

# In-person session 6

**February 16, 2023**

PMAP 8521: Program evaluation  
Andrew Young School of Policy Studies

# Plan for today

**Exam 1**

**FAQs**

**Confidence intervals, credible intervals,  
and a crash course on Bayesian statistics**

# Exam 1

**Tell us about Exam 1!**

# FAQs

**What's the difference  
between + and %>%?**

**Are *p*-values really misinterpreted  
in published research?**

# Power calculations and sample size

Won't we always be able to find a significant effect if the sample size is big enough?

Yes!



# Math with computers

[andhs.co/live](https://andhs.co/live)

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

**Do people actually post  
their preregistrations?**

**Yes!**

**OSF**

See **this** and **this** for examples

**As Predicted**

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?**

**Confidence intervals,  
credible intervals,  
and a crash course on  
Bayesian statistics**

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



# Math with computers

[andhs.co/live](https://andhs.co/live)

# 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.)**

**Uncertainty**

**A range of possible values**

# Greek, Latin, and extra markings

Statistics: use a sample to make inferences about a population

## Greek

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

## Latin

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

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

# Estimating truth

Data → Calculation → Estimate → Truth

Data	$X$
Calculation	$\bar{X} = \frac{\sum X}{N}$
Estimate	$\hat{\mu}$
Truth	$\mu$

$$\bar{X} = \hat{\mu}$$

$$X \rightarrow \bar{X} \rightarrow \hat{\mu} \xrightarrow{\text{👉 hopefully 👉}} \mu$$

# 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 NYC

Proportion of red M&Ms produced in a factory

ATE 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-value 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

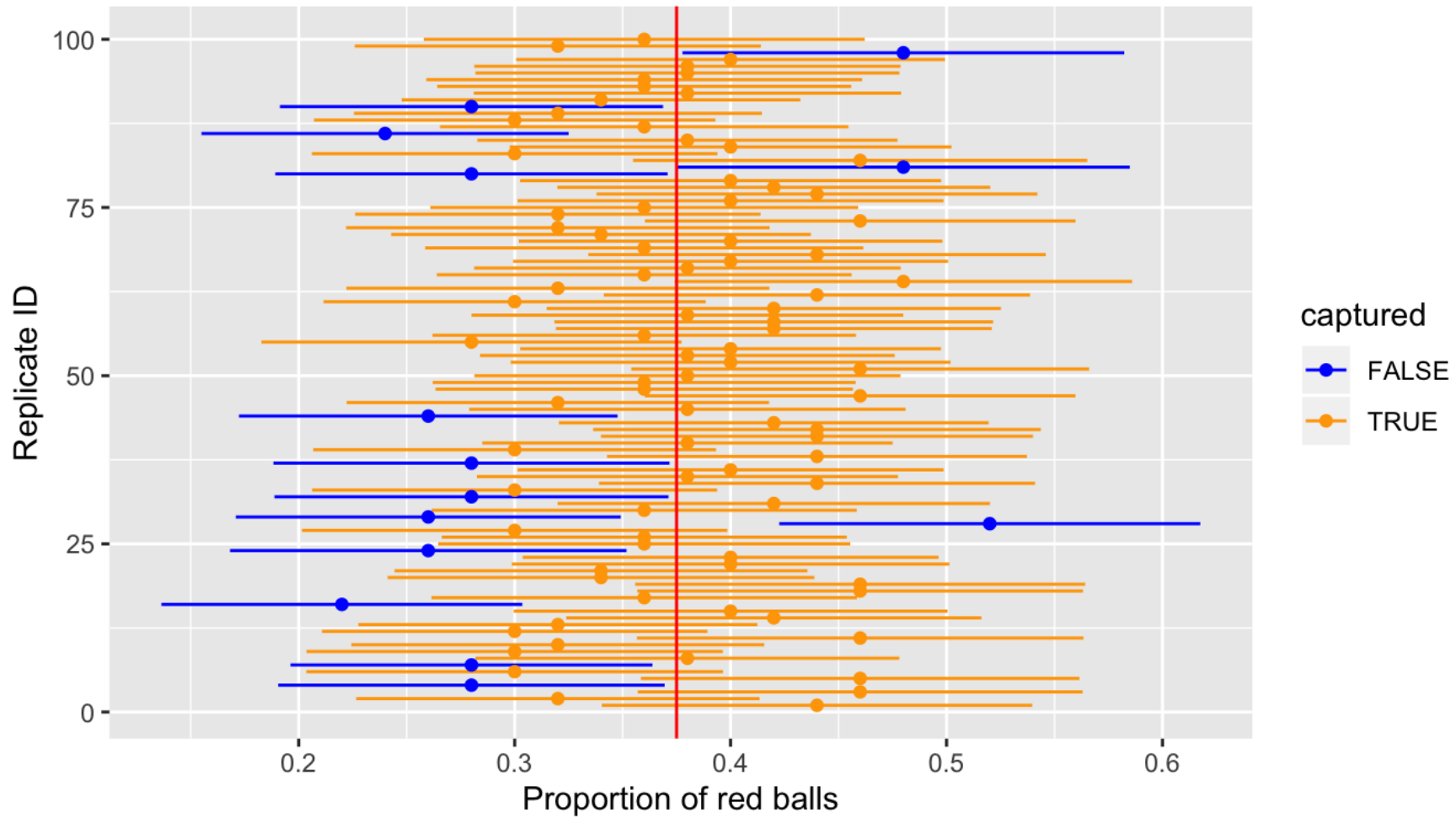
It is way too tempting to say  
“We’re 95% sure that the  
population parameter is  $X$ ”

People do this all the time! People with PhDs!

YOU will try to do this too

# Nets

**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"~~



# **Weekends and restaurant scores**

# Bayesian statistics



Rev. Thomas Bayes

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

$$P(H \mid E) = \frac{P(H) \times P(E \mid H)}{P(E)}$$

# Bayesianism in WWII



Alan Turing



An enigma machine

$$P(\mathbf{H} \mid \mathbf{E}) = \frac{P(\mathbf{H}) \times P(\mathbf{E} \mid \mathbf{H})}{P(\mathbf{E})}$$

$$P(\text{Hypothesis} \mid \text{Evidence}) = \frac{P(\text{Hypothesis}) \times P(\text{Evidence} \mid \text{Hypothesis})}{P(\text{Evidence})}$$

$$P(H | E) = \frac{P(H) \times P(E | H)}{P(E)}$$



$$\overbrace{P(\text{Unknown} | \text{Data})}^{\text{Posterior}} = \frac{\overbrace{P(\text{Unknown})}^{\text{Prior}} \times \overbrace{P(\text{Data} | \text{Unknown})}^{\text{Likelihood}}}{\underbrace{P(\text{Data})}_{\text{Average likelihood}}}$$

$$\overbrace{P(\text{Unknown} \mid \text{Data})}^{\text{Posterior}} = \frac{\overbrace{P(\text{Unknown})}^{\text{Prior}} \times \overbrace{P(\text{Data} \mid \text{Unknown})}^{\text{Likelihood}}}{\underbrace{P(\text{Data})}_{\text{Average likelihood}}}$$



$$\overbrace{P(\text{Unknown})}^{\text{Prior}} \times \overbrace{P(\text{Data} \mid \text{Unknown})}^{\text{Likelihood}} \propto \overbrace{P(\text{Unknown} \mid \text{Data})}^{\text{Posterior}}$$

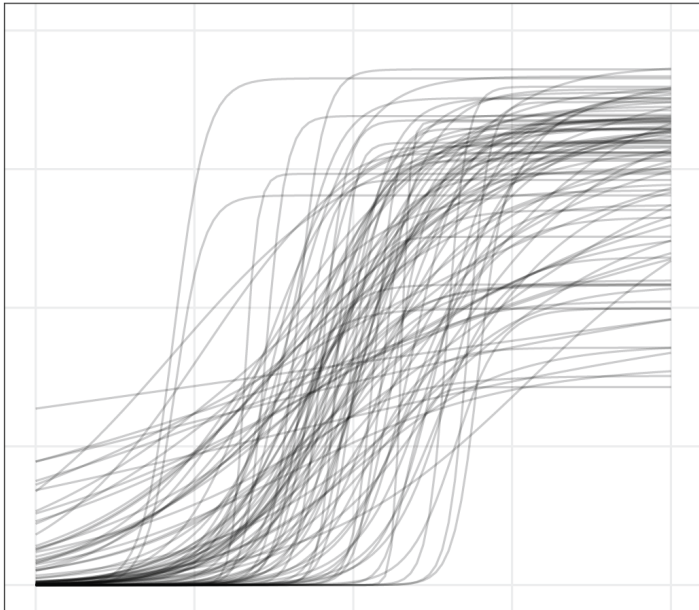
# Bayesian statistics and more complex questions

$$\overbrace{P(\text{Unknown})}^{\text{Prior}} \times \overbrace{P(\text{Data} \mid \text{Unknown})}^{\text{Likelihood}} \propto \overbrace{P(\text{Unknown} \mid \text{Data})}^{\text{Posterior}}$$



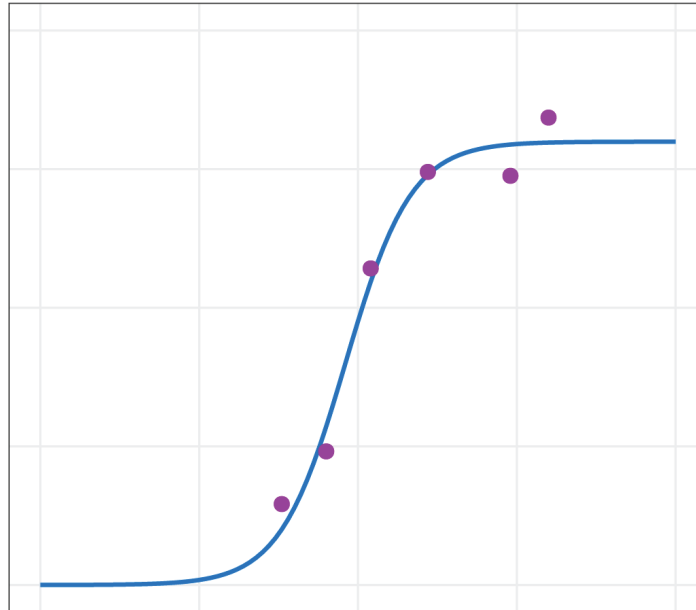
### Plausible curves before seeing the data

The prior:  $P(\text{Unknown})$



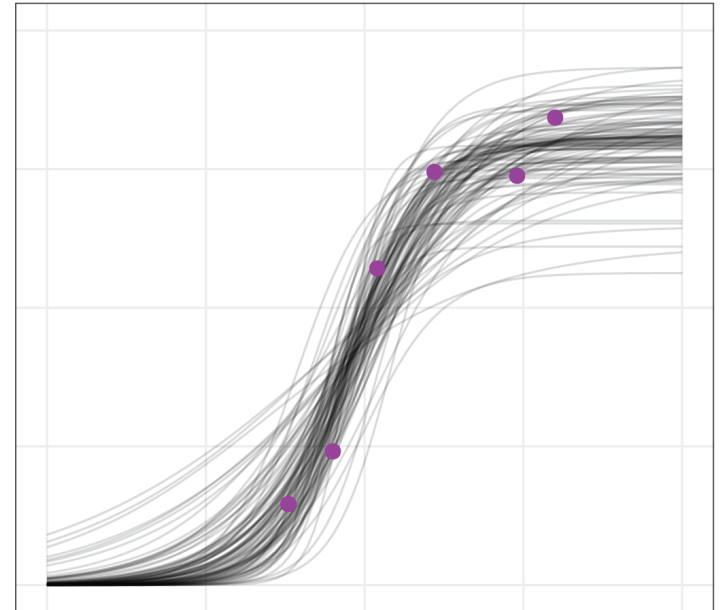
### How well the curves fit the data

The likelihood:  $P(\text{Data} \mid \text{Unknown})$



### Plausible curves after seeing the data

The posterior:  $P(\text{Unknown} \mid \text{Data})$





**But the math is too hard!**

**So we simulate!**

**(Monte Carlo Markov Chains, or MCMC)**

# **Weekends and restaurant scores again**

# Bayesianism and parameters

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

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

In the world of Bayesianism,  
the data is fixed (you collected it just once!)  
and the population parameter can vary

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

# 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

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

Probability of the range

Few people naturally  
think like this

## Bayesianism

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

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!

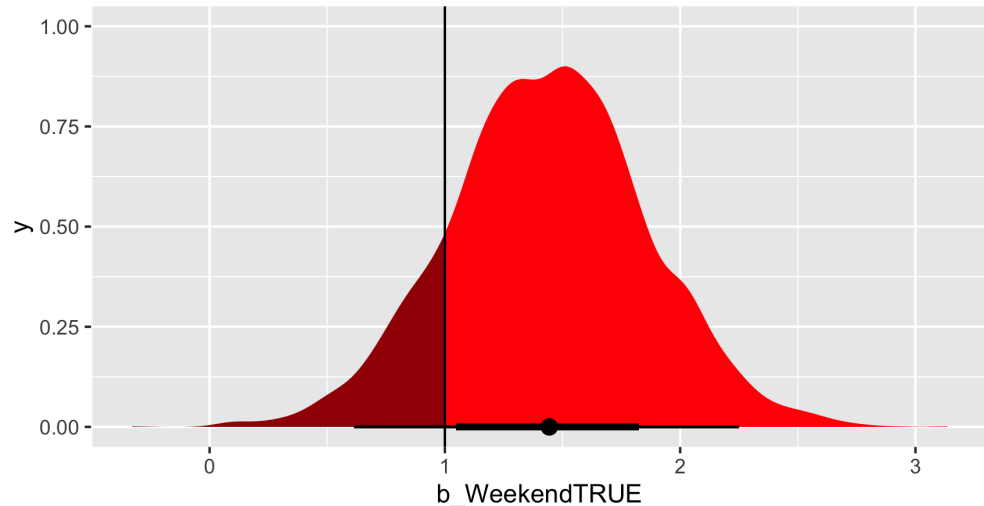
**BUT** Imbens cites research that  
that's actually generally okay

Often credible intervals are super similar to confidence intervals

# Bayesian inference

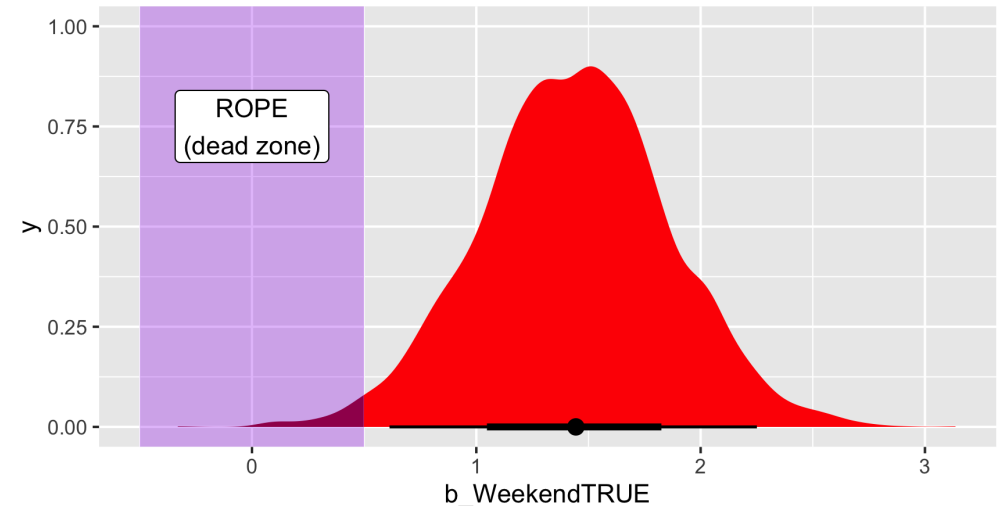
What do you do without p-values then?

Probability of direction



Point shows median value;  
thick black bar shows 66% credible interval;  
thin black bar shows 95% credible interval

Region of practical equivalence (ROPE)



Point shows median value;  
thick black bar shows 66% credible interval;  
thin black bar shows 95% credible interval

**Weekends and  
restaurant scores  
once more**