# FLS 6415 - Causal Inference for the Political Economy of Development

# Week 6 - Social Accountability, Information & Instrumental Variables

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- What can we do when the treatment assignment mechanism is not 'as-if' random?
- Natural experiments focus on a specific part of treatment assignment that is 'as-if' random
- An 'instrument' is a variable which assigns treatment in an 'as-if' random way
  - Or at least in a way which is 'exogenous' not related to confounders
  - Even if other confounding variables **also** affect treatment

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- We can use the instrument to isolate 'as-if' random variation in treatment, and use that to estimate the effect of treatment on the outcome
- NOT the effect of the instrument on the outcome

- ► Example Instruments:
  - Rainfall for conflict
  - Sex-composition for effect of third child
  - Distance from the coast for exposure to slave trade

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  - Strong First Stage: The Instrument must affect the treatment
  - We can test this with a simple regression: Treatment ~ Instrument
  - The instrument should be a significant predictor of treatment
  - Rule-of-thumb: F statistic > 10

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  - We cannot test or prove this assumption!
  - Theory and qualitative evidence needed to argue that the instrument is not correlated with any other factors affecting the outcome
  - Sometimes, the exclusion restriction may be more credible if we include controls

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    - Interpret the coefficient on  $\hat{D}$

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- We call our causal effect estimate a 'Local Average Treatment Effect' (LATE)
- 'Local' to the units whose treatment status actually changed
- Remember, those 'Local' units are not representative so we can't generalize

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  - 5. **Reduced-Form Regression:** Estimate of the Instrument on the Outcome, ignoring treatment mediation: *Y* ~ *IV*

Instruments for Non-compliance

- Instruments for Non-compliance
  - With an instrument and treatment we can divide our units into four types:

Treatment Sta- tus if Instru- ment=0	Treatment Sta- tus if Instru- ment=1	Unit Type
0	1	Complier
0	0	Never-taker
1	1	Always-taker
1	0	Defier

- LATE just means we don't learn anything about Never-takers and Always-takers from Instrumental Variables
  - Because the instrument doesn't do anything to affect treatment for these units
- We also need to assume Defiers don't exist
- ► So LATE = Causal Effect for Compliers

# Instruments for Non-compliance in Experiments
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  - Normally we analyze experiments based on randomized treatment
  - But what if **assignment** to treatment is different from **taking** the treatment?
    - ► Eg. If government implementation failed in some places

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- Instruments for Non-compliance in Experiments
  - We can still use randomization as an instrument for treatment
  - The causal effect estimate of our experiment is now LATE
    - These estimates are internally valid for compliers
    - But they are NOT externally valid for non-compliers
    - Since whether you accepted treatment is probably confounded/subject to self-selection
  - We can also estimate the Intention-to-Treat effect, the effect of the instrument itself
    - But this will be **conservative**, i.e. less than the LATE estimate

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  - Where the instrument is an arbitrary rule, there is often sorting as people re-adjust

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  - Protests, lobbying
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  - The short-route of accountability: Client power in demanding public service improvements
- ► Information & Media also influence electoral accountability

## Social Accountability



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  - ▶ 1995: Only 24% of grants to schools arrive
  - 2002: 82% of grants to schools arrive
- This wasn't elite corruption, but diversions within the bureaucracy (centre -> district -> school)
- What changed? A Government newspaper campaign to publicize grants

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- What is the challenge to inference here?
- Information is not randomly assigned; eg. checks and balances on the bureaucracy may also be stronger in places where headteachers have more information

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- Treatment Assignment Mechanism: Messy! Influenced by confounders and instrument

Instrumental Variables Assumptions:

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  - First-Stage: Distance of school to newspaper seller -> -Headteacher knowledge of grant amount/timing
    - Verifiable
  - Exclusion Restriction: Distance to newspaper seller ONLY affects grant access and learning through information, not directly
    - Unverifiable
    - But more likely when we include controls for distance to nearest bank, district headquarters etc.
- They actually combine this with a difference-in-differences method to look at *changes* in information and grant receipt over time.

- Methodology:
  - Information<sub>i</sub> =  $\alpha + \beta_0 Distance_{to} Newspaper_i + \epsilon_i$
  - Grant\_Received<sub>i</sub> =  $\alpha + \beta_1 Information_i + \epsilon_i$
- Methodology:
  - Information<sub>i</sub> =  $\alpha + \beta_0 Distance_{to_Newspaper_i} + \epsilon_i$
  - Grant\_Received<sub>i</sub> =  $\alpha + \beta_1 Information_i + \epsilon_i$
  - Alternative:
  - Grant\_Received<sub>i</sub> =  $\alpha + \beta_0$ Distance\_to\_Newspaper<sub>i</sub> +  $\epsilon_i$
  - Enrolment =  $\alpha + \beta_1 Grant \hat{Received}_i + \epsilon_i$

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- ► A one standard deviation increase in information leads to
  - 44.2% points more funding received
  - 297 students per school
  - ▶ 6% better in exams

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  - Enrollment doubled in 1997 when school became free
  - ► WB support conditional on better systems, transparency
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  - Grants were also displayed on 90% of school notice-boards
- Where did these headteachers gain the political power to demand their grants?

# Enikolopov et al (2011)

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- Does independent media encourage voting for the opposition?
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- What is the inference problem?
- People who watch NTV might be more anti-government in the first place
- Or NTV may choose to broadcast in anti-government areas

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- **Outcome:** Vote-share for each government/opposition party

- Instrumental Variables Assumptions:
  - First Stage: Availability of signal clearly correlated with watching NTV
  - Exclusion Restriction: Availability of the signal only affects voting through watching NTV

- ► Exclusion Restriction Supporting Evidence:
  - History: The transmitters were located for a Soviet education channel, not chosen by the opposition
  - Controls: Transmitters are correlated with socioeconomic characteristics, but we can control for this (urban, population, wage)
  - Placebo: If the instrument only operates through treatment, it should have no effect when treatment is impossible, eg. in 1995

 Estimate signal availability using Irregular Terrain Model and transmitter power/frequency

#### Social Accountability



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# Aggregate Level Data (effect of NTV availability):

- Predicted\_NTV\_available<sub>i</sub> =  $\alpha + \beta_0 + Signal_Strength_i + \epsilon_i$
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# Individual Level Data (effect of watching NTV):

- $Watch_NTV_i = \alpha + \beta_0 Predicted_NTV_Available_i + \epsilon_i$
- $vote_i = \alpha + \beta_1 Watc\hat{h}_N TV_i + \beta_2 X_i + Region_FEs + \epsilon_i$

Results:

- Results:
  - NTV broadcast availability reduces pro-government 'Unity' voting by 8.9% points (official results)
  - NTV broadcast availability reduces turnout by 3.8% points (official results)
  - Watching NTV broadcast reduces pro-government 'Unity' voting by 26% (survey results)

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- ► Institutions depend on powerful elites, esp. colonial settlers
- Extractive vs. Settler Institutions
- Colonial Strategy -> Institutions -> Growth
- What is the inferential problem here?

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- Treatment: Settler institutions (measured by 'risk of expropriation' index 1985-95)
- Control: Extractive institutions
- ► Instrument: Settler (soldier...) mortality rates
- Treatment Assignment Mechniams: Messy! Confounders plus Instrument
- Outcome: Growth rates in 1995

- Instrumental Variables Assumptions:
  - First Stage: Settler Mortality explains Current Institutions
  - Exclusion Restriction: Settler Mortality only affects growth through institutions

- Exclusion Restriction Supporting Evidence:
  - Disease environment doesn't affect human capital/growth directly because locals have adapted

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  - Disease environment doesn't affect human capital/growth directly because locals have adapted
  - Control for possible correlates geography, climate,

- Methodology:
  - Institutions<sub>i</sub> =  $\alpha + \beta_0$ Settler\_Mortality<sub>i</sub> +  $\epsilon_i$
  - Growth<sub>i</sub> =  $\alpha + \beta_1$ Institutions<sub>i</sub> +  $\epsilon_i$

 Results: Improving Nigeria's institutions to Chile's level would raise GDP 7-fold

- 'Social' Accountability can dramatically affect public services, voting behaviour and growth
  - Client Power to demand more from government
  - Exposure to information/Media
  - Checks and Balances on expropriation