FLS 6441 - Methods III: Explanation and Causation

Week 8 - Difference-in-Differences

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Classification of Research Designs

		Independence of Treatment Assignment	Researcher Con- trols Treatment Assignment?
Controlled Experiments	Field Experiments	✓	√
	Survey and Lab Experiments	√	√
Natural Experiments	Natural Experiments	√	
	Instrumental Variables	√	
	Discontinuities	√	
Observational Studies	Difference-in-Differences		
	Controlling for Confounding		
	Matching		
	Comparative Cases and Process Tracing		

Section 1

What if we have NO variation in treatment that is independent of potential outcomes?

- ► What if we have *NO* variation in treatment that is independent of potential outcomes?
- ► Then we have an *Observational* study

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 - ► BUT Omitted variable bias
- 2. **Time-series:** Compare units **before** and **after** treatment
 - ► BUT Outcomes might change over time for reasons other than treatment ('Overall Trend Bias')

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 - ► Even *unobserved* characteristics

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- Comparing across units and across time
- Removing the risks from both overall trends and omitted variables

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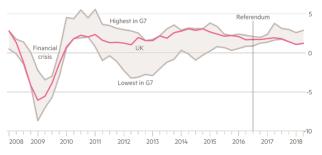
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Reversal of fortune: since the EU referendum, strong growth relative to other G7 economies has tailed off

Annual % change in GDP



Source: Thomson Reuters Datastream
© FT

▶ But can we really say this was the effect of Brexit?

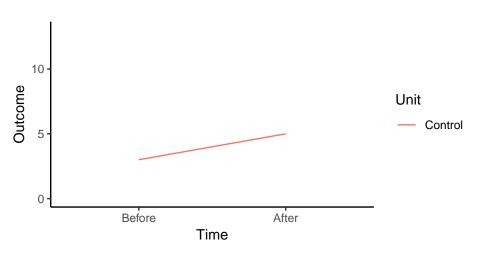
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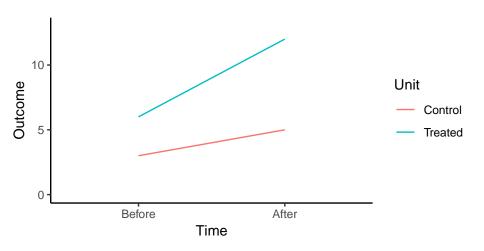
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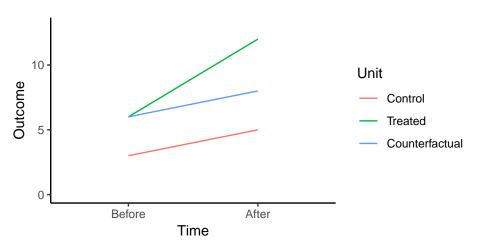
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 - We have to check there are no compound treatments

	Add caption	ption		
	Balances time-invariant 'fixed' unit characteris-tics		Balances time-varyi	
	Observed	Unobserved	Overall trends	Un
Field Experiments	\checkmark	\checkmark	\checkmark	
Survey and Lab Experiments	\checkmark	\checkmark	\checkmark	
Natural Experiments	\checkmark	\checkmark	\checkmark	
Instrumental Variables	\checkmark	\checkmark	\checkmark	
Regression Discontinuity	\checkmark	\checkmark	\checkmark	
Cross-sectional comparisons	Χ	X	\checkmark	
Before-After comparisons	\checkmark	\checkmark	Χ	
Difference-in-Differences	\checkmark	\checkmark	\checkmark	







Regression for the cross-unit effect of treatment

$$Y_{it} = \alpha + \gamma D_i$$

► Regression for the cross-unit effect of treatment

$$Y_{it} = \alpha + \gamma D_i$$

► Regression for the before-after treatment comparison

$$Y_{it} = \alpha + \gamma T_i$$

pause

► The difference-in-differences estimate is just the interaction of time and treatment status

$$Y_{it} = \alpha + \gamma D_i + \delta T_t + \beta D_i * T_t$$

ightharpoonup is our **Average Treatment Effect** estimate

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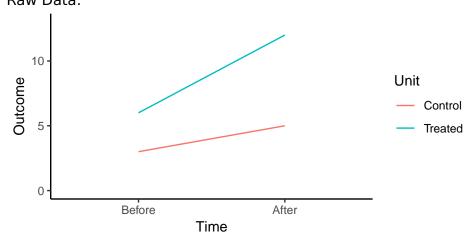
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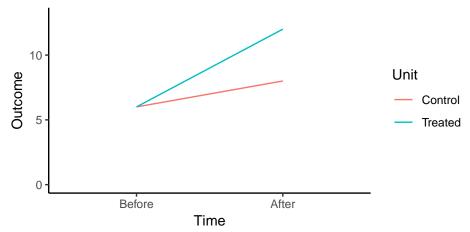
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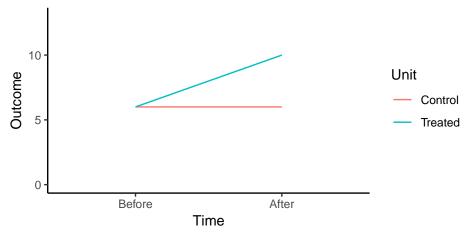




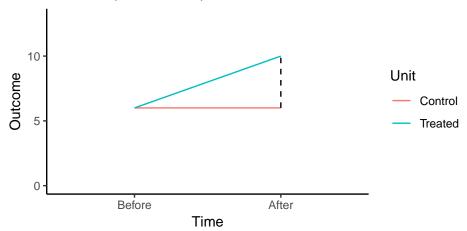
Add a variable (fixed effect) for treated/control:



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- So crucial to cluster standard errors by each cross-sectional unit (eg. each country)

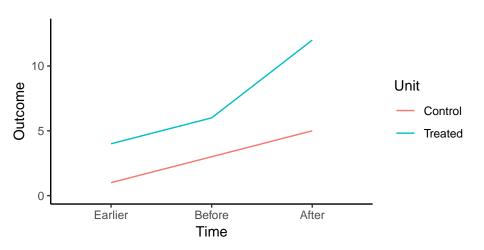
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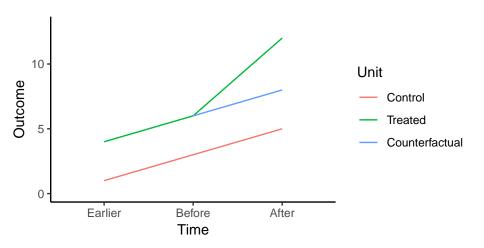
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- One test of this is to check if pre-treatment trends are parallel





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- 4. Group membership is stable (no migration from control to treatment)

Section 2

Chimeli and Soares 2017

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- How does an activity being illegal affect violence?
- How did Brazil's ban on mahogany affect homicides?
- What are the challenges to explanation?
 - Omitted variables, eg. State capacity
 - Overall trends, eg. national decrease in homicides
- Comparing the change in violence in mahogany-growing areas to the change in violence in non-growing areas

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- ► **Before:** Pre-1998
- ► After: Post-1998
- Outcome: Rate of Homicides

- Multiple treatment timings:
 - ► 1st policy change
 - ► 2nd policy change
 - ► Reverse treatment: Better policing of mahogany regulations

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$$Homicides_{it} = \gamma_t + \delta_i + \beta(Post - 1998_t * Mahogany_i) + \epsilon_i$$

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- Cluster standard errors by municipality
- Apply more complex state-specific trends for covariates to minimize risk of non-parallel trends

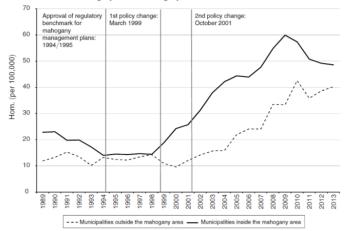
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- Cluster standard errors by municipality
- Apply more complex state-specific trends for covariates to minimize risk of non-parallel trends
- Supporting evidence: The 'extra' homicides were the type we'd expect from illegal activity

Difference-in-Differences

Panel A. Homicides in mahogany and non-mahogany areas



- Interpretation
 - ► Illegal activity prevents 'peaceful' contract enforcement
 - Competition between loggers
 - Contract enforcement with buyers
 - ► Intimidation of communities to not report logging