## What Drives Wage Stagnation: Monopsony or Monopoly?

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## Wage Stagnation

U.S Census: Tradeable sectors


## Mechanisms

- Explore two mechanisms behind wage stagnation:

1. Monopsony: direct effect from imperfect labor market
$\rightarrow$ Lower firm-specific wages for own workers
2. Monopoly: output market power affects labor demand - General Equilibrium effect
$\rightarrow$ Lowers aggregate, economy-wide wages

## Mechanisms

- Explore two mechanisms behind wage stagnation:

1. Monopsony: direct effect from imperfect labor market
$\rightarrow$ Lower firm-specific wages for own workers
2. Monopoly: output market power affects labor demand - General Equilibrium effect
$\rightarrow$ Lowers aggregate, economy-wide wages
$\therefore$ Objective:
3. Explain mechanism behind decoupling of wages and productivity
4. 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

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- Challenge for measurement: marginal cost directly not observable
- Challenge for measurement: we don't observe who competes
- 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 $i=1, \ldots, l$
- Finite numbers of firms in each market $n=1, \ldots, N$ (set of establishments $i$ in firm $n: \mathcal{I}_{n j}$ )


## Household Preferences

- maximizes static utility

$$
\max _{C_{i n j}, L_{i n j}} U\left(C-\frac{1}{\bar{\phi}^{\frac{1}{\phi}}} \frac{L^{\frac{\phi+1}{\phi}}}{\frac{\phi+1}{\phi}}\right)
$$

s.t. $P C=L W+\Pi$

- CES preferences over Consumption and Labor

$$
\begin{aligned}
& C=\left(\int_{j} J^{\frac{1}{\theta}} C_{j}^{\frac{\theta-1}{\theta}} d j\right)^{\frac{\theta}{\theta-1}}, \quad C_{j}=\left(\sum_{i} I^{-\frac{1}{\eta}} C_{i n j}^{\frac{\eta-1}{\eta}}\right)^{\frac{\eta}{\eta-1}} \\
& L=\left(\int_{j} J^{\frac{1}{\theta}} L_{j}^{\frac{\hat{\theta}+1}{\theta}} d j\right)^{\frac{\hat{\theta}}{\hat{\theta}+1}}, \quad L_{j}=\left(\sum_{i} I^{\frac{1}{\eta}} L_{i n j}^{\frac{\hat{n}}{\eta}}\right)^{\frac{\hat{\eta}}{\hat{\eta}+1}}
\end{aligned}
$$

## Model Setup

## Technology

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

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

subject to

$$
Y_{i n j}=A_{i n j} L_{i n j}
$$

## 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:

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

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$$

- Markups and Markdowns

$$
\begin{array}{cl}
\mu_{i n j}=\frac{P_{i n j}}{M C_{i n j}}=\frac{1}{1+\varepsilon_{i n j}^{P}} ; \quad & \varepsilon_{i n j}^{P}=-\left[\frac{1}{\theta} s_{n j}+\frac{1}{\eta}\left(1-s_{n j}\right)\right] \\
\delta_{i n j}=\frac{M R P L_{i n j}}{W_{i n j}}=1+\varepsilon_{i n j}^{W} ; & \varepsilon_{i n j}^{W}=\left[\frac{1}{\hat{\theta}} e_{n j}+\frac{1}{\hat{\eta}}\left(1-e_{n j}\right)\right]
\end{array}
$$

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\end{array}
$$

- Mechanism

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

## Quantitative Exercise

- U.S. Census Bureau Longitudinal Business Database (LBD): Tradeable Sectors
- In the data we observe

1. Employment by establishment: $L_{i n j}$
2. Average Wages by establishment: $W_{i n j}=\frac{\text { Wage } B_{i i l} i_{i j j}}{L_{i n j}}$
3. Revenue: $R_{i n j}$
4. Industry classification NAICS, SIC

- Market Assignment: Randomly assign $l_{j}$ establishments in same industry into a market. Randomly assign $l_{j}$ establishments into $N$ subsets of size $l_{j} / N$


## Quantitative Exercise

## Estimation

|  | Input $/$ data | Output |  |
| :--- | :---: | :---: | :--- |
| 1. Common elasticities | $W_{i n j}, L_{i n j}$ | $\hat{\theta}, \hat{\eta}$ |  |
| 2. Firm-specific technology | $L_{i n j}$ | $A_{i n j}, \mu_{i n j}, \delta_{i n j}$ | system of FOCs given $N$ |
| 3. Market Structure | $R_{i n j} / W_{i n j} L_{i n j}$ | $N$ |  |

## Estimating Labor Elasticities

Estimating Within and Between Market Substitutability

$$
\ln W_{i n j t}^{*}=\mathrm{c}_{j t}+\gamma \ln L_{j t}+\beta \ln L_{i n j t}+\underbrace{\alpha_{i n j}+\epsilon_{i n j t}}_{\varepsilon_{i n j t}}
$$

where we define $\beta=\frac{1}{\hat{\eta}}$ and $\gamma=\left(\frac{1}{\hat{\theta}}-\beta\right)$
Use Two-Stage Least Squares to estimate $\beta$ and $\gamma$, sequentially.
Rely on Berger, Herkenhoff and Mongey (2021) and Giroud and Rauh (2019)

- Exploit variation in state corporate taxes as instruments for employment


## Preference Estimates and Parameters

| Variable | Value |  | Source |
| :---: | :---: | :--- | :--- |
| $\hat{\theta}$ | 1.71 | Input market: Between-market elasticity | estimated |
| $\hat{\eta}$ | 3.49 | 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) |
| $\boldsymbol{l}$ | 32 | Establishments in each market | Externally set |

## Backing out $\left\{A_{i n j}, \mu_{i n j}, \delta_{i n j}\right\}$

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

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

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

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

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

Estimated Technology Distribution
$A_{i n j}$


Estimated $N$


Average Markups and Markdowns


## Decoupling Wages-Productivity


(a) Data

(b) Model

## Decoupling Wages-Productivity

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



## Counterfactual Economies

Wage Decomposition




## Counterfactual Economies

## Wage Growth/Stagnation




## 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 1.45 to 1.93
- Markdowns are stable, increase only marginally from 1.33 to 1.38

2. Wage stagnation: decoupling wages-productivity
3. Decomposition: indirect effect from monopoly dominates direct effect from monopsony $69 \%$ of wage level; $80 \%$ of the wage stagnation

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## Producer Optimality

$$
\begin{gathered}
P_{i n j}+\frac{\partial P_{i n j}}{\partial Y_{i n j}} Y_{i n j}+\sum_{i^{\prime} \in \mathcal{I}_{n j} / i}\left(\frac{\partial P_{i^{\prime} n j}}{\partial Y_{i n j}} Y_{i^{\prime} n j}\right)=\frac{1}{A_{i n j}}\left[W_{i n j}+\frac{\partial W_{i n j}}{\partial L_{i n j}} L_{i n j}+\sum_{i^{\prime} \in \mathcal{I}_{n j} / i}\left(\frac{\partial W_{i^{\prime} n j}}{\partial L_{i n j}} L_{i^{\prime} n j}\right)\right] \\
P_{i n j}[1-\underbrace{1-\frac{1}{\theta} s_{n j}-\frac{1}{\eta}\left(1-s_{n j}\right)}_{\epsilon_{i n j}^{P}}] A_{i n j}=W_{i n j}[1+\underbrace{\frac{1}{\hat{\theta}} e_{n j}+\frac{1}{\hat{\eta}}\left(1-e_{n j}\right)}_{\epsilon_{i n j}^{W}}]
\end{gathered}
$$

We define our markup $\mu_{i n j}=\frac{P_{i n j}}{M C_{i n j}}$ and markdown $\delta_{i n j}=\frac{M R P L_{i n j}}{W_{i n j}}$

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

## Model Solution

Rearranging FOC, we get:

$$
\begin{gathered}
P_{i n j}=\frac{\left[1+\frac{1}{\hat{\theta}} e_{n j}+\frac{1}{\hat{\eta}}\left(1-e_{n j}\right)\right]}{\left[1-\frac{1}{\theta} s_{n j}-\frac{1}{\eta}\left(1-s_{n j}\right)\right]} \frac{W_{i n j}}{A_{i n j}} \\
\left.s_{i n j}=\frac{P_{i n j}^{1-\eta}}{\sum_{i, n} P_{i n j}^{1-\eta}}=\frac{\left[\frac{1+\frac{1}{\hat{\theta}} e_{n j}+\frac{1}{\hat{\eta}}\left(1-e_{n j}\right)}{1-\frac{1}{\theta} s_{n j}-\frac{1}{\eta}\left(1-s_{n j}\right)} \frac{e_{i n j}^{\frac{1}{1+\hat{\eta}}}}{A_{i n j}}\right]^{1-\eta}}{\sum_{i^{\prime}, n^{\prime}}\left[\frac{1+\frac{1}{\hat{\theta}} e_{n^{\prime} j}+\frac{1}{\hat{\eta}}\left(1-e_{n^{\prime} j}\right)}{1-\frac{1}{\theta} s_{n^{\prime} j}-\frac{1}{\eta}\left(1-s_{n^{\prime} j}\right)} e_{i^{\prime} n^{\prime} n^{\prime}}^{\frac{1}{1+\hat{\eta}}}\right]_{i^{\prime} n^{\prime} j}^{1-\eta}}\right]^{1-\eta}
\end{gathered}
$$

where

$$
e_{i n j}=\left[\sum_{i^{\prime}, n^{\prime}}\left(\left(\frac{s_{i^{\prime} n^{\prime} j}}{s_{i n j}}\right)^{\frac{\eta}{\eta-1}} \frac{A_{i n j}}{A_{i^{\prime} n^{\prime} j}}\right)^{\frac{\hat{\eta}+1}{\hat{\eta}}}\right]^{-1}=\frac{\left(s_{i n j}^{\frac{-\eta}{1-\eta}} / A_{i n j}\right)^{\frac{1+\hat{\eta}}{\hat{\eta}}}}{\sum_{i^{\prime}, n^{\prime}}\left(s_{i^{\prime} n^{\prime} n^{\prime} j}^{\frac{-\eta}{1-\eta}} / A_{i^{\prime} n^{\prime} n^{\prime} j}\right)^{\frac{1+\hat{\eta}}{\hat{\eta}}}} .
$$

## 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

$$
\begin{equation*}
\ln W_{i n j t}^{*}=k_{j t}+\gamma \ln L_{j t}+\beta \ln L_{i n j t}+\underbrace{\alpha_{i n j}+\epsilon_{i n j t}}_{\varepsilon_{i n j t}} \tag{1}
\end{equation*}
$$

- $\hat{\theta}$ Estimation

$$
\begin{equation*}
\bar{\Omega}_{S j t}=k_{j t}+\gamma_{S} \ln S_{j t}+\bar{\varepsilon}_{S j t} \tag{2}
\end{equation*}
$$

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

## First and Second Stage Results

Table: Estimates of reduced-form parameters: Tradeables

| A. OLS and Second-Stage IV Estimates |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  | OLS | IV |  | OLS | IV |
|  | (1) | (2) |  | (3) | (4) |
| $\frac{1}{\hat{\eta}}$ | $\begin{gathered} -0.187 \\ (3.8 \mathrm{e}-4) \end{gathered}$ | $\begin{gathered} 0.287 \\ (0.048) \end{gathered}$ | $\frac{1}{\hat{\theta}}-\frac{1}{\hat{\eta}}$ | $\begin{gathered} 0.180 \\ (1.3 \mathrm{e}-4) \end{gathered}$ | $\begin{gathered} 0.298 \\ (0.001) \end{gathered}$ |
| Sector $\times$ Year FE | Yes | Yes | Sector FE | Yes | Yes |
| Establishment FE | Yes | Yes | Year FE | Yes | Yes |
| B. First-Stage Regressions for the IV |  |  |  |  |  |
| $\tau_{X(i) t}$ |  | $\begin{gathered} -0.003 \\ (1.9 \mathrm{e}-4) \end{gathered}$ | $\bar{\tau}_{j t}$ | - | $\begin{gathered} -0.138 \\ (3.8 \mathrm{e}-4) \end{gathered}$ |
| Sector $\times$ Year FE | - | Yes | Sector FE | - | Yes |
| Establishment FE | - | Yes | Year FE | - | Yes |
| No. of obs. | 3,921,000 | 3,921,000 | No. of obs. | 3,921,000 | 3,921,000 |

## Wage Distribution



Wage Distribution 1997


Wage Distribution 2016

N Estimation Fit


Figure: Model Fit-N estimation

