Data-intensive Innovation and the State: Evidence from AI Firms in China

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 \implies Government data can exceed privately-collected data in magnitude/scope; or lack good substitutes altogether

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- A common way in which Al firms gain access to valuable government data is by providing services to the state
- Think about facial recognition AI firms in China...
 - ▶ Train algorithms with, e.g., video streams of faces from many angles
 - The state's public security units collect this form of data through their surveillance apparatus, and contract AI firms for services
 - Al firms gaining access to surveillance data can use it to train algorithms and develop software

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- 2. Firms may learn to manage and utilize large datasets too

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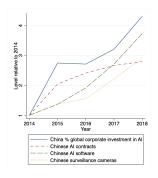
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Evidence of this in China's facial recognition AI sector

Two implications

- 1. Access to gov't data contributed to Chinese firms' emergence as leading innovators in facial recognition Al
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2. Novel role for the state in data-intensive economies

- So far, emphasis on the regulation of privately-collected data due to antitrust or privacy concerns (Tirole, 2020; Aridor et al., 2020)
- Al procurement and policies of gov't data collection and provision could, whether intentionally or not, stimulate and shape the direction of innovation in a range of sectors

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Would like to compare software output changes after receipt of gov't procurement contracts giving access to more v. less data

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- 1. Dataset linking AI firms to govt. contracts did not exist
- 2. Dataset on AI firms' software did not exist (our measure of *product innovation*). Also, critical for us to classify by use (commercial or not)
- 3. No available direct measures of firm-level use of gov't data

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Identification challenges

- 1. Non-random assignment of gov't contracts
- 2. Contracts work through other mechanisms unrelated to data

Data 1: linking AI firms to govt. contracts

- 1. Identify all facial recognition AI firms
 - 7,837 firms
 - Two sources: Tianyancha (People's Bank of China) and PitchBook (Morningstar)
 - Include: (*i*) firms specialized in facial recognition AI (e.g., Yitu); (*ii*) hardware firms that devote substantial resources to develop AI software (e.g., Hik-Vision); (*iii*) facial recognition AI units of large tech conglomerates (e.g., Baidu AI)

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- 3. Link government buyers to Al suppliers



Data 2: AI firms' software production

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Categorize by intended customers:

- 1. Commercial: e.g., visual recognition system for smart retail;
- 2. **Government:** e.g., *smart city real time monitoring system on main traffic routes*;
- 3. General: e.g., a synchronization method for multi-view cameras based on FPGA chips.

Categorization: analyze text using machine learning

Recurrent Neural Network (RNN) model using tensorflow

- Corpus: 13,000 manually labeled software programs
- Word-embedding: converted sentences to vectors based on word frequencies and used the words from full datasets as dictionary
- Long Short-Term Memory (LSTM) algorithm: 2 layers of 32 nodes
- 90% of corpus for training, 10% for validating
- 10,000 training cycles are run for gradient descent on loss function

Results robust to perturbing parameters of learning model

Within AI public security contracts: variation in the data collection capacity of the public security agency's local surveillance network

- 1. Identify non-AI contracts: police department purchases of street cameras
- 2. Measure quantity of advanced cameras in a prefecture at a given time
- 3. Categorize public security contracts as coming from "high" or "low" camera capacity prefectures

Baseline empirical strategy

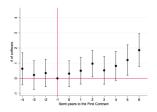
Triple diffs: compare cumulative software releases before and after firms received 1st data-rich contracts, relative to the data-scarce ones

$$y_{it} = \sum_{T} \beta_{1T} T_{it} Data_i + \sum_{T} \beta_{2T} T_{it} + \alpha_t + \gamma_i + \sum_{T} \beta_{3T} T_{it} X_i + \epsilon_{it}$$

- *T_{it}*: 1 if, at time *t*, *T* semi-years have passed before/since firm *i* received 1st contract
- *Data*_i: 1 if firm *i* receives "data rich" contract (i.e., from "high" camera capacity prefecture at time of contract receipt)
- X_i controls for pre-contract firm characteristics: age, size (cap), and software production

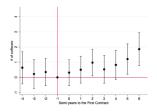
Public security contract "richer in data" & firm innovation

Commercial use cumulative software releases



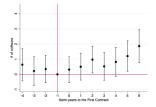
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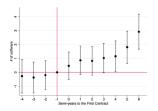


Magnitude: 2 new software products over 3 years (20% of pre-contract software)

Public security contract "richer in data" & firm innovation Commercial use cumulative software releases



Government use cumulative software releases



Commercial innovation overcomes crowd-out of inputs by gov't

Evaluating alternative hypotheses

- 1. Selection at a given time differs by contract
 - No differential pre-contract levels/trends of software
 - Control for time-varying effects of proxies for firms' underlying productivity: index constructed from establishment year, pre-contract capitalization, pre-contract rounds of external financing, pre-contract software production

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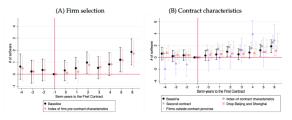
- Value of contract; tasks of contract; market access: we control for time-varying effects of an index of non-data contract characteristics (dollar value; prefecture income; tasks coded using NLP)
- Signaling value: examine second contracts within parent firms
- Political value: drop Beijing/Shanghai contracts; drop firms receiving contracts in home province

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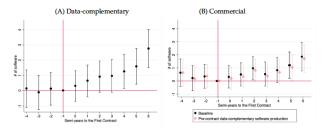
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Additional evidence for our mechanism(s)

Data-complementary software (e.g., storage/transmission) differentially increases after data-rich contract (learning); but, accounting for pre-contract data-complementary software does not greatly affect our findings (sharable data and algorithms)



Contributions to literature

- 1. To the literature on the economics of Al and data (e.g., Aghion et al., 2017; Agrawal et al., 2018; Farboodi et al., 2019; Jones and Tonetti, 2019)
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- 3. To the literature on the rise of China emphasizing the role of the state (e.g., Lau et al., 2000; Brandt and Rawski, 2008; Song et al., 2011)
 - Highlight the role of the surveillance apparatus in commercial innovation
 - Next project: Al-tocracy. Alignment between innovation and autocracy? Contrasts with e.g., North (1991); Acemoglu and Robinson (2006, 2012)

China's export of AI

Dominate global trade (> 50%), different from other frontier tech



export_x

China's export of AI

High number of autocratic destinations



export_x

US's export of AI

Much fewer links, higher share of democratic destinations



export_x