## Are Managers Paid for Market Power?

<sup>1</sup> Princeton
<sup>2</sup> KU Leuven
<sup>3</sup> UPF Barcelona

#### SOLE

May 12, 2023

#### MOTIVATION

• Increase in income inequality – most in top percentiles (Piketty and Saez, 2003)

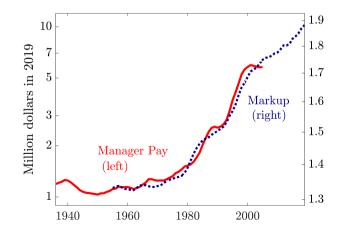
#### MOTIVATION

- Increase in income inequality most in top percentiles (Piketty and Saez, 2003)
- Manager pay: top earnings; rise since the late 1970s (Frydman and Saks, 2010)



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#### MANAGER PAY FROM <u>FIRM SIZE</u> AND <u>MARKET POWER</u>

- Decompose Manager Pay into two channels
  - 1. Firm Size: conventional wisdom
  - 2. Market Power: new mechanism outperform competitors

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|               |                  | log Manager Pay  |                  |                  |                  |                  |
|---------------|------------------|------------------|------------------|------------------|------------------|------------------|
|               | A.size = sales   |                  | B. size = COGS   |                  | C. size = employ |                  |
|               | (1)              | (2)              | (3)              | (4)              | (5)              | (6)              |
| log Firm Size | 0.438<br>(0.003) | 0.450<br>(0.003) | 0.387<br>(0.003) | 0.450<br>(0.003) | 0.391<br>(0.003) | 0.410<br>(0.003) |
| log Markup    |                  | 0.405<br>(0.015) |                  | 0.854<br>(0.016) |                  | 0.472<br>(0.016) |
| R-squared     | 0.551            | 0.565            | 0.502            | 0.565            | 0.489            | 0.508            |
| Observations  | 32,930           | 31,982           | 32,930           | 31,982           | 32,930           | 31,982           |

### MANAGER PAY FROM <u>FIRM SIZE</u> AND <u>MARKET POWER</u>

- Decompose Manager Pay into two channels
  - 1. Firm Size: conventional wisdom
  - 2. Market Power: new mechanism outperform competitors
- Method
  - 1. Structural model: a combination of
    - Competitive matching market (Gabaix and Landier, 2008; Terviö, 2008)
    - Oligopolistic competition (Atkeson and Burstein, 2008)  $\Rightarrow$  compete for market power
  - 2. Estimation
    - Technology: productivity and complementarity
    - Market structure: the number of firms competing with each other

### Contribution of <u>Firm Size</u> and <u>Market Power</u>

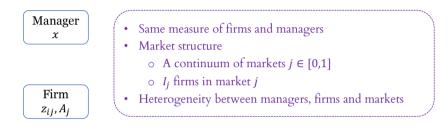
- The rise of Manager Pay:
  - 1. On average, Market Power 45.8% vs. Firm Size 54.2%
  - 2. Over time, market power contributes from 38.0% (1994) to 48.8% (2019)
    - accounts for 57.8% of increase in Pay

### Contribution of <u>Firm Size</u> and <u>Market Power</u>

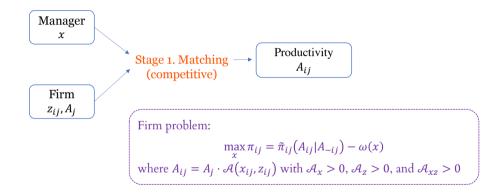
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    accounts for 57.8% of increase in Pay
- Cross-section of managers: *heterogeneity* 
  - Low-ability managers: Firm Size channel dominates  $\approx 100\%$
  - Top-ability managers: Market Power channel dominates 80.3%

# Model

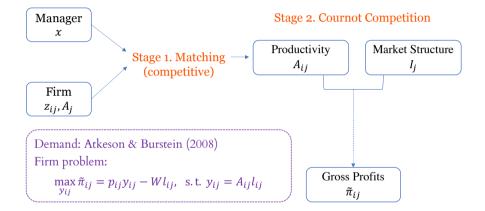
### A Sketch of Setup – Primitives



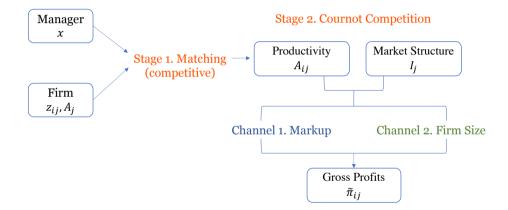
#### A Sketch of Setup - Manager Market



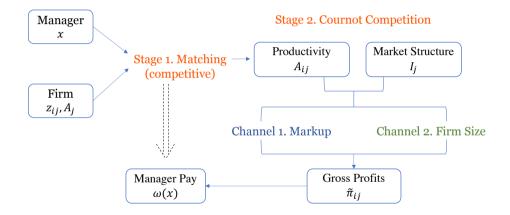
#### A Sketch of Setup - Output Market



#### A Sketch of Setup - Output Market



#### A Sketch of Setup - Determinants of Manager Pay



#### Manager Pay - First Order Conditions

#### MANAGER PAY - FIRST ORDER CONDITIONS

Stage 2:
$$p_{ij} \underbrace{\left(1 + \varepsilon_{ij}^{P}\right)}_{\mu_{ij}^{-1}} = W/A_{ij} \quad \Leftrightarrow \quad \tilde{\pi}_{ij} = (\mu_{ij} - 1)Wl_{ij}$$
Stage 1: $\max_{x} \pi_{ij} = \tilde{\pi}_{ij} - \omega(x) \quad \Rightarrow \quad \frac{\partial \tilde{\pi}_{ij}}{\partial A_{ij}} \frac{\partial A_{ij}}{\partial x_{ij}} = \frac{d}{dx} \omega(x_{ij})$ 

• Managers contribute in two channels: • Elasticity

$$\tilde{\pi}_{ij} = (\mu_{ij} - 1)Wl_{ij} \quad \Rightarrow \quad \frac{\partial \tilde{\pi}_{ij}}{\partial A_{ij}} = \underbrace{\frac{\partial \mu_{ij}}{\partial A_{ij}}Wl_{ij}}_{\text{Market power}} + \underbrace{(\mu_{ij} - 1)W\frac{\partial l_{ij}}{\partial A_{ij}}}_{\text{Firm size}}$$

#### MANAGER PAY - FIRST ORDER CONDITIONS

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• Managers contribute in two channels: • Elasticity

$$\begin{split} \tilde{\pi}_{ij} &= (\mu_{ij} - 1)Wl_{ij} \quad \Rightarrow \quad \frac{\partial \tilde{\pi}_{ij}}{\partial A_{ij}} = \underbrace{\frac{\partial \mu_{ij}}{\partial A_{ij}}Wl_{ij}}_{\text{Market power}} + \underbrace{(\mu_{ij} - 1)W\frac{\partial l_{ij}}{\partial A_{ij}}}_{\text{Firm size}} \\ \Rightarrow \quad \omega(x_{ij}) &= \omega_0 + \int_{\underline{x}}^{x_{ij}} \left[ \frac{\partial \mu_{i'j'}}{\partial A_{i'j'}}Wl_{i'j'} + \left(\mu_{i'j'} - 1\right)W\frac{\partial l_{i'j'}}{\partial A_{i'j'}} \right] \times \left[ \frac{\partial A_{i'j'}}{\partial x_{ij}} \right] dx_{i'j'} \end{split}$$

#### MATCHING - Algorithm with Externality

• Complementarity  $\Rightarrow$  PAM between managers and firms...

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- Complementarity  $\Rightarrow$  PAM between managers and firms...
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#### MATCHING - Algorithm with Externality

- Complementarity  $\Rightarrow$  PAM between managers and firms...
- But, externality from competition
  - Productivity is not the correct criterion for firm ranking
  - Impossible to find the exact matching with a large number of firms
- Approximate stable matching: find a proxy for firms' profitability with externality
  - 1. Compute  $\partial \tilde{\pi}_{ij} / \partial x_{ij}|_{\overline{x}}$  by assigning all firms the *average* manager
  - 2. Construct PAM allocation between the manager types x and  $\partial \tilde{\pi}_{ij} / \partial x_{ij}|_{\overline{x}}$

#### ▶ Efficiency

## Quantitative Exercise

#### Assumptions & Parametrization

- Simulated Methods of Moments year by year
- Market structure:  $I_j \sim \mathcal{N}(m_I, \sigma_I)$  and  $I_j \in \{1, 2, ...\}$
- Types  $\{x_{ij}, z_{ij}, A_j\}$ : independently drawn from lognormal distribution
- TFP CES form:

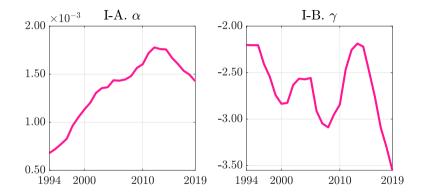
$$A_{ij} = A_j \left[ \frac{\alpha x_{ij}^{\gamma} + (1 - \alpha) z_{ij}^{\gamma} \right]^{\frac{1}{\gamma}},$$

 $\Rightarrow$  flexibility of CES setup

#### TARGETED MOMENTS - MANAGER RELATED PARAMETERS

|           |                               |                  | Key Paramet                 | er Meaning                                                                                             |
|-----------|-------------------------------|------------------|-----------------------------|--------------------------------------------------------------------------------------------------------|
| I. Match  | Average salary share          | 1                | $\alpha$                    | $A_{ij} = A_j \left[ \alpha x_{ij}^{\gamma} + (1 - \alpha) z_{ij}^{\gamma} \right]^{\frac{1}{\gamma}}$ |
|           | Sales elasticity of sala      | ary share        | $\gamma$                    | $\Pi_{ij} = \Pi_j \left[ \alpha \omega_{ij} + (1 - \alpha) \omega_{ij} \right]$                        |
|           | A. Data 2019                  | в                | Model: $\alpha$             | C. Model: $\gamma$                                                                                     |
| (Joo)     | -2                            | -2<br>-2         | $\alpha = 2 \times 10^{-3}$ |                                                                                                        |
|           |                               | -4               |                             | $-4$ $\gamma = -4$                                                                                     |
| anala<br> |                               | -6               |                             | -6                                                                                                     |
| sh<br>sh  | -8                            | -8               |                             | -8                                                                                                     |
| Salarw    | -10                           | -10 $\alpha = 6$ | $\times 10^{-4}$            | $\gamma = -2$                                                                                          |
|           | $16 \ 18 \ 20 \ 22 \ 24 \ 26$ |                  | 20 $22$ $24$ $26$           | $16 \ 18 \ 20 \ 22 \ 24 \ 26$                                                                          |
|           | Sales $r_{ij}$ (log)          | S                | Sales (log)                 | Sales (log)                                                                                            |

#### **ESTIMATION - MANAGER RELATED PARAMETERS**



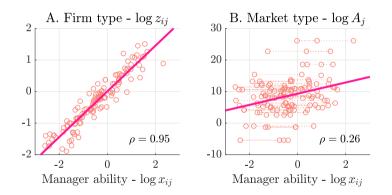
#### **ESTIMATION - OTHER PARAMETERS**

- Other parameters are consistent with the literature Identification
  - Increasingly concentrated market structure
  - Higher heterogeneity across firms

# Main Results

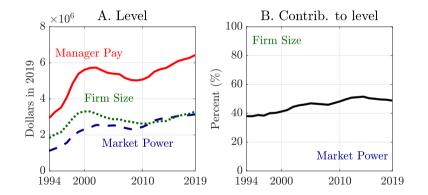
#### MATCHING CORRELATION

Estimated Economy (2019)



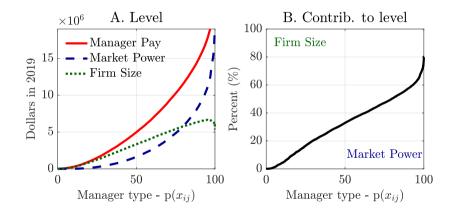
#### MARKET POWER VS. FIRM SIZE

Time Series



#### MARKET POWER VS. FIRM SIZE

Crosssectional Heterogeneity (2019)







Top managers are hired by firms with market power



Top managers are hired by firms with market power And they get rewarded for it

#### Conclusion

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#### CONCLUSION

Top managers are hired by firms with market power And they get rewarded for it Increasingly so

- Market Power contributes 45.8% to Manager Pay, from 38.0% (1994) to 48.8% (2019)
- Heterogeneity: the bottom (all via Firm Size) and the top (80.3% via Market Power)
- A general story for all managers and superstar workers (coders, athletes,...)

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Appendix

#### **ROBUSTNESS & ADDITIONAL EXERCISES**

- Elasticity of productivity Elasticity
- Cournot vs. Bertrand Bertrand
- An alternative decomposition: interpreting revenue as firm size Revenue

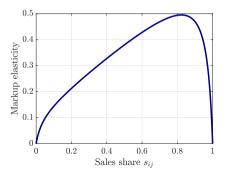
#### MARKET POWER VS. FIRM SIZE

Markup Elasticity of Productivity

$$\varepsilon_{ij}^{\mu} \coloneqq \frac{\partial \mu_{ij}}{\partial A_{ij}} \frac{A_{ij}}{\mu_{ij}} = \underbrace{\left[\frac{(\eta - 1)\left(1 - \phi_{ij}\right)}{1 + (\eta - 1)\left(\frac{1}{\theta} - \frac{1}{\eta}\right)\mu_{ij}s_{ij}}\right]}_{\frac{\partial s_{ij}}{\partial A_{ij}} \frac{A_{ij}}{s_{ij}}, \, \downarrow \, \text{in} \, s_{ij}} \times \underbrace{\left[\left(\frac{1}{\theta} - \frac{1}{\eta}\right)\mu_{ij}s_{ij}\right]}_{\frac{d\mu_{ij}}{ds_{ij}} \frac{s_{ij}}{\mu_{ij}}, \, \uparrow \, \text{in} \, s_{ij}} \in [0, 1)$$

- $\phi_{ij}$  is a weight for firm's importance
- First increase with  $s_{ij}$ , then decreases, where

$$\lim_{s_{ij}\to 0}\varepsilon^{\mu}_{ij}=\lim_{s_{ij}\to 1}\varepsilon^{\mu}_{ij}=0$$



#### MARKET POWER VS. FIRM SIZE

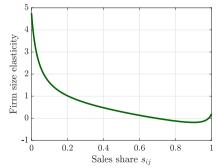
Firm Size Elasticity of Productivity

$$\varepsilon_{ij}^{l} \coloneqq \frac{\partial l_{ij}}{\partial A_{ij}} \frac{A_{ij}}{l_{ij}} = \phi_{ij} \underbrace{\left[ \theta - 1 \right]}_{\text{Monopoly}} + \left(1 - \phi_{ij}\right) \underbrace{\left[ \frac{\eta}{1 + \left(\frac{1}{\theta} - \frac{1}{\eta}\right) \left(\eta - 1\right) \mu_{ij} s_{ij}} - 1 \right]}_{\text{Strategic interaction, \downarrow in } A_{ij}},$$

- +  $\varepsilon_{ij}^l$  can be negative when  $s_{ij}$  is moderately large
- First decreases with  $s_{ij}$ , then increases, with

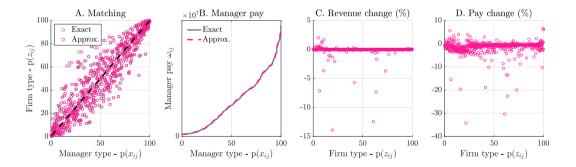
$$\lim_{s_{ij}\to 0}\varepsilon_{ij}^l=\eta-1>0\quad,\quad \lim_{s_{ij}\to 1}\varepsilon_{ij}^l=\theta-1>0$$

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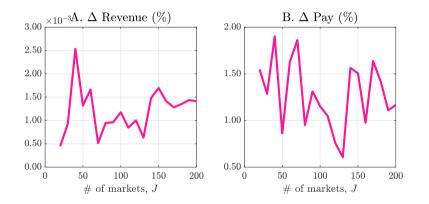


#### Efficiency: Matching Algorithm

- An example with J = 200
  - The average revenue difference is 0.001%
  - The average manager pay difference is 1.17%



#### Efficiency: Matching Algorithm

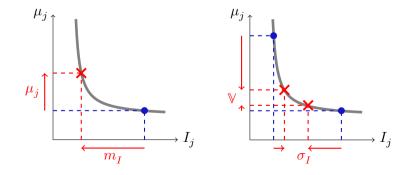


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### TARGETED MOMENTS

II. Market

|            |                                             | Moment(s)                                                                    | Key Parameter(s) |
|------------|---------------------------------------------|------------------------------------------------------------------------------|------------------|
| II. Market | Average markup<br>Variance markup (between) | $ \begin{split} \mathbb{E}(\mu_{ij}) \\ \mathbb{V}(\log \mu_j) \end{split} $ | $m_I \ \sigma_I$ |



### TARGETED MOMENTS

III. Firm

|                                                                                                                                                                |                                                                     | Moment(s)                                                                                              | Key Parameter(s)            |
|----------------------------------------------------------------------------------------------------------------------------------------------------------------|---------------------------------------------------------------------|--------------------------------------------------------------------------------------------------------|-----------------------------|
| III. Firm                                                                                                                                                      | Variance markup (within)<br>Average worker's wage<br>Variance sales | $\begin{array}{l} \mathbb{V}(\log \mu_{ij} j) \\ \mathbb{E}(W) \\ \mathbb{V}(\log r_{ij}) \end{array}$ | $\sigma_z \ m_A \ \sigma_A$ |
| $ \begin{array}{c} \mu_{ij} \\ \downarrow \\ $ | W<br>W<br>W<br>w<br>$\sigma_z$                                      | $\begin{array}{c} (MRPL) \uparrow \\ S \\ D_2 \\ D_1 \\ \hline \\ L \end{array}$                       | $r_{j}$                     |

#### TARGETED MOMENTS

IV. Aggregates

|                |                                                                    | Moment(s)                                                                                         | Key Parameter(s)                          |
|----------------|--------------------------------------------------------------------|---------------------------------------------------------------------------------------------------|-------------------------------------------|
| IV. Aggregates | Average employment<br>Average manager salary<br>Manager salary, p1 | $ \begin{split} & \mathbb{E}(l_{ij}) \\ & \mathbb{E}_x(\omega(x)) \\ & \omega(x p1) \end{split} $ | $rac{\overline{arphi}}{\psi} \ \omega_0$ |

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