DD v1

DD OLS

M: Set-up

M: Estimation

Lecture 3: Difference in Difference 1 of 2

January 29, 2023

Potential Outcome

DD OLS

M: Set-u

M: Estimation

Course Administration

- 1. Summaries are graded
- 2. Thanks for proposals will grade by next week

- 3. Problem set 2 posted
- 4. Will post answers to PS 1, including code
- 5. Any other issues?

▲□▶ ▲□▶ ▲目▶ ▲目▶ ▲目 ● ● ●



Potential Outcomes

DD OLS

M: Set-

M: Estimation

Today

Diff-in-diff overview

- $1. \ \mbox{When to use diff-in-diff}$
- 2. Simplest formulation: before and after only
- 3. With multiple obs before and after

Next time: including a "trend" in a regression

Milligan and the Stork

- 1. Estimation problem
- 2. Data
- 3. Diff-in-diff in chart
- 4. Diff-in-diff in table
- 5. Diff-in-diff in regression

・ロト・日本・日本・日本・日本・日本

When DD

DD v1

Potential Outcomes

DD OLS

M: Set-up

M: Estimation

Motivating Diff-in-Diff

otential Outcome

DD OLS

M: Set-up

▲□▶ ▲□▶ ▲ □▶ ▲ □▶ ▲ □ ● ● ● ●

M: Estimation

Motivating Diff-in-Diff

- 1. When should you use diff-in-diff?
- 2. Motivating example
- 3. Diff-in-diff v1
- 4. With potential outcomes notation
- 5. Writing and interpreting a diff-in-diff regression

Next time: validity tests and trends

0. Why Bother? Or, Why Not Regression with Covariates?

• OLS with covariates is unlikely to deliver a causal estimate of \hat{eta}



0. Why Bother? Or, Why Not Regression with Covariates?

- OLS with covariates is unlikely to deliver a causal estimate of \hat{eta}
- So we need a causal strategy
- Diff-in-diff is a causal strategy

DD v1

DD OLS

M: Set-up

M: Estimation

1. When to use diff-in-diff?

When DD

et-up

M: Estimation

When to Use a Difference in Difference Methodology?

- To evaluate the impact of a policy at an aggregate level
- Where you have some potential control group

Set-up

M: Estimation

When to Use a Difference in Difference Methodology?

- To evaluate the impact of a policy at an aggregate level
- Where you have some potential control group
- Groups are frequently but not necessarily geographic
- For example: national policy that affects some groups by not others
- Examples?

When DD

When DD

Set-up

M: Estimation

When to Use a Difference in Difference Methodology?

- To evaluate the impact of a policy at an aggregate level
- Where you have some potential control group
- Groups are frequently but not necessarily geographic
- For example: national policy that affects some groups by not others
- Examples? EITC evaluation that compares women with children versus those without

・ロト・西ト・山田・山田・山下

When DD

Motivating Ex.

DD v1

Potential Outcomes

DD OLS

M: Set-up

M: Estimation

2. Motivating Example

◆□ > ◆□ > ◆ □ > ◆ □ > ◆ □ > ● ● ●

▲□▶ ▲□▶ ▲ □▶ ▲ □▶ ▲ □ ● ● ● ●

Motivating Example: Card and Krueger, AER, 1991

Policy

- April 1992
 - NJ and PA have the same minimum wage of \$4.25/hour
- April 1992 onward
 - NJ raises state minimum wage to \$5.05/hour, no change in PA

Motivating Example: Card and Krueger, AER, 1991

Policy

- April 1992
 - NJ and PA have the same minimum wage of \$4.25/hour
- April 1992 onward
 - NJ raises state minimum wage to \$5.05/hour, no change in PA

Data

- C&K collect data on employment and wages at fast food places in NJ and E PA
- observe data from February to November 1992

・ロト・日本・日本・日本・日本

DD M

ivating Ex.

DD v1

Potential Outcomes

DD OLS

M: Set-up

M: Estimation

3. Diff-in-Diff Version 1

M: Estimation

With This Setup, How Do We Estimate?

We observe

- Employment in NJ before and after
 - NJ_B and NJ_A
- Employment in PA before and after
 - PA_B and PA_A

With This Setup, How Do We Estimate?

Why not $(NJ_A - NJ_B)$?

We observe

- Employment in NJ before and after
 - NJ_B and NJ_A
- Employment in PA before and after
 - PA_B and PA_A

With This Setup, How Do We Estimate?

Why not $(NJ_A - NJ_B)$? Estimating

We observe

- Employment in NJ before and after
 - NJ_B and NJ_A
- Employment in PA before and after
 - PA_B and PA_A

▲□▶ ▲□▶ ▲ □▶ ▲ □▶ ▲ □ ● ● ● ●

With This Setup, How Do We Estimate?

Why not $(NJ_A - NJ_B)$? Estimating

We observe

- Employment in NJ before and after
 - NJ_B and NJ_A
- Employment in PA before and after
 - PA_B and PA_A

 $(NJ_A - NJ_B) - (PA_A - PA_B)$

With This Setup, How Do We Estimate?

Why not $(NJ_A - NJ_B)$? Estimating

We observe

- Employment in NJ before and after
 - NJ_B and NJ_A
- Employment in PA before and after
 - PA_B and PA_A

 $(NJ_A - NJ_B) - (PA_A - PA_B)$

or $(NJ_A - PA_A) - (NJ_B - PA_B)$

▲□▶ ▲□▶ ▲三▶ ▲三▶ 三三 - シスペ

M: Estimation

In Graph Form



This is NJ Only – Why Not This Comparison?



M: Estimation

Here are Both: Where is Double Difference?



M: Estimation

Double Difference v.1



M: Estimation

Double Difference v.2



M: Estimation

Or, the Implicit Comparison



Whe

DD v1

Potential Outcomes

DD OLS

M: Set-up

M: Estimation

4. Potential Outcomes Framework

・ロト ・日・・日・・日・・ のくの

▲□▶ ▲□▶ ▲ □▶ ▲ □▶ ▲ □ ● ● ● ●

Card and Krueger in a Potential Outcomes Framework

- $Y_{0ist} \equiv$ fast food employment at restaurant *i*, state *s*, period *t* with the low minimum wage
- $Y_{1ist} \equiv$ fast food employment at restaurant *i*, state *s*, period *t* with the high minimum wage
- Recall that we only observe one of these for any given t
- State $s \in {NJ, PA}$
- Time period $t \in \{\text{before, after}\}$

◆□ > ◆□ > ◆ Ξ > ◆ Ξ > → Ξ = の < @

M: Estimation

Key Assumptions

1.
$$E[Y_{0ist}|s,t] = \gamma_s + \lambda_t$$

DD M

M: Set-up

M: Estimation

Key Assumptions

- 1. $E[Y_{0ist}|s,t] = \gamma_s + \lambda_t$
 - $\gamma_s \equiv$ state fixed effects
 - $\lambda_t \equiv \text{time fixed effects}$

Motivating Ex

DD OLS

M: Set-ı

▲□▶ ▲□▶ ▲□▶ ▲□▶ □ の00

M: Estimation

Key Assumptions

- 1. $E[Y_{0ist}|s,t] = \gamma_s + \lambda_t$
 - $\gamma_s \equiv$ state fixed effects
 - $\lambda_t \equiv \text{time fixed effects}$
 - In words: the outcome, conditional on state and time, can be explained by something fixed about the state, and something fixed in a given time period for all states
 - Note that γ_s does not have to be the same for all states
 - · Give an example where you think this isn't true

DD Mot

▲□▶ ▲□▶ ▲□▶ ▲□▶ □ の00

M: Estimation

Key Assumptions

1. $E[Y_{0ist}|s,t] = \gamma_s + \lambda_t$

- $\gamma_s \equiv$ state fixed effects
- $\lambda_t \equiv \text{time fixed effects}$
- In words: the outcome, conditional on state and time, can be explained by something fixed about the state, and something fixed in a given time period for all states
- Note that γ_s does not have to be the same for all states
- Give an example where you think this isn't true
- This is the "common" or "parallel trends" assumption

Motivating

ig Ex.

DD OLS

M: Set-up

M: Estimation

Key Assumptions

1. $E[Y_{0ist}|s,t] = \gamma_s + \lambda_t$

- $\gamma_s \equiv$ state fixed effects
- $\lambda_t \equiv \text{time fixed effects}$
- In words: the outcome, conditional on state and time, can be explained by something fixed about the state, and something fixed in a given time period for all states
- Note that γ_s does not have to be the same for all states
- Give an example where you think this isn't true
- This is the "common" or "parallel trends" assumption
- 2. $E[Y_{1ist} Y_{0ist}|s, t] = \delta$
 - Change between treated and untreated states is a level difference it's additive, not multiplicative, or some other function

Motivatin

DD v1

DD OLS

M: Set-up

M: Estimation

5. Difference in difference estimation

◆□ > ◆□ > ◆ Ξ > ◆ Ξ > → Ξ → のへで

Regression Specification and Interpretation

• In the regression world, we write the regression equation as

Regression Specification and Interpretation

• In the regression world, we write the regression equation as

$$Y_{\textit{ist}} = lpha + \gamma \textit{NJ}_{\textit{s}} + \lambda \textit{d}_{t} + \delta \textit{NJ}_{\textit{s}} * \textit{d}_{t} + \epsilon_{\textit{ist}}$$

and note that $NJ_s * d_t$ is the treatment


• In the regression world, we write the regression equation as

$$Y_{\textit{ist}} = lpha + \gamma \textit{NJ}_{\textit{s}} + \lambda \textit{d}_{t} + \delta \textit{NJ}_{\textit{s}} * \textit{d}_{t} + \epsilon_{\textit{ist}}$$

and note that $NJ_s * d_t$ is the treatment

- Break down this equation
 - PA before?

◆□ ▶ ◆□ ▶ ◆ 三 ▶ ◆ 三 ▶ ● ○ ○ ○ ○

• In the regression world, we write the regression equation as

$$Y_{ist} = lpha + \gamma N J_s + \lambda d_t + \delta N J_s * d_t + \epsilon_{ist}$$

and note that $NJ_s * d_t$ is the treatment

- Break down this equation
 - PA before? α

Regression Specification and Interpretation

• In the regression world, we write the regression equation as

$$Y_{\textit{ist}} = lpha + \gamma \textit{NJ}_{\textit{s}} + \lambda \textit{d}_{t} + \delta \textit{NJ}_{\textit{s}} * \textit{d}_{t} + \epsilon_{\textit{ist}}$$

- Break down this equation
 - PA before? α
 - PA after?

• In the regression world, we write the regression equation as

$$Y_{\textit{ist}} = lpha + \gamma \textit{NJ}_{\textit{s}} + \lambda \textit{d}_{t} + \delta \textit{NJ}_{\textit{s}} * \textit{d}_{t} + \epsilon_{\textit{ist}}$$

and note that $NJ_s * d_t$ is the treatment

- Break down this equation
 - PA before? α
 - PA after? $\alpha + \lambda$

・ロト・日本・日本・日本・日本・日本・日本

▲□▶ ▲□▶ ▲ □▶ ▲ □▶ ▲ □ ● ● ● ●

Regression Specification and Interpretation

• In the regression world, we write the regression equation as

$$Y_{\textit{ist}} = lpha + \gamma \textit{NJ}_{\textit{s}} + \lambda \textit{d}_{t} + \delta \textit{NJ}_{\textit{s}} * \textit{d}_{t} + \epsilon_{\textit{ist}}$$

- Break down this equation
 - PA before? α
 - PA after? $\alpha + \lambda$
 - NJ before?

▲□▶ ▲□▶ ▲□▶ ▲□▶ □ の00

Regression Specification and Interpretation

• In the regression world, we write the regression equation as

$$Y_{\textit{ist}} = lpha + \gamma \textit{NJ}_{\textit{s}} + \lambda \textit{d}_{t} + \delta \textit{NJ}_{\textit{s}} * \textit{d}_{t} + \epsilon_{\textit{ist}}$$

- Break down this equation
 - PA before? α
 - PA after? $\alpha + \lambda$
 - NJ before? $\alpha + \gamma$

▲□▶ ▲□▶ ▲□▶ ▲□▶ □ の00

Regression Specification and Interpretation

• In the regression world, we write the regression equation as

$$Y_{ist} = lpha + \gamma N J_s + \lambda d_t + \delta N J_s * d_t + \epsilon_{ist}$$

- Break down this equation
 - PA before? α
 - PA after? $\alpha + \lambda$
 - NJ before? $\alpha + \gamma$
 - NJ after?

▲□▶ ▲□▶ ▲□▶ ▲□▶ □ の00

Regression Specification and Interpretation

• In the regression world, we write the regression equation as

$$Y_{\textit{ist}} = lpha + \gamma \textit{NJ}_{\textit{s}} + \lambda \textit{d}_{t} + \delta \textit{NJ}_{\textit{s}} * \textit{d}_{t} + \epsilon_{\textit{ist}}$$

- Break down this equation
 - PA before? α
 - PA after? $\alpha + \lambda$
 - NJ before? $\alpha + \gamma$
 - NJ after? $\alpha + \gamma + \lambda + \delta$

▲□▶ ▲□▶ ▲ □▶ ▲ □▶ ▲ □ ● ● ● ●

Regression Specification and Interpretation

• In the regression world, we write the regression equation as

$$Y_{\textit{ist}} = lpha + \gamma \textit{NJ}_{\textit{s}} + \lambda \textit{d}_{t} + \delta \textit{NJ}_{\textit{s}} * \textit{d}_{t} + \epsilon_{\textit{ist}}$$

- Break down this equation
 - PA before? α
 - PA after? $\alpha + \lambda$
 - NJ before? $\alpha + \gamma$
 - NJ after? $\alpha + \gamma + \lambda + \delta$
 - Can you see diff-in-diff?

• In the regression world, we write the regression equation as

$$Y_{\textit{ist}} = lpha + \gamma \textit{NJ}_{\textit{s}} + \lambda \textit{d}_{t} + \delta \textit{NJ}_{\textit{s}} * \textit{d}_{t} + \epsilon_{\textit{ist}}$$

and note that $NJ_s * d_t$ is the treatment

- Break down this equation
 - PA before? α
 - PA after? $\alpha + \lambda$
 - NJ before? $\alpha + \gamma$
 - NJ after? $\alpha + \gamma + \lambda + \delta$
 - Can you see diff-in-diff?

$$(NJ after - NJ before) - (PA after - PA before) =$$

・ロト ・ 日 ・ ・ ヨ ・ ・ 日 ・ うへの

• In the regression world, we write the regression equation as

$$Y_{\textit{ist}} = lpha + \gamma \textit{NJ}_{\textit{s}} + \lambda \textit{d}_{t} + \delta \textit{NJ}_{\textit{s}} * \textit{d}_{t} + \epsilon_{\textit{ist}}$$

and note that $NJ_s * d_t$ is the treatment

- Break down this equation
 - PA before? α
 - PA after? $\alpha + \lambda$
 - NJ before? $\alpha + \gamma$
 - NJ after? $\alpha + \gamma + \lambda + \delta$
 - Can you see diff-in-diff?

・ロト・西ト・山田・山田・山下

• In the regression world, we write the regression equation as

$$Y_{\textit{ist}} = lpha + \gamma \textit{NJ}_{\textit{s}} + \lambda \textit{d}_{t} + \delta \textit{NJ}_{\textit{s}} * \textit{d}_{t} + \epsilon_{\textit{ist}}$$

and note that $NJ_s * d_t$ is the treatment

- Break down this equation
 - PA before? α
 - PA after? $\alpha + \lambda$
 - NJ before? $\alpha + \gamma$
 - NJ after? $\alpha + \gamma + \lambda + \delta$
 - Can you see diff-in-diff?

・ロト・「「「・」」・ 「」・ 「」・ (「」・

• In the regression world, we write the regression equation as

$$Y_{\textit{ist}} = lpha + \gamma \textit{NJ}_{\textit{s}} + \lambda \textit{d}_{t} + \delta \textit{NJ}_{\textit{s}} * \textit{d}_{t} + \epsilon_{\textit{ist}}$$

and note that $NJ_s * d_t$ is the treatment

- Break down this equation
 - PA before? α
 - PA after? $\alpha + \lambda$
 - NJ before? $\alpha + \gamma$
 - NJ after? $\alpha + \gamma + \lambda + \delta$
 - Can you see diff-in-diff?

• Note that you can estimate this with sample means! A very good place to start, for reasons we will talk about next week

VI: Set-up

▲□▶ ▲□▶ ▲□▶ ▲□▶ □ の00

M: Estimation

Recap: Key Parts

Key Assumption

- In the absence of treatment, treatment and control observations would have evolved in parallel fashion
- AKA, "parallel trends"
- Fundamentally untestable
- Phrased differently: the only difference between treatment and control, apart from any level differences, is treatment

otential Outcome

DD OLS

M: Set-up

M: Estimation

Recap: Key Parts

Key Assumption

- In the absence of treatment, treatment and control observations would have evolved in parallel fashion
- AKA, "parallel trends"
- Fundamentally untestable
- Phrased differently: the only difference between treatment and control, apart from any level differences, is treatment

Why a regression?

- a convenient way to get estimates and standard errors
- can do more policies (e.g. put in value of wage changes)
- can add controls, if parallel trend assumption is only valid conditionally, or if we want to reduce variance

Admin

When DD

DD v1

Potential Outcomes

DD OLS

M: Set-up

M: Estimation

Setting up the Milligan et al paper

◆□ > ◆□ > ◆ Ξ > ◆ Ξ > → Ξ = の < @

M: Estimation

Research Question and Estimation Problem

1. What is the research question?



▲□▶ ▲□▶ ▲ □▶ ▲ □▶ ▲ □ ● ● ● ●

Research Question and Estimation Problem

- 1. What is the research question?
- 2. What does this mean? "... empirical researchers have shown great interest in trying to uncover evidence of a relationship between prices and fertility. The endogeneity of key variables has frustrated this effort.

▲□▶ ▲□▶ ▲□▶ ▲□▶ □ の00

M: Estimation

Research Question and Estimation Problem

- 1. What is the research question?
- 2. What does this mean? "... empirical researchers have shown great interest in trying to uncover evidence of a relationship between prices and fertility. The endogeneity of key variables has frustrated this effort. Women may have unobserved proclivities for different family sizes.

Research Question and Estimation Problem

- 1. What is the research question?
- 2. What does this mean? "... empirical researchers have shown great interest in trying to uncover evidence of a relationship between prices and fertility. The endogeneity of key variables has frustrated this effort. Women may have unobserved proclivities for different family sizes. If differences in these proclivities lead to different human capital accumulation and marital decisions, then the opportunity cost of time out of the labor market will be jointly determined with fertility."

Omitted Variable Bias

• What are the two components of an omitted variable problem/story? An omitted variable is



Omitted Variable Bias

- What are the two components of an omitted variable problem/story? An omitted variable is
 - 1. correlated with the treatment
 - 2. and with the outcome conditional on covariates (aka the error)

Omitted Variable Bias

- What are the two components of an omitted variable problem/story? An omitted variable is
 - 1. correlated with the treatment
 - 2. and with the outcome conditional on covariates (aka the error)
- Give an example of a potential omitted variable in this paper

What's the Argument for How the ANC Solves This Problem?

M: Estimation

What's the Argument for How the ANC Solves This Problem?

- Large subsidy at peak > \$15k today
- Plausibly not related to fertility
- Everyone in Quebec treated similarly
- $\bullet\, \rightarrow$ compare Quebec and Rest of Canada
- Can also use a triple difference

▲□▶ ▲□▶ ▲□▶ ▲□▶ ▲□ ▶ ▲□

M: Estimation

Data and Units of Observation and Analysis

What are the two data sources?

Vital statistics data

M: Estimation

Data and Units of Observation and Analysis

What are the two data sources?

Vital statistics data

• from birth records

M: Estimation

Data and Units of Observation and Analysis

What are the two data sources?

Vital statistics data

- from birth records
 - \rightarrow unit of observation is woman in a year

M: Estimation

Data and Units of Observation and Analysis

What are the two data sources?

Vital statistics data

- from birth records
 - \rightarrow unit of observation is woman in a year
- aggregates to fertility rates by cohort/province/parity

M: Estimation

Data and Units of Observation and Analysis

What are the two data sources?

Vital statistics data

- from birth records
 - \rightarrow unit of observation is woman in a year
- aggregates to fertility rates by cohort/province/parity
 - $\bullet \ \to \ {\rm unit \ of \ analysis \ is \ province/year}$

M: Estimation

Data and Units of Observation and Analysis

What are the two data sources?

Vital statistics data

- from birth records
 - \rightarrow unit of observation is woman in a year
- aggregates to fertility rates by cohort/province/parity
 - $\bullet \ \to \ {\rm unit \ of \ analysis \ is \ province/year}$

Canadian Census data

M: Estimation

Data and Units of Observation and Analysis

What are the two data sources?

Vital statistics data

- from birth records
 - \rightarrow unit of observation is woman in a year
- aggregates to fertility rates by cohort/province/parity
 - $\bullet \ \to {\rm unit \ of \ analysis \ is \ province/year}$

Canadian Census data

- from 1991 and 1996
- covering five prior years
- unit of observation and analysis is family

M: Estimation

Basic Diff-in-diff

- We need
 - treated and untreated
 - before and after

Potential Outcome

DD OLS

M: Set-up

M: Estimation

Basic Diff-in-diff

- We need
 - treated and untreated
 - before and after
- What are these here?

otential Outcome

DD OLS

M: Set-up

M: Estimation

Basic Diff-in-diff

- We need
 - treated and untreated
 - before and after
- What are these here?
 - before and after: before ANC and during ANC

M: Estimation

Basic Diff-in-diff

- We need
 - treated and untreated
 - before and after
- What are these here?
 - before and after: before ANC and during ANC
 - treated and untreated: Quebec and Rest of Canada

・ロト ・日・・日・・日・・ つくぐ
Admin

DD v1

DD OLS

M: Set-up

M: Estimation

Estimation in Milligan

▲□▶ ▲□▶ ▲ □▶ ▲ □▶ ▲ □ ● ● ● ●

Basic Diff-in-diff in Figure 1

For the simplest diff-in-diff, what are the two comparisons?



Basic Diff-in-diff in Figure 1

For the simplest diff-in-diff, what are the two comparisons?



M: Estimation

The Diff-in-diff in Table 5

Region	м	Mean		Diff	
	1991 (1)	1996 (2)	(2) - (1) = (3)	Differences (4)	Increase (5)
A. All Parities					
Ouebec	0.418	0.451	0.033		
	(0.003)	(0.004)	(0.005)		
n	20,285	16,453			
Rest of Canada	0.432	0.441	0.009	0.024	5.5%
	(0.002)	(0.002)	(0.003)	(0.006)	
n	54,115	46,032		. ,	

• How do you calculate 0.418?

M: Estimation

The Diff-in-diff in Table 5

Mean		Trend Difference	Difference in	Descenteres
1991 (1)	1996 (2)	(2) - (1) = (3)	Differences (4)	Increase (5)
0.418 (0.003)	0.451 (0.004)	0.033 (0.005)		
20,285	16,453			
0.432 (0.002)	0.441 (0.002)	0.009 (0.003)	0.024 (0.006)	5.5%
	0.418 (0.003) 20,285 0.432 (0.002)	Mean 1991 1996 (1) (2) 0.418 0.451 (0.003) (0.004) 20.285 16.453 0.432 0.441 (0.002) (0.002)	$\begin{tabular}{ c c c c c c c c c c c c c c c c c c c$	Mean Trend Difference in Means, (2) - (1) = Difference in Difference in (2) - (1) = 0.418 0.451 0.033 (0.003) (0.004) (0.005) 20,285 16,453 0.432 0.441 0.009 0.024 (0.002) (0.003) (0.006) 0.024

- How do you calculate 0.418?
- And 0.441?

M: Estimation

The Diff-in-diff in Table 5

Region	М	Mean		Difference in	Descenteres
	1991 (1)	1996 (2)	(2) $-(1) =$ (3)	Differences (4)	Increase (5)
A. All Parities					
Quebec	0.418 (0.003)	0.451 (0.004)	0.033 (0.005)		
n Rest of Canada	20,285 0.432 (0.002)	16,453 0.441 (0.002)	0.009 (0.003)	0.024 (0.006)	5.5%
n	54,115	46,032			

- How do you calculate 0.418?
- And 0.441?
- And Col. 3, 0.033?

▲□▶ ▲圖▶ ▲≣▶ ▲≣▶ ▲国 ● ④�♡

M: Estimation

The Diff-in-diff in Table 5

Region	М	Mean		Difference in	Description
	1991 (1)	1996 (2)	(2) - (1) = (3)	Differences (4)	Increase (5)
A. All Parities					
Quebec	0.418 (0.003)	0.451 (0.004)	0.033 (0.005)		
n Rest of Canada	20,285 0.432 (0.002)	16,453 0.441 (0.002)	0.009 (0.003)	0.024 (0.006)	5.5%
n	54,115	46,032			

- How do you calculate 0.418?
- And 0.441?
- And Col. 3, 0.033?

• How do we find 0.024?

M: Estimation

The Diff-in-diff in Table 5

Region	М	Mean		Difference in	Percentage
	1991 (1)	1996 (2)	(2) - (1) = (3)	Differences (4)	Increase (5)
A. All Parities					
Quebec	0.418 (0.003)	0.451 (0.004)	0.033 (0.005)		
n	20,285	16,453	0.000	0.024	6.60
Rest of Canada	(0.002) 54,115	(0.002) 46,032	(0.003)	(0.006)	5.5%

- How do you calculate 0.418?
- And 0.441?
- And Col. 3, 0.033?

- How do we find 0.024?
- And 5.5%?

M: Estimation

The Diff-in-diff in Table 5

Region	М	Mean		Difference in	Percentage
	1991 (1)	1996 (2)	(2) - (1) = (3)	Differences (4)	Increase (5)
A. All Parities					
Quebec	0.418 (0.003)	0.451 (0.004)	0.033 (0.005)		
n Dart of Conside	20,285	16,453	0.000	0.024	5 50
Rest of Canada	(0.002) 54,115	(0.002) 46,032	(0.003)	(0.006)	5.5%

- How do you calculate 0.418?
- And 0.441?
- And Col. 3, 0.033?

- How do we find 0.024?
- And 5.5%? (0.024)/(0.418+0.009)

M: Estimation

What Regression Equation Parallels the Diff-in-diff?

Region	м	Mean		Difference in	Percentage
	1991 (1)	1996 (2)	(2) - (1) = (3)	Differences (4)	Increase (5)
A. All Parities					
Quebec	0.418 (0.003)	0.451 (0.004)	0.033 (0.005)		
n	20,285	16,453			
Rest of Canada	0.432 (0.002)	0.441 (0.002)	0.009 (0.003)	0.024 (0.006)	5.5%
n	54,115	46,032			



-up

M: Estimation

What Regression Equation Parallels the Diff-in-diff?

Region	м	Mean		Difference in	Percentage
	1991 (1)	1996 (2)	(2) - (1) = (3)	Differences (4)	Increase (5)
A. All Parities					
Quebec	0.418 (0.003)	0.451 (0.004)	0.033 (0.005)		
n	20,285	16,453			
Rest of Canada	0.432 (0.002)	0.441 (0.002)	0.009 (0.003)	0.024 (0.006)	5.5%
n	54,115	46,032			

 $\mathsf{fertility}_{i,j,t} = \beta_0 + \frac{\beta_1}{\mathsf{Quebec}_i} * 1\{t = 1996\}_t + \beta_2 \mathsf{Quebec}_i + \beta_3 1\{t = 1996\}_t + \beta_4 X_{i,j,t} + \epsilon_{i,j,t} + \beta_4 X_{i,j,t} + \beta_4 X_{$

◆□ > ◆□ > ◆ Ξ > ◆ Ξ > → Ξ → のへで

-up

M: Estimation

What Regression Equation Parallels the Diff-in-diff?

	Mean		Trend Difference	Difference in	Percentage
Region	1991 (1)	1996 (2)	(2) - (1) = (3)	Differences (4)	Increase (5)
A. All Parities					
Quebec	0.418 (0.003)	0.451 (0.004)	0.033 (0.005)		
n	20,285	16,453			
Rest of Canada	0.432 (0.002)	0.441 (0.002)	0.009 (0.003)	0.024 (0.006)	5.5%
n	54,115	46,032			

 $\mathsf{fertility}_{i,j,t} = \beta_0 + \frac{\beta_1}{\mathsf{Quebec}} \mathsf{Quebec}_i + \beta_2 \mathsf{Quebec}_i + \beta_3 \mathsf{I}\{t = 1996\}_t + \beta_4 X_{i,j,t} + \epsilon_{i,j,t} + \beta_4 \mathsf{Quebec}_i + \beta_4 \mathsf{$

When estimated without covariates $X_{i,j,t}$, β_1 is **the same** as the estimate above.

otential Outcomes

DD OLS

M: Set-up

M: Estimation

What is the Underlying Assumption Here?

Can state the assumption many different ways

◆□ > ◆□ > ◆ Ξ > ◆ Ξ > → Ξ = の < @

▲□▶ ▲□▶ ▲□▶ ▲□▶ □ の00

M: Estimation

What is the Underlying Assumption Here?

Can state the assumption many different ways

- Only differences between Quebec and ROC are time-invariant
- Fertility in Quebec would have evolved like that in ROC absent the policy
- There are no pre-treatment trends in fertility in Quebec

M: Estimation

What is the Underlying Assumption Here?

Can state the assumption many different ways

- Only differences between Quebec and ROC are time-invariant
- Fertility in Quebec would have evolved like that in ROC absent the policy
- There are no pre-treatment trends in fertility in Quebec

Can you test with Census data?

M: Estimation

What is the Underlying Assumption Here?

Can state the assumption many different ways

- Only differences between Quebec and ROC are time-invariant
- Fertility in Quebec would have evolved like that in ROC absent the policy
- There are no pre-treatment trends in fertility in Quebec

Can you test with Census data? no - we only have one pre-treatment period

◆□ > ◆□ > ◆ Ξ > ◆ Ξ > → Ξ = の < @

M: Estimation

Table 6: Regression Version

Independent Variable	(a)	(b)
Pseudo R ² 1996 dummy × Quebec 1996 dummy Implied percentage increase in probability of having a child	0.0003 0.024* (0.005) 0.009 (0.005) 5.6%	0.058 0.034* (0.006 0.013* (0.006 7.8%
Quebec One older child Two or more older children	-0.014* (0.007)	-0.021* (0.007 0.205* (0.016 -0.163* (0.011
Female age 25-34 Female Immigrant Female Inglophone Female high school Female high school Female post-high school		0.187* (0.009 0.032* (0.007 -0.047* (0.010 -0.049* (0.012 -0.015* (0.006 -0.086* (0.004 -0.192* (0.005
Male age 25-34 Male age 35-44 Male age 45+ Male Immigrant Male Anglophone Male high school		
maie pois-mgn school Male university degree Married Lives in urban area Family income (CS10,000) Provincial GDP growth		
Provincial migration rate Provincial education spending		_

- interpret coefficient for 1996 dummy
- interpret coefficient for 1996 dummy X Quebec

M: Estimation

And a Triple Difference!

	Mean		Trend Difference	Difference in	Percentage	Triple
Region	1991 (1)	1996 (2)	(2) - (1) = (3)	Differences (4)	Increase (5)	Difference (6)
A. All Parities						
Quebec	0.418	0.451	0.033			
	(0.003)	(0.004)	(0.005)			
n	20,285	16,453				
Rest of Canada	0.432	0.441	0.009	0.024	5.5%	
	(0.002)	(0.002)	(0.003)	(0.006)		
n	54,115	46,032				
B. Zero older children						
Ouebec	0.393	0.418	0.025			
4	(0.004)	(0.004)	(0.006)			
n	15.017	12,399	(<i>-</i>)			
Rest of Canada	0.398	0.407	0.009	0.016	4.0%	
	(0.002)	(0.003)	(0.003)	(0.007)		
n	38,754	33,338				
C. One older child						
Ouebec	0.627	0.677	0.050			
4	(0.009)	(0.009)	(0.013)			
n	3.207	2.475	(0.010)			
Rest of Canada	0.691	0.681	-0.010	0.060	9.7%	
	(0.005)	(0.006)	(0.008)	(0.015)		
n	8,262	7,088				
D. Two or more older children						
Ouebec	0.278	0.353	0.075			
******	(0.010)	(0.012)	(0.015)			
n	2,061	1.579	(
Rest of Canada	0.321	0.344	0.023	0.052	17.2%	0.036
_	(0.006)	(0.006)	(0.008)	(0.018)		(0.020)
n	7,099	5,606	,	,		, , , , , , , , , , , , , , , , , , , ,



Motivating Ex.

M: Set-up

◆□ > ◆□ > ◆ Ξ > ◆ Ξ > → Ξ = の < @

M: Estimation

Next Lecture

Read

- Janssen and Zhang, selected pages
- just through Section 4
- Summary due next week if you're on the list