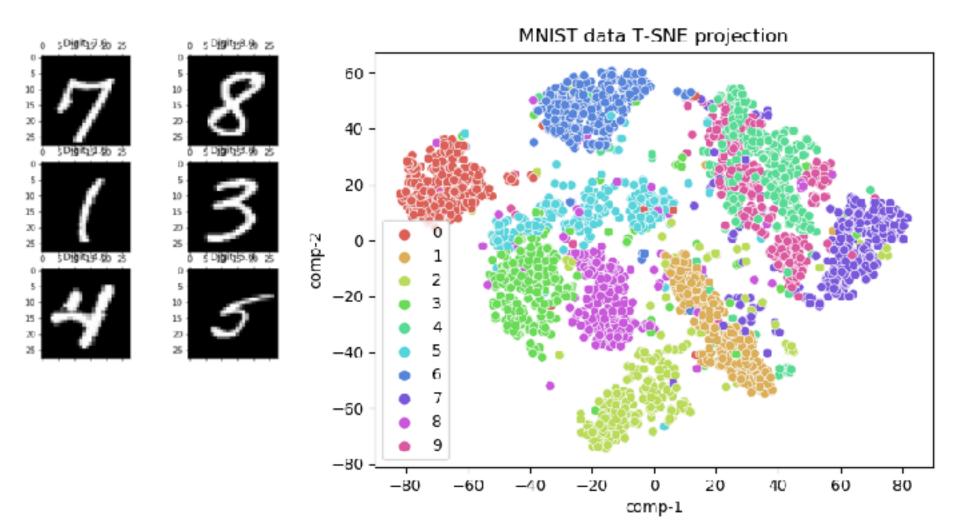
Course Announcements

- Due Friday (11:59 PM)
 - o **D4**
 - o A2

Grading underway: Project Proposals



- 1. Hyperparameters really matter
- 2. Cluster sizes in a UMAP plot mean nothing
- 3. Distances between clusters might not mean anything
- 4. Random noise doesn't always look random.
- 5. You may need more than one plot

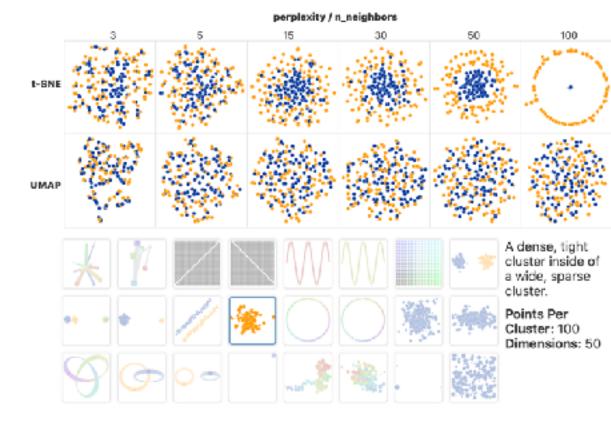


Figure 7: Comparison between UMAP and t-SNE projecting various toy datasets.

Inferential analysis

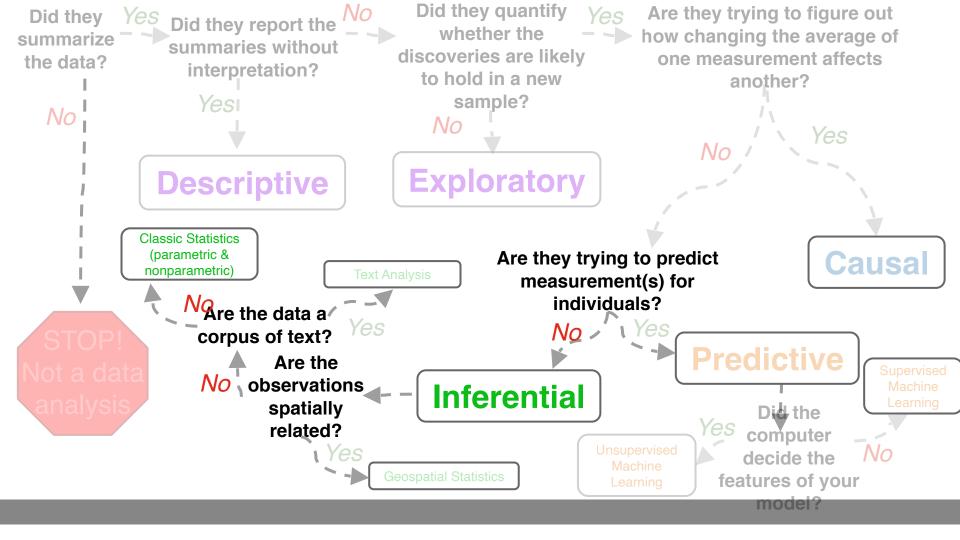
Jason G. Fleischer, Ph.D.

Asst. Teaching Professor Department of Cognitive Science, UC San Diego

ifleischer@ucsd.edu



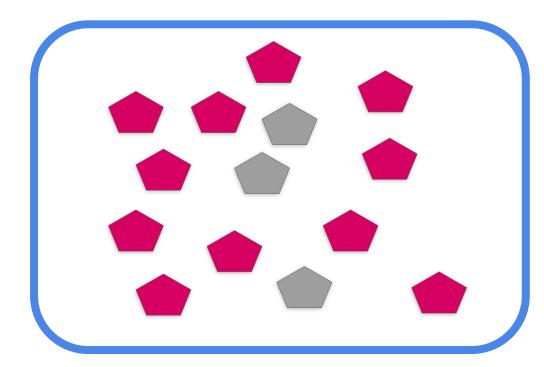
https://jgfleischer.com



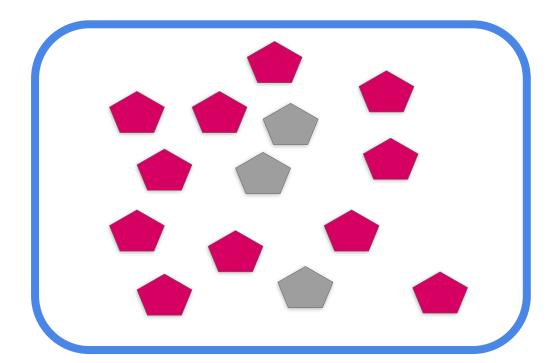
- Problem: Does Sesame Street affect kids brain development?
- Data science question: Is there a relationship between watching Sesame Street and test scores among children?
- Type of analysis: Inferential analysis



Sesame Street ?? Test scores viewership

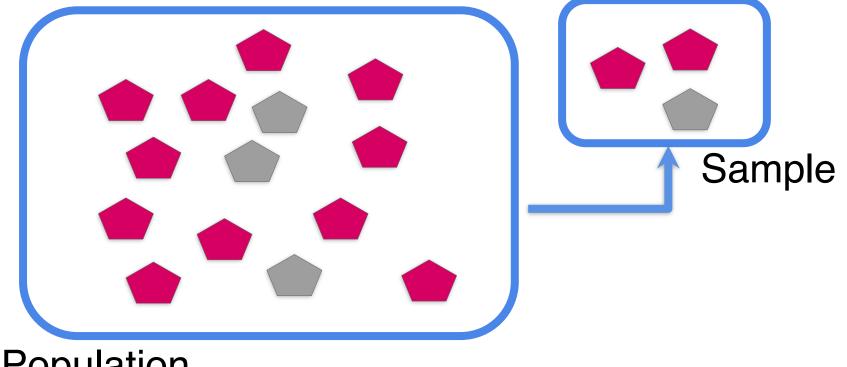


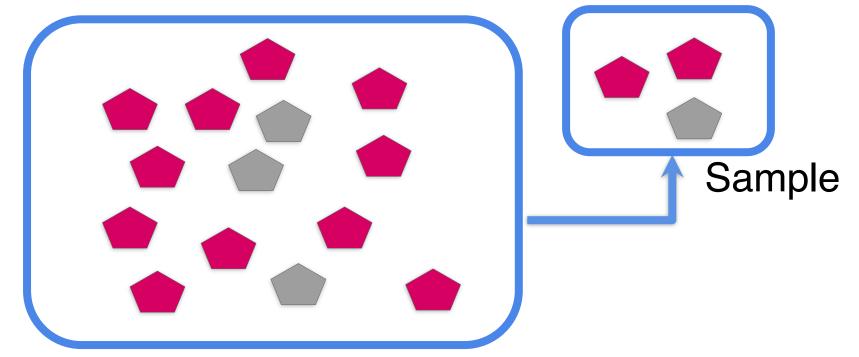
Population





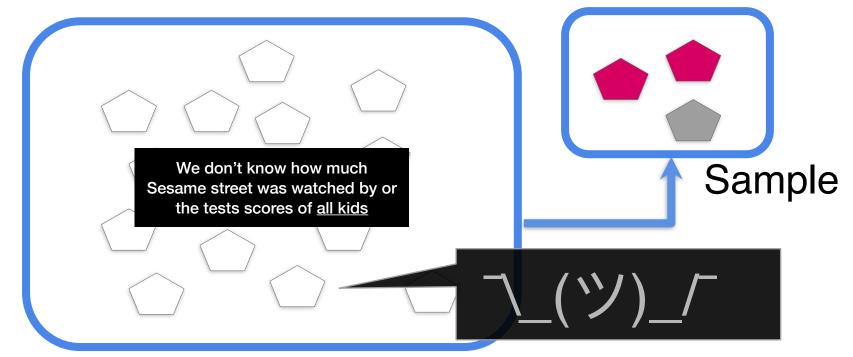
In our Sesame street example, the population would be all children





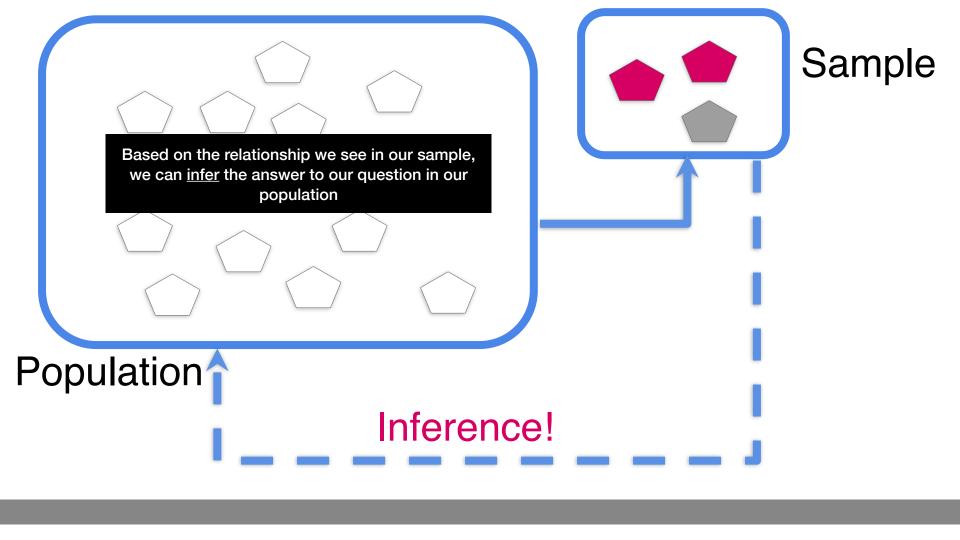


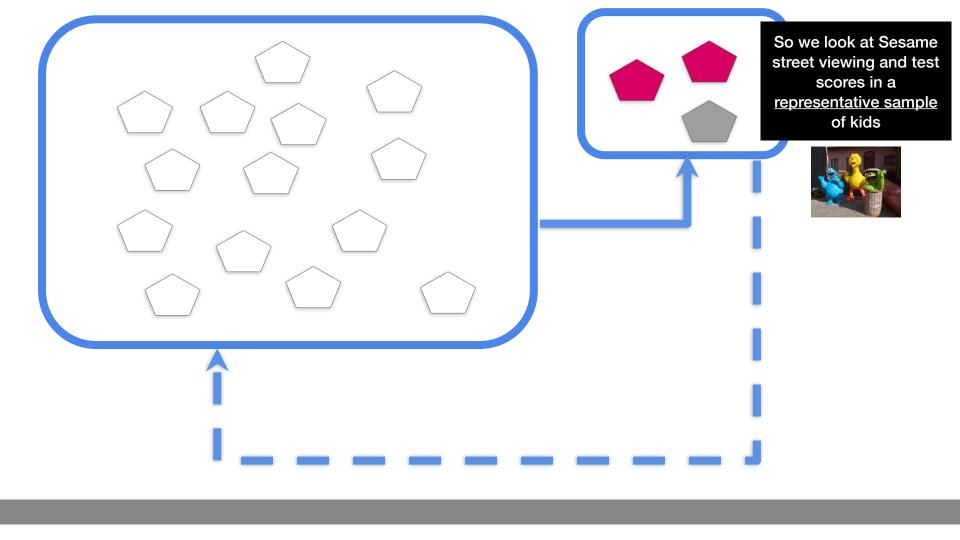
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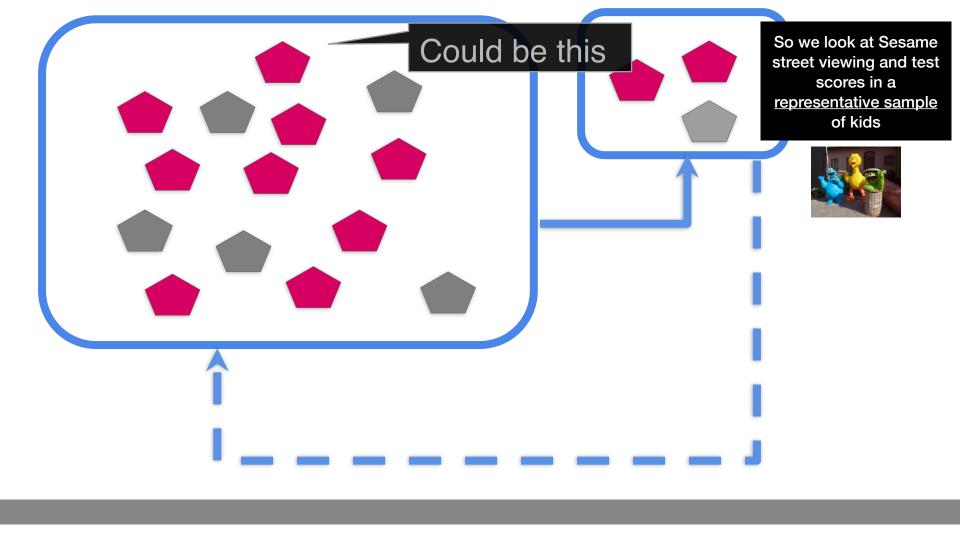


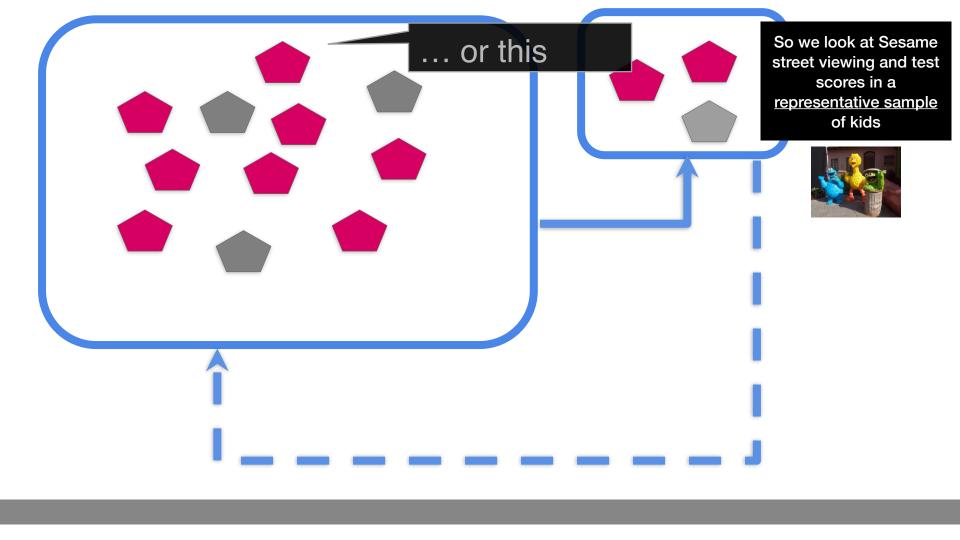


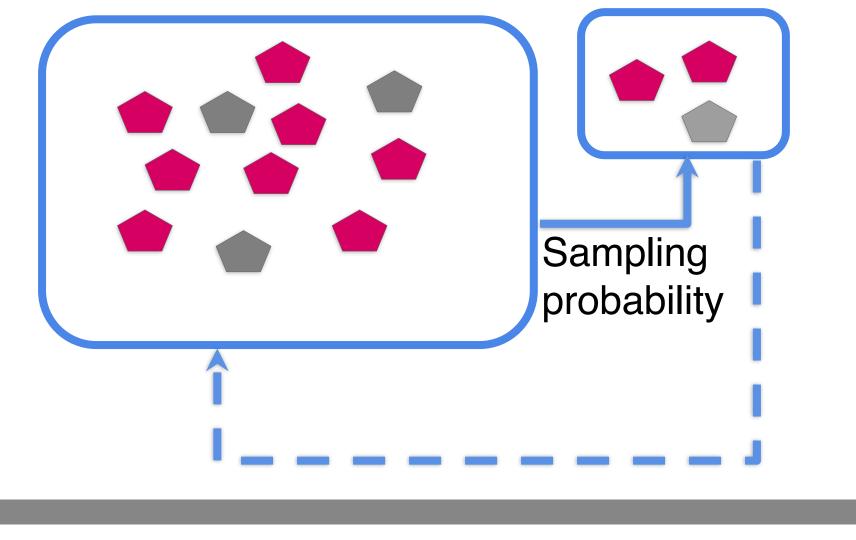
In our Sesame street example, the population would be all children

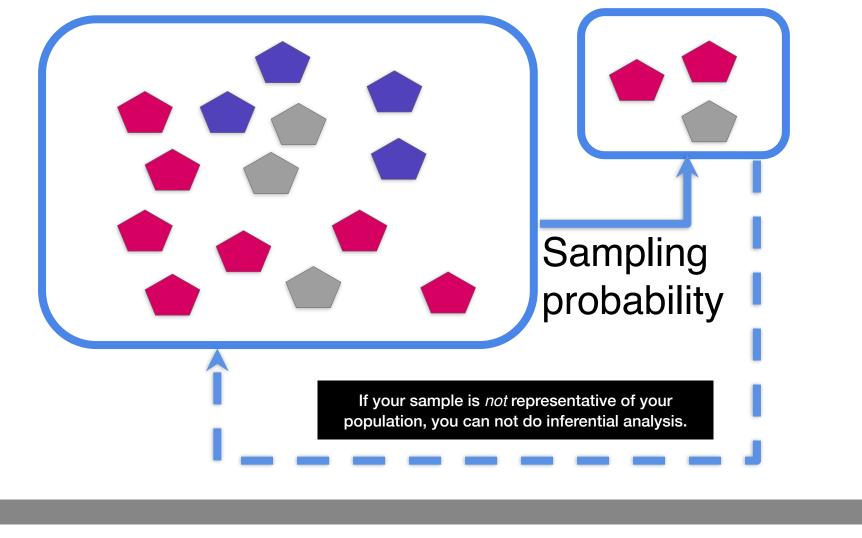














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The Effect of Air Pollution Control on Life Expectancy in the United States: An Analysis of 545 US counties for the period 2000 to 2007

Andrew W. Correia.

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Majid Ezzati, and

MRC-HPA Centre for Environment and Health and Department of Epidemiology and Biostatistics, Imperial College London, Norfolk Place, St Mary's Campus, London W2 1PG

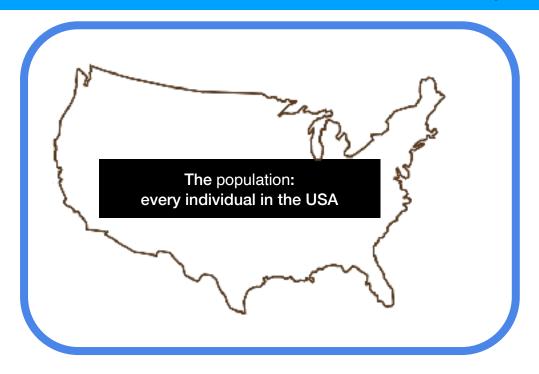
Francesca Dominici

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Air pollution ?? Lifespan

Is there a relationship between air pollution control and lifespan?

What if we want to know the effect of air pollution on everyone in the United States?





Random Sampler

"Girl Scouts were randomly sampled from across the United States"

What would be the likely population given this sample?

Approaches to Inference

CORRELATION

COMPARISON OF MEANS

REGRESSION

NON-PARAMETRIC TESTS

ASSOCIATION BETWEEN VARIABLES

i.e. Pearson Correlation, Spearman Correlation, chisquare test

DIFFERENCE IN MEANS BETWEEN VARIABLES

i.e. t-test, ANOVA

DOES CHANGE IN ONE VARIABLE MEAN CHANGE IN ANOTHER?

I.e. simple regression, multiple regression

FOR WHEN ASSUMPTIONS
IN THESE OTHER 3
CATEGORIES ARE NOT
MET

i.e. Wilcoxon ranksum test, Wilcoxon sign-rank test, sign test

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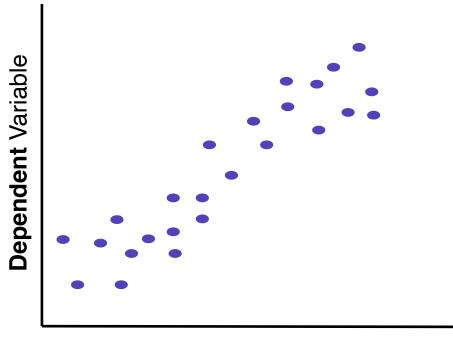
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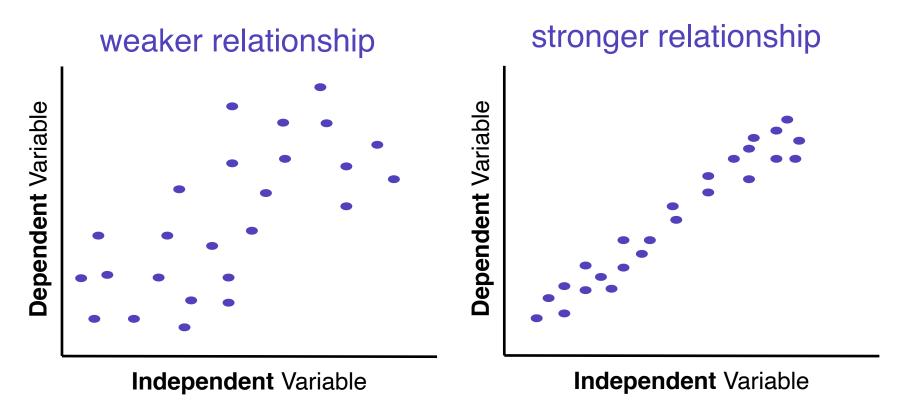
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i.e. Wilcoxon ranksum test, Wilcoxon sign-rank test, sign test



Independent Variable



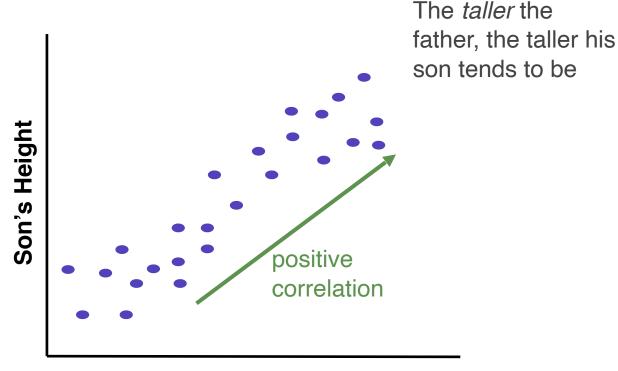
stronger relationship = higher correlation

Dependent Variable positive correlation

The *larger* the independent variable value, the *larger* the dependent variable tends to be

The *smaller* the independent variable value, the *smaller* the dependent variable tends to be

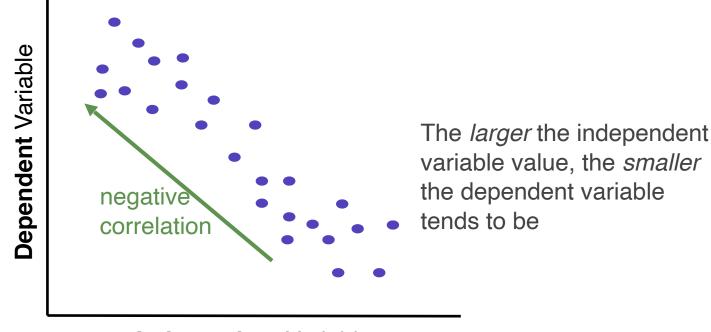
Independent Variable



The *shorter* the father, the shorter his son tends to be

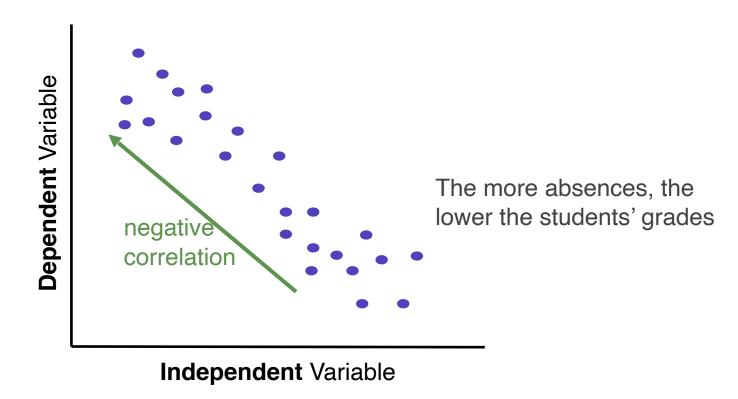
Father's Height

The *smaller* the independent variable value, the *larger* the dependent variable tends to be



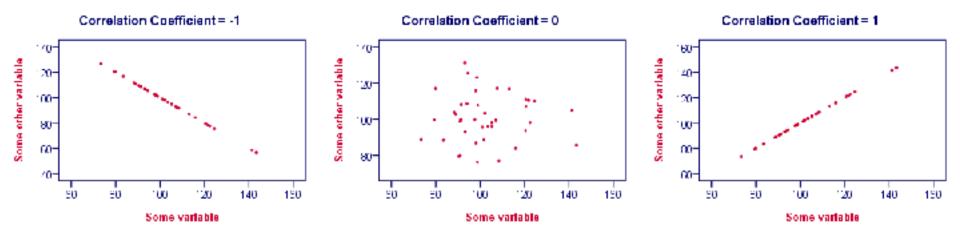
Independent Variable

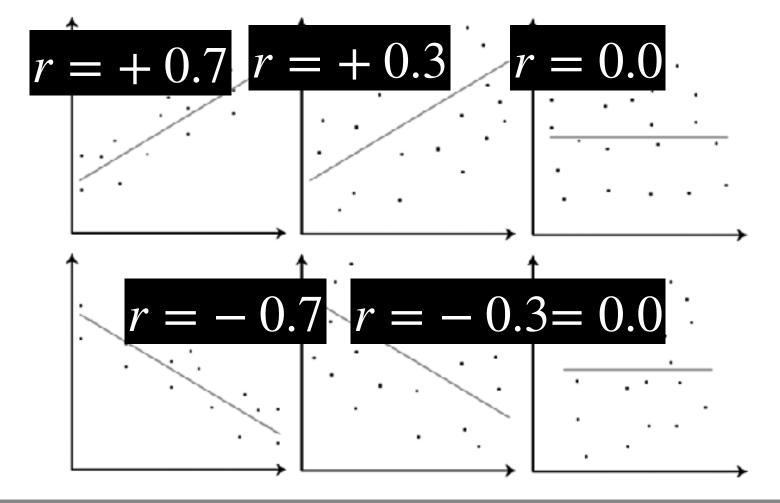
The *lower* the number of absences, the *higher* the students' grades tend to be



Pearson's *r*: linear correlation between two variables takes values [-1,1]

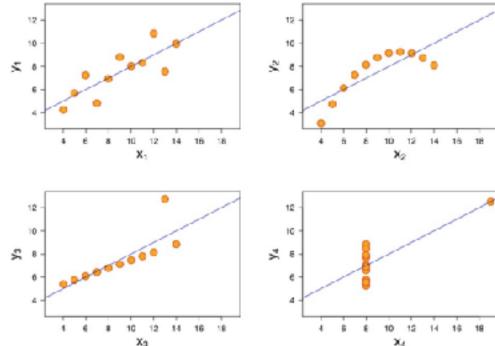
Correlation is how close the data are to being in a line... BUT IT HAS NOTHING TO DO WITH THE SLOPE





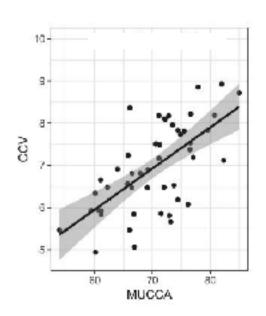
Anscombe's Quartet

Property	Value	
Meen of x in each case	0 экиеп	
Variance of a in each case	11 (pract)	
Meen of y in each case	7.50 to a continui otaces	
Variance of yin each case	4.122 Gr 4.127 (a 6 decimal picom)	
Correlation between x and y in each case	0.516 je 3 direktorá přanový	
Linear regression line in each case	g = 3.00 + 0.500 кго з выстанства раме, марокачер	

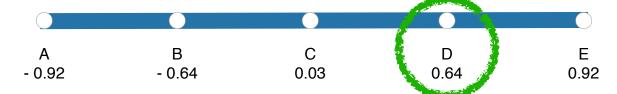


Correlation Champ



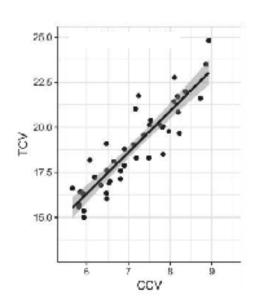


Which of the following is the Pearson correlation coefficient (r) for this relationship?

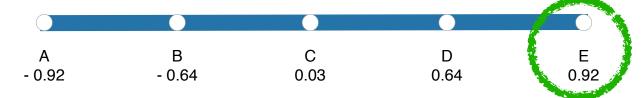


Correlation Champ



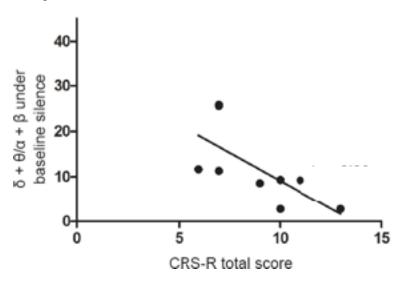


Which of the following is the Pearson correlation coefficient (r) for this relationship?

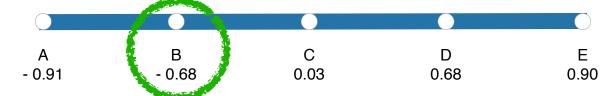


Correlation Champ





Which of the following is the Pearson correlation coefficient (r) for this relationship?



Correlation != Causation

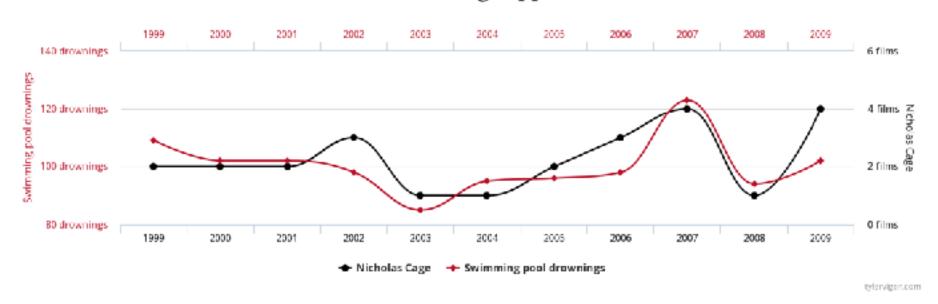
Correlation establishes a relationship.

It does **NOT** establish causation.

Number of people who drowned by falling into a pool

correlates with

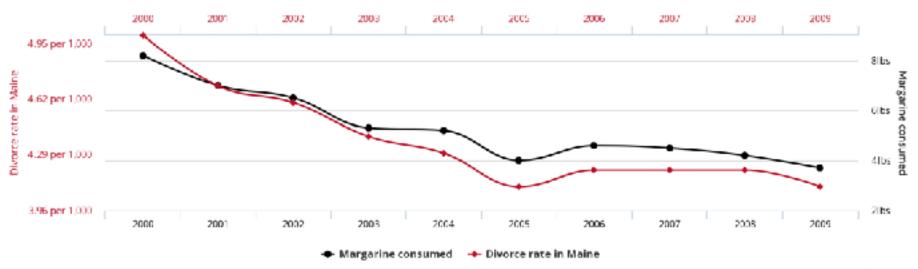
Films Nicolas Cage appeared in



Divorce rate in Maine

correlates with

Per capita consumption of margarine



CORRELATION

ASSOCIATION BETWEEN VARIABLES

i.e. Pearson Correlation, Spearman Correlation, chisquare test

COMPARISON OF MEANS

DIFFERENCE IN MEANS BETWEEN CONDITIONS

i.e. t-test, ANOVA

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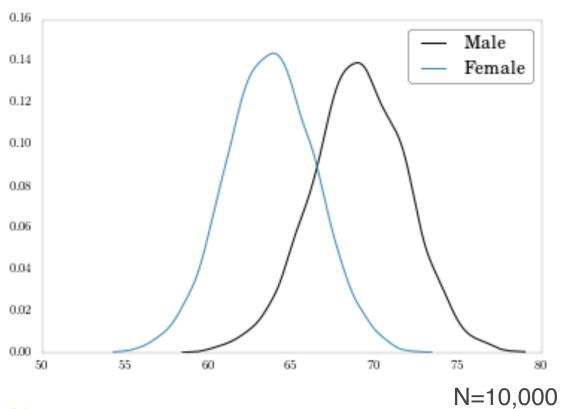
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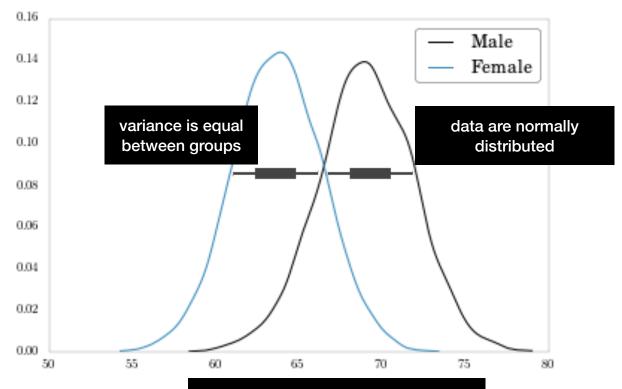
t-test

tests for difference in means between groups



t-test Assumptions

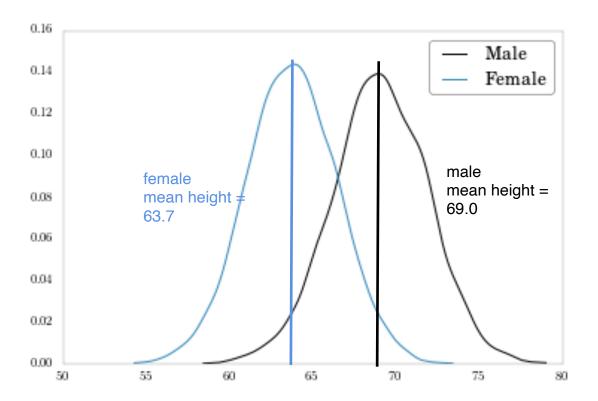
- 1. Data are continuous
- 2. Normally distributed
- 3. Large enough sample size
- 4. Equal variance b/w groups



sample size is large enough

N=10,000

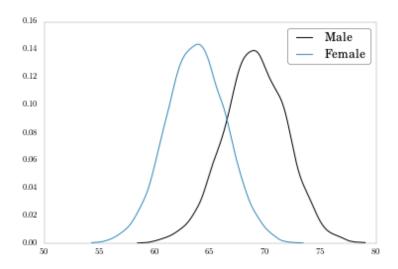
data are continuous



N=10,000

t-statistic: -95.6

p-value << 0.001



95% CI for true difference in means [-5.43, -5.21]

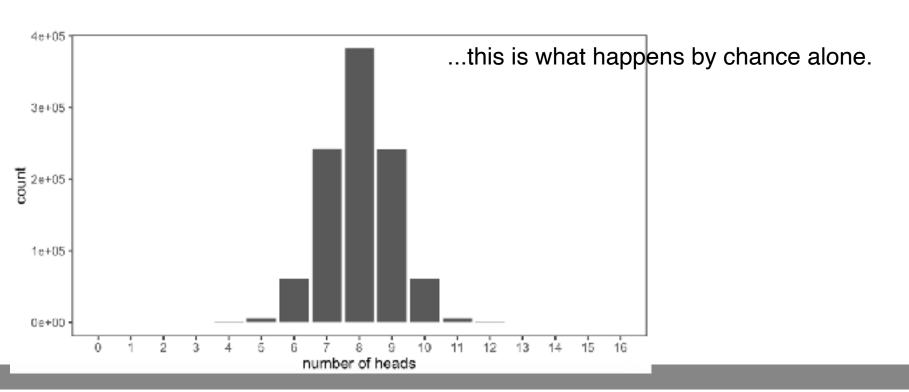
Yes.

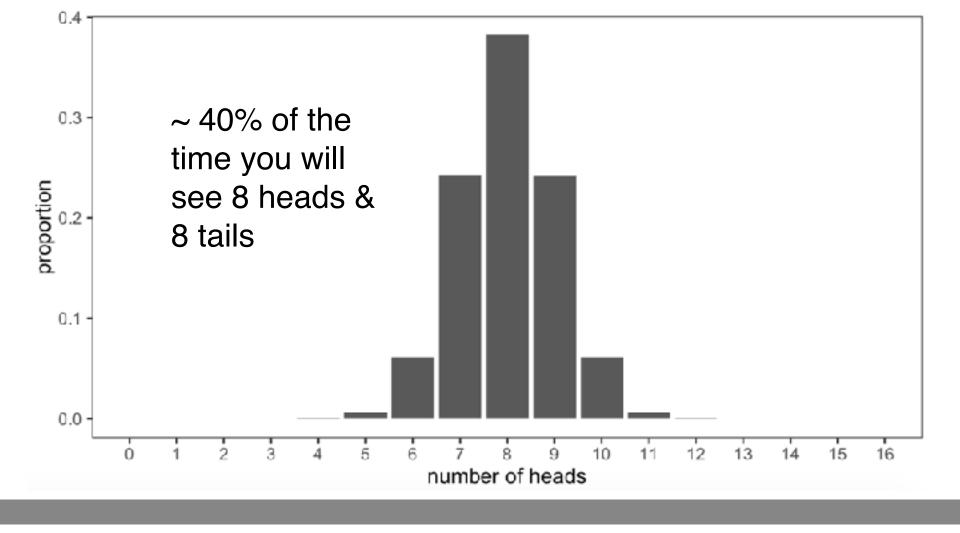
p-value: the probability of getting the observed results (or results more extreme) by chance alone

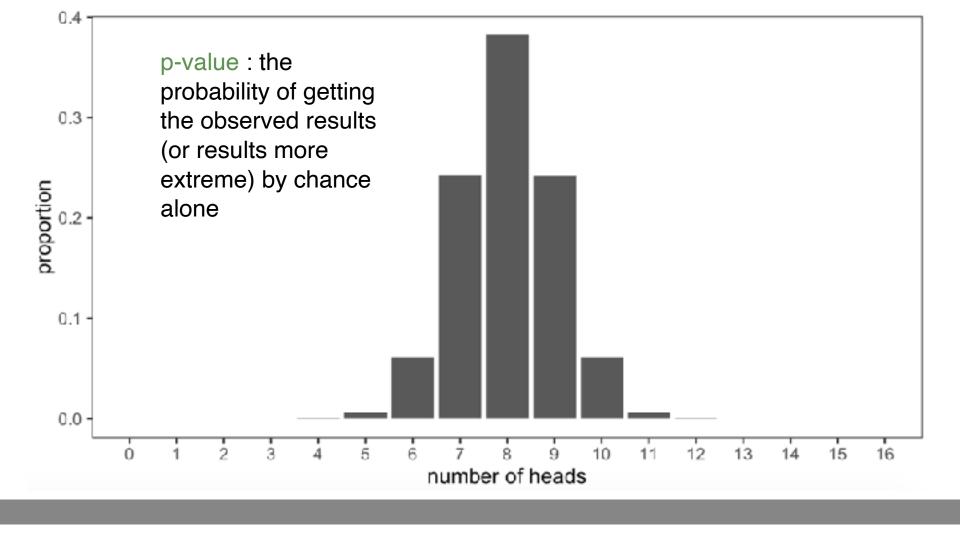


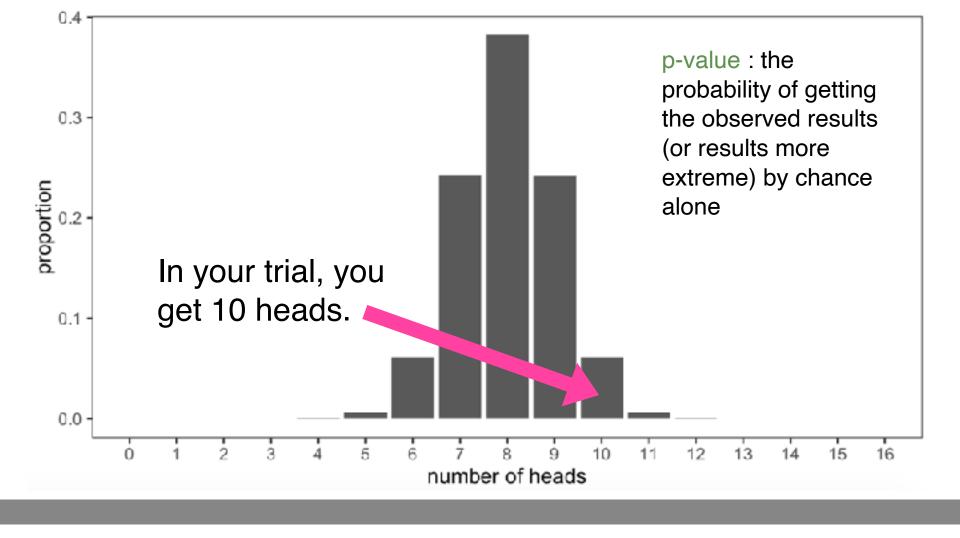
https://forms.gle/ 6MCyp7qFsaHgGKi5A

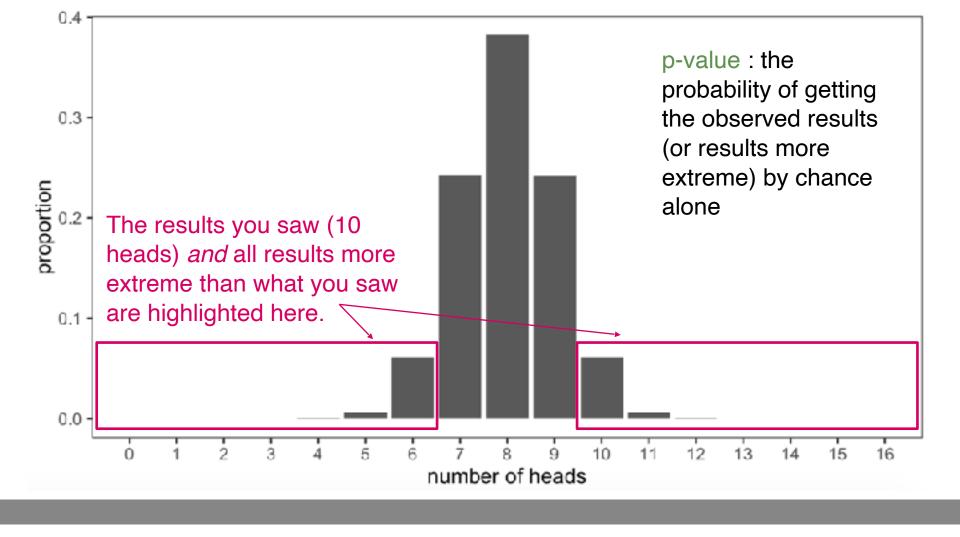
If we flip a coin 16 times and record the number of heads....and then do that 1M times

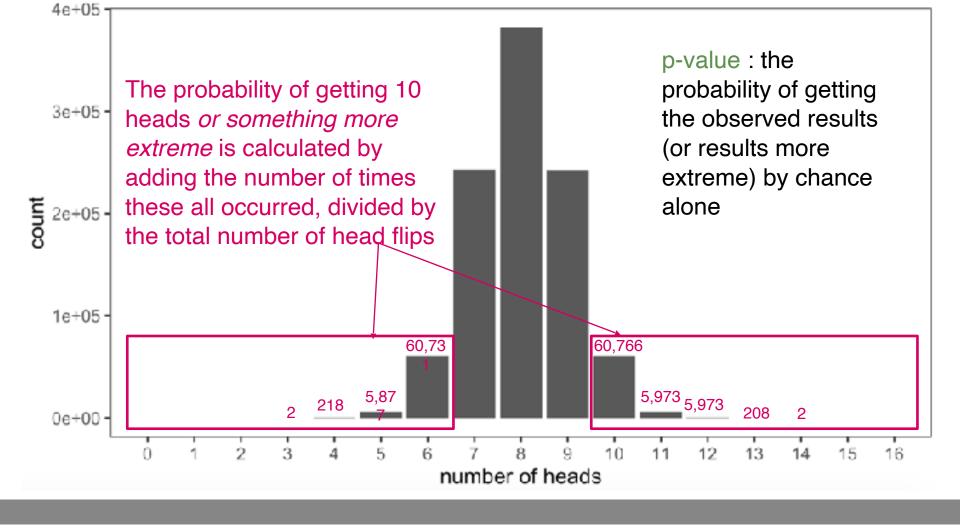


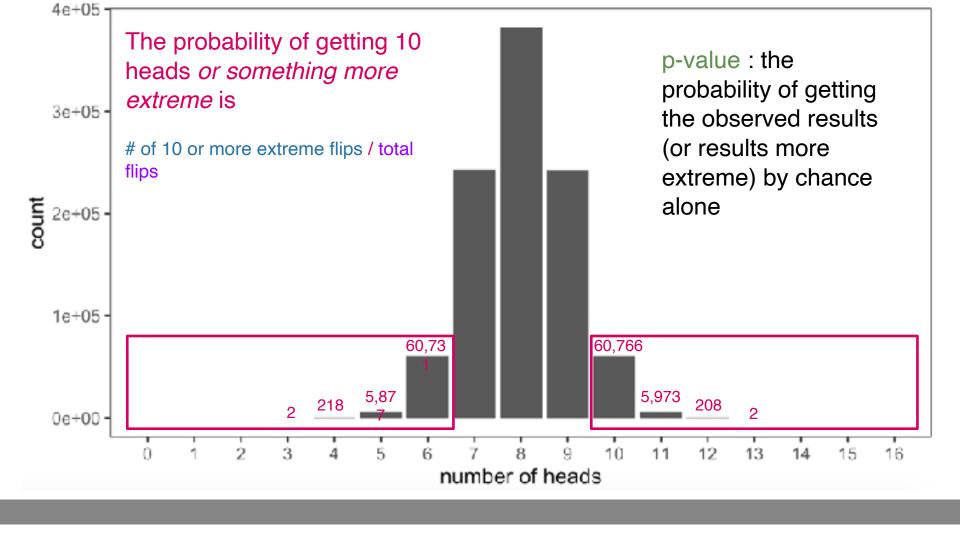


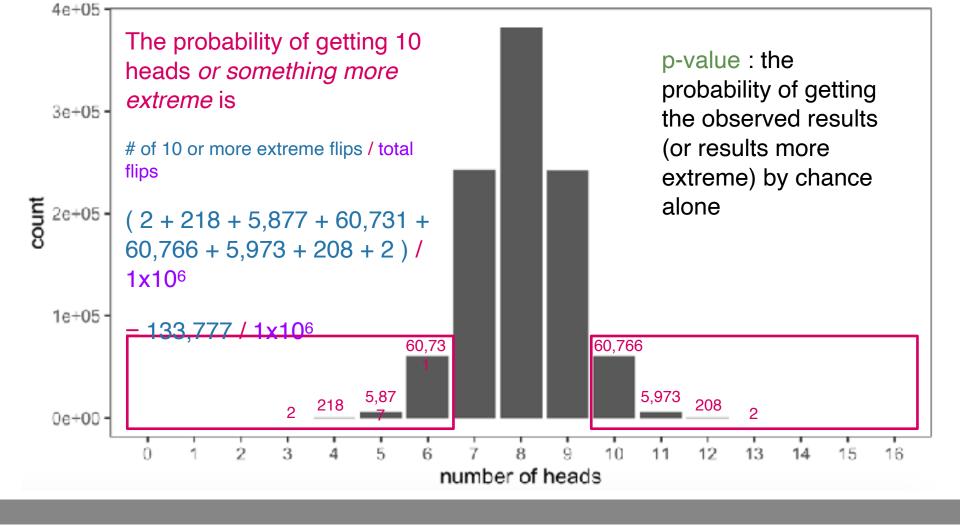


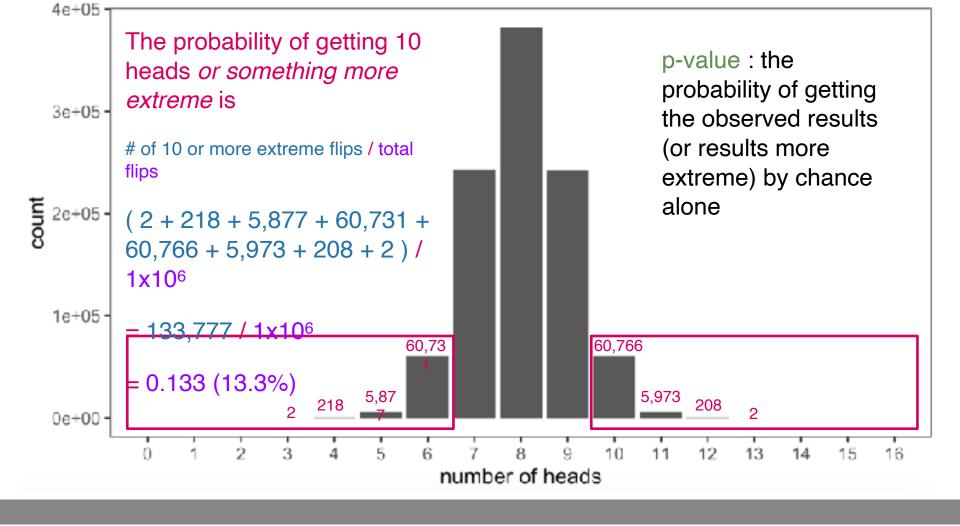


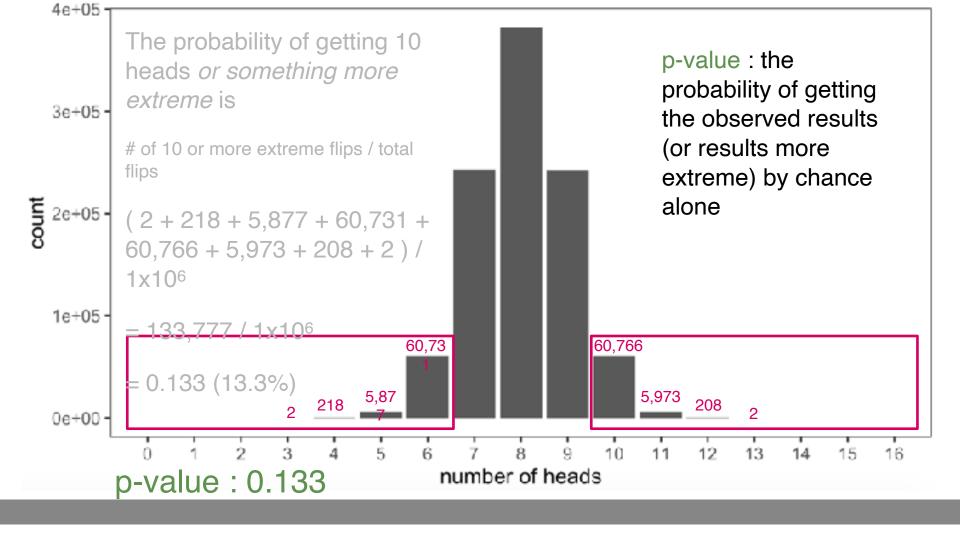


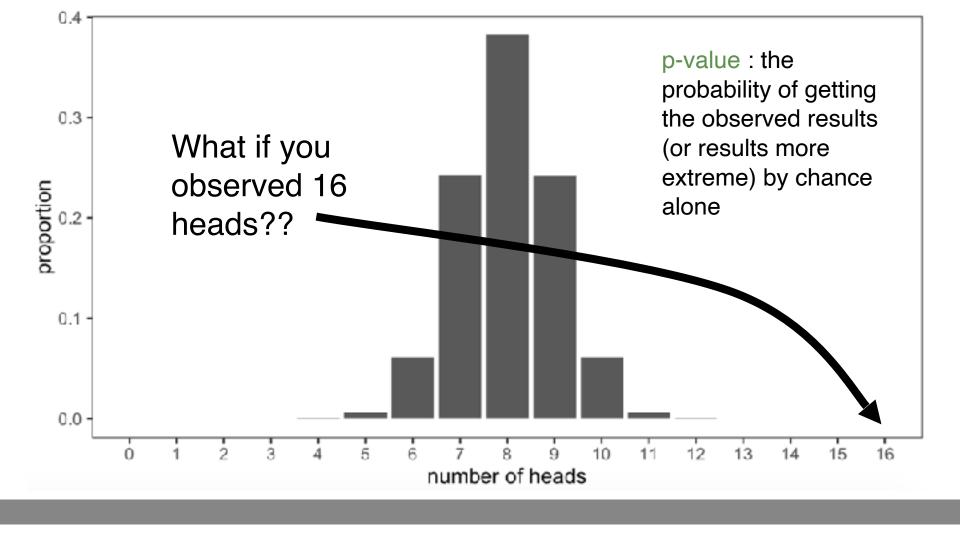


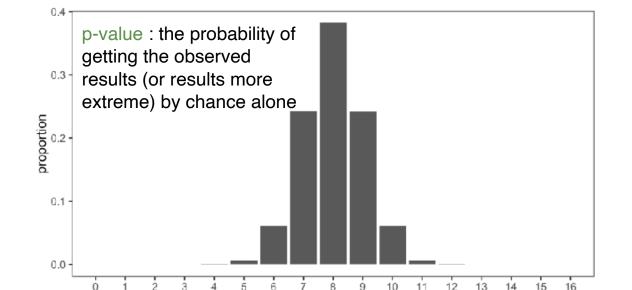














What would be the p-value of you flipping 16 heads?

number of heads



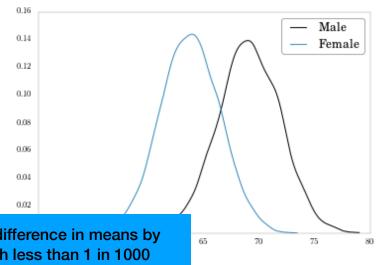
t-statistic: -95.6

p-value << 0.001

The probability of seeing this difference in means by random chance alone is much less than 1 in 1000

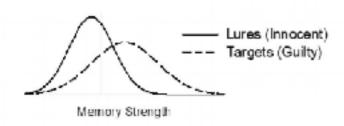
95% CI for true difference in means [-5.43, -5.21]

Yes.

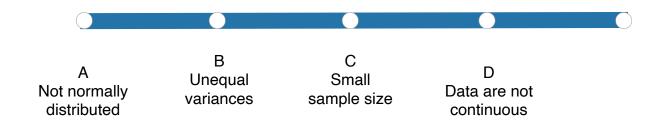


Difference in Means

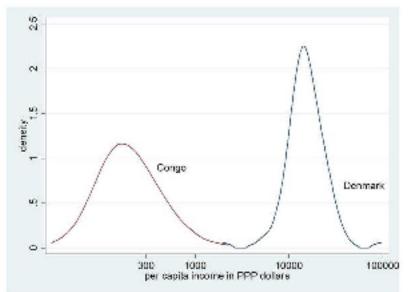




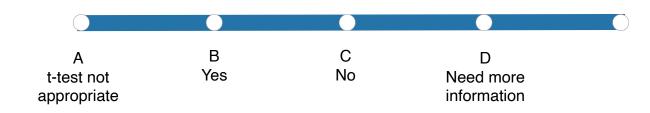
Why would a t-test *not* be appropriate for these data?







Would a t-test find a significant difference in means?



CORRELATION

COMPARISON OF MEANS

REGRESSION

NON-PARAMETRIC TESTS

ASSOCIATION BETWEEN VARIABLES

i.e. Pearson Correlation, Spearman Correlation, chisquare test

DIFFERENCE IN MEANS BETWEEN VARIABLES

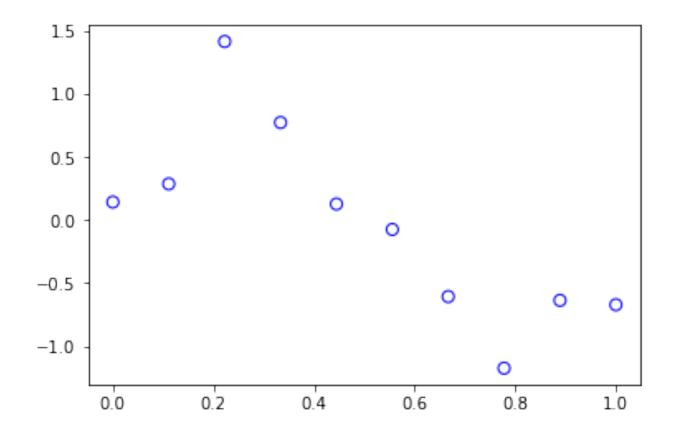
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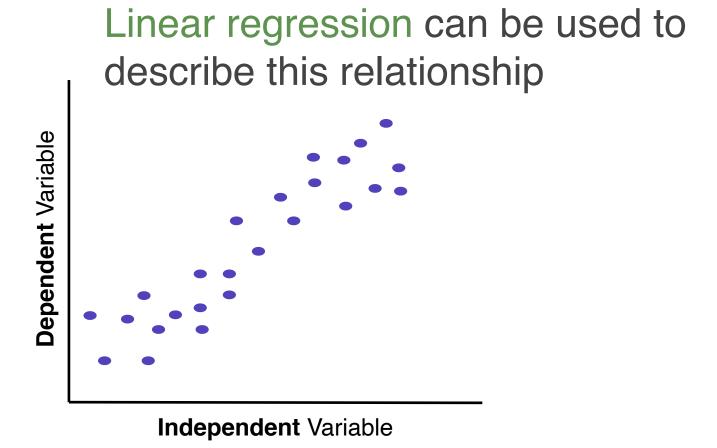
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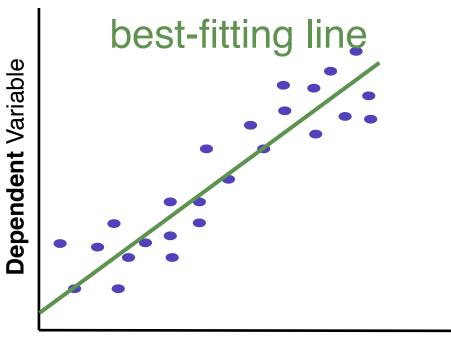
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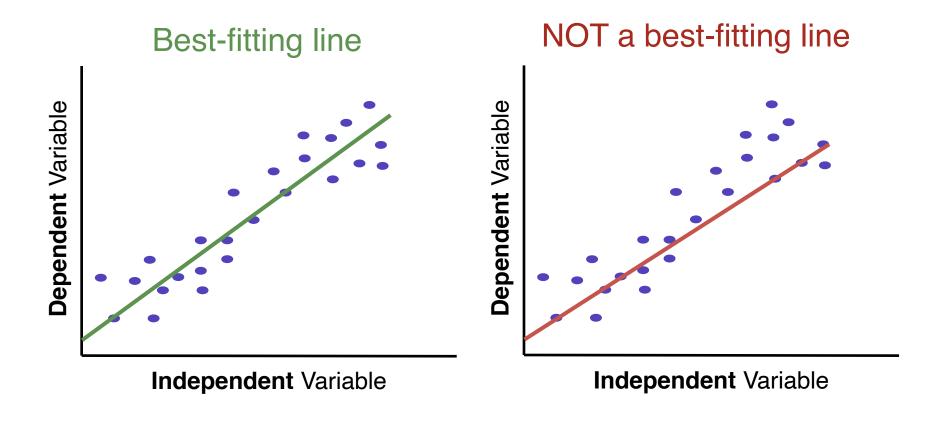
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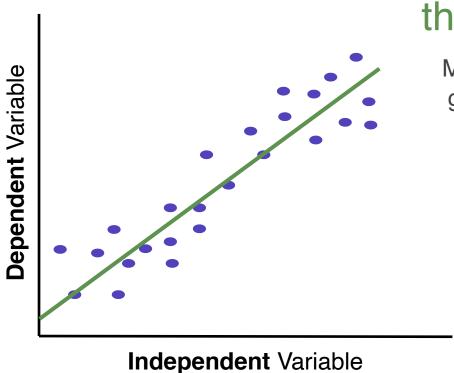
i.e. Wilcoxon ranksum test, Wilcoxon sign-rank test, sign test





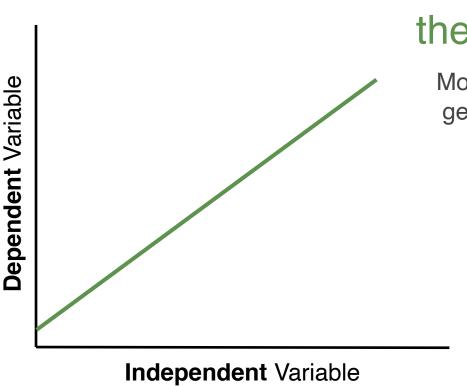
Independent Variable





This line is a model of the data

Models are mathematical equations generated to *represent* the real life situation



This line is a model of the data

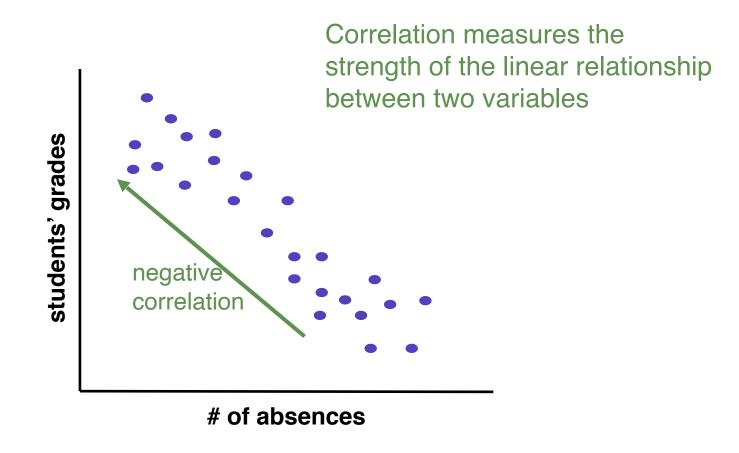
Models are mathematical equations generated to *represent* the real life situation

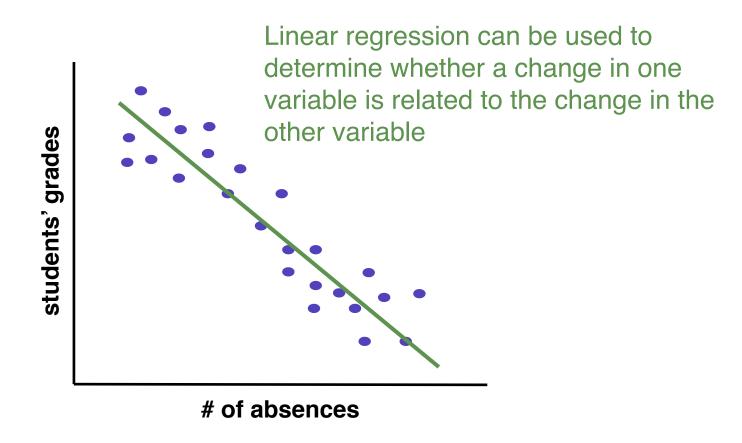
2.3 Parsimony

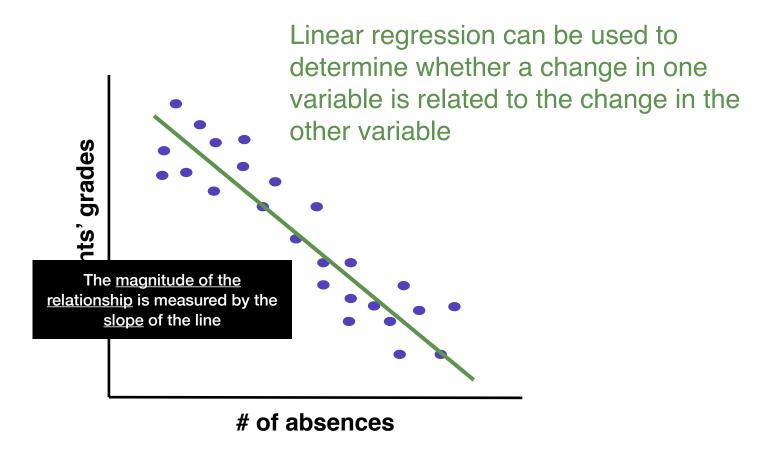
Since all models are wrong the scientist cannot obtain a "correct" one by excessive elaboration. On the contrary following William of Occam he should seek an economical description of natural phenomena. Just as the ability to devise simple but evocative models is the signature of the great scientist so overelaboration and overparameterization is often the mark of mediocrity.

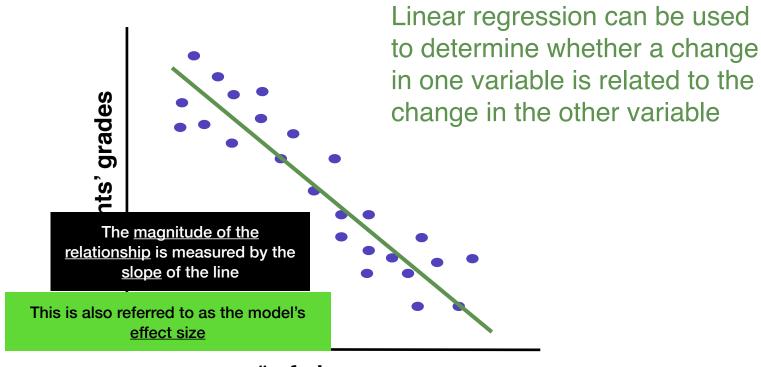
2.4 Worrying Selectively

Since all models are wrong the scientist must be alert to what is importantly wrong. It is inappropriate to be concerned about mice when there are tigers abroad.



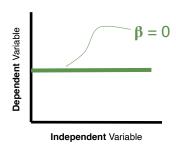




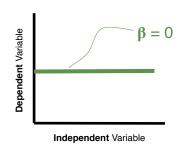


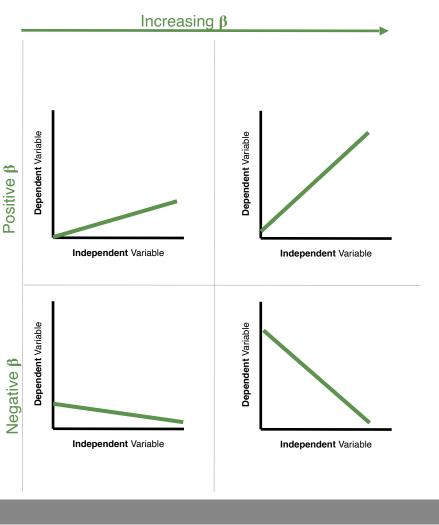
of absences

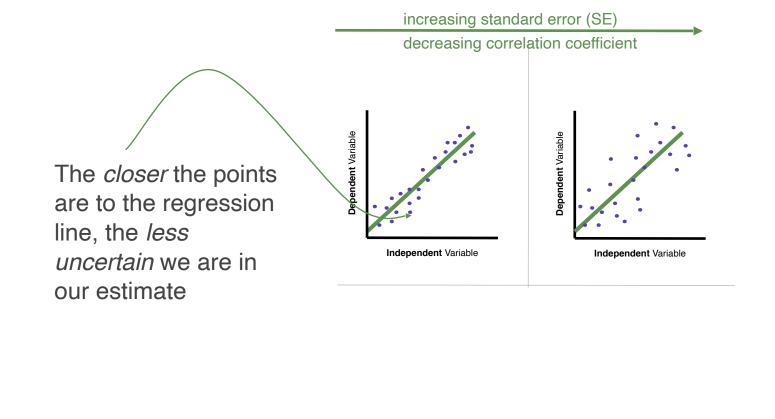
Effect size (β) can be estimated using the slope of the line



Effect size (β) can be estimated using the slope of the line

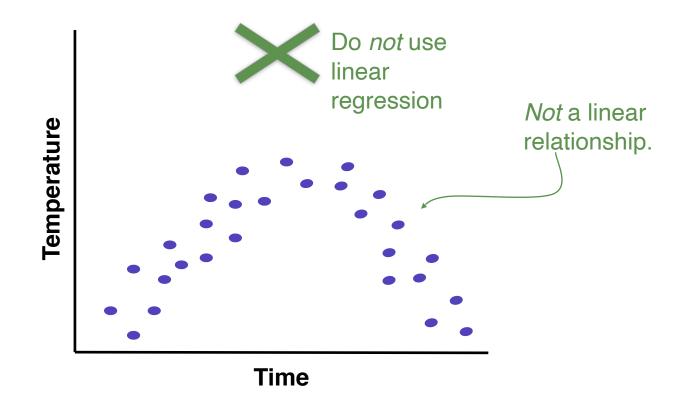


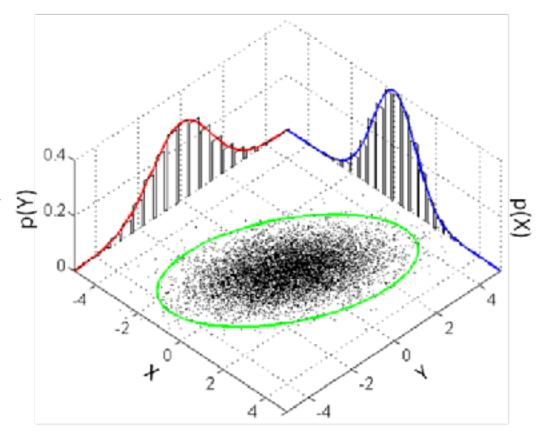




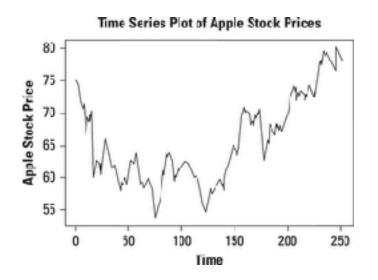
Assumptions of linear regression

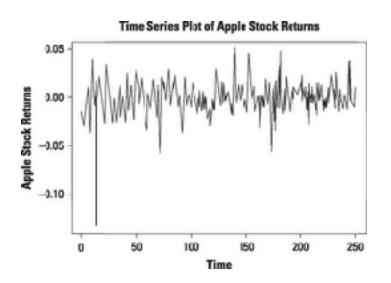
- 1. Linear relationship
- 2. Multivariate normality
- 3. No multicollinearity
- 4. No autocorrelation
- 5. Homoscedasticity



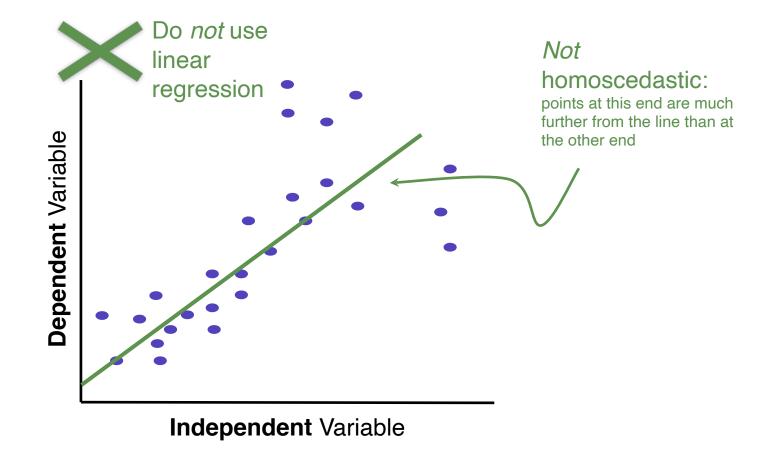


Linear regression assumes no multicollinearity. Multicollinearity occurs when the independent variables (in multiple linear regression) are too highly correlated with each other.

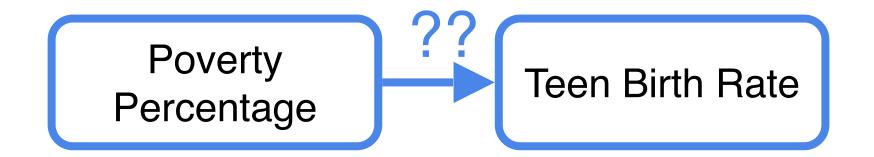




Autocorrelation occurs when the observations are *not* independent of one another (i.e. stock prices)



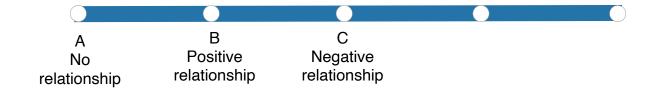
Does Poverty Percentage affect Teen Birth Rate?





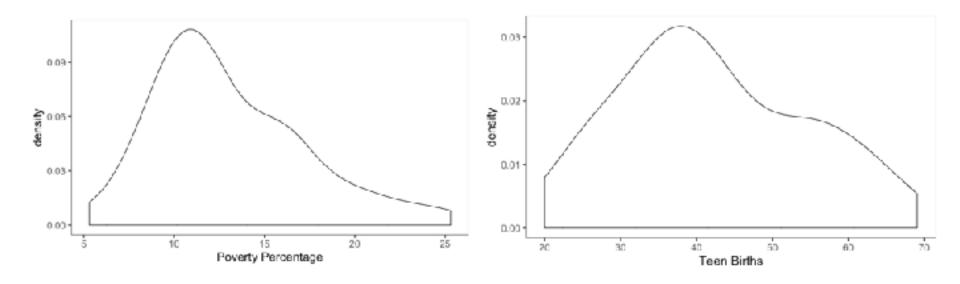
What is the relationship between Poverty Percentage & Teen Birth Rate?

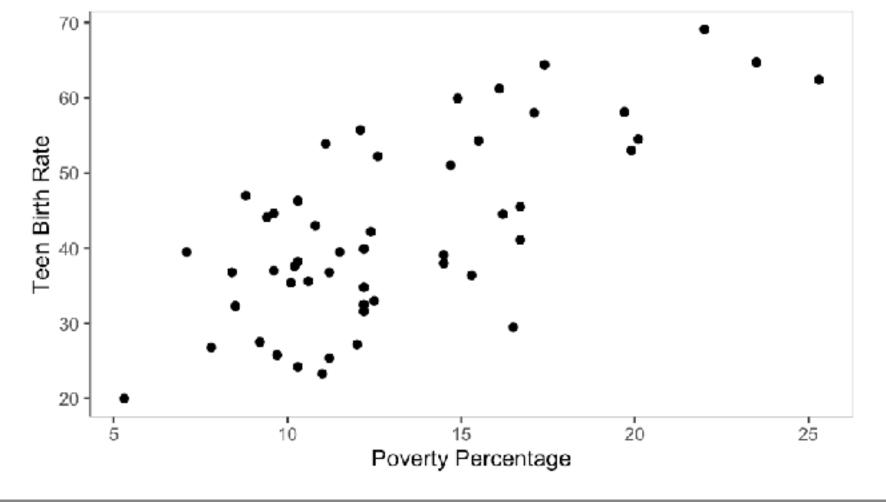
What's your hypothesis?

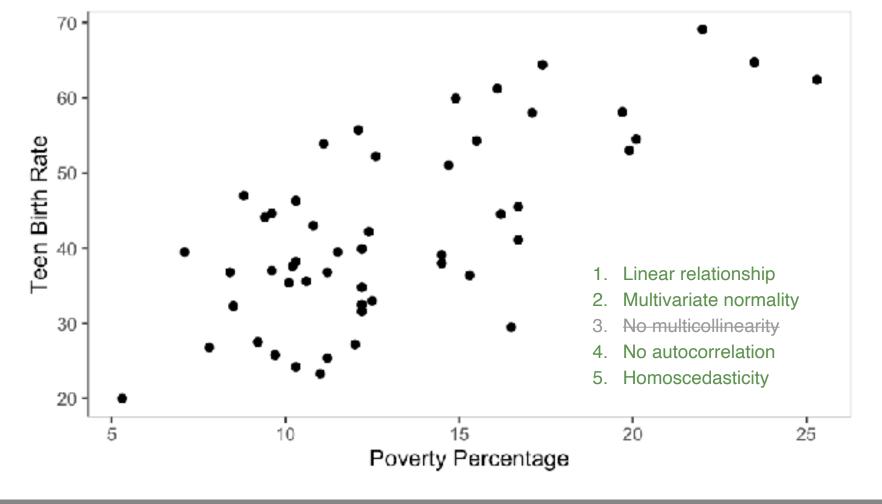


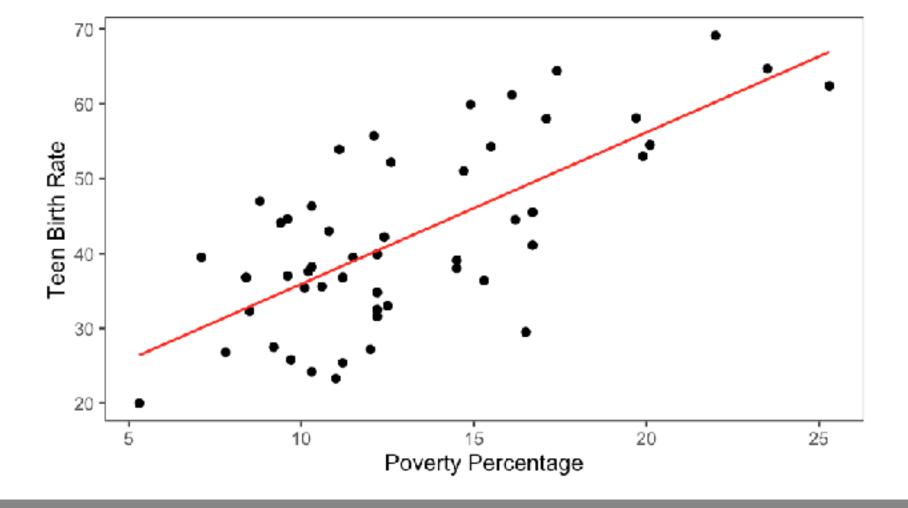
	Location [‡]	PovPct [‡]	Brth15to17	Brth18to19	ViolCrimê	TeenBrth
1	Alabama	20.1	31.5	88.7	11.2	54.5
2	Alaska	7.1	18.9	73.7	9.1	39.5
3	Arizona	16.1	35.0	102.5	10.4	61.2
4	Arkansas	14.9	31.6	101.7	10.4	59.9
5	California	16.7	22.6	69.1	11.2	41.1
6	Colorado	8.8	26.2	79.1	5.8	47.0
7	Connecticut	9.7	14.1	45.1	4.6	25.8
8	Delaware	10.3	24.7	77.8	3.5	46.3
9	District_of_Columbia	22.0	44.8	101.5	65.0	69.1
10	Florida	16.2	23.2	78.4	7.3	44.5
11	Georgia	12.1	31.4	92.8	9.5	55.7
12	Hawaii	10.3	17.7	66.4	4.7	38.2
13	Idaho	14.5	18.4	69.1	4.1	39.1
14	Illinois	12.4	23.4	70.5	10.3	42.2
15	Indiana	9.6	22.6	78.5	8.0	44.6
16	Iowa	12.2	16.4	55.4	1.8	32.5
17	Kansas	10.8	21.4	74.2	6.2	43.0

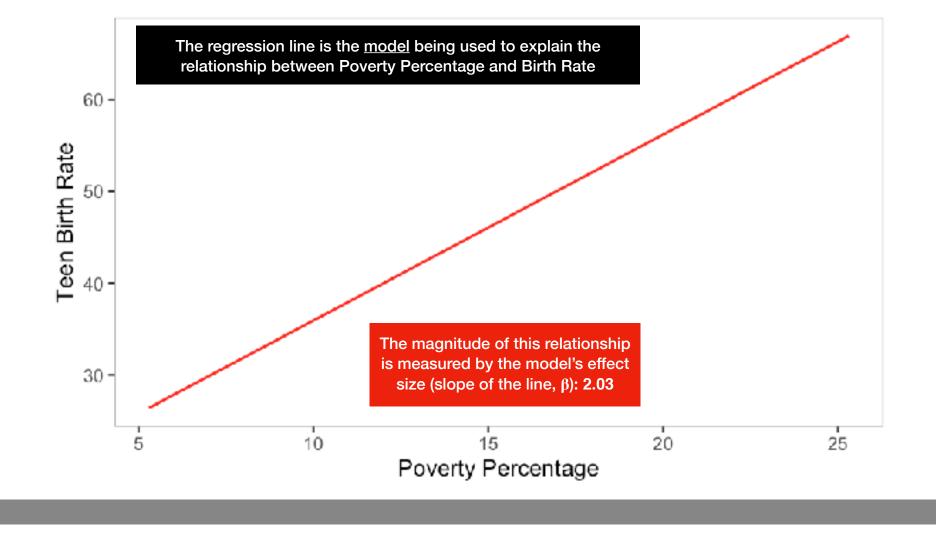
Normal(ish) distributions

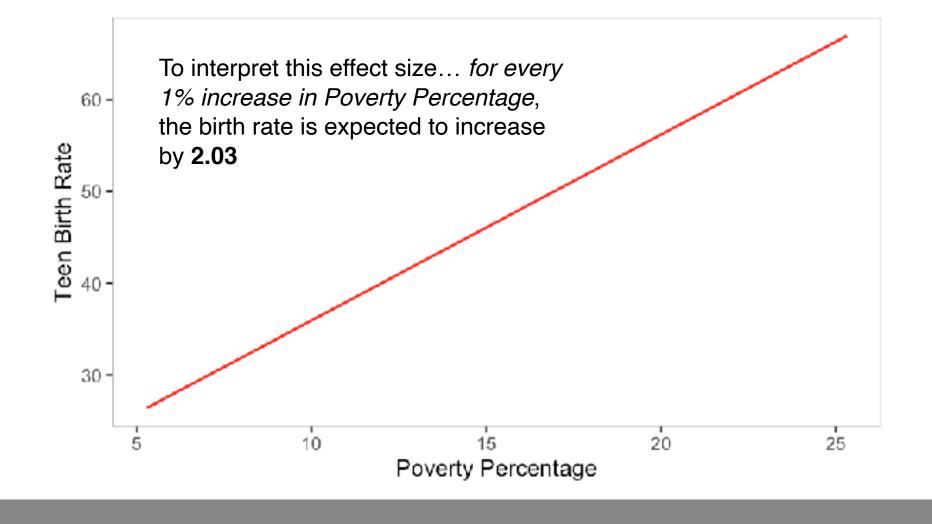


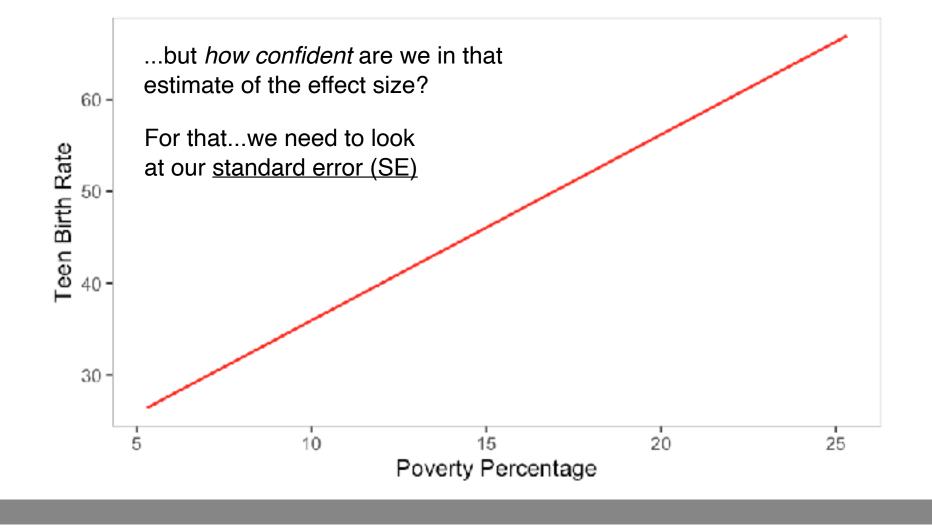


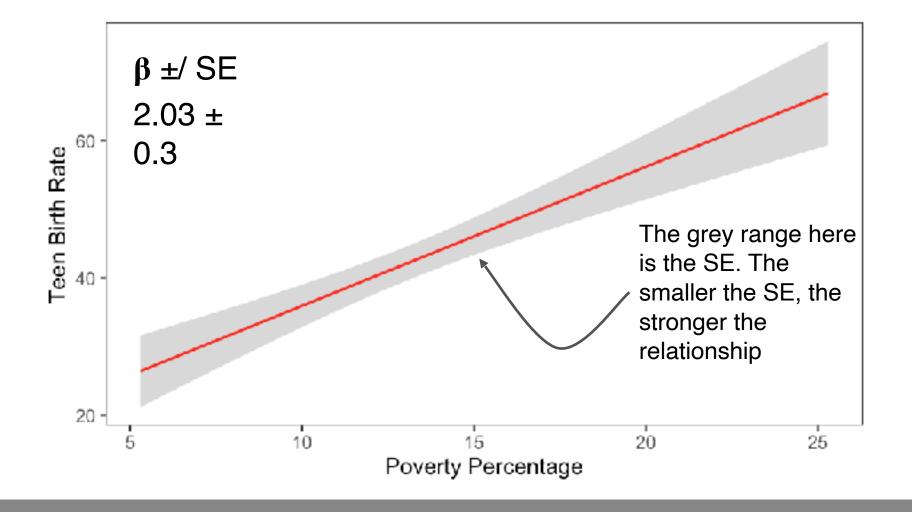


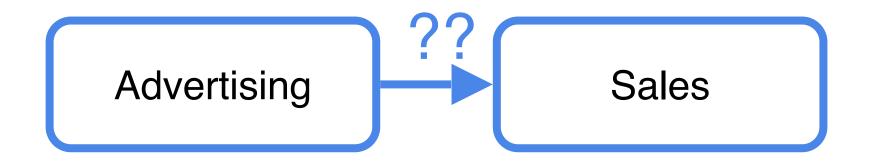






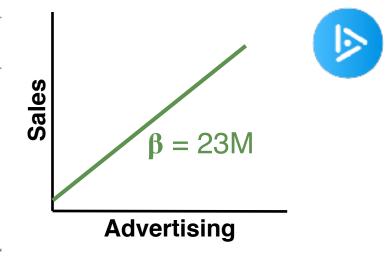




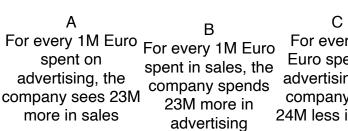


Effect size interpretation

Sales	
(Million	Advertising
Euro)	(Million Euro)
651	23
762	26
856	30
1,063	34
1,190	43
1,298	48
1,421	52
1,440	57
1,518	58

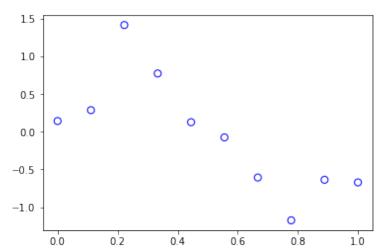


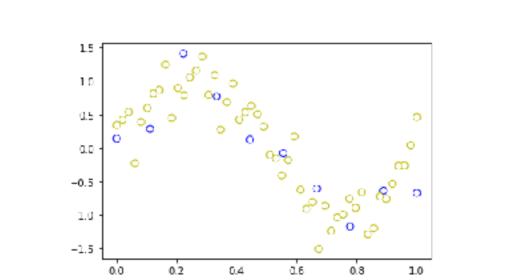
The effect size (β) between the advertising and sales is 23M. What does this mean?

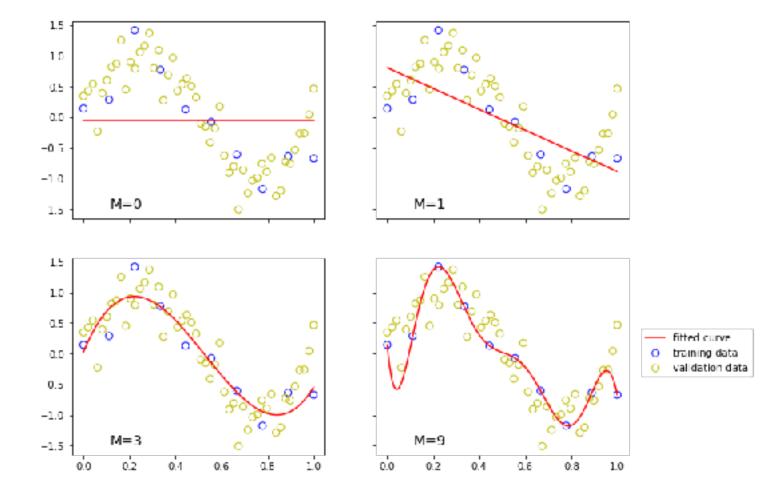




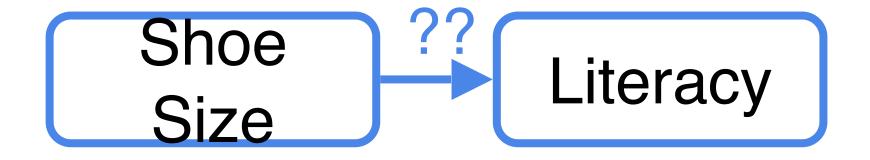






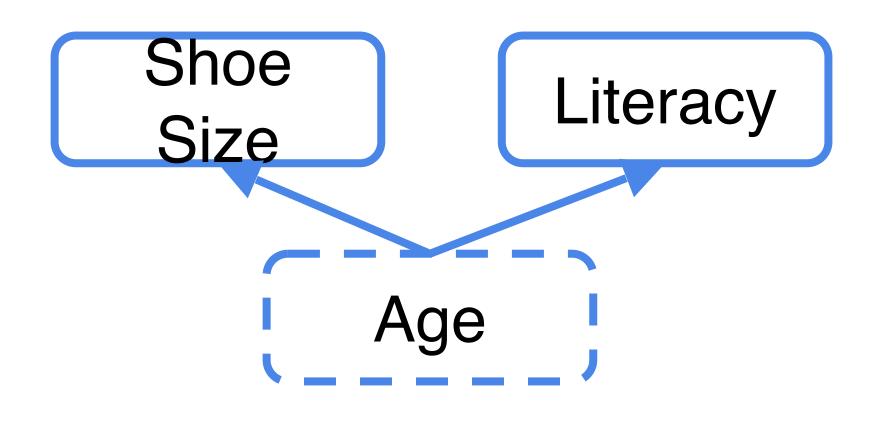


Confounding





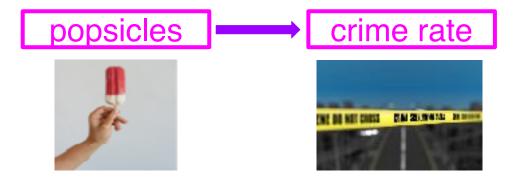
Big shoes Literate Adult



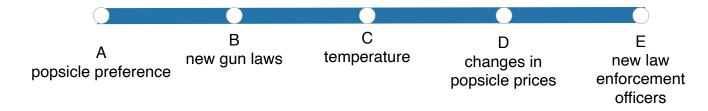
Variable2 Variable1 Confounder

Confounding





Your analysis sees an increase in crime rate whenever popsicle sales increase. What could confound this analysis?



You can plan ahead to avoid confounding and/or include confounders in your models to account for their role on the outcome variable.

Ignoring confounders will lead you to draw incorrect conclusions

Spine Surgery Results

Sample: 400 patients with index vertebral fractures

...looks like vertebroplasty was *way* worse for patients!

Vertebroplasty	Conservative care	Relative risk (95% confidence interval)
30/200 (15%)	15/200 (7.5%)	2.0 (1.1–3.6)

subsequent fractures

But wait...at time of initial fracture...

	Vertebroplasty	Conservative care
	N = 200	N = 200
Age, y, mean \pm SD	78.2 ± 4.1	79.0 ± 5.2
Weight, kg, mean \pm SD	54.4 ± 2.3	53.9 ± 2.1
Smoking status, No. (%)	110 (55)	16 (8)

Age and weight are similar between groups. **Smoking Status** differs vastly.

So...let's stratify those results real quick

Smoke		No smoke		
Vertebroplasty Conse	ervative RR (95% confidence	Vertebroplasty	Conservative	RR (95% confidence
	interval)			interval)
23/110 (21%) 3/16	(19%) (1.1 (0.4, 3.3)	7/90 (8%)	12/184(7%)	1.2 (0.5, 2.9)

Risk of re-fracture is now similar within group