

# **Monopsony Power and Creative Destruction**

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# Introduction

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- **Research Question:**
  - How does labor market power affect output and growth?
- **Key trade-off:**
  - Monopsony → markdown distribution
  - Static misallocation (lower current output)
  - Innovation incentives (higher output growth)

- Productivity growth in developed countries:
  - Slowdown over last decades broadly
- One approach in existing research: product market power
  - Aghion et al., 2023, De Ridder, 2022
- We incorporate labor market power: monopsony
  - Studied e.g. by Berger et al., 2022, Bachmann et al., 2022
  - Focus in existing literature: static misallocation
  - This paper: incorporate long-run growth implications

- Framework builds on existing firm dynamics & growth models:
  - Klette and Kortum, 2004, Aghion et al., 2023
  - Growth model of creative destruction and product market power
- We expand this with labor market power:
  - Discrete choice workplaces & home production, Card et al., 2018
- Note on monopsony:
  - 'New classical monopsony' as in Card et al., 2018, Manning, 2021
  - Wage setting power: upward sloping labor supply curve facing firm

# Model Setup

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# Workers

- Mass  $\mathcal{L}$  workers, no savings
- Choose to work ( $g = e$ ) at firm  $j \in \{1, \dots, \mathcal{J}\}$ , or at home ( $g = u$ )
- Utility of worker  $o$ , choosing to work at firm  $j$ :

$$u_{oj} = \beta \log C_j + \xi_{oe} + (1 - \sigma)\varepsilon_{oj}. \quad \xi_{og}, \varepsilon_{oj} \sim EVT1$$

- From logit-choice then follows labor supply given net wage:

$$L_j(W_j) = z W_j^{\frac{\beta}{1-\sigma}},$$

Details

- where  $z$  includes the option value of all wages and the outside option
- The labor supply elasticity is:

$$\frac{\partial \log L_j}{\partial \log W_j} = \frac{\beta}{1 - \sigma}$$

- **Final goods production:**  $Y = \exp \int_0^1 \ln(q_i y_i) di$ .
  - $q_i$  is quality level of good  $i$
- **Intermediate good demand:**  $p_i y_i = PY$ , normalize  $P \equiv 1$ .
- **Competition:** Details
  - Bertrand competition in product lines, quality breaks ties.
  - Quality leader in line  $i$  is  $j(i)$ , follower  $j'(i)$
  - Leader's quality is one  $\gamma$ -step above follower's:  $q_{j(i)} = \gamma q_{j'(i)}$
  - Nash equilibrium: Leader fulfills line demand,  $p_i = \gamma m_{Cj'(i)}$ .
- **Intermediate goods production:**  $y_{i,j(i)} = s_{j(i)} l_{i,j(i)}$ .
- **Key link:**  $m_{Cj'(i)}$  depends on firm size due to monopsony! Details
- **Firm types:** Top 10% with productivity  $s_h$ , remaining with  $s_l$



## Dynamic Block

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# Dynamic decision: Research effort

- Given line-level solutions:
  - $n_{j,t}$ : number of product lines where firm  $j$  is quality leader
  - This is firms' only state variable,  $L_{jt}$  &  $W_{jt}$  follow from it
  - Markups, markdowns function of firm size [Details](#)
- The dynamic problem is how much to invest in research:
  - Stock of lines develops according to:  $n_{j,t+1} = (1 - X_t)n_{j,t} + x_{jt}$
  - Aggregate rate of creative destruction:  $X_t = \sum_j x_{jt}$
  - Cost of drawing  $x_t$  new lines:  $R(x_{jt}) = \psi Y x_{jt}^\phi$ .

# Firm Problem on BGP

- Focus on a balanced growth path
  - Constant  $\mathcal{J}, X$ , constant Top 10% concentration  $h$
  - Quality growth  $Q_{t+1}/Q_t = g = \gamma^X$
  - $Y_t, mc_{jt}, W_{jt}$  all grow at  $g$  &  $z$  at  $g_z = g^{-\frac{\beta}{1-\sigma}}$
- Restate firm problem, relative to output  $Y$ :

$$v_j(n_j) = \max_{n'_j} n_j - \frac{W_j}{Y} L_j \\ - \psi(n'_j - (1 - X)n_j)^\Phi + \rho v(n'_j),$$

- $W_j, L_j$  are functions of  $n_j$ , which is constant on BGP

# Output Decomposition

- Define  $S \equiv \int_0^1 s_{j(i)} di$ ,  $L \equiv \sum_j L_j$

$$\begin{aligned} Y &= \exp \int_0^1 \ln(q_i y_i) di = Q \exp \int_0^1 \ln(s_{j(i)}) di \exp \int_0^1 \ln(l_{j(i)}) di \\ &= Q \cdot S \cdot M \cdot L \end{aligned}$$

Details

- Where  $M = \frac{\exp \int_0^1 \ln \mu_{j(i)} di}{\int_0^1 \mu_{j(i)} di}$  is misallocation from price dispersion
- Decomposition of present value, accounting for  $g$ :

$$PV \left\{ \frac{Y}{L} \right\}_{t=0}^{\infty} \approx \underbrace{\frac{Q_0}{1 - \rho(1 + g)}}_{TFP} \cdot S \cdot M \cdot L$$

- Tension between static- and dynamic efficiency. Higher  $h$ :
  - Increases  $S$ , but also R&D spending  $Y \sum_j \psi(X n_j)^{\phi}$  for given  $X$

## Quantitative Results

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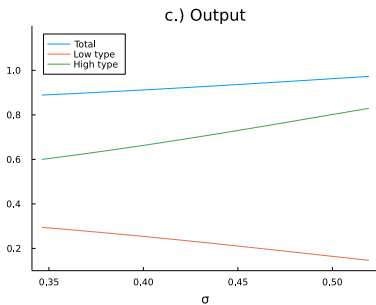
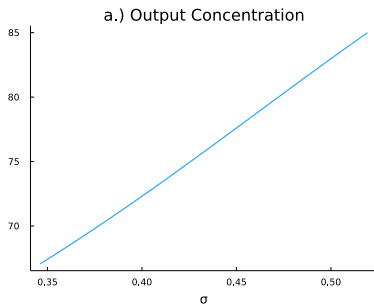
# Model Fit

- Match U.S. economy 1954 – 2007: [Details](#)

Definition	Data	Model
Average Markup	1.24	1.28
Growth rate	1.078%	1.077%
R&D spending (% of GDP)	2.45%	6.33%
Share of Output, top 10% firms	75.59%	75.74%
Labor Market Participation	83.4%	83.41%
Profit Share	5.45%	5.45%
Top 10% wage premium	21%	25.4%

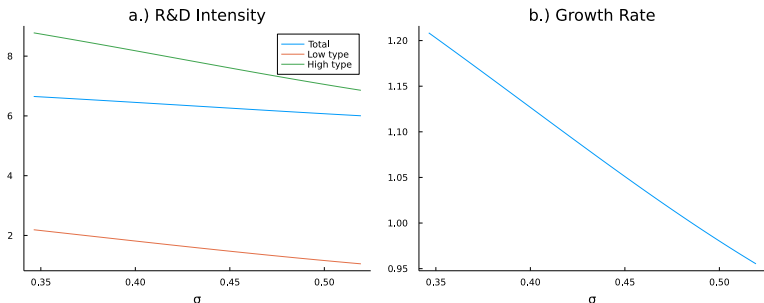
- Good overall match, although R&D spending too high

# Static Results



- Concentration increases as  $\sigma$  increases (labor elasticity increases)
- Production (by large firms) increases

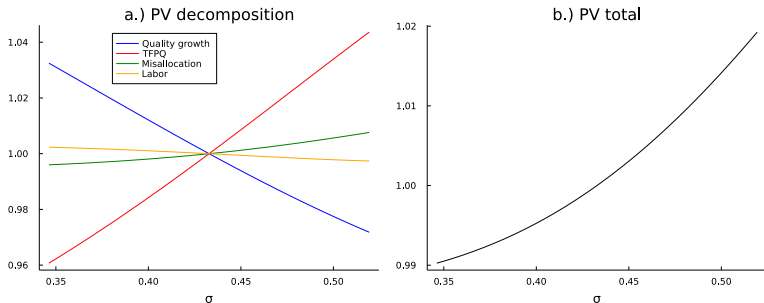
# Dynamic Results



- R&D by large firms increases, but not in line with Output increases
- Small firm R&D declines
- More concentrated R&D also less efficient
- Strong decline in productivity growth



# Present Value Decomposition



- Main channels: Quality growth versus Static TFPQ
- Here: PV maximized for high concentration – low growth scenario!
- Preference change, so no welfare analysis here

- Contribution:
  - Importance of labor supply elasticities for output, wages and growth
  - Key result: Static-dynamic tradeoff
- Omitted in this presentation:
  - Detailed results w.r.t markups, markdowns and wages
  - Policy exercise: Income taxes
  - Amenity-heterogeneity (WIP)

## **Extension: Taxes**

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- Tax function as in Borella et al., 2022, but here paid by firm:

$$T\left(\frac{W_j}{\bar{W}}\right) = \left(\frac{1}{1-\lambda} \frac{W_j^\tau}{\bar{W}^\tau}\right)^{\frac{1}{1-\tau}} - 1$$

- $\lambda$  governs average tax level,  $\tau$  progressivity
- $1 - \tau$  is the elasticity of post tax income w.r.t pre tax income
- Reference wage:  $\bar{W} = \sum_j L_j W_j / \sum_j L_j$
- The budget balances, government spending  $G$  per household:

$$\mathcal{L}G = \sum_j T(W_j/\bar{W}) W_j L_j$$

# Gross Wage Labor Supply Elasticity

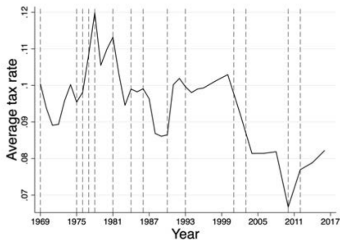
- Gross wage:  $W^G = (1 + T(W_j/\bar{W}))W_j$
- Labor supply elasticity wrt the gross wage  $W^G$ :

$$\frac{\partial \log(L_j)}{\partial \log(W^G)} = \underbrace{\frac{\beta}{1-\sigma}}_{\text{preferences}} \underbrace{(1-\tau)}_{\text{policy}} \quad (1)$$

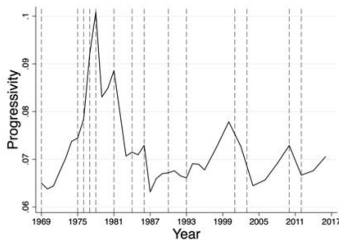
- This is the elasticity relevant to the firm
- Can be directly affected by changing  $\tau$

# Income Taxation

- Tax level  $\lambda$  and progressivity  $\tau$  from Borella et al., 2022



(a) Average tax rate at median income ( $\lambda$ )



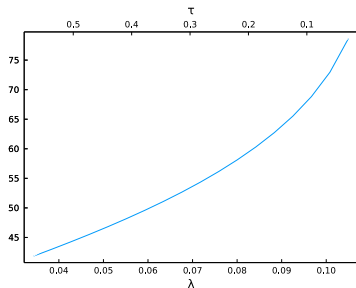
(b) Progressivity parameter ( $\tau$ )

- Macnamara et al., 2024 suggest tax cuts should increase TFP growth
- TFP growth in data not high(er) post tax cuts, according to model:
  - $\lambda \downarrow$  has no effect on  $h$ , slightly increases R&D for all firms
  - $\tau \downarrow$  increases labor elasticity, increases  $h$ , decreases (small) firm R&D

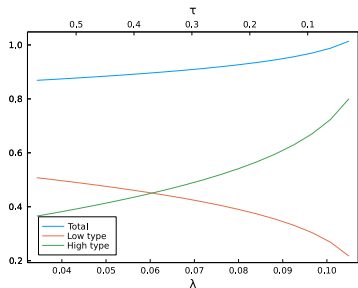
# Comparing Tax Regimes

- Before: Little effect from historical reforms
- Now: Show that tax regime can strongly affect growth
- To discipline this exercise, we fix today's  $G$  at its base level
- Increasing  $\lambda$ , decreasing  $\tau$  makes taxes less progressive
- Concentration (almost) entirely from  $\tau$ , through labor elasticity

# Static Results



(a) Top 10% Concentration (in %)

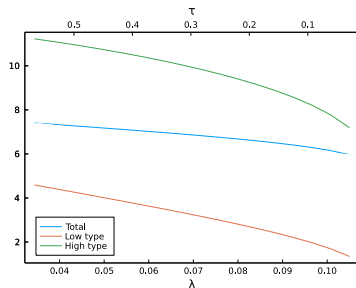


(b) Output, relative to base

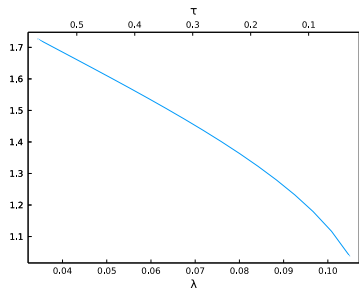
- Concentration increases as  $\tau$  decreases (labor elasticities increase)
- Note: Higher  $\lambda$  decreases Output
- Effect from  $\tau$  (higher  $S$ ) dominates, Output increases overall



# Dynamic Results



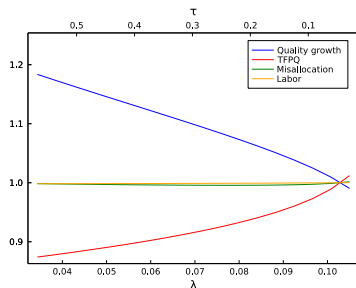
(a) R&D intensity (in %)



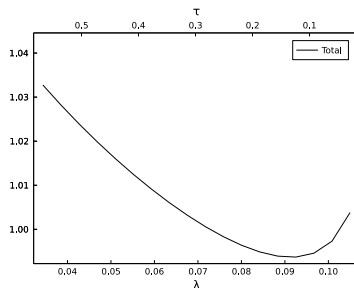
(b) Productivity Growth (in %)

- R&D by large firms increases, but not in line with Output increases
- Small firm R&D declines
- more concentrated R&D also less efficient
- strong decline in productivity growth

# Present Value Decomposition



(a) PV Decomposition, relative to base



(b) Total PV, relative to base

- Main channels: Quality growth versus Static TFPQ
- Present value maximized in low base – high progressivity regime
- PV U-Shape, but  $S$  capped at  $h = 1$  (requires regressive  $\tau < 0$ !)

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# Appendix

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## Labor Supply: details

- Using  $D_e = \sum_{k=1}^{\mathcal{J}} W_k^{\frac{\beta}{1-\sigma}}$  and  $D_u = (\omega Y)^{\frac{\beta}{1-\sigma}}$ :

$$P(g = e) = \frac{D_e^{1-\sigma}}{D_e^{1-\sigma} + D_u^{1-\sigma}}$$

$$P(j|g = e) = \frac{\exp(\beta \frac{\log W_j}{1-\sigma})}{D_e} = \frac{W_j^{\frac{\beta}{1-\sigma}}}{D_e}$$

$$P(g = e)P(j|g = e) = \frac{W_j^{\frac{\beta}{1-\sigma}}}{D_e^{\sigma}(D_e^{1-\sigma} + D_u^{1-\sigma})}$$

which implies:

$$L_j(W_j) = \mathcal{L}P(W_j) = \mathcal{L} \frac{W_j^{\frac{\beta}{1-\sigma}}}{\left(\sum_{k=1}^{\mathcal{J}} W_k^{\frac{\beta}{1-\sigma}}\right)^{\sigma} (\omega Y)^{\beta} + \sum_{k=1}^{\mathcal{J}} W_k^{\frac{\beta}{1-\sigma}}}$$

# Within-line Nash equilibrium

- There are other equilibria, in which  $j'$  threatens price  $< mc_{j'}$
- This feature exists in all Klette-Kortum type models
- Competition is in prices, firms commit to produce by setting price
  - Is this a crazy assumption with our increasing marginal cost?
  - Recall that lines are atomistic....
  - ... and that acquiring them is costly!
  - Producing in a single additional line has little of effect on cost
  - In addition, acquiring a line and then not producing in it is clearly not optimal

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# Note on marginal costs

- Firm-level employment:  $L_j = \frac{Y_j}{s_j}$ ,
- Firm-level output:  $Y_j = \int_0^{n_j} y_i di = \int_0^{n_j} \frac{Y}{\gamma mc_{j'(i)}} di$ .
  - On BGP, every firm faces the same distribution of 'followers' marginal costs.
  - Therefore,  $Y_j = \int_0^{n_j} \frac{Y}{\gamma mc_{j'(i)}} di = \frac{Y}{\gamma m} n_j$ , where  $m^{-1} \equiv \int_0^1 \frac{1}{mc_{j'(i)}} di$
- Wage:  $W_j = \left(\frac{L_j}{z}\right)^{\frac{1-\sigma}{\beta}} = \left(\frac{Y_j}{s_j z}\right)^{\frac{1-\sigma}{\beta}}$ 
  - Recall  $z \equiv \frac{\mathcal{L}}{D_e^\sigma (\bar{W} Y)^\beta + D_e}$
- Production costs:  $C(Y_j) = (1 + T(W_j(Y_j)/\bar{W}))W_j(Y_j)L_j(Y_j)$
- Marginal cost of increasing production:  $mc_j = C'(Y_j)$ .

# Markups and Markdowns

- From line-level equilibrium:  $p_i = \gamma mc_{j'(i)}$
- Line-level markups  $p/mc$  thus depend on leader, follower:

$$\mu_{j(i)j'(i)} = \gamma \frac{mc_{j'(i)}}{mc_{j(i)}}$$

- Firm-level markups additionally a function of  $m = \left( \int_0^1 mc_{j(i)}^{-1} di \right)^{-1}$

$$\mu_j \equiv \frac{\int_0^{n_j} y_i p_i di}{mc_j \cdot \int_0^{n_j} y_i di} = \frac{\gamma m}{mc_j}.$$

- Gross wage markdown is then a function of markup, taxes:

$$\frac{W_j \cdot \left( 1 + T \left( \frac{W_j}{W} \right) \right)}{\gamma ms_j} = \frac{1}{\mu_j} \cdot \frac{\frac{\beta}{1-\sigma}}{1 + \frac{\beta}{1-\sigma} + \frac{\tau}{1-\tau}}$$

## Closing the model

- Final output is spent on private consumption  $C$ , government consumption  $\mathcal{L}G$ , research spending  $X$ , and rents  $R$ .
  1.  $X = Y \sum_j \psi(n'_j - (1 - X)n_j)^\phi$
  2.  $C = \int_o W_o$
  3.  $R = \sum_j (Y - (1 + T(W_j/\bar{W}))L_j W_j - Y \psi(n'_j - (1 - X)n_j)^\phi)$
- Growth rate depends on aggregate rate  $X$  of creative destruction:

$$X = \sum_j x_j, \quad g = \gamma^X.$$

# Algorithm, Outer Loop

- **Outer loop:** Guess  $J_{\text{guess}}$
- **Inner loop:** Fully solve model given  $J_{\text{guess}}$
- **Compute:**  $V_{\text{entry}} = \frac{\alpha \tilde{v}_h(n_h) + (1-\alpha) \tilde{v}_l(n_l)}{1-\rho}$
- **Outer Check:**  $|V_{\text{entry}} - \text{entry cost}|$

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# Algorithm, Inner Loop

**Inner loop:** Guess  $h_{\text{guess}}, \left(\frac{Y}{mz}\right)_{\text{guess}}$

- Compute  $n_h = \frac{h_{\text{guess}}}{J_h}, n_l = \frac{1-h_{\text{guess}}}{J_l}$
- Get  $w_j = \left(n_h \left(\frac{Y}{mz}\right)_{\text{guess}} \frac{1}{\gamma s_j}\right)^{\frac{1-\sigma}{\beta}}$  and  $\bar{W} = f_w(h, w_h, w_l)$
- $mc_j = f_{mc}(n_j, s_j, \frac{Y}{mz}, \bar{W})$  and  $m = \left[\frac{h}{mc_h} + \frac{1-h}{mc_l}\right]^{-1}$
- $D_e = J_h w_h^{\frac{\beta}{1-\sigma}} + J_l w_l^{\frac{\beta}{1-\sigma}}$
- Find  $Y$  such that  $\left(\frac{Y}{mz}\right)_{\text{guess}} = \frac{Y^{1+\beta} \omega D_e^\sigma + Y D_e}{m L s}$
- $D_0 = (\omega Y)^\beta$
- $z = \frac{L s}{D_0 D_e^\sigma + D_e}, L_j = w_j^{\frac{\beta}{1-\sigma}} z$

**Inner Check:**  $\left| n_h^{\phi-1} (mc_l - \gamma m) - n_l^{\phi-1} (mc_h - \gamma m) \right| + \left| mc_h^h mc_l^{1-h} - \frac{Q}{\gamma} \right|$

- Solve for  $X \in (0, 1)$  using  $\frac{mc_j - \gamma m}{\gamma m} \frac{1}{\psi \phi} \frac{1}{n_j^{\phi-1}} = X^{\phi-1} \frac{\rho-1}{\rho} - X^\phi$

# Calibration Details

Parameter	Value	Moment	Moment source
$\beta$	7.19	Top 10% Output share	Computestat: Standard & Poor's, 2020
$\sigma$	0.43	Top 10% Wage Premium	Wong, 2023
$\omega$	0.61	Labor Market Participation	BLS, 2024a, 1986 – 1999 average
$\psi$	3.09	TFP growth rate	BLS, 2024b, 1954 – 2007 average
$\phi$	1.48	R&D Spending (% of GDP)	World Bank, 2024, 1996
$\gamma$	1.26	Average Markup	Autor et al., 2020
$\zeta$	0.92	Profit share	BEA, 2024a, 1986 – 1999 average

Parameter	Value	Source
$\lambda$	0.103	Borella et al., 2022, 1969 – 1981 average
$\tau$	0.078	Borella et al., 2022, 1969 – 1981 average
$s_h$	1.49	Compustat: Standard & Poor's, 2020, $s_h/s_l$ , 1954 – 2007 average
$\eta$	0.32	BEA, 2024b, $G/Y$ , 1969 – 2007 average

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## Decomposition details

$$\begin{aligned} Y &= Q \cdot \exp \int_0^1 \ln s_{j(i)} di \cdot \exp \int_0^1 \ln l_i di \\ &\approx Q \cdot \exp \int_0^1 \ln s_{j(i)} di \cdot \left( \ln \bar{l} + \int_0^1 \frac{1}{\bar{l}} (l_i - \bar{l}) - \frac{1}{2\bar{l}^2} (l_i - \bar{l})^2 di \right) \\ &= Q \cdot S \cdot \left( 1 - \frac{CV^2}{2} \right) \int_0^1 l_i di \\ &= \underbrace{Q \cdot S \cdot \left( 1 - \frac{CV^2}{2} \right)}_{TFP} \cdot \sum_{j \in \mathcal{J}} L_j \\ &= Q \cdot S \cdot M \cdot L, \quad \text{where } M \text{ follows from price/markup dispersion:} \end{aligned}$$

$$M = \left( 1 - \frac{CV^2}{2} \right) = \left( \frac{3}{2} - \frac{\mathbb{E} \left( \frac{1}{(s_j m c_{j'})^2} \right)}{2 \cdot \mathbb{E} \left( \frac{1}{s_j m c_{j'}} \right)^2} \right)$$