

Discussion of IMF SDN on  
“Digitalization during the Covid-19 Crisis”  
by Jaumotte et al.

Anton Korinek

[anton@korinek.com](mailto:anton@korinek.com)

Rubenstein Fellow, Brookings  
Professor, University of Virginia  
Economics of AI Lead, GovAI

<http://www.korinek.com>    [@akorinek](https://twitter.com/akorinek)

# Main Take-Away: COVID-19 and Digitalization

COVID-19 acted as a catalyst for the adoption of digital technologies  
→ what can we learn from the “natural experiment”

- The SDN carefully documents effects on digitalization
  - reduced gap between more- and less-digitalized advanced countries
- considers resulting productivity effects
- identifies structural effects on the labor market:
  - no permanent shift in labor demand
  - some indications of permanent supply effects:  
work-from-home ↑, male labor market participation ↓

# Framing: Digitalization as a “Silver Lining”

Increased digitalization = “silver lining” of COVID-19

Authors’ Hypothesis: this is desirable because of positive efficiency & distributive effects

Question: how do we square this with the invisible hand?

In other words, why did businesses need a pandemic to do what’s good for them?

- Argument relies on uninternalized externalities or internalities
- Identifying these is helpful for industrial policy more generally

# Focus on Productivity

$$\textit{Productivity} = \frac{\textit{Output}}{\textit{Factor Inputs}}$$

- During COVID pandemic, lots of effects on both numerator and denominator, e.g., labor hoarding, stark demand fluctuations, etc.
- Productivity measures most useful when markets are in equilibrium, esp. as an indicator for long-run productive potential
- We are yet to see the long-run productivity effects of COVID

# The Road Ahead

Concern of the 2000s/2010s – the beginning of the “Digital Age:”

- workers who are non-digitized, lesser-educated may be left behind by technological advances

→ the opposite happened during the recovery from COVID

New Concern of 2023 – the “Age of Generative AI:”

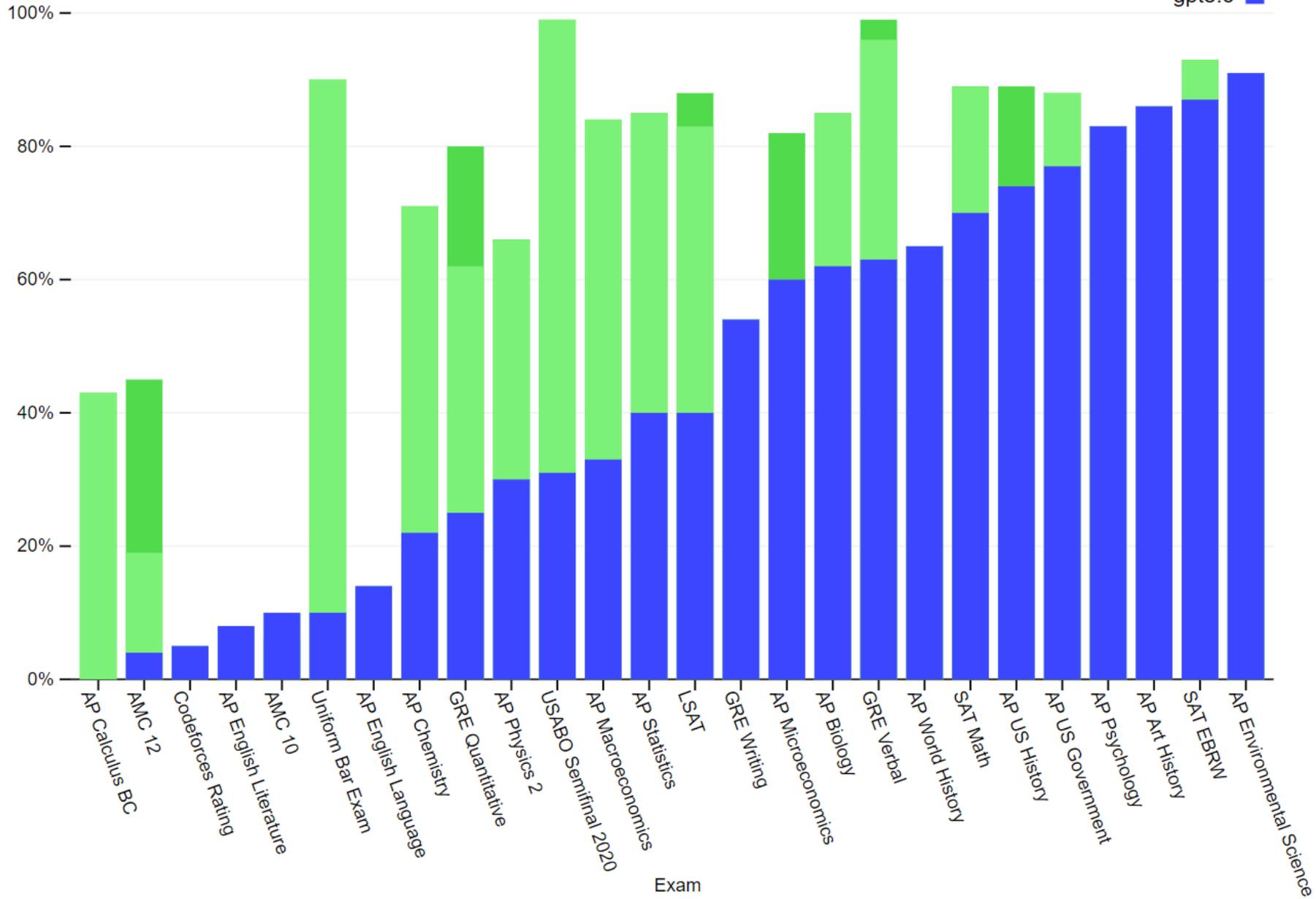
- cognitive workers suddenly seem at greatest risk of automation

→ “ChatGPT is taking my job”

# Exam results (ordered by GPT-3.5 performance)

Estimated percentile lower bound (among test takers)

gpt-4  
gpt-4 (no vision)  
gpt3.5



# Digitalization as a Steppingstone for Automation

Digitalization as a Steppingstone:

traditional job → work-at-home job → automated job

once a job has become “virtual,” cognitive automation may be easier to accomplish going forward

# Generative AI Capabilities for Cognitive Workers

1. Ideation
2. Writing
3. Background research
4. Coding
5. Data analysis
6. Math

→ see my [NBER Paper #30957](#) on “Language Models and Cognitive Automation for Economic Research”

<b>Category</b>	<b>Task</b>	<b>Usefulness</b>
Ideation	Brainstorming	3
	Evaluating ideas	2
	Providing counterarguments	3
Writing	Synthesizing text	3
	Editing text	3
	Evaluating text	3
	Generating catchy titles & headlines	3
	Generating tweets to promote a paper	3
Background Research	Summarizing Text	3
	Literature Research	1
	Formatting References	3
	Translating Text	3
	Explaining Concepts	2

<b>Category</b>	<b>Task</b>	<b>Usefulness</b>
Coding	Writing code	2
	Explaining code	2
	Translating code	3
	Debugging code	2
Data Analysis	Extracting data from text	3
	Reformatting data	3
	Classifying and scoring text	2
	Extracting sentiment	2
	Simulating human subjects	2
Math	Setting up models	2
	Deriving equations	1
	Explaining models	1

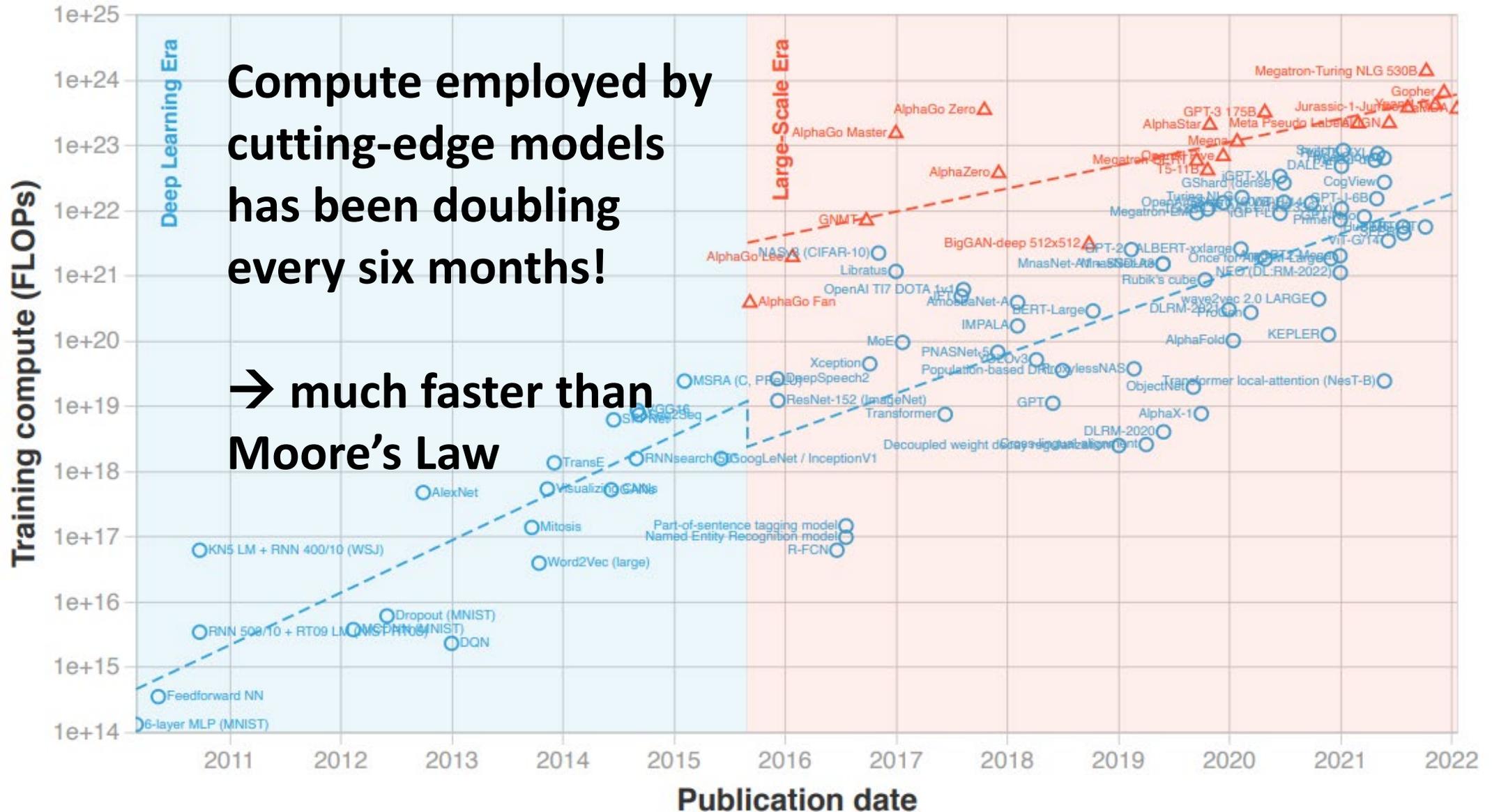
# Useful Capabilities for Cognitive Workers

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# Training compute (FLOPs) of milestone Machine Learning systems over time

n = 102



Compute employed by cutting-edge models has been doubling every six months!

→ much faster than Moore's Law