Artificial Intelligence, Firm Growth, and Industry Concentration

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Motivation

- Explosion in Artificial Intelligence (AI) investment in recent years
 - From \$7.6 billion in 2010 to \$48 billion in 2018 in the US + \$24 billion targeted by EU,
 \$150 billion by China
- How does AI affect firms and industries?
 - Huge potential to transform production, yet sluggish productivity growth recently

Motivation

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- How does AI affect firms and industries?
 - Huge potential to transform production, yet sluggish productivity growth recently
- Potential channels:
 - Productivity growth (Brynjolfsson, Hitt, Kim 2011; Tanaka, Bloom, David, Koga 2019)
 - Monopoly power (Varian 2018; Mihet and Philippon 2019)
 - Economies of scale (Autor, Dorn, Katz, Patterson, Van Reenen, 2020; Aghion, Bergeaud, Boppart, Klenow, Li 2019; Farboodi, Mihet, Philippon, Veldkamp 2019)

This paper

- Goals:
 - Which firms invest in AI?
 - How do Al investments affect firm growth?
 - Which firms benefit most, and how does Al affect industry concentration?
- Our approach:
 - Data: detailed, firm-level information on demand for Al talent (job postings) + hiring of Al talent (resumes) of US public firms
 - Measure: data-driven measure of Al-relatedness of each skill and job, without pre-specifying Al-related words
 - Identification: instrument AI investments with:
 - a) industry-level AI investments in Europe
 - b) shift-share instrument relying in differences in local exposure to Al

Preview of findings

- 1. Rapid growth in AI technologies across a wide range of industries
- 2. Large firms are more likely to invest in Al
- 3. Firm-level: Al increases sales, employment, market share
 - No effect on productivity or markups
 - Growth is concentrated in ex-ante largest and most productive firms
- 4. Industry-level: Al increases aggregate sales, employment and industry concentration

Overall, Al leads to the expansion of ex-ante most efficient firms and contributes to the rise of "superstar" firms

Data

- Job postings: Burning Glass Technologies
 - 180 million job postings
 - Comprehensive coverage of online job openings in 2007 and 2010–2018
 - Detailed taxonomy of required skills
- Employment profiles (resumes): Cognism, aggregator of public profile information
 - 145 million full profiles; global coverage
 - Job histories, skills, education, publications, patents, awards, references
 - Captures actual hiring, not just demand

Measuring Al

- Identify relevant skills: Burning Glass
 - Four core Al skills: Artificial Intelligence, Machine Learning, Natural Language Processing, Computer Vision
 - Al-relatedness score of skill s =
 % of jobs requiring skill s that also require at least one core Al skill

Highly related (score>0.7) (N=68)	Less related (0.05 <score<0.7) (N=533)</score<0.7) 	Not related (score<0.05) (N=13,577)		
Tensorflow (0.90)	Information retrieval (0.37)	Communication skills (0.003)		
Unsupervised learning (0.89)	Logistic regression (0.26)	Microsoft Office (0.001)		
Deep learning (0.86)	Speech recognition (0.22)	Lawn mowing (0.000)		
Random forests (0.84)	Python (0.12)			

Measuring Al

Burning Glass job postings

A job posting is Al-related if average score of all required skills > 0.1

Example:

Required skills: Machine learning (1), Text mining (0.63), MapReduce (0.29), Apache Hadoop (0.21), Data mining (0.16), Software engineering (0.04), Research (0.01), Communication skills (0.00)

Average score: 0.29

Cognism online resumes

A job in a resume is AI-related if it contains a highly related skill (score>0.7) in profile

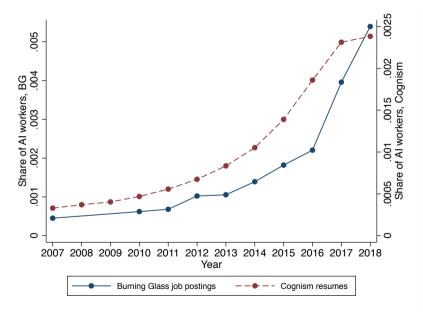
Examples:

Job title: "Senior Machine Learning Developer"
Job description: "develop Chatbots using Python with scikit learn, tensorflow and deep learning models..."
Publications: "A New Cluster-Aware Regularization of Neural Networks"

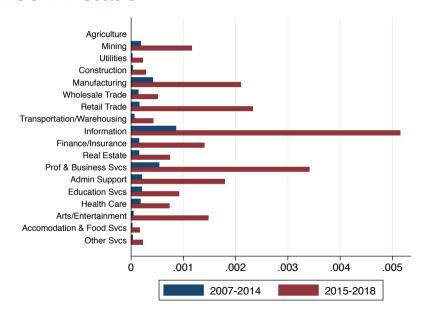
Patents: "Systems and methods for prime design using machine learning"

- Match employers to Compustat and calculate % of AI jobs at firm level
- Al measures from two datasets highly correlated (65%) and yield consistent results

Fast Growth in Share of Al Workers in the Last Decade



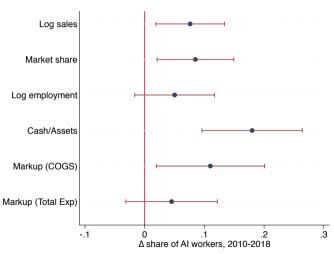
More Al Workers in All Sectors



Which Firms Adopt AI?

- Predict change in AI share from 2010–2018 based on 2010 firm characteristics:

$$\Delta Share Al Workers_i^{2010-2018} = \beta X_i^{2010} + Industry FE + \varepsilon_i$$



Effect of AI on Firm Growth

Long-differences regression:

$$\Delta Y_i^{2010-2018} = \beta \Delta ShareAlWorkers_i^{2010-2018} + \gamma X_i^{2010} + IndustryFE + \varepsilon_i$$

- Two IV strategies to identify causal effect of AI
- Instrument #1: industry-level Al growth in Europe (Autor, Dorn, Hanson 2013; Acemoglu and Restrepo 2020)
 - Build 5-digit-NAICS-level AI measure from granular European firm-level data
 - Identifying assumption: Al growth in US and Europe driven by similar technological factors that vary across industries
 - First stage F-statistics between 22 and 38
- Instrument #2: Bartik IV based on firms' differences in local exposure to AI

Effect of Al on Firm Growth, 2010–2018

- OLS results:

	Δ Log Sales		Δ Log Employment		Δ Market Share	
Δ Share Al Workers	0.106**	0.146***	0.114*	0.133**	0.012	0.013
	(0.052)	(0.045)	(0.064)	(0.053)	(0.012)	(0.009)
Ind FE	Y	Y	Y	Y	Y	Y
Controls	N	Y	N	Y	N	Y

- IV results (Foreign-industry IV):

	Δ Log Sales		Δ Log Employment		∆ Market Share	
Δ Share Al Workers	0.173 0.274**		0.217	0.327*	0.024**	0.053**
	(0.185)	(0.123)	(0.177)	(0.173)	(0.011)	(0.026)
Ind FE	Υ	Υ	Υ	Υ	Υ	Υ
Controls	N	Υ	N	Υ	N	Υ

Al Investments Associated with Expansion at Industry Level

- OLS results:

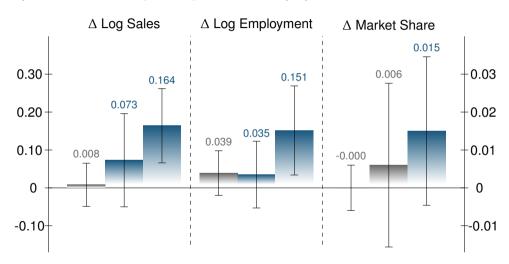
	∆ Log	Sales	Δ Log Employment		
Δ Share Al Workers	0.125*** (0.039)	0.151*** (0.029)	0.130** (0.063)	0.158*** (0.056)	
Sector FE	Υ	Υ	Υ	Υ	
Controls	N	Υ	N	Υ	

- IV results (Foreign-industry IV):

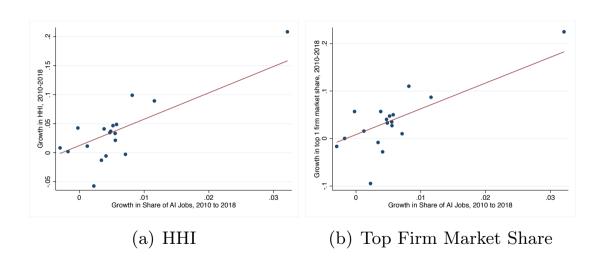
	∆ Log	Sales	Δ Log Employment		
Δ Share Al Workers	0.210*** 0.248*** (0.047) (0.038)		0.177*	0.218***	
			(0.106)	(0.074)	
Sector FE	Υ	Υ	Υ	Υ	
Controls	N	Υ	N	Υ	

Growth is Concentrated in Big Firms

Subsamples: firms ranked (terciles) based on employment as of 2010



Al Investments Associated with Increased Industry Concentration



1. Productivity growth

	∆ Log Sal	es Per Worker	Δ Revenue TFP		
Δ Share Al Workers	s -0.028 -0.006 (0.038) (0.035)		-0.015	0.004	
			(0.033)	(0.035)	
Ind FE	Υ	Υ	Υ	Υ	
Controls	Ν	Υ	Ν	Υ	

1. Productivity growth

	Δ Log Sal	es Per Worker	Δ Revenue TFP		
Δ Share Al Workers	-0.028 (0.038)	-0.006 (0.035)	-0.015 (0.033)	0.004 (0.035)	
Ind FE	Υ	Υ	Υ	Υ	
Controls	N		Ν	Υ	

2. Market power

	Δ Log Markup (COGS)		Δ Log Markup (Total Exp)		Lerne	\ r Index
Δ Share Al Workers	0.001	0.031	-0.001	0.010	-0.005	0.004
	(0.016)	(0.028)	(0.006)	(0.011)	(0.003)	(0.006)
Ind FE	Y	Y	Y	Y	Y	Y
Controls	N	Y	N	Y	N	Y

3. Economies of scale: Al allows ex-ante large and efficient firms to expand

- Al and big data can induce economies of scale and reinforce winner-take-all effects (Farboodi, Mihet, Philippon, Veldkamp 2019)
- Aghion, Bergeaud, Boppart, Klenow, Li (2019): falling costs of spanning multiple markets due to IT ⇒ most efficient firms expand into new markets

	Δ Log # of Counties		,	g # of Managers
Δ Share Al Workers	0.067*** 0.078***		0.165***	0.134***
	(0.020) (0.024)		(0.046)	(0.067)
Ind FE	Y	Y	Y	Y
Controls	N	Y	N	Y

Conclusion

- Al leads to higher firm growth
 - Al allows firms to grow even larger, capturing more market share
 - No significant impact on productivity and markups
- Industry-level shift towards ex-ante larger firms
 - The impact of AI is most pronounced among the largest, most productive firms
 - All investments are associated with higher industry concentration
- Our results highlight the role of technology in shaping industry concentration and superstar firms