## Bartik instruments

Vasily Rusanov

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- ► General setup and definition (using Borusyak et al. notation)
- Some examples
- Growth rates as instruments (Borusyak et al.)
- Industry shares as instruments (Goldsmith-Pinkham et al.)
- Conclusion

## General setup

- $y_l = \beta x_l + \omega_l' \gamma + \varepsilon$ , possibly a panel
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- $y_l = \beta x_l + \omega_l' \gamma + \varepsilon$ , possibly a panel
- $ightharpoonup x_l$  is the endogenous regressor of interest, scalar at "locations" l
- ▶ Suppose we can decompose  $x_l = \sum_{n=1}^N s_{ln}g_{ln}$ , over "industries" n
- ▶ If  $g_{ln} = g_n + \tilde{g}_{ln}$ , then intuitive instrument:  $z_l = \sum_n s_{ln} g_n$ .
- lacktriangle Often not feasible:  $g_n$  is not observed. Use Leave-One-Out estimators.
- Decomposition  $x_l = \sum_n s_{ln} g_{ln}$  is nice for intuition, but not necessary. Bartik is whenever the instrument is a weighted average:  $z_l = \sum_n s_{ln} g_n$

### Simple example: Acemoglu & Linn, Market Size in Innovation

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- ▶  $y_l$ : Number of new drugs approved in category l  $x_l$ : growth in consumption of drugs in l in the US
- Decompose  $x_l = \sum_{n=1}^{N} s_{ln} g_{ln}$   $s_{ln}$  is the share of spending on drug in category l by age group n $g_{ln}$  is the growth in spending of age group n on drug category l.

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- ▶ Can decompose  $g_{ln} = g_n + g_{ln}$  , where  $g_n$  is simply the growth of population in group n.
- ▶ Plain Bartik is often not feasible as:  $g_n$  is not observed. Here,  $g_l n$  is not observed either, but it's ok!

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- $y_l = \beta x_l + \omega'_l \gamma + \varepsilon$ , instrument  $z_l = \sum_n s_{ln} g_n$ .
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- ▶ By definition,  $x_{lt} = \sum_{n=1}^{N} s_{lnt}g_{nt}$ , where  $s_{lnt}$  is the share of people in state l who work in industry n at time t  $g_{nt}$  is the growth of imports to the US (not location specific)

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- ▶ Instrument  $x_l$  with  $z_{lt} = \sum_{n=1}^{N} s_{lnt-1} \hat{g}_{nt}$ , where  $\hat{g}_{nt}$  is the growth of Chinese imports in industry n to 8 other countries.
- Note that  $\sum_{n} s_{lnt} \neq 1$  because some people work in non-manufacturing (but still tradeables).

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- ▶ We have  $s_{ln}$  and  $g_n$ . The assumptions required can be roughly classified into "shares are exogenous" or "growth rates" are exogenous.
- ▶ Turns out, using  $z_l = \sum_n s_{ln} g_n$  (a scalar) is numerically equivalent to using a vector  $g_n$  or a vector of  $s_l n$  in weighted regressions, which motivates two sets of restrictions.

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- ▶ Often  $\sum_n s_{ln} = 1$ . We usually don't think that  $\sum_l s_{ln} = 1$ .
- ▶ Bartik is num. equivalent to  $\bar{y}_n^{\perp} = \beta \bar{x}_n^{\perp} + \varepsilon^{\perp}$ , where  $\bar{y}_n = \sum_l s_{ln} y_l / \sum_l s_{ln}$  and the regression is weighted by  $\hat{s}_n$

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### Consistency when growth rates are instruments

- ▶ Under shocks-as-instruments approach,  $\hat{\beta}_{Bartik}$  is consistent iff  $\sum_n s_n g_n \phi_n \to 0$ , where  $s_n = \mathbb{E}(s_{ln})$  and  $\phi_n = \mathbb{E}(s_{ln}\varepsilon_l)/\mathbb{E}(s_{ln})$ . I don't understand it.
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$$\mathbb{E}(g_n|\phi_n) = \mu$$
,  $\forall n$ 

A2 1) 
$$\mathbb{E}[(g_n - \mu)(g_m - \mu)|\phi_n, \phi_m] = 0$$
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- ▶ This can be modified to have clusters with random assignment within clusters and to the case where  $\sum_{n} s_{ln} \neq 1$
- Similar for panel data.

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- $\begin{array}{c} \blacktriangleright \text{ Vectorize } \underbrace{S}: \text{ matrix of shares, } \underbrace{G}: \text{ vector of growth rates, } \underbrace{Y}: \\ \text{outcomes. Then } \underbrace{B}_{L\times 1} = ZG \text{ is the Bartik instrument.} \\ \end{array}$

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- $\qquad \qquad \text{We have } \hat{\beta}_{Bartik} = \frac{B'Y^{\perp}}{B'X^{\perp}} = \frac{G'Z'Y^{\perp}}{G'Z'X^{\perp}} = \underbrace{\overbrace{X^{\perp'}ZG}G'Z'Y^{\perp}}_{X^{\perp'}ZGG'Z'X^{\perp}}$
- ▶ Using Barik is the same as running GMM with weight matrix W = GG' and industry shares as instruments!

- ▶ This can be extended to a panel, but notation is long
- ► They provide several consistency conditions similar to Borusyak et al., but this doesn't give much intuition
- ▶ Key contribution: the GMM estimator gives way to Rotemberg weights.

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$$\hat{\beta}_k = (Z'_kX^{\perp})^{-1}Z'_kY^{\perp}$$

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- ▶ Then  $\hat{\beta}(\hat{W}) = \sum_{n=1}^{N} \hat{\alpha}_k(\hat{W}) \hat{\beta}_k$
- ▶ For the Bartik instruments, the weights are  $\hat{\alpha}_k(\hat{W}) = \frac{\hat{X}_k^{Bartik} X^{\perp}}{\hat{X}^{Bartik} X^{\perp}}$  where  $\hat{X}^{Bartik}$  is the residualized X after the first stage.
- ► This is a way to see whether some industries are more important than others. It also allows to estimate the bias.

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- ► Little "theory" is needed to use Bartik. It's an empirical tool, not a modeling technique.
- ➤ You need those two papers to justify your instrument, not to define it. They way they define Bartik is the same.
- ▶ It's bad if one industry *n* gives a significant part of variation (under both approaches).