

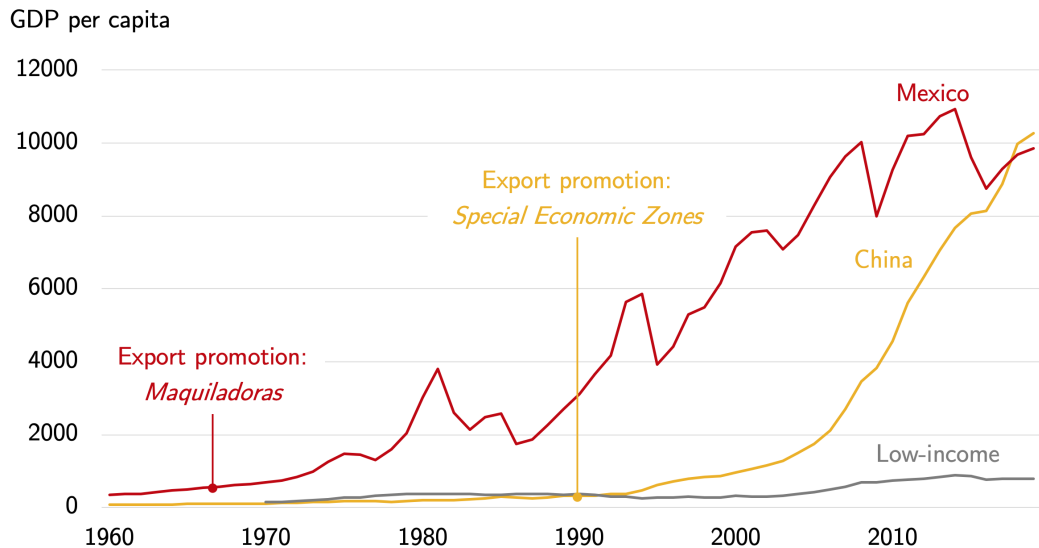
Robots and reshoring: Evidence from Mexican labor markets

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In last 50 years, some countries have made leap from low-income to middle-income by using cheap, manual labor to their advantage



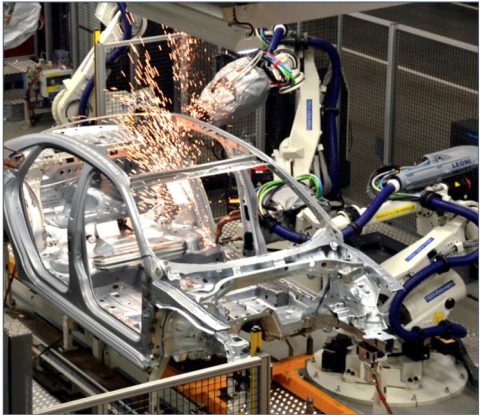
More recently, industrial robots have become powerful alternative to perform many routine, manual tasks



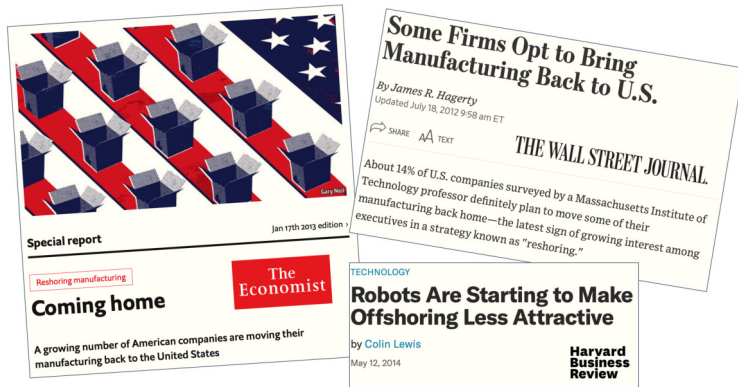
1990s: Puebla, Mexico



2010s: Tennessee, USA



Worries whether export-led growth still a viable path to economic growth in age of automation, but empirical evidence focuses on developed world



Literature so far

- OECD countries: Gratz & Michaels (2018, REStat)
- United States: Acemoglu & Restrepo (2020, JPE)
- Germany: Dauth, Findeisen, Suedekum & Woessner (2020, R&R JEEA)

Research question

How do robots affect employment in offshoring countries?

Method

- Use context of Mexico and its trading relations with US
- Exploit variation in *exposure to foreign robots* across Mexican local labor markets between 1990 and 2015
- Construct instrumental variable based on robot adoption outside of US and Mexico to purge results from reverse causality

Results

- US robots reduce employment in Mexico
- Low-educated machine operators in manufacturing most affected
- Evidence for reshoring (reverse of offshoring) as mechanism

1. Empirical strategy and data

2. Main results

3. Mechanism

4. Conclusion

Accounting for foreign robots in Acemoglu & Restrepo (2020) framework

- Equilibrium response of employment in commuting zone c (L_c) to advances in robotic automation technology from AR (2020):

$$d \ln L_c = \beta_c \sum_{i \in I} \ell_{ci} \frac{dR_i}{L_i} + \epsilon_c, \quad \text{where } \ell_{ci} = \frac{L_{ci}}{L_c} \quad (1)$$

- In offshoring country, considerable share of employment in export-producing sector ($\sim 30\%$ of GDP in Mexico)
- Workers compete not only with *domestic* robots (R_i^d), but also *foreign* robots (R_i^f) in offshorable industries (indicator $O_i = 1$)
- Account for this by including *exposure to foreign robots*:

$$d \ln L_c = \beta_c \underbrace{\sum_{i \in I} \ell_{ci} \frac{dR_i^d}{L_i}}_{\text{Exposure to domestic robots}_c \text{ as in AR (2020)}} + \beta_c \underbrace{\sum_{i \in I} \ell_{ci}^f \frac{dR_i^f O_i}{L_i^f}}_{\text{Exposure to foreign robots}_c} + \epsilon_c \quad (2)$$

Taking this to the data using four sources

1. Robots (IFR):

- Shipments of *industrial robots* for 11 countries and 19 industries since 1993
- Typical applications:
 - Handling, welding, assembling, packaging, dispensing (Manufacturing)
 - Harvesting (Agriculture)
 - Inspecting of structures and equipment (Utilities)

2. Mexican census (INEGI):

- Employment status, municipality of residence and work place, and education level, among others
- Can construct Commuting Zones (CZs)

3. Maquiladoras (UN CEPAL):

- Factories in Mexico required by law to export all goods they produce
- Information on number of Maquiladora employees by industry and municipality in 1990

4. Exports (Mexico's Tax Administration Service, SAT):

- Value of exports and number of export-producing plants
- By municipality and product code, from 2004-2014

$$\text{Exposure to domestic robots } c, (t_0, t_1) \equiv \sum_{i \in I} \ell_{ci, 1990} \left(\frac{R_{i, t_1}^{MX} - R_{i, t_0}^{MX}}{L_{i, 1990}} \right)$$

$$\text{Exposure to foreign robots } c, (t_0, t_1) \equiv \sum_{i \in I} \ell_{ci, 1990}^f \left(\frac{R_{i, t_1}^{US} - R_{i, t_0}^{US}}{L_{i, 1990}^f} \right) O_{i, 1992}$$

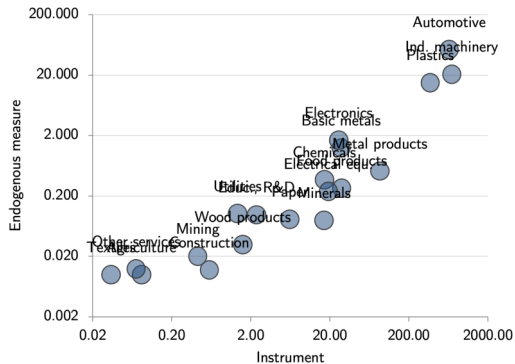
- $\ell_{ci, 1990}$: 1990 share of Commuting Zone c employment in industry i (f =exports sector)
- $L_{i, 1990}$: 1990 employment in industry i (f =exports sector)
- $R_{i, t}^j$: Industrial robots in country j and industry i at time t
- $O_{i, 1992}$: 1992 share of Mexican imports of US output in industry i

$$\text{External exposure to domestic robots } c, (t_0, t_1) \equiv \sum_{i \in I} \ell_{ci, 1990} \left(\frac{R_{i, t_1}^{WLD} - R_{i, t_0}^{WLD}}{L_{i, 1990}} \right)$$

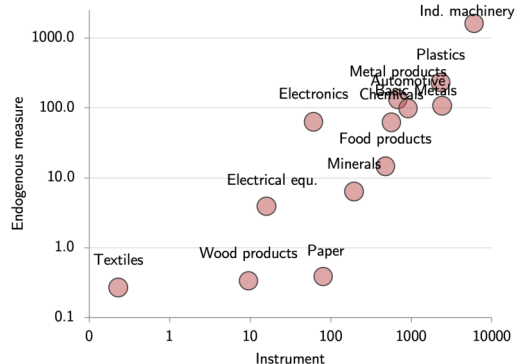
$$\text{External exposure to foreign robots } c, (t_0, t_1) \equiv \sum_{i \in I} \ell_{ci}^f, 1990 \left(\frac{R_{i, t_1}^{WLD} - R_{i, t_0}^{WLD}}{L_{i, 1990}^f} \right) \tilde{O}_{i, 1990}$$

- $\ell_{ci, 1990}$: 1990 share of Commuting Zone c employment in industry i (f =exports sector)
- $L_{i, 1990}$: 1990 employment in industry i (f =exports sector)
- $R_{i, t}^j$: Industrial robots in country j and industry i at time t
- $\tilde{O}_{i, 1990}$: 1990 share of offshorable intermediates in industry i in US (Feenstra & Hanson, 1999)

First-stage industry variation

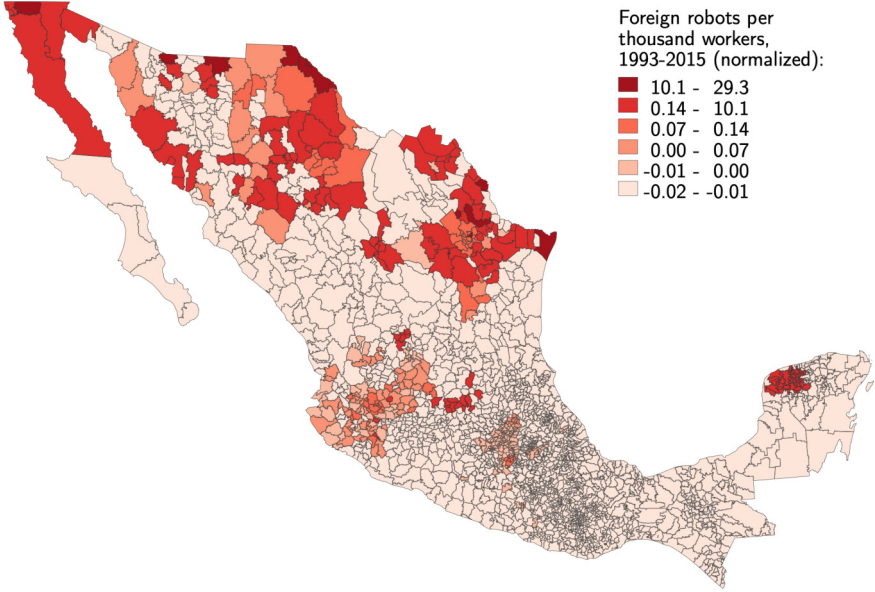


A. Domestic robots



B. Foreign robots

Regional variation in external exposure to foreign robots



Estimating equation

$$\Delta y_{c,(t_0,t_1)} = \beta^d \text{Exposure to domestic robots}_{c,(t_0,t_1)} + \beta^f \text{Exposure to foreign robots}_{c,(t_0,t_1)} + \gamma \mathbf{X}_{c,1990} + \delta_{(t_0,t_1)} + \varepsilon_{c,(t_0,t_1)}$$

- c : Commuting Zone c (1,805 CZs)
- (t_0, t_1) : Two stacked time periods (1990–2000 & 2000–15)
- y : Employment-to-population ratio as main dependent variable
- $\mathbf{X}_{c,1990}$: Vector of covariates for Commuting Zone c in 1990
 - Region dummies
 - Main effects (Maquiladoras, US import reliance)
 - Demographic characteristics & initial conditions
 - Broad industry shares
 - Contemporaneous changes (Chinese imports, NAFTA, computers)
 - Commuting Zone trends (in stacked differences)

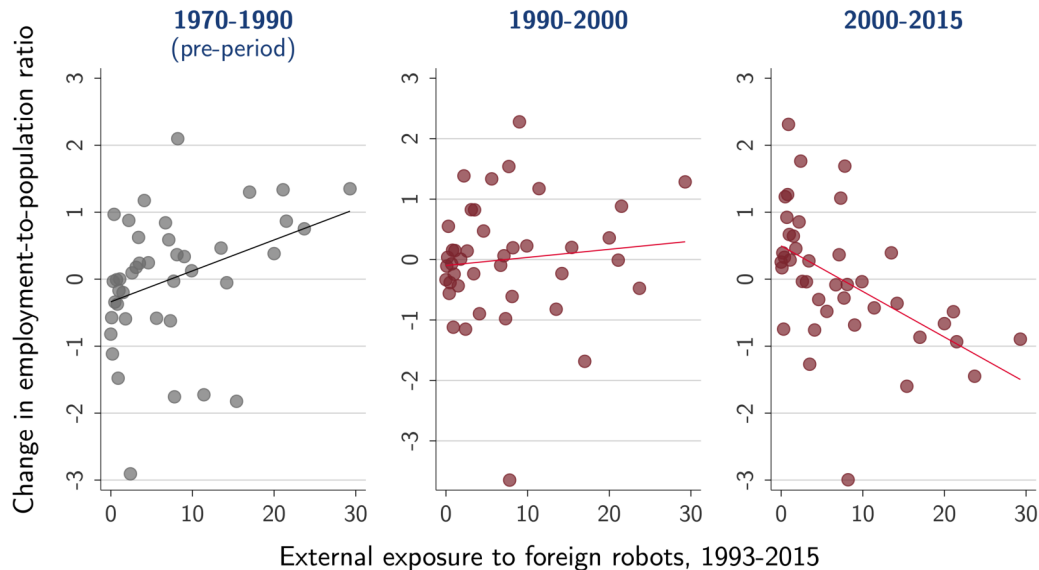
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Raw binned scatter plot



Stacked differences regressions (1990–2000 & 2000–15)

	(1)	(2)	(3)	(4)	(5)	(6)
	Δ Employment-to-population ratio (2SLS)					
Exposure to <i>domestic</i> robots	-0.57* (0.33)	-0.07 (0.23)	0.30 (0.24)	-0.17 (0.24)	0.58** (0.29)	-0.11 (0.33)
Exposure to <i>foreign</i> robots	-0.67*** (0.18)	-0.75*** (0.19)	-0.58*** (0.14)	-0.61*** (0.16)	-0.72*** (0.13)	-0.52** (0.23)
Kleibergen-Paap rank F	706	222	198	1318	159	104
Period dummies	✓	✓	✓	✓	✓	✓
Region & main effects	✓	✓	✓	✓	✓	✓
Baseline covariates		✓	✓	✓	✓	✓
Contemp. changes			✓	✓	✓	✓
Unweighted				✓		
Only Maquiladora CZs					✓	
CZ trends						✓
Observations	3,610	3,610	3,610	3,610	502	3,610

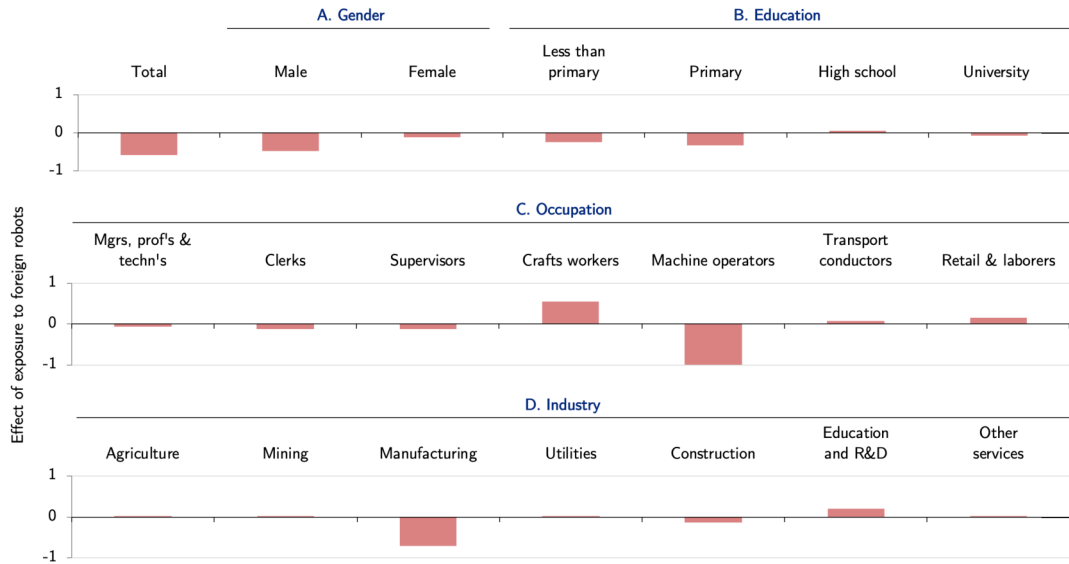
Effects by time period (1990–2000, 2000–15)

	(1)	(2)	(3)	(4)	(5)	(6)
	Δ Employment-to-population ratio (2SLS)					
	1990-2000			2000-2015		
Exposure to <i>domestic</i> robots	-4.73 (5.10)	0.69 (3.60)	3.14 (4.36)	-0.27 (0.33)	0.27 (0.26)	0.44 (0.29)
Exposure to <i>foreign</i> robots	-0.23 (1.23)	-0.04 (1.52)	-0.02 (1.48)	-0.28** (0.14)	-0.44*** (0.15)	-0.66*** (0.13)
Kleiberger-Paap rank F	84	98	99	62	75	82
Region & main effects	✓	✓	✓	✓	✓	✓
Baseline covariates		✓	✓		✓	✓
Contemp. changes			✓			✓
Observations	1,805	1,805	1,805	1,805	1,805	1,805

Results robust to several alternative explanations:

- No significant pre-trends 1970–90
- Not driven by contemporaneous shocks to single industries
- Pattern arises from changes in employment, not migration
- Visible using alternative instruments
- Excluding top 1% of observations with respect to exposure to foreign robots
- Using fixed effects for 31 states instead of nine broad regions
- Using LASSO procedure for covariate selection

Effect heterogeneity



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2. Main results

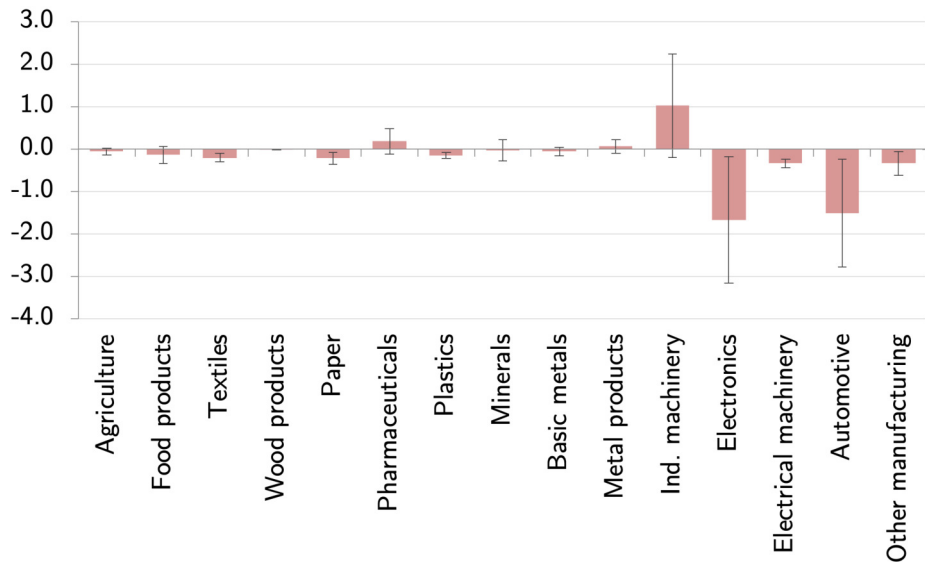
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Reshoring as mechanism: Reduction in exports

	(1)	(2)	(3)	(4)	(5)	(6)
	Δ 2004-2014 (2SLS)					
	Exports value per worker			Exports plants per worker		
Exposure to <i>domestic</i> robots	5.23*** (1.87)	3.44 (2.89)	2.84 (3.03)	0.15*** (0.05)	0.02 (0.06)	0.05 (0.05)
Exposure to <i>foreign</i> robots	-4.07*** (1.06)	-3.15*** (1.07)	-2.61** (1.03)	-0.40*** (0.08)	-0.14*** (0.04)	-0.13*** (0.04)
Kleibergen-Paap rank F	57	120	116	57	58	69
Region & main effects	✓	✓	✓	✓	✓	✓
Baseline covariates		✓	✓		✓	✓
Contemp. changes			✓			✓
Observations	1,805	1,805	1,805	1,805	1,805	1,805

Effect of US robots on exports by industry



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Conclusion

- US robots reduce employment in Mexico
 - Regions with average exposure to foreign robots have 0.4 percentage points lower emp-to-pop ratio
 - Nationally, amounts to roughly 270,000 fewer jobs
 - Negative employment effect strongest for
 - men
 - less educated
 - machine operators
 - workers in manufacturing industry
 - Reshoring as mechanism: Employment effects mirrored in reduced export volumes
- ⇒ **Automation technologies capable of changing comparative advantages across countries**