

# STA 235 - Causal Inference: Natural Experiments

Spring 2021

McCombs School of Business, UT Austin

# Reminders

**Homework 2 is due next Monday (noon)**

- Remember to ask your questions **before the weekend**
- I will make **additional OH** available if they get full.
- Remember to **contact me** if you need to meet another time.

# Reminders (cont.)

**Class participation increases learning**



**Jamin Speer**  
@JaminSpeer



I think students sometimes think that asking question will annoy their professor. No! I love it when you ask questions because it means you are actually paying attention and thinking in class!

12:50 PM · Feb 25, 2021 · Twitter Web App

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# Last class

- **Selection on observables**
  - Assumption of non-random selection, but selection on observables.
- **Matching and Weighting:**
  - Use of other adjustment methods beyond regression.
  - Advantages and disadvantages of matching.
  - Use of weighting for approximating different populations.



# Today



- **Natural Experiments:**
  - Identifying random assignment\* in observational studies
  - Use exogenous variation to identify causal effects.
- **Differences in Differences:**
  - Using two dimensions for identification.
  - Assumptions and shortcomings.

Is there randomness out there?

# Finding "RCTs" in the wild

- Given that we can't run RCTs for everything, the next best thing is finding a source of random variation that, for all practical purposes, **would work as an RCT**

## Natural Experiments

You, as a researcher, did not assign units to treatment levels

1. **Random**: Assignment to an intervention is random (e.g. lottery).
2. **As if random**: Assignment to an intervention is not random, but it's not correlated with potential confounders.

**Context matters!**

# Potential outcomes in Observational Studies

- The same **potential outcomes framework** that we reviewed for RCTs also work with observational studies.

## Steps to identify a Natural Experiment:

1) Identify treatment groups: What is the control status?

2) Identify your estimand of interest: Write it down in terms of PO!

3) Identify potential threats to causality: Is this as good as random?



# Let's talk about the JITT example

- A retailer provides a 15% discount to first 1,000 customers, 10% to customers 1,001-2,000 (and no discount after).

**Is this a natural experiment?**

Mixed answers

**Let's think about this more carefully**

# An example: Timely discounts

- **Two treatments:** 10% discount (1) and 15% discount (2) (control is no discount).
- **Outcome:** Total sales (\$)
- **Estimand:** Average Treatment Effects,

$$ATE_1 = E[Y(1) - Y(0)]$$

and

$$ATE_2 = E[Y(2) - Y(0)]$$

How are people assigned to treatment?

# An example: Timely discounts

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**Could there be confounding? Why?**

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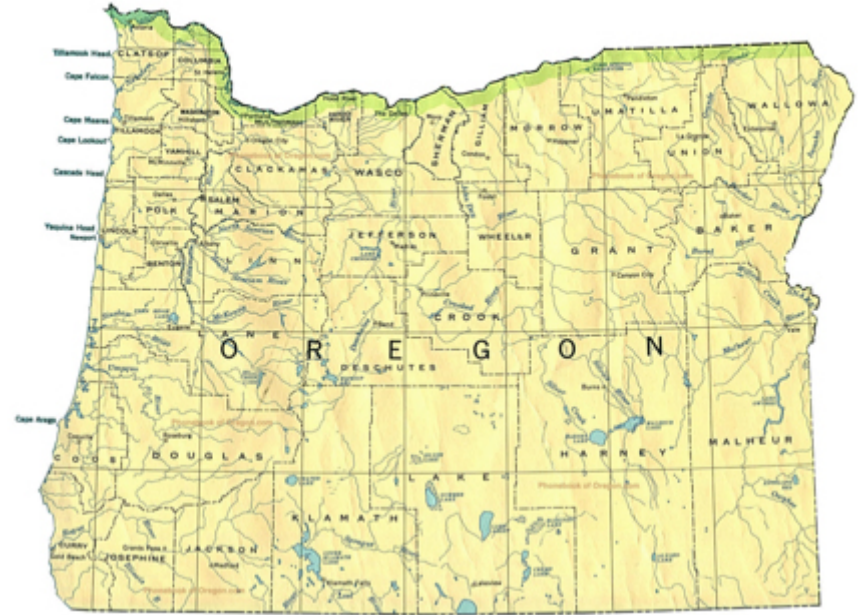
and

$$ATE_2 = E[Y(2) - Y(0)]$$

What if customers didn't know about the discount until they get there?

# A true natural experiment: The Oregon Health Plan

- In 2008, Oregon implemented a **limited expansion of Medicaid**.
- **Target population**: Low-income adults.
- People selected through a **lottery**.



# The Oregon Health Plan

What is the treatment in this case? What is *randomized*\*?

What is our estimand of interest?

What about external validity? For what population is this effect generalizable?

What is the first thing you would do with the data?

**Let's go to R**

# Other natural experiments

- Natural experiments can arise from **anything**!
  - E.g. glitches in systems, allocation based on "random" variables.

**Treatment assignment has to be [conditionally] independent of potential outcomes!**



# Additional Examples: Fish and Phones

- **Context:** Kerala, 1997.
- **Problem:** Fishermen out at sea cannot observe prices in the different markets, and due to time constraints can only visit one.
- **Identification Strategy:** Natural source of variation when cellphones were introduced and rolled out geographically.

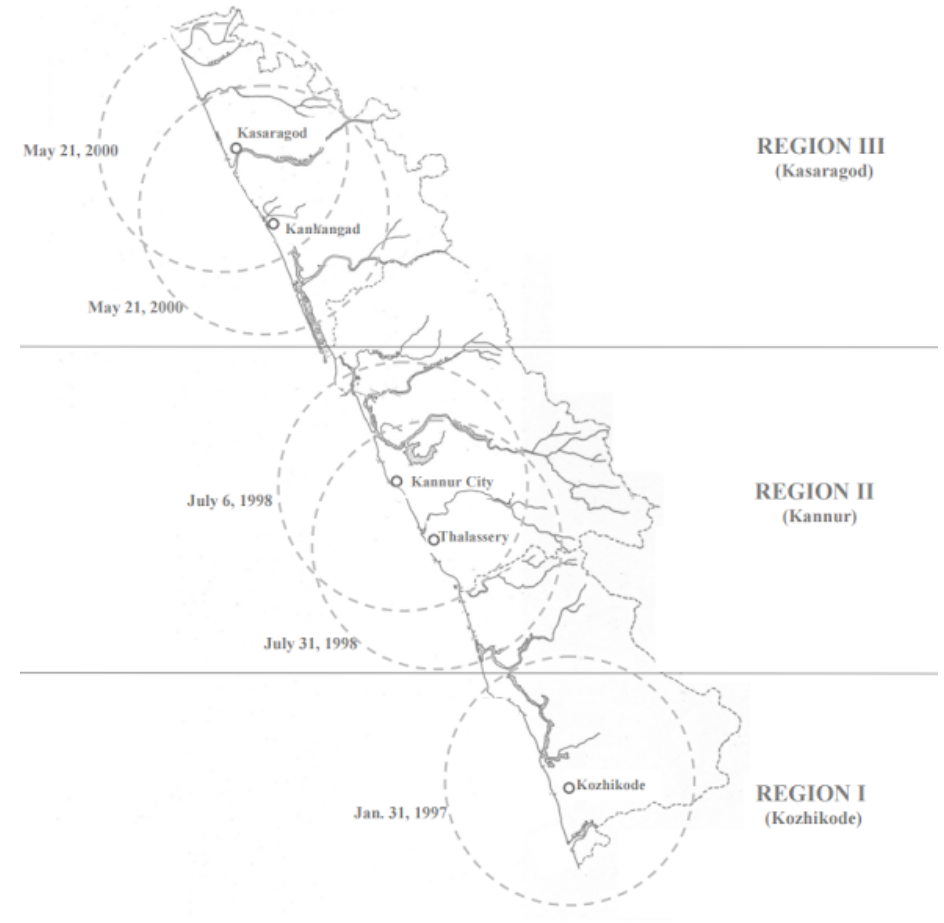


FIGURE II  
SPREAD OF MOBILE PHONE COVERAGE IN KASARAGOD, KANNUR AND KOZHIKODE DISTRICTS

# Additional Examples: Fish and Phones

- **Study:** Jensen (2007) uses the variation to study prices in time depending on location.
- **Results:** Significant decrease in price variation!

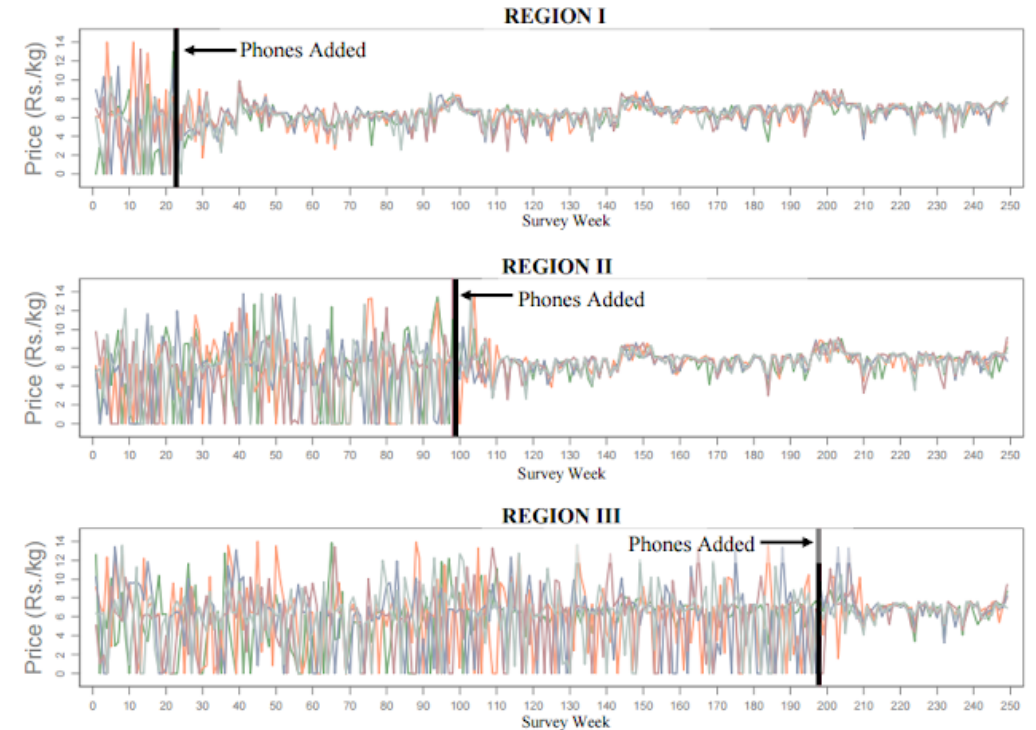


FIGURE IV  
PRICES AND MOBILE PHONE SERVICE IN KERALA

# Additional Examples: Natural disasters

- **Natural disasters** such as hurricanes or earthquakes usually make for great natural experiments
  - As long as you can **credibly** convince people they are not correlated with potential outcomes!
- Imagine that you have two cities, A and B, which both have equal probability of being affected by a hurricane:

City A: Large city

City B: Smaller town

- You want to study the effect of this natural disaster on unemployment. **Could you use this setting?**

# Takeaway points



- We don't always need to **"randomize"**.
- We can exploit **natural variation**.
  - In **natural experiments** we are approximating an ideal RCT very straightforwardly.

# References

- Angrist, J. and S. Pischke. (2015). "Mastering Metrics". *Chapter 2*.
- NBER (2021). "Oregon Health Insurance Experiment Background". *Summary of the policy*.
- Jensen, R. (2007). "The Digital Divide: Information (Technology), Market Performance, and Welfare in the South Indian Fisheries Sector". *Quarterly Journal of Economics*.