# Lab 5: Simulation Challenges

2024 - 05 - 01

## Split into 3 groups

Each group will work together to simulate one of 3 scenarios over 30 minutes, and then 1 member from each group will informally present their simulation for about 5 minutes.

#### Group 1: Difference in Differences and Parallel Trends

The **parallel trends** assumption in the difference-in-differences model says that absent treatment, the treatment group *would have* followed the same trend that the control group did.

Consider the following scenario: three states (A, B, and C) are considering implenting near identical investment policies. It is passed in states A and B, not in C. However, in state B, the day before the policy was to be implemented, it was halted by a Federal Judge. Additionally, state C is hit with a devastating natural disaster, plumeting them into a recession.

Suppose GDP (Y) in each state i at time t is determined by the following data generating process:

$$Y_{i,t} = \alpha + \gamma_i + \beta \cdot t + \delta \cdot \mathbb{I}(policy_{i,t}) + \eta \cdot t \cdot \mathbb{I}(i = C, t \ge 0) + \varepsilon_{i,t}$$

Where:

- $\alpha = 10$
- Fixed effects:  $\gamma_A = 2, \gamma_B = -1, \gamma_C = 0.5$
- Time trend:  $\beta = 0.5$
- $\delta = 1$ , and  $\mathbb{I}(policy_{i,t}) = 1$  if the policy is implemented, 0 otherwise
- $\eta = -0.7$
- $\varepsilon_{i,t} \sim N(0, 0.3)$

You, as an economist interested in the causal effect of the policy's implementation.

**1.1 Create data** Write a function that generates data for states A, B, and C over periods -10 to 10, where the policy is enacted at time t = 0.

Graph an example dataset, showing each state's output (Y) overtime on the same graph, depicting when the policy was implemented.

- 1.2 Simulate with C as Control Run 1000 iterations of a simulation where you run a difference in differences regression with state A as the treatment group and C as the control. Then graph the estimates of  $\delta$  in a density plot.
- 1.3 Simulate with B as Control Repeat 1.2, this time with B as the control.
- **1.4 BONUS!** Run 1.2/1.3 again, this time using both B and C as control groups.

### Group 2: Instrumental Variable

Consider an agriculture market where equilibrium is determined by the two following equations for Supply and Demand:

Supply: 
$$q_t = \gamma \cdot p_t + \eta \cdot w_t + \nu_t$$
  
Demand:  $q_t = \delta \cdot p_t + \varepsilon_t$ 

Where  $\gamma, \eta > 0$ , and  $\delta < 0$ .  $p_t$  and  $q_t$  are de-meaned measures of price and quantity of the good,  $w_t$  represents a de-meaned measure of weather, where higher levels of  $w_t$  increase crop yields.

**2.1 Create Data** To generate the data, first solve for the market clearing price  $p_t$ . The exogenous variables are drawn i.i.d. (each period) from the following distributions:

- $w_t \sim U(-3,3)$
- $\nu_t \sim N(0,1)$
- $\varepsilon \sim N(0,2)$

Generate the data for 100 periods t = 1 to t = 100 where:

$$\gamma = 0.5, \quad \eta = 1.5, \quad \delta = -1$$

Graph and example data set with  $p_t$  against quantity demanded  $(q_t)$ .

**2.2 Estimating Demand** Write a function that simulates 1000 iterations of this data set, running a regression estimating  $\delta$ . Collect the estimates and graph on a density plot.

**2.3 IV** Now repeat **2.2** using  $w_t$  as an instrument for price.

**2.4 Invalid Instrument?** Generate the data again, this time splitting observations into two groups of equal size, and make the following correction:

- In odd periods:  $\varepsilon_t \sim N(0,2)$
- In even periods:  $\varepsilon_t \sim N(0,2) + 0.2 \cdot w_t$

That is, there is correlation between the demand disturbances and the weather (bad weather makes consumers discouraged).

Repeat 2.2 and 2.3 again with this data generating process

#### Group 3: Regression Discontinuity

Consider the following scenario:

You want to estimate the effect of college on earnings. A state college only accepts SAT math scores above 400, you have access to a high school's record of 1000 students' SAT scores  $(SAT_i)$  and annual income many years later  $(Y_i)$ . (Assume everyone who scored over 400 went to college).

The data generating process for income is determined by the following equation:

$$Y_i = \alpha + \delta \cdot D_i + \beta_1 \cdot SAT_i \cdot (1 - D_i) + \beta_2 \cdot SAT_i \cdot D_i + \varepsilon_i$$

Where : 
$$D_i = \begin{cases} 1 & \text{if} \quad SAT_i \ge 400 \\ 0 & \text{if} \quad SAT_i < 400 \end{cases}$$

Assume  $\beta_1 < \beta_2$ ,  $\varepsilon_i \sim N(0, \sigma_{\varepsilon})$ .

Additionally, suppose SAT scores are distributed according to  $SAT_i \sim N(500, 120)$  (pretty close to reality) Which variable determines the causal effect of going to college on income?

#### 3.1 Create Data: Let:

- $\alpha = 10,000$
- $\delta = 500$
- $\beta_1 = 2$
- $\beta_2 = 3$
- $\sigma_{\varepsilon} = 300$

Create a function that generates a dataset of 1,000 students according to this scenario.

Plot a sample generation, labeling the cut off and whether an observation went to college or not.

**3.2 Regression with Same Slope** Write a function that simulates generating the data and running a *Regression Discontinuity Design* linear regression that assumes the slope of the line is **the same on either side** of the cut off.

Run this simulation 1,000 times, collecting the estimates of  $\delta$ , and plot the density of these estimates.

Are these estimates biased, unbiased, or ambiguous?

**3.3 Regression with Different Slopes** Repeat **3.2**, this time **allowing for differences** in the slope of the regression line on either side of the cut off.

Are these estimates biased, unbiased, or ambiguous?