Welcome to Week #2!

General Survey Results

Most of programming experience at about 1 year

Exercises will have wide range of options for different levels.

Wide range of data interests

Will use "real" datasets, wide variety of topics (feel free to suggest some!)

People are excited about learning more about programming

We will do a lot of collaborative coding in class, different options for different levels.

Week	Topic	Reading
1	 Data, Models, and Information Elementary statistics: Definitions Overview of R 	OIS 1 (ISL 1)
2	Elementary statistics: Applications & Plots	OIS 1 (ISL 1)
3	 Introduction to data analysis with R Review of tabular and graphical displays of data 	ITR 1, 2, 5, 6, 7, 12
4	 Random variables: expectation and variance Joint and conditional probability Bayes rule 	OIS 2
5	Random variables: distributions (normal, binomial, poisson)	OIS 3

Definitions, basic concepts, R practice

HW and Exam Formats

File name structure: lastname-first-module.ext (e.g, naiman-jill-assignment1.pdf).

The submission must include:

1) A narrative document as a PDF file (to be read by a human). To preserve the natural flow of the narrative, figures (e.g., screenshots, code snippets) and tables should be embedded into the document near their first mention. Any supplementary files containing R programs or data should be referenced in the text and separately uploaded.

AND

2) All R code as separate files with an .R extension (to be read by a computer).

ALSO: make sure you include any files needed to run your R-script (data files for example)

Last time...



Last time...

Summary Statistic Definitions!

Mean (Sample) = sum of all data values divided by number of data points

$$\text{Mean} = \frac{\text{Sum of all values}}{\text{Number of values}}$$
 Symbolically, $\overline{x} = \frac{\sum x}{n}$

where \overline{x} (read as 'x bar') is the mean of the set of x values, $\sum x$ is the sum of all the x values, and n is the number of x values.

(note - only works with "numerical" data types... more about data types later)

Median = if we order the data from smallest to largest, this is the observation in the middle (splits the data in 2 halves)

First/Third Quartiles = where 25% of the data falls below/above

Standard Deviation = this is the square root of the variance, where the variance is roughly the average distance of data values from the mean

Standard Deviation (sample) =
$$\sqrt{\frac{\sum_{i=1}^{n} (x_i - \text{mean})^2}{n-1}}$$

Last time...

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Lets play with *your* survey data!

IS542 Spring 2020 Poll

This is a quick, informal poll to gauge skill with statistics & programming to guide development of the course.

* Required

Name *

Your answer



How would you gauge your familiarity with Statistics? *

1

2

5

Novice

C

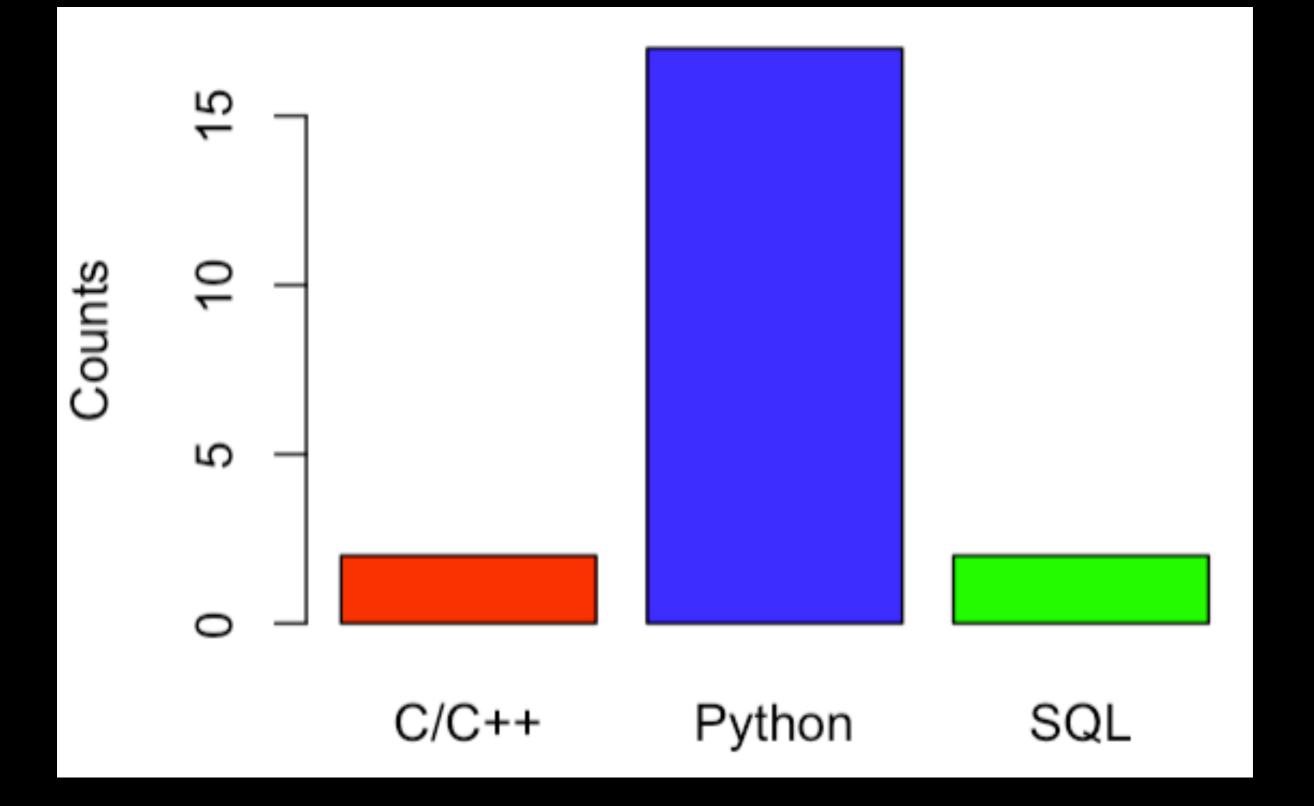
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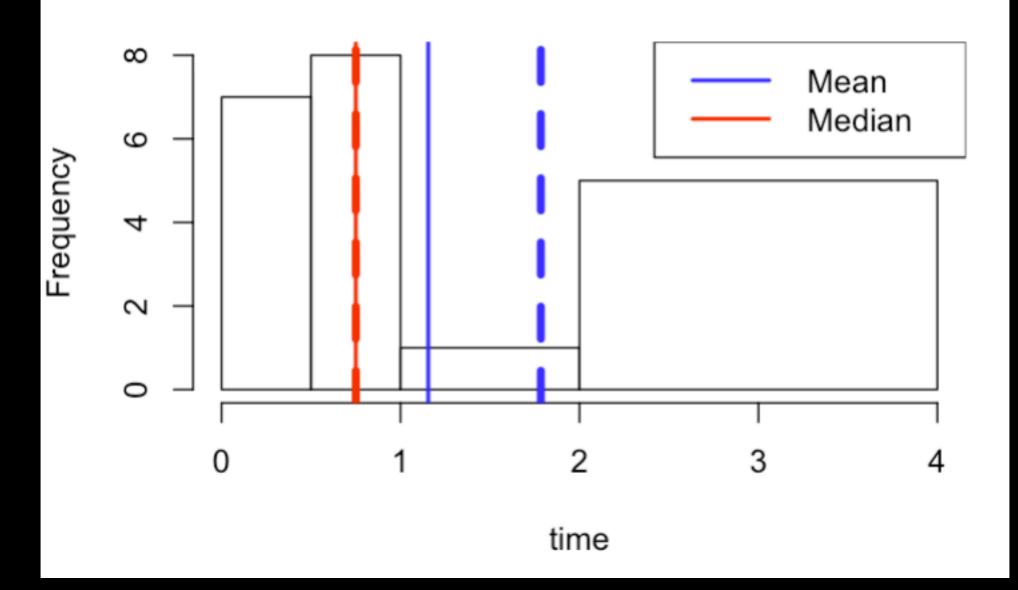
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Expert

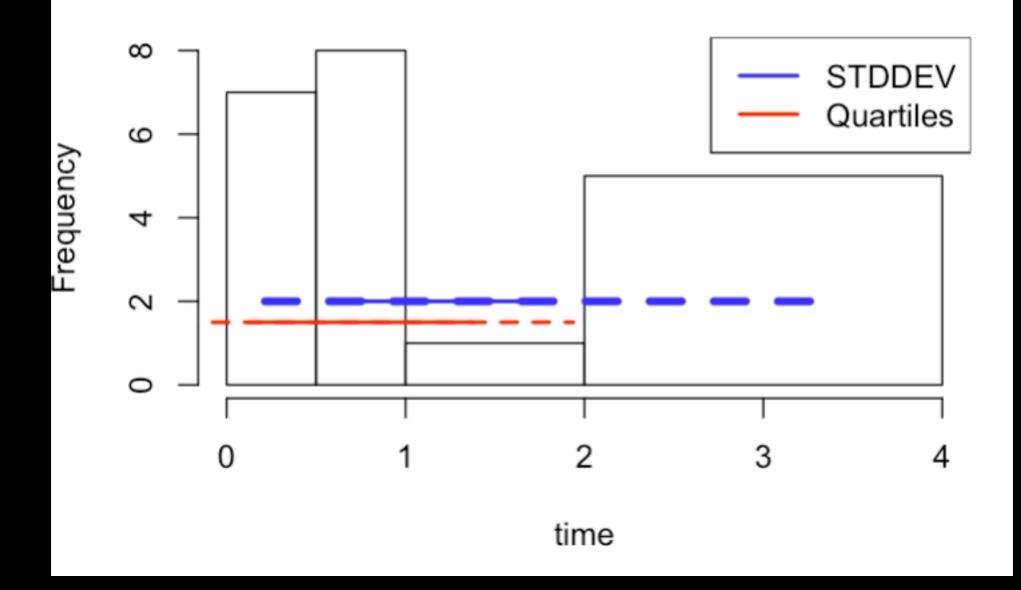
How would you gauge your familiarity with programming?

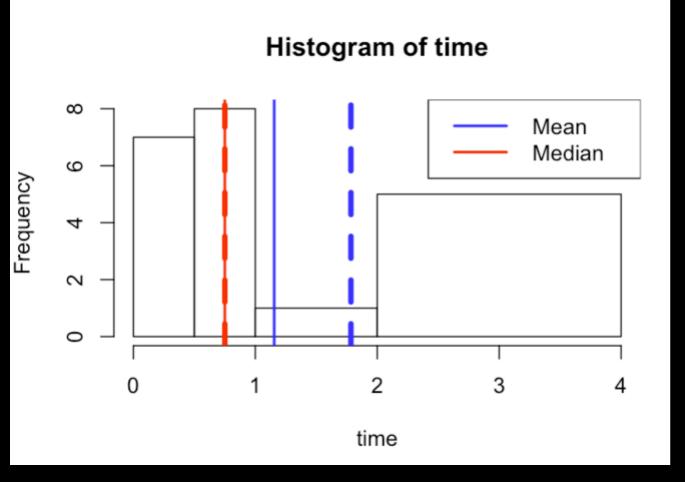


Histogram of time

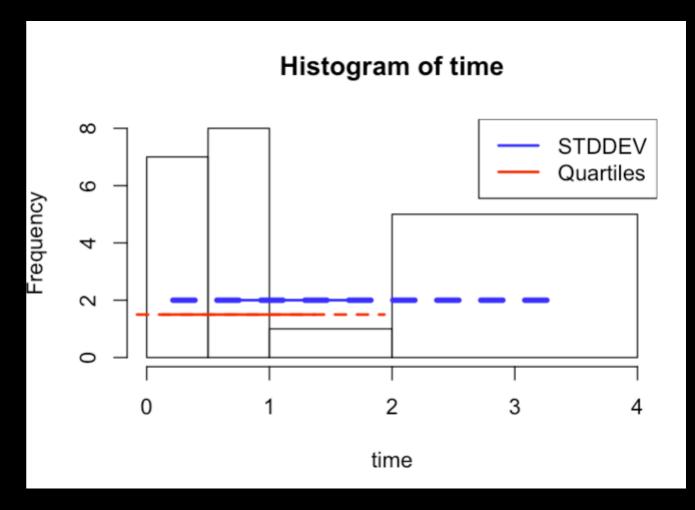


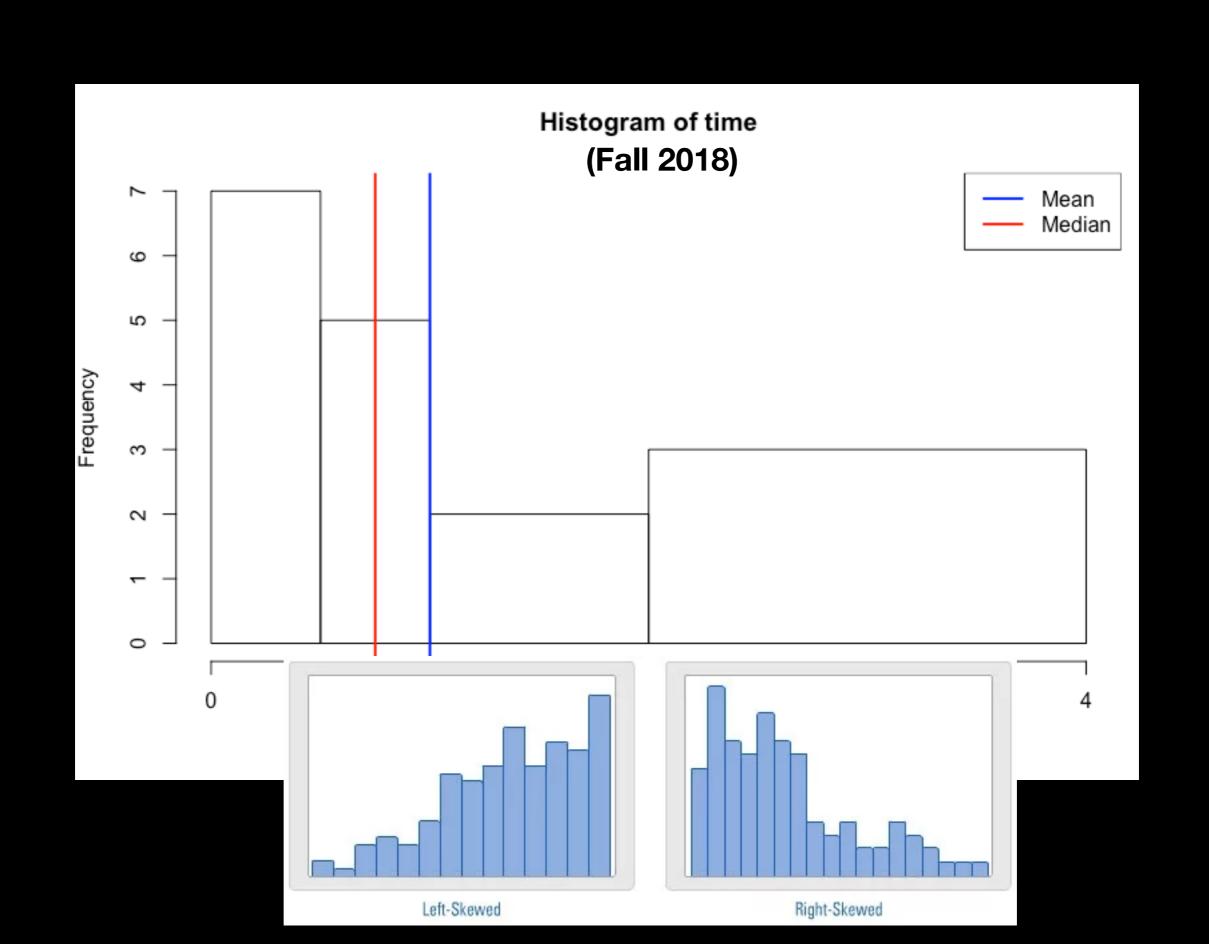
Histogram of time

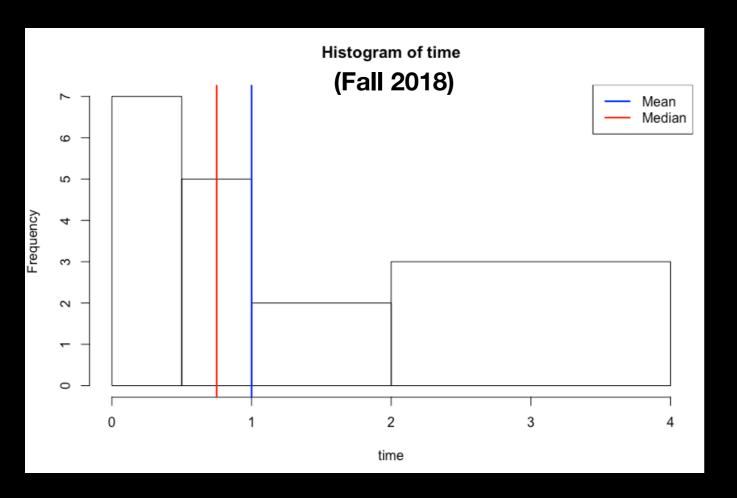


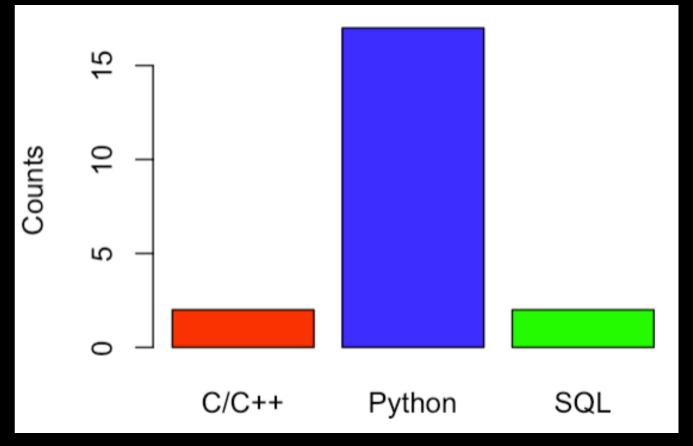


Median and quartile ranges are "robust" statistics - i.e. not as sensitive to variability









Variables

Data Matrix

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The second secon			
Fam.Stats =	language [‡]	time_yr.min 🗦	time_yr.max
4	Python	0.0	0.5
4	Python	2.0	4.0
3	Python	0.0	0.5
4	Matlab	2.0	4.0
2	Python	2.0	4.0
3	Python	1.0	2.0
3	Python	0.5	1.0
3	Python	0.5	1.0
3	C/C++	0.5	1.0
3	Java	0.0	0.5
	Fam.Stats 4 4 3 4 2 3 3 3 3 3	4 Python 4 Python 3 Python 4 Matlab 2 Python 3 Python 3 Python 3 Python 3 Python 3 C/C++	Eam.Stats language time_yr.min 4 Python 0.0 4 Python 0.0 3 Python 0.0 4 Matlab 2.0 2 Python 2.0 3 Python 1.0 3 Python 0.5 3 Python 0.5 3 C/C++ 0.5

Variable Types

NOT Numerical

Numerical (+/-, means, etc)

	Fam.Stats	language [‡]	time_yr.min ÷	time_yr.max =
1	4	Python	0.0	0.5
2	4	Python	2.0	4.0
3	3	Python	0.0	0.5
4	4	Matlab	2.0	4.0
5	2	Python	2.0	4.0
6	3	Python	1.0	2.0
7	3	Python	0.5	1.0
8	3	Python	0.5	1.0
9	3	C/C++	0.5	1.0
10	3	Java	0.0	0.5

<u>discrete</u> numerical

VS.

continuous numerical

Variable Types

Categorical (levels)

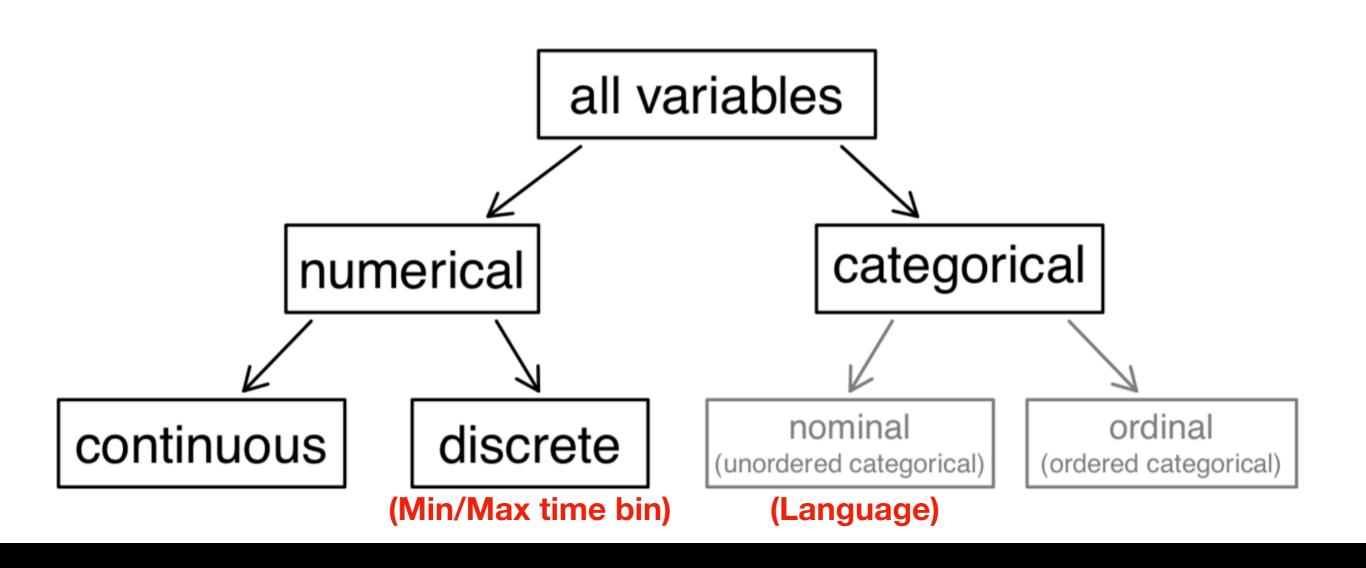
Numerical (+/-, means, etc)

•	Fam.Stats 🗦	language [‡]	time_yr.min ÷	time_yr.max ÷
1	4	Python	0.0	0.5
2	4	Python	2.0	4.0
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7	3	Python	0.5	1.0
8	3	Python	0.5	1.0
9	3	C/C++	0.5	1.0
10	3	Java	0.0	0.5

nominal categorical

VS.

ordinal categorical



Population of interest

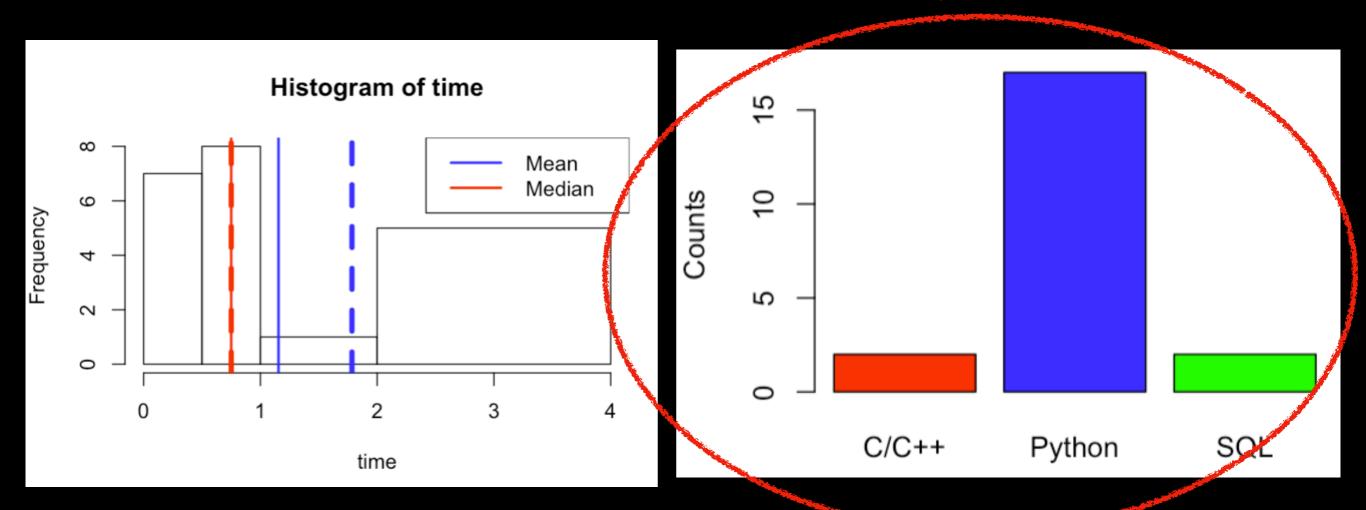
Can we use this data to ask the following questions about the typical iSchool student?

- 1. Can we say typical iSchool student codes in python?
- 2. Can we say typical iSchool student has been programming for <1 year? >2 years?

Why or why not?

Share: biases of sample?

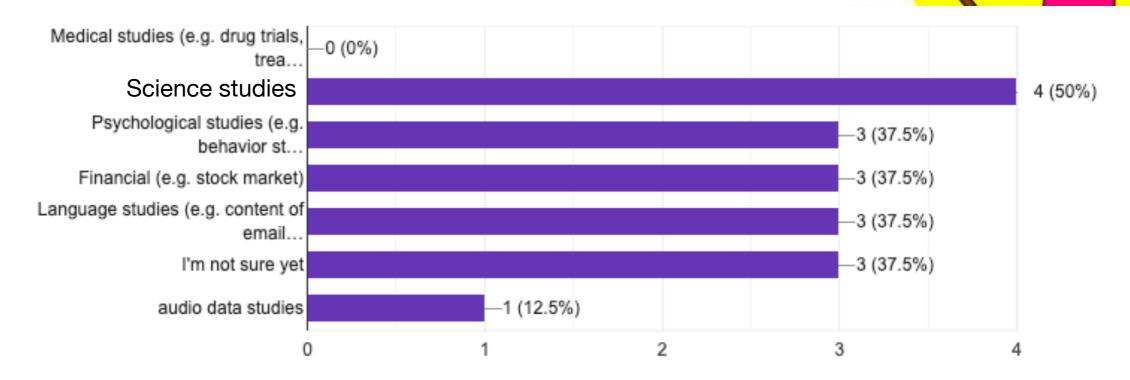
Our survey was just a **sample**



In fact a **convenience sample** (not a **random sample**)

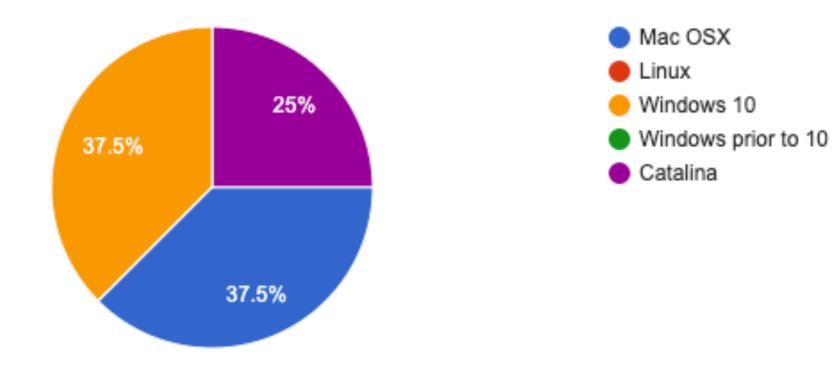
What kinds of datasets are you looking forward to visualizing? Check all the





Which operation system do you use on your laptop?

8 responses



Real Data! (ooooo!)

Dataset #1:

http://data.un.org/

to

http://data.un.org/Explorer.aspx

to

Commodity trade stats => Fish, crustaceans, mollusks, etc

Dataset #2:

Excerpt of the Introduction of "A Void" by Georges Perec

Georges Perec

A VOID

Translated from the French by Gilbert Adair

http://data.un.org/Data.aspx?d=ComTrade&f=_I1Code%3a4

or

https://tinyurl.com/y8vo7v8u

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							336777 records Page 1 of 6
Country or Area	Year	Commodity	Flow	Trade (USD)	Weight (kg)	Quantity Name	Quantity
Afghanistan	2010	Trout, fresh or chilled, whole	Import	8,600	9,000	Weight in kilograms	9,000
Albania	2016	Fish live, except trout, eel or carp	Import	2,202,944	39,896	Weight in kilograms	39,896
Albania	2016	Trout, fresh or chilled, whole	Export	1,973,381	266,283	Weight in kilograms	266,283
Albania	2016	Salmon fresh or chilled, whole	Import	387,938	96,083	Weight in kilograms	96,083
Albania	2016	Salmon fresh or chilled, whole	Export	39,162	5,974	Weight in kilograms	5,974
Albania	2016	Salmonidae, not trout or salmon,fresh or chilled whol	Import	70,993	11,065	Weight in kilograms	11,065
Albania	2016	Salmonidae, not trout or salmon,fresh or chilled whol	Export	49,287	10,808	Weight in kilograms	10,808
Albania	2016	Sardines, brisling, sprats, fresh or chilled, whole	Import	305,172	457,234	Weight in kilograms	457,234
Albania	2016	Fish nes, fresh or chilled, whole	Import	2,654,089	1,322,338	Weight in kilograms	1,322,338
Albania	2016	Fish nes, fresh or chilled, whole	Export	1,353,539	241,436	Weight in kilograms	241,436
Albania	2016	Fish livers and roes, fresh or chilled	Import	3,525	51	Weight in kilograms	51
Albania	2016	Salmon Atlantic or Danube, frozen, whole	Export	22,526	2,445	Weight in kilograms	2,445
Albania	2016	Fish nes, frozen, whole	Import	1,114,294	698,868	Weight in kilograms	698,868
Albania	2016	Fish nes, frozen, whole	Export	631,255	119,205	Weight in kilograms	119,205
Albania	2016	Fish fillets, frozen	Export	147,044	22,523	Weight in kilograms	22,523
Albania	2016	Fish meat & mince, except liver, roe & fillets, froze	Export	111,045	20,212	Weight in kilograms	20,212
Albania	2016	Fish fillets, dried, salted or in brine, not smoked	Export	7,529,293	558,797	Weight in kilograms	558,797
Albania	2016	Salmon, smoked, including fillets	Import	14,497	779	Weight in kilograms	779
Albania	2016	Anchovies, salted or in brine, not dried or smoked	Import	18,934,237	5,799,550	Weight in kilograms	5,799,550

Discuss:

What is the population?

What is a case?

What are each type of variable (numerical, categorical, etc)

Note: could be multiple answers!

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							336777 records Page 1 of 6736
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Poll: Type of variable "Country or Area"

- 1. Continuous Numerical
- 2. Discrete Numerical
- 3. Nominal (unordered) Categorical
- 4. Ordinal (ordered) Categorical

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						336	777 records Page 1 of 6736 🕨
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Poll: Type of variable "Trade USD"

- 1. Continuous Numerical
- 2. Discrete Numerical
- 3. Nominal (unordered) Categorical
- 4. Ordinal (ordered) Categorical

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						3	336777 records Page 1 of 6736 >
Country or Area	Year	Commodity	Flow	Trade (USD)	Weight (kg)	Quantity Name	Quantity
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Poll: Type of variable "Year"

- 1. Continuous Numerical
- 2. Discrete Numerical
- 3. Nominal (unordered) Categorical
- 4. Ordinal (ordered) Categorical

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336777 records Page 1 of 6736							
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What are some questions you might want to use your new R-skills (plotting, calculations) and summary statistics knowledge (mean, median, standard deviation, quartiles) to probe with this dataset?

INTRODUCTION

In which, as you will soon find out, Damnation has its origin

Today, by radio, and also on giant hoardings, a rabbi, an admiral notorious for his links to Masonry, a trio of cardinals, a trio, too, of insignificant politicians (bought and paid for by a rich and corrupt Anglo-Canadian banking corporation), inform us all of how our country now risks dying of starvation. A rumour, that's my initial thought as I switch off my radio, a rumour or possibly a hoax. Propaganda, I murmur anxiously - as though, just by saying so, I might allay my doubts - typical politicians' propaganda. But public opinion gradually absorbs it as a fact. Individuals start strutting around with stout clubs. "Food, glorious food!" is a common cry (occasionally sung to Bart's music), with ordinary hard-working folk harassing officials, both local and national, and cursing capitalists and captains of industry. Cops shrink from going out on night shift. In Macon a mob storms a municipal building. In Rocadamour ruffians rob a hangar full of foodstuffs, pillaging tons of tuna fish, milk and cocoa, as also a vast quantity of corn - all of it, alas, totally unfit for human consumption. Without fuss or ado, and naturally without any sort of trial, an indignant crowd hangs 26 solicitors on a hastily built scaffold in front of Nancy's law courts (this Nancy is a town, not a woman) and ransacks a local journal, a disgusting right-wing rag that is siding against it. Up and down this land of ours looting has brought docks, shops and farms to a virtual standstill.

Discuss:
What is the population?
What is a case?
What are each type of variable
(numerical, categorical, etc)

Note: could be multiple answers or somewhat nebulous answers!

INTRODUCTION

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What are some questions you might want to use your new R-skills (plotting, calculations) and summary statistics knowledge (mean, median, standard deviation, quartiles) to probe with this dataset?

Real Data! (ooooo!)

We'll go through this one

Dataset #1:

http://data.un.org/

to

http://data.un.org/Explorer.aspx

to

Commodity trade stats => Fish, crustaceans, mollusks, etc

We may or may not get to this one

Dataset #2:

Excerpt of the Introduction of "A Void" by Georges Perec

Georges Perec

A VOID

Translated from the French by Gilbert Adair

Real Data! (00000!)

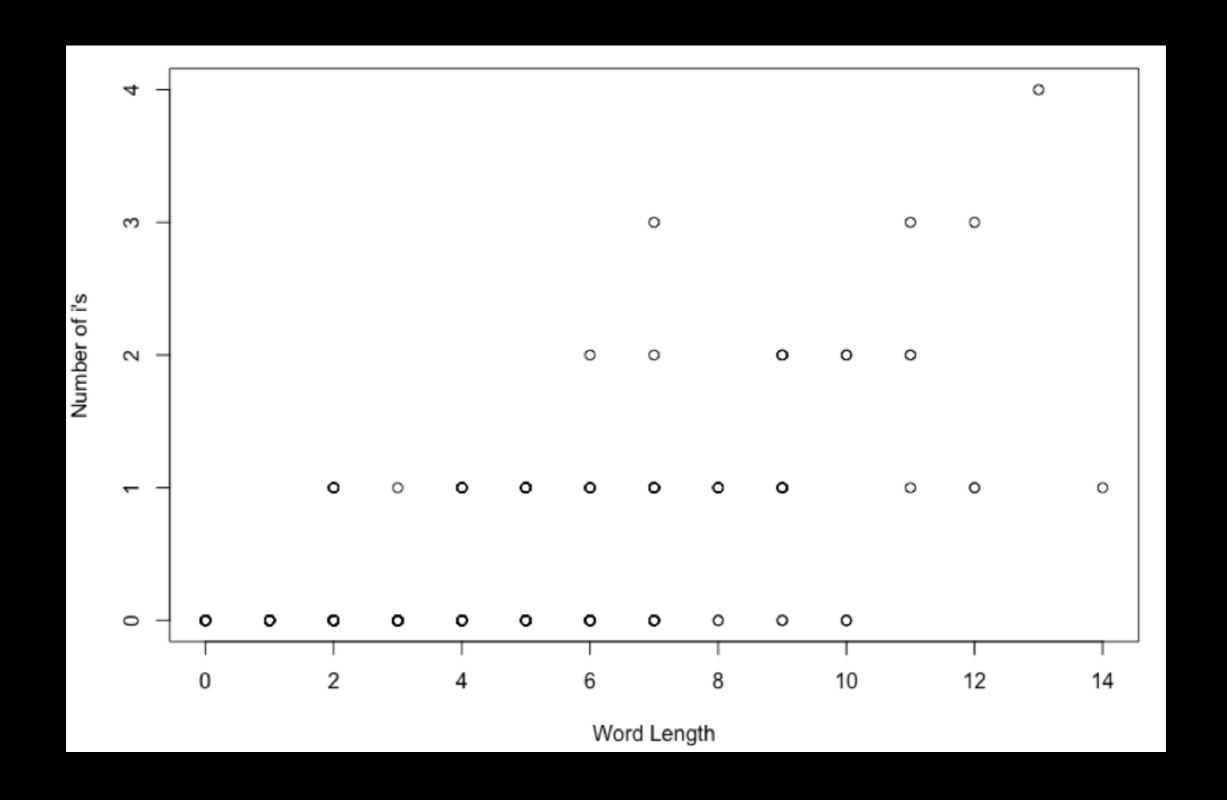


Steps:

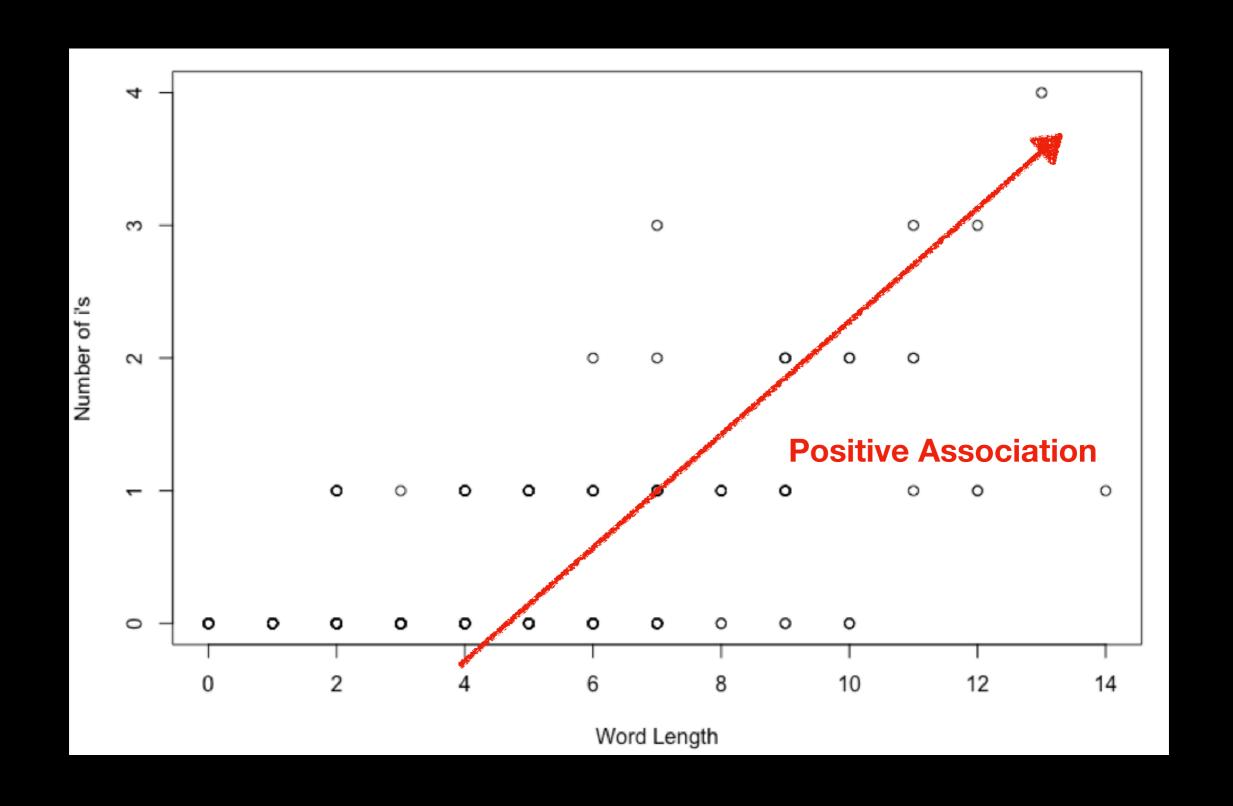
- 1. Make sure everybody can download and open the data.
- 2. Make sure everybody can run the example code and look at the plots.
- 3. Come up with 1 (or more) questions/ideas/features you want to explore more with your dataset using R, plotting, summary statistics.
- 4. Write down the idea in the chat window so we can discuss as a group.
- 5. I'll go through some basic statistics things with this dataset feel free to follow along or explore your idea. If exploring your own idea, please put in the chat window what you are doing and any questions you have.

Confusion at this stage is TOTALLY normal!

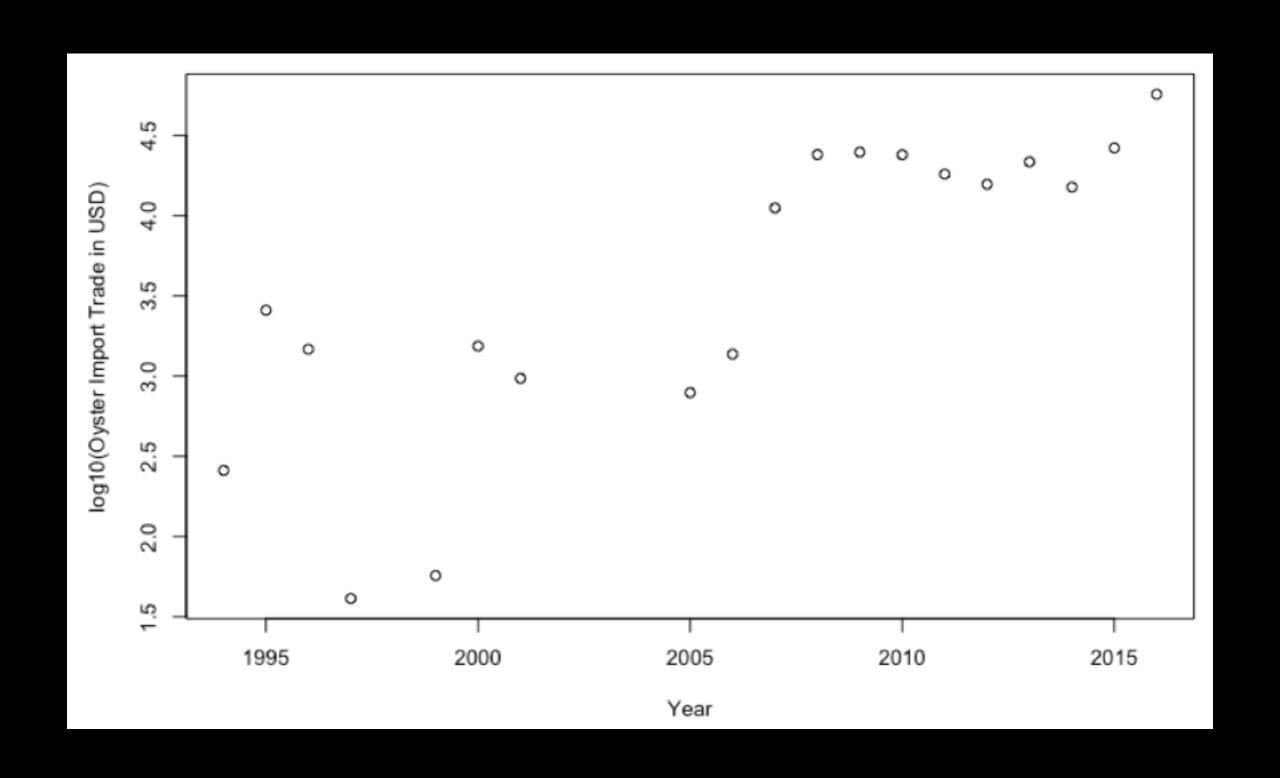
A Void Dataset: Number of "i"s vs. Length of Word



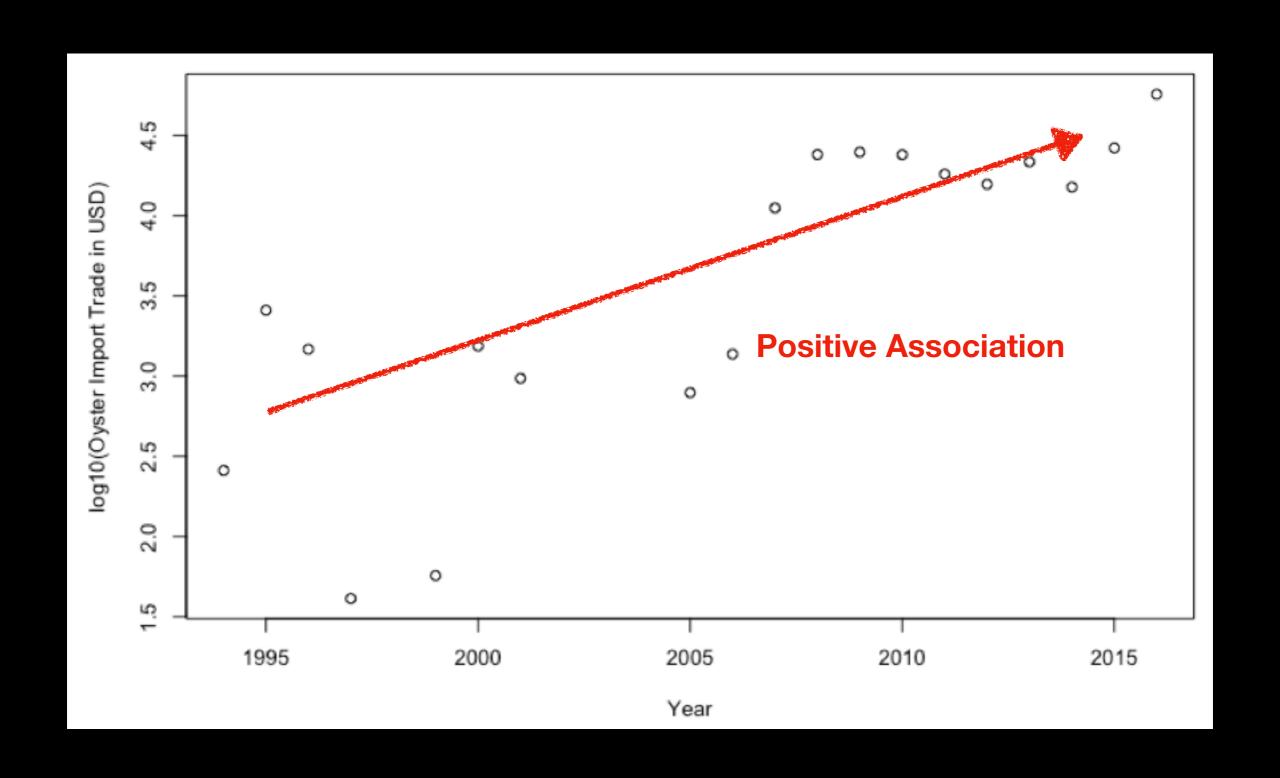
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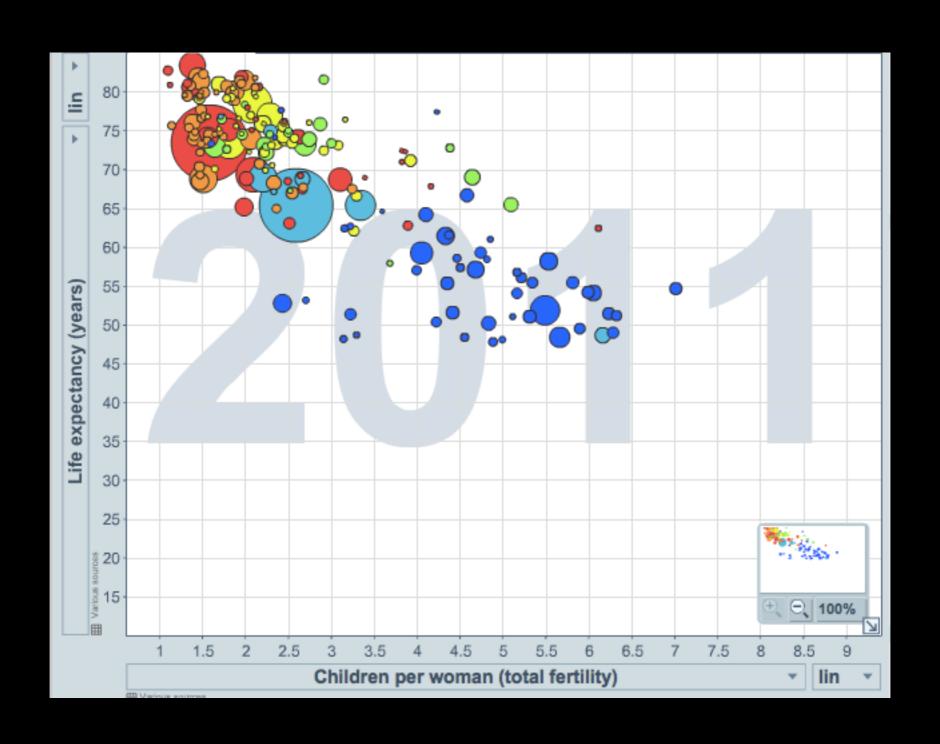
Fish Dataset: Croatian Oyster Trade in USD vs. Time



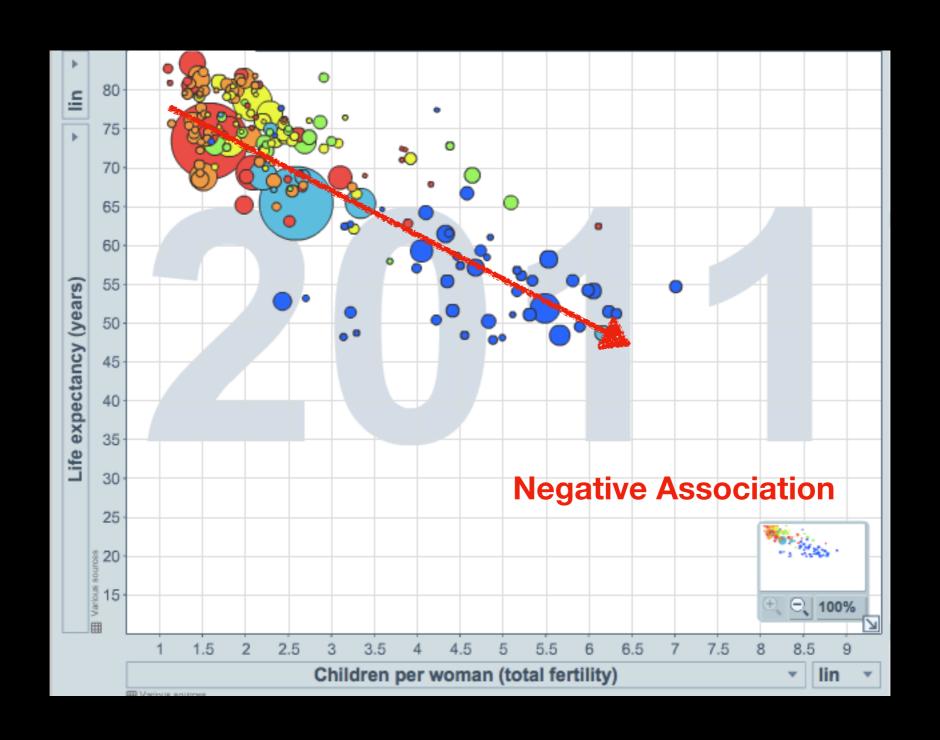
Fish Dataset: Croatian Oyster Trade in USD vs. Time



Fish Dataset: Life expectancy vs. Children Birthed per Woman



Fish Dataset: Life expectancy vs. Children Birthed per Woman



Fish Dataset: Life expectancy vs. Children Birthed per Woman

Associated or independent, not both

A pair of variables are either related in some way (associated) or not (independent). No pair of variables is both associated and independent.



There are more airplane deaths now than there were 50 years ago. Does that mean airplane travel is becoming more dangerous?

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The death rate in the Navy during the Spanish-American war was 9/1000. For civilians in the same time period in New York City it was 16/1000. Does that mean it is safer to be in the Navy?

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The death rate in the Navy during the Spanish-American war was 9/1000. For civilians in the same time period in New York City it was 16/1000. Does that mean it is safer to be in the Navy?

No, civilians include the old and sick while the Navy was comprised of healthy, young men.

Make sure to compare like to like.

1975 study of 1286 British women.

- ★ 23% of smokers died by 20-year follow-up
- ★ 29% of nonsmokers died by 20-year follow-up

- So do smokers tend to live longer?
- What is a potential confounding factor?

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Explanatory

Response

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- What is a potential confounding factor?

1975 study of 1286 British women.

- ★ 23% of smokers died by 20-year follow-up
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- So do smokers tend to live longer?
- What is a potential confounding factor?

TIP: Explanatory and response variables

To identify the explanatory variable in a pair of variables, identify which of the two is suspected of affecting the other and plan an appropriate analysis.

```
explanatory might affect response variable variable
```

	Died in 20 yrs	Alive	Total Smoker or Non
Non-smokers	205	499	704
Smokers	138	444	582
Total Alive or Dead	343	943	1286

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Non-smokers	205	499	704
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What percentage of the study participants are smokers? Non-smokers?

	Died in 20 yrs	Alive	Total Smoker or Non
Non-smokers	205	499	704/1286 = 55%
Smokers	138	444	582/1286 = 45%
Total Alive or Dead	343	943	1286

What percentage of the study participants are smokers? Non-smokers?

	Died in 20 yrs	Alive	Total Smoker or Non
Non-smokers	205	499	704
Smokers	138	444	582
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What percentage of the study participants are alive? Dead?

	Died in 20 yrs	Alive	Total Smoker or Non
Non-smokers	205	499	704
Smokers	138	444	582
Total Alive or Dead	343/1286 = 27%	943/1286 = 73%	1286

What percentage of the study participants are alive? Dead?

	Died in 20 yrs	Alive	Total Smoker or Non
Non-smokers	205	499	704
Smokers	138	444	582
Total Alive or Dead	343	943	1286

What percentage of non-smokers (and smokers) are alive?

	Died in 20 yrs	Alive	Total Smoker or Non
Non-smokers	205	499/704 = 71%	704
Smokers	138	444/582 = 76%	582
Total Alive or Dead	343	943	1286

What percentage of non-smokers (and smokers) are alive?

	Died in 20 yrs	Alive	Total Smoker or Non
Non-smokers	205	499/704 = 71%	704
Smokers	138	444/582 = 76%	582
Total Alive or Dead	343	943	1286

What percentage of non-smokers (and smokers) are alive?

More alive smokers?

	Died in 20 yrs	Alive	Total in Row
Non-smokers Age 18-64	65	474	539
Non-smokers Age 65+	140	25	165
Smokers Age 18-64	96	437	533
Smokers Age 65+	42	7	49
Total In Column	343	943	1286

What percentage of non-smokers (and smokers) are alive from the <65 age group?

	Died in 20 yrs	Alive	Total in Row
Non-smokers Age 18-64	65	474/539 = 88%	539
Non-smokers Age 65+	140	25	165
Smokers Age 18-64	96	437/533 = 82%	533
Smokers Age 65+	42	7	49
Total In Column	343	943	1286

What percentage of non-smokers (and smokers) are alive from the <65 age group?

Now looks like non-smokers fair better than smokers.

Age is a confounding factor

	Died in 20 yrs	Alive	Total in Row
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What percentage of non-smokers (and smokers) are alive from the <65 age group?

Now looks like non-smokers fair better than smokers.

- A confounding factor is a variable associated with the both the explanatory <u>and</u> the response variable.
- Because of this, the response could be due to the supposed explanatory variable or to the confounding factor - the two are confounded.

- In the previous example the supposed <u>explanatory</u> variable is:
 - ★ Smoking

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 - ★ Smoking
- The <u>response</u> variable is:
 - ★ Dying within 20 years
- The <u>confounding</u> factor is:
 - ★ Age there were more older people in the non-smoking group and older people are more likely to pass away within 20 years.

Eating more fruit, particularly blueberries, apples and grapes, is linked to a reduced risk of developing type-2 diabetes, suggests a study in the British Medical Journal.



http://www.bbc.com/news/health-23880701

"Blueberries cut the risk (of type-2 diabetes) by 26%..."

"The research looked at the diets of more than 187,000 people in the US..."

"The studies used food frequency questionnaires to follow up the participants every four years, asking how often, on average, they ate a standard portion of each fruit..."

 Can we conclude that eating blueberries will reduce risk of type-2 diabetes?

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 - **★** NO
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- Can we conclude that eating blueberries will reduce risk of type-2 diabetes?
 - **★** NO
- Why not?
 - ★ This is an <u>observational study</u>, so we cannot draw causal conclusions. There are many potential confounding factors.
- Identify a possible confounding factor and explain how it could confound. Make sure to explicitly connect it with risk of type-2 diabetes.

Observation study vs. Experiment

 In an observational study researches watch/record information without imposing any treatment.

 In an experiment, researchers impose a treatment to try to draw a causal conclusion about the effect of the treatment.

Why would we carry out one or the other?

- You have a hypothesis: Prolonged smoking leads to dental problems.
- How will you collect data to test this hypothesis?
- Should we do an experiment?

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A case study

- You have a hypothesis: Prolonged smoking leads to dental problems.
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A case study

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- You could survey people and ask about smoking habits and dental health, but will this prove a causal relationship?
 - ★ Why not?
- Lots of possible confounding factors. Give an example.
- The best we do is an observational study and try to compare like to like older people to older people, women to women, etc.

Women, age, etc are blocking factors

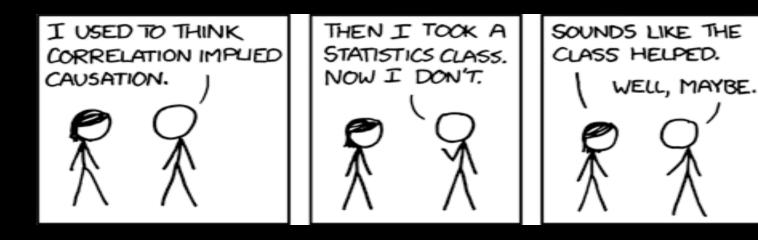
Blocking Variables

Blocking variables are characteristics that the experimental units come with, that we would like to control for.

Blocking is an effort to minimize confounding factors.

Correlation is not causation

Correlation is not causation!



http://xkcd.com/ 552/

 Observational studies alone cannot prove causation; only well designed experiments can prove causation.

More Experimental Design Terminology...

Placebo: fake treatment, often used as the control group for medical studies

Placebo effect: experimental units showing improvement simply because they believe they are receiving a special treatment

Blinding: when experimental units do not know whether they are in the control or treatment group

Double-blind: when both the experimental units and the researchers who interact with the patients do not know who is in the control and who is in the treatment group

A study is designed to test the effect of light level and noise level on exam performance of students. The researcher also believes that light and noise levels might have different effects on males and females, so wants to make sure both genders are equally represented in each group. Which of the below is correct?

- (1) There are 3 explanatory variables (light, noise, gender) and 1 response variable (exam performance)
- (2) There are 2 explanatory variables (light and noise), 1 blocking variable (gender), and 1 response variable (exam performance)
- (3) There is 1 explanatory variable (gender) and 3 response variables (light, noise, exam performance)
- (4) There are 2 blocking variables (light and noise), 1 explanatory variable (gender), and 1 response variable (exam performance)

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- (4) There are 2 blocking variables (light and noise), 1 explanatory variable (gender), and 1 response variable (exam performance)

Protein May Hold the Key to Who Gets Alzheimer's By PAM BELLUCK, MARCH 19, 2014, New York Times

It is one of the big scientific mysteries of Alzheimer's disease: Why do some people whose brains accumulate the plaques and tangles so strongly associated with Alzheimer's not develop the disease?

Now, a series of studies by Harvard scientists suggests a possible answer, one that could lead to new treatments if confirmed by other research....

The research, <u>published on Wednesday</u> in the journal Nature, focuses on a protein previously thought to act mostly in the brains of developing fetuses. The scientists found that the protein also appears to protect neurons in healthy older people from aging-related stresses. But in people with Alzheimer's and other dementias, the protein is sharply depleted in key brain regions.

Experts said if other scientists could replicate and expand upon the findings, the role of the protein, called REST, could spur development of new drugs for dementia, which has so far been virtually impossible to treat. But they cautioned that much more needed to be determined...

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• What type of study is this, observational study or an experiment?

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Yes, there is an observed association.

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Is there an association between lack of REST protein and having Alzheimer's?

Yes, there is an observed association.

What can we conclude?

4 possible explanations

1. The lack of REST protein causes Alzheimer's.

2. Alzheimer's causes a depletion in the REST protein.

3. A third variable is responsible for both the lack of REST protein and the development of Alzheimer's. What could it be? An extraneous variable that affects both the explanatory and the response variable and that make it seem like there is a relationship between the two (confounding factor).

4. It is pure coincidence.