DiD Session Discussion, or How I Learned to Stop Worrying and

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Love Untestable Assumptions

DiD/panel data: popular identification/estimation strategies

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- 3. Under what conditions are they the same?

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Follow up: can we find an estimator that estimates what we want, under the assumptions that we're willing to make?

A core result

A core result For a 2x2 DiD

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XZD: No anticipation, both groups treated \Rightarrow pre-ATT is 0 for both groups

- insight: useful to define add'l layer of POs based on group $Y_t(g,z)$

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= E[Y_{\text{post}}(1, 1) - Y_{\text{post}}(1, 0) \mid G = 1] - \mathbb{E}[Y_{\text{post}}(0, 1) - Y_{\text{post}}(0, 0) \mid G = 0]$$

What do we want to estimate?

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XZD: The **causal interaction** between group status and treatment

$$interaction \ effect = \mathbb{E}[(Y_{post}(1,1) - Y_{post}(1,0)) - (Y_{post}(0,1) - Y_{post}(0,0))]$$

- What would be the average difference in effects, if everyone was in group 1 vs. group 0?

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XZD: Under generalized DiD, the 2x2 DiD estimator is estimating a weird thing that is really just a description of heterogeneity



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LS: Restrictions on heterogeneity

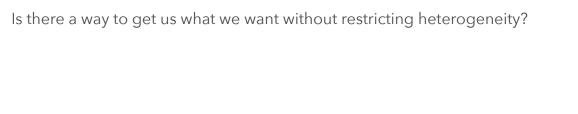
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- Generalized parallel trends assumptions (Assumption 4)
- Parallel trends across both levels of both treatments ($2 \times 2 = 4$ sets of constraints)
- What this buys us: conditioning on group doesn't matter, and so description = causal
- Is this assumption palatable?
 - Possible diagnostic with multiple periods and add'l (strong!) assumption of no effect carryover



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LS: Review of heterogeneity-robust estimators

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XZD: Unclear

- Are there opportunities to use multiple periods?
- Are we just out of luck?

XZD: Exclusion restriction and cannonical DiD

GDID setting recovers typical DiD setting under exclusion restriction (Assumption 5)

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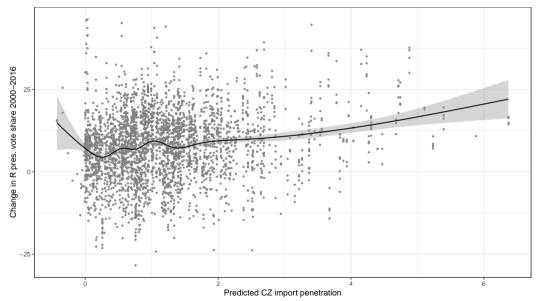
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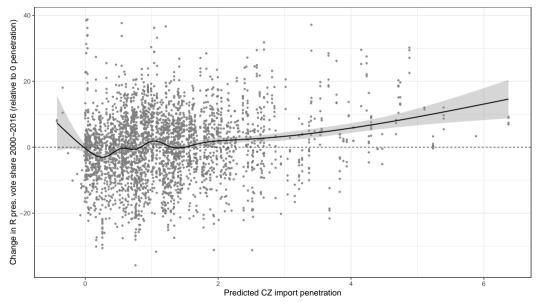
Are Bartik instruments/shift-share designs the other side of the coin?

- Groups defined by G have different level of exposure to Z
 - e.g. the "China shock" [Autor et al., 2013, 2020]
 - IV-style identification, but also fixed-effects/first differencing
 [Goldsmith-Pinkham et al., 2020; Borusyak et al., 2022; Borusyak and Hull, 2023]
 - Focus is on estimating effect of Z, not interaction, but does the GDID-style of analysis have something to say about shift-share designs, and vice versa?

Heterogeneity in impacts of "China shock" on presidential elections?



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- Discussion about omitted relative time period
 - "Omitting a baseline period that does not appear for all units... places a much greater weight on the effect homogeneity assumption" (pg 28)
 - "Omitting pre-treatment relative time indicators that are too far from the start of treatment risks placing greater weight on the effect homogeneity assumption" (pg 35)

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Including never-treated units vs sometimes-treated units

- Paglayan and Hall & Yoder examples: evidence of pre-trends with sometimes-treated units
- Q in paper: does including never treated units make it worse?
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Strong form of parallel trends \Rightarrow need to check for pre-trends across all pairs of groups?

- Omnibus diagnostic for strong parallel trends?
- Removing comparison pairs w/o parallel trends?
- But also pre-testing in problematic [Roth, 2018]

Your turn: Q&A

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