

A SLIPPERY SLOPE:

TOPOGRAPHIC VARIATION AS AN INSTRUMENTAL VARIABLE

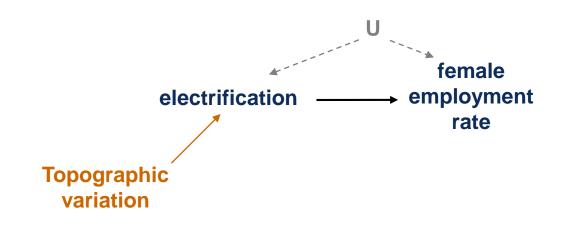
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Leibniz Open Science Day, Berlin, 25 November 2024

Motivation

- > Topographic variation: measures of elevation changes, land slope, ruggedness...
 - > Arguably exogenous to any social behavior and political decision making
- Popular Instrumental Variable (IV) for impact evaluations of hard-to-evaluate policy interventions, e.g energy-infrastructure (Dinkelman 2011, Duflo & Pande 2007, Mian & Sufi 2011)

Running example: Dinkelman's electrification effects on employment in rural South Africa (2011, AER)



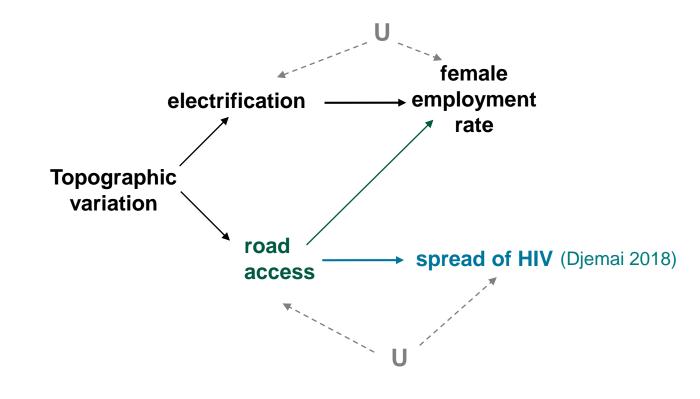
Idea: Exploit topographic variation as a natural experiment

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Motivation

The Exclusion Restriction



In this setup, instrumenting for electrification and road access is mutually exclusive.

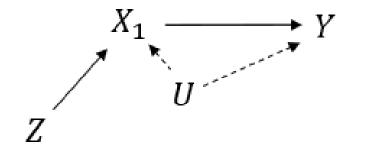
→ Repeated use of the same instrument in different contexts likely renders the instrument invalid (Gallen 2023, Mellon 2024)

\rightarrow How problematic is the repeated use of similar IVs in general?

- > Theoretical framework
 - 1. Exclusion Restriction violations
 - 2. Post-instrument bias



1. Recap: Basic IV model using the DAG framework (Pearl 2009)



For Dinkelman (2011): X = electrification Y = female employment rate Z = land gradient U = unobservables

Identifying assumptions:

- 1. IV-relevance: $Cov(X_1, Z) \neq 0$
- 2. IV-validity: **Cov(U, Z | X_{exo}) = 0**
- › IV's Unconfoundedness
- > Exclusion Restriction

→ There should be no effects $\underline{unmediated}$ through the treatment X1 !

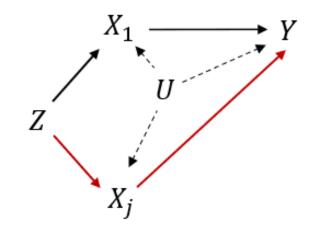
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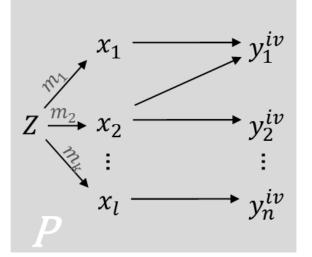
1. Exclusion Restriction violations

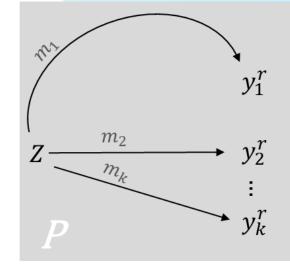
a) Exclusion restriction violation by Xj



c) Multiple non-IV studies on reduced-form effects



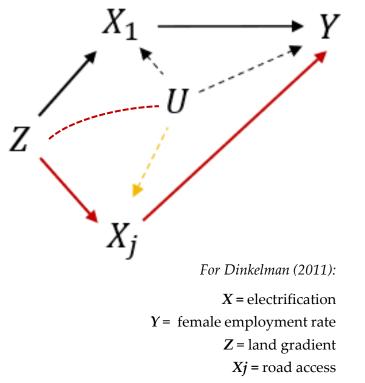






2. Post-instrument bias

(a) X_j is a collider



U = business potentials

Conventional practice: Block exclusion restriction
 variables by conditioning on them (Felton & Steward 2022)

Problem:

- Controlling for Xj introduces "post-instrument bias" (Deuchert & Huber, 2017; Glynn et al., 2023)
- \rightarrow similar to a "bad-control" problem in OLS

> Resembles a catch-22:

Omitting X_j is inappropriate, but including X_j as covariates violates the IV's exogeneity requirement

Do we see a repeated use of topographic IVs?

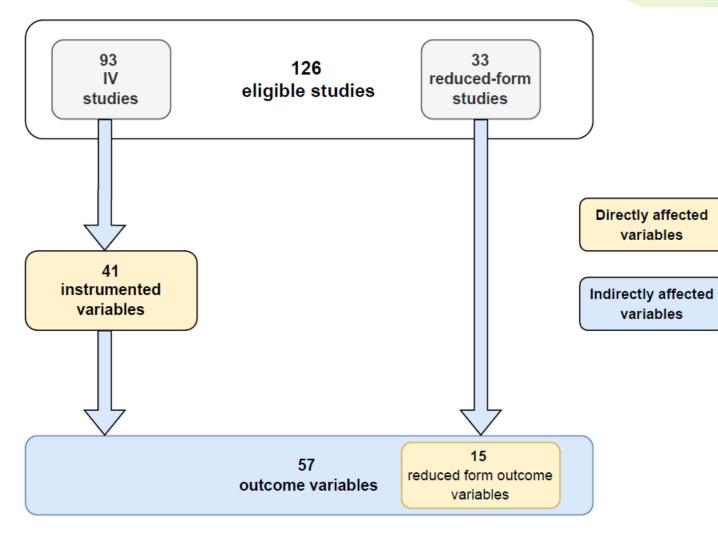
- > Systematic literature review
 - 1. The causal web of Topographic variation
 - 2. Prevalence of post-instrument bias
 - 3. Co-occurrence with weak IVs

Eligibility Criteria

- IV studies using topographic instruments (elevation changes, terrain slope, ruggedness...)
 → corr = 0.6 0.95 (Amatulli et al., 2018)
- 2. "Reduced-form" articles that assign a direct effect of topography on outcomes
- 3. Published between **2012 and 2021**
- 4. Published in Top120 Economics journals

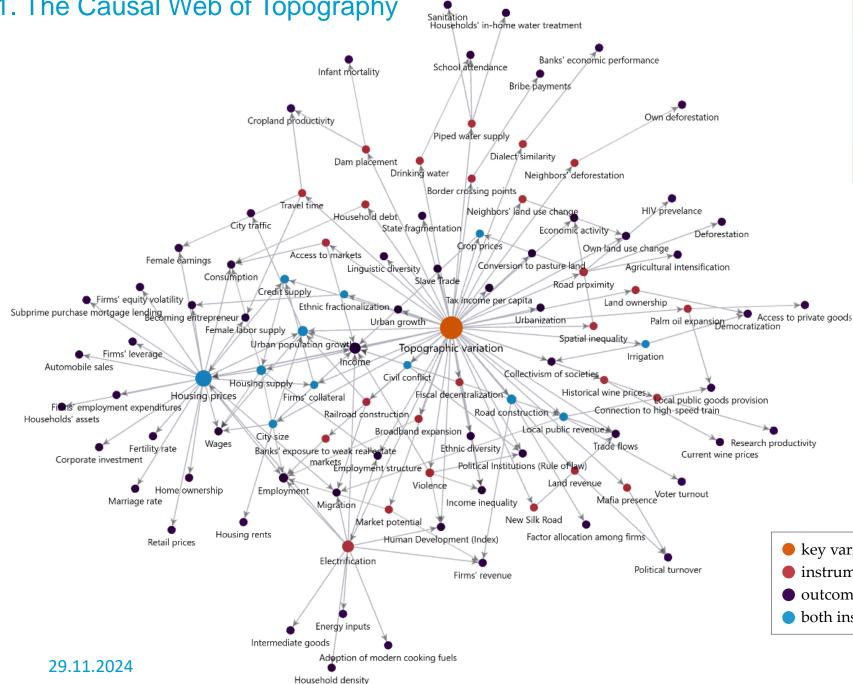


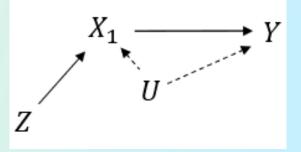
1. Results



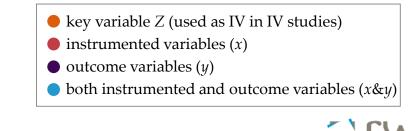


1. The Causal Web of Topography

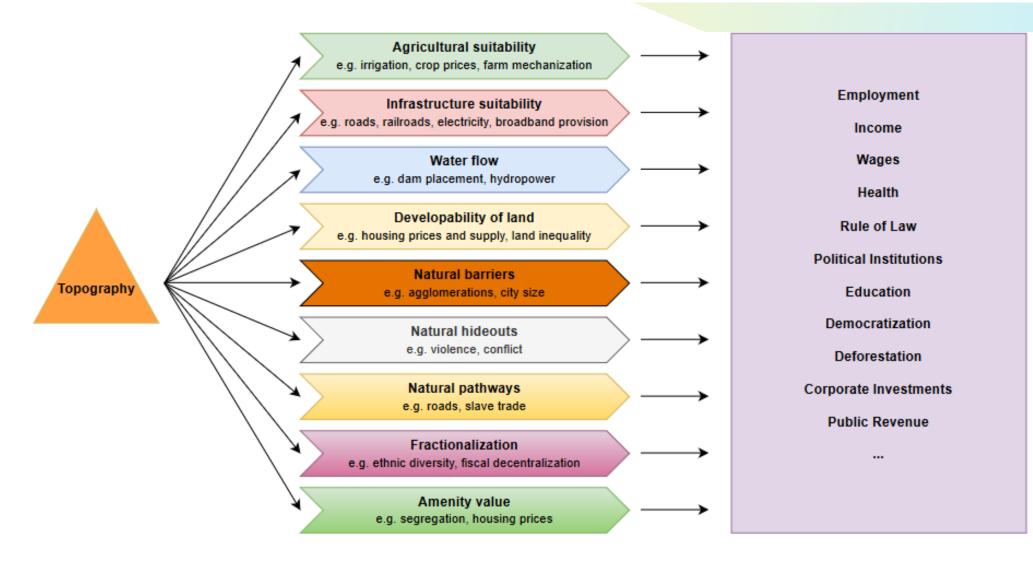




111 different pathways from topography to different outcomes

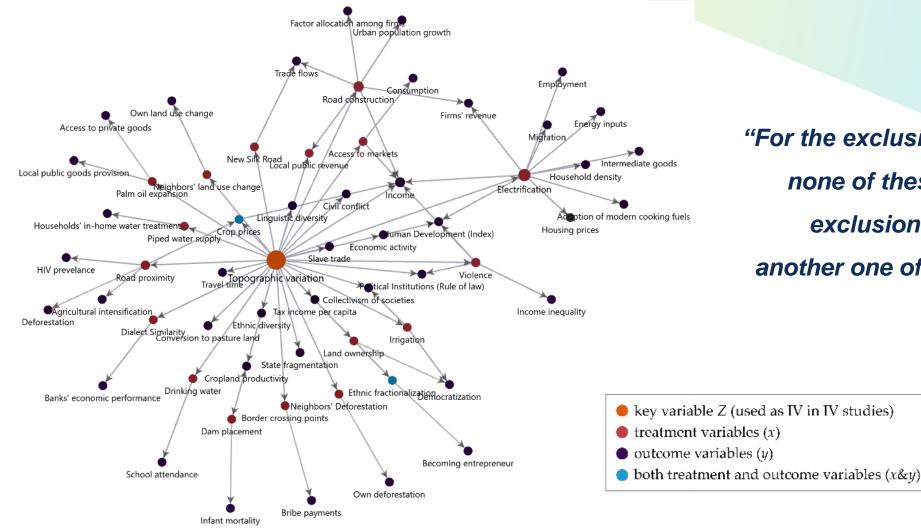


Main pathways from topography, via mechanisms and instrumented variables to outcomes





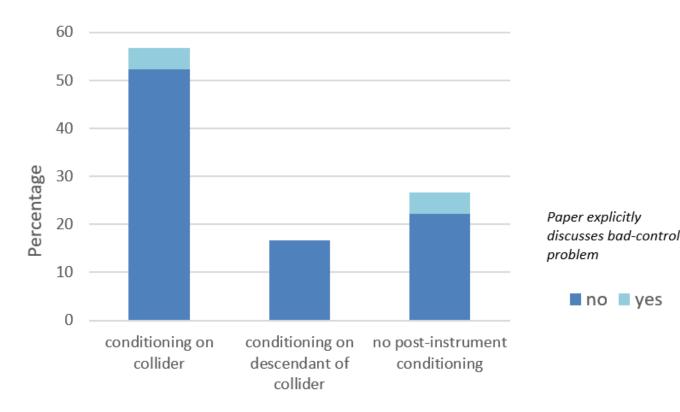
1. Causal Web - Robustness Check: Topographic IV papers from rural areas in low-income countries



"For the exclusion restriction to hold, none of these variables can be an exclusion-restriction variable, if another one of them is picked as the treatment variable"

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2. Post-instrument bias

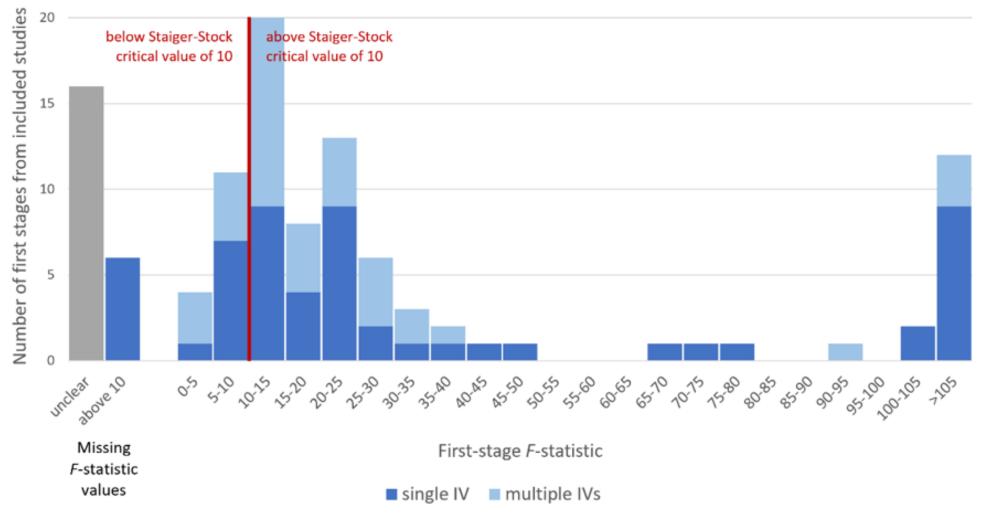


Almost 75 % of our IV studies condition on what are likely **bad controls**

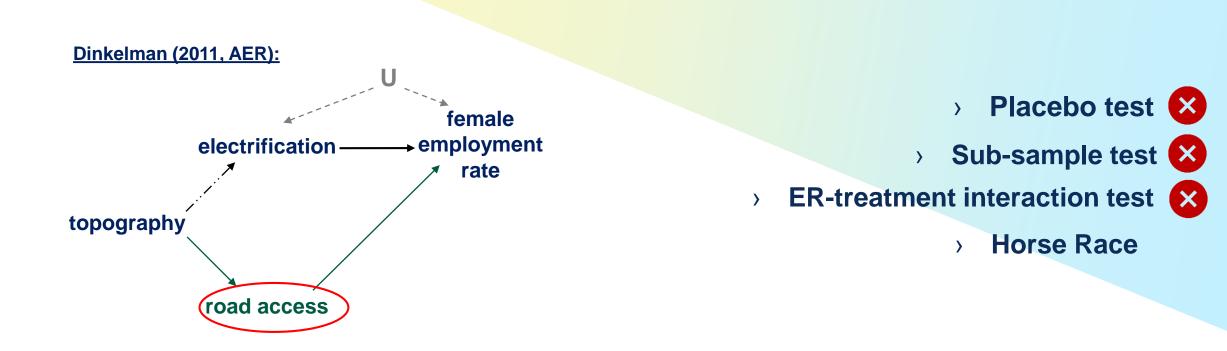
Note: The graph shows information on post-instrument conditioning for 90 of the 93 IV studies, for which the necessary information on control variables is provided in the articles. We define problematic controls as all variables from our causal web. Colliders are directly, descendants of colliders indirectly linked to topographic variation.



4. Concomittance with IV weakness







How to test the severity for a specific case?

Application to Dinkelman (2011)



APPLICATION TO DINKELMAN (2011)

Horse race

 comparing the original treatment variable (electrification) with the alternative (road access) in the spirit of a horse race – each using the other one as a control

Caution:

 Same post-instrument bias problem & IV weakness

IV results for road access as alternative treatment variable

Dependent variable	∆ female employment rate					
Estimation method	IV (gradient)					
Treatment definition	Access to roads within		Electrification			
	5 km distance		(Dinkelman (2011)			
	(1)	(2)	(3)	(4)		
Treatment	0.104*	0.106*	0.090*	0.095*		
	(0.059)	(0.058)	(0.054)	(0.055)		
	[0.080]	[0.069]	[0.097]	[0.083]		
AR 90% Confidence Interval	{0.03; 0.29}	{0.03; 0.28}	{0.02; 0.25}	{0.02; 0.26}		
Baseline controls	Yes	Yes	Yes	Yes		
District FE	Yes	Yes	Yes	Yes		
Controls for other services*	No	Yes	No	Yes		
First-stage results						
Gradient x 10 coefficient	-0.068***	-0.071***	-0.072***	-0.074***		
	(0.026)	(0.025)	(0.024)	(0.024)		
R-Squared	0.15	0.16	0.19	0.20		
Partial F-statistic	7.11	7.77	8.34	8.26		
N communities	1,816	1,816	1,816	1,816		

Notes: All model specifications are equal to columns 5 to 8 of table 4 from the original study with the only difference that the instrumented variable is access to roads instead of electrification and that we excluded binary road access from the control variables while adding electrification status; robust standard errors clustered at village level in parentheses and *p*-values in square brackets.

* significant at the 10% level

CONCLUSION

- We identify 126 studies that deliver 56 different potential exclusion-restriction variables linked to topographic variation
 - ...encompassing economically powerful variables such as infrastructure placement, population trajectories, agriculture, and housing prices
- "Controlling-away" exclusion restriction variables is typically necessary but at the same time invalid
 - Findings suggest that strict exogeneity assumptions might not hold in most topographical IV studies of our sample
- > Yet, no "strict proof" that topographical IVs should be excluded from the literature
 - > Findings do not disprove of any particular empirical claims from topographic-IV papers
- Rather, the findings suggest higher standard in transparently discussing, systematically testing, and ultimately choosing to use topographic variation as an IV
 - > Our paper as contribution to provide guidance in this process



REFERENCES

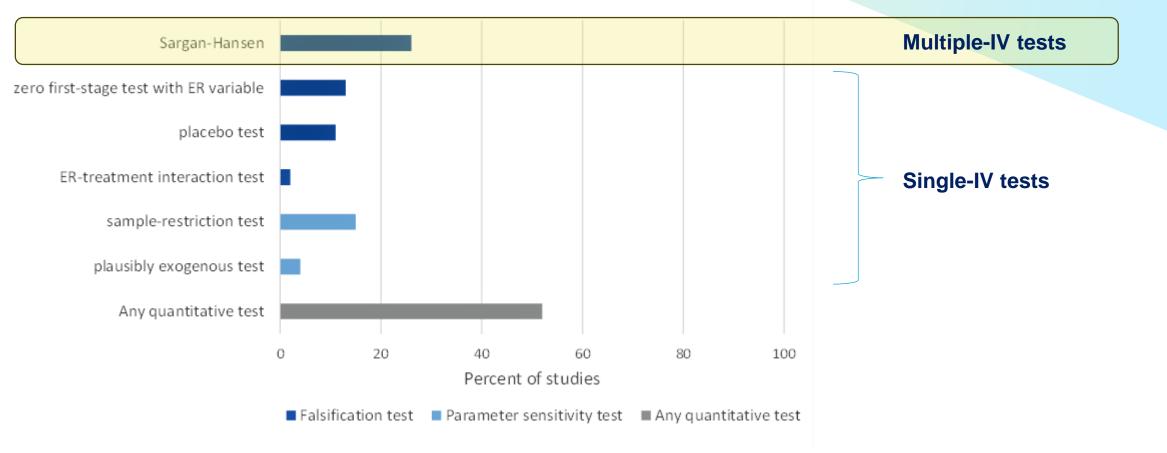
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SUPPLEMENTARY MATERIAL



3. Use of quantitative tests to validate the exclusion restriction



Quantitative exclusion-restriction testing



How to test the severity for a specific case?

Application to Dinkelman (2011)



APPLICATION TO DINKELMAN (2011)

Placebo experiment for effect of land gradient on the female employment rate

- > "zero-first-stage" test
- direct effect of the IV on the outcome in a population where effect is expected to be absent

Dependent variable	∆ female employment rate						
Estimation method	OLS						
Sample	Areas electrified before 1996			Non-electrified communities (no Eskom project)			
	(1)	(2)		(3)	(4)		
Land gradient x 10	-0.001 (0.001)	-0.001 (0.001)		-0.007* (0.0004)	-0.007** (0.0004)		
	[0.36]	[0.43]		[0.07]	[0.05]		
Baseline controls	Yes	Yes		Yes	Yes		
District FE	Yes	Yes		Yes	Yes		
Controls for other services*	No	Yes		No	Yes		
R²	0.068	0.106		0.065	0.081		
N communities	373	373		1,451	1,451		

Reduced form regression of female employment rate on land gradient in placebo areas



WHAT WE DO IN THIS PAPER (EXCERPT)

→ How problematic is the repeated use of the same IV in general?

- > Theoretical framework
 - > Exclusion Restriction violations
 - > Post-instrument bias

\rightarrow Do we see such repeated use for topographic IVs?

- > Systematic literature review
 - > The causal web of Topography
 - > Prevalence of post-instrument bias
 - > How do authors defend the exclusion restriction?

\rightarrow How to test the severity for a specific case?

- > Application to Dinkelman (2011)
 - > Bias analysis

WHAT WE DO IN THIS PAPER

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 - 3. Concomitance with weak IVs

\rightarrow Do we see such repeated use for topographic IVs?

- > Systematic literature review
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 - 2. Prevalence of post-instrument bias in our sample
 - 3. How do authors defend the exclusion restriction?
 - 4. Prevalence of weak IVs in our sample

\rightarrow How to test the severity for a specific case?

- > Application to Dinkelman (2011)
 - > Bias analysis

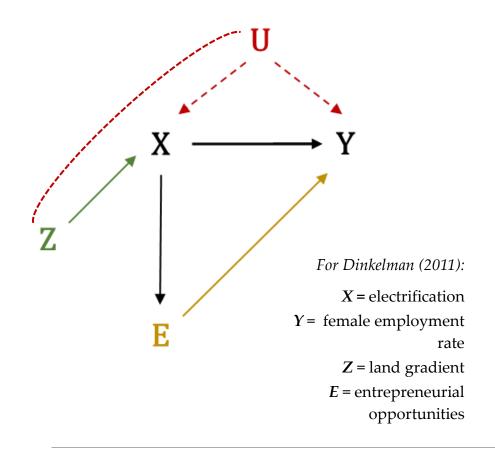
RELATED LITERATURE

Mellon's review (2023) of 289 social science studies reveal 195
 variables linked to weather

- Gallen & Raymond (2023) find six groups of commonly-used instruments (e.g. rainfall, sibling structure) suggesting likely exclusion restriction violations
- Deuchert & Huber (2017) and Glynn et al. (2023) point towards bad control problems in IV analysis
- Felton & Steward (2022) highlight bias sensitivity to even minor
 violations of the exclusion restriction in a weak-instrument setting

Post-instrument bias

b) *E* is a descendant of a collider



 Conditioning on descendants of a collider imports similar problems of unconfoundedness (Glynn et al., 2023)

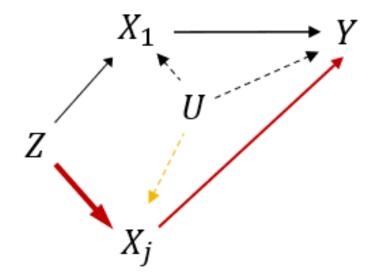
Solution: not control for E! (Cinelli et al., 2022; Glynn et al., 2023)

 Yet, it might be plausible that many (topographic)
 IV studies condition on colliders and their descendants...



Exclusion restriction violation & concomitance with IV weakness

>



- Weak instruments excaberate biases from Exclusion Restriction violations (Felton & Steward 2022)
- → "Identification bias" (Felton & Steward 2022)

 Note: Weak-IV robust AR confidence intervals do not provide a panacea here (Andrews et al. 2019)

For Dinkelman (2011):

X = electrification Y = female employment rate Z = land gradient D = road access V = business potentials

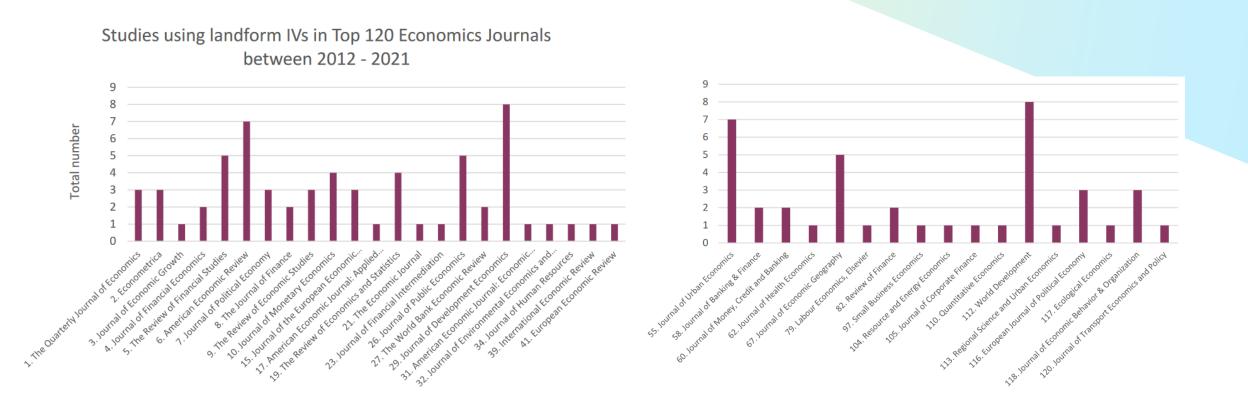
DESCRIPTIVES ON STUDIES OF SYSTEMATIC REVIEW

	ıv	reduced form		IV		IV	reduced form
Year of publication, %			Purpose of IV, %		Mechanisms mentioned,		
2012-2016	43	48	main specification	73	#		
2017-2021	57	52	robustness check	27	water flow	7	0
					agricultural suitability	9	11
Journal ranking, %			Number of IVs, %		infrastructure suitability	25	4
among top-10 journals	32	24	single IV	64	developability of land	46	10
			multiple IVs	36	natural barriers	5	3
Country context w.r.t.					natural hideouts	3	0
income, %			IV composition [‡] , %		natural pathways	3	16
higher/ upper-middle	59	52	IV with topographic-		fractionalization	4	3
low/ lower-middle	30	30	variation components	11	amenity value	1	2
both	11	18	only				
			IV including non-	72			
Rural context, %	44	45	geographic components	72			
Number of studies	93	33	Number of studies	93	Number of mentions	103	49

Table 1: Descriptive statistics on studies included in systematic review

Note: [‡] 'Topographic-variation components only' means that the IV does not include other geographic or economic components.

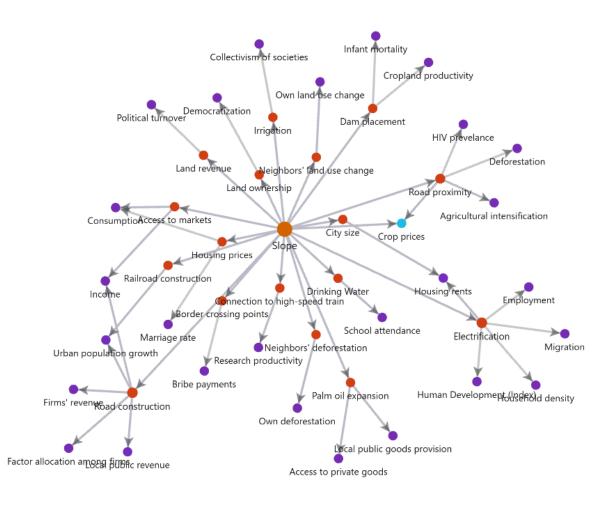
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Robustness Check

a) Land gradient/ slope IV studies from low-income countries

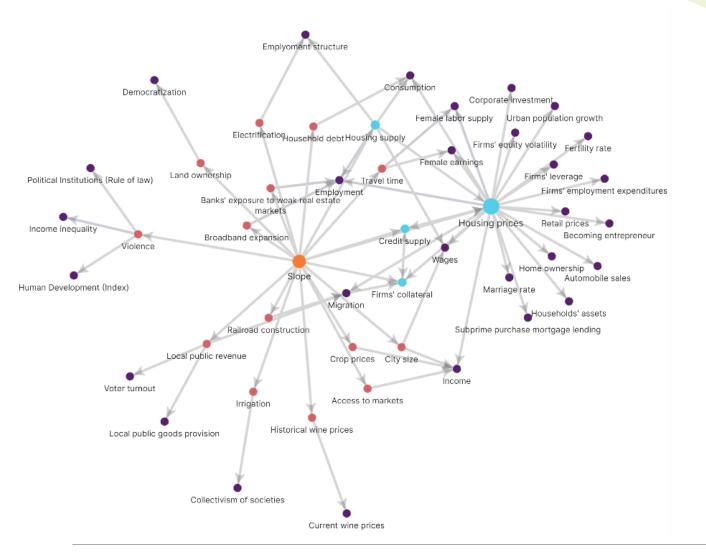


b) Ruggedness / elevation IV studies from low-income countries

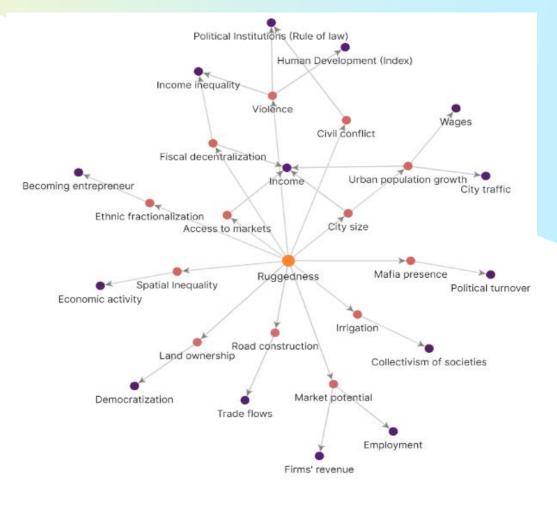


APPENDIX

Slope IV studies from high-income countries

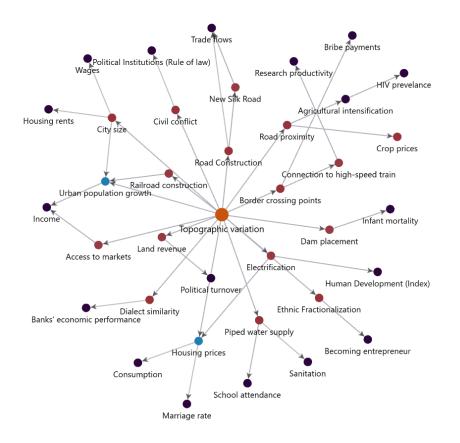


Ruggedness/ elevation IV studies from high-income countries

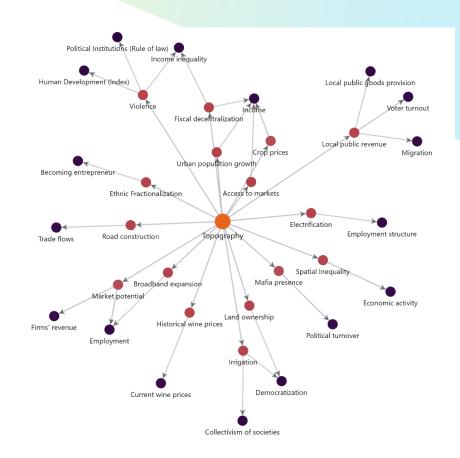




Urban low-income context

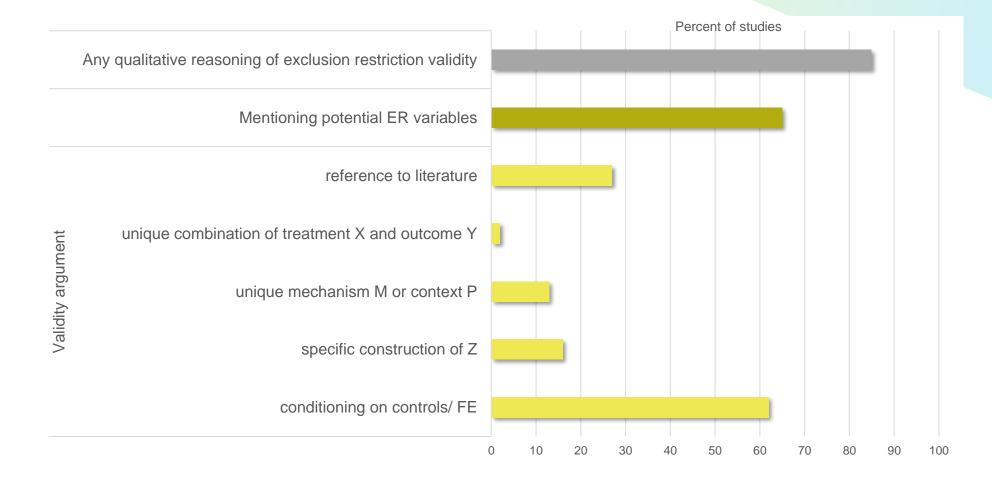


Rural high-income context



() rwi

3. Qualitative exclusion restriction reasoning



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APPLICATION TO DINKELMAN (2011)

Sample-isolation test

- test that seeks to isolate a subsample from the sample populatior with no road access
- Dinkelman (2011) excludes communities directly cut by a major national road
- More appropriate measure of road access for African context:

5 km distance (Raballand et al. 2010)

 ...for which electrification effect becomes insignificant IV estimates of electrification effect on employment, by maximum distance to roads of communities excluded from the sample

