# FLS 6441 - Methods III: Explanation and Causation

Week 11 - Comparative Case Studies & Process Tracing

Jonathan Phillips

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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

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- We need counterfactuals to estimate treatment effects: Comparative Cases

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- ► Common error: "research that tries to explain the outbreak of war with studies only of wars" (KKV)

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  - ► Matching: More useful

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- Our Large-N dataset after matching might look very similar to comparative case studies

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  - 4. Slow economic growth

#### Does Development cause Democracy?

	Variable	Case A	Case B
Outcome	Democracy	?	?
Treatment	Development	Low	High
Controls	Religion	Christian	Christian
	Continent	Europe	Europe
	Inequality	0.45	0.44
	Economic growth	1.2%	5%
	National dish	Pasta	Corn
	Length of Railways	400km	120km

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  - Statistical Inference: Non-random case-selection, so generalization is harder

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  - Our goal is not to explain why outcome Y happened in one case, but why it happens generally

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    - Randomization does not guarantee balance on confounders in small samples
    - Randomized sampling is not the same as randomized treatment
  - Probably easier to 'block' on key confounders and impose variation in treatment - purposive sampling

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  - ▶ But: If we select cases explicitly for a range of values of the outcome, that's better

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  - ► At the same time as balancing confounders hard!

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  - ► Extreme cases: Highest and lowest values of treatment, eg. Lieberman

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  - ► Influential cases: How sensitive is our relationship to mismeasurement of a key case?

## Section 2

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  - 2. Comparative cases to identify explanation, then **tested for generalizability** in Large-N sample (Lieberman)
  - Large-N analysis to show causal effect within-case, then generalized using comparative case studies (Ziblatt and Slater)

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  - ► Is the evidence consistent with theory A? Or inconsistent with theory B?

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- 5. Can we eliminate all other theories except our treatment?
  - ► Sherlock Holmes' Method of Elimination

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- ► The outcome could instead have been caused by a confounder

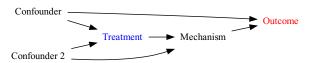


- ➤ One way to support our theory is to test the mechanisms along the causal path of treatment:
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- ► If there are no other possible confounders consistent with this mechanism, this is a 'Smoking Gun' test



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- ► This is a 'straw in the wind' test



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- ► This is a 'Doubly-Decisive' test



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  - Elites choosing the time and form of democratization

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- We're substituting assumptions/theory for a counterfactual
  - We 'assume' that the only way our treatment could work is through the mechanism we specify
  - ► And we assume the only way confounding works is through the mechanism we specify
- So everything depends on how confident we are in our theory/assumptions about mechanisms

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  - Imperfect measurement and data availability

- ► In practice, process tracing is made harder by:
  - ► Imprecise, or non-discriminating theory
  - ► Imperfect measurement and data availability
  - ► Subjective judgment on the weight of each piece of evidence

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- ► But how representative is our case?
- ▶ Will the same causal effect occur in other contexts?

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  - ► How long was left for the election after treatment?: 10 minutes
  - ► How many voters were **potentially influenced**:

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  - ► How many voters were **probably treated**:

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  - ► How long was left for the election after treatment?: 10 minutes
  - ► How many voters were **potentially influenced**: 4,200 voters
  - ► How many voters were **probably treated**: 560 voters

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  - ► How long was left for the election after treatment?: 10 minutes
  - ► How many voters were **potentially influenced**: 4,200 voters
  - ► How many voters were **probably treated**: 560 voters
  - ► How many voters likely complied with treatment:

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- ► The only way the causal effect is true is if there is a causal mechanism connecting the treatment to the outcome:
  - ► How long was left for the election after treatment?: 10 minutes
  - ► How many voters were **potentially influenced**: 4,200 voters
  - ► How many voters were **probably treated**: 560 voters
  - ► How many voters likely complied with treatment: 56 voters

- ► Brady (2010)
- ▶ Difference-in-differences evidence that the early announcement of a Democrat victory in Florida led to reduced Republican voting
- ► Estimated 10,000 lost Republican votes
- ► The only way the causal effect is true is if there is a causal mechanism connecting the treatment to the outcome:
  - ► How long was left for the election after treatment?: 10 minutes
  - ► How many voters were **potentially influenced**: 4,200 voters
  - ► How many voters were **probably treated**: 560 voters
  - ► How many voters **likely complied with treatment**: 56 voters < 10,000