

Discussion: The Changing Economics of Knowledge Production

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Overview

- Agenda: How does AI affect the financial industry?
- This paper: AI reduces the labor share of (knowledge) production

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- This paper: AI reduces the labor share of (knowledge) production
- Contributions
 - 1 Unequal distribution of AI's benefits
Fuster, Goldsmith-Pinkham, Ramadorai, Walther 2022
 - 2 By-product: Measurement of opaque intangible investment

Model

- Production with N technologies

$$\max_{K, L_j} \sum_{j=1}^N Y_j - \rho K - \sum_{j=1}^N w_j L_j, \quad Y_j = A_j K^{\alpha_j} L_j^{1-\alpha_j}$$

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- Optimal expenditure shares

$$(1 - \alpha_j) Y_j = w_j L_j$$

$$\sum_{j=1}^N \alpha_j Y_j = \rho K$$

- Estimation equation

$$K = \sum_{j=1}^N \underbrace{\frac{1}{\rho} \frac{\alpha_j}{1 - \alpha_j}}_{=\beta_j} \underbrace{w_j L_j}_{=X_j} \Rightarrow \frac{\beta_j}{\beta_k} = \frac{\alpha_j}{1 - \alpha_j} \frac{1 - \alpha_k}{\alpha_k}$$

Measurement in Finance

- Old-school analysts L_1 and AI/ML analysts L_2
- Data capital K is a moving average of past data cleaners L_0
- Infer L_j from job listings across firms, w_j from online survey

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- Data capital K is a moving average of past data cleaners L_0
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- Full structural estimation

$$\alpha_1 - \alpha_2 \simeq 0.05$$

Interpretation

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 - Structuring / sales / executive jobs that convert knowledge into profit could have *greater* shares in the age of AI

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- Data or labor? K , L_1 and L_2 are all estimated MAs of job postings
- GE effects? $P_Y A_j$ is identified, diminishing returns could be severe

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- Workers' opportunity set
 - Work old-school for a higher wage
 - Work in AI for an even higher wage
- Pareto improvement due to Cobb-Douglas
- Other functional forms allow task *replacement*

Specifications: Rivalrous data capital

- Production with heterogeneous labor

$$\max_{K, L_j} \sum_{j=1}^N Y_j - \rho \sum_{j=1}^N K_j - \sum_{j=1}^N w_j L_j, \quad Y_j = A_j K_j^{\alpha_j} L_j^{1-\alpha_j}$$

- Optimal expenditure shares

$$\begin{aligned}(1 - \alpha_j) Y_j &= w_j L_j \\ \alpha_j Y_j &= \rho K_j\end{aligned}$$

- Estimation equation

$$\sum_{j=1}^N K_j = \sum_{j=1}^N \underbrace{\frac{1}{\rho} \frac{\alpha_j}{1 - \alpha_j}}_{=\beta_j} \underbrace{w_j L_j}_{=X_j}$$

Specifications: Single technology

- Production with heterogeneous labor

$$\max_{K, L_j} \sum_{j=1}^N Y - \rho K - \sum_{j=1}^N w_j L_j, \quad Y = A_j K^{1 - \sum_{j=1}^N \gamma_j} \prod_j L_j^{\gamma_j}$$

- Optimal expenditure shares

$$\beta_j Y = w_j L_j$$

$$(1 - \sum_{j=1}^N \gamma_j) Y = \rho K$$

- Estimation equation

$$K = \sum_{j=1}^N \frac{1}{\rho} \underbrace{\frac{(\frac{1}{N} - \gamma_j)}{\gamma_j}}_{=\beta_j} \underbrace{w_j L_j}_{=X_j}$$

Conclusions

- Huge progress in serious modelling of data-driven technology
- Tight implications for knowledge production
- Potential estimation of a more flexible model