Inequality and Misallocation under Production Networks

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Research Question & Motivation

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What is the effect of variations in the distributions of labor income and consumption expenditure on TFP?

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Upper Decile vs The Rest

Higher Expenditure Share in Education and Entertainment

Lower Expenditure Share in Shelter, Utilities, Healthcare

Data from Consumer Expenditure Survey

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Higher Expenditure Share in Education and Entertainment

Lower Expenditure Share in Shelter, Utilities, Healthcare

Data from Consumer Expenditure Survey

Income share for the top has increased

In this Presentation

In economies with distortions, variations in distributions (labor income & expenditure) can influence misallocation

Novel TFP decomposition and sufficient statistics that measures aggregate misallocations effects

Implementation of the model with US data: between 2010-2019, distributional sources of variation reduced TFP growth by 7.5%



Motivation

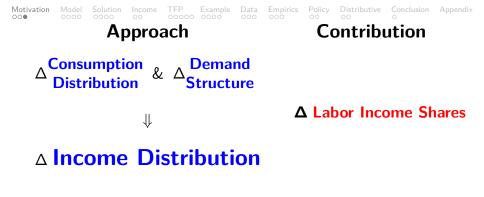


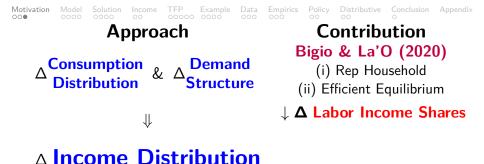
Δ Consumption Distribution & Δ Demand Structure

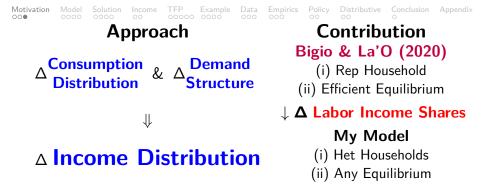
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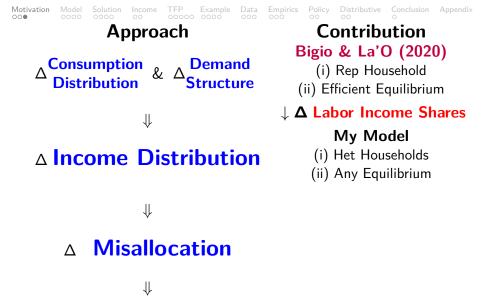
△ Income Distribution

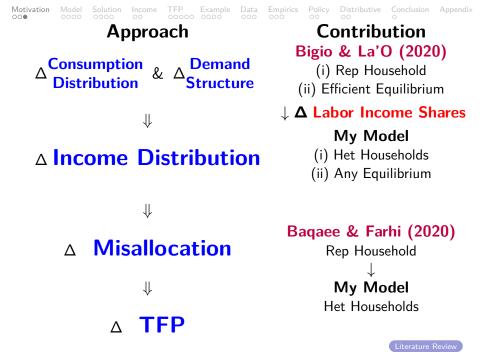
Literature Review











Static General Equilibrium Model with...

Two
FirmsMore Competitive: H
Less Competitive: LTwo
WorkersHigh-Skill: h
Low-Skill: ICaveat: Paper is more general than this case

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Static General Equilibrium Model with...

- Two
Firms : $\begin{cases} More Competitive: H \\ Less Competitive: L \end{cases}$ Two
Workers : $\begin{cases} High-Skill: h \\ Low-Skill: I \end{cases}$ Caveat:Paper is more general than this case
 - 1. Good markets face exogenous distortions

Cost
$$= \mu$$
 $imes$ Revenue

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Cost
$$= \mu \times Revenue$$

2. Labor markets are competitive

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Static General Equilibrium Model with...

- - 1. Good markets face exogenous distortions

Cost
$$= \mu \times Revenue$$

2. Labor markets are competitive

3. Correlations "HhH":

- \pmb{H} has high $\pmb{\mu}$
- *H* requires more *h*
- **h** have a higher expenditure in **H**

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Mechanism's Intuition

1. μ heterogeneity \longrightarrow allocates more workers to H

- H operates with low marginal productivity
- L operates with high marginal productivity

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Mechanism's Intuition

Model

1. μ heterogeneity \longrightarrow allocates more workers to H

- H operates with low marginal productivity
- L operates with high marginal productivity
- 2. Skill-bias heterogeneity \longrightarrow asymmetries in the income exposure in response to local perturbations
- 3. Preference heterogeneity expenditure flows
 - As **h** income increase, expenditure in **H** rises
 - Workers relocate from *L* to *H*
 - Misallocation is accentuated

$$\begin{array}{c} \begin{array}{c} \begin{array}{c} \text{Weight bindling} \\ \text{With bindling} \\ \text{With bindling} \\ \end{array} \end{array} \xrightarrow{} \begin{array}{c} \text{Time Heterogeneity} \\ i \in \{H, L\} \end{array} \end{array} \begin{array}{c} \begin{array}{c} \text{a. Skill Bias} \\ \text{b. Distortions} \end{array} \\ \begin{array}{c} \text{b. Distortions} \end{array} \end{array}$$

 $0 < \mu_L \leq \mu_H \leq 1$

$$Cost_i = \mu_i \times Revenue_i$$

Model 000 $\begin{array}{l} \mbox{Household} \\ \mbox{Heterogeneity} \\ \mbox{$r \in \{h, l\}$} \end{array} \begin{cases} \mbox{Preferences} & \rightarrow & \mbox{Aggregate} \\ \mbox{Non-Homotheticity} \\ \mbox{Graphic Argument} \\ \mbox{Unique Skill} & \rightarrow & \mbox{Structural Income} \\ \mbox{Heterogeneity} \\ \mbox{Heterogeneity} \\ \mbox{C}_{r, \ C_{rH}, \ C_{rL}} U_r(C_r) \ \mbox{s.t.} \ \box{$\frac{C_r}{\overline{C_r}} = \left(\beta_r \left(\frac{C_r \mu}{\overline{C_r \mu}} \right)^{\frac{\rho-1}{\rho}} + \left(1 - \beta_r \right) \left(\frac{C_r \underline{L}}{\overline{C_r L}} \right)^{\frac{\rho-1}{\rho}} \right)^{\frac{\rho-1}{\rho-1}} } \end{array} \end{cases}$ Non-Homotheticity

 $E_r = p_r^c C_r = p_H C_{rH} + p_L C_{rL} \leq w_r L_r + 0.5$ profits

Consumption Bias $\beta_l \leq \beta_h$

Solve for Equilibrium Distributions

From FOC of households and firms

 $p_H C_{rH} = \beta_r p_r^c C_r \qquad w_h \ell_{ih} = \alpha_i \mu_i p_i y_i$



Solve for Equilibrium Distributions

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In market clearing conditions

$$y_i = C_{hi} + C_{li} \qquad \qquad L_r = \ell_{Hr} + \ell_{Lr}$$

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Equilibrium in terms of

Solution

$$\begin{split} \boldsymbol{\lambda_i} &= \frac{p_i \, y_i}{GDP} \\ \text{Sales (Domar weights)} \\ \boldsymbol{\Lambda_h} &= \frac{w_h \, L_h}{GDP} = \sum \alpha_i \mu_i \lambda_i \\ \text{Labor income} \\ \text{Equilibrium Definition} \end{split}$$

 $\chi_{r} = \frac{p_{r}^{r} C_{r}}{GDP}$ Expenditure $\tilde{\Lambda}_{h} = \sum \alpha_{i} \lambda_{i}$ Value added

Necessary and sufficient conditions for equilibrium

Source of Misallocation

Parameter Space Restrictions

$$\alpha_H + \alpha_L = \beta_h + \beta_l = 1$$

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$$\frac{U_{C_{rH}}}{U_{C_{rL}}} = \frac{\mu_L}{\mu_H} \frac{d y_L/d \ell_{Lr}}{d y_H/d \ell_{Hr}}$$

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Additional Restriction

$$\mu_H + \mu_L = 1$$





What I Don't Do

- Misallocation literature distorted vs. efficient equilibrium
- Parametric assumptions (usually CD) ightarrow analytic TFP
- Evaluate how getting rid of distortions has an effect on TFP

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What I Do

- Local TFP Δ around distorted equilibrium to any shock
- Distributional $\Delta \rightarrow$ Misallocation $\Delta \rightarrow \Delta$ TFP
- To ilustrate: d log A



Local Variation to *d log A*

$$d \chi_h = \frac{(\alpha_H - \alpha_L)}{2} d \lambda_H$$

Expenditure elasticity requires $\alpha_H \neq \alpha_L$



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Sales elasticity requires $\rho \neq 1$

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 $\begin{cases} \text{Under } \rho > 1: \text{ consumers increase expenditure on } H \& \lambda_H \uparrow \\ \text{Under } \rho < 1: \text{ consumers increase expenditure on } L \& \lambda_H \downarrow \\ \hline \\ \text{Baumol's Cost Disease} \end{cases}$

Income Distribution & Bilateral Centralities

$$\boldsymbol{\Lambda}_{h} = \underbrace{\left(\underbrace{\alpha_{H} \ \mu_{H}}_{f_{H \to h}} \ \beta_{h} + \underbrace{\alpha_{L} \ \mu_{L}}_{f_{L \to h}} \ (1 - \beta_{h})\right)}_{+ \underbrace{\left(\alpha_{H} \ \mu_{H}}_{m_{H}} \ \beta_{l} + \widehat{\alpha_{L} \ \mu_{L}} \ (1 - \beta_{l})\right)}_{m_{l \to h}} \boldsymbol{\chi}_{l}$$

Comparative Statics $\Lambda_{h} = m_{h \to h} \chi_{h} + m_{l \to h} \chi_{l}$ $m_{r \to h} = \beta_{r} f_{H \to h} + (1 - \beta_{r}) f_{L \to h} \qquad f_{i \to h} = \alpha_{i} \mu_{i}$

Income 0. **Comparative Statics** $\Lambda_{h} = m_{h \rightarrow h} \chi_{h} + m_{l \rightarrow h} \chi_{l}$ $m_{r \to h} = \beta_r f_{H \to h} + (1 - \beta_r) f_{L \to h}$ $f_{i \to h} = \alpha_i \mu_i$ Take total derivative $\boldsymbol{d} \boldsymbol{\Lambda}_{h} = \chi_{h} \boldsymbol{d} \boldsymbol{m}_{h \to h} + \chi_{l} \boldsymbol{d} \boldsymbol{m}_{l \to h} + \boldsymbol{m}_{h \to h} \boldsymbol{d} \boldsymbol{\chi}_{h} + \boldsymbol{m}_{l \to h} \boldsymbol{d} \boldsymbol{\chi}_{l}$ **Income Centrality**_h **Distributive Income**_h

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Baqaee & Fahri (2020)

$$d \log TFP = \underbrace{\lambda_{H} d \log A_{H} + \lambda_{L} d \log A_{L}}_{\text{Technology}} \\ + \underbrace{\lambda_{H} d \log \mu_{H} + \lambda_{L} d \log \mu_{L}}_{\text{Competitiveness}} \\ - \underbrace{\left(\widetilde{\Lambda}_{h} d \log \Lambda_{h} + \widetilde{\Lambda}_{I} d \log \Lambda_{I}\right)}_{\text{Competitiveness}}$$

Misallocation

Without distortions \rightarrow Hulten (1978)

Distortion Centralities δ

$$\begin{aligned} \textbf{Misallocation} &= \tilde{\Lambda}_h \ d \ \log \Lambda_h + \tilde{\Lambda}_l \ d \ \log \Lambda_l \\ &= \frac{\tilde{\Lambda}_h}{\Lambda_h} \ d \ \Lambda_h + \frac{\tilde{\Lambda}_l}{\Lambda_l} \ d \ \Lambda_l \\ &= \delta_h \ d \ \Lambda_h + \delta_l \ d \ \Lambda_l \end{aligned}$$

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 δ measures how **undervalue** a worker is

Distortion Centralities δ

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δ measures how **undervalue** a worker is

$$\boldsymbol{\delta_{l}} - \boldsymbol{\delta_{h}} = \overbrace{(\mu_{H} - \mu_{L})}^{\geq 0} \overbrace{(\alpha_{H} - \alpha_{L})}^{\geq 0} \overbrace{\boldsymbol{a}}^{\geq 0}$$



Introduce $d \Lambda_h$ and $d \Lambda_l$

Misallocation = $(F_H - F_L) \sum \chi_r d\beta_r + (\underline{M_h - M_l}) d\chi_h$

1. Final Demand

2. Expenditure



Introduce $d \Lambda_h$ and $d \Lambda_l$

$$Misallocation = \underbrace{(F_H - F_L) \sum \chi_r \, d \, \beta_r}_{1 \text{ First Particle}} + \underbrace{(M_h - M_l) \, d \, \chi_h}_{2 \text{ First Particle}}$$

1. Final Demand

2. Expenditure

Sufficient Statistics

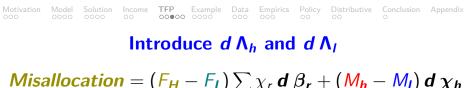
Revenue Centrality *F*

 $F_i = f_{i \to h} \, \delta_h + f_{i \to l} \, \delta_l$



$$Misallocation = \underbrace{(F_H - F_L) \sum \chi_r \, d \, \beta_r}_{1. \ Final \ Demand} + \underbrace{(M_h - M_l) \, d \, \chi_h}_{2. \ Expenditure}$$

Sufficient Statistics Revenue Centrality *F* Expenditure Centrality *M* $F_i = f_{i \rightarrow h} \, \delta_h + f_{i \rightarrow l} \, \delta_l$ $M_r = m_{r \rightarrow h} \, \delta_h + m_{r \rightarrow l} \, \delta_l$



$$Misallocation = \underbrace{(F_H - F_L) \sum \chi_r \, d \, \beta_r}_{1. \ Final \ Demand} + \underbrace{(M_h - M_l) \, d \, \chi_h}_{2. \ Expenditure}$$

Sufficient Statistics Revenue Centrality *F* Expenditure Centrality *M*

 $F_{i} = f_{i \to h} \, \delta_{h} + f_{i \to l} \, \delta_{l} \qquad M_{r} = m_{r \to h} \, \delta_{h} + m_{r \to l} \, \delta_{l}$

- **1.** F_i is high for firms that operate in relatively competitive supply chains and directly or indirectly demand high δ workers
- 2. M_r is high for households that consume from relatively competitive supply chains that demand workers with high δ

Final Demand $\uparrow \rightarrow$ Misallocation $\uparrow \rightarrow$ TFP \downarrow

 $(F_H - F_L) \sum \chi_r d \beta_r$

Final Demand

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Final Demand $\uparrow \rightarrow$ Misallocation $\uparrow \rightarrow$ TFP \downarrow

$$\underbrace{(F_H - F_L) \sum \chi_r \ \boldsymbol{d} \ \boldsymbol{\beta}_r}_{F_r}$$

Final Demand

$$F_{H} - F_{L} = \underbrace{\left(\begin{array}{c} \geq & 0 \\ (\mu_{H} - \mu_{L}) \end{array}\right)}_{\times \left[\delta_{I} - (\alpha_{H} - \alpha_{L}) (\alpha_{H} \mu_{H} - \alpha_{L} \mu_{L}) \mathbf{a} \right]}_{> & 0}$$

Final Demand $\uparrow \rightarrow$ Misallocation $\uparrow \rightarrow$ TFP \downarrow

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$$\begin{array}{c|c} \mu_{H} = \mu_{L} & \alpha_{H} = \alpha_{L} & \beta_{h} = \beta_{I} \\ \hline Final Demand \\ Recomposition & \checkmark & \checkmark & \checkmark \\ \end{array}$$



 $d\chi_h = \frac{(\alpha_H - \alpha_L)}{2} d\lambda_H$ $(M_h - M_l) d \chi_h$ Distributive



Expenditure $\uparrow \rightarrow$ Misallocation $\uparrow \rightarrow$ TFP \downarrow

$$(\underline{M_h - M_l}) d\chi_h$$

$$d \chi_h = \frac{(\alpha_H - \alpha_L)}{2} d \lambda_H$$

Distributive

$$\underbrace{\stackrel{\geq}{\mathbf{M_h} - \mathbf{M_l}}_{\mathbf{M_h} - \mathbf{M_l}} = \underbrace{\stackrel{\geq}{\mathbf{(\mu_H - \mu_L)}}_{\mathbf{(\mu_H - \mu_L)}} \underbrace{\stackrel{\geq}{\mathbf{(\beta_h - \beta_l)}}_{\mathbf{(\beta_h - \beta_l)}$$





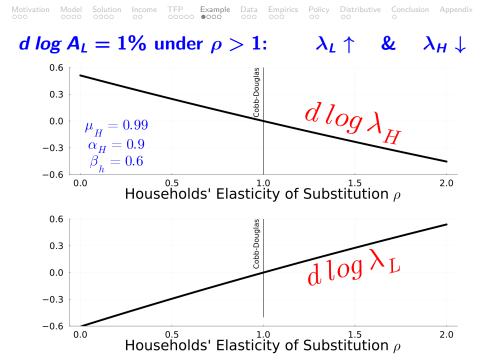
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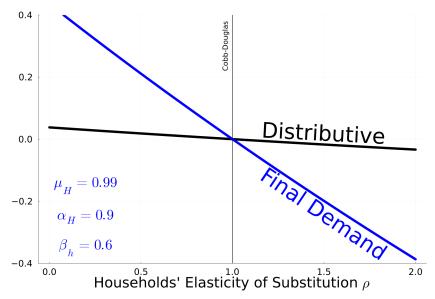
Distributive

$$\underbrace{\stackrel{\geq}{M_{h} - M_{l}}_{M_{h} - M_{l}} = \underbrace{\stackrel{\geq}{(\mu_{H} - \mu_{L})}_{(\mu_{H} - \mu_{L})} \underbrace{\stackrel{\geq}{(\beta_{h} - \beta_{l})}_{(\beta_{h} - \beta_{l})}_{\times \underbrace{[\delta_{l} - (\alpha_{H} - \alpha_{L})(\alpha_{H}\mu_{H} - \alpha_{L}\mu_{L}) \mathbf{a}]}_{> 0}}_{= 0}$$



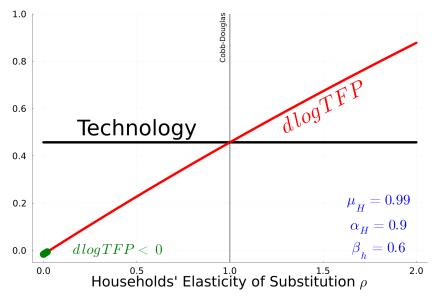


Under $\rho > 1$: Distributive \downarrow & Final Demand \downarrow



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Under $\rho > 1$: Misallocation \downarrow & d log TFP > λ_L





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- General Non-Parametric CRS model for production & consumption
- General Input-Output Networks
- General Equity Distribution

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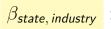
Additional Channels

- **1.** μ \uparrow & stronger for sectors with high λ_i F_i
- **2.** α \uparrow for high δ workers & stronger if $\mu_i \lambda_i$ high
- **3.** Intermediate demand \uparrow on sectors with high F_i

Data

Data for money flows from 1997 to 2021 Household to Firm

1. State level Personal Consumption Expenditure (BEA)



 $\frac{\beta_{state, industry}}{\beta_{state, industry}} : \begin{cases} PCE \text{ provides expenditure on types of goods} \\ IO \text{ Make matrix: type of good} \rightarrow \text{industry} \end{cases}$

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Data for money flows from 1997 to 2021 Household to Firm

1. State level Personal Consumption Expenditure (BEA)

 $\beta_{\textit{state, industry}}: \left\{ \begin{array}{c} \mathsf{PCE} \text{ provides expenditure on types of goods} \\ \mathsf{IO} \text{ Make matrix: type of good} \rightarrow \mathsf{industry} \end{array} \right.$

Firm to Firm

2. Input-Output tables (BEA) for 66 NAICS industries

$$\mu_i = \frac{\text{Total Cost}_i}{\text{Sales}_i}$$
 Intermediate
Intensity $_{ij} = \frac{p_j x_{ij}}{\text{Total Cost}_i}$

Total $Cost_i = Labor Costs_i + Intermediate Cost_i$



3. Industry Level Production Accounts (BEA)

 $d \log A_i$

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4. County Business Patterns (Census) Industry specific geographic (state) bias in labor

Antisupression Algorithm Missing Private Employment

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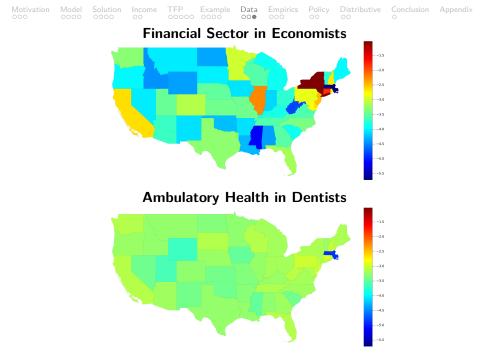
4. County Business Patterns (Census) Industry specific geographic (state) bias in labor

Antisupression Algorithm Missing Private Employment

5. Occupational Employment & Wage Statistics

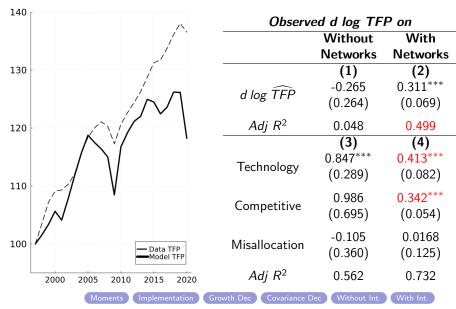
- Industry specific occupational demand bias
- State specific occupational supply bias

From 4 & 5 \rightarrow industry specific heterogeneity by worker type. Worker type comes from State & Occupational interactions $H = 38,189 \ (\approx 1.5 \text{ bill } m's)$



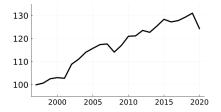
Motivation Model Solution Income TFP Example Data Construction October Conclusion Appendix October Conclusion Conclusio

R^2 rises from 5% to 50% with IO Networks

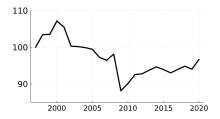


 $d \log TFP = \text{Technology} + \text{Competitiveness} - \text{Misallocation}$ $\begin{array}{c} \textbf{Technology} \uparrow & \textbf{Competitiveness} \downarrow \\ \sum_{i \in \mathcal{N}} \widetilde{\lambda}_i \ d \log A_i & \sum_{i \in \mathcal{N}} \widetilde{\lambda}_i \ d \log \mu_i \end{array}$

Empirics



Between 1997 and 2020Oil & gas extraction-11.1%Computer & electronic-6.6%

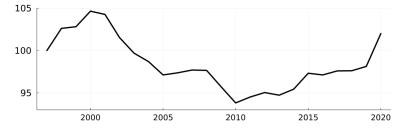


Between 1997 and	2020			
Credit intermediation	4.1%			
Between 2002 and 2009				
Between 2002 and	2009			

Without Misallocation[↑] after 2009, TFP[↑] 7.5%

Empirics

- Misallocation↓ between 2001 and 2010 by -8.2%
- Misallocation↑ between 2010 and 2020 by 7.5%



Increasing profit margins

- Oil & gas extraction: -1.5%
- Computer & electronics: -1.1%

Increasing labor demand

• Credit intermediation: 2.4%

Final and Intermediate Demand

• Wholesale Trade: 2.2%

Distribution of δ Sources of Misallocation (Graph) Sources of Misallocation (Counterfactual)

Industry variation from μ and Labor Demand) Industry variation from Final and Intermediate Demand

Normalized Nested CES

Policy

Introduced by de La Grandville (1989) and Klump & de La Grandville (1989) and as in Baqaee & Farhi (2019,a,b, 2020, 2022)

Normalized Nested CES

Parameters - Atalay (2017), Boehm et al. (2014)

- 1. Elasticity of substitution between worker types: 1.0
- 2. Elasticity of substitution between sectoral intermediate inputs: 0.2
- 3. Elasticity of substitution between labor and intermediate inputs: 0.5
- 4. Elasticity of substitution in final consumption: 0.9
- 5. Substitution effect in labor supply: 2
- 6. Income effect in labor supply: 2

$d \log TFP = \underbrace{\text{Technology}}_{=1\%} - Misallocation$

Best Sectors	d log TFP						
1. Nursing & Residential Care	1.041%		0.359***	d log T	FP on	0.207	
2. Social Assistance	1.039%	μ_i	(0.09)			(0.13)	
3. General Merchandise Store	1.029%			0.170		0.854*	
4. Ambulatory health care	1.027%	λ_i		(0.56)		(0.50)	
5. Hospitals	1.026%	Fi			0.212***	0.148**	
		Γ_i			(0.05)	(0.07)	
Worst Sectors	d log TFP	R^2	0.00	$1e^{-3}$	0.21	0.07	
1. Oil & Gas extraction	0.587%		0.20	le -	0.21	0.27	
		N			66		
2. Primary Metals	0.610%						
3. Chemical Products	0.618%						
4. Mining, except Oil & Gas	0.630%	We want productivity shocks in					

5. Utilities 0.647%

We want productivity shocks in sectors with high *F_i*!

ANTITRUST POLICY: Shocks in Markdowns

Positional Terms of Trade

 $C_r = \mathbf{PTT}_r \times f_r(L_h, L_l)$

Positional Terms of Trade

 $C_r = \mathbf{PTT}_r \times f_r(L_h, L_l)$

d log TFP = $\sum \chi_r d \log \mathsf{PTT}_r$

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Positional Terms of Trade

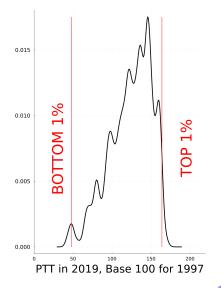
$$C_r = \mathbf{PTT}_r \times f_r(L_h, L_l)$$

$d \log TFP = \sum \chi_r d \log \mathbf{PTT}_r$

$$d \log PTT_{h} = \underbrace{\beta_{h} d \log A_{H} + (1 - \beta_{h}) d \log A_{L}}_{\text{Technology}_{h}} + \underbrace{\beta_{h} d \log \mu_{H} + (1 - \beta_{h}) d \log \mu_{L}}_{\text{Competitiveness}_{h}} - \underbrace{\left(\frac{\widetilde{m}_{h \leftarrow h}}{\Lambda_{h}} d \Lambda_{h} + \frac{\widetilde{m}_{h \leftarrow I}}{\Lambda_{I}} d \Lambda_{I} - d \log \chi_{h}\right)}_{\text{Misallocation}_{h}}$$



C_r = Positional Terms of Trade_r × $f_r(L_h, L_l)$



Top 1% Occupation					
Computer Occupations	13%				
Mathematical Sciences Occupations	10%				
Compensation Managers	7%				
Bottom 1%					
Occupation					
Printing Workers	40%				
Shoe & Leather Operator	26%				
Textile Machine Operator	15%				
Miscellaneous Textile	12%				

COUNTERFACTUAL SHOOCK: Antitrust Policy in Housing

Conclusion

- First comprehensive study for join heterogeneity in multisector economies with distortions and input-output networks
- **Theoretical Contribution** in production network + distortions + heterogeneous households:
 - Variation of the income distribution
 - Variations for TFP
 - Variations for PTT
- **Empirical Contribution**: First implementation of a production network model with household heterogeneity for the US
 - In the absence of distributional sources of misallocation, TFP would have grown 7.5% more after Great Recession

Upper Decile vs The Rest (Consumer Expenditure Survey 2021)

Higher Expenditure Share in

Lower Expenditure Share in

- Education: 3.4% vs 1.3%
- Entertainment: 6.5% vs 4.9%
- Pensions: 17.4% vs 9.1%
- Lodging: 2.6% vs 1.1%

- Shelter: 17.6% vs 20.5%
- Home Food: 5.9% vs 8.5%
- Utilities: 4.1% vs 7.0%
- Healthcare: 6.2% vs 8.3%

From 2004 to 2019

Income share for top quintile \uparrow from 48% to 53%

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Literature Review

Diggregated National Accounts

Cantillon (1756), Quesnay (1758), Leontief (1928), Meade & Stone (1941), Kuznetz (1946), Stone (1961), Andersen et al. (2022)

Production Networks

Hulten (1978), Long & Plosset (1983), Gabaix (2011), Jones (2011, 2013), Acemoglu et al. (2012), Baqaee (2018), Baqaee & Farhi (2019, 2020, 2023), Bigio & La'O (2020)

Growth Accounting

Solow (1957), Domar (1961), Jorgenson et al. (1987), Basu & Fernanld (2022), Petrin & Levinsohn (2012), Baqaee & Farhi (2020)



Dixit-Stiglitz Aggregation

• Sector *i* has a sectoral aggregator for $z_i \in [0, 1]$

$$y_i = \left(\int y_{z_i}^{\boldsymbol{\mu_i}} d z_i\right)^{\frac{1}{\boldsymbol{\mu_i}}}$$

Demand for variaties

$$y_{z_i} = \left(\frac{p_i}{p_{z_i}}\right)^{\frac{1}{1-\mu_i}} y_i$$

Intermediate's problem

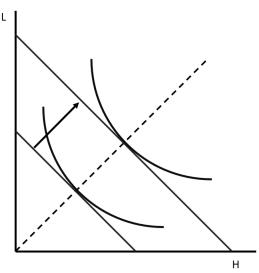
$$\underset{p_{z_i}, y_{z_i}, \ell_{z_i h}, \ell_{z_i I}}{\text{Max}} \quad \pi_{z_i} = p_{z_i} y_{z_i} - w_h \ell_{z_i h} - w_I \ell_{z_i I}$$

$$y_{z_i} = A_i \ \ell_{z_i \ \boldsymbol{h}}^{\boldsymbol{\alpha}_i} \ \ell_{z_i \ \boldsymbol{l}}^{1-\boldsymbol{\alpha}_i}$$
 (Back

Appendix



Aggregate Non-Homotheticity





Equilibrium Definition

For $e = (A, \mu, \beta, \alpha) \in \mathscr{E}$, prices and allocations:

- (i) **Firms'** labor demand and output decisions maximize profits;
- (ii) **Households'** consumption and labor supply maximize utility satisfying budget constraints;

(iii) Goods and labor markets clear.

Equilibrium Definition

$$e = (A, \mu, eta, lpha) \in \mathscr{E}$$
 into

$$\vartheta \equiv \left\{ \left\{ \mathbf{y}_{i}, \left\{ \ell_{ir}, C_{ri} \right\}_{r \in \{h, l\}} \right\}_{i \in \{H, L\}}, \left\{ C_{r}, L_{r} \right\}_{r \in \{h, l\}} \right\}$$

$$\rho \equiv \{p_H, p_L, w_h, w_l, p_h^c, p_l^c\}$$



Necessary & sufficient equilibrium conditions

(artheta, ho) are an equilibrium iff

 $\begin{aligned} \frac{U_{C_{rj}}}{U_{C_{ri}}} &= \frac{\mu_i \ \partial \ y_i / \partial \ \ell_{ib}}{\mu_j \ \partial \ y_j / \partial \ \ell_{jb}} \quad i,j \in \{H,L\} \ , r,b \in \{h,I\} \ ,\\ \text{such that } C_{ri} &> 0, \text{ and } \ell_{ib} > 0, \end{aligned}$

and resource constraints

$$egin{aligned} y_i\left(e
ight) &= C_{hi}\left(e
ight) + C_{li}\left(e
ight) & i \in \{H,L\} \ L_r\left(e
ight) &= \ell_{Hr}\left(e
ight) + \ell_{Lr}\left(e
ight) & r \in \{h,I\} \,. \end{aligned}$$





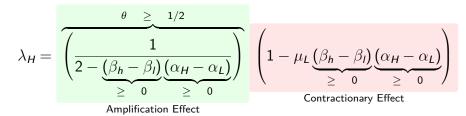
Sales Share

$$\lambda_H = \beta_h \chi_h + \beta_l \chi_l \qquad \lambda_L = 1 - \lambda_H$$

$$\begin{array}{c} \text{Model Solution hoome TFP concerces Frample Data Empirics Policy Distributive Conclusion Appendix}\\ \hline \textbf{Sales Share}\\ \hline \lambda_{H} = \beta_{h} \ \chi_{h} + \beta_{I} \ \chi_{I} \qquad \lambda_{L} = 1 - \lambda_{H}\\ \hline \textbf{Labor Income Share}\\ \hline \Lambda_{h} = \alpha_{H} \ \mu_{H} \ \lambda_{H} + \alpha_{L} \ \mu_{L} \ \lambda_{L} \end{array}$$



Sales Distribution



Consumption Expenditure Distribution

$$\chi_h = \theta \left(1 - \underbrace{(\alpha_H - \alpha_L)}_{\geq 0} \underbrace{(\beta_h - \mu_H)}_{?} \right)$$



Labor Income Distribution

$$\Lambda_{h} = \theta \left[\alpha_{H}\mu_{H} + \alpha_{L}\mu_{L} - \mu_{H}\mu_{L} \underbrace{(\beta_{h} - \beta_{l})}_{\geq 0} \underbrace{(\alpha_{H} - \alpha_{L})}_{\geq 0} \right]$$
$$\Lambda_{l} = \theta \left[\alpha_{L}\mu_{H} + \alpha_{H}\mu_{L} - \mu_{H}\mu_{L} \underbrace{(\beta_{h} - \beta_{l})}_{\geq 0} \underbrace{(\alpha_{H} - \alpha_{L})}_{\geq 0} \right]$$

Value-Added Distribution Back

$$\widetilde{\Lambda}_{h} = \alpha_{H}\lambda_{H} + \alpha_{L}\lambda_{L}$$

$$= \theta \Big(1 - \underbrace{(\beta_{h} - \beta_{I})}_{\geq 0} \underbrace{(\alpha_{H} - \alpha_{L})}_{\geq 0} \Big(\alpha_{H}\mu_{L} + \alpha_{L}\mu_{H} \Big) \Big)$$

$$\widetilde{\Lambda}_{I} = \theta \Big(1 - \underbrace{(\beta_{h} - \beta_{I})}_{\geq 0} \underbrace{(\alpha_{H} - \alpha_{L})}_{\geq 0} \Big(\alpha_{H}\mu_{H} + \alpha_{L}\mu_{L} \Big) \Big)$$

3 Effects from Distortions on Labor

1. Misallocation comes from *MRS* wedges

$$\frac{U_{C_{rH}}}{U_{C_{rL}}} = \frac{\mu_L}{\mu_H} \frac{d y_L/d \ell_{Lr}}{d y_H/d \ell_{Hr}}$$



3 Effects from Distortions on Labor

1. Misallocation comes from *MRS* wedges

$$\frac{U_{C_{rH}}}{U_{C_{rL}}} = \frac{\mu_L}{\mu_H} \frac{d y_L / d \ell_{Lr}}{d y_H / d \ell_{Hr}}$$

2. Allocative differences \neq **Misallocation**

$$\frac{\ell_{Hh}}{L_h} \neq \alpha_H$$

 $\begin{array}{c} \mbox{Intuition}\\ \mbox{For the undistorted case}\\ \mu_{H}{=}\mu_{L}{=}1/2\\ \mbox{there is a continuum}\\ \mbox{of property rights on firms} \end{array}$





Data Empirics Policy Distributive Conclusion Appendix

 $\frac{\ell_{Hh}}{L_H} \neq \alpha_H \quad \text{not the same} \quad \frac{U_{C_{rH}}}{U_{C_{rL}}} = \frac{\mu_L}{\mu_H} \frac{d y_L/d \ell_{Lr}}{d y_H/d \ell_{Hr}}$ Misallocation



$$\frac{\ell_{Hh}}{L_{H}} \neq \alpha_{H} \text{ not the same } \underbrace{\frac{U_{C_{rH}}}{U_{C_{rL}}} = \frac{\mu_{L}}{\mu_{H}} \frac{d y_{L}/d \ell_{Lr}}{d y_{H}/d \ell_{Hr}}}_{\text{Misallocation}}$$
Case 1 Case 2 Case 3 Case 4
$$\mu_{H} = \mu_{L}$$
Symmetric π All π for h



$$\frac{\ell_{Hh}}{L_{H}} \neq \alpha_{H} \quad \text{not the same} \quad \underbrace{\frac{U_{C_{rH}}}{U_{C_{rL}}} = \frac{\mu_{L}}{\mu_{H}} \frac{d y_{L}/d \ell_{Lr}}{d y_{H}/d \ell_{Hr}}}_{\text{Misallocation}}}$$
Case 1 Case 2 Case 3 Case 4
$$\mu_{H} = \mu_{L}$$
Symmetric π All π for h

$$\frac{\ell_{Hh}}{L_{h}} = \alpha_{H} \quad \alpha_{H} + \alpha_{H}\alpha_{L} \left(\beta_{h} - \beta_{l}\right)$$



$$\frac{\ell_{Hh}}{L_{H}} \neq \alpha_{H} \text{ not the same } \underbrace{\frac{U_{C_{rH}}}{U_{C_{rL}}} = \frac{\mu_{L}}{\mu_{H}} \frac{d y_{L}/d \ell_{Lr}}{d y_{H}/d \ell_{Hr}}}_{\text{Misallocation}}$$
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$$\frac{\ell_{Hh}}{L_h} = \alpha_H \quad \alpha_H + \alpha_H \alpha_L \left(\beta_h - \beta_l\right)$$

$$\frac{U_{C_{rH}}}{U_{C_{rL}}} = \frac{d y_L/d \ell_{Lr}}{d y_H/d \ell_{Hr}}$$

$$\frac{\ell_{Hh}}{L_{H}} \neq \alpha_{H} \text{ not the same } \underbrace{\frac{U_{C_{rH}}}{U_{C_{rL}}} = \frac{\mu_{L}}{\mu_{H}} \frac{d y_{L}/d \ell_{Lr}}{d y_{H}/d \ell_{Hr}}}_{\text{Misallocation}}$$
Case 1 Case 2 Case 3 Case 4
$$\mu_{H} = \mu_{L}$$
Symmetric π All π for h

$$\frac{\ell_{Hh}}{L_h} = \alpha_H \quad \alpha_H + \alpha_H \alpha_L \left(\beta_h - \beta_l\right)$$

$$\frac{U_{C_{rH}}}{U_{C_{rL}}} = \frac{d y_L/d \ell_{Lr}}{d y_H/d \ell_{Hr}}$$

$$\frac{\ell_{Hh}}{L_{H}} \neq \alpha_{H} \text{ not the same } \underbrace{\frac{U_{C_{rH}}}{U_{C_{rL}}} = \frac{\mu_{L}}{\mu_{H}} \frac{d y_{L}/d \ell_{Lr}}{d y_{H}/d \ell_{Hr}}}_{\text{Misallocation}}$$
Case 1 Case 2 Case 3 Case 4
$$\mu_{H} = \mu_{L}$$
Symmetric π All π for h Symmetric π

$$\frac{\ell_{Hh}}{L_h} = \alpha_H \quad \alpha_H + \alpha_H \alpha_L \left(\beta_h - \beta_l\right)$$

$$\frac{U_{C_{rH}}}{U_{C_{rL}}} = \frac{d y_L/d \ell_{Lr}}{d y_H/d \ell_{Hr}}$$

$$\frac{\ell_{Hh}}{L_{H}} \neq \alpha_{H} \text{ not the same } \underbrace{\underbrace{\bigcup_{C_{rH}}}_{U_{C_{rL}}} = \frac{\mu_{L}}{\mu_{H}} \frac{d y_{L}/d \ell_{Lr}}{d y_{H}/d \ell_{Hr}}}_{\text{Misallocation}}$$
Case 1 Case 2 Case 3 Case 4
$$\mu_{H} = \mu_{L} \qquad \alpha_{H} = \alpha_{L} \quad \beta_{h} = \beta_{I}$$
Symmetric π All π for h Symmetric π

$$\frac{\ell_{Hh}}{L_{h}} = \alpha_{H} \quad \alpha_{H} + \alpha_{H}\alpha_{L}(\beta_{h} - \beta_{I}) \qquad \frac{\ell_{Hh}}{L_{h}} > \alpha_{H}$$

$$\frac{U_{C_{rH}}}{U_{C_{rL}}} = \frac{d y_L/d \ell_{Lr}}{d y_H/d \ell_{Hr}}$$

$$\frac{\ell_{Hh}}{L_{H}} \neq \alpha_{H} \text{ not the same } \underbrace{\underbrace{U_{C_{rH}}}_{U_{C_{rL}}} = \frac{\mu_{L}}{\mu_{H}} \frac{d y_{L}/d \ell_{Lr}}{d y_{H}/d \ell_{Hr}}}_{\text{Misallocation}}$$
Case 1 Case 2 Case 3 Case 4
$$\mu_{H} = \mu_{L}$$
Symmetric π All π for h Symmetric π

$$\frac{\ell_{Hh}}{L_h} = \alpha_H \quad \alpha_H + \alpha_H \alpha_L \left(\beta_h - \beta_l\right) \qquad \frac{\ell_{Hh}}{L_h} > \alpha_H$$
$$\frac{U_{C_{rH}}}{U_{C_{rL}}} = \frac{d_{YL}/d\,\ell_{Lr}}{d_{YH}/d\,\ell_{Hr}} \qquad \frac{U_{C_{rH}}}{U_{C_{rL}}} = \frac{\mu_L}{\mu_H} \frac{d_{YL}/d\,\ell_{Lr}}{d_{YH}/d\,\ell_{Hr}}$$



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Linear Approximation in response to $d \log A_H / A_L$

$$d \Lambda_h = (\alpha_H - \mu_L) d \lambda_H, \qquad d \Lambda_I = (\mu_H - \alpha_H) d \lambda_H,$$

$$d \chi_h = d \Lambda_h + \frac{1}{2} (\mu_L - \mu_H) d \lambda_H,$$

$$d \lambda_{H} = (\rho - 1) (\chi_{h} - \chi_{l}) \beta_{h} \beta_{l} d \log \frac{A_{H}}{A_{L}} + (\beta_{h} - \beta_{l}) d \chi_{h}$$
$$+ (\rho - 1) (\alpha_{H} - \alpha_{L}) \beta_{h} \beta_{l} d \log \frac{\Lambda_{l}}{\Lambda_{h}}$$

Constant
$$\Phi$$
 in $d\lambda_H = (\rho - 1) (\chi_h - \chi_l) \beta_h \beta_l \Phi d \log \frac{A_H}{A_l}$

$$\Phi = \frac{1 + \Upsilon \left(\phi_h \phi_l + (\rho - 1)\beta_h \beta_l (\alpha_H - \alpha_L)(\alpha_H - \mu_H)(\alpha_H - \mu_L)((\beta_h - \beta_l)\Lambda_h - (\rho - 1)\beta_h \beta_l (\alpha_H - \alpha_L))\right)^{-1}}{1 + \frac{1}{2} \left(\beta_h - \beta_l\right) \left(\mu_H - \mu_L\right)}$$

with

$$\begin{split} \Upsilon &= ((\beta_h - \beta_l)\Lambda_h - (\rho - 1)\beta_h\beta_l(\alpha_H - \alpha_L))(\alpha_H - \mu_L)\Lambda_l \left(1 + \frac{1}{2}(\beta_h - \beta_l)(\mu_H - \mu_L)\right) \\ &+ (\rho - 1)\beta_h\beta_l(\alpha_H - \alpha_L)(\mu_H - \alpha_H)\Lambda_h \left(1 - \frac{1}{2}(\beta_h - \beta_l)(\alpha_H - \alpha_L)\right), \end{split}$$

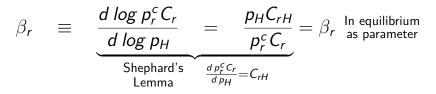
$$\begin{split} \phi_h &= \Lambda_h + \left(\beta_h - \beta_l\right) \Lambda_h \left(\mu_L - \alpha_H + \frac{1}{2} \left(\mu_H - \mu_L\right)\right) + \left(\varrho - 1\right) \beta_h \beta_l \left(\alpha_H - \alpha_L\right) \left(\alpha_H - \mu_L\right), \\ \phi_l &= \Lambda_l + \left(\beta_h - \beta_l\right) \frac{\Lambda_l}{2} \left(\mu_H - \mu_L\right) + \left(\varrho - 1\right) \beta_h \beta_l \left(\alpha_H - \alpha_L\right) \left(\alpha_H - \mu_H\right). \end{split}$$



β_{r}	≡	$\frac{d \log p_r^c C_r}{d \log p_H}$	$= \frac{p_H C_{rH}}{p_r^c C_r} = \beta_r$	In equilibrium as parameter
		Shephard's Lemma	$\frac{d p_r^c C_r}{d p_H} = C_{rH}$	



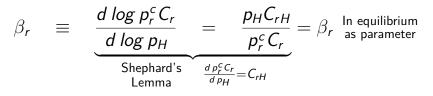




1. New equilibrium with local approximations keep α and β fixed





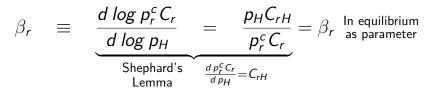


1. New equilibrium with local approximations keep α and β fixed

2. Estimate β 's consistent with the new equilibrium







1. New equilibrium with local approximations keep α and β fixed

2. Estimate β 's consistent with the new equilibrium

Exact delta hat - Dekle, Eaton & Kortum (2008)

$$\frac{p_{H}C_{rH}}{E_{r}} = \beta_{r}^{\rho} \left(\frac{p_{r}^{c}\overline{C}_{r}}{p_{H}\overline{C}_{rH}}\right)^{\rho-1} \rightarrow \underbrace{d \beta_{r} = (\rho-1)\beta_{r} (1-\beta_{r}) \ d \log \frac{p_{L}}{p_{H}}}_{P_{H}}$$

Increases under substitutability when p_L/p_H \uparrow

Theorem 1: labor income share variation

$$\boldsymbol{d} \ \boldsymbol{\Lambda}_{l} = \underbrace{\underbrace{(m_{h \to l} - m_{l \to l})}_{\text{Distributive Income}_{l}}^{2} \boldsymbol{d} \ \boldsymbol{\chi}_{h}}_{\text{Distributive Income}_{l}} + \underbrace{\underbrace{(\mu_{H} - \alpha_{H})}_{(\mu_{H} - \alpha_{H})}}_{\text{Income Centrality}_{l}} \sum \chi_{r} \ \boldsymbol{d} \ \boldsymbol{\beta}_{r}$$

Labor Wedge

For factors with endogenous supply...

$$-\frac{U_{L_h}}{U_{C_h}} = \Gamma_h \frac{C_h}{L_h} \qquad \text{with} \qquad \Gamma_h = \frac{\Lambda_h}{\chi_h}$$
Proof

$d \log \Gamma_h$ - Extension of Bigio & La'O (2020)

(i) Representative Household (i) Around Efficient Equilibrium \longrightarrow (i) Heterogenous Households (ii) Any Equilibrium

$$d \log \Gamma_h = d \log \Lambda_h - d \log \chi_h$$



Distributive Conclusion Appendix

Proof of Theorem 1 for $d \log \Gamma_h$

Distributive Conclusion Appendix

From goods market clearing

$$\begin{pmatrix} y_H \\ y_L \end{pmatrix} = \begin{pmatrix} C_{hH} + C_{IH} \\ C_{hL} + C_{IL} \end{pmatrix}$$

$$\begin{pmatrix} \frac{\beta_{h}}{C_{hH}} y_{H} \\ \frac{(1-\beta_{h})}{C_{hL}} y_{L} \end{pmatrix} = \begin{pmatrix} \frac{\beta_{h}}{C_{hH}} (C_{hH} + C_{IH}) \\ \frac{(1-\beta_{h})}{C_{hL}} (C_{hL} + C_{IL}) \end{pmatrix}$$

From FOC and equilibrium $\beta_h \frac{\chi_h}{C_{hH}} = p_H = \beta_I \frac{\chi_I}{C_{IH}}$

$$\begin{pmatrix} \frac{\beta_{h}}{C_{hH}} \, \mathcal{Y}_{H} \\ \frac{(1-\beta_{h})}{C_{hH}} \, \mathcal{Y}_{L} \end{pmatrix} = \begin{pmatrix} \beta_{h} \frac{\chi_{h}}{\chi_{h}} + \beta_{l} \frac{\chi_{l}}{\chi_{h}} \\ (1-\beta_{h}) \frac{\chi_{h}}{\chi_{h}} + (1-\beta_{l}) \frac{\chi_{l}}{\chi_{h}} \end{pmatrix}$$

Proof of Theorem 1 for $d \log \Gamma_h$

Distributive Conclusion Appendix

From FOC and equilibrium $-\frac{1}{\beta_h}\frac{U_{L_h}}{U_{C_h}}\frac{C_{hH}}{C_h} = \frac{w_h}{p_H} = \mu_H \alpha_H \frac{y_H}{\ell_{Hh}}$

$$\begin{pmatrix} \boldsymbol{\ell}_{Hh} \\ \boldsymbol{\ell}_{Lh} \end{pmatrix} = \begin{pmatrix} -\frac{U_{C_h}}{U_{L_h}} \alpha_H \mu_H y_H \beta_h \frac{C_h}{C_{hH}} \\ -\frac{U_{C_h}}{U_{L_h}} \alpha_L \mu_L y_L (1 - \beta_h) \frac{C_h}{C_{hL}} \end{pmatrix}$$

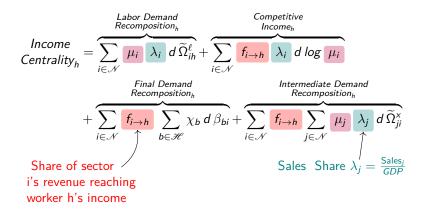
From labor market clearing condition

$$L_{h} = \ell_{Hh} + \ell_{Lh} = -\frac{U_{C_{h}}}{U_{L_{h}}} C_{h} \left(\alpha_{H} \mu_{H} \quad \alpha_{L} \mu_{L} \right) \left(\frac{\frac{\beta_{h}}{C_{hH}} y_{H}}{\frac{(1-\beta_{h})}{C_{hH}} y_{L}} \right)$$

$$= -\frac{U_{C_h}}{U_{L_h}}C_h \underbrace{\left(\alpha_H \,\mu_H \sum_{r \in \{h,l\}} \beta_r \frac{\chi_r}{\chi_h} + \alpha_L \,\mu_L \sum_{r \in \{h,l\}} \left(1 - \beta_r\right) \frac{\chi_r}{\chi_h}\right)}_{= \Gamma_h}_{\text{Back}}$$

Motivation ooo Solution Income TFP Example Data Empirics Policy Distributive Conclusion Appendix

Income Centrality







Δ TFP

Α.

$GDP = P_Y Y = p_h^c C_h + p_l^c C_l$

$d \log GDP = \chi_h d \log p_h^c C_h + \chi_l d \log p_l^c C_l$



Δ TFP

Α.

$$GDP = P_Y Y = p_h^c C_h + p_l^c C_l$$

$$d \log GDP = \chi_h d \log p_h^c C_h + \chi_l d \log p_l^c C_l$$

B. Divisia Index GDP deflator



Motivation Model Solution Income TFP Example Data Empirics Policy Distributive Conclusion Appendix

Additional Steps for $d \log P_Y$

Start from

$$p_{H} = \frac{w_{h} \ell_{Hh} + w_{I} \ell_{HI}}{\mu_{H} A_{H} \ell_{Hh}^{\alpha_{H}} \ell_{HI}^{1-\alpha_{H}}}$$

Take first-order approximation

$$\hat{p}_{H} = -\hat{A}_{H} - \hat{\mu}_{H} + \alpha_{H} \hat{\alpha}_{Hh} + (1 - \alpha_{H}) \hat{\alpha}_{HI}$$

Do the same for bundle prices

$$\widehat{p}_{h}^{c} = -\beta_{h}\left(\widehat{A}_{H} + \widehat{\mu}_{H}\right) - (1 - \beta_{h})\left(\widehat{A}_{L} + \widehat{\mu}_{L}\right) + \widetilde{h}_{h}\widehat{w}_{h} + \widetilde{h}_{h}\widehat{w}_{h}$$

Back

Motivation Model Solution Income TFP Example Data Empirics Policy Distributive Conclusion Appendix

Distortion Centrality Heterogeneity

$$d \Lambda = d \Lambda_h + d \Lambda_l$$

$$Misallocation = \overbrace{(\delta_l - \delta_h)}^{\geq 0} d \Lambda_l + \delta_h d \Lambda$$

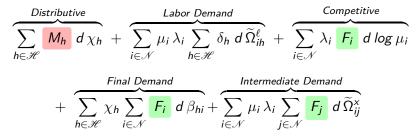
$$\boldsymbol{\delta_{l}} - \boldsymbol{\delta_{h}} = \overbrace{\left(\boldsymbol{\mu_{H}} - \boldsymbol{\mu_{L}}\right)}^{\geq 0} \overbrace{\left(\boldsymbol{\alpha_{H}} - \boldsymbol{\alpha_{L}}\right)}^{\geq 0} \overbrace{\boldsymbol{a}}^{\geq 0}$$

$$\boldsymbol{a} = \frac{1 - (\alpha_H - \alpha_L) \left(1 - (\beta_h - \beta_I) \overline{\mu_H \mu_L (\alpha_H - \alpha_L) (\beta_h - \beta_I)} \right)}{(\alpha_H \mu_H + \alpha_L \mu_L - \boldsymbol{b}) (\alpha_L \mu_H + \alpha_H \mu_L - \boldsymbol{b})}$$

Misallocation Decomposition

1. Misallocation \uparrow as expenditure rises for households with high M_h

- 2. Misallocation \uparrow as labor demand for workers with high δ rises
- **3.** Misallocation \uparrow as profit margins fall in sector with high F_i



4. Misallocation \uparrow as demand of goods \uparrow from sectors with high F_i

Appendix



Antisupression Algorithm

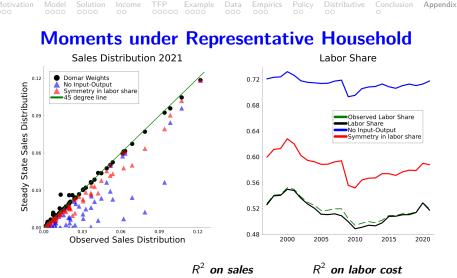
- 1. Significant portion of data supressed to protect confidentiality
- **2.** Since 2007 non-suppressed observations have a random noise infusion multiplier
- 3. Use information available due to to the industrial and geographical hierarchical nature \rightarrow manifold of bound and aggregation constraints across hierarchies
- 4. Two gold standards:
 - i. Two-staged algorithm from Isserman & Westervelt (2006)
 - ii. Linear programming solution from Eckert et al. (2020)
- These two methods estimate the number of workers, not their compensation. I develop a three-staged algorithm that starting from the guess Eckert et al. (2020) extends Isserman & Westervelt (2006) to the estimation of labor compensation



Missing Private Employment

- 1. The CBP only covers some forms of private employment
- 2. It does not include workers in
 - Agriculture production
 - Railroads
 - Government
 - Private household
- **3.** To fill this gap, I use the BEA's Regional Economic Information System to obtain state-level employment and income measures for agricultural and production workers
- 4. Data sources for REIS are the Quarterly Census of Employment and Statistics from the BLS
- 5. Main limitation from REIS is that it is only provided at the 2-digit NAICS level





	distribution	share
Base Model	0.994	0.981
No Input-Output	0.730	0.733
Symmetry in Labor	0.978	0.933



Contribution from each component

Table: Counterfactual TFP Growth Differential in the Absence of Components

A. Between 1997 and 2020					
Technology	Competitiveness	Misallocation			
-23.4%	2.5%	2.8%			
	B. Between 2002 and 200	19			
Technology	Competitiveness	Misallocation			
-13.0%	19.3%	-8.2%			
	C. Between 2010 and 202	0			
Technology	Competitiveness	Misallocation			
-6.3%	-9.8%	7.6%			



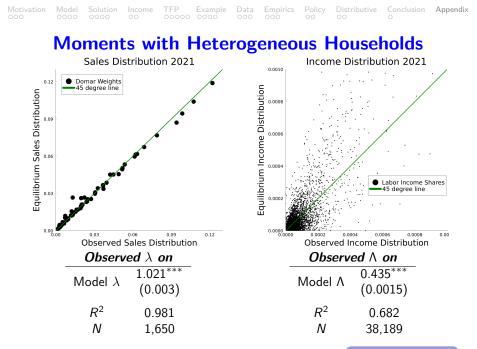
Covariance Decomposition

Table: Covariance Decomposition

	A. Between 1997 and 202	20
Technology	Competitiveness	-Misallocation
44.4%	34.6%	21.0%

	B. Between 2002 and 20	09
Technology	Competitiveness	-Misallocation
28.3%	61.2%	10.5%

	C. Between 2010 and 202	20
Technology	Competitiveness	-Misallocation
58.1%	4.9%	37.0%



Moments under different RH



Implementation

$$d \log TFP_{t} = \sum_{i} \widetilde{\lambda}_{i,t-1} d \log A_{i,t}$$

$$Technology_{t}$$

$$+ \sum_{i} \widetilde{\lambda}_{i,t-1} d \log \mu_{i,t}$$

$$Competitiveness_{t}$$

$$- \sum_{r} \widetilde{\Lambda}_{r,t-1} d \log \Lambda_{r,t}$$

$$Misallocation_{t}$$





Model without Intermediate Inputs

	Rep. H	ousehold	Occup	ation	Cou	nty	State &	Ocupation
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
$d \log TFP$	0.523		0.503		0.388		-0.265	
<i>u tog 1 F F</i>	(0.366)		(0.350)		(0.316)		(0.264)	
Technology		1.341^{***}		0.789^{***}		0.796^{***}		0.847^{***}
rechnology		(0.308)		(0.267)		(0.266)		(0.289)
Competitiveness		0.212		0.320		0.454		0.986
Competitiveness		(0.423)		(0.489)		(0.373)		(0.695)
Misallocation		0.573^{*}		0.450		0.335		-0.105
Misanocation		(0.329)		(0.437)		(0.315)		(0.360)
Intercept	0.012***	0.011^{***}	0.012^{***}	0.012^{***}	0.013^{***}	0.012^{***}	0.015^{***}	0.012^{***}
intercept	(3.2e-3)	(2.0e-3)	(3.2e-3)	(2.2e-3)	(3.2e-3)	(2.1e-3)	(3.0e-3)	(2.2e-3)
Observations					22			
Ν					66			
Η		1	75	0	3,1	36	3	8,190
R^2	9.2%	71.4%	9.35%	62.4%	7.00%	62.5%	4.8%	60.4%
$Adj. R^2$	9.2%	68.4%	9.35%	58.4%	7.00%	58.6%	4.8%	56.2%



Model with Intermediate Inputs

	Rep. H	ousehold	Occup	ation	Cou	nty	State &	Ocupation
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
$d \log TFP$	0.370***		0.311^{***}		0.316^{***}		0.311^{***}	
<i>a i 0g 1 1 1</i>	(0.072)		(0.069)		(0.065)		(0.069)	
Technology		0.478^{***}		0.414^{***}		0.416^{***}		0.413^{***}
rechnology		(0.097)		(0.081)		(0.083)		(0.082)
Competitiveness		0.398^{***}		0.341^{***}		0.350^{***}		0.342^{***}
Competitiveness		(0.062)		(0.054)		(0.053)		(0.054)
Misallocation		0.074		0.172		0.164		0.168
Misanocation		(0.138)		(0.125)		(0.135)		(0.125)
Intercept	0.010***	0.009	0.011^{***}	0.010^{***}	0.011^{***}	0.010^{***}	0.011^{***}	0.010^{***}
mercept	(2.1e-3)	(2.0e-3)	(2.2e-3)	(1.8e-3)	(2.1e-3)	(1.9e-3)	(2.3e-3)	(1.9e-3)
Observations					22			
Ν					66			
Н		1	75	0	3,1	36	3	8,190
R^2	56.9%	75.2%	49.9%	75.8%	54.0%	75.4%	49.9%	75.5%
$Adj. R^2$	56.9%	72.6%	49.9%	73.3%	54.0%	72.8%	49.9%	73.2%



Technological Sources

	A. Between 1998 and 2	020
1	Oil & gas extraction	-11.11%
2	Computer & electronics	-6.64%
3	Telecommunications	-2.85%
4	Computer systems design	-2.30%
5	Administrative services	-1.74%
6	Insurance carriers	-1.45%
7	Farms	-1.34%
8	Primary metals	-1.28%
	:	
63	Rental & leasing	1.41%
64	Credit intermediation	1.77%
65	Chemical Products	2.84%
66	Construction	2.87%

C. Between 2010 and 2020			
1	Oil & gas extraction	-5.41%	
2	Computer systems design	-1.29%	
3	Management of companies	-1.26%	
4	Housing	-1.14%	
5	Other real estate	-1.01%	
64	Air transportation	1.03%	
65	Chemical products	1.90%	
66	Credit intermediation	2.73%	

	B. Between 2002 and 2	009
1	Oil & gas extraction	-5.35%
2	Computer & electronics	-2.84%
3	Telecommunications	-2.27%
4	Utilities	-1.92%
5	Administrative services	-1.06%
66	Construction	1.76%



Competitiveness Sources

A. Between 1998 and 2020

1	Housing	-1.65%
2	Insurance carriers	-1.53%
3	Misc. professional services	-1.10%
4	Other services	-0.89%
	:	
63	Publishing industries	0.80%
64	Computer and electronics	1.34%
65	Chemical products	2.57%
66	Credit intermediation	4.10%

C. Between 2010 and 2020

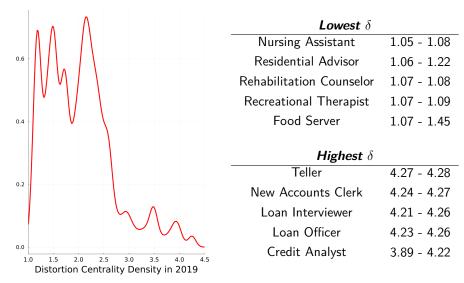
1	Oil & gas extraction	-6.34%
2	Housing	-3.09%
3	Insurance carriers	-0.98%
4	Misc. professional services	-0.87%
5	Administrative services	-0.82%
	•	
64	Primary metals	0.80%
65	Chemical products	0.84%
66	Credit intermediation	3.86%

B. Between 2002 and 2009

1	Securities & investment	-0.86%
	:	
58	Wholesale trade	0.92%
59	Publishing industries	0.93%
60	Internet, & inf. services	0.99%
61	Chemical products	1.35%
62	Telecommunications	1.43%
63	Computer and electronics	1.48%
64	Housing	1.57%
65	Utilities	1.87%
66	Oil & gas extraction	6.59%

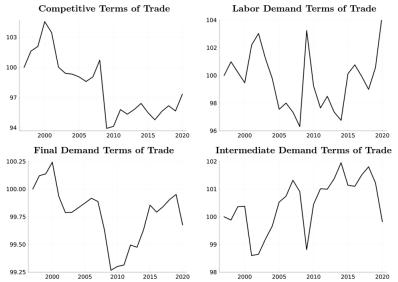


Distortion Centralities δ





Sources of Misallocation



Motivation	Model	Solution	Income	TFP	Example	Data	Empirics	Policy	Distributive	Conclusion	Appendix

Table 11: Counterfactual TFP Growth Differential in the Absence of Misallocation Components

Heterogeneity	Distributiv	e Competitive	Labor	Final	Intermediate
Heleroyeneity	TT	TT	DTT	DTT	DTT
Rep. Household	0%	-3.4%	6.3%	0.4%	-1.3%
Occupation	0%	-5.9%	15.1%	-2.0%	-4.2%
County	0.1%	-5.2%	14.2%	-0.9%	-4.4%
State & Occupation	0.1%	-5.9%	15.6%	-2.6%	-4.5%

A. Between 1998 and 2020

B. Between 2002 and 2009

Heterogeneity	Distributive TT	Competitive TT	$Labor \\ DTT$	Final DTT	Intermediate DTT
Rep. Household	0%	-9.3%	1.1%	-0.9%	-0.2%
Occupation	0%	-11.0%	3.4%	-1.9%	-0.8%
County	0.1%	-10.4%	3.4%	-0.7%	-1.0%
State & Occupation	0.1%	-11.1%	3.4%	-2.0%	-0.9%

C. Between 2010 and 2020

Heterogeneity	Distributive TT	c Competitive TT	$Labor \\ DTT$	$Final \\ DTT$	Intermediate DTT
Rep. Household	0%	3.9%	1.2%	1.7%	0.9%
Occupation	0%	2.9%	7.2%	0.2%	-1.8%
County	0.1%	3.0%	3.5%	2.1%	-1.5%
State & Occupation	0.1%	2.8%	7.4%	-0.1%	-1.7%

Motivation	Model	Solution	Income	TFP	Example	Data	Empirics	Policy	Distributive	Conclusion	Appendix

Table 12: Counterfactual TFP Growth Without Sectoral Competitive TT

A. Between 1998 and 2020

1	Credit intermediation	-2.16%
2	Chemical products	-1.06%
3	Computer & electronics	-0.98%
4	Publishing industries	-0.80%
5	Internet & inf. services	-0.69%
	:	
64	Insurance carriers	0.77%
65	Other services	0.81%
66	Misc. professional services	0.87%

B. Between 2002 and 2009

1	Oil & gas extraction	-1.46%
2	Computer & electronics	-1.11%
3	Internet & inf. services	-1.01%
4	Wholesale trade	-0.92%
5	Telecommunications	-0.86%
6	Utilities	-0.84%
7	Publishing industries	-0.82%

C. Between 2010 and 2020

1	Credit intermediation	-2.0%
2	Securities & investment	-0.52%
64	Administrative services	0.62%
64 65	Misc. professional services	0.82%
66	Oil & gas extraction	1.91%

Table 13: Counterfactual TFP Growth Without Sectoral Labor Demand TT

A. Between 1998 and 2020

1	Wholesale trade	-1.62%
2	Insurance carriers	-1.61%
3	Other retail	-1.07%
61	Utilities	0.69%
62	Computer systems design	0.82%
63	Publishing industries	1.34%
64	Oil & gas extraction	1.79%
65	Computer & electronics	2.28%
66	Credit intermediation	2.40%

B. Between 2002 and 2009

1	Securities & investment	-0.96%
	:	
64	Computer & electronicss	0.85%
65	Utilities	1.02%
66	Oil & gas extraction	2.20%

C. Between 2010 and 2020

1	Wholesale trade	-1.70%
2	Insurance carriers	-1.03%
3	Administrative services	-0.93%
4	Other retail	-0.83%
64	Publishing industries	0.89%
65	Computer & electronics	0.98%
66	Credit intermediation	2.44%

Motivation 000	Model 0000	Solution		FP Exa	ample 00	Data 000	Empirics 000	Policy 00	Distribu 00	tive	Conclusion O	Appendix
		Tab	le 14: Count	erfactual	TFP	\mathbf{Tal}	ble 15: (Counterf	actual 7	FFP		
	Growth Without Sectoral Final											
					Growth Without Sectoral Intermediate Demand TT							
	Demand TT						nterme	diate Der	mand T	Т.		
			A. Between 1998 and 2020				A. Between 1998 and 2020					
		1	Computer & el		-1.50%	1		ter & electro		.24%		
		2	Motor veh		-0.91%	2		intermediat		.90%		
		3	Machine	ry	-0.88%	3	Publis	hing industr	ries -0	.76%		
		4	Apparel & l	eather	-0.51%	4	Compute	er systems d	lesign -0	.45%		
			:			5	Ambu	bulatory health -0.42%		.42%		
		62	Securities & in	vestment	0.87%							
		63	Misc. profession	al services	0.94%	61	Teleco	ommunicatio	ons 0.	.52%		
		64	Hospita	ls	0.95%	62	Admini	strative serv	vices 0.	.54%		
		65	Internet & inf.		1.01%	63	1	Hospitals	0.	.56%		
		66	Wholesale	trade	1.18%	64		rance carrier	rs 0.	.74%		
						65	0	ther retail	0.	.90%		
			B. Between 2002 and 2009			66	Wh	olesale trade	e 1.	.21%		
		1			-							
			2 Motor vehicles		-0.82%		B. Between 2002 and 2009					
		-	:		,.	1	Comput	ter & electro	onics -0	.48%	-	
		66	Hospita	le	0.58%			:				
		00	nospita	15	0.5670	66	Securiti	es & investr	nent 0.	.49%		
	C. Between 2010 and 2020											
		1	1 Computer & electronis -0.52%			-	C. Between 2010 and 2020					
			· ·			1	Credit	intermediat	tion -0	.97%		
					04	2	Publis	hing industr	ries -0	.51%		
		63	Other ret		0.59%	3	Comput	ter & electro	onics -0	.49%		
		64	Internet & inf.		0.60%							
		65	Construct		0.89%			:				
		66	Wholesale	trade	1.08%	63		rance carrier		.52%		
						64		strative serv		.63%		
						65	0	ther retail	0.	.66%		

66

Wholesale trade

1.12%

Back

Motivation Model Solution Income TFP Example Data Empirics Policy Distributive Conclusion Appendix

Normalized nested CES environment - Firms

Firms

$$\frac{y_i}{\overline{y}_i} = A_i \left(\sum_{h \in \mathscr{H}} \widetilde{\Omega}_{ih}^{\ell} \left(\frac{\ell_{ih}}{\overline{\ell}_{ih}} \right)^{\frac{\theta_i - 1}{\theta_i}} + \sum_{j \in \mathscr{N}} \widetilde{\Omega}_{ij}^{x} \left(\frac{x_{ij}}{\overline{x}_{ij}} \right)^{\frac{\theta_i - 1}{\theta_i}} \right)^{\frac{\theta_i}{\theta_i - 1}}$$

Back

Normalized nested CES environment - Households Households

Policy Distributive Conclusion Appendix

$$U_h\left(c_h,\widetilde{L}_h\right) = \frac{\left(c_h\left(1 - E_h^{-\gamma_h}\widetilde{L}_h\right)^{\varphi_h}\right)^{1-\sigma} - 1}{1-\sigma} \quad s.t. \quad \frac{C_h}{\overline{C}_h} = \left(\sum_{i \in \mathcal{N}} \beta_{hi}\left(\frac{C_{hi}}{\overline{C}_{hi}}\right)^{\frac{\rho_h - 1}{\rho_h}}\right)^{\frac{\rho_h}{\rho_h - 1}}$$

with $C_h = n_h c_h$ and $L_h = n_h \widetilde{L}_h$

Normalized nested CES environment - Households Households

$$U_{h}\left(c_{h},\widetilde{L}_{h}\right)=\frac{\left(c_{h}\left(1-E_{h}^{-\gamma_{h}}\widetilde{L}_{h}\right)^{\varphi_{h}}\right)^{1-\sigma}-1}{1-\sigma}\quad s.t.\quad \frac{C_{h}}{\overline{C}_{h}}=\left(\sum_{i\in\mathcal{N}}\beta_{hi}\left(\frac{C_{hi}}{\overline{C}_{hi}}\right)^{\frac{\rho_{h}-1}{\rho_{h}}}\right)^{\frac{\rho_{h}-1}{\rho_{h}-1}}$$

with $C_h = n_h c_h$ and $L_h = n_h \widetilde{L}_h$

The change in labor supply from type h workers is, to a first-order

$$d \log L_h = \zeta_h^n d \log n_h + \zeta_h^w d \log w_h - \zeta_h^e d \log E_h$$

Where the corresponding elasticities are given by

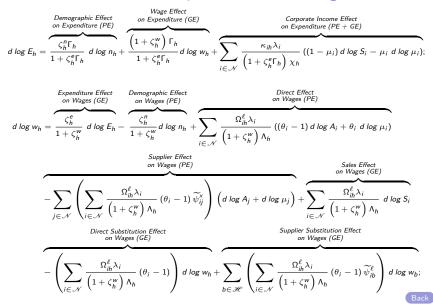
$$\zeta_h^n = \frac{E_h^{\gamma_h}}{1 - \varphi_h \gamma_h} \frac{n_h}{L_h}, \qquad \zeta_h^w = \frac{1}{1 - \varphi_h \gamma_h} \frac{\varphi_h}{\Gamma_h}, \qquad \zeta_h^e = \zeta_h^w - \gamma_h \zeta_h^n.$$



Distributive Conclusion Appendix

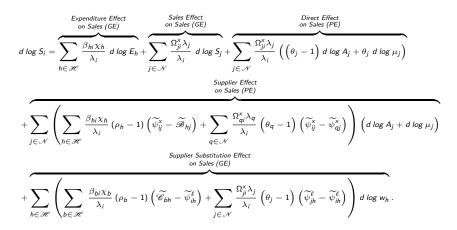


Solution - Expenditure & Wages





Solution - Sales



Back

 $d \log TFP = Competitiveness - Misallocation$

=1%

Best Sectors	d log TFP					
1. Housing	0.766%			d log	TFP on	
11 110000118	0.10070		-0.974***			-0.919***
2. Credit Intermediation	0.414%	μ_i	(0.12)			(0.18)
3. Oil & Gas extraction	0.384%	,		1.351		-0.132
4. Furniture	0.370%	λ_i		(0.95)		(0.73)
5. Mining, except Oil & Gas	0.364%	_			-0.427***	-0.046
		Fi			(0.08)	(0.11)
Worst Sectors	d log TFP				(0.00)	(••==)
		R^2	0.48	0.03	0.29	0.48
1. Nursing & Residential Care	-0.329%	N			66	
2. Social Assistance	-0.303%				00	
3. General Merchandise Store	-0.274%					

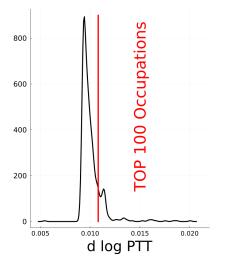
 4. Hospitals
 -0.219%

 5. Ambulatory health care
 -0.201%

We want competition shocks in sectors with low $\mu_i!$

Motivation Model Solution Income TFP Example Data Empirics Policy Distributive Conclusion Appendix

Effects from more competition in Housing



Top 100 occupations	
Construction workers	48
Painters, Carpet Installer, Tile Setter,	
Stonemason, Plasterer, Drywall Installer,	
Septic Servicer, Construction Supervisor	
Financial specialist	7
Property appraiser, Loan Officer	
Credit Analyst, Financial Examiner	
Extraction Workers Rock Splitter, Roof Bolter	7
Woodworkers Cabinetmaker, Furnite Finisher	6
Installation & Maintenance Heating & AC, Mobile Home Installer	5

Ton 100 occupations