

Lecture#11  
Difference – in – difference  
regression  
Causal parameters part II

# Difference estimator

- Let's assume a discrete treatment:

$X_i = 1$  if individual  $i$  gets the treatment

$X_i = 0$  if individual  $i$  does not get the treatment

# Randomized Control Trial (RCT)

- In economics:
  1. The researcher decides on what the experiment is.
  2. The researcher decides what the population of interest is.
  3. The researcher draws a random sample.
  4. Individuals in the random sample are randomly allocated into control and treatment groups.

$$Y_i = \beta_0 + \beta_1 X_i + u_i$$

- Q1: what is the interpretation of  $\beta_1$ ?
- Q2: is there an omitted variable problem?
- Q3: what if individuals truly randomized and the researcher observes other characteristics besides  $Y_i, X_i$ ?

- Q4: is there any reason to include control variables ( $W_i$ )?
  1. Efficiency -  $cor(X_i, W_i)=0$  by design.
  2. Control for randomization: if  $\beta_1$  without controls  $\neq$   $\beta_1$  with controls, then RCT has failed.

# RCT and Difference estimator

Control:  $Y_i = \beta_0 + u_i$

Treatment:  $Y_i = \beta_0 + \beta_1 + u_i$

Difference:  $\Delta Y_i = \beta_1$

# So what is DiD?

- Imagine the researcher has 2 consecutive observations / individual:
  1. Before experiment
  2. After experiment

# Examples of DiD setups

- WFH:

1. performance of workers before and after WFH allowed.
2. difference in change of performance between WFH and those who stay in office.



# Examples of DiD setups

- VI:
  1. Ticket prices in theatres before and after VI is allowed (notice difference to Gil data).
  2. difference in change of ticket prices between VI and non-VI theatres.

# Regression framework

Write generally

$$Y_{it} = \beta_0 + \beta_{group}X_i + \beta_{period}D_{after,it} + \beta_1X_iD_{after,it} + u_i + u_{it}$$

$X_i$  = 1 if in control group, 0 otherwise.

$D_{after,it}$  = 1 if period after, 0 otherwise.

$u_i$  = observation unit unobservables ("fixed effect").

# Regression framework – control group

For control group,  $X_i=0$ .

Therefore can rewrite

$$Y_{it} = \beta_0 + \beta_{group}X_i + \beta_{period}D_{after,it} + \beta_1X_iD_{after,it} + u_i + u_{it}$$

as

$$Y_{it} = \beta_0 + \beta_{period}D_{after,it} + u_i + u_{it}$$

# Regression framework – control group

$$Y_{it} = \beta_0 + \beta_{period}D_{after,it} + u_i + u_{it}$$

1. After  $Y_{i2} = \beta_0 + \beta_{period} + u_i + u_{i2}$

2. Before  $Y_{i1} = \beta_0 + u_i + u_{i1}$

3. Diff  $\Delta Y_i = \beta_{period} + u_{i2} - u_{i1}$

# Regression framework – treatment group

For treatment group,  $X_i=1$ .

Therefore can rewrite

$$Y_{it} = \beta_0 + \beta_{group}X_i + \beta_{period}D_{after,it} + \beta_1X_iD_{after,it} + u_i + u_{it}$$

as

$$Y_{it} = \beta_0 + \beta_{group} + \beta_{period}D_{after,it} + \beta_1D_{after,it} + u_i + u_{it}$$

# Regression framework – treatment group

$$Y_{it} = \beta_0 + \beta_{group} + \beta_{period}D_{after,it} + \beta_1 D_{after,it} + u_i + u_{it}$$

1. After

$$Y_{i2} = \beta_0 + \beta_{group} + \beta_{period} + \beta_1 + u_i + u_{i2}$$

2. Before

$$Y_{i1} = \beta_0 + \beta_{group} + u_i + u_{i1}$$

3. Diff

$$\Delta Y_i = \beta_{period} + \beta_1 + u_{i2} - u_{i1}$$

# DiD

1. Treatment group:  $\Delta Y_i = \beta_{period} + \beta_1$
2. Control group:  $\Delta Y_i = \beta_{period}$
3. DiD:  $\Delta \Delta Y_i = \beta_1$

# Diff vs. DiD

1. DiD needs data over at least 2 periods.
2. DiD allows for individual-specific constants.
3. → DiD doesn't necessitate randomization.
4. Identifying assumption #1: "common trends".
5. Identifying assumption #2:  $E[u_{it} | X_{it}, D_{after,it}, u_i] = 0$



# Diff vs. DiD

- Identifying assumption #1: common trends.
- When would this be violated?
- Technically,  $\beta_{period|control} \neq \beta_{period|treatment}$ .
- Call  $\beta_{period|treatment} - \beta_{period|control} = \Delta\beta_{period}$

# DiD

1. Treatment group:

$$\Delta Y_{it} = \beta_1 + \beta_{period|treatment}$$

2. Control group:

$$\Delta Y_{it} = \beta_{period|control}$$

3. DiD:

$$\Delta\Delta Y_{it} = \beta_1 + \Delta\beta_{period}$$

# Diff vs. DiD

- Substantively?
- Example #1: WFH: those that know their productivity is (permanently) declining decide to work from home (or office).
- Example #2: Think of the effect of hiring a new CEO on firm performance.
- Firm observes performance is (permanently) declining compared to peers, and therefore hires a new CEO.

# Diff vs. DiD

- Identifying assumption #2:  $E[u_{it} | X_{it}, D_{after,it}, u_i] = 0$ .
- → selection into treatment can depend on individual-specific “things” that are constant over the periods.
- Even 2-period DiD allows control variables.
- Controls may be more important than in an RCT.

# Diff vs. DiD

- Identifying assumption #2.
- When would this be violated?
- Technically, the "shock" in the 1st period leads somebody to (not) choose the treatment.

# Diff vs. DiD

- Substantively?
- Example #1: WFH: those that know their productivity was (temporarily) lower decide to work from home (or office).
- Example #2: Think of the effect of hiring a new CEO on firm performance.
- Firm observes a shock to performance compared to peers, and therefore hires a new CEO.

# What data for more than 2 periods?

- More data always a plus.
- Makes distinction between (differential) trends and temporary shocks clearer.
- Can allow for more flexible models (e.g. treatment of time; testing of common trends).
- BUT: notice that one stretches what  $u_i$  captures.
- Remember: Even 2-period DiD allows control variables.

# A complication on all causal estimators

- What have we assumed about the effect of treatment on the control group?
- That there is none.
- = "no general equilibrium effects"
- = "**Stable Unit Treatment Value Assumption (SUTVA)**"



# A complication on all causal estimators

- When is this an issue?
- In a lab, think of infectious diseases.
- Regarding human behavior, think of interactions (markets).
- Important but difficult topic. We will neglect it.