

(Multi-Agent) Reinforcement Learning: a primer, with applications in economics

Introduction and Basics of Reinforcement Learning



Aldo Glielmo*

Applied Research Team
General Directorate for IT



*All views and opinions are those of the speaker(s) and do not necessarily reflect the position of Bank of Italy



BANCA D'ITALIA
EUROSISTEMA

Meet the team

Applied Research Team



established 2013, boosted 2017



www.bankit.art

Permanent staff:
25 people

- 10 [PhDs in Computer Science/Engineering]
- 7 [MSc in Computer Science/Engineering]
- 3 [ongoing (executive) PhDs in Computer Science]
- 2 [PhD in Computational Physics/Mathematics]
- 1 [PhD in Economics]
- 1 [MSc Mathematics]
- 1 [Designer & Scientific Secretary]



Permanent staff:
25 people

- 10** [PhDs in Computer Science/Engineering]
- 7** [MSc in Computer Science/Engineering]
- 3** [ongoing (executive) PhDs in Computer Science]
- 2** [PhD in Computational Physics/Mathematics]
- 1** [PhD in Economics]
- 1** [MSc Mathematics]
- 1** [Designer & Scientific Secretary]



academic profile

- research-oriented, post-graduates
- oft ambitious/unprecedented projects

+

Academic Network



UNIVERSITY OF
OXFORD

UNIVERSITÀ
DELLA CALABRIA

POLITECNICO
MILANO 1863



SAPIENZA
UNIVERSITÀ DI ROMA

TU
WIEN

TECHNISCHE
UNIVERSITÄT
WIEN
Vienna | Austria

LUISS

ROMA
TRE
UNIVERSITÀ DEGLI STUDI



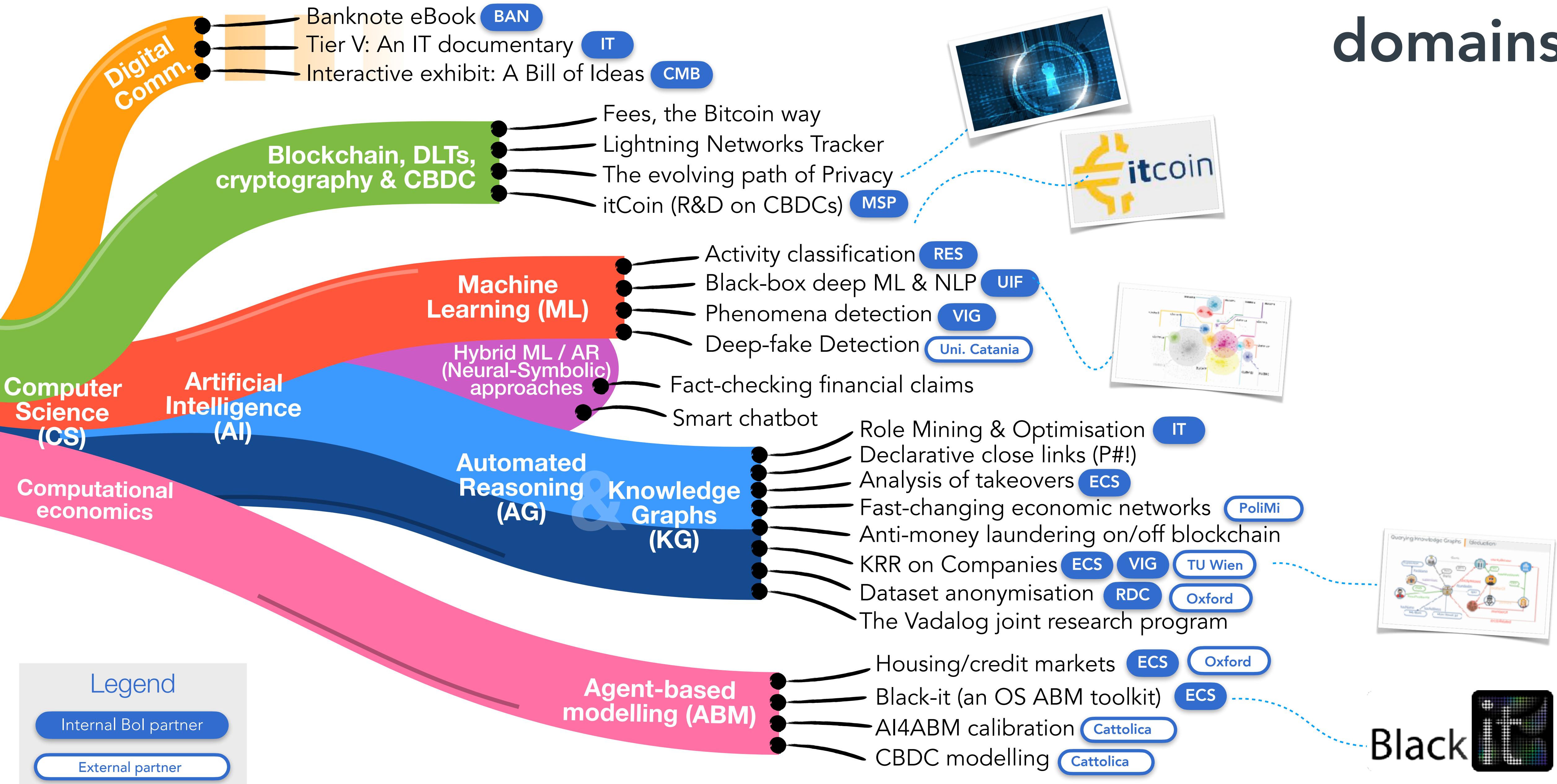
UNIVERSITÀ
DEGLI STUDI
DI MILANO

EURECOM
Sophia Antipolis



UNIVERSITÀ
degli STUDI
di CATANIA

ART R&D domains



My timeline

KCL Cond. Matter



PhD - AI4Science

- Accurate interatomic force fields via machine learning with covariant kernels, **PRB** (2017)
- Efficient nonparametric n-body force fields from machine learning, **PRB** (2018)
- SPONGE: A generalized eigenproblem for clustering signed networks, **AISTATS** (2019)
- Gaussian Process States: A data-driven representation of quantum many-body physics, **PRX** (2020)

SISSA Data Science / Stat. Bio.

Sep-2019



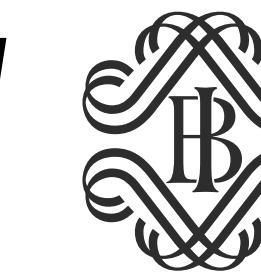
Scuola Internazionale Superiore
di Studi Avanzati

Postdoc - AI & Data Science

- Hierarchical nucleation in deep neural networks, **NeurIPS** (2020)
- Unsupervised learning methods for molecular simulation data, **Chemical Reviews** (2021)
- Ranking the information content of distance measures, **PNAS Nexus** (2022)
- Redundant representations help generalization in wide neural networks, **NeurIPS** (2022)
- DADAPy: Distance-based analysis of data-manifolds in Python, **Patterns** (2022)

Bol Applied Research Team

Oct-2021



BANCA D'ITALIA
EUROSISTEMA

Research Scientist - AI4Econ

- Intrinsic dimension estimation for discrete metrics, **PRL** (2023)
- Simulating the economic impact of rationality with reinforcement learning ..., **ICAI** (2024)
- Investigating the price determinants of the European Carbon Trading System, **Quantitative Finance** (2024)
- Natural-gas storage modelling by deep reinforcement learning, **ICAI** (2025)
- Heterogeneous RBCs via deep multi-agent reinforcement learning, **AAMAS** (2026)



Science - PINS Prize for Neuromodulation

Get published in Science and win a \$25,000 Grand Prize!

Deadline: March 15, 2019

SHARE

RESEARCH ARTICLES | PHYSICS



295



Solving the quantum many-body problem with artificial neural networks

Giuseppe Carleo^{1,*}, Matthias Troyer^{1,2}

nature

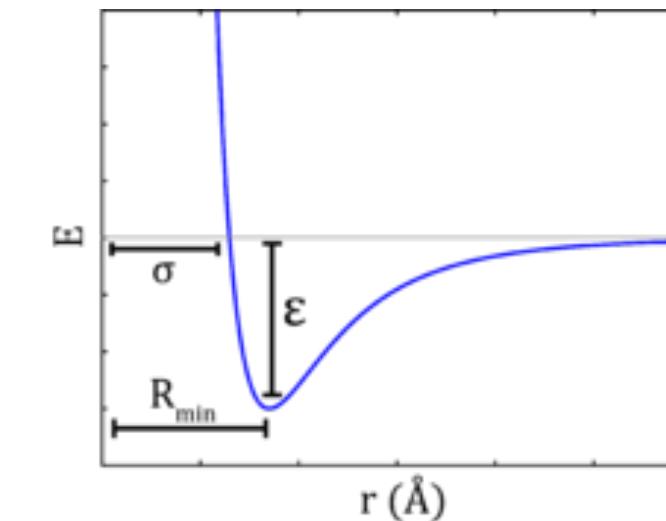
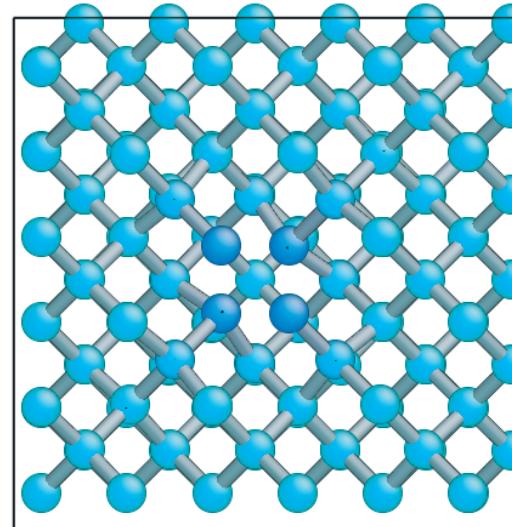
Explore content ▾ About the journal ▾ Publish with us ▾

Review Article | Published: 25 July 2018

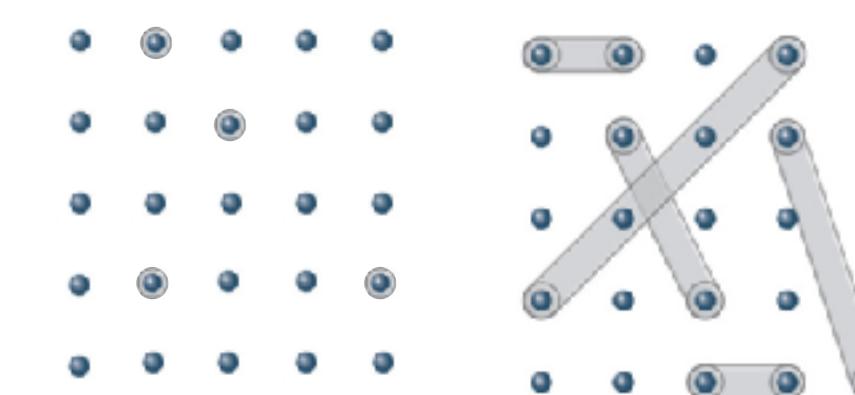
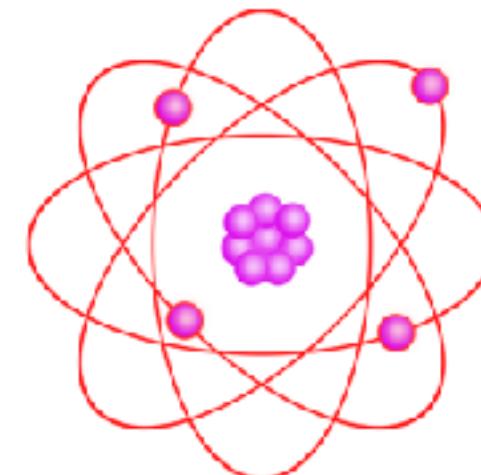
Machine learning for molecular and materials science

Keith T. Butler, Daniel W. Davies, Hugh Cartwright, Olexandr Isayev✉ & Aron Walsh✉

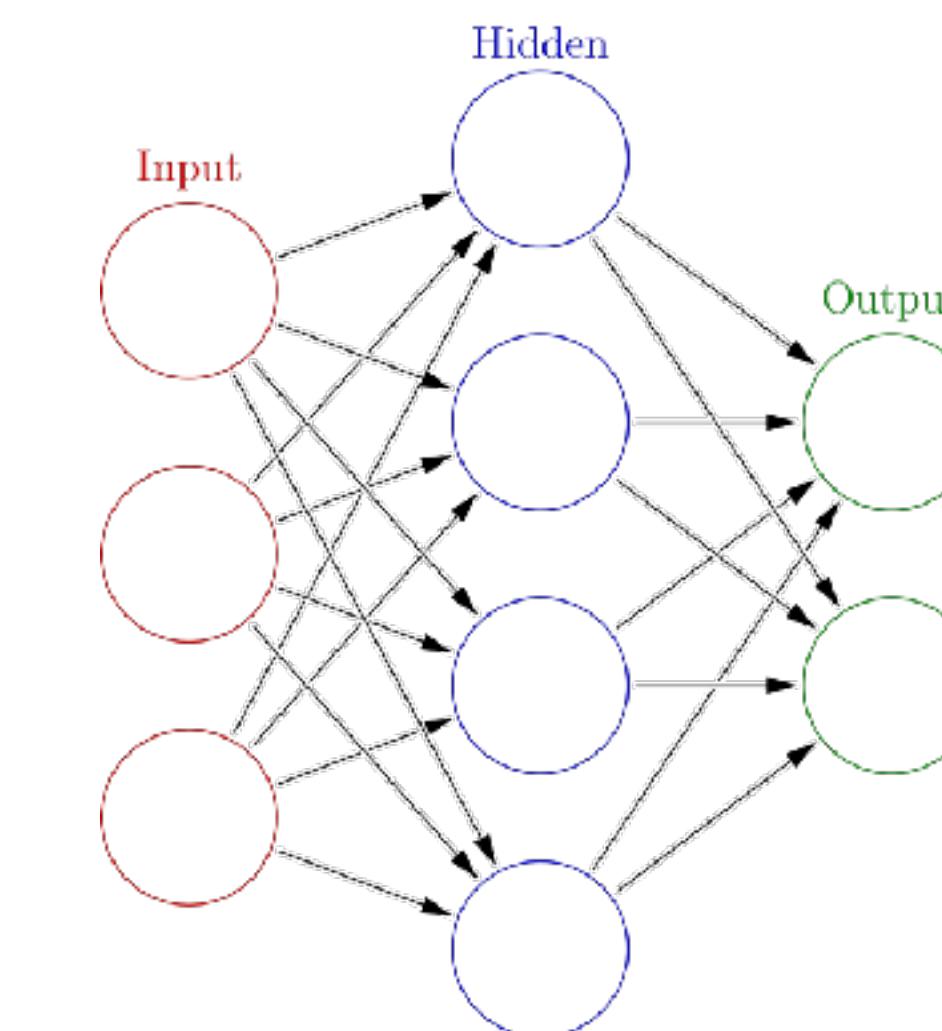
Parametric potential for atoms



Parametric wave functions for electrons



Neural network (ML) models trained with supervised or reinforcement learning

Kristof T. Schütt · Stefan Chmiela
O. Anatole von Lilienfeld
Alexandre Tkatchenko · Koji Tsuda
Klaus-Robert Müller Editors

Machine Learning Meets Quantum Physics

Generalized Neural-Network Representation of High-Dimensional Potential-Energy Surfaces

Jörg Behler¹ and Michele Parrinello²
Phys. Rev. Lett. **98**, 146401 – Published 2 April 2007Featured in Physics
Gaussian Approximation Potentials: The Accuracy of Quantum Mechanics, without the ElectronsAlbert P. Bartók, Mike C. Payne, Risi Kondor, and Gábor Csányi
Phys. Rev. Lett. **104**, 136403 – Published 1 April 2010

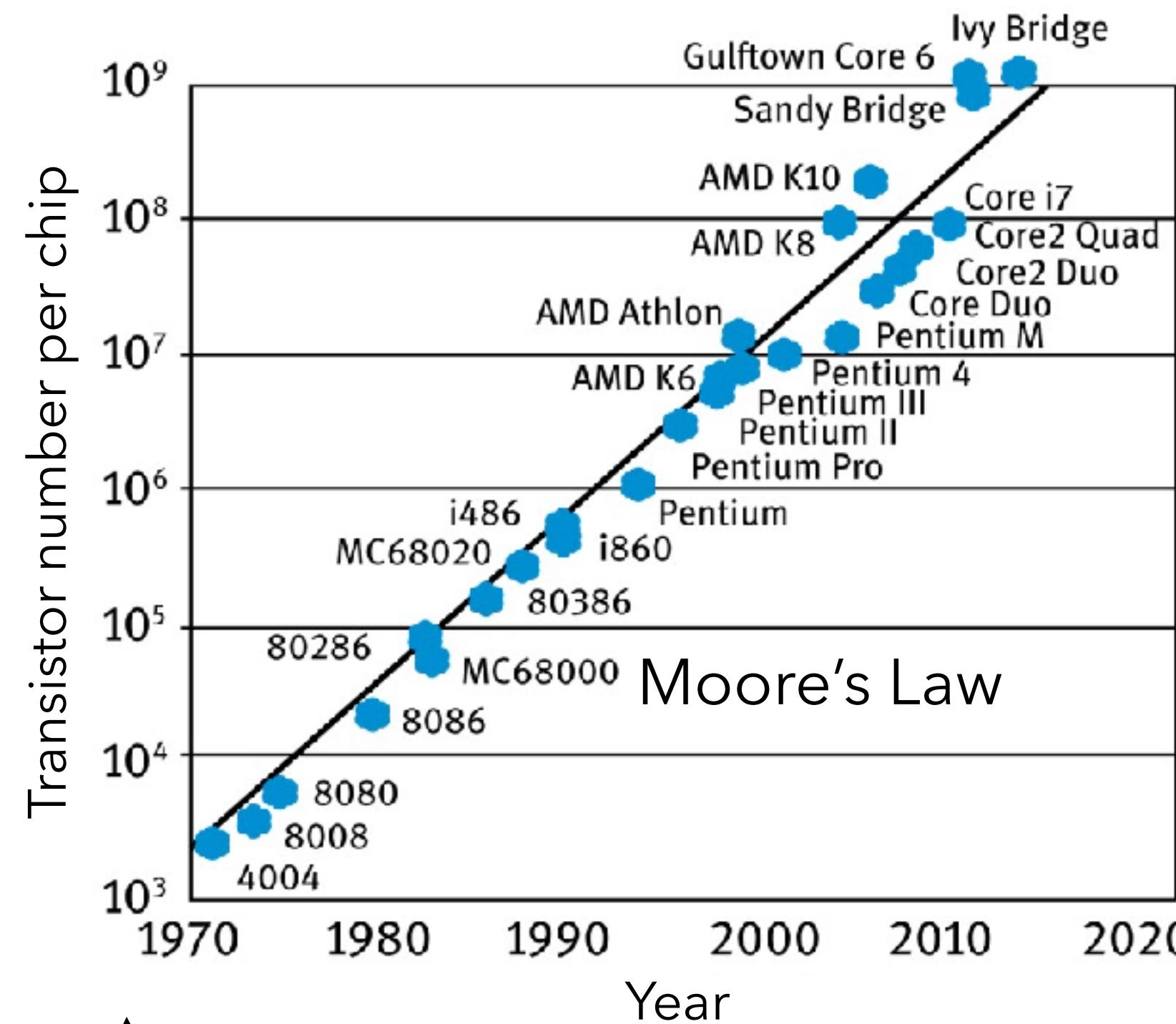
Fast and Accurate Modeling of Molecular Atomization Energies with Machine Learning

Matthias Rupp, Alexandre Tkatchenko, Klaus-Robert Müller, and O. Anatole von Lilienfeld
Phys. Rev. Lett. **108**, 058301 – Published 31 January 2012

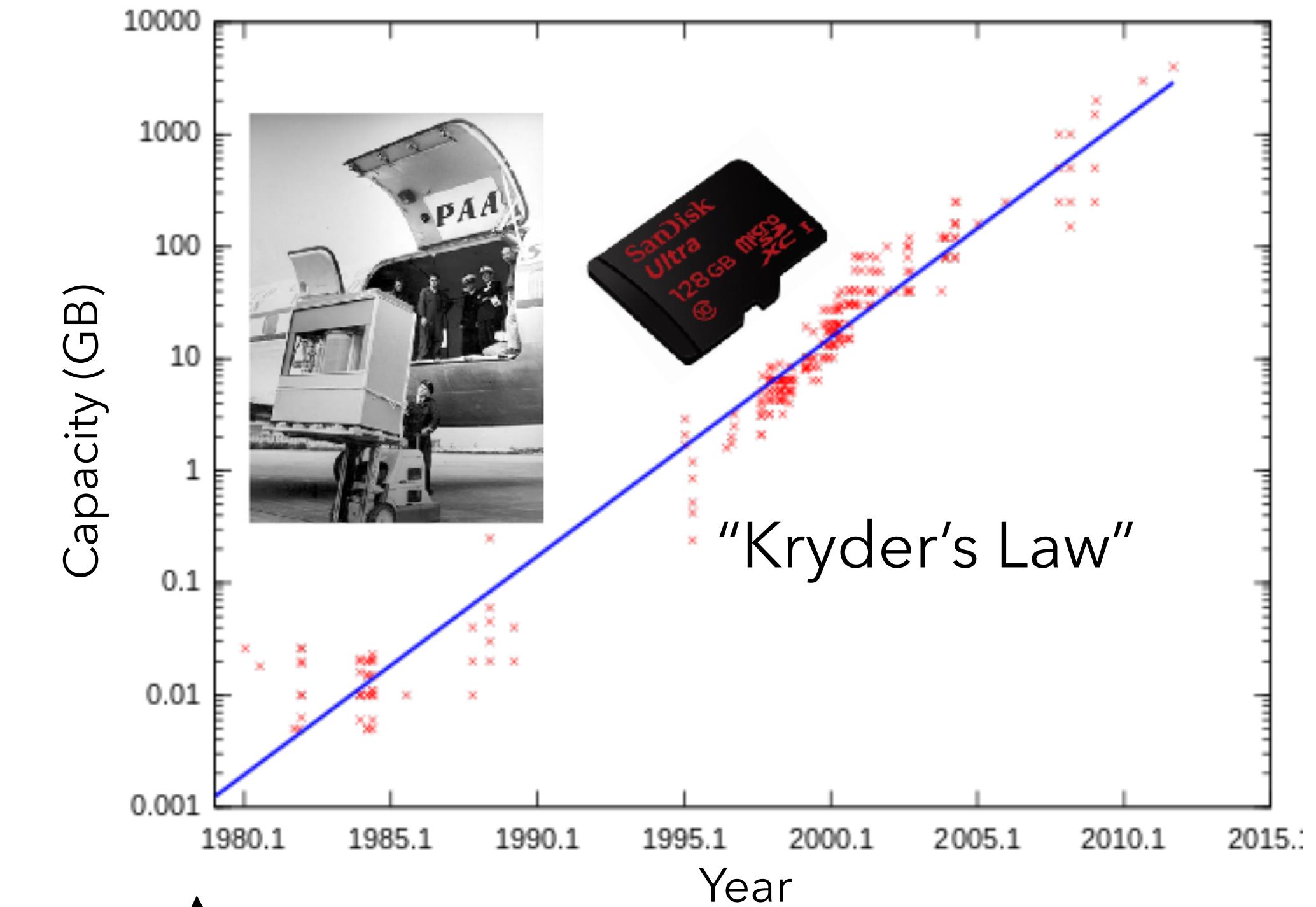
Machine learning and the physical sciences*

Giuseppe Carleo, Ignacio Cirac, Kyle Cranmer, Laurent Daudet, Maria Schuld, Naftali Tishby, Leslie Vogt-Mario, and Lenka Zdeborová
Rev. Mod. Phys. **91**, 045002 – Published 6 December 2019

Computation and data availability

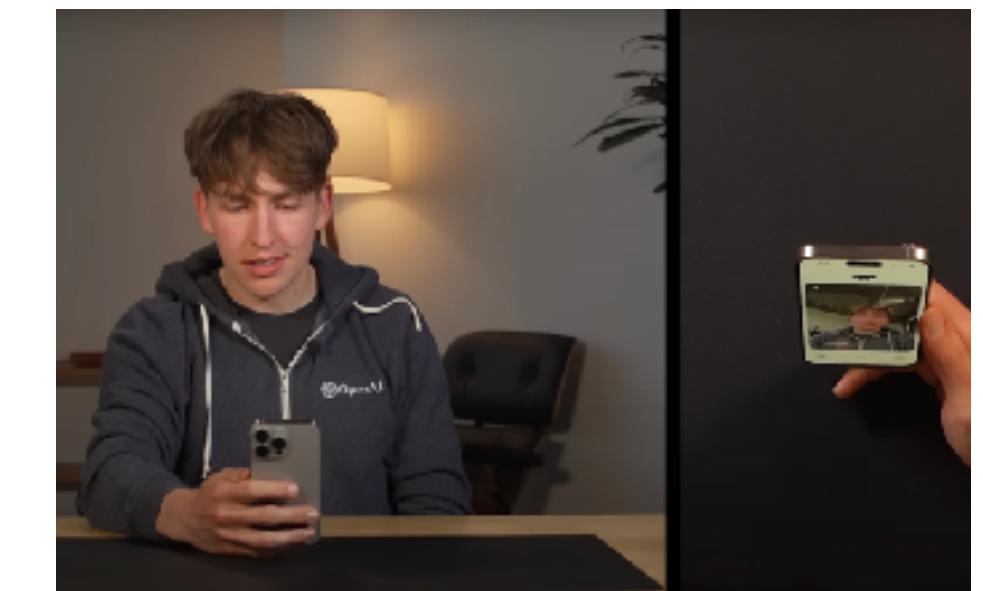
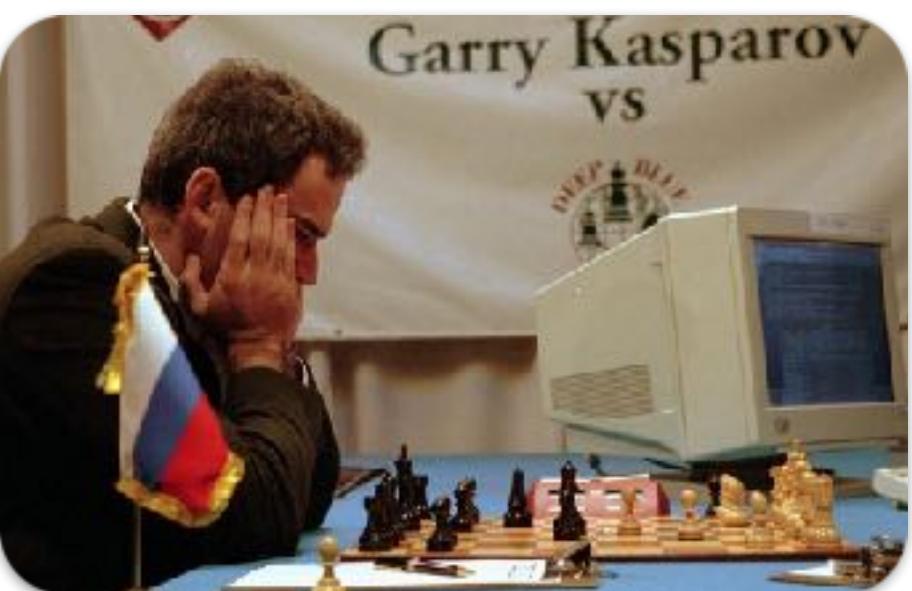
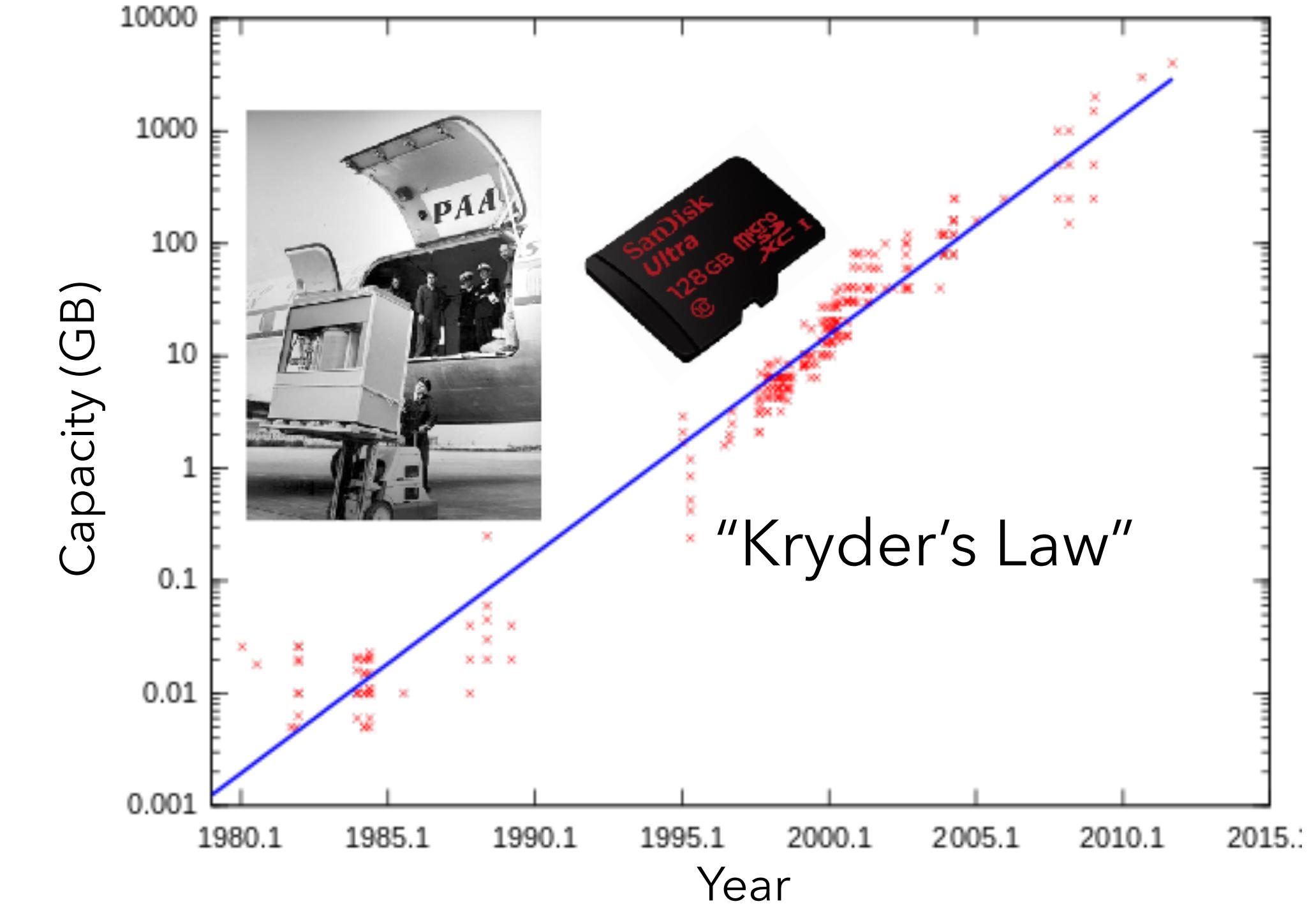
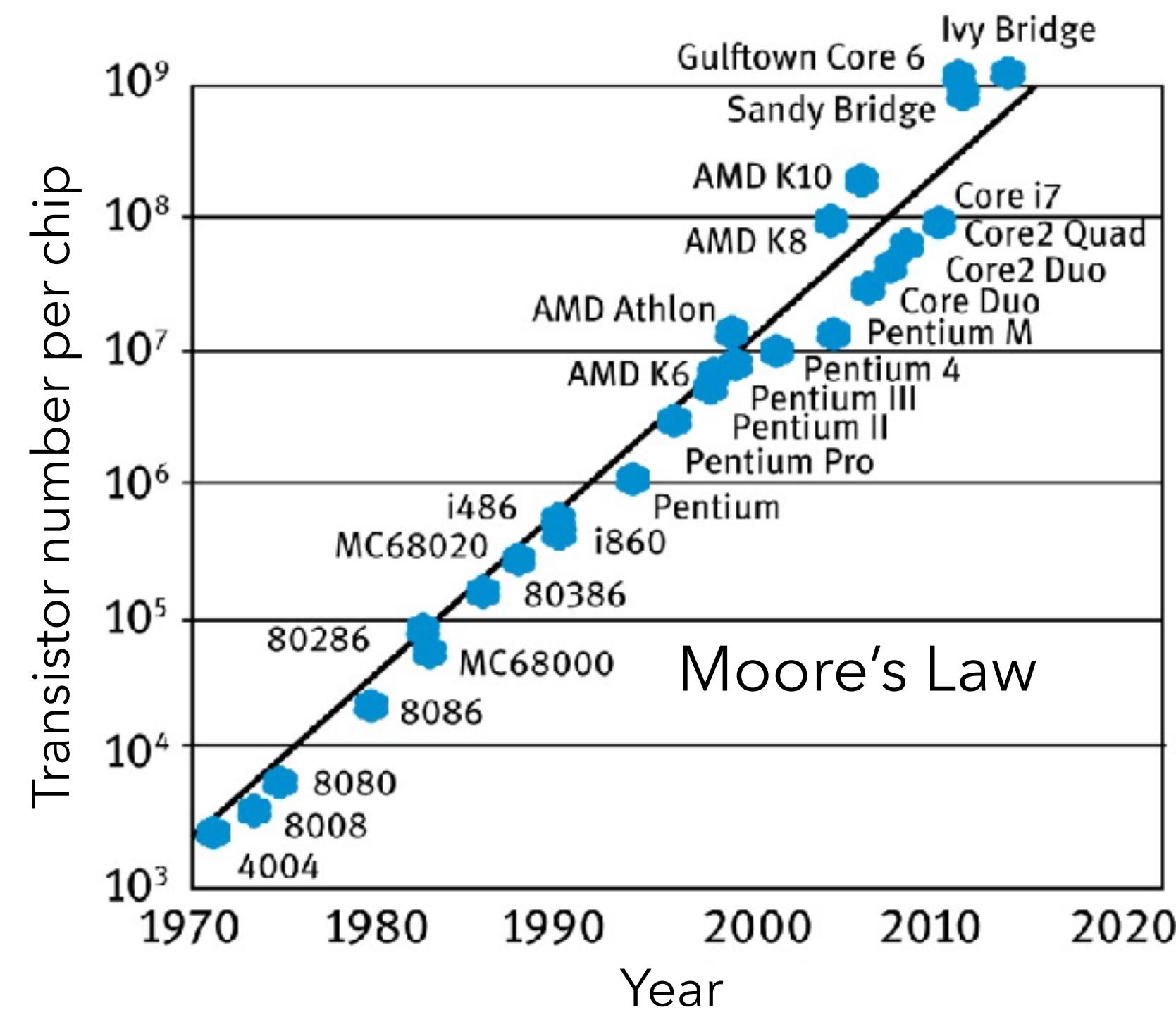


exponential increase in
computational power

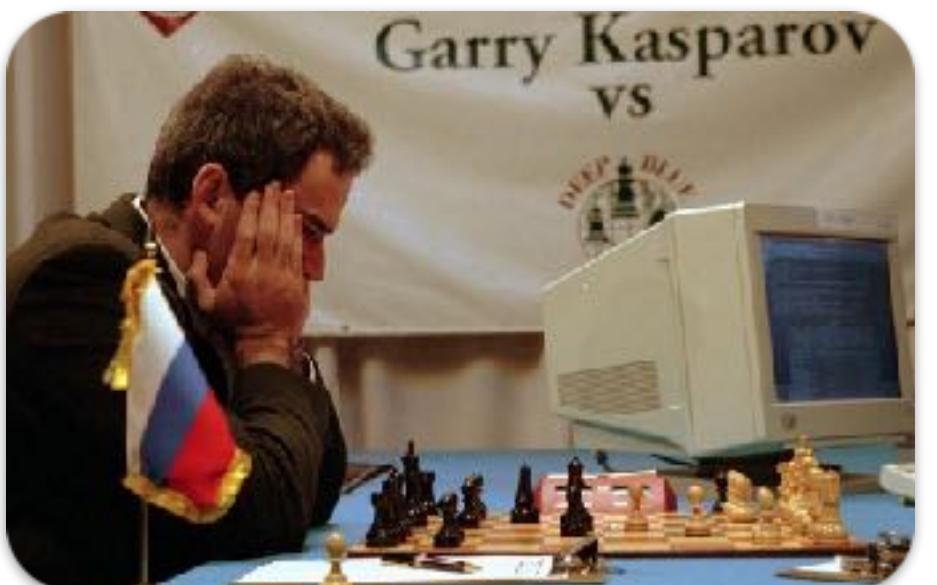


exponential increase in
data availability

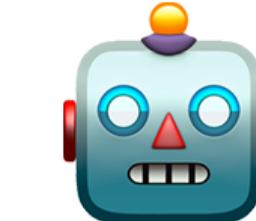
ABMs and computation: more ML



Econ and computation: more ML



Traditional
Economics



ML-driven
Economics



Based on sophisticated
human-designed carefully
crafted rules and databases

Based on algorithms that can
self-improve using *more data*
and/or *more computation*

How can we enable this transition?

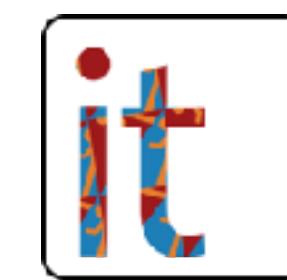
 GitHub
bancaditalia

our focus

Line

1.

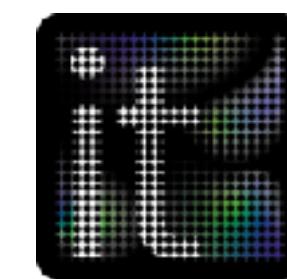
High-quality **open-source** software



Line

2.

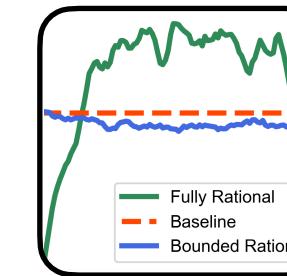
Faster ML-based **calibration**



Line

3.

AI-driven economic agents

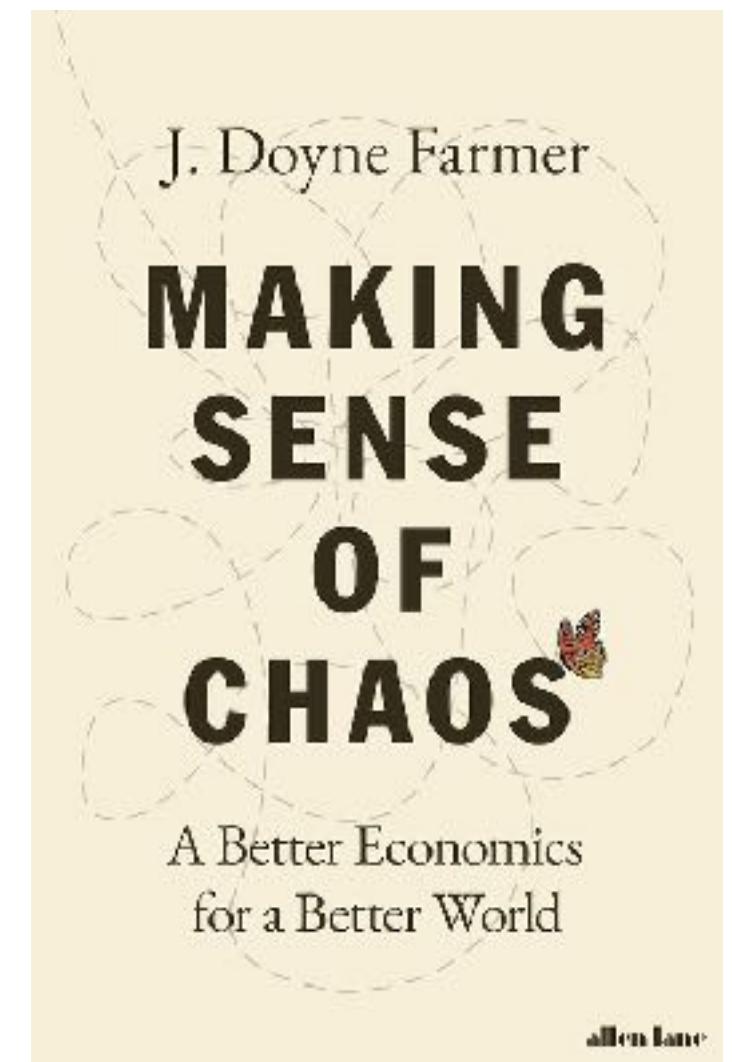


Economics = Accounting + Behaviour

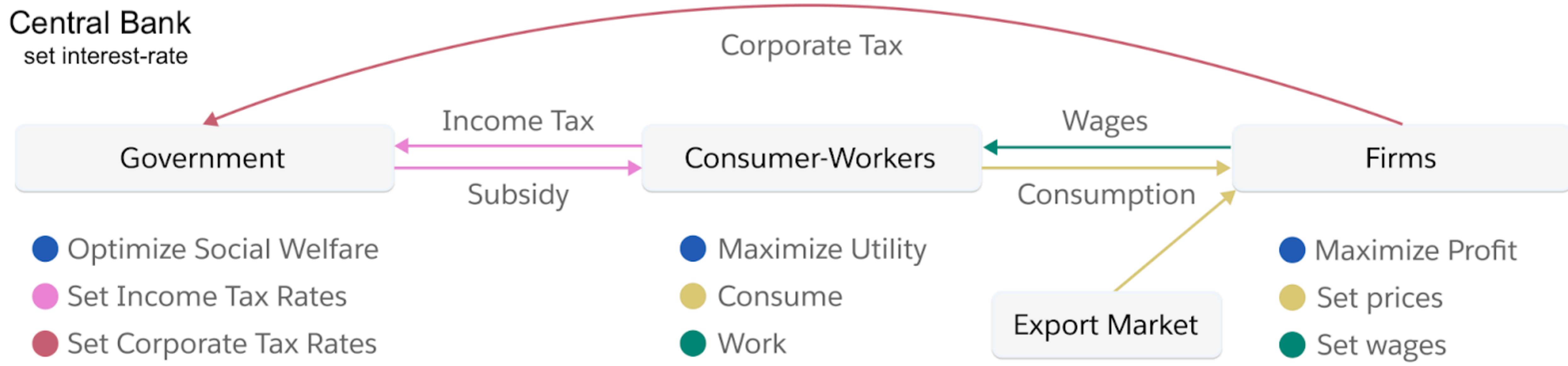
We can think about the **economy** schematically as an **interaction of accounting and human decision-making**.

Accounting is represented by the network of *balance sheets*, which continually changes as people make economic decisions. These *decisions take many forms*: What product to consume? What product to offer? Whom to hire? When to borrow? How much and from whom? All these decisions constitute economic activity. To understand the economy, we have to understand human behavior as it relates to *economic decision-making*, and we need to understand how this interacts with the underlying network of balance sheets.

Accounting and decision-making **both pose difficult problems but in very different ways**. Accounting is complicated but well-understood, whereas decision-making requires an understanding of human nature that is still incomplete.



Economics in a nutshell

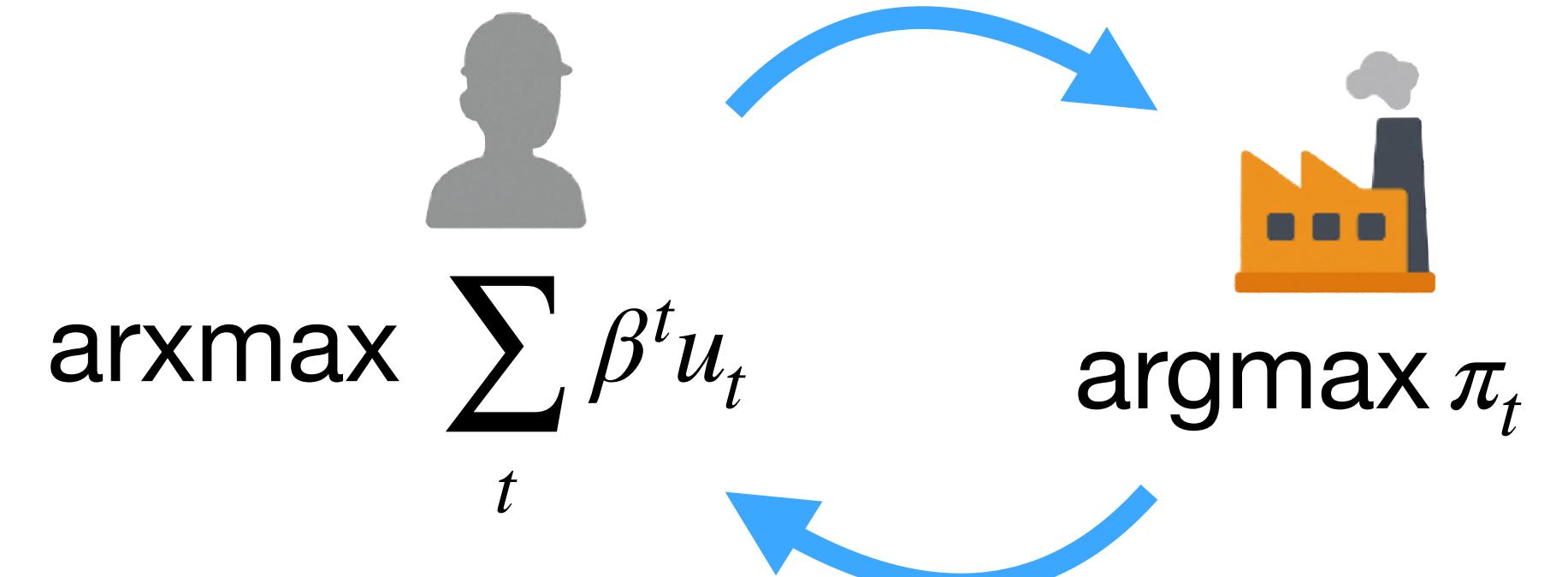


[Curry et al., Analyzing Micro-Founded General Equilibrium Models ..., 2022]

- Economics = Accounting + Behaviour
- The economy is a system of heterogeneous decision-making agents interacting through markets and making economic decisions

Traditional econ in a nutshell: the RBC

- ✗ Very limited or completely absent heterogeneity
- ✗ Unrealistic behaviour coming from assumptions of perfect "rationality"
- ✓ Behavioural rules are easy to specify in advance
- ✓ Behaviour optimally adapts to changes in policy
- ✓ Models are simple to design and analyse



$$\mathcal{L} = \mathbb{E}_0 \sum_{t=0}^{\infty} \beta^t \left[u(c_t, 1 - \ell_t) + \lambda_t (Y(k_t, \ell_t) - c_t - i_t) + \mu_t ((1 - \delta)k_t + i_t - k_{t+1}) \right]$$

- Traditional economics has focussed on decision making and has invented the "homo economicus"
- And frameworks that represent the economy as a general equilibrium outcome of intertemporally optimising representative agents facing stochastic shocks

ABMs in a nutshell

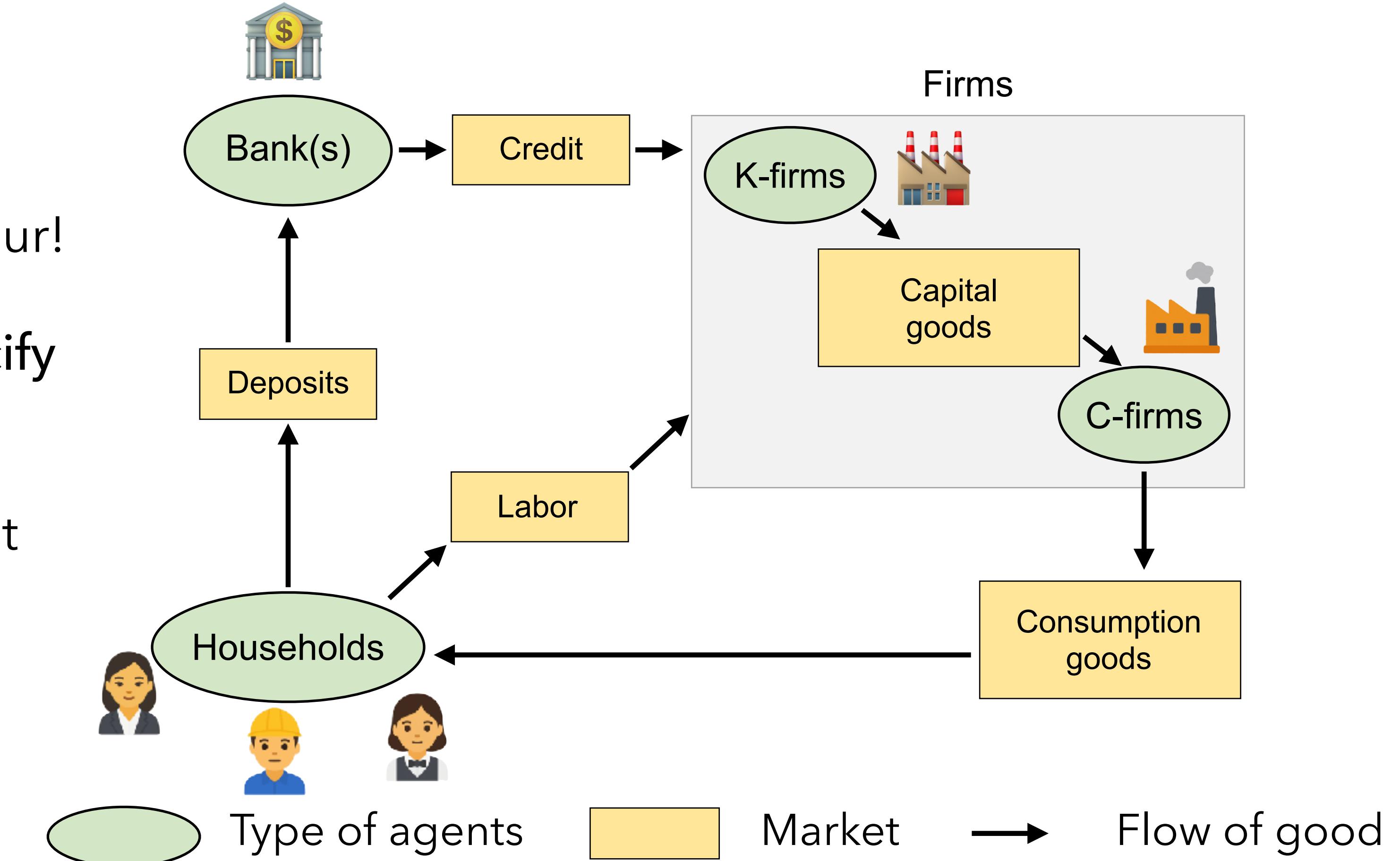
✓ Unconstrained heterogeneity!

✓ Realistic, boundedly rational behaviour!

✗ Behavioural rules are difficult to specify
in advance...

✗ Behaviour does not necessarily adjust
as a result of a changing policy

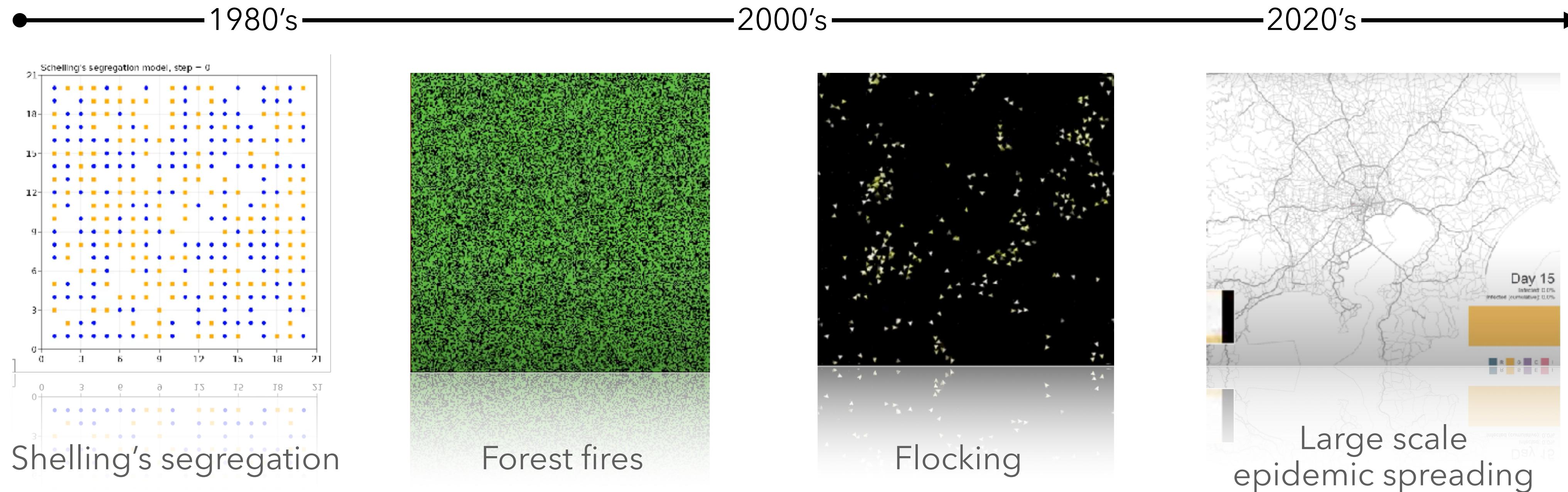
✗ Models are hard to design and
analyse...



- Agent-based economics has focussed on multiple agents, heterogeneity, and accounting
- And has built computational frameworks that represents an economy as the **outcome of interactions** among heterogeneous, boundedly rational agents whose behavior and expectations emerge from adaptive rules rather than imposed equilibrium conditions

ABMs in a nutshell

Agent-based models (ABMs) grew from toy models to realistic simulations with **thousands of heterogeneous interacting** agents



ABMs in a nutshell: central banks

High modelling **flexibility**
Intrinsically **non-equilibrium**

heterogeneity
bounded rationality

Complementary to traditional models



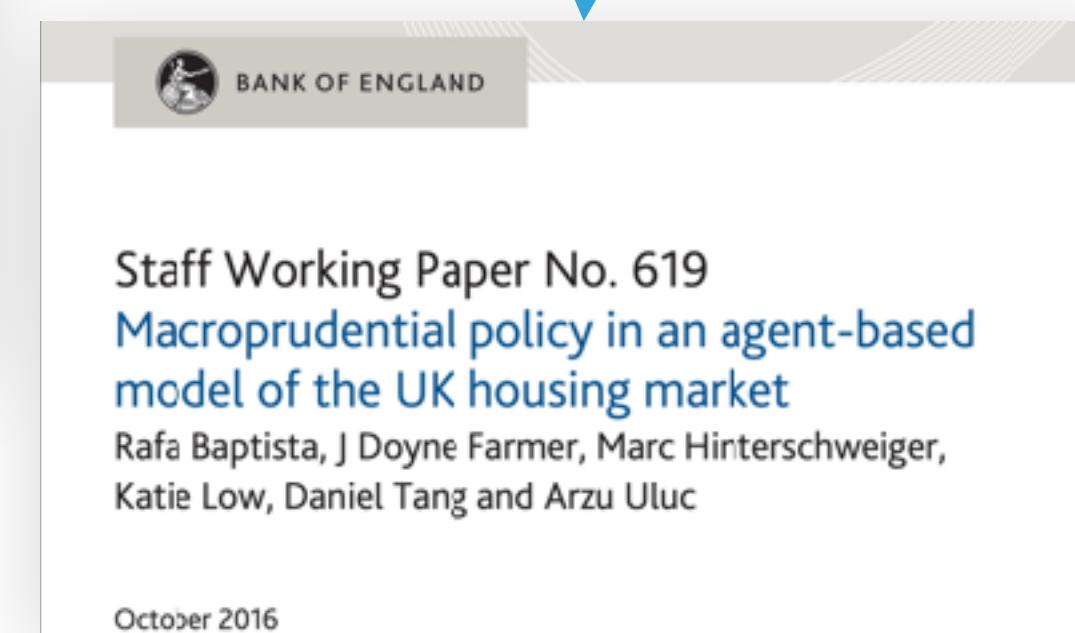
2008

No. 686 - Exploring agent-based methods for the analysis of payment systems: A crisis model for StarLogo TNG



2008

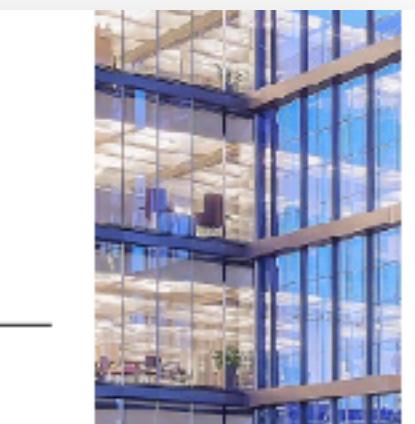
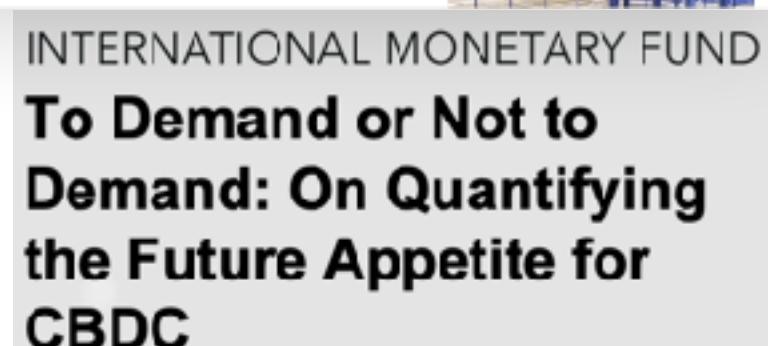
Payment systems



2016 Housing market



2021



2022-2023

Macro and CBDC

ABMs in a nutshell: central banks

Published on 12 August 2008

Working Paper No. 352

By Marco Galbiati and Kimmo Soramäki

This paper lays out and simulates a multi-agent, multi-period model of an RTGS payment system. At the beginning of the day, banks choose how much costly liquidity to allocate to the settlement process. Then, they use it to execute an exogenous, random stream of payment orders. If a bank's liquidity stock is depleted, payments are queued until new liquidity arrives from other banks, imposing costs on the delaying bank. The paper studies the equilibrium level of liquidity pooled in the system, performing some comparative statics and obtaining: i) a liquidity demand curve which links liquidity to delay costs and iii) insights on the efficiency of alternative system configurations.

An agent-based model of payment systems



BANK OF ENGLAND

Staff Working Paper No. 619 Macroprudential policy in an agent-based model of the UK housing market

Rafa Baptista, J Doyne Farmer, Marc Hinterschweiger, Katie Low, Daniel Tang and Arzu Uluc

October 2016

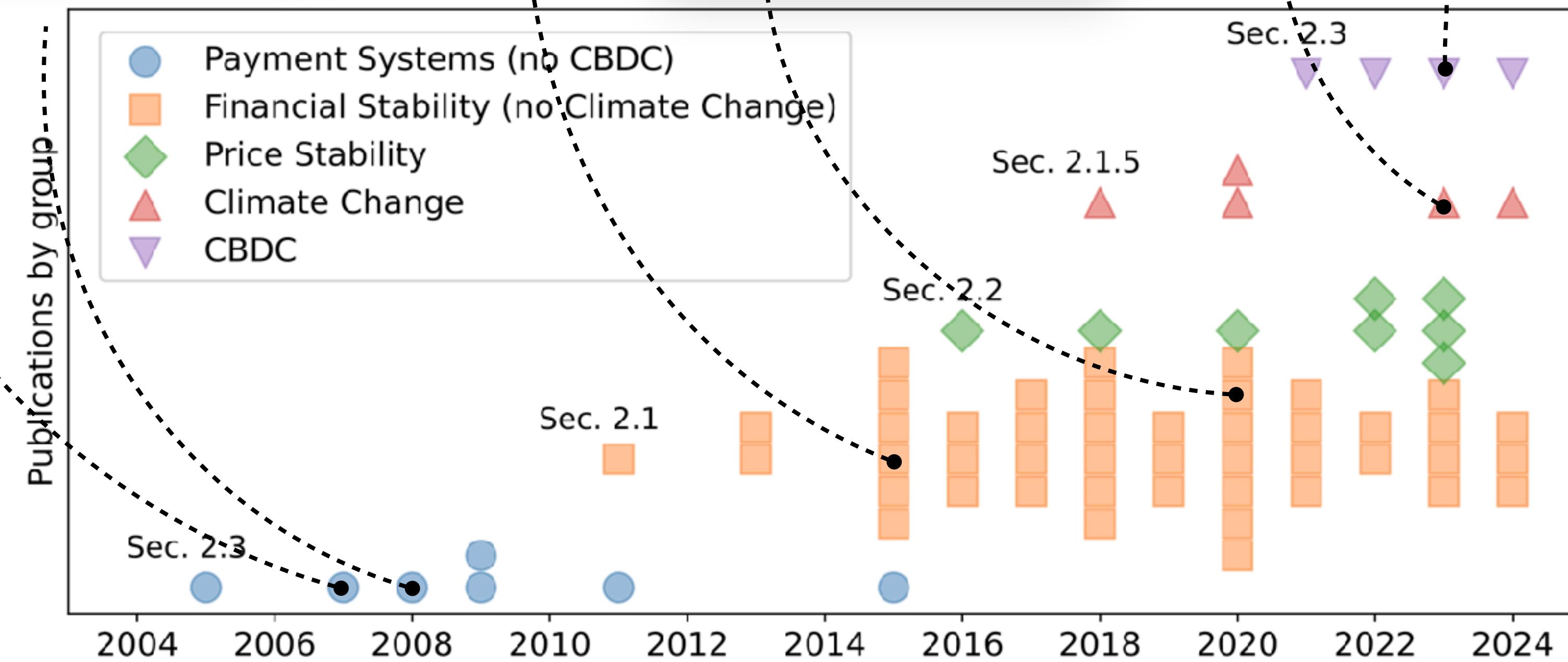
No. 1338 - Macroprudential policy analysis via an agent-based model of the real estate sector



CANVAS: A Canadian Behavioral Agent-Based Model

by Cars Hommes,¹ Mario He,² Sebastian Poledna,⁴ Melissa Siquiera⁴ and Yang Zhang²

INTERNATIONAL MONETARY FUND To Demand or Not to Demand: On Quantifying the Future Appetite for CBDC



Agent-based modeling at central banks: Recent developments and new challenges, w. Borsos (BoH), Carro (BoS), Hinterschweiger (BoE), Kaszowska (BoP), Uluc (BoE)



Book available!

What's next ?

ABMs and Traditional Econ

✓ Unconstrained heterogeneity

✓ Realistic, boundedly rational behaviour

✗ Behavioural rules are difficult to specify in advance...

✗ Behaviour does not necessarily adjust as a result of a changing policy...

✗ Models are hard to design and analyse...

✗ Very limited or completely absent heterogeneity...

✗ Unrealistic behaviour coming from assumptions of perfect "rationality"...

✓ Behavioural rules are easy to specify in advance

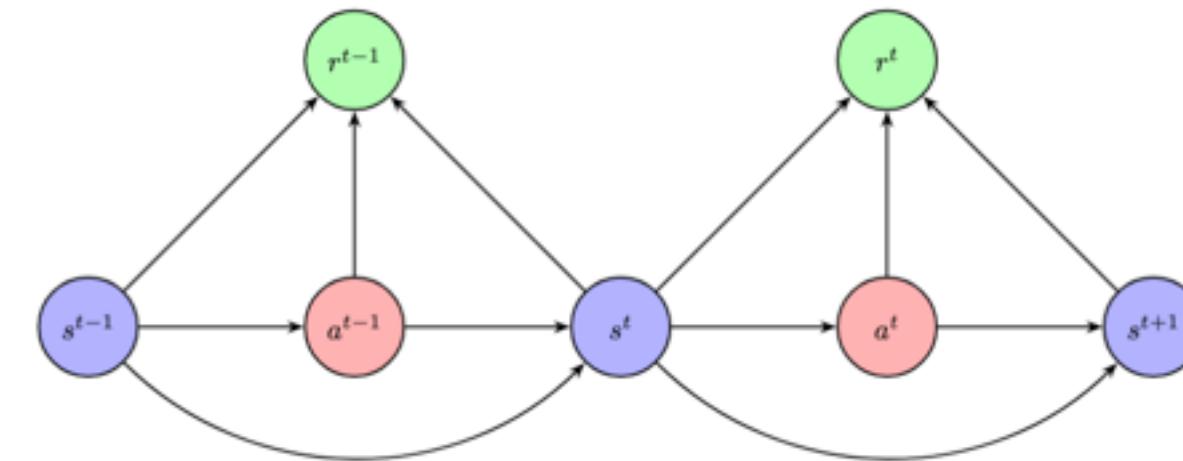
✓ Behaviour optimally adapts to changes in policy

✓ Models are simple to design and analyse

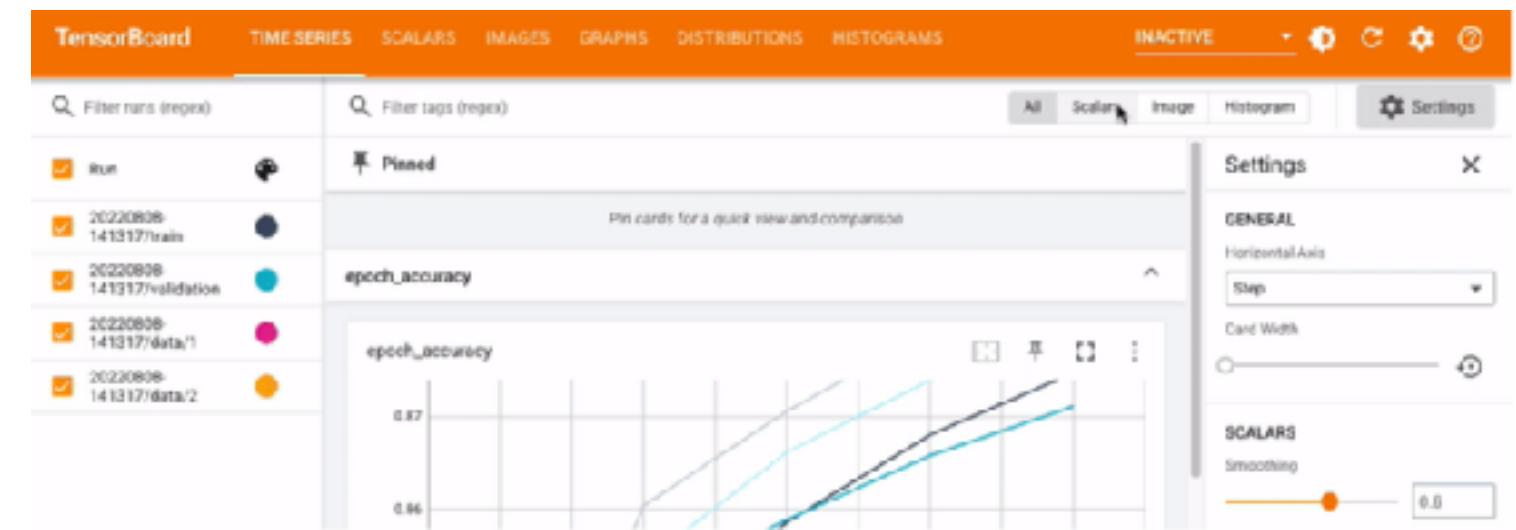
Can we do better by leveraging modern computation?

Course objectives

1. Learn the ***basic theoretical concepts*** behind Reinforcement Learning and Multi-Agent Reinforcement Learning



2. Experiment with the ***practical application*** of RL and MARL to different problems



3. Understand how these techniques are used for research at the intersection of ***economics and computer science***

