



THE POLITICAL CROSSROADS OF DIGITAL CAPITALISM

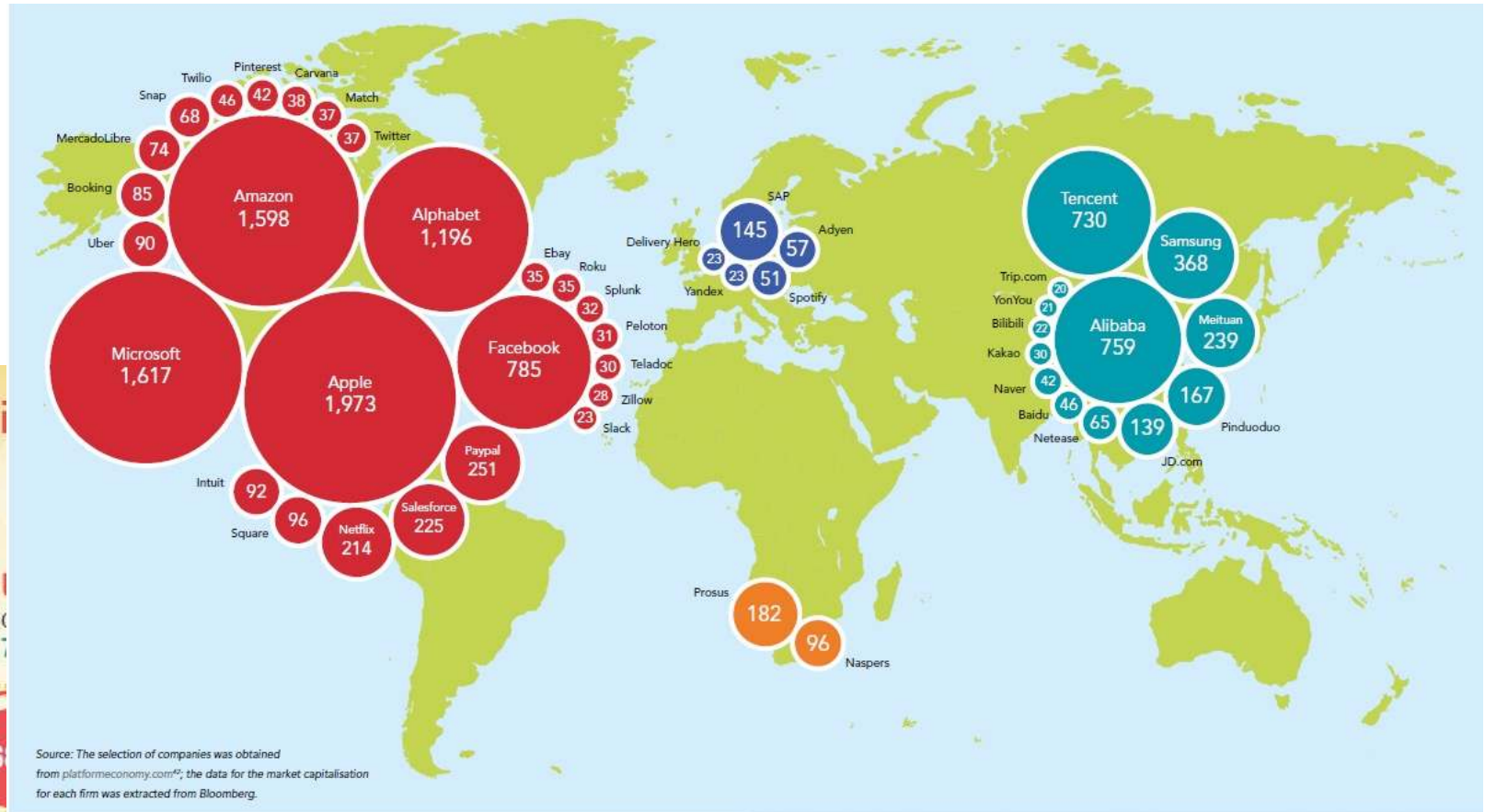
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Digital capitalis concentration



Source: SOMO (2021)

Geography of the digital economy in two countries

● United States and China ● Rest of the world

75% of all patents related to blockchain technologies

50% of global spending on IoT

>75% of the cloud computing market



Still huge digital divides

Half of the world remains offline

In LDCs only 1 in 5 people are online

Gender gap is the widest in the poorest economies

Source: UNCTAD (2019)

Industrial policy for the Digital Economy

- In an EC whitepaper on AI called “A European approach to excellence and trust”: EC proposes STI-policy initiatives to strengthen Europe’s position.
- Goal: “to achieve an ‘ecosystem of excellence’ along the entire value chain, starting in research and innovation, and to create the right incentives to accelerate the adoption of solutions based on AI, including by SMEs”
- Since 2020, EU-commission has raised issues in relation to the wave of GAFAM-acquisitions

“What we will try to do is to find ways to make sure that we see these acquisitions because sometimes these businesses are quite small, and maybe the sums, the turnover will not meet our thresholds”



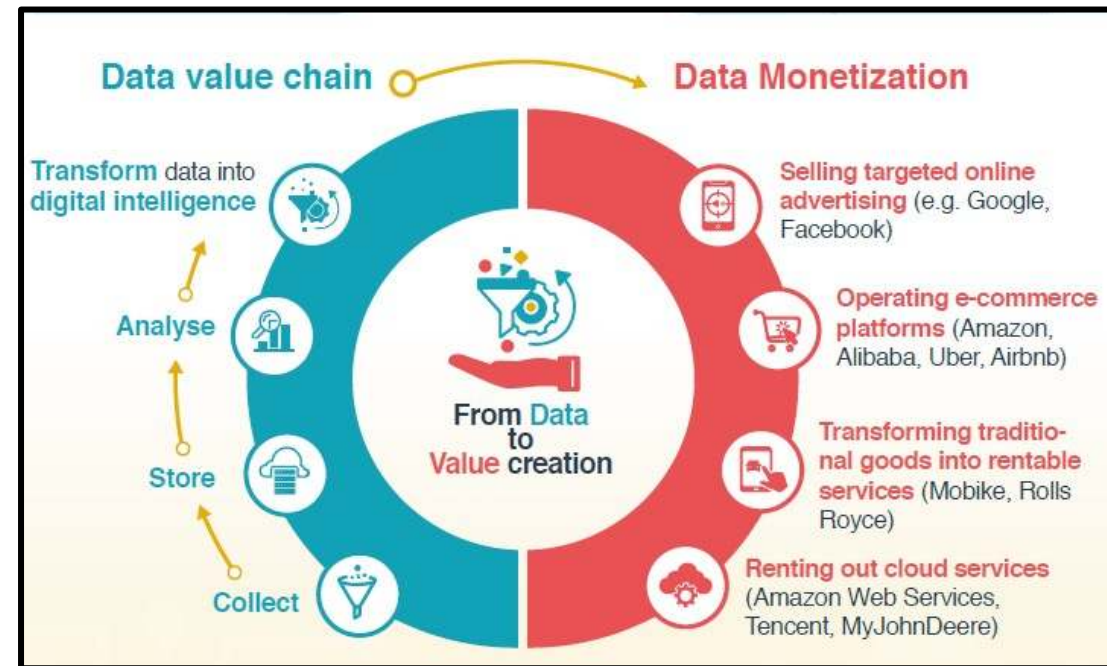
Digital Markets Act

- “A single set of rules” for the EU to have more bargaining power
- Focus on core platform services with a market capitalisation of at least € 75b or an annual turnover of € 7.5b .
- Regulation: platforms’ potential role as gatekeepers that are not necessarily dominant in competition-law terms.
- Antitrust at the EU level: carry out market investigations and sanction non-compliant behaviour.



Will this policy framework work?

- Past fines were never cashed. Google Shopping (2010), Google's Android (2015) and Google AdSense (2016).
- EC ruled against Apple and Ireland (illegal state aid through selective tax break) but the EU general court annulled the decision
- Tech giants can afford to operate without privileging their products over others in their marketplaces because of the popularity of their products.
- Even if each firm (and person) gets access to its data, this will not alter the balance of power.



Source: UNCTAD (2019)

trust:

ogle

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Why are these policies ill-equipped?

- The limits of Antitrust. Is more competition the solution?

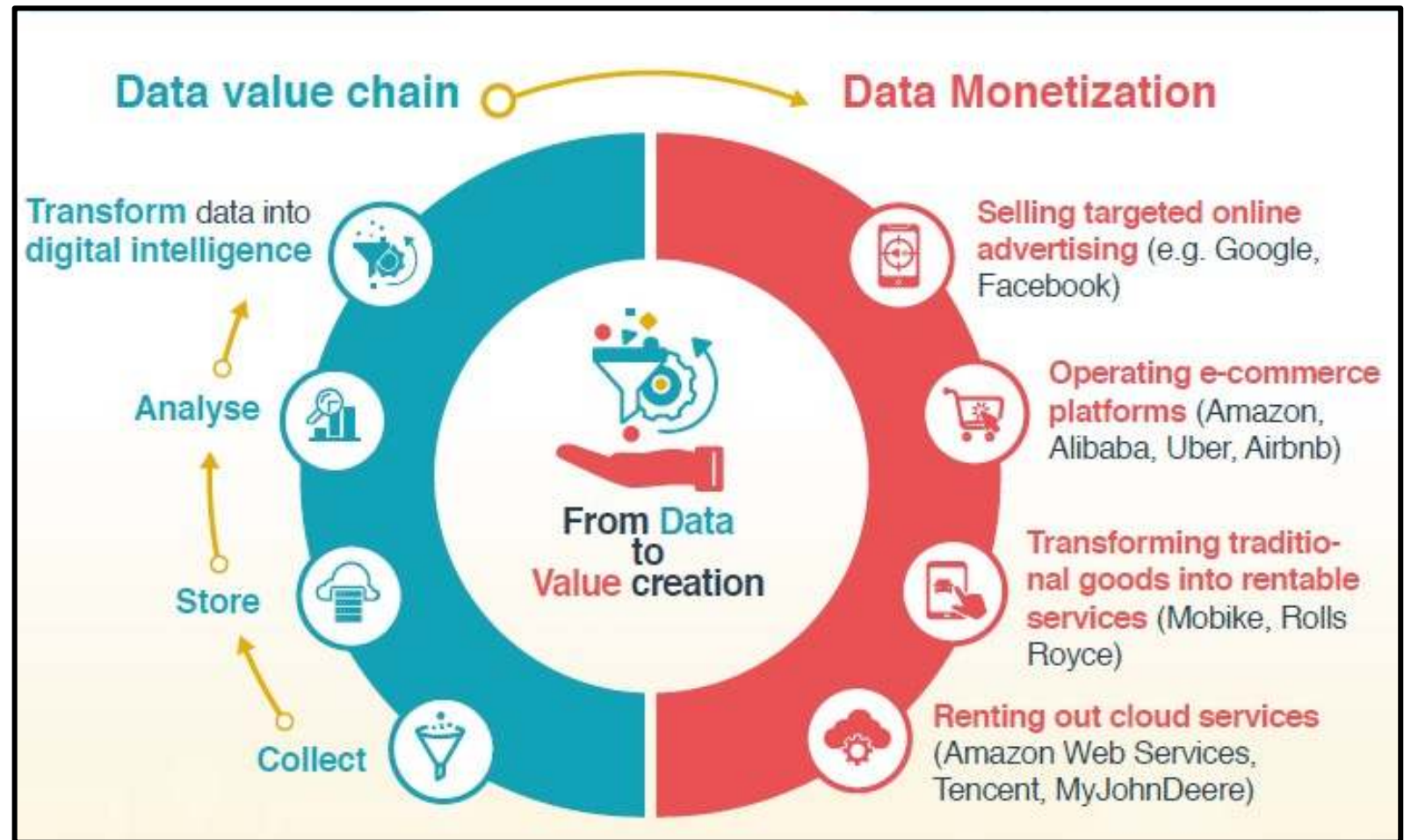


intellectual monopolies


How probable is the IMs scenario?

- Cumulative causation + Absorptive capacity → firms' technological differentiation
- Higher costs of doing R&D
- Recent technological change: ICT revolution + more S&T based production enabling knowledge modularity
- Institutional / Political Transformations:
 - **IPRs regime**: within the US (From Goodwill to Workplace knowledge law incl/ Trade Secret & then Bayh-Dole Act + others) and at the global level (TRIPS)
 - **Strategic tax planning/regulatory arbitrage**: particularly beneficial for MNEs intensive in intangible assets
 - **Anti-trust control was weakened**: Chicago Boys & the “consumer welfare” doctrine. Focus: consumer prices
 - Hidden **industrial policy**, among others connected to the military in the US

Data-Driven Intellectual Monopolies: Big Data + ML



Source: UNCTAD (2019)

A complex network graph with numerous nodes and edges, primarily in shades of red and pink, with some nodes in blue, green, and yellow. The nodes vary in size, and the edges are thin lines connecting them. The background is a light gray gradient.

KNOWLEDGE & DATA APPROPRIATION



- Data stored in public cloud: 5% of worldwide data storage in 2010 → 20% by 2018.

- Amazon, Microsoft, Google and Alibaba stored in their public clouds around 4.9% of the global data stored worldwide in 2015 and already 22.8% in 2020

- Between 2010 and 2018, Amazon's datacentres grew 1,337% in surface area

- Microsoft has over 100 datacentres in 54 regions of the world

- 50% undersea internet cable GAM+ Meta

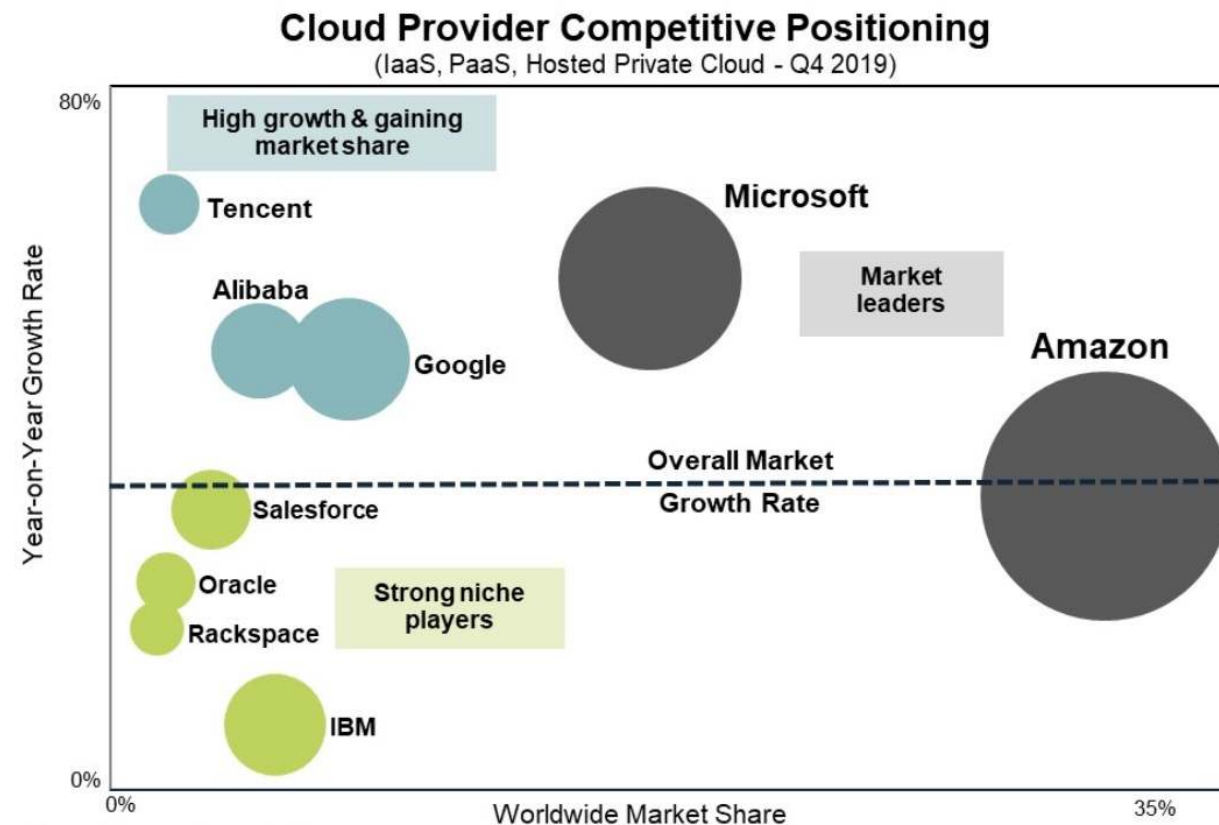
Scientific publications' semantic analysis

(2014-2019) Source: Web of Science

| Google | Amazon | Microsoft | Tencent | Alibaba |
|-------------------------------|------------------------------|------------------------------|------------------------------|------------------------------|
| machine learning | machine learning | machine learning | neural network | neural network |
| neural networks | deep neural networks | speech recognition | convolutional neural network | recommender systems |
| speech recognition | neural network | data sets | social networks | reinforcement learning |
| Deep learning | genetic algorithm | training data | machine learning | user behavior |
| deep neural networks | data sets | neural networks | benchmark datasets | deep neural network |
| language model | cloud computing | video coding | training data | convolutional neural network |
| acoustic models | natural language | language model | Neural Machine Translation | social networks |
| approximation algorithms | speech recognition | social networks | image retrieval | data sets |
| learning algorithms | knowledge graph | search engine | big data | natural language |
| reinforcement learning | convolutional neural network | based approach | topic model | e-commerce platforms |
| training data | acoustic model | data center | attention mechanism | proposed algorithm |
| mobile devices | training data | image retrieval | representation learning | big data |
| recurrent neural networks | data centers | natural language | computer vision | search engine |
| natural language | predictive models | computer vision | target domain | attention mechanism |
| search engines | social media | deep neural networks | domain adaptation | Online Shopping |
| computer vision | computer vision | mobile devices | transfer learning | benchmark datasets |
| automatic speech recognition | approximation algorithms | data structures | feature learning | question answering |
| efficient algorithms | data streams | web search | reinforcement learning | network based |
| convolutional neural networks | learning algorithms | convolutional neural network | learning approach | display advertising |
| energy efficiency | sentiment analysis | learning algorithms | learning algorithm | user experience |
| data mining | object detection | programming language | search results | representation learning |
| voice search | Big Data | search results | face alignment | transfer learning |
| language processing | topic models | software engineers | face images | short text |
| computational cost | transfer learning | recurrent neural network | community detection | data analytics |
| cloud computing | word embeddings | approximation algorithms | learning methods | recurrent neural networks |

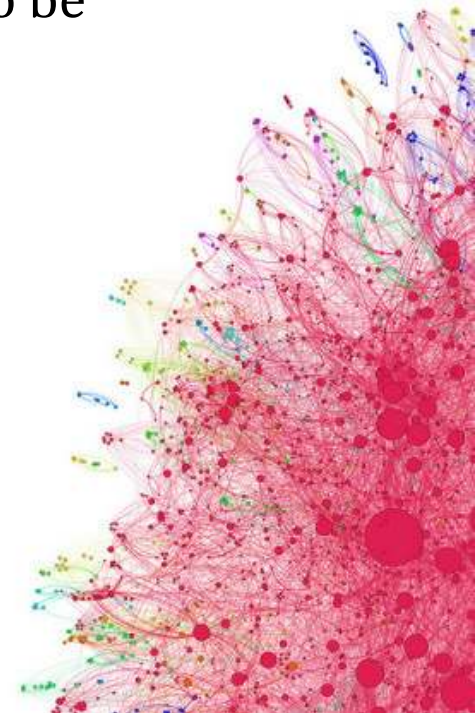
CLOUD COMPUTING, OR HOW TO CURTAIL LEARNING BY USING

- AI in the cloud: i) GPU for AI and store data, ii) access to standardized databases to train algorithms, and iii) AI models
- While clients pay as they use the services and entering fees are low, long-term contracts and high exiting fees lock them in
- **Services sold as black-boxes**: innovation's user-producer interactions as a power relation
- Cloud computing dependence allows tech giants to early notice when companies (or certain initiatives) are thriving



IM's Corporate Innovation System

- A (usually global) system organized and controlled by the IM, constituted also by subordinate organizations (such as innovating companies, universities and other PROs) participating in innovation networks
- IM defines the general R&D orientations without anticipating every step to be followed and leaving degrees of autonomy to subordinate actors
- Subordinate participants are in charge of a stage/s of the innovation process. Most of the CIS associated rents are garnered by the IM
(knowledge predation)
- Intellectual rents' concentration → polarization



Tech giants' co-authorships and co-ownerships

Source: Web of Science & Derwent Innovation

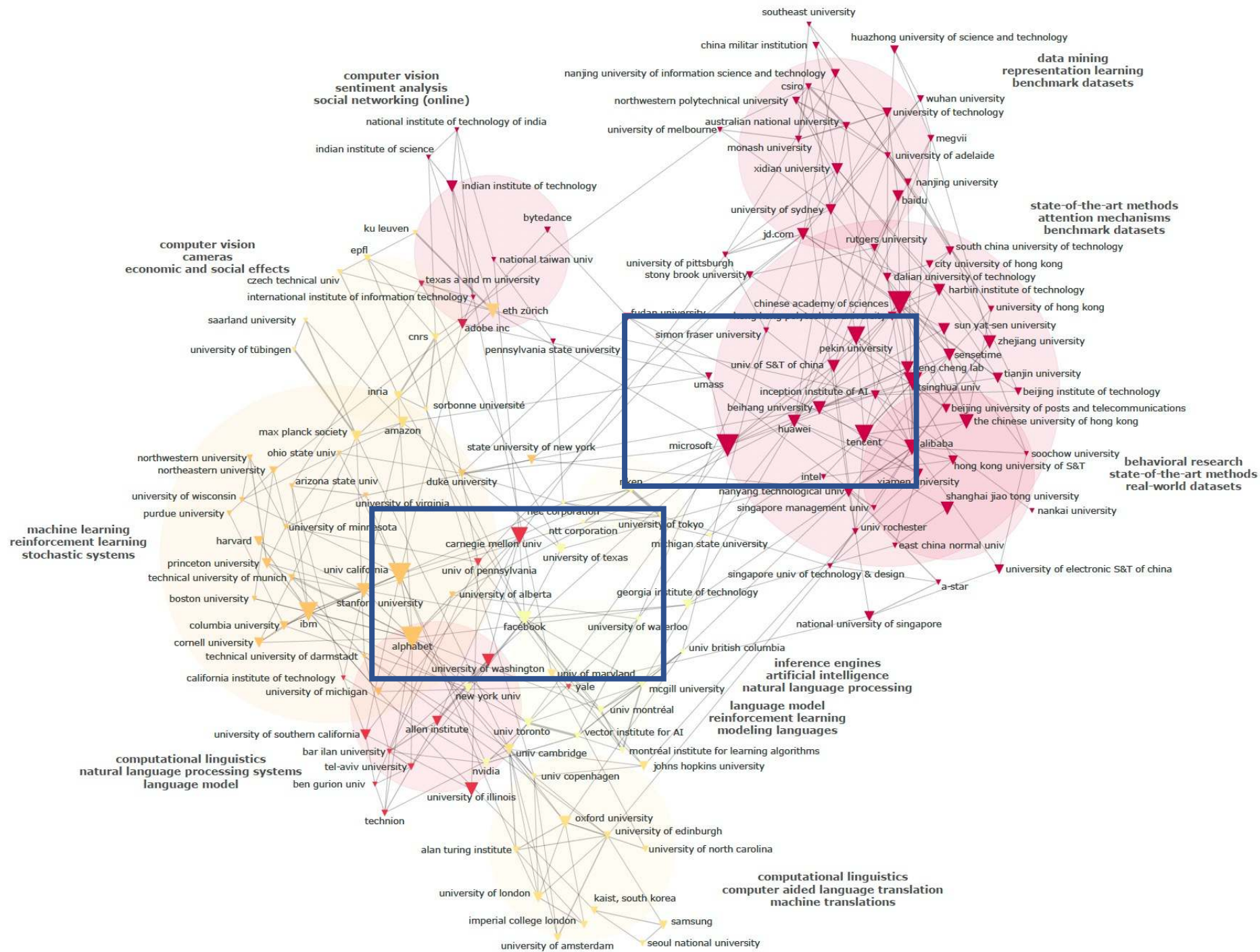
| Company | Publications (until 2019 included) | Co-authoring organizations | Applied and granted patents (until 2017 included) | Co-owned patents with other organizations |
|-----------|------------------------------------|----------------------------|---|---|
| Amazon | 824 | 766 | 10063 | 13 (0.1%) |
| Microsoft | 17405 | 4025 | 76109 | 160 (0.2%) |
| Google | 6447 | 3397 | 25538 | 65 (0.3%) |
| Tencent | 643 | 366 | 5462 | 13 (0.2%) |
| Alibaba | 685 | 427 | 3532 | 0 (0%) |

Dominating the AI research agenda

Top AI conferences

(2018-2020)

Source: Scopus



What is to be done? STI policy & more

STI policies by themselves will most likely favour IMs → knowledge predation + IM set STI agendas. Market regulations are ill-equipped

A new common knowledge regime (+ public & free education). Global & regional public natural monopolies (Socialize data and algorithms)

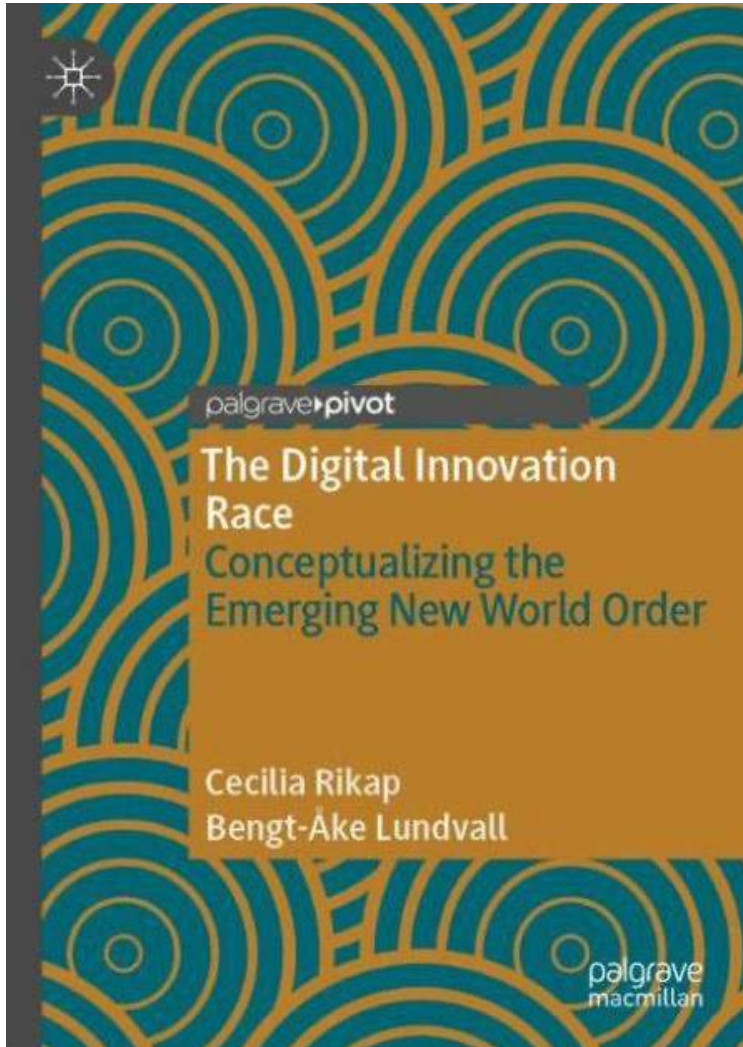
Feasible: automatic waiver for knowledge that could alleviate social, health, ecological hardships

What is to be done? Taxing

Progressive tax on owners of digital databases created from centralizing third-party data.

Taxing leading corporations' revenues and financial gains: shareholders & asset managers.

A recovery plan: Integrated economic, political, social and ecological plan for humanity.



CAPITALISM, POWER AND INNOVATION

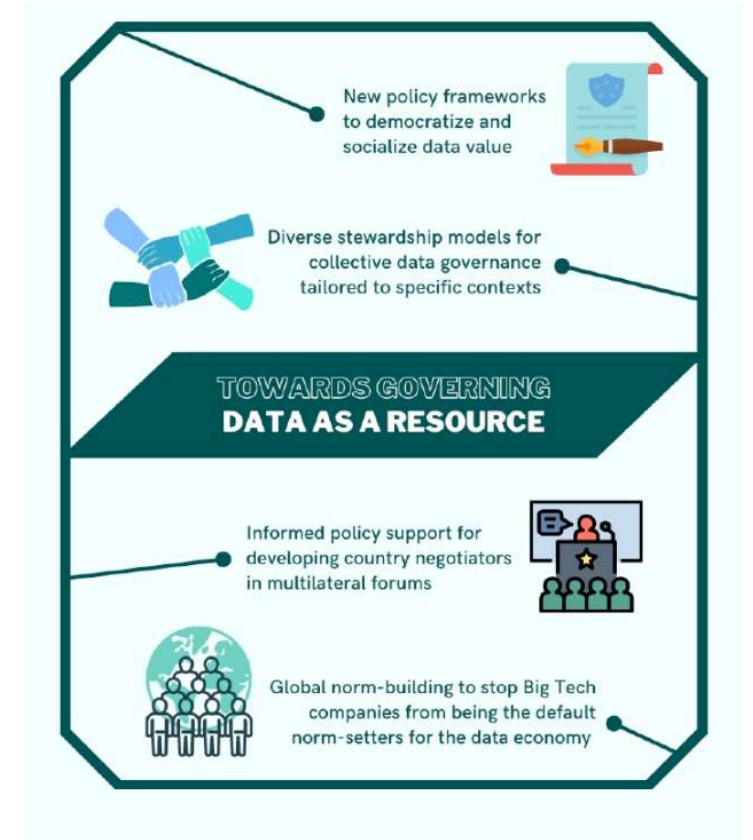
**INTELLECTUAL MONOPOLY CAPITALISM
UNCOVERED**

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Thanks

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Governing Data

Innovation & capitalist competition

| | Schumpeter Mark I: creative destruction | Schumpeter Mark II: Big corporation as the innovator | Intellectual Monopoly |
|---|--|--|---|
| Who innovates | Any firm in the industry | Big firms in the industry | The same leading corporations with maximum absorptive capacity and concentration of intangible assets |
| Firms' technological differentiation | No | Yes | Yes |
| Learning by using of the adopter (adapter)? | Yes | Yes | Yes |
| Predominant Learning Mode(s) | DUI learning (also STI learning) | STI learning (also DUI learning) | STI learning (also DUI learning) |
| Temporality of the innovation privilege | Temporary | Temporary | Permanent |
| Innovation pace | Discrete | Discrete | Systematic |
| Schematic model | T ₁ : Firm "i" innovates | T ₁ : Big firm "i" innovates | T ₁ to T _n : The same leading firm innovates systematically. |
| | T ₂ : The rest of the industry either adopts or leaves the market. Adopters pay intellectual rents only temporary | T ₂ : The rest of the industry either adopts or leaves the market. Adopters pay intellectual rents only temporary | T ₂ to T _{n+1} : Complier firms adopt some modular innovations but always paying intellectual rents |
| | T ₃ : Firm "j" innovates | T ₃ : Big firm "j" innovates | T _{1+x} to T _{n+x} : The rest of the industry only occasionally adopts innovations remaining as laggards or leaves the market |
| | T ₄ : The rest of the industry either adopts or leaves the market. Adopters pay intellectual rents only temporary | T ₄ : The rest of the industry either adopts or leaves the market. Adopters pay intellectual rents only temporary | |
| Innovation-driven economic growth (system level) | Yes | Yes | Shrunked |