## n-person Session 7

#### February 22, 2024

PMAP 8521: Program evaluation Andrew Young School of Policy Studies

#### **Plan for today**



#### *p*-values and confidence intervals



#### Matching and IPW



#### What exactly is a data generating process?

#### Can we make another DAG together?

The opera!

#### Randomness

#### How do we use random.org for things in R?

Are the results from p-hacking actually a threat to validity?

## Is a little exploratory p-hacking okay?

# Do people actually post their preregistrations?





#### See this and this for examples



See this

#### Do you have any tips for identifying the threats to validity in articles since they're often not super clear?

Especially things like spillovers, Hawthorne effects, and John Henry effects? Using a control group of some kind seems to be the common fix for all of these issues.

What happens if you can't do that? Is the study just a lost cause?

# *p*-values and confidence intervals

In the absence of *p*-values, I'm confused about how we report... significance?

#### Imbens and *p*-values

Nobody really cares about *p*-values

Decision makers want to know a number or a range of numbers some sort of effect and uncertainty

Nobody cares how likely a number would be in an imaginary null world!

#### Imbens's solution

**Report point estimates and some sort of range** 

"It would be preferable if reporting standards emphasized confidence intervals or standard errors, and, even better, Bayesian posterior intervals."

**Point estimate** 

The single number you calculate (mean, coefficient, etc.)



## Greek, Latin, and extra markings

Statistics: use a sample to make inferences about a population



Letters like  $\beta_1$  are the **truth** 

Letters with extra markings like  $\hat{\beta}_1$  are our **estimate** of the truth based on our sample



Letters like X are **actual data** from our sample

Letters with extra markings like  $\bar{X}$  are **calculations** from our sample

### **Estimating truth**

#### $\textbf{Data} \rightarrow \textbf{Calculation} \rightarrow \textbf{Estimate} \rightarrow \textbf{Truth}$

X

Data	X	
Calculation	$ar{X} = rac{\sum X}{N}$	
Estimate	$\hat{oldsymbol{\mu}}$	
Truth	$\mu$	

$$ar{X} = \hat{\mu}$$
  
 $ightarrow ar{X} 
ightarrow \hat{\mu} \xrightarrow{\hspace{0.1cm} ext{ bopefully } e$ 

### **Population parameter**

### Truth = Greek letter

#### An single unknown number that is true for the entire population

**Proportion of left-handed students at GSU** 

Median rent of apartments in Atlanta

**Proportion of red M&Ms produced in a factory** 

**Treatment effect of your program** 

#### Samples and estimates

#### We take a sample and make a guess

This single value is a *point estimate* 

(This is the Greek letter with a hat)

### Variability

You have an estimate, but how different might that estimate be if you take another sample?

#### Left-handedness

You take a random sample of 50 GSU students and 5 are left-handed.

If you take a different random sample of 50 GSU students, how many would you expect to be left-handed?

3 are left-handed. Is that surprising?

40 are left-handed. Is that surprising?

#### Nets and confidence intervals

How confident are we that the sample picked up the population parameter?

#### **Confidence interval is a net**

We can be X% confident that our net is picking up that population parameter

If we took 100 samples, at least 95 of them would have the true population parameter in their 95% confidence intervals

A city manager wants to know the true average property value of single-owner homes in her city. She takes a random sample of 200 houses and builds a 95% confidence interval. The interval is (\$180,000, \$300,000).

> We're 95% confident that the interval (\$180,000, \$300,000) captured the true mean value

#### WARNING

It is way too tempting to say "We're 95% sure that the population parameter is X"

People do this all the time! People with PhDs!

YOU will do this too



If you took lots of samples, 95% of their confidence intervals would have the single true value in them



#### Frequentism

#### This kind of statistics is called "frequentism"

The population parameter  $\theta$  is fixed and singular while the data can vary

#### $P(\text{Data} \mid \theta)$

You can do an experiment over and over again; take more and more samples and polls

#### Frequentist confidence intervals

"We are 95% confident that this net captures the true population parameter"

> "There's a 95% chance that the true value falls in this range"

#### **Bayesian statistics**



 $P(\theta \mid \text{Data})$ 

## $P(\mathbf{H} \mid \mathbf{E}) = rac{P(\mathbf{H}) imes P(\mathbf{E} \mid \mathbf{H})}{P(\mathbf{E})}$

**Rev. Thomas Bayes** 



#### P(Hypothesis | Evidence) =

 $rac{P( ext{Hypothesis}) imes P( ext{Evidence} \mid ext{Hypothesis})}{P( ext{Evidence})}$ 

## But the math is too hard!

So we simulate!

(Monte Carlo Markov Chains, or MCMC)

### **Bayesianism and parameters**

In the world of frequentism, there's a fixed population parameter and the data can hypothetically vary

In the world of Bayesianism, the data is fixed (you collected it just once!) and the population parameter can vary  $P(\text{Data} \mid \theta)$ 

 $P(\theta \mid \text{Data})$ 

### **Bayesian credible intervals**

(AKA posterior intervals)

"Given the data, there is a 95% probability that the true population parameter falls in the credible interval"

#### Intervals

#### Frequentism

#### Bayesianism

There's a 95% probability that the range contains the true value There's a 95% probability that the true value falls in this range

#### Probability of the range

Few people naturally think like this

#### Probability of the actual value

People *do* naturally think like this!

## **Thinking Bayesianly**

We all think Bayesianly, even if you've never heard of Bayesian stats

Every time you look at a confidence interval, you inherently think that the parameter is around that value, but that's wrong!

> BUT Imbens cites research that that's actually generally okay

Often credible intervals are super similar to confidence intervals

#### **Bayesian inference**

#### **Inference without** *p***-values!**

#### Probability of direction



#### **Region of practical** equivalence (ROPE)



Point shows median value; thick black bar shows 66% credible interval thin black bar shows 95% credible interval 37 / 53 RCTS

# Do we really not control for things in an RCT?

### Randomness and arrow deletion



## **Balance tests**



Chelsea Parlett-Pelleriti @ChelseaParlett

Trying to convince someone NOT to do t-tests to compare randomly assigned groups at baseline

no context the good place @nocontexttgp · Mar 10



1:04 PM · Mar 13, 2021 · Twitter for iPhone



Chelsea Parlett-Pelleriti @ChelseaParlett · Mar 13 THE RANDOMIZATION WORKED. RANDOMIZATION DOESN'T MEAN GROUPS WILL ALWAYS BE EQUAL

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Chelsea Parlett-Pelleriti @ChelseaParlett

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#### YOU DONT NEED A HYPOTHESIS TEST IF YOU KNOW THE DATA GENERATING PROCESS

1:18 PM · Mar 13, 2021 · Twitter for iPhone

# Can you walk through an example of RCTs in class?

## Matching and IPW

Can you talk more about propensity scores and "weirdness" weights?

Lecture slide





## Why not just control for confounders instead of doing the whole matching/IPW dance?

# Do you have to use logistic regression + OLS for IPW?



## Which should we use? Matching or IPW?

# Can you walk through an example of IPW and matching in class?