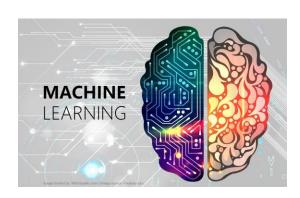
Many other fields impacted by 'Data Science

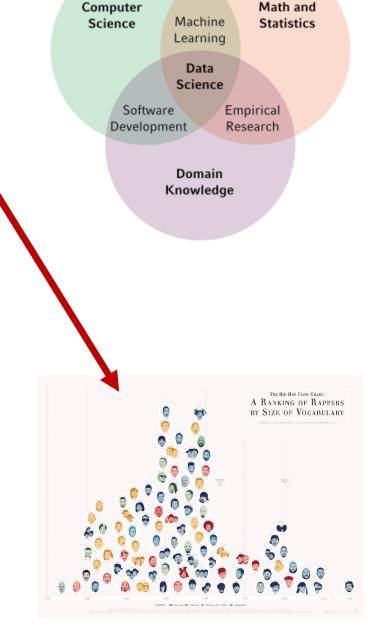
- Making business decisions
- Predictive medicine
- Fraud detection
- Etc.

Examples:

- NYC city bike visualization
- Wind map visualization







Ethical concerns around privacy, fairness and other issues





Observational and Experimental Studies

Is it best to run in the afternoon?

The New York Times

PHYS ED

The Best Time of Day to Exercise

Men at risk for diabetes had greater blood sugar control and lost more belly fat when they exercised in the afternoon than in the morning.





Getty Images

Should we drink a lot of coffee?

Three coffees a day linked to a range of health benefits

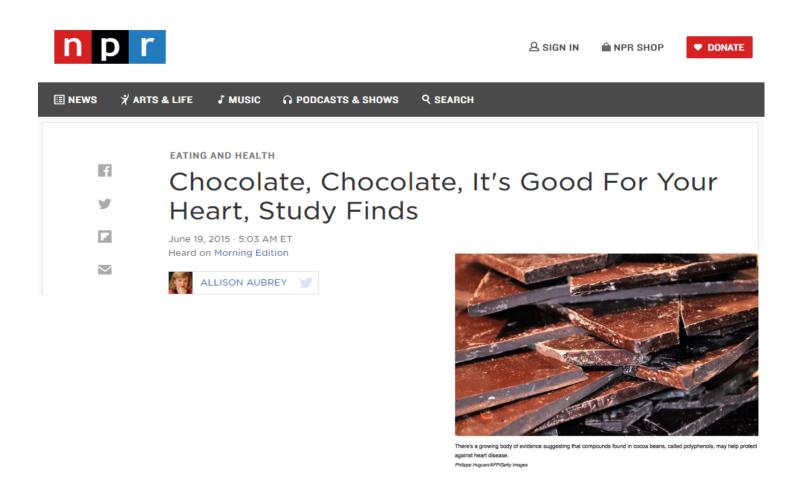
Research based on 200 previous studies worldwide says frequent drinkers less likely to get diabetes, heart disease, dementia and some cancers

Staff and agencies Wed 22 Nov 2017 19.54 EST



The findings supported other studies showing the health benefits of drinking coffee. Photograph: Wu Hong/EPA

Is chocolate good for your heart?



Terminology

Cases

- Units that we take measurements from
 - E.g., European adults

Treatment (explanatory variable)

- A property that differs between groups
 - E.g., chocolate consumption

Outcome (response variable)

- A measurement of interest
- We want to see if the outcome differs depending on the treatment
 - E.g., heart disease differ between groups



Treatment (explanatory variable)	Outcome (response variable)
Chocolate	Cardiovascular disease
No Chocolate	Cardiovascular disease
Chocolate	No cardiovascular disease
Chocolate	No cardiovascular disease

Association

An association is the presence of <u>a reliable</u> relationship between the treatments an outcome



E.g., Do people who eat chocolate have lower rates of heart disease?

Some data:

 "Among those in the top tier of chocolate consumption, 12 percent developed or died of cardiovascular disease during the study, compared to 17.4 percent of those who didn't eat chocolate."

This suggests there may be an association!

Causation

A causal relationship is when changing the value of a treatment variable influences the value outcome variable

E.g., Consuming chocolate leads to a reduction in heart disease

Association does not ≠ causation!

Incorrectly implying causation is one of the most common mistakes in Statistics and Data Science

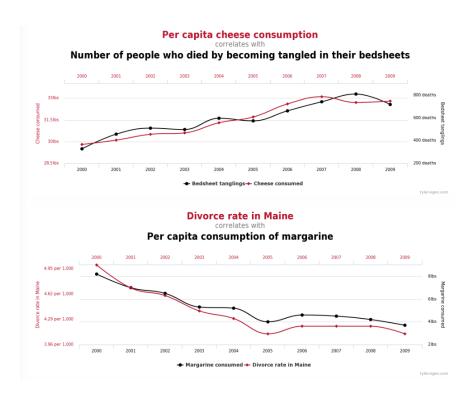
Causal relationships?

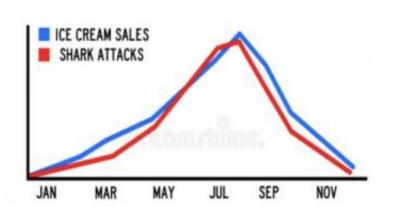
Since the 1950's both the atmospheric CO₂ level and the obesity levels have increased sharply. CO₂ causes obesity?

• Spurious correlation

As ice cream sales increase, the rate of shark attacks increases. Does eating ice cream cause shark attacks?

Confounding



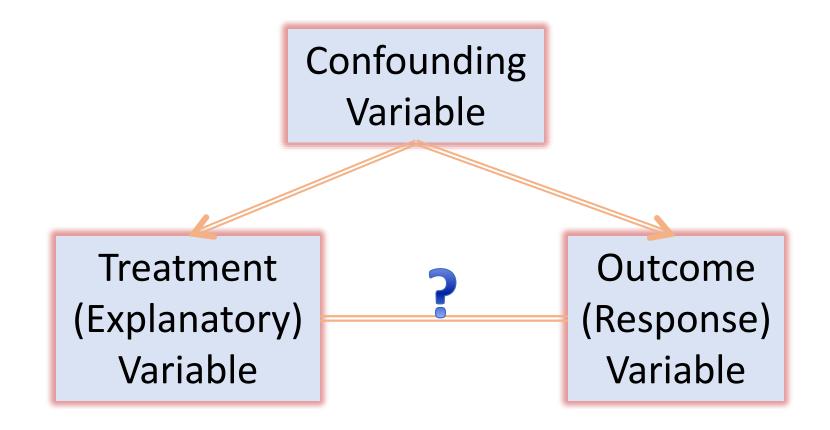


Confounding

A **confounding variable** (also known as a **lurking variable**) is a third variable that is associated with both the treatment (explanatory) variable and the outcome (response) variable

A confounding variable can offer a plausible explanation for an association between the other two variables of interest

Confounding



Observational and experimental studies

An **observational study** is a study in which the researcher does not actively control the value of any treatment variable but simply observes the values as they naturally exist

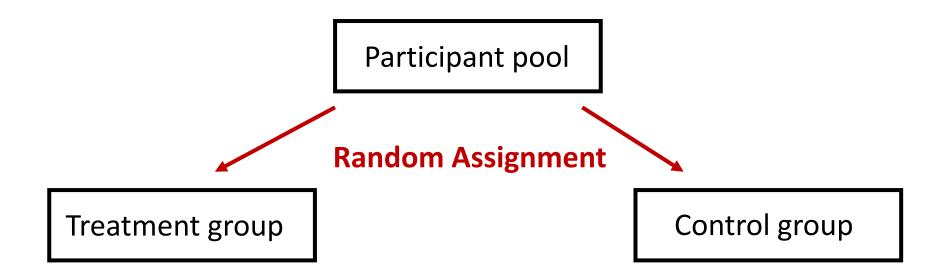
An **experiment** is a study in which the researcher actively controls one or more of the <u>treatment</u> variables

- Randomly assigns treatments to cases
- Allows one to get at questions of causation!

Randomized Controlled Experiment

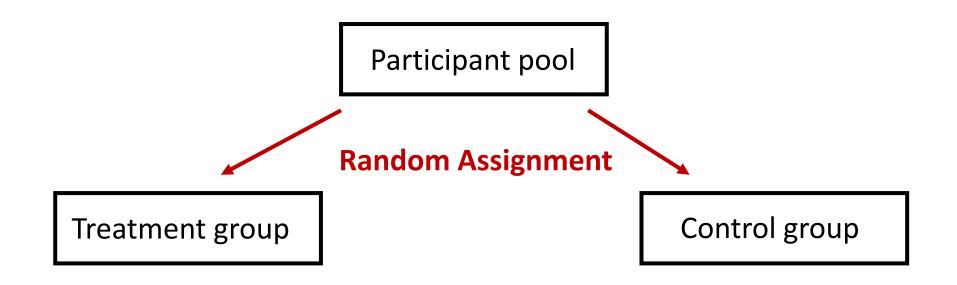
Take a group of participant and *randomly assign*:

- Half to a *treatment group* where they get chocolate
- Half in a control group where they get a fake chocolate (placebo)
- See if there is more improvement in the treatment group compared to the control group

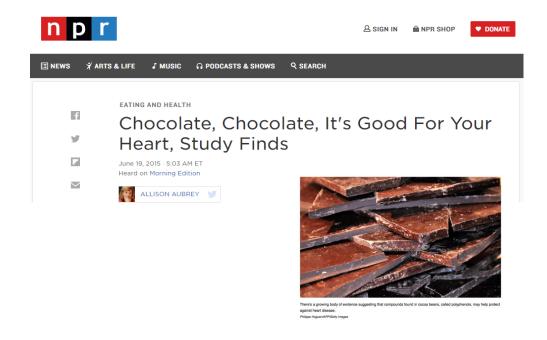


Randomized Controlled Experiment

If chocolate was not effective at reducing heart disease, would we expect a difference in the amount (proportion) of heart disease between the treatment and control groups?



Central question: is chocolate good for your heart?



Does the cholate study show there is a causal link?

Why or why not?

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Next class: Cholera and Python

Things to do:

- Try to install Anaconda/Python on your computer and/or try out Google colabs
- Try out homework 0
 - Does not need to be turned in

