

BEYOND THE HYPE: PROMISES, PITFALLS, AND ILLUSIONS OF AI IN POLICYMAKING

EMPOWERING POLICY EXCELLENCE, UCD GEARY INSTITUTE FOR PUBLIC POLICY 4 DECEMBER 2025

Dr Omar A Guerrero University of Helsinki



ARE WE RIDING A WAVE?





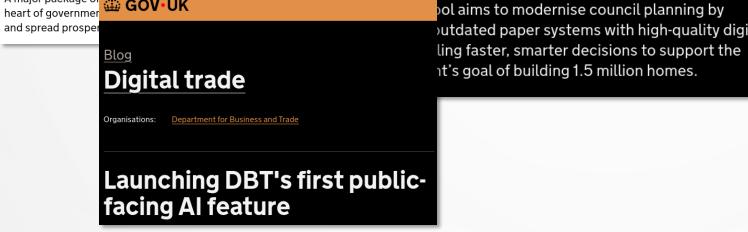
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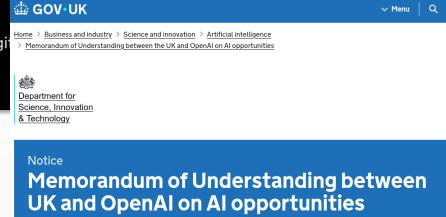
Memorandum of Understanding between the UK and Anthropic on Al opportunities

Memorandum of Understanding on Al opportunities between the UK government and Anthropic.

From: Department for Science, Innovation and Technology
Published 14 February 2025

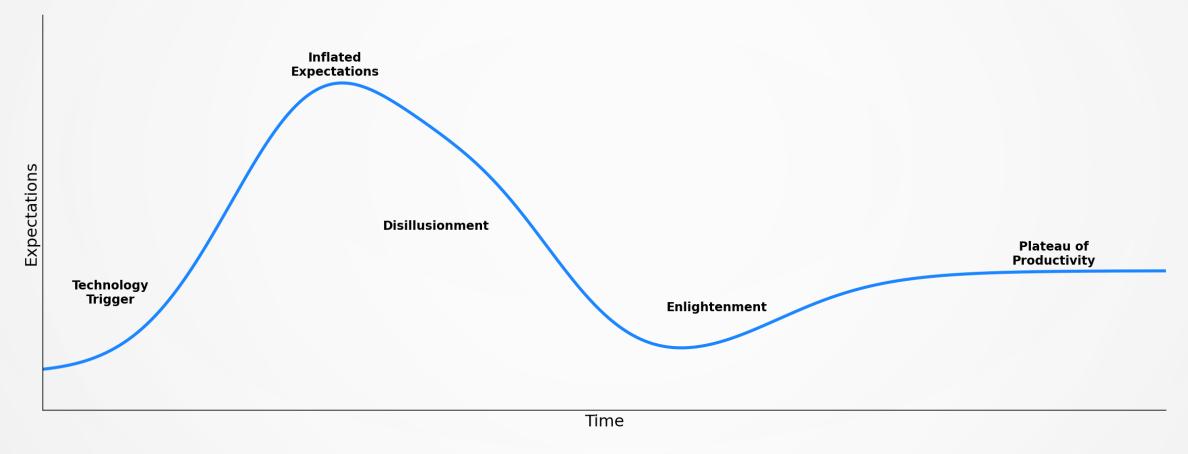
♠ GOV•UK







THE GARTNER HYPE CYCLE

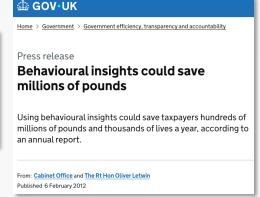


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HYPES AND ADOPTIONS IN UK POLICY DURING THE LAST DECADE





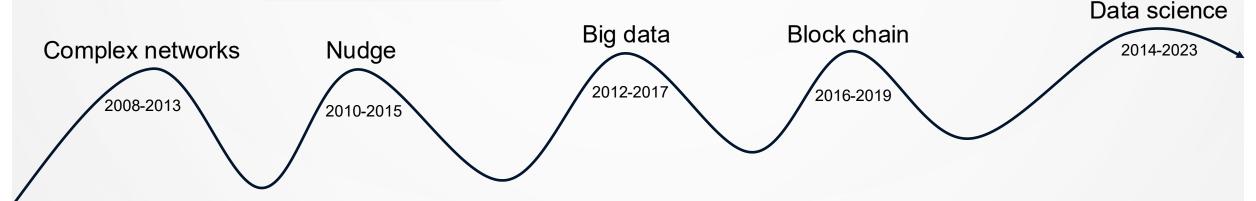


Published 6 February 2014





'spurs' around the country.





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COMMON PHRASES ABOUT AI IN MEDIA

- Unleashing creativity
- Unlocking jobs
- Multiplying productivity
- Transforming economies
- Reshaping labour market skills
- Boosting investment
- Supporting policy goals
- Enhancing better spending
- Securing the country





EVIDENCE ON AI'S IMPACT IS MIXED



Contents lists available at ScienceDirect

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journal homepage: www.elsevier.com/locate/jebo



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d Department of Economics, KU Leuven, Belgium

^a Dept. of Management, Strategy and Innovation, KU Leuven, Belgium

b Centre for R&D Monitoring (ECOOM) at KU Leuven, Belgium

Generative AI at Work 6

Artificial intelligence and firm-level productivity

Dirk Czarnitzki a,b,c,*, Gastón P. Fernández d, Christian Rammer c

Erik Brynjolfsson, Danielle Li, Lindsey Raymond

^c Department Economics of Innovation and Industrial Dynamics, ZEW - Leibniz Centre for European Economic Research, Germany

The Quarterly Journal of Economics, Volume 140, Issue 2, May 2025, Pages https://doi.org/10.1093/qje/qjae044

Published: 04 February 2025 Article history •

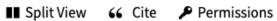
















The simple macroeconomics of Al

Daron Acemoglu

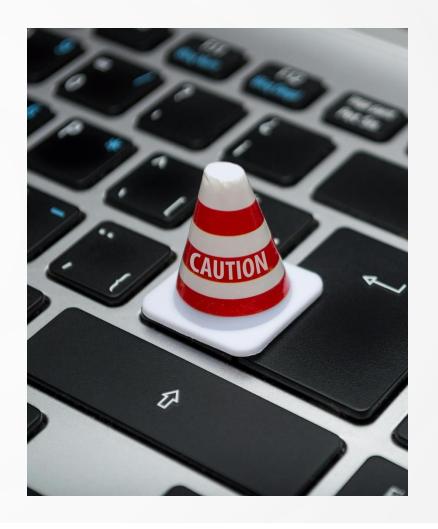
Massachusetts Institute of Technology

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HOW TO NAVIGATE A SEA OF PROMISES?

- Ask questions
- Understanding one's problems
- Knowing the data
- Understanding the ideas/epistemology behind methods
- Working around institutional constraints and incentives misalignments
- Avoiding pitfalls
- Seeing through illusions





WHAT DO WE MEAN BY AI?

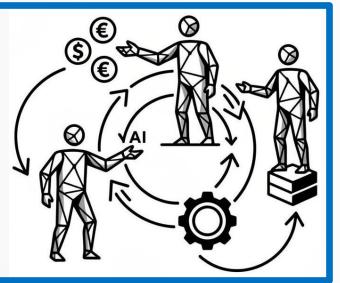
Machine learning

Agency

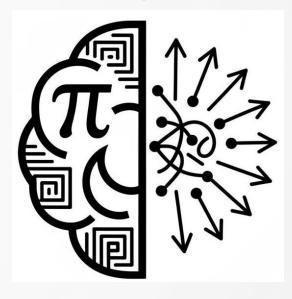
Cognition

Logic









Classification according to Russell & Norvig (2002)

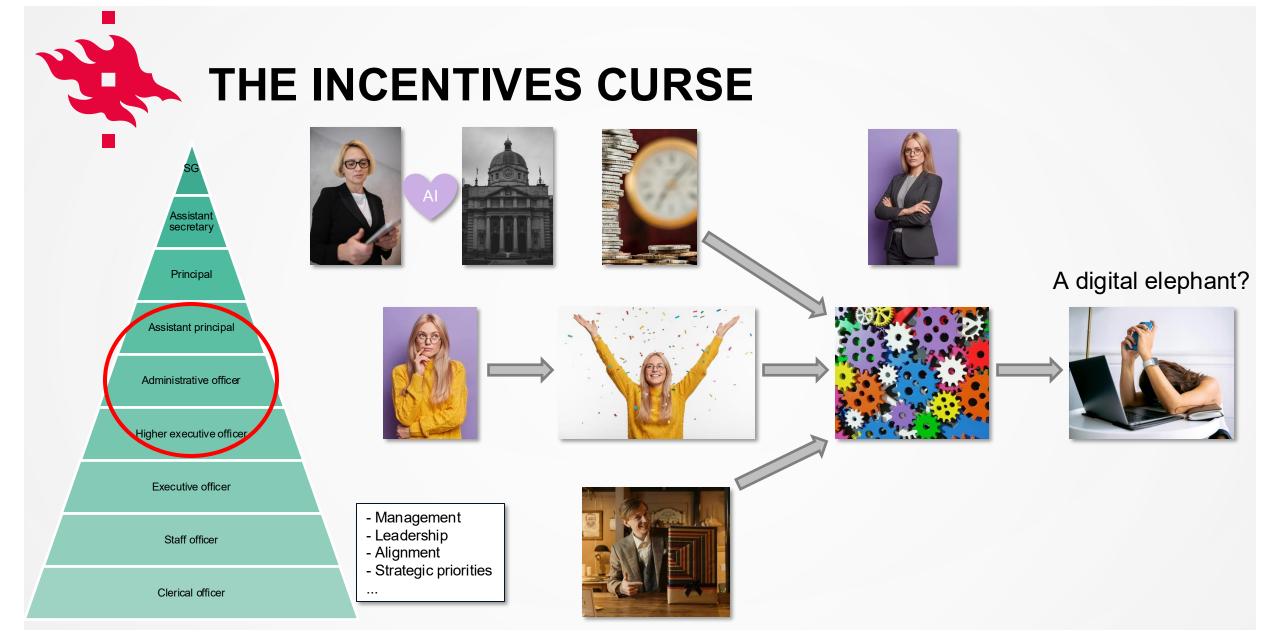


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COMMON PITFALLS WHEN ADOPTING AI

- Some common issues include bias reinforcement, hallucinations, poor replicability, lack of robustness, model drift ...
- However, most of the important issues are not inherent to methods/models, but to wrongful or careless adoptions
 - Task does not need the tool (overengineering)
 - Data are not good or the tool is not meant to work with the data
 - Hardware is not adequate
 - Contracts are imperfect
 - Skills are not ready
 - Priorities are upside down
 - Gains are marginal
 - Changes in practice are not profound



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SO MANY THINGS CAN GO WRONG

- Most of the times, inadequate adoptions are not intentional
- Institutional constraints and incentives structures set the conditions for things to go wrong:
 - Poor handover procedures
 - Bad contracts
 - Barriers to data access
 - Overwhelming paperwork
 - Prior commitments
 - Limited understanding of the method/tool

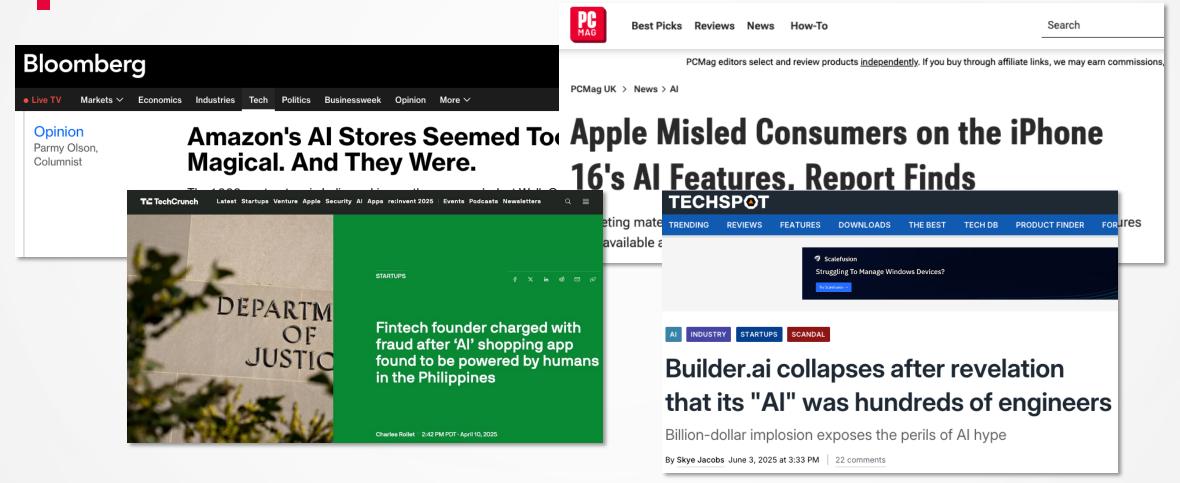




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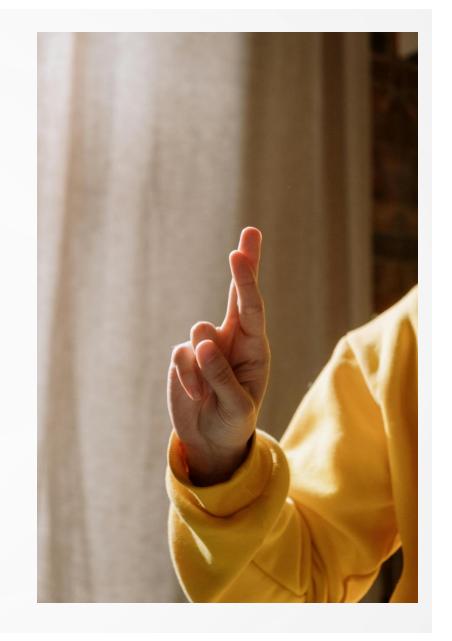
DISGUISED PITFALLS ARE NOT NEW





ON THE ILLUSIONS

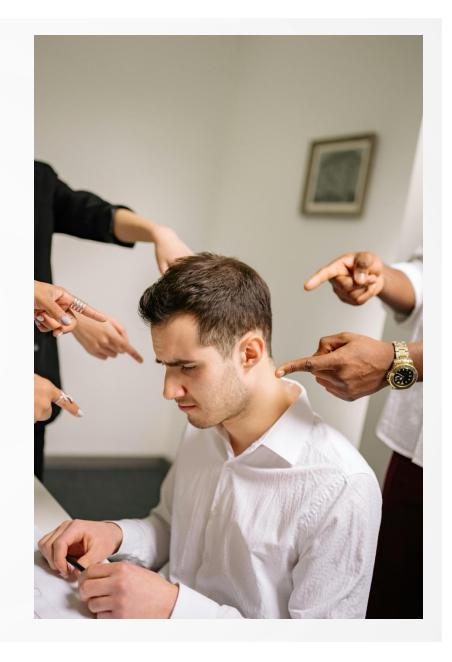
- Not necessarily intentionally misleading
- Products designed for market demand but not necessarily for policy needs
- Overblown claims and demos are not representative
- Important details are often omitted
 - Robustness
 - Uncertainty quantification
 - Training data
- The incentives curse may lead to overlook the illusions, event if one does not believe them





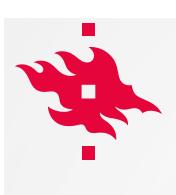
AI FOR POLICYMAKING

- Public institutions should be held accountable for their policy decisions
- Al for policymaking should be subjected to the same principles of accountability and transparency
- The misalignment of incentives creates a mined field
- Seeing through the illusions can be challenging, especially without strong internal capacity
 - Short vs long term vision
 - Capacity building vs outsourcing
 - Ability to discriminate between different technologies

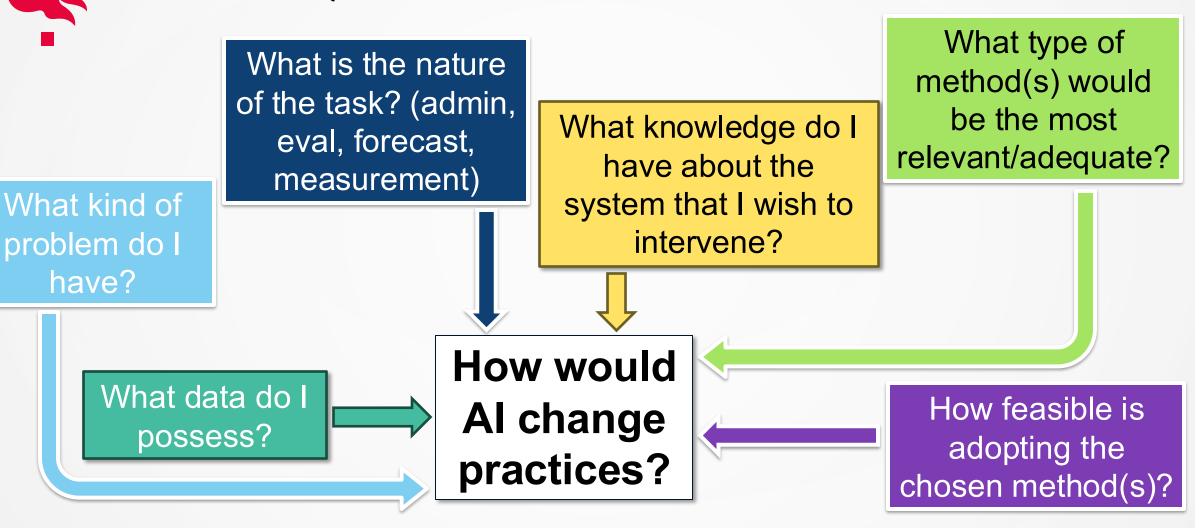




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SOME QUESTIONS WE SHOULD BE ASKING



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have?



THE 3PS APPROACH

Three dimensions to think about the adoption of AI in policymaking:

- Pace: Can some methods be adopted and deployed at a faster pace than others?
- Penetration: Can some tools achieve a deeper degree of penetration in public organisations in terms of how generalised is their adoption?
- Profundity: Can adoption translate into fundamental changes in the policymakers' understanding of the world and everyday practices?

LSE Polic

LSE Public Policy Review

Are All Policymakers Data Scientists Now? Data, Data Science and Evidence in Policymaking

RESEARCH

OMAR A. GUERRERO
HELEN MARGETTS

*Author affiliations can be found in the back matter of this article



ABSTRACT

Traditional government bureoucracies operated in a world of official statistics and filing systems, but without transactional data that might be used to inform policymaking, administrative design or the provision of public services. In a world powered by data science – there are many ways in which policymaking can be informed by data and 'data-driven'. There has been a step change in the availability of data and methods of analysis. It is no longer acceptable to rely on custom-built statistics and consultants' reports to inform high level policy discussions. We need to embed data science models into every stage of the policy process; understanding a population that the policy is to serve; running predictions on demand; elaborating detailed and sophisticated models for counter factual analysis and intervention; and simulating policy interventions to discover unintended consequences before policy interventions are introduced. These opportunities for policymaking place new demands on civil servants, and rapid updating of practices, standards and tools and the breaking down of boundaries between technical/analysis and non-technical/generalist teams.

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EYWORDS:

Data science; policymaking; artificial intelligence; digital government; information capacity

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THE 3PS SLIDE

- All models have strengths and limitations
- More specialised models may provide a better understanding at the cost of a lower capacity to be socialised
- In the short term, movement in the 3P space is constraint
- Trade-off between pace of adoption, how broad is its penetration, and how profound is the change it produces on practices

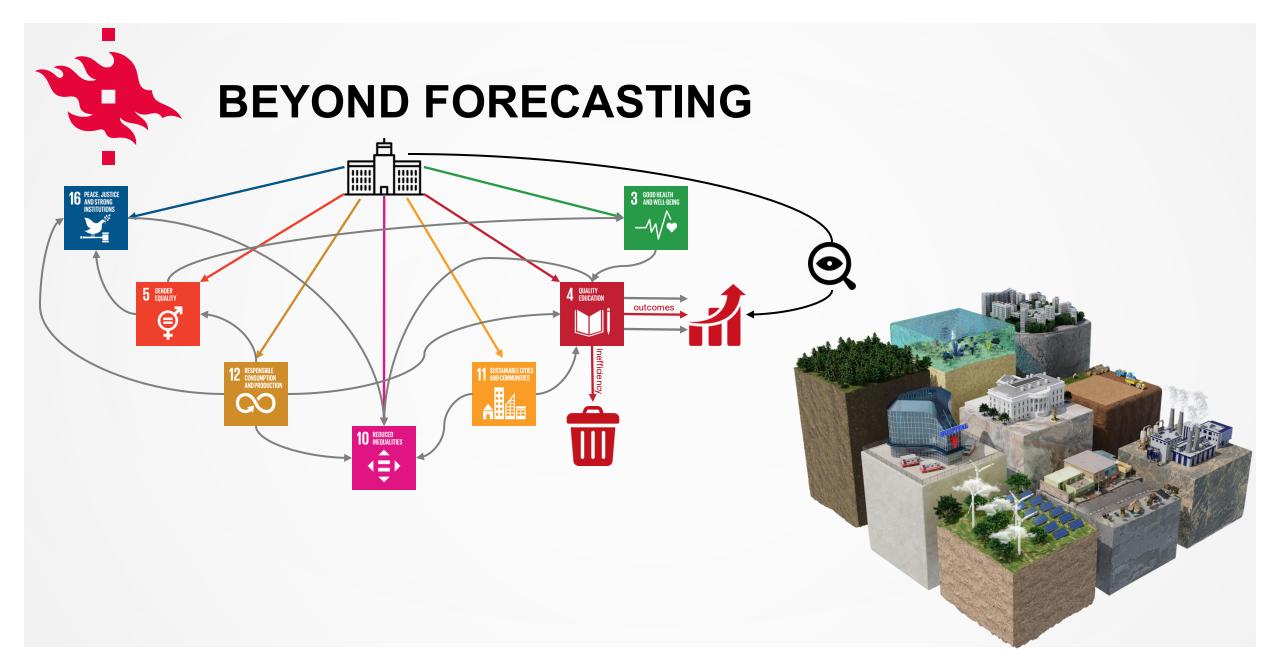




EXAMPLE: AGENCY

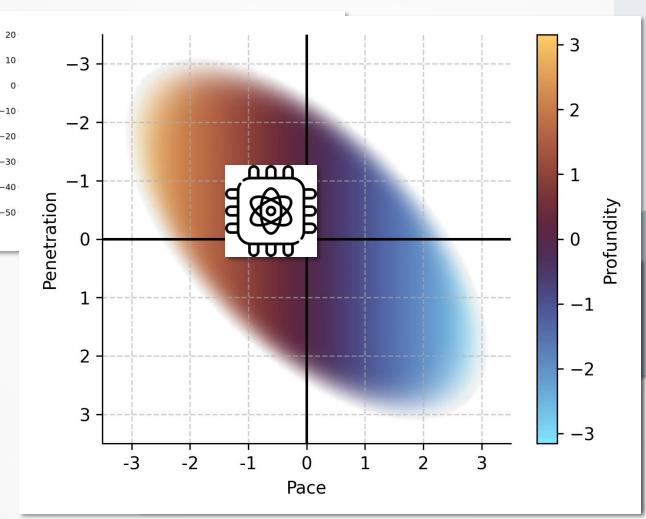
- Problem: evaluating the impact of government spending on the SDGs (UNDP)
- Nature of task: measurement and evaluation
- Data: short time series on indicators and spending
- Potential methods: econometrics, machine learning, integrated assessment frameworks, CGEs, network analysis, agent computing
- Considerations: complex interdependencies, scalability, transparency, theories of change, data intensiveness, counterfactual analysis
- Best choice: agent computing





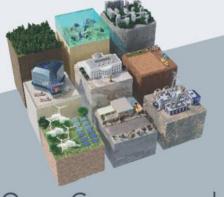


POLICY TASKS ARE DIVERSE



COMPLEXITY ECONOMICS AND SUSTAINABLE DEVELOPMENT

A Computational Framework for Policy Priority Inference



Omar Guerrero and Gonzalo Castañeda

policypriority.org

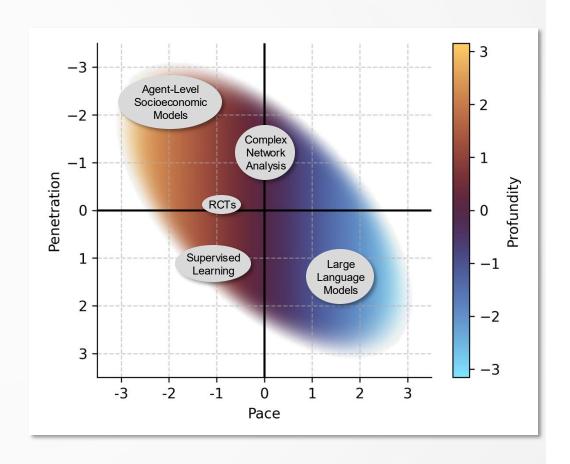
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SOME USEFUL QUESTIONS

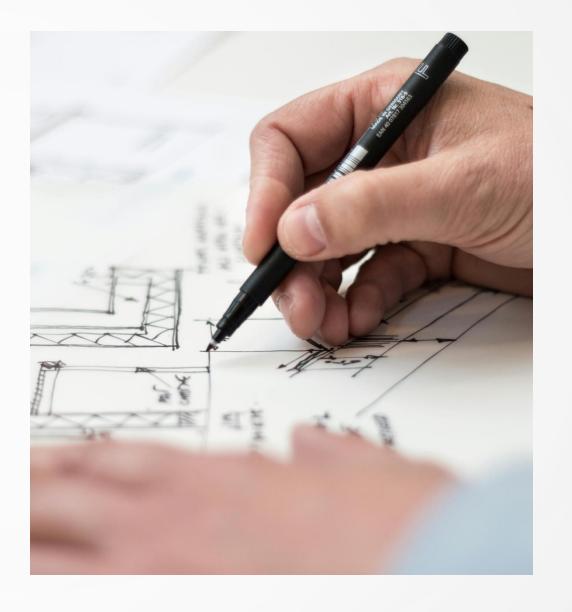
- What is the nature of the task?
- Is the method/tool adequate?
- Are the data adequate?
- What pragmatic issues do I need to consider (e.g., computational resources, human capital, data access)?
- · How robust is the approach?
- How feasible is its training/calibration?
- What is the value for money compared to alternatives that show marginally lower performance?
- Can we exploit economies of scale?
- Positioning in the 3P slide is dynamic (there is such thing as right timing)
- Complexity of the method/tool does not determine the position in the 3Ps slide





CONCLUSIONS

- Successfully adopting Al and related tools is challenging
- While it is impossible to overcome all institutional constraints and incentives misalignments, asking the right questions can help in a substantial way
- As methods mature and are socialised, they move along the 3P slide, so there are good and bad times to adopt them
- The challenge is to always have internal capacity to ask the questions, avoid pitfalls, and see through illusions





THANK YOU

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