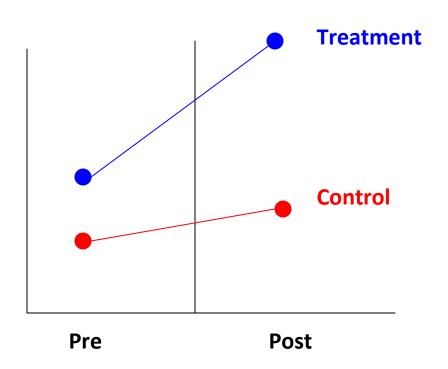
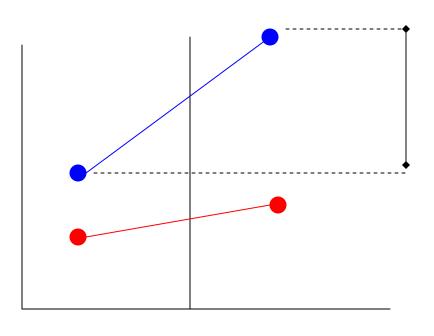
## Quasi experimental methods: Difference in differences

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University of California, San Diego
March 24th, 2010

## Quick re-cap

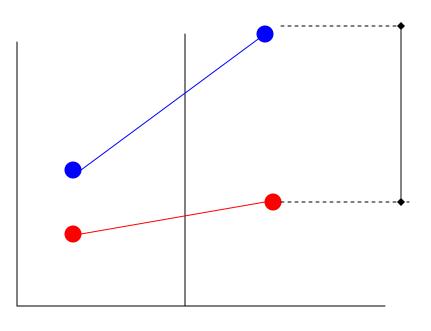
- While using non experimental data to infer causal relationships, we must think through sample selection and omitted variables bias
- 2. Comparing just pre-post or participant vs nonparticipant is not enough
- 3. This lecture is about *differencing* out the potential omitted variables bias





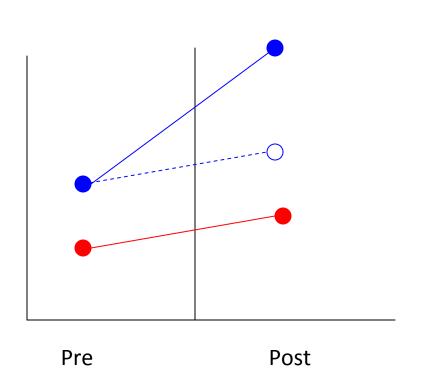
Effect of program using only pre- & post- data from T group (ignoring general time trend).

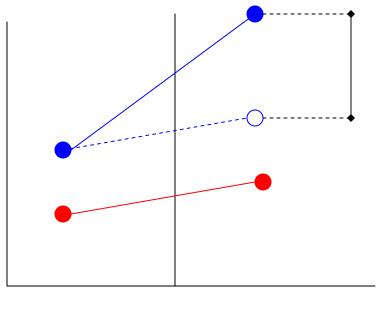
Pre Post



Effect of program using only T & C comparison from post-intervention (ignoring pre-existing differences between T & C groups).

Pre Post



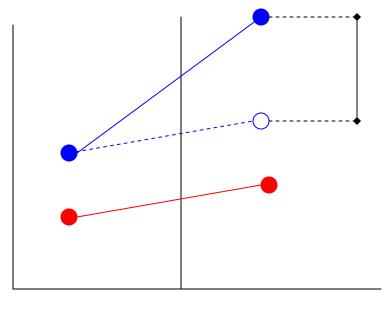


Effect of program
difference-in-difference
(taking into account preexisting differences
between T & C and general
time trend).

Pre Post

## **Identifying Assumption**

 Whatever happened to the control group over time is what would have happened to the treatment group in the absence of the program.



Effect of program difference-in-difference (taking into account pre-existing differences between T & C and general time trend).

Pre Post

## DD: Simple Example

	Pre	Post	Difference
Treatment	100	170	70
Control	65	100	35
Difference in Difference			35

## Example: Change in marriage laws

- 1. In 1957 Mississippi amended its marriage law
  - 1. Raised minimum age for men and women
  - 2. Introduced parental consent laws
  - 3. Proof of age, blood tests, other restrictions

## Impact of changes in marriage law

- 1. How can we figure out the impact of this marriage law on outcomes such as
  - 1. Marriages
  - 2. Fertility
  - 3. Education
- 2. Relevant for our context in terms of empowering women does postponement of marriage empower women by delaying fertility and increasing education?

## What happened after 1957?

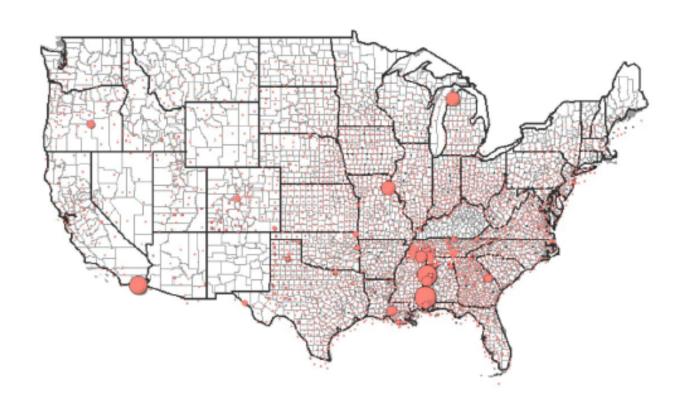


#### DD in this context

- 1. Marriage law affected Mississippi residents and immediate neighbors (*treated* group)
- 2. But did other changes happen during this time that would affect marriage rates, fertility and education?
  - 1. Changes in other laws compulsory schooling etc
  - 2. Changes in work opportunities for women
  - 3. Migration
  - 4. Pre 1957 trends in marriage?
- 3. Nearby states, counties are a potential control group

## Pre and post marriage law

Figure 4 - Change in Marriages 1954-1960



### Appropriate control groups

- 1. Counties next to Mississippi
- 2. Age groups not affected by the law
- 3. Triple difference: DD estimates for younger vs older age groups
  - 1. In the event of treatment specific trends affecting all age groups

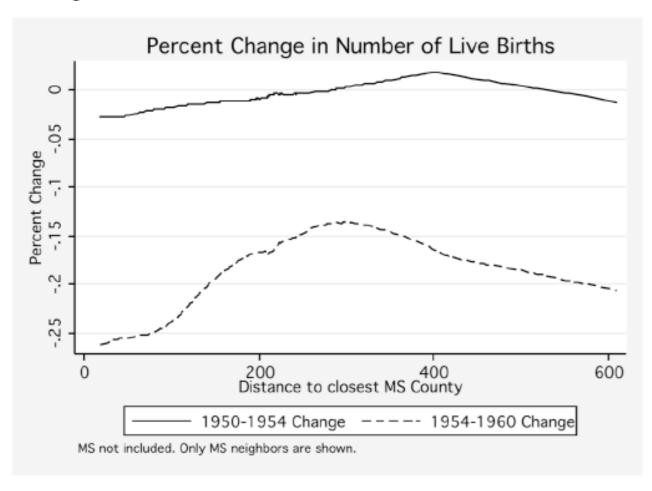
#### Trends in treatment and control

Figure 2



## Distance to Mississippi

Figure 6



# Did other things change differentially

Table 6a: Checking for differential trends

	Manufacturing wage	Manufacturing Employment	% of Farms with tractors	Farms per 1000 in population
Treatment X Post	-0.131	1.017	0.528	-0.013
	[0.683]	[2.323]	[0.852]	[0.002]***
Treatment (1 if MS or border county, 0 ow)	1.101	9.582	13.864	-0.038
	[0.302]***	[1.658]***	[0.577]***	[0.001]***
Post (=1 if year ≥ 1957)	1.476	-9.029	0.468	0.015
	[3.785]	[5.650]	[2.347]	[0.008]*
R-squared	0.03	0.06	0.26	0.26
Observations	1149	1532	766	1532

<sup>\*</sup> significant at 10%; \*\* significant at 5%; \*\*\* significant at 1%, robust std errors, clustered at the county level

## Placebo assignment of law

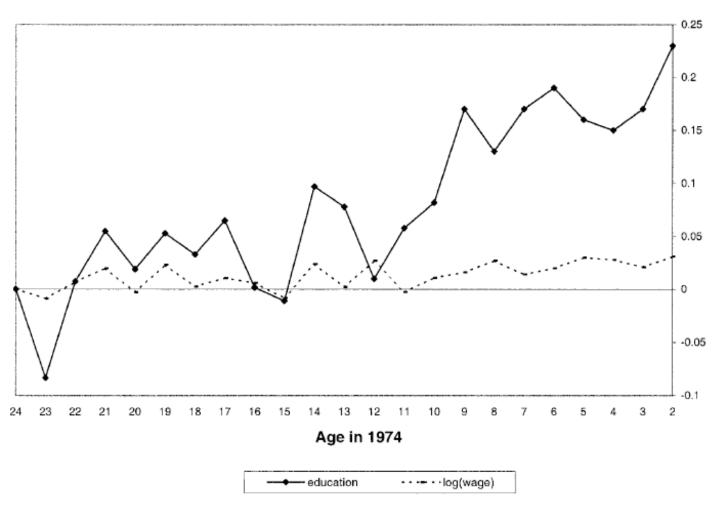
Table 6b: Random Assignment of Treatment Group

Random assignment of treatment group and 1000 repetitions	Marriages	Crude Birth Rate	School Enrollment
Average coefficient size under random assignment	-0.016	-0.048	0.033
True coefficient size	-13.389	-1.869	2.547

## Impact of marriage law

- 1. Marriage decline
- 2. Immediate rates of fertility decline
  - 1. Unintended effect: Some evidence of increases in illegitimate childbearing
  - 2. No effect on completed fertility
- 3. Enrollment in school increases
  - 1. Overall high school completion rates higher

## Indonesia Schooling (Duflo)



#### Conclusions

Difference in differences is a good non-experimental method for impact evaluation if:

- 1. Randomization is difficult or expensive
- 2. Control group is well defined
- 3. Assumption that trends in treatment and control in absence of treatment would have been the same is true