DD v1

M: Set-up

M: Estimation

Lecture 3: Difference in Difference 1 of 2

September 20, 2023

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Course Administration

- 1. Summaries are graded
- 2. One page proposals graded

Course Administration

- 1. Summaries are graded
- 2. One page proposals graded

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



Today

Diff-in-diff overview

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

Including a "trend" in a regression

Milligan and the Stork

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

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Potential Outcomes

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Motivating Diff-in-Diff

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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

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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

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1. When to use diff-in-diff?

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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

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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?

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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

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2. Motivating Example

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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

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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

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3. Diff-in-Diff Version 1

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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
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With This Setup, How Do We Estimate?

Why not $(NJ_A - NJ_B)$? Estimating

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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

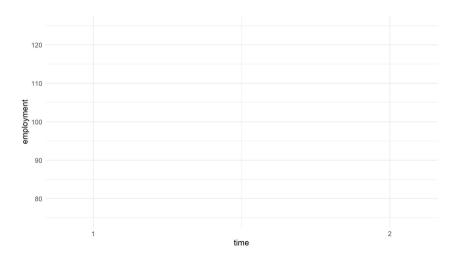
 $(NJ_A - NJ_B) - (PA_A - PA_B)$

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

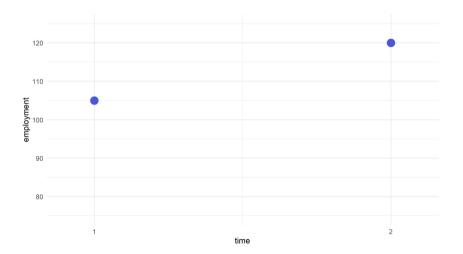
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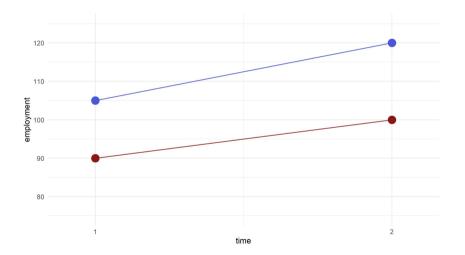
In Graph Form



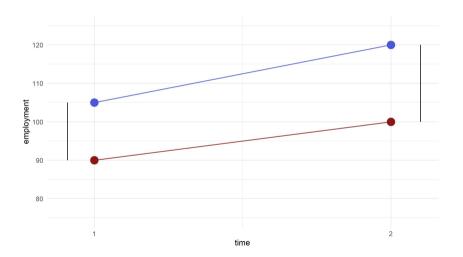
This is NJ Only – Why Not This Comparison?



Here are Both: Where is Double Difference?

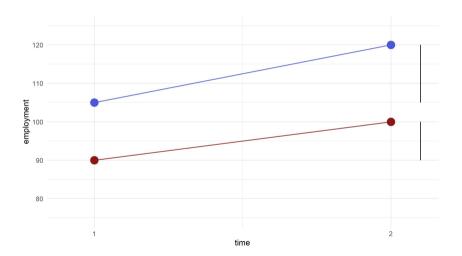


Double Difference v.1



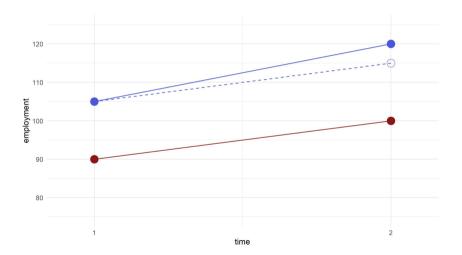
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Double Difference v.2



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Or, the Implicit Comparison



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4. Potential Outcomes Framework

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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 $t \in \{\text{before, after}\}$

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

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



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- 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 $\gamma_{\rm s}$ does not have to be the same for all states
 - · Give an example where you think this isn't true

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 - This is the "common" or "parallel trends" assumption

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 - 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

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Potential Outcomes

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5. Difference in difference estimation

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Regression Specification and Interpretation

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

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



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

and note that $NJ_s * d_t$ is the treatment

- Break down this equation
 - PA before?

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- Break down this equation
 - PA before? α

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 - PA before? α
 - PA after?

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

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

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 - PA before? α
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Regression Specification and Interpretation

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 - Can you see diff-in-diff?

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 - Can you see diff-in-diff?

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

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M: Estimation

Regression Specification and Interpretation

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 - 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

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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

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M: Set-up

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Recap: Key Parts

Key Assumption

- In the absence of treatment, treatment and control observations would have evolved in parallel fashion
- AKA, "parallel trends"
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- 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

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Setting up the Milligan et al paper

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Research Question and Estimation Problem

1. What is the research question?



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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.

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Research Question and Estimation Problem

- 1. What is the research question?
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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

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Data and Units of Observation and Analysis

What are the two data sources?

Vital statistics data

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Data and Units of Observation and Analysis

What are the two data sources?

Vital statistics data

• from birth records

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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

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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

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Data and Units of Observation and Analysis

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Data and Units of Observation and Analysis

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Canadian Census data

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Data and Units of Observation and Analysis

What are the two data sources?

Vital statistics data

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 - \rightarrow unit of observation is woman in a year
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 - $\bullet \ \rightarrow \ {\rm unit} \ {\rm of} \ {\rm analysis} \ {\rm is} \ {\rm province/year}$

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

M: Estimation

Basic Diff-in-diff

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

Potential Outcome

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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

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Potential Outcomes

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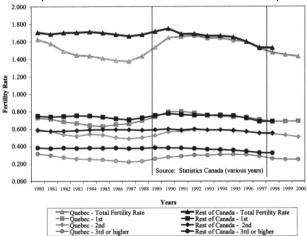
Estimation in Milligan

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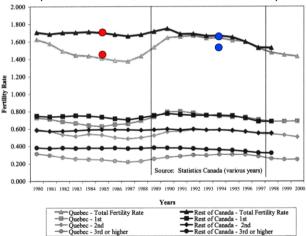
Basic Diff-in-diff in Figure 1

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		Difference in	
	1991 (1)	1996 (2)	in Means, (2) $-$ (1) $=$ (3)	Differences (4)	Percentage 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			

• How do you calculate 0.418?

M: Estimation

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- How do you calculate 0.418?
- And 0.441?

M: Estimation

The Diff-in-diff in Table 5

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- And 0.441?
- And Col. 3, 0.033?

M: Estimation

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

• How do we find 0.024?

M: Estimation

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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?
- And 0.441?
- And Col. 3, 0.033?

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

∕I: Set-up

M: Estimation

The Diff-in-diff in Table 5

Region	м	Mean		Difference in	
	1991 (1)	1996 (2)	in Means, (2) $-$ (1) $=$ (3)	Differences (4)	Percentage 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			

- 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?

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	1991 (1)	1996 (2)	in Means, (2) $-$ (1) $=$ (3)	Differences (4)	Percentage Increase (5)
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M: Estimation

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

up

M: Estimation

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

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

M: Estimation

What is the Underlying Assumption Here?

Can state the assumption many different ways

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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

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Can you test with Census data?

M: Estimation

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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

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 innnigrant Female Anglophone 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 Francophone Male Anglophone Male high school Male past-high school Male aniversity degree				
Married Lives in urban area Family income (CS10,000) Provincial GDP growth Provincial migration rate Provincial migration rate				

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

M: Estimation

And a Triple Difference!

	м	ean	Trend Difference in Means.	Difference in	Percentage	Triple
	1991	1996	(2) - (1) =	Differences	Increase	Difference
Region	(1)	(2)	(3)	(4)	(5)	(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			
•	(0.004)	(0.004)	(0.006)			
n	15,017	12,399				
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			
	(0.009)	(0.009)	(0.013)			
n	3,207	2,475				
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						
Quebec	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

M: Estimation

Next Lecture

Read

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