

ORGANIZATIONS, SKILLS, AND WAGE INEQUALITY

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WHAT IS IT ABOUT?

$$\text{Var}(w_{it}) = \underbrace{\mathbb{E} \text{Var}(w_{it}|z_{it})}_{\text{within}} + \underbrace{\text{Var} \mathbb{E}(w_{it}|z_{it})}_{\text{between}}$$

- ▶ A model with worker and firm heterogeneity accounting for wage dispersion
- ▶ Within- and between-group components (+ Δ from mid-1980s to late 2000s) driven by:
 - ▶ TFP distribution
 - ▶ Value added per worker
 - ▶ Labor market frictions
 - ▶ Education

OUTLINE

1. Reviewing the properties of the model
2. A few comments on the estimation
3. Comments on the results and experiments
 - 3.1 how to disentangle labor market frictions and technology?
 - 3.2 what does 'technological change' actually measure/capture?

THE MODEL

- ▶ Burdett & Mortensen [IER '98] with two-sided heterogeneity
- ▶ Workers differ by skill-type, indexed by $i = 1, \dots, N$
 - ▶ population measure $m(i)$
 - ▶ home production $b(i)$
 - ▶ friction parameters $\lambda(i)$ and $\delta(i)$
- ▶ Firms differ by TFP level x
 - ▶ x is drawn from distribution $\Gamma(\cdot)$
 - ▶ production function is $p(x, i)$

THE MODEL

Worker's problem:

$$rU(i) = b(i) + \lambda(i)f(i) \int_{R(i)} (W(w', i) - U(i)) dF_i(w')$$

and

$$rW(w, i) = w + \lambda(i)f(i) \int_w (W(w', i) - W(w, i)) dF_i(w') + \delta(i)(U(i) - W(w, i))$$

- ▶ $f(i) = 1 - \Gamma(\underline{x}(i))$, where $\underline{x}(i)$ is the lowest x hiring type- i workers
- ▶ $F_i(w)$ is the distribution of posted wage offers
- ▶ On-the-job search is as effective as off-the-job search, implying that

$$R(i) = b(i)$$

THE MODEL

- ▶ $\frac{\partial p}{\partial x}(x, i) \geq 0$; $\underline{x}(i)$ is pinned down by

$$p(\underline{x}(i), i) = b(i)$$

- ▶ $\frac{\partial p}{\partial i}(x, i) \geq 0$; the sign of $\frac{\partial \underline{x}}{\partial i}(i)$ is given by

$$b'(i) - \frac{\partial p}{\partial i}(\underline{x}(i), i)$$

Firm's problem:

$$\pi(x, i) = \max_w (p(x, i) - w) \ell(w; i)$$

Standard trade-off for wage posting

- ▶ higher $w \Rightarrow \downarrow$ profit per worker
- ▶ higher $w \Rightarrow \uparrow \ell(w; i)$ (\uparrow hires and retention)

THE MODEL

Theorem 1: Firms with higher x are bigger and

1. hire more at all i 's, and proportionally more at high i
2. hire i 's who are hired by lower- x firms
3. hire i 's that are beyond the reach of lower- x firms
4. pay higher w at all i 's

Propositions 2 and 3: If $i > j$, then $F_i(w)$ (resp. $G_i(w)$) dominates $F_j(w)$ (resp. $G_j(w)$)

- ▶ corollary: average wages (posted and observed) increase with i

ESTIMATION/CALIBRATION

- ▶ Assumption: the economy is at a steady state in 1985 and 2009

Reminder:

$$f(i) = 1 - \Gamma(\underline{x}(i)) \quad \text{and} \quad p(\underline{x}(i), i) = b(i)$$

- ▶ $m(i) = 5$ education levels
- ▶ $\delta(i) = \text{EU transition rate}$
- ▶ $\lambda(i)f(i)(1 - \delta(i)) = \text{UE transition rate}$
- ▶ $b(i) = \text{lowest wage by education}$

Note: no EE transition, no ‘reallocation shock’ (Jolivet, Postel-Vinay & Robin [EER '06])

ESTIMATION/CALIBRATION

1- Nonparametric approach

- ▶ follows Bontemps, Robin & van den Berg [IER '00]
- ▶ use wage distributions to recover the productivity distribution
- ▶ caveat: cannot separately estimate x and $p(x, i)$

2- Parametric approach

- ▶ suppose that $p(x, i) = x^\alpha A(i)^\beta$, with $\alpha + \beta = 1$ and $\beta \sim$ labor share
- ▶ fit the TFP distribution taken from an external source (İmrohoroğlu & Tuzel [MS '14])
- ▶ adjust $A(i)$ to match average wages by skill type

NONPARAMETRIC ESTIMATION

- ▶ Must impose some estimation restrictions:
 - ▶ single-peaked wage distribution for each type
however, it seems that this condition is not fulfilled at the lower end of the wage distribution among low-skill workers
 - ▶ all firms hire all skill levels, i.e. $\Gamma(\underline{x}(i)) = 0$ for all i
- ▶ Under this approach, they are able to recover the distribution of $p(x, i)$
 - ▶ in other words, the results cannot be related to a given firm

PARAMETRIC ESTIMATION

- ▶ External distribution of firm-level TFP is possibly biased towards high TFP firms
- ▶ Conditional on this TFP distribution, part of the results become more heavily driven by the imposed theoretical structure
 - ▶ in the Burdett-Mortensen model, high TFP firms compete little with low TFP firms
 - ▶ ... so this TFP distribution boosts the competition among high TFP firms
 - ▶ ... and high TFP firms in the model hire proportionally more at the top skill level

PARAMETRIC ESTIMATION

Table 8: Variance Decomposition

<i>Firm Decomposition</i>		
	Within-group	Between-group
1985	0.48	0.52
2009	0.60	0.40

<i>Skill Decomposition</i>		
	Within-group	Between-group
1985 - Model	0.29	0.71
1985 - Data	0.81	0.19
2009 - Model	0.21	0.79
2009 - Data	0.77	0.23

- ▶ Size of the within-firm component of wage dispersion? Lentz, Piyapromdee & Robin ['18] find very small firm effects (much smaller than AKM firm effects)
- ▶ 2/3 of the rise in the variance of (log) earnings occurred between firms (Song, Price, Guvenen, Bloom & von Wachter [QJE '19])

OTHER COMMENTS ON THE ESTIMATION

- ▶ The estimation draws little on the results from Theorem 1

- ▶ Suppose that

$$p(x, i) = p_0 + p_1x + p_2x^2 + p_3i + p_4i^2 + p_5x \cdot i$$

- ▶ To exploit this formula, you would need firm-level data moments, such as, e.g., moments on the joint distribution of firms' size and education composition of their workforce
- ▶ The new draft uses data from the Quarterly Workforce Indicators. Going in this direction?

RESULTS (NONPARAMETRIC APPROACH)

Table 5: St. Deviation Decomposition - within-group
(Values in \$1000 2009 USD)

	HS Dropout	HS Grads	Some College	College Grads	Post Grads
1985 Benchmark	1.10	1.29	1.59	2.52	3.67
2009 Benchmark	.98	1.49	1.76	4.33	7.64
Difference	-0.12	0.19	0.17	1.81	3.98
<i>Counterfactuals:</i>					
<i>Production</i>	234.25%	38.81%	118.26%	110.30%	117.06%
<i>Labor</i>	-154.47%	49.90%	-15.56%	-4.65%	-6.83%
<i>Education</i>	0.00%	0.00%	0.00%	0.00%	0.00%
<i>Production + Labor</i>	100%	100%	100%	100%	100%

RESULTS (PARAMETRIC APPROACH)

Table 10: St. Deviation Decomposition - within-group
(Calculations using values in \$1000 2009 USD)

	HS Dropout	HS Graduate	Some College	College	Post-Graduate
1985 - Benchmark	0.54	0.57	0.62	0.77	0.95
2009 - Benchmark	0.55	0.65	0.72	1.02	1.36
Difference	0.01	0.08	0.09	0.26	0.41
<i>Counterfactuals:</i>					
<i>TFP</i>	-75.76%	9.02%	28.62%	52.68%	69.13%
<i>Labor</i>	-42.73%	-5.53%	-2.42%	-0.02%	-0.09%
<i>Education</i>	0.00%	0.00%	0.00%	0.00%	0.00%
<i>Labor Productivity</i>	-387.74%	58.65%	55.99%	79.13%	86.92%
<i>Production Fcn</i>	-44.37%	0.16%	-0.00%	-0.85%	-2.34%

COUNTERFACTUAL EXERCISES

- ▶ ‘Labor market frictions helped to attenuate the impact of SBTC’
 - ▶ separate out the effects of higher $\lambda(i)$ vs. lower $\delta(i)$
 - ▶ ... and their impact on earnings through the competition among firms
- ▶ Technological change (i.e. changes in the distribution of $p(x, i)$)
 - ▶ flesh out the interpretation of technological change
 - ▶ does technological change affect labor market frictions?
- ▶ Changes in permanent vs. transitory components of wage earnings

POSSIBLE EXPLANATIONS

1. Increased sorting of high-wage workers across high-wage firms
Song, Price, Guvenen, Bloom & von Wachter [QJE '19]
2. Rising segregation of similar workers between firms
3. Changes in pay schemes (incidence of performance pay contracts)
Card, Cardoso, Heining & Kline [JoLE '18]

OTHER POSSIBLY RELEVANT ISSUES

- ▶ Lower end of the wage distribution: role of changes in the minimum wage
 - ▶ during the 1980s, this explains 1/3 of the change in overall residual wage inequality
DiNardo, Fortin & Lemieux [ECMA '96], Card & DiNardo [JoLE '02]
 - ▶ ideal theoretical framework to examine this issue

- ▶ Inequality at the top end of the wage distribution
 - ▶ a distinct model for large firms with 1000+ employees?
 - ▶ however, your data is not ideal to study this issue

CONCLUDING REMARKS

- ▶ This is really interesting and relevant work
 - ▶ a structural model showing the role of firms in explaining wage inequality
 - ▶ match key properties discussed in the empirical literature on organizations
- ▶ The authors take this structure to data over a long period
- ▶ Expect additional gains from using moments on firms' size, composition, and value added