

# **Robot, Employment, and Population: Evidence from Articulated Robot in Japan's Local Labor Markets**

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July 28, 2019

# Automation technology and employment

- How does the automation affect our future work?
- Half of current occupations disappear due to the penetration of AI (Frey and Osborne, 2017)
- Penetration of robots reduced the local employment-population ratio in the US (Acemoglu and Restrepo, 2019)
- Penetration of robots reduced the local manufacturing employment but increased local service employment in Germany (Dauth et al., 2018)
- Impacts in Japan? - Most advanced robot technology, Declining population

# Overview

- Regional analysis based on 300 commuting zones (CZs)
- Heterogeneous robot penetration across CZs depending on initial industrial structure and heterogeneous robot penetration across industries
- Instrumented by German penetration index
- Robot penetration increased local employment as well as local population
- In total, robot penetration reduced the employment-population ratio

## Variety of robots

- Cartesian robot
- SCARA robot
- **Articulated robot**
- Parallel robot
- Cylindrical robot

Figure: Articulated robot

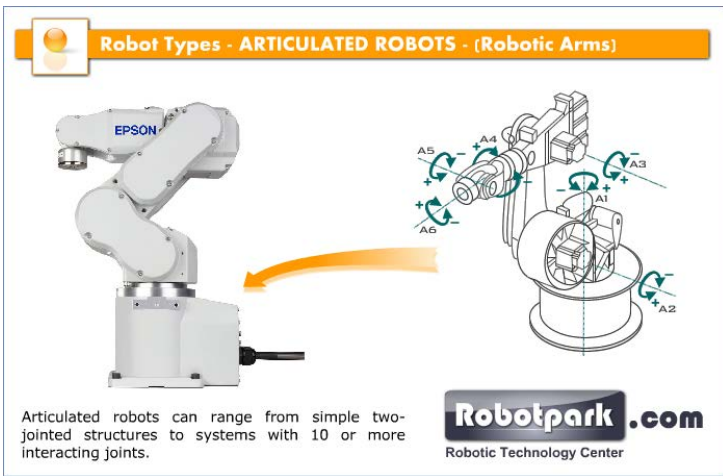
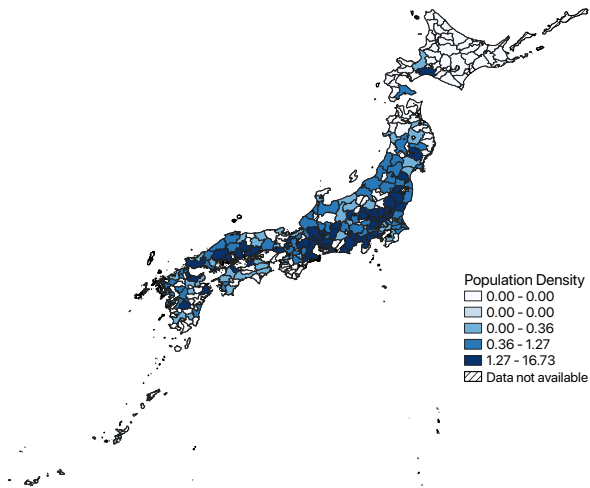


Table: Per-thousand-worker Penetration of Robot, Articulated Robot, 2002-17

Industry	Penetration of Robot
Transport machinery	19.033
Chemical products	0.326
Food / Beverage / Tobacco	0.275
General machinery	-5.863
Precision and optics	-7.481
Non-ferrous metal	-12.253

Figure: Density of automobile industry as of 2002



## Construction of robot exposure

**Robot growth in industry  $i$ :**

$$PR_{i,(2002,2017)}^s \equiv \frac{R_{i2017}^s - R_{i2002}^s}{L_{i2002}}, \quad (1)$$

where  $R_{i,t}^s$  is industry- $i$ , structure- $s$ , year- $t$  stock of robot.

**Robot exposure in CZ  $c$ :**

$$ER_{c,(2002,2017)}^s \equiv \sum_i l_{ci2002} PR_{i,(2002,2017)}^s, \quad (2)$$

where  $l_{ci2002} \equiv \frac{L_{ci2002}}{\sum_k L_{ck2002}}$  is the industry- $i$  share of employment in CZ  $c$  and year 2002.



## Empirical strategy

### Estimation equation:

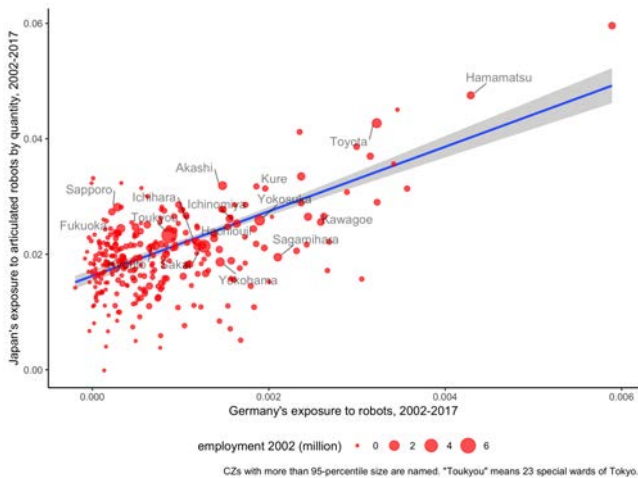
$$y_c^g = \alpha^{gs} + \beta^{gs} ER_c^s + X_c' \gamma^{gs} + \varepsilon_c^{gs}, \quad (3)$$

where  $y_c^g$  is one of CZ  $c$ -demographic group  $g$  outcome variables,  $X_c$  is the (column) vector of control variables of CZ  $c$ , and  $\varepsilon_c^{gs}$  is the error term.

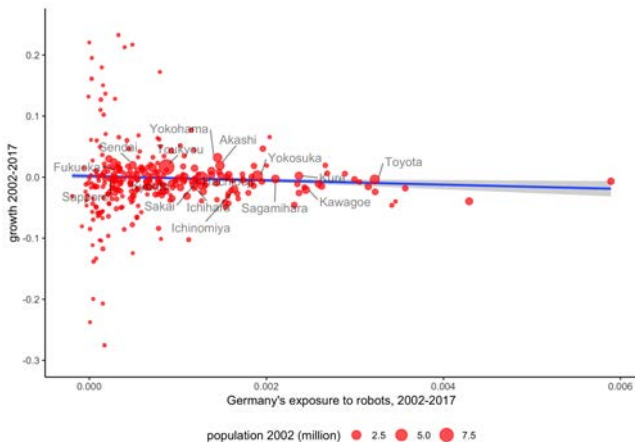
### Instrumental variable:

$$ER_{c,(t_0,t_1)}^{DEU} \equiv \sum_i l_{cit-1} PR_{i,(t_0,t_1)}^{DEU}, \quad (4)$$

# First stage

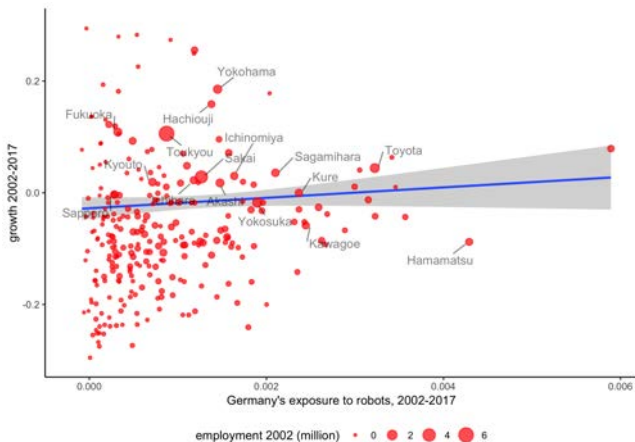


# Reduced form, employment to population ratio



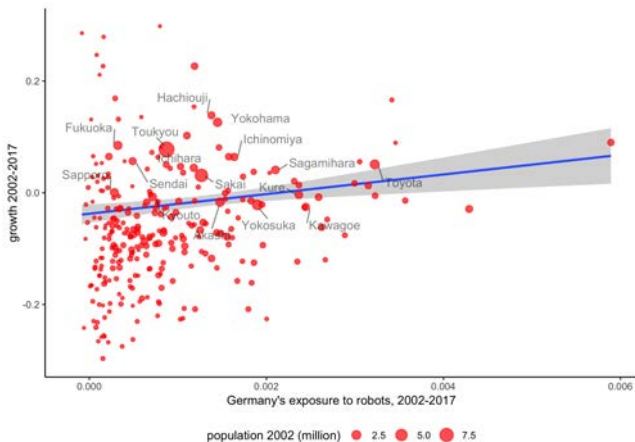
CZs with more than 30 percent change are trimmed from the figure.  
CZs with more than 95-percentile size are named. "Toukyou" means 23 special wards of Tokyo.

# Reduced form, employment growth rate



CZs with more than 30 percent change are trimmed from the figure.  
CZs with more than 95-percentile size are named. "Toukyou" means 23 special wards of Tokyo.

# Reduced form, population growth rate



CZs with more than 30 percent change are trimmed from the figure.  
CZs with more than 95-percentile size are named. "Toukyou" means 23 special wards of Tokyo.

Table: IV First Stage and Reduced Form

	<i>Dependent variable:</i>			
	$ER$	$\Delta \frac{L}{P}$	$g^L$	$g^P$
	(1)	(2)	(3)	(4)
$ER^{DEU}$	0.220*** (0.012)	-0.234* (0.121)	1.289** (0.632)	1.731*** (0.523)
Observations	302	302	302	302
R <sup>2</sup>	0.514	0.012	0.014	0.035
Adjusted R <sup>2</sup>	0.513	0.009	0.010	0.032

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01. See the main text for the definition of each dependent and independent variables. All regressions are weighted by base-year populations in each CZ.

Table: IV regressions

	<i>Dependent variable:</i>		
	$\Delta \frac{L}{P}$	$g^L$	$g^P$
	(1)	(2)	(3)
<i>ER</i>	-1.067* (0.553)	5.868** (2.892)	7.883*** (2.418)
Observations	302	302	302

*Notes:* \* $p < 0.1$ ; \*\* $p < 0.05$ ; \*\*\* $p < 0.01$ . See the main text for the definition of each dependent and independent variables. All regressions are weighted by base-year populations in each CZ. *ER* is instrumented by  $ER^{DEU}$ .

## Tentative conclusion

- Articulated robot penetration affected different regions heterogeneously.
- Penetration of articulated robot decreased employment/population ratio.
- Penetration of articulated robot increased both employment and population.
- The impact on population dominated the impact on employment.
- Automation technology help local area to sustain employment and population.



- Acemoglu, Daron and Pascual Restrepo**, “Automation and New Tasks: How Technology Displaces and Reinstates Labor,” Technical Report, National Bureau of Economic Research 2019.
- Dauth, Wolfgang, Sebastian Findeisen, Jens Suedekum, and Nicole Woessner**, “Adjusting to Robots: Worker-Level Evidence,” Technical Report Institute Working Paper 13 2018.
- Frey, Carl Benedikt and Michael A. Osborne**, “The future of employment: How susceptible are jobs to computerisation?,” *Technological Forecasting and Social Change*, 2017, 114 (C), 254–280.