The Granular Origins of Agglomeration

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MIT

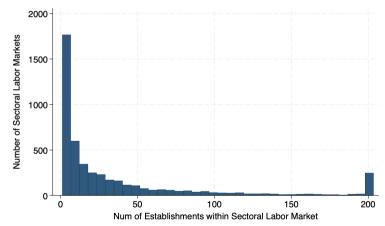
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Motivation

- Individual firms play a key role in local labor markets
 - Kodak in Rochester, Toyota in Toyota, Microsoft for engineers in Seattle

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 - Japanese local labor market (2-digit mfg \times CZ): median of 13 plants



Motivation

- Individual firms play a key role in local labor markets
 - Kodak in Rochester, Toyota in Toyota, Microsoft for engineers in Seattle
 - Japanese local labor market (2-digit mfg \times CZ): median of 13 plants
- Firm-specific shocks can have a big impact on the whole labor market
 - People can end up unemployed because a single firm had a bad year
 - Firms can have a tough time finding workers to expand

What We do

- 1. A model of a labor market with a finite number of firms s.t. idios, shocks
 - Show that there are increasing returns to scale
 - Derive testable empirical predictions that speak directly to the mechanism
- 2. Tests of the empirical predictions in Japanese microdata
 - The variance of the log wage bill decreases in the size of the labor market
 - The variance of log firm employment increases in the size of the labor market
 - Firms with a larger employment share respond less to demand shocks
- 3. A quantitative model of economic geography to quantify the mechanism

Related Literature

 Labor Market Pooling: Theory: Marshall (1890), Krugman (1991), Duranton and Puga (2004), Stahl and Walz (2001); Empirics: Overman and Puga (2010), Nakajima and Okazaki (2012), Almeida and Rocha (2018)

This paper: Stylized model for empirical predictions, direct quantification of the mechanism

- Granularity: Gabaix (2011), Hottman, Redding, Weinstein (2016), Gaubert and Itskhoki (2021)

This paper: Spatial implications, relevant for medium-sized cities, not just small towns

- Job Search in Large/Thick Markets: Moretti and Yi (WP), Andersson et al. (2014), Gan and Zhang (2006)

This paper: Similar implications, different mechanism

- Japan: Nakamura (1985), Tabuchi and Yoshida (2000), Nakajima et al. (2012), Miyauchi (2018)

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Baseline Model

The Model

- Small, open region with E establishments (firms) and a mass ℓ of workers
- Ex-ante homogeneous firms (for now)
- In a pre-period, the state of the world $s \in \mathcal{S}$ is revealed o firm productivity
- Firms then choose labor to maximize profits taking wages and prices as given

$$\ell_{e}(s) \in \operatorname*{argmax} \quad a_{e}(s) f\left(\ell'\right) - w(s) \ell'$$

where $a_e(s)$ are iid across firms, $f(x) = x^{\eta}$

- Workers inelastically supply labor

$$\ell = \sum_{e} \ell_{e}(s)$$

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Characterization: Expected Production

- Labor demand is characterized by the FOC

$$a_e(s)f'(\ell_e(s)) = w(s)$$

- Wages adjust to clear the labor market in every state of the world s

$$oldsymbol{w}(oldsymbol{s}) = \eta \ell^{\eta-1} \left[\sum_{oldsymbol{e} \in \mathcal{E}} (oldsymbol{z}_{oldsymbol{e}} oldsymbol{a}_{oldsymbol{e}}(oldsymbol{s}))^{rac{1}{1-\eta}}
ight]^{1-\eta}$$

- Then expected production is

$$Y(\ell,\mathcal{E}) = \mathbb{E}\left[\ell^{\eta}\left[\sum_{m{e}\in\mathcal{E}}(z_{m{e}}a_{m{e}}(m{s}))^{rac{1}{1-\eta}}
ight]^{1-\eta}
ight]$$

Increasing returns to scale

Proposition

If $Var(a_e(s)) > 0$, then expected production has increasing returns to scale. In math, for any $\ell > 0$, $E \in \mathbb{N}$, and $\alpha > 1$ so that $\alpha E \in \mathbb{N}$,

$$Y(\alpha \ell, \alpha E) > \alpha Y(\ell, E)$$
.

Comments:

- Larger markets are more productive!
- Without uncertainty, no benefit to being in a larger labor market

Sketch of proof when $\alpha = 2$

- Two separate labor markets each with E establishments and ℓ workers \implies double the production.
- Idiosyncratic firm shocks \implies sometimes the wages in labor market 1 will be higher
- Move a small number of workers from labor market 2 to labor market 1 when wages are higher. This must increase production.

Firm Side Intuition: In response to shock to $a_e(s)$

- Recall labor demand is

$$a_e(s)f'(\ell_e(s)) = w(s)$$

Firm Side Intuition: In response to shock to $a_e(s)$

- Recall labor demand is

$$a_e(s)f'(\ell_e(s)) = w(s)$$

- Suppose that there is one firm. It must always hire everyone even if unproductive

$$a_e(s)f'({\color{red}\ell})=w(s)$$

Firm Side Intuition: In response to shock to $a_e(s)$

- Recall labor demand is

$$a_e(s)f'(\ell_e(s)) = w(s)$$

- Suppose that there is one firm. It must always hire everyone even if unproductive

$$a_e(s)f'(\ell) = w(s)$$

- Suppose that there are many firms so that wages are constant. Firms can adjust labor as they wish

$$a_e(s)f'(\ell_e(s)) = w$$

Disappearing Agglomeration

Proposition

As the labor market becomes infinitely large, production converges to constant returns to scale.

In math, suppose that $\ell > 0$, E > 0 and $\alpha > 1$. Then

$$\frac{Y(\alpha\kappa\ell,\alpha\kappa E)}{\alpha Y(\kappa\ell,\kappa E)}\to 1$$

as $\kappa \to \infty$.

Comments:

- By using models with a continuum of firms, we miss this force.
- Larger market would be largely unaffected

New Reason for Spatial Policy

Proposition

Adding new firms increases expected production more than the profits those firms would earn.

In math, for $\alpha > 1$,

$$\mathbb{E}[\sum_{\boldsymbol{e} \in \alpha \mathcal{E} \setminus \mathcal{E}} \pi_{\boldsymbol{e}}(\boldsymbol{s})] < Y(\ell, \alpha \mathcal{E}) - Y(\ell, \mathcal{E}),$$

where $\pi_e(s) = z_e a_e(s) \ell_e(s)^{\eta} - w(s) \ell_e(s)$ are the profits earned when there are $\alpha \mathcal{E}$ set of firms operating.

Comments:

- If the firm entry is somewhat elastic, under-entry
- Violates FWT because it's not Walrasian entry: firms internalize the increase in wages when they enter

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Cross-sectional Implications of the Model

Proposition

To a first-order log-linear approximation around a symmetric equilibrium:

- The variance of log wage bill is decreasing *E*:

$$Var(\log w(s)\ell) \approx \frac{\sigma^2}{E};$$

- The variance of log employment for an establishment is increasing in *E*:

$$Var(\log \ell_{\boldsymbol{e}}(\boldsymbol{s})) pprox rac{\sigma^2}{(1-\eta)^2} \left(1 - rac{1}{E}\right)$$

where
$$\sigma^2 = var(\log a_e(s))$$
.

Comparative Statics Implied by the Model

Proposition

In response to a productivity shock, firms that have a larger share of the labor market expand less.

In math:

$$\Delta \log \ell_{\mathbf{e}}(s) pprox rac{1}{1-\eta} \left[1-\mu_{\mathbf{e}}\right] \Delta \log a_{\mathbf{e}}(s)$$

where $\mu_e = \frac{\ell_e(s)}{\sum_{e'}\ell_{e'}(s)}$ is the share of labor hired by establishment e

Comments:

- In larger labor markets, firms are a smaller share of the market and so can expand without issue

Robustness

- Imperfect mobility across establishments and labor markets Details
- Monopsony power Details
- Labor hoarding/employer insurance Potalis
- Wage rigidity Petails

Empirical Evidence

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- Japanese Census of Manufactures (CoM)
 - Annual survey of all manufacturing establishments with at least 4 employees
 - For 2011, 2016 (Economic Census)
 - Employment, product sales, export sales by establishment
- Sample Construction: 724,417 unique establishments
 - 1986-2016
 - Manufacturing
 - Must appear for at least 5 years consecutively
- Local Labor Market:
 - JSIC 2 digit manufacturing industry × commuting zone
 - 25 unique 2-digit manufacturing industries (robust using 3-digit or just MFG)
 - 256 commuting zones

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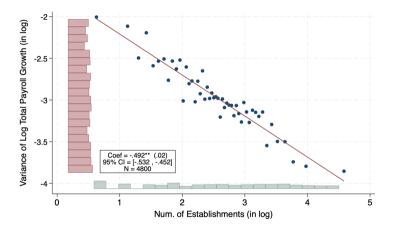
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Fact 1: Volatility of Log Total Payment to Labor

- For each LLM,
 - Compute one-year log growth of total payroll in each year
 - Take LLM-level variance over time
- Correlation with number of establishments in each LLM

Fact 1: Volatility of Log Total Payment to Labor





Fact 2: Volatility of Establishment-level Employment

- Establishments in larger markets adjust employment more flexibly?
 - Variance of log growth in establishment-level employment
- First residualize estab. yearly employment year FEs

In
$$\ell_{m{e},t} = \eta_t + arepsilon_{m{e},t}^\ell$$

- Second, compute yearly change

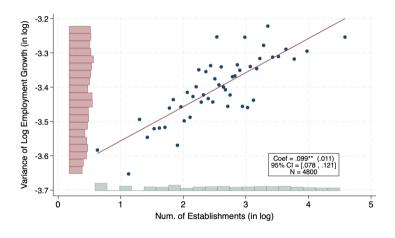
$$\Delta arepsilon_{e,t,t+1}^\ell \equiv \hat{arepsilon}_{e,t+t}^\ell - \hat{arepsilon}_{e,t}^\ell$$

- Then residualize by estab. employment and estab-age FEs

$$\Delta \varepsilon_{e,t,t+1}^{\ell} = \gamma \ln \ell_{e,t} + \eta_{age(e)} + \zeta_{e,t,t+1}$$

- Finally take variance $Var(\hat{\zeta}_{e,t,t+1})$ across time

Fact 2: Volatility of Estab-level Employment





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Causal Analysis: Empirical Specification

- Specification

$$\Delta \ln \ell_{e,t,t+1} = \beta \Delta \mu_{e,t,t+1} + \mathbf{X}'_{e,t} \Gamma + \zeta_e + \zeta_t + \varepsilon_{e,t}, \tag{1}$$

where $\Delta \ln \ell_{e,t}$ is the change in employment

- $\Delta \mu_{e,t,t+1}$: the shift-share demand shock

$$\Delta\mu_{e,t,t+1} = \overline{\mathsf{EXP}_e} \times \left(\sum_{c} \overline{\omega_{e,c}} \cdot \Delta REX_{c,t,t+1}^{JPN} \right) \tag{2}$$

- $\overline{\text{EXP}_e}$: median export ratio
- $\overline{\omega_{e,c}}$ median exposure of establishment e to country c from product mix
- $\Delta REX_{c,t,t+1}^{JPN}$ is the change in real exchange rate of the currency
- $\mathbf{X}_{e,t}$: Establishment age squares, log payroll

Regression without the interaction term for the proof of concept of the shock

Table: Effects of JPY Appreciation on Employment Growth

	Dep. Var.: Log Changes Employment by Types			
	Sales	Employment	Regular	Non-Regular
AREER Shock				
Observations	1,164,363	1,164,363	1,164,363	1,164,363
Covariates	✓	✓	✓	√
Year FEs	\checkmark	\checkmark	\checkmark	\checkmark
Establishment FEs	✓	✓	✓	✓

Regression without the interaction term for the proof of concept of the shock

Table: Effects of JPY Appreciation on Employment Growth

	Dep. Var.: Log Changes Employment by Types			
	Sales	Employment	Regular	Non-Regular
AREER Shock	-3.46			
	(0.17)			
Observations	1,164,363	1,164,363	1,164,363	1,164,363
Covariates	✓	✓	✓	✓
Year FEs	\checkmark	\checkmark	\checkmark	\checkmark
Establishment FEs	✓	✓	✓	✓

Regression without the interaction term for the proof of concept of the shock

Table: Effects of JPY Appreciation on Employment Growth

	Dep. Var.: Log Changes Employment by Types			
	Sales	Employment	Regular	Non-Regular
AREER Shock	-3.46	-0.25		
	(0.17)	(0.09)		
Observations	1,164,363	1,164,363	1,164,363	1,164,363
Covariates	✓	✓	✓	√
Year FEs	\checkmark	\checkmark	\checkmark	✓
Establishment FEs	✓	✓	✓	✓

Regression without the interaction term for the proof of concept of the shock

Table: Effects of JPY Appreciation on Employment Growth

	Dep. Var.: Log Changes			
		Employment by Types		
	Sales	Employment	Regular	Non-Regular
AREER Shock	-3.46	-0.25	-0.29	
	(0.17)	(0.09)	(0.12)	
Observations	1,164,363	1,164,363	1,164,363	1,164,363
Covariates	✓	✓	✓	✓
Year FEs	\checkmark	\checkmark	\checkmark	\checkmark
Establishment FEs	✓	✓	✓	✓

JPY Appreciation Decreases Establishments' Employment

Regression without the interaction term for the proof of concept of the shock

Table: Effects of JPY Appreciation on Employment Growth

	Dep. Var.: Log Changes				
		Employment by Types			
	Sales	Employment	Regular	Non-Regular	
AREER Shock	-3.46	-0.25	-0.29	-2.62	
	(0.17)	(0.09)	(0.12)	(0.23)	
Observations	1,164,363	1,164,363	1,164,363	1,164,363	
Covariates	✓	✓	✓	✓	
Year FEs	\checkmark	\checkmark	\checkmark	\checkmark	
Establishment FEs	✓	✓	✓	✓	

Interaction with Employment Share

- Focus on Non-Regular workers
- Interaction with employment share within LLM $s_{e,t-1}$

$$\Delta \ln \ell_{e,t,t+1} = (\beta_1 + \beta_2 \cdot s_{e,t-1}) \cdot \Delta \mu_{e,t,t+1} + \mathbf{X}'_{e,t} \Gamma + \zeta_e + \zeta_t + \varepsilon_{e,t}, \quad (3)$$

- β_2 : Heterogeneous responses with employment share

Table: Effects of JPY Appreciation on Employment Growth

	Dep. Var.: Log Changes in Non-Regular Emp.			
	(1)	(2)	(3)	(4)
AREER Shock	-2.62			
	(0.23)			

AREER Shock × Payroll Share

Observations	1,164,363	1,164,363	1,164,363	1,164,363
Covariates	√	✓	✓	$\overline{\hspace{1cm}}$
Year FEs	\checkmark	\checkmark	\checkmark	\checkmark
Establishment FEs	✓	✓	✓	✓

Table: Effects of JPY Appreciation on Employment Growth

	Dep. Var.: Log Changes in Non-Regular Emp.			
	(1)	(2)	(3)	(4)
AREER Shock	-2.62	-2.98		
	(0.23)	(0.27)		
AREER Shock $ imes$ Payroll Share		3.35		
,		(1.26)		
Observations	1,164,363	1,164,363	1,164,363	1,164,363
Covariates	√	√	√	✓
Year FEs	\checkmark	\checkmark	\checkmark	\checkmark
Establishment FEs	\checkmark	\checkmark	\checkmark	\checkmark

Table: Effects of JPY Appreciation on Employment Growth

	Dep. Var	Dep. Var.: Log Changes in Non-Regular Emp.		
	(1)	(2)	(3)	(4)
AREER Shock	-2.62	-2.98	-0.31	
	(0.23)	(0.27)	(0.44)	
AREER Shock \times Log Payroll			-1.08	
- '			(0.14)	
AREER Shock \times Payroll Share		3.35	8.26	
•		(1.26)	(1.41)	

Observations	1,164,363	1,164,363	1,164,363	1,164,363
Covariates	√	✓	✓	$\overline{\hspace{1cm}}$
Year FEs	\checkmark	\checkmark	\checkmark	\checkmark
Establishment FEs	✓	✓	✓	✓

Table: Effects of JPY Appreciation on Employment Growth

	Don Var	.: Log Change	s in Non Pos	ular Emp
	(1)	(2)	(3)	(4)
AREER Shock	-2.62	-2.98	-0.31	-0.55
	(0.23)	(0.27)	(0.44)	(0.44)
AREER Shock $ imes$ Log Payroll			-1.08	-1.12
			(0.14)	(0.15)
AREER Shock $ imes$ Payroll Share		3.35	8.26	
		(1.26)	(1.41)	
AREER Shock \times (Payroll Share $>$ 3%)				2.43
				(0.50)
Observations	1,164,363	1,164,363	1,164,363	1,164,363
Covariates	√	✓	✓	√
Year FEs	\checkmark	\checkmark	\checkmark	\checkmark
Establishment FEs	\checkmark	\checkmark	\checkmark	\checkmark

Quantitative Model of Granularity

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The Model Overview

- Small open economy
 - N regions $n \in \mathcal{N}$
 - continuum of sectors $j \in \mathcal{J}$
- Timing of the Model:
 - 1. Continuum of firms m_{nj} can pay a fixed cost ψ to attempt an entrance in sector j
 - 2. Random, finite number of firms enter E_{nj} (Poisson)
 - 3. Firms get an ex-ante productivity draw z_{nje} (Pareto)
 - 4. Workers decide where to live n, and how much to invest in sector-specific skills s_{nj}
 - 5. Firm ex-post productivity shocks revealed $a_{nje}(s)$ (Log-normal)
 - 6. Workers move labor across establishments and sectors subject to migration frictions



Equilibrium

- Firms:

- earn zero expected profits, conditional on trying to enter;
- maximize profits taking as given wages, conditional on entering.

- Workers:

- choose the utility-maximizing location;
- choose sector-specific skills to maximize expected utility;
- choose where to work to maximize utility.

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Intended Calibration

Description	Parameter	Value	Source
Short run labor elasticity across sectors	ν	0.42	Berger et al. (2022)
Short run labor elasticity across firms	κ	10.85	Berger et al. (2022)
Long run labor elasticity across sectors	$\overline{ u}$	1	Burstein et al. (2020)
Elasticity of production to labor	η	0.5	Labor Share (CoM)
Ex-ante firm prod. tail	$\dot{\lambda}$	2.8	Direct from Regression
Ex-post shock log variance	σ^2	0.25	Variance of log wages
Migration elasticity	heta	3	Redding (2016)
Congestion externality	$\gamma_{\it u}$	-0.25	Redding (2016)
Production externality	γ_z	0.0025	Combes et al. (2011)

Size of Externality

1. Agglomeration Externality: Elasticities of wages to the population

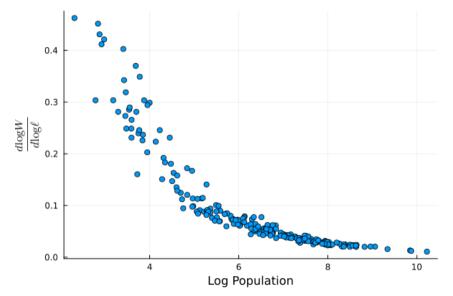
$$\frac{d \log W_n}{d \log \ell_n} := \frac{\gamma_z + \frac{\Psi'(m_n)m_n}{\Psi(m_n)} - (1-\eta)}{1 - \frac{\Psi'(m_n)m_n}{\Psi(m_n)}},$$

2. Firm Entry Wedge: The percentage difference between expected profits and the expected benefits on production.

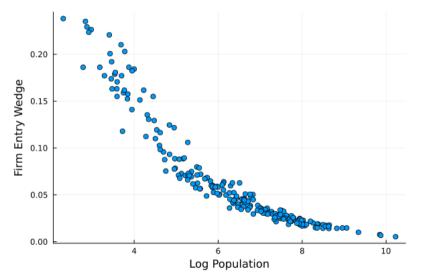
$$\frac{rac{\Pi_n}{m_n}}{\psi_n}-1:=rac{1-\eta}{rac{\Psi'(m_n)m_n}{\Psi(m_n)}}-1,$$

where $\Pi_n \equiv (1 - \eta) Y_n$ is the expected profit

Aggolomeration Externality is 0.4 in Small Locations



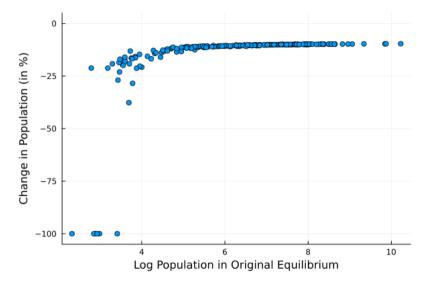
Firm in Small Locations Capture Less than 80% of Production Benefits



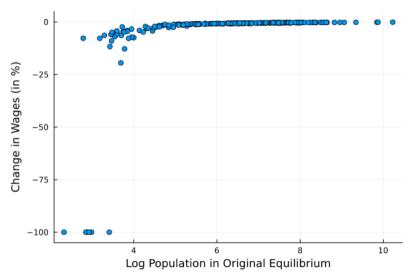
Counterfactual

- The Japanese working-age population is decreasing
 - NRPSSR: 87 million in 1995, 75 million in 2020, 70 million in 2032
- Simulate 10% drop in population (which is not a crazy scenario)
- See changes in
 - 1. Population
 - 2. Wages

Initially Smaller Locations Become Even Smaller



Initially Smaller Locations Hit Harder



Tokyo: 9.6% drop in population but 0.1% drop in wages (externality is small)

Conclusion

- Granularity is an important reason for agglomeration
- Standard economic geography models miss this and give incorrect counterfactual predictions because of it
 - Effects of Demographic Changes on Spatial Distribution
- Lots left to do!
 - How does granularity affect skill acquisition?
 - What is the optimal industrial mix?

Appendix

Imperfect Mobility Across Establishments and Labor Markets

- **Key Assumption:** easier to move across establishments within a labor market than moving across labor markets
- We show that this is the case
- We account for this in our quantitative model



Monopsony Power

- Another force for agglomeration
 - Firms would rather open in small labor markets
 - Workers would rather live in large labor markets
 - Workers "usually" win the tug of war since larger labor markets are more efficient
- Makes our mechanism stronger because distortions are especially bad for good shocks
- Variance of wages understates our mechanism



Labor Hoarding/Employer Insurance

- If firms have monopsony power, then they should
 - 1. Hold onto workers during bad years so they can have them when they need them
 - 2. Provide wage insurance for workers so wages represent "average" contribution
- Both cases strengthen our mechanism
 - In larger labor markets, monopsony power is lower, easier to find workers when you need them, less need for insurance
- Variance of wages understates gains

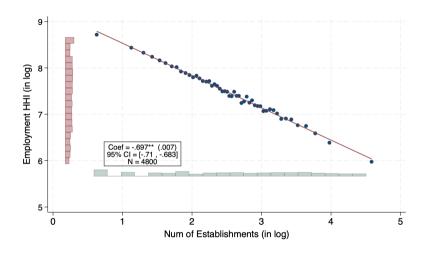


Wage Rigidity

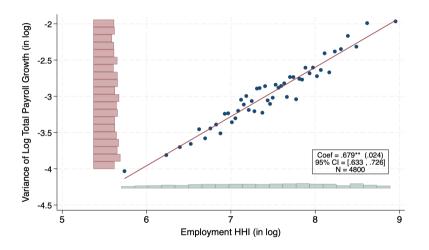
- In large labor markets, variance of marginal product is low
 - Wage rigidity rarely matters
- In small labor markets, will matter a lot!
- Even more inefficient because people become unemployed rather than underemployed
- Wage variance understates the mechanism.



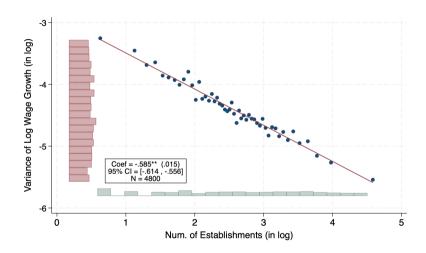
Number of Establishments and HHI



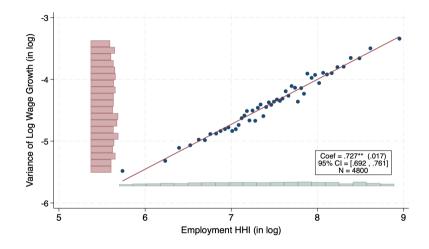
Fact 1: HHI and Volatility of LLM-level Payroll



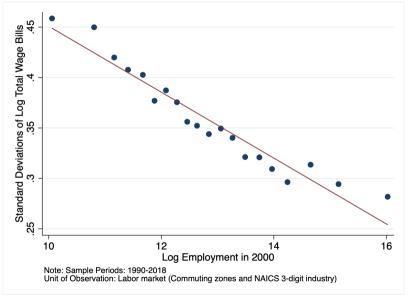
Fact 1: Volatility of LLM-level Wage



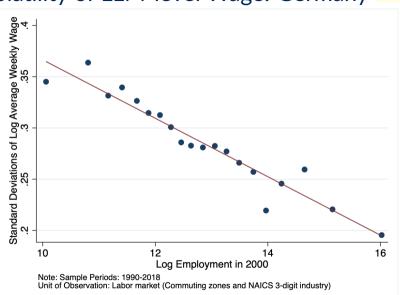
Fact 1: HHI and Volatility of LLM-level Wage



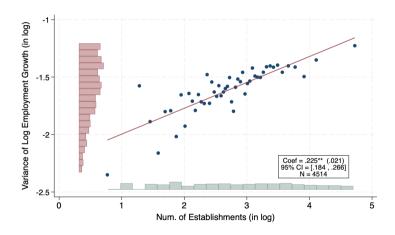
Fact 1: Volatility of LLM-level Payroll: Germany



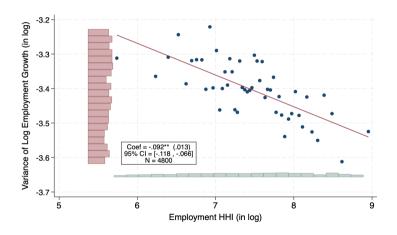
Fact 1: Volatility of LLM-level Wage: Germany •



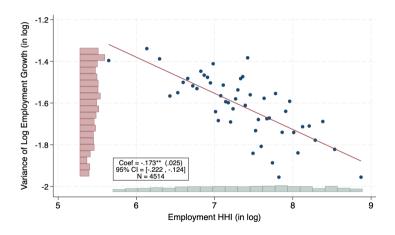
Fact 2:Volatility of Establishment-level Non-Regular Employment



Fact 2: HHI and Volatility of Establishment-level Employment

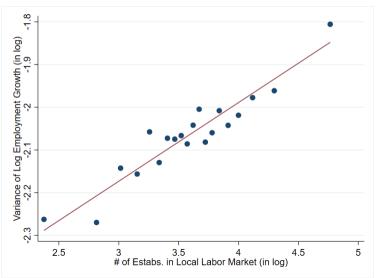


Fact 2: HHI and Volatility of Establishment-level Non-Regular Employment



Fact 2: Volatility of Establishment-level Employment:

Germany • Back



Workers - Location Choice

- Fundamental utility of location *n* is

$$U_n = u_n W_n$$

- Amenities are also subject to spillovers (congestion, $\gamma_u < 0$)

$$u_n = \overline{u}_n(\ell_n)^{\gamma_u}$$
.

- Workers have Fréchet utility shocks over the different locations

$$\ell_n = \left(\frac{U_n}{U}\right)^{\theta} \ell$$

where

$$U = \left[\sum_{n} (U_n)^{\theta}\right]^{\frac{1}{\theta}}$$

Workers - Ex-ante Skills Choice - Back

- Workers choose skill investments to maximize expected wages

$$\{s_{nj}\}_{j\in\mathcal{J}} \in \operatorname{argmax} \ W_n(\{s_j'\})$$
 s_j'
 $s.t. \quad 1 = \int_{\mathcal{J}} (s_j')^{\frac{1+\overline{\nu}}{\overline{\nu}-\nu}} dj$

- This takes as given number of firms in each sector and ex-ante productivity shocks z_{nie}
- ν is the short-run elasticity across sectors
- $\overline{\nu} > \nu$ is the long-run elasticity across sectors
- Denote solution by W_n

Workers - Ex-post Labor Choice

- After the shocks are revealed, workers maximize earnings, taking wages and skills as given

$$L_{nje}(s), L_{nj}(s) \in \operatorname*{argmax} \int_{\mathcal{J}} \left[\sum_{e \in \mathcal{E}_{nj}} w_{nje}(s) L'_{je} \right] dj$$

$$s.t. \quad L'_{j} = \left[\sum_{e \in \mathcal{E}_{nj}} b_{nje}^{-1/\kappa} (L'_{je})^{\frac{1+\kappa}{\kappa}} \right]^{\frac{\kappa}{1+\kappa}}$$

$$1 = \left[\int_{\mathcal{J}} s_{nj}^{-1/\nu} (L'_{j})^{\frac{1+\nu}{\nu}} \right]^{\frac{\nu}{1+\nu}}$$

- Denote solution by $W_n(\{s_{nj}\})$

Firms Back

Firms maximize profits by taking wages as given

$$\pi_{\textit{nje}}(s) = \max_{\ell'(s)} \;\; \textit{z}_{\textit{nje}} \textit{a}_{\textit{nje}}(s) \ell'(s)^{\eta_j} - \textit{w}_{\textit{nje}}(s) \ell'(s)$$

Free entry is

$$\psi = rac{1}{m_n} \mathbb{E}\left[\sum_{e}^{\mathcal{E}_{nj}} \pi_{nje}(s) \middle| m_{nj}
ight]$$

- Entry is Poisson

$$\mathbb{P}[E_{nj} = k] = \frac{(m_n)^k e^{-m_n}}{k!}$$

- Ex-ante shocks are distributed Pareto

$$egin{aligned} oldsymbol{z_{nie}} \sim \mathcal{P}(oldsymbol{z_{ni}}, \lambda); \quad oldsymbol{z_{ni}} = \overline{oldsymbol{z}}_{ni}(\ell_n)^{\gamma_{oldsymbol{z}}} \end{aligned}$$

Ex-post shocks are distributed log-normal

$$a_{ extit{nje}}(s) \sim \mathcal{LN}\left(-\sigma^2/2,\sigma^2
ight)$$