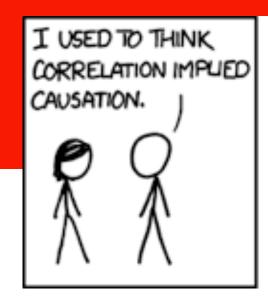
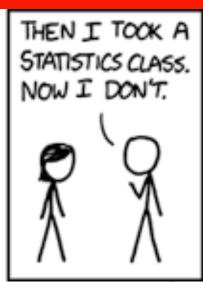
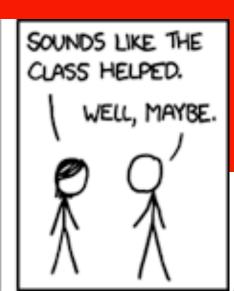
Analytical Issues With Correlated Data







Analytical Issues With Correlated Data

Outline

- 1. What is correlation?
- 2. How does it arise in our data?
- 3. Why is it important?

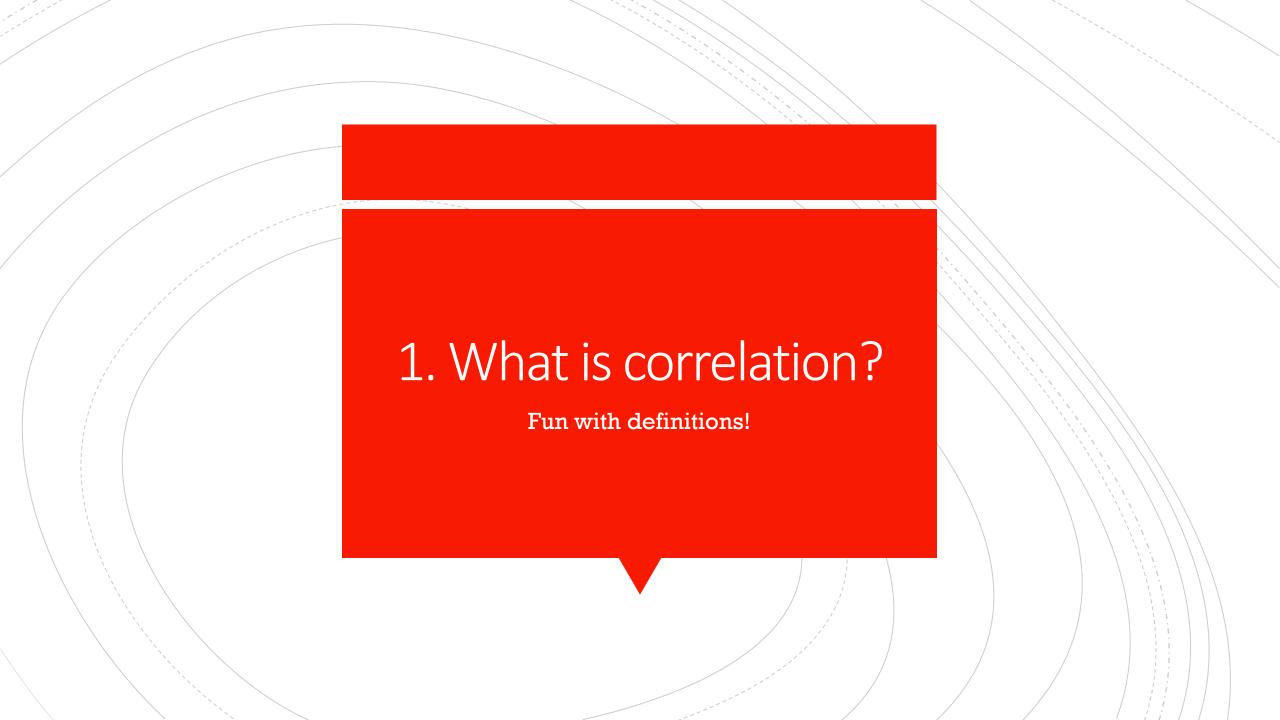
Analytical Issues With Correlated Data

Disclaimer: There will be a few formulas introduced in this lecture, but do not focus on them too much, we will be exploring them further in future lectures.

Fi BO X BI * Xi

t-statistic: β / v [var(β)]





Poll: Collecting more data, (if budget allows) is always better:

- 1) Duh (Lawful good)
- 2) It depends (True Neutral)
- 3) You may be getting more than you bargained for (Chaotic Evil)

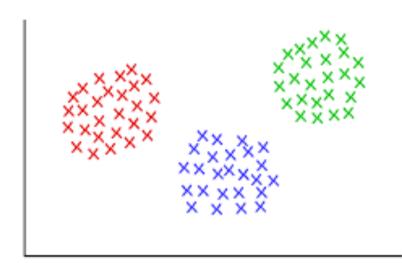
We can talk about correlation in 2 ways:

l) Relationship between two variables



We can talk about correlation in 2 ways:

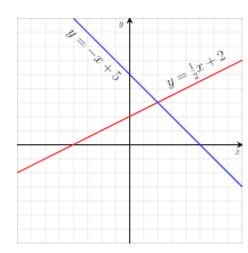
2) Correlation is the degree to which <u>outcomes</u> move together



- When we model data, we are often interested in determining the relationship between an outcome Y and one or more exposures X
- We often simplified this as the equation of a line

$$Yi = \beta 0 + \beta 1 * Xi$$

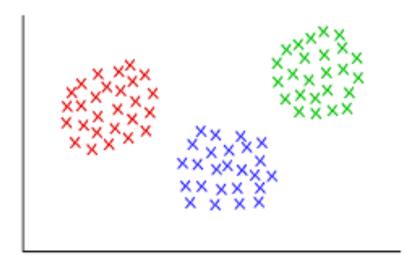
- Y is our outcome
- X is our exposure
- β0 is our intercept
- Bl is our slope





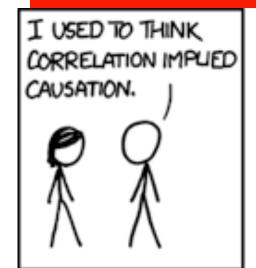
2) Correlation is the degree to which **outcomes** move together

$$Yi = \beta 0 + \beta 1 * Xi$$

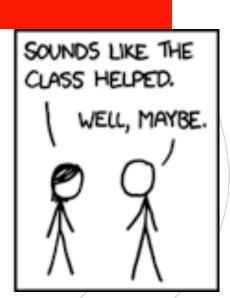


Each of these little dots is a Yi value



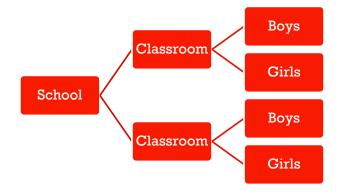




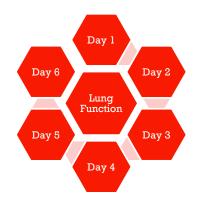


- We generally have two types of data: cross sectional and longitudinal.
 - Cross-sectional data is collected at a single timepoint. It is a "snapshot"
 - Longitudinal data is collected multiple times on a single entity over a period of time
- Example 1: If we collect the info on student IQ and GPA in a single at any one given time, it is cross-sectional
- Example 2: If we collect a student's GPA over time, it is longitudinal

HIERARCHICAL STRUCTURE



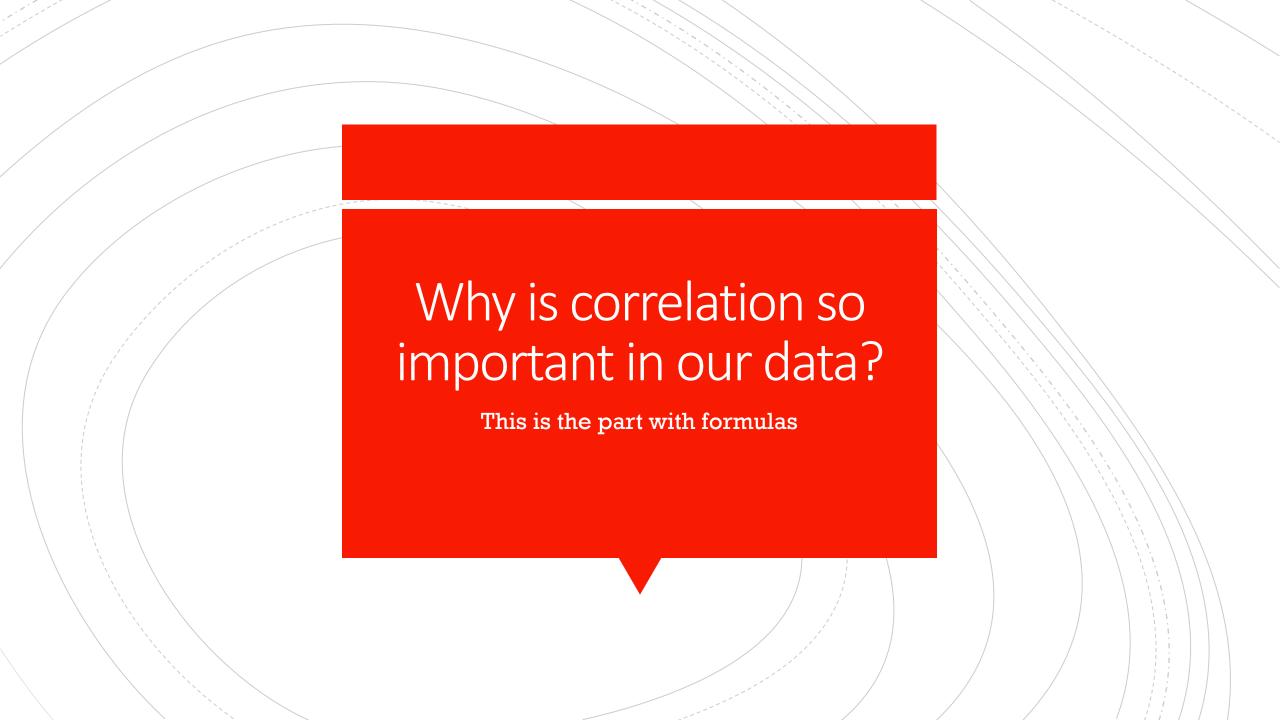
CLUSTERED STRUCTURE



- Some Examples
 - Looking at trends in grades over the semester
 - Comparing the number of goals scored by each striker within a league
 - Changes in biological samples assays between two sites

Some Examples

- Looking at trends in grades over the semester
 - Each student will have multiple grades (Yi) which would be correlated
- Comparing the number of goals scored by each striker within a league
 - Each striker could score a similar amount of goals (Yi)
 during each game
- Changes in biological samples assays between multiple sites
 - Samples from each site might be more similar to one another than compared to another site



Let's take another look at our old friend

$$Yi = \beta 0 + \beta 1 * Xi$$

- You may recall some assumptions that came with it:
 - L: The relationship between X and Y is linear
 - I: Independence of the outcome values Y (or residuals)
 - N: All variables are normally distributed
 - **E**: Equality of residuals

Let's take another look at our old friend

$$Yi = \beta 0 + \beta 1 * Xi$$

- You may recall some assumptions that came with it:
 - L: The relationship between X and Y is linear
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- If two events Y1 and Y2 are independent, then
 - Probability (Y1 happens) & Probability (Y2 happens)
 = Probability (Y1 happens) * Probability (Y2 happens)
 - Covariance (Y1,Y2) = 0
- The variance of Y1 and Y2 is
 - Var (Y1 + Y2) = Var (Y1) + Var (Y2) + 2 * Covariance (Y1, Y2)
- If Y1 and Y2 are independent
 - Var(Y1 + Y2) = Var(Y1) + Var(Y2) + 0

We would like to go from this formula

$$Yi = \beta 0 + \beta 1 * Xi$$

■ To these formula (T-statistic and 95% CI of the mean)

T-statistic:
$$\beta / \sqrt{[var(\beta)]}$$

95% CI:
$$\beta$$
 +/- 1.96* $\sqrt{[var(\beta)]}$

- β is an estimate of our effect or mean value for Y in our sample. It can also be written as ($\beta = Y2-Y1$) if we are just looking at two observations. Looking at the t-statistic can tell us:
 - Is there an association?
 - What is the magnitude of the association?

Let's focus on the variance portion:

T-statistic =
$$\beta / \sqrt{[var(\beta)]}$$

T-statistic = Y2 - Y1 / $\sqrt{[var(Y1,Y2)]}$

If we assume Y1 and Y2 are independent

T-statistic =
$$Y2 - Y1 / \sqrt{[var(Y1) + var(Y2)]}$$

However, if Y1 and Y2 are not independent

T statistic = Y2 - Y1 /
$$\sqrt{[var(Y1) + var(Y2) + 2*covariance(Y1,Y2)]}$$

• If we use the naïve version of the T-statistic:

T-statistic =
$$Y2 - Y1 / \sqrt{[var(Y1) + var(Y2)]}$$

Instead of

T-statistic =
$$Y2 - Y1 / \sqrt{[var(Y1) + var(Y2) + 2*covariance(Y1,Y2)]}$$

- Our interpretation of the results will change:
 - Is there an association? Biased, our T-statistic will be falsely increased!
 - What is the magnitude of the association? Biased, our Tstatistic will be falsely increased!

- Groups are not always constituted at random, but they can have some physical, geographic or social traits in common.
- We want to investigate correlation between 2 variables,
 and address correlation within one variable
- If we do not address correlation in our data, it can bias our analyses!

