

# WHAT DRIVES WAGE STAGNATION: MONOPSONY OR MONOPOLY?

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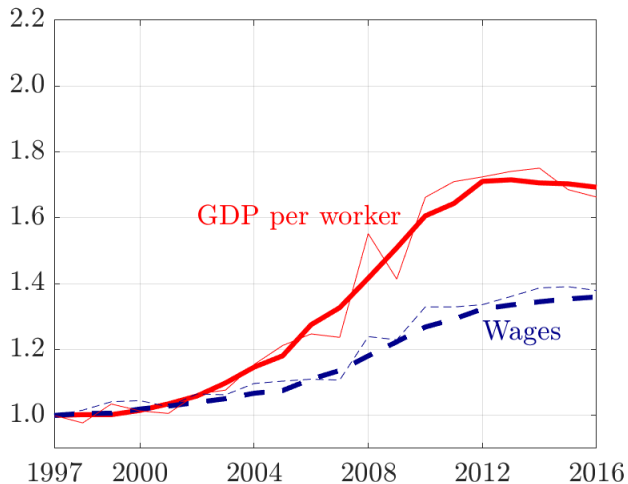
JEEA Supplemental Slides

2022

*Any opinions and conclusions expressed herein are those of the authors and do not represent the views of the U.S. Census Bureau. All results have been reviewed to ensure that no confidential information is disclosed. Data Management System (DMS) number: P-7083300, Subproject number: 7508369. Disclosure Review Board number: CBDRB-FY22-CED006-0027.*

# Wage Stagnation

U.S. Census : Tradeable sectors



# Mechanisms

- Explore two mechanisms behind wage stagnation:
  1. **Monopsony**: direct effect from imperfect labor market
    - Lower firm-specific wages for own workers
  2. **Monopoly**: output market power affects labor demand – **General Equilibrium** effect
    - Lowers aggregate, economy-wide wages

# Mechanisms

- Explore two mechanisms behind wage stagnation:
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  2. **Monopoly**: output market power affects labor demand – **General Equilibrium** effect
    - Lowers aggregate, economy-wide wages
- ∴ Objective:
  1. Explain mechanism behind **decoupling of wages and productivity**
  2. **Decomposition**: measure contribution from Monopsony (markdowns) vs. Monopoly (markups)

# Motivation

- Evidence on market power:
  1. Monopoly power (markups)  
De Loecker, Eeckhout, Unger (2020); Hall (2018)
  2. Monopsony power: (markdowns)  
Berger, Herkenhoff, Mongey (2020); Hershbein, Macaluso, Yeh (2018)

# Motivation

- Evidence on market power:
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  2. Monopsony power: (markdowns)  
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- Challenges for measurement:
  1. Marginal cost not directly observable
  2. We don't observe who competes with whom
- Our approach: structurally estimate Strategic Competition in GE:
  1. Jointly Measure Markups and Markdowns
  2. Estimate Market Structure

# Findings

1. Competition has decreased over time:
  - Markups increase substantially
  - Markdowns are stable, increase only marginally
2. Wage stagnation: decoupling wages-productivity
3. Decomposition monopoly vs. monopsony: dominant force is monopoly

# Model Setup

## MARKETS

- Continuum of markets  $j \in [0, J]$
- Finite number of establishments  $i = 1, \dots, I$
- Finite numbers of firms in each market  $n = 1, \dots, N$  (set of establishments  $i$  in firm  $n$ :  $\mathcal{I}_{nj}$ )

## HOUSEHOLD PREFERENCES

- maximizes static utility

$$\max_{C_{inj}, L_{inj}} U \left( C - \frac{1}{\phi^{\frac{1}{\phi}}} \frac{L^{\frac{\phi+1}{\phi}}}{\frac{\phi+1}{\phi}} \right) \quad \text{s.t. } PC = LW + \Pi$$

- CES preferences over Consumption and Labor

$$C = \left( \int_j J^{-\frac{1}{\theta}} C_j^{\frac{\theta-1}{\theta}} dj \right)^{\frac{\theta}{\theta-1}}, \quad C_j = \left( \sum_i I^{-\frac{1}{\eta}} C_{inj}^{\frac{\eta-1}{\eta}} \right)^{\frac{\eta}{\eta-1}}$$
$$L = \left( \int_j J^{\frac{1}{\hat{\theta}}} L_j^{\frac{\hat{\theta}+1}{\hat{\theta}}} dj \right)^{\frac{\hat{\theta}}{\hat{\theta}+1}}, \quad L_j = \left( \sum_i I^{\frac{1}{\hat{\eta}}} L_{inj}^{\frac{\hat{\eta}+1}{\hat{\eta}}} \right)^{\frac{\hat{\eta}}{\hat{\eta}+1}}$$



# Model Setup

## TECHNOLOGY

Firm  $n \in \{1, \dots, N\}$  in sector  $j \in [0, J]$

$$\Pi_{nj} = \max_{\{Y_{inj}\}_{i \in \mathcal{I}_{nj}}} \sum_{i \in \mathcal{I}_{nj}} \left[ \underbrace{P_{inj}(Y_{inj}, Y_{-inj})Y_{inj}}_{\text{Sales}} - \underbrace{W_{inj}(L_{inj}, L_{-inj})L_{inj}}_{\text{Variable costs}} \right]$$

subject to

$$Y_{inj} = A_{inj}L_{inj}$$

## MARKET STRUCTURE

The same set of  $N$  firms compete in goods and labor market

## PRICES AND EQUILIBRIUM

Cournot-Nash Competition in goods markets and labor markets

# Equilibrium Solution

## Producer Optimality

- The firm's first order condition for establishment  $i$  can be written as:

$$\underbrace{P_{inj} \left(1 + \varepsilon_{inj}^P\right)}_{\mu_{inj}^{-1}} A_{inj} = W_{inj} \underbrace{\left(1 + \varepsilon_{inj}^W\right)}_{\delta_{inj}}$$

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- Markups and Markdowns

$$\begin{aligned} \mu_{inj} &= \frac{P_{inj}}{MC_{inj}} = \frac{1}{1 + \varepsilon_{inj}^P}; & \varepsilon_{inj}^P &= - \left[ \frac{1}{\theta} s_{nj} + \frac{1}{\eta} (1 - s_{nj}) \right] \\ \delta_{inj} &= \frac{MRPL_{inj}}{W_{inj}} = 1 + \varepsilon_{inj}^W; & \varepsilon_{inj}^W &= \left[ \frac{1}{\hat{\theta}} e_{nj} + \frac{1}{\hat{\eta}} (1 - e_{nj}) \right] \end{aligned}$$

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- Mechanism

$$P_{inj} A_{inj} \times \mu_{inj}^{-1} = W_{inj} \times \delta_{inj} \Rightarrow \underbrace{W_{inj}}_{\text{Wage}} = \underbrace{\frac{R_{inj}}{L_{inj}}}_{\text{Rev/worker}} \times \underbrace{\mu_{inj}^{-1}}_{\text{Markup}} \times \underbrace{\delta_{inj}^{-1}}_{\text{Markdown}}$$

## Limit Cases

### NO HETEROGENEITY IN PRODUCTIVITY

$$\mu_{inj} = \left( 1 - \frac{1}{\theta} \frac{1}{N} - \frac{1}{\eta} \left( 1 - \frac{1}{N} \right) \right)^{-1}$$

$$\delta_{inj} = \left( 1 + \frac{1}{\hat{\theta}} \frac{1}{N} + \frac{1}{\hat{\eta}} \left( 1 - \frac{1}{N} \right) \right)$$

## Limit Cases

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### INCREASING COMPETITION

1. Perfect Substitutability :  $\eta, \theta, \hat{\eta}, \hat{\theta} \rightarrow \infty$

$$\mu_{inj} = 1 ; \delta_{inj} = 1$$

# Limit Cases

## NO HETEROGENEITY IN PRODUCTIVITY

$$\mu_{inj} = \left( 1 - \frac{1}{\theta} \frac{1}{N} - \frac{1}{\eta} \left( 1 - \frac{1}{N} \right) \right)^{-1}$$
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## INCREASING COMPETITION

1. Perfect Substitutability :  $\eta, \theta, \hat{\eta}, \hat{\theta} \rightarrow \infty$

$$\mu_{inj} = 1 ; \delta_{inj} = 1$$

2. Number of firms in a market :  $N \rightarrow \infty \implies s_{inj} \rightarrow 0$

$$\mu_{inj} = \frac{\eta}{\eta - 1} ; \delta_{inj} = \frac{\hat{\eta} + 1}{\hat{\eta}}$$

## MONOPOLISTIC COMPETITION ( $N = 1$ )

$$\mu_{inj} = \frac{\theta}{\theta - 1} ; \delta_{inj} = \frac{\hat{\theta} + 1}{\hat{\theta}}$$

# Quantitative Exercise

- U.S. Census Bureau Longitudinal Business Database (LBD): Tradeable Sectors
- In the data we observe
  1. Employment by establishment:  $L_{inj}$
  2. Average Wages by establishment:  $W_{inj} = \frac{\text{Wage Bill}_{inj}}{L_{inj}}$
  3. Revenue:  $R_{inj}$
  4. Industry classification NAICS, SIC



# Quantitative Exercise

## Estimation

1. Market Assignment: Randomly assign  $I_j$  establishments from NAICS 6 industry into a market, and  $I_j$  establishments into  $N$  “firms” of size  $I_j/N$ .
2. Elasticity Estimation: Estimate labor elasticities  $\hat{\eta}$  and  $\hat{\theta}$  using labor supply equation.
3. Backout Productivity: Given establishment-level employment  $L_{inj}$ , market assignment, and parameters, back out technology  $A_{inj}$  for each establishment, for each  $N$ .
4. Estimate Market Structure: Given a distribution of  $A_{inj}$  for each  $N$ , pick  $N$  which minimizes distance between model and data to find the market structure for each year.

## Preference Estimates and Parameters

| Variable       | Value                   |  | Source             |
|----------------|-------------------------|--|--------------------|
| $\hat{\theta}$ | 1.89                    | Input market: Between-market elasticity  | estimated          |
| $\hat{\eta}$   | 3.10                    | Input market: Within market elasticity   | estimated          |
| $\theta$       | 1.2                     | Output market: Between-market elasticity | DLEM (2021)        |
| $\eta$         | 5.75                    | Output market: Within market elasticity  | DLEM (2021)        |
| $\phi$         | 0.25                    | Elast. Aggregate LS                      | Chetty e.a. (2011) |
| $I$            | 32                      | Establishments in each market            | Externally set     |
| $N$            | $\in [2, 4, 8, 16, 32]$ | Firms competing within each market       | estimated          |

# Estimating Labor Elasticities

## Estimating Within and Between Market Substitutability

Modify the baseline labor supply equation along two dimensions to estimate  $\hat{\eta}$  and  $\hat{\theta}$ :

1. Exploit the panel dimension of the micro-data
2. Account for measurement error in wages:  $W_{inj}^* = W_{inj} \times \exp(\varepsilon_{inj})$

$$\ln W_{inj}^* = c_{jt} + \gamma \ln L_{jt} + \beta \ln L_{inj} + \underbrace{\alpha_{inj} + \epsilon_{inj}}_{\varepsilon_{inj}}$$

where we define  $\beta = \frac{1}{\hat{\eta}}$  and  $\gamma = (\frac{1}{\hat{\theta}} - \beta)$

Use Two-Stage Least Squares to estimate  $\beta$  and  $\gamma$ , sequentially.

Rely on Berger, Herkenhoff and Mongey (2021) and Giroud and Rauh (2019)

- Exploit time-series variation in state corporate taxes as instruments for employment

# Labor Elasticities Estimates

Exogenous variation from tax differences over time

| Parameter      | Description               | Estimate<br>IV |
|----------------|---------------------------|----------------|
| $\hat{\eta}$   | Within-market elasticity  | 3.10           |
| $\hat{\theta}$ | Between-market elasticity | 1.89           |

First Stage

Wage Distribution

## Backing out $\{A_{inj}, \mu_{inj}, \delta_{inj}\}$

- For given market structure (N) and preferences  $\{\eta, \theta, \hat{\eta}, \hat{\theta}\}$ , using data on  $\{L_{inj}\}$  we can recover  $\{A_{inj}, \mu_{inj}, \delta_{inj}\}$ .
- System of  $I$  equations and  $I$  unknowns for all establishments  $i$  in each market  $j$

$$P_{inj} \underbrace{\left(1 + \varepsilon_{inj}^P\right)}_{\mu_{inj}^{-1}} A_{inj} = W_{inj} \underbrace{\left(1 + \varepsilon_{inj}^W\right)}_{\delta_{inj}}$$

## Backing out $\{A_{inj}, \mu_{inj}, \delta_{inj}\}$

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- System of  $I$  equations and  $I$  unknowns for all establishments  $i$  in each market  $j$

$$\begin{aligned}
 & \frac{1}{J} \frac{1}{I} \frac{1}{\theta} \frac{1}{\eta} (A_{inj} L_{inj})^{\frac{1}{\eta}} \left[ \left( \frac{1}{I} \sum_i (A_{inj} L_{inj})^{\frac{\eta-1}{\eta}} \right)^{\frac{\theta-\eta}{(\eta-1)\theta}} \right] \underbrace{\left[ 1 - \frac{1}{\theta} \frac{\sum_{i \in \mathcal{I}_{nj}} (A_{inj} L_{inj})^{\frac{\eta-1}{\eta}}}{\sum_i (A_{inj} L_{inj})^{\frac{\eta-1}{\eta}}} - \frac{1}{\eta} \left[ 1 - \frac{\sum_{i \in \mathcal{I}_{nj}} (A_{inj} L_{inj})^{\frac{\eta-1}{\eta}}}{\sum_i (A_{inj} L_{inj})^{\frac{\eta-1}{\eta}}} \right] \right]}_{\text{Inverse Markup: } \mu_{inj}^{-1}} \\
 &= \frac{1}{Z} \frac{1}{J} \frac{1}{I} \frac{1}{\theta} \frac{1}{\eta} \frac{(L_{inj})^{\frac{1}{\eta}}}{A_{inj}} \left[ \left( \frac{1}{I} \sum_i (L_{inj})^{\frac{\hat{\eta}+1}{\hat{\eta}}} \right)^{\frac{\hat{\eta}-\hat{\theta}}{(\hat{\eta}+1)\hat{\theta}}} \right] \underbrace{\left[ 1 + \frac{1}{\hat{\theta}} \frac{\sum_{i \in \mathcal{I}_{nj}} (L_{inj})^{\frac{\hat{\eta}+1}{\hat{\eta}}}}{\sum_i (L_{inj})^{\frac{\hat{\eta}+1}{\hat{\eta}}}} + \frac{1}{\hat{\eta}} \left[ 1 - \frac{\sum_{i \in \mathcal{I}_{nj}} (L_{inj})^{\frac{\hat{\eta}+1}{\hat{\eta}}}}{\sum_i (L_{inj})^{\frac{\hat{\eta}+1}{\hat{\eta}}}} \right] \right]}_{\text{Markdown: } \delta_{inj}}
 \end{aligned}$$

where  $Z = W^{-1} L^{\frac{1}{\theta}} Y^{\frac{1}{\theta}}$  and the aggregate price  $P$  is normalized to 1.

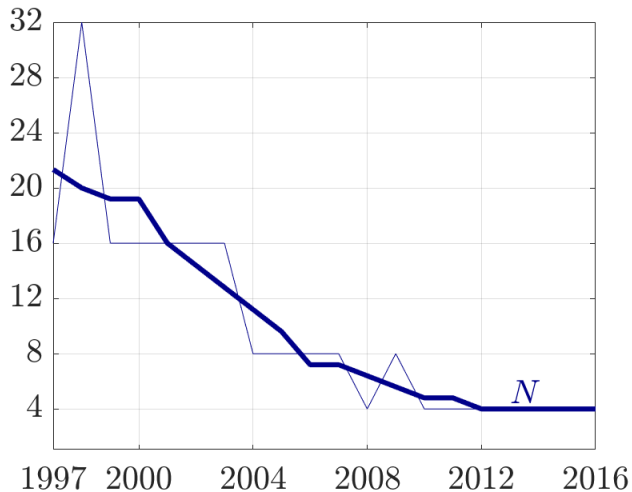
## N Estimation

$$N_t^* = \min_{N \in \{2, 4, 8, 16, 32\}} \left[ \int_j \sum_i m_{inj}^D \psi_{inj}^D dj - \int_j \sum_i m_{inj}^M(N) \psi_{inj}^M(N) dj \right]^2$$

Where  $m_{inj}^{D,M} = \frac{R_{inj}}{\int_j \sum_i R_{inj} dj}$  is the sales share of establishment  $i$  in the Data and Model

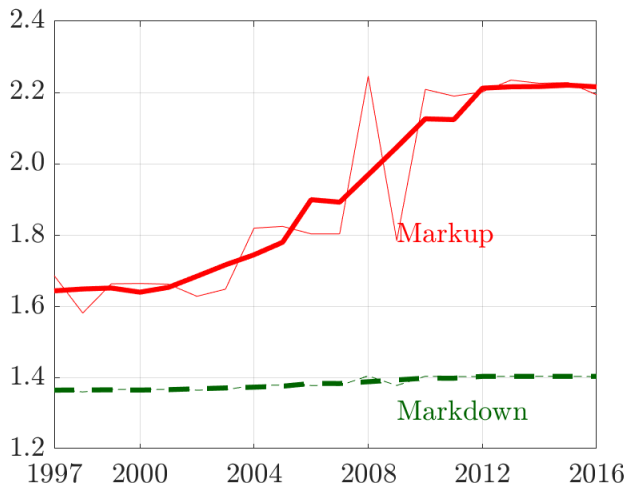
And  $\psi_{inj}^{D,M}$  is the revenue over wage bill for establishment  $i$  in the Data and Model

## Estimated $N$

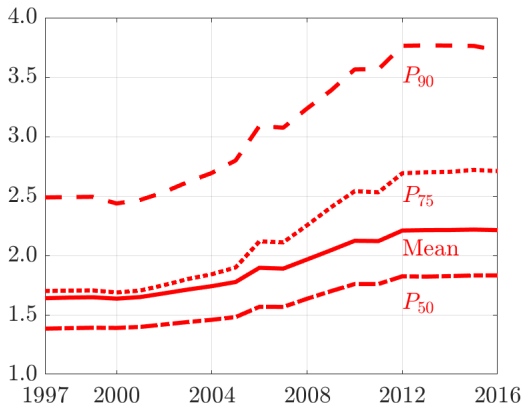




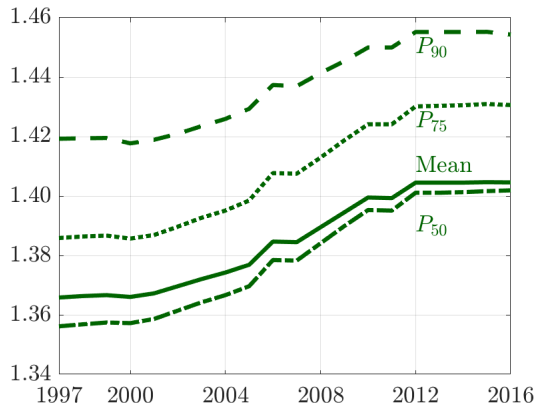
## Average Markups and Markdowns



# Markup and Markdown Distributions



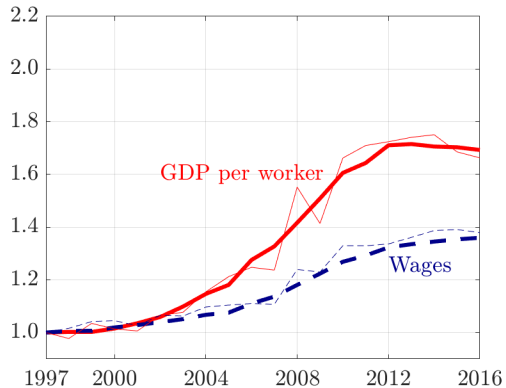
(a) Markups



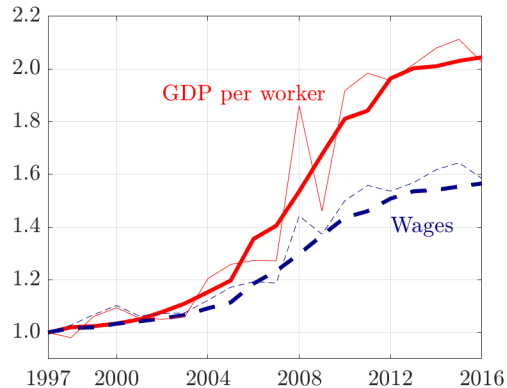
(b) Markdowns

# Wage Stagnation

## Decoupling Wages-Productivity



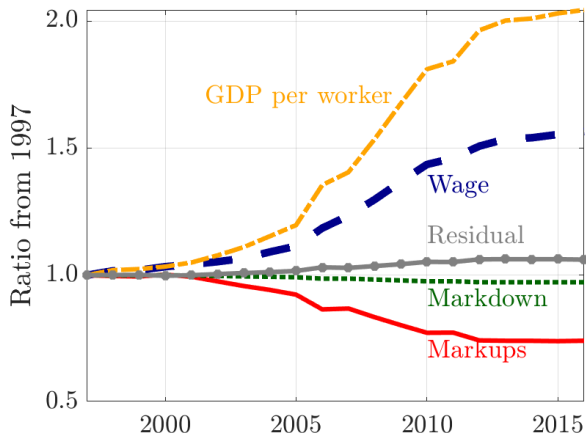
(a) Data



(b) Model

# Decoupling Wages-Productivity

$$W = \text{GDP/Worker} \times \mu^{-1} \times \delta^{-1} \times \Omega$$



## Social Planner's Problem

$$V = \max_{\{C_{inj}, L_{inj}\}} U \left( C - \frac{1}{\phi^{\frac{1}{\phi}}} \frac{L^{\frac{\phi+1}{\phi}}}{\frac{\phi+1}{\phi}} \right)$$

$$\text{s.t.} \quad C_{inj} = Y_{inj} = A_{inj} L_{inj}$$

# Counterfactual Economies

1. DECENTRALIZED EQUILIBRIUM:  $L_{inj}^{\mu, \delta}$

$$A_{inj} P_{inj} \mu_{inj}^{-1} = W_{inj} \delta_{inj}$$

# Counterfactual Economies

2. SOCIAL PLANNER'S SOLUTION:  $L_{inj}^{1,1}$

$$A_{inj}P_{inj} = W_{inj}$$

# Counterfactual Economies

3. MONOPOLY; NO MONOPSONY:  $L_{inj}^{\mu,1}$

$$A_{inj} P_{inj} \mu_{inj}^{-1} = W_{inj}$$



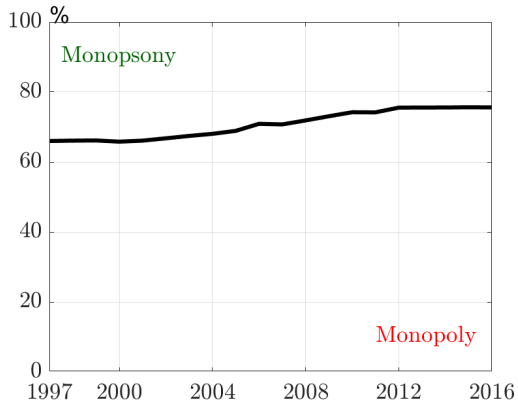
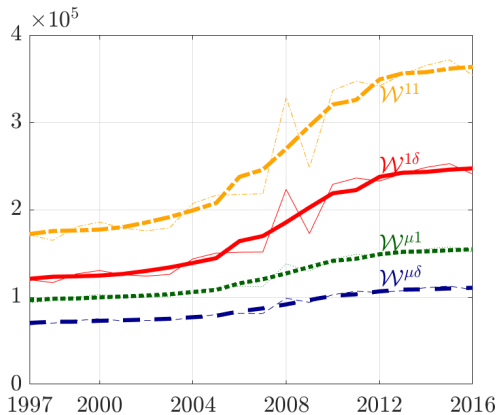
# Counterfactual Economies

4. NO MONOPOLY; MONOPSONY:  $L_{inj}^{1,\delta}$

$$A_{inj} P_{inj} = W_{inj} \delta_{inj}$$

# Counterfactual Economies

## Wage Decomposition



# Conclusion

- We propose a novel method to:
  1. Jointly model and measure monopsony and monopoly
  2. Back out market structure
- Our Main Findings:
  1. Market Power has increased over time:
    - Markups increase from 1.69 to 2.2
    - Markdowns are stable, increase only marginally from 1.37 to 1.4
  2. Wage stagnation: decoupling wages-productivity
  3. Decomposition: indirect effect from monopoly dominates direct effect from monopsony  
Monopoly's relative share of the wage decline is 75% in 2016

# WHAT DRIVES WAGE STAGNATION: MONOPSONY OR MONOPOLY?

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## Producer Optimality

$$P_{inj} + \frac{\partial P_{inj}}{\partial Y_{inj}} Y_{inj} + \sum_{i' \in \mathcal{I}_{nj} \setminus i} \left( \frac{\partial P_{i'nj}}{\partial Y_{inj}} Y_{i'nj} \right) = \frac{1}{A_{inj}} \left[ W_{inj} + \frac{\partial W_{inj}}{\partial L_{inj}} L_{inj} + \sum_{i' \in \mathcal{I}_{nj} \setminus i} \left( \frac{\partial W_{i'nj}}{\partial L_{inj}} L_{i'nj} \right) \right]$$

$$P_{inj} \left[ \underbrace{1 - \frac{1}{\theta} s_{nj} - \frac{1}{\eta} (1 - s_{nj})}_{\epsilon_{inj}^P} \right] A_{inj} = W_{inj} \left[ \underbrace{1 + \frac{1}{\hat{\theta}} e_{nj} + \frac{1}{\hat{\eta}} (1 - e_{nj})}_{\epsilon_{inj}^W} \right]$$

We define our markup  $\mu_{inj} = \frac{P_{inj}}{MC_{inj}}$  and markdown  $\delta_{inj} = \frac{MRPL_{inj}}{W_{inj}}$

$$\mu_{inj} = \frac{1}{1 + \epsilon_{inj}^P} = \left[ 1 - \frac{1}{\theta} s_{nj} - \frac{1}{\eta} (1 - s_{nj}) \right]^{-1} \quad \text{and} \quad \delta_{inj} = 1 + \epsilon_{inj}^W = \left[ 1 + \frac{1}{\hat{\theta}} e_{nj} + \frac{1}{\hat{\eta}} (1 - e_{nj}) \right]$$

## Model Solution

Rearranging FOC, we get:

$$P_{inj} = \frac{\left[1 + \frac{1}{\hat{\theta}} e_{nj} + \frac{1}{\hat{\eta}} (1 - e_{nj})\right]}{\left[1 - \frac{1}{\hat{\theta}} s_{nj} - \frac{1}{\hat{\eta}} (1 - s_{nj})\right]} \frac{W_{inj}}{A_{inj}}$$

$$s_{inj} = \frac{P_{inj}^{1-\eta}}{\sum_{i'} P_{i'nj}^{1-\eta}} = \frac{\left[ \frac{1 + \frac{1}{\hat{\theta}} e_{nj} + \frac{1}{\hat{\eta}} (1 - e_{nj})}{1 - \frac{1}{\hat{\theta}} s_{nj} - \frac{1}{\hat{\eta}} (1 - s_{nj})} \frac{e_{inj}^{\frac{1}{1+\hat{\eta}}}}{A_{inj}} \right]^{1-\eta}}{\sum_{i'} \left[ \frac{1 + \frac{1}{\hat{\theta}} e_{n'j} + \frac{1}{\hat{\eta}} (1 - e_{n'j})}{1 - \frac{1}{\hat{\theta}} s_{n'j} - \frac{1}{\hat{\eta}} (1 - s_{n'j})} \frac{e_{i'n'j}^{\frac{1}{1+\hat{\eta}}}}{A_{i'n'j}} \right]^{1-\eta}}$$

where

$$e_{inj} = \frac{\left( s_{inj}^{\frac{\eta}{\eta-1}} / A_{inj} \right)^{\frac{\hat{\eta}+1}{\hat{\eta}}}}{\sum_{i'} \left( s_{i'nj}^{\frac{\eta}{\eta-1}} / A_{i'nj} \right)^{\frac{\hat{\eta}+1}{\hat{\eta}}}} = \left[ \sum_{i'} \left( \left( \frac{s_{i'nj}}{s_{inj}} \right)^{\frac{\eta}{\eta-1}} \frac{A_{inj}}{A_{i'nj}} \right)^{\frac{\hat{\eta}+1}{\hat{\eta}}} \right]^{-1}$$

# Regression Specification

We use Two-Stage Least Squares (2SLS) on the following equations to get the estimate of  $\hat{\eta}$  and  $\hat{\theta}$ .

- $\hat{\eta}$  Estimation

$$\ln W_{injt}^* = k_{jt} + \gamma \ln L_{jt} + \beta \ln L_{injt} + \underbrace{\alpha_{inj} + \epsilon_{injt}}_{\epsilon_{injt}} \quad (1)$$

- $\hat{\theta}$  Estimation

$$\bar{\Omega}_{Sjt} = k_{jt} + \gamma_S \ln S_{jt} + \bar{\epsilon}_{Sjt} \quad (2)$$

where we define  $\beta = \frac{1}{\hat{\eta}}$  and  $\gamma = (\frac{1}{\hat{\theta}} - \beta)$ .

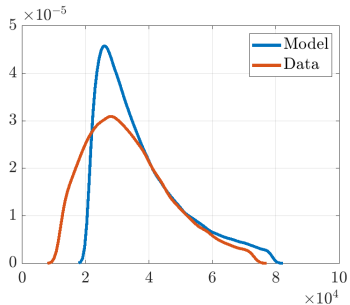
# First and Second Stage Results

**Table:** Estimates of reduced-form parameters: Tradeables

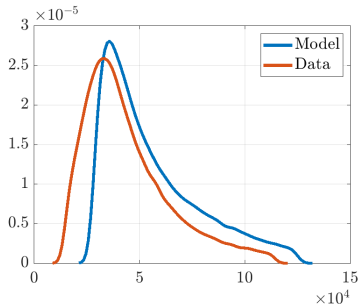
| <b>A. OLS and Second-Stage IV Estimates</b>  |                    |                    |   |                   |                    |
|--|--------------------|--------------------|---|-------------------|--------------------|
|  | OLS                | IV                 |   | OLS               | IV                 |
|  | (1)                | (2)                |   | (3)               | (4)                |
| $\frac{1}{\bar{\eta}}$                       | -0.197<br>(5.0e-4) | 0.323<br>(0.051)   | $\frac{1}{\bar{\theta}} - \frac{1}{\bar{\eta}}$ | 0.110<br>(2.0e-4) | 0.207<br>(0.006)   |
| Market x Year FE                             | Yes                | Yes                | Market FE                                       | Yes               | Yes                |
| Establishment FE                             | Yes                | Yes                | Year FE   | Yes               | Yes                |
| <b>B. First-Stage Regressions for the IV</b> |                    |                    |   |                   |                    |
| $\tau_{X(i)t}$                               | -                  | -0.003<br>(2.0e-4) | $\bar{\tau}_{jt}$                               | -                 | -0.023<br>(3.0e-4) |
| Market x Year FE                             | -                  | Yes                | Market FE                                       | -                 | Yes                |
| Establishment FE                             | -                  | Yes                | Year FE   | -                 | Yes                |
| No. of obs.                                  | 2,559,000          | 2,559,000          | No. of obs.                                     | 2,674,000         | 2,674,000          |



# Wage Distribution



Wage Distribution 1997



Wage Distribution 2016

# Estimated Technology Distribution

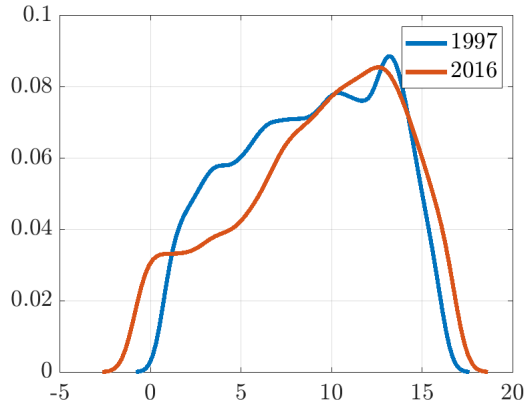


Figure:  $A_{inj}$

## N Estimation Fit

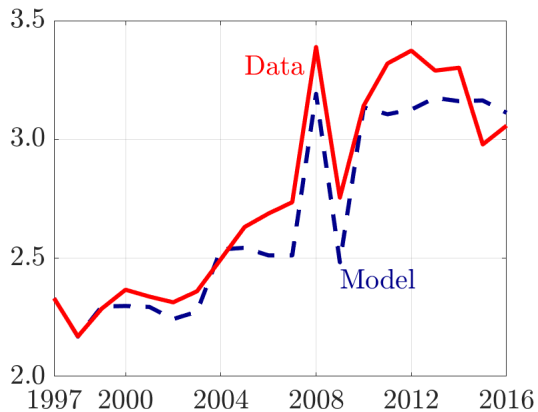


Figure: Model Fit-N estimation