

Epistemic Risk reduction: regulating online spaces

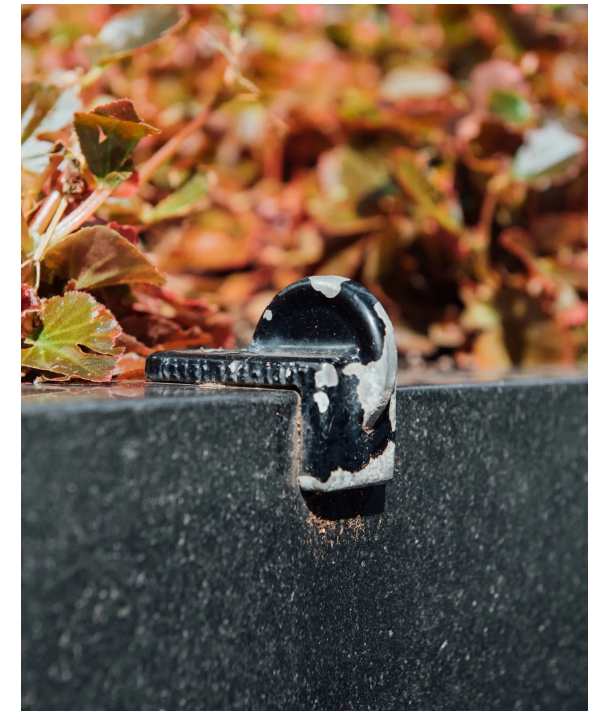
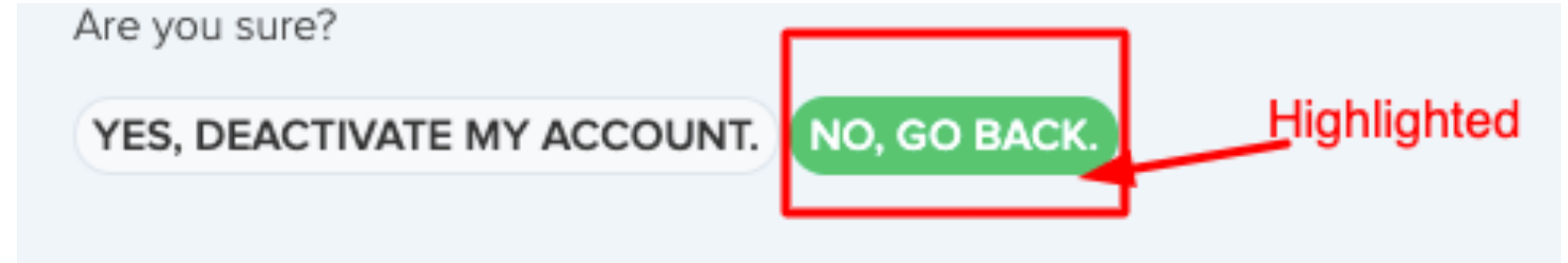
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Processing Lab, Centre for the Governance of AI



A framework

Opinion formation as
behavior

Every environment
promotes and curbs certain
behaviors



Behavior in space

Spaces embed values and determine political outcomes.

Mark Beissinger (Princeton University Press, 2022).



Epistemic Risk & Resilience

A longterm, tail-risk approach

Episteme – knowledge / wisdom

Carla Cremer, talk @ UK DCMS Committee Online Harms and Disinformation , July 2022

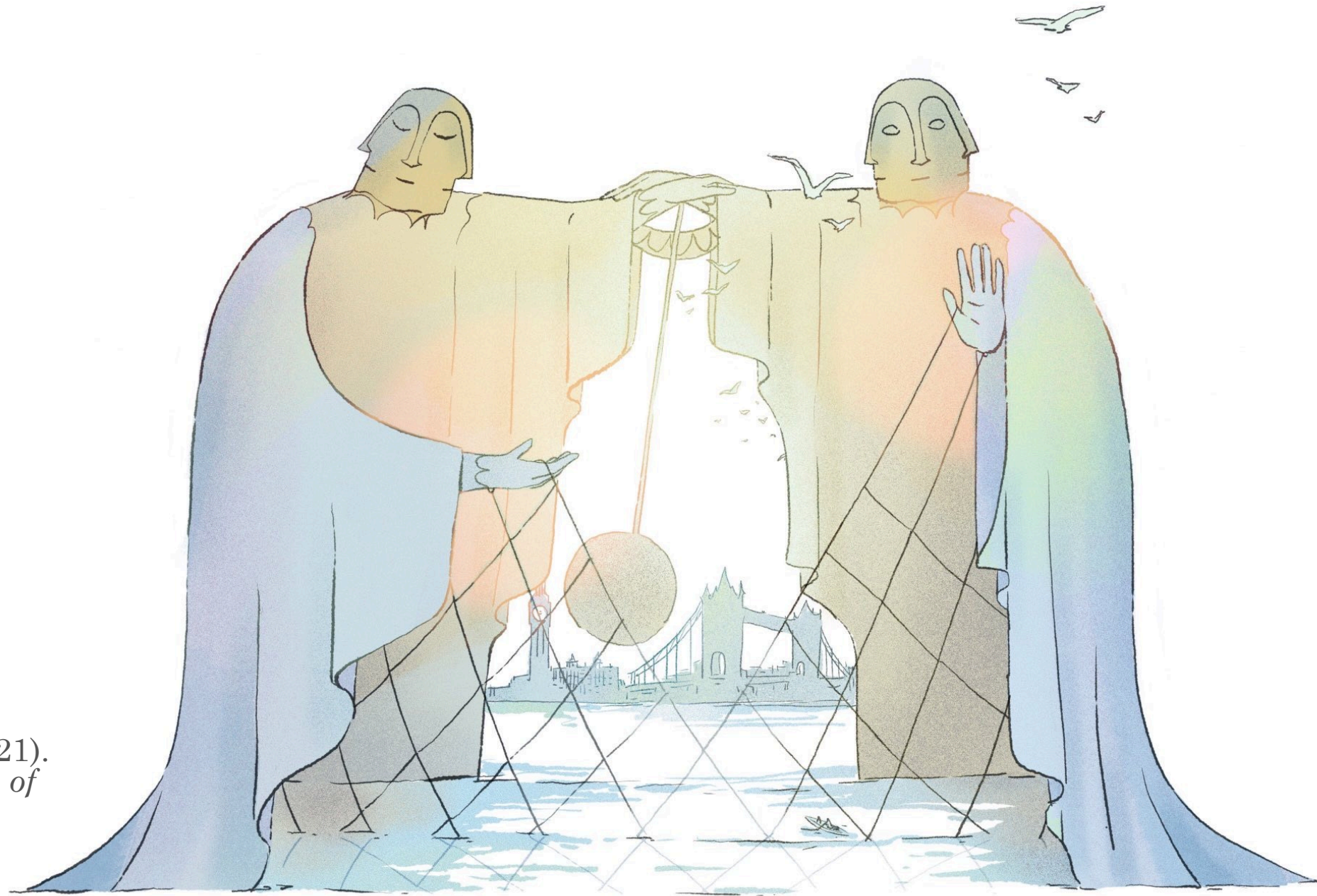
@noemidrexler

Epistemic Risk

&

Resilience

Cremer, C. Z., & Kemp, L. (2021).
*Democratising Risk: In Search of
a Methodology to Study
Existential Risk*



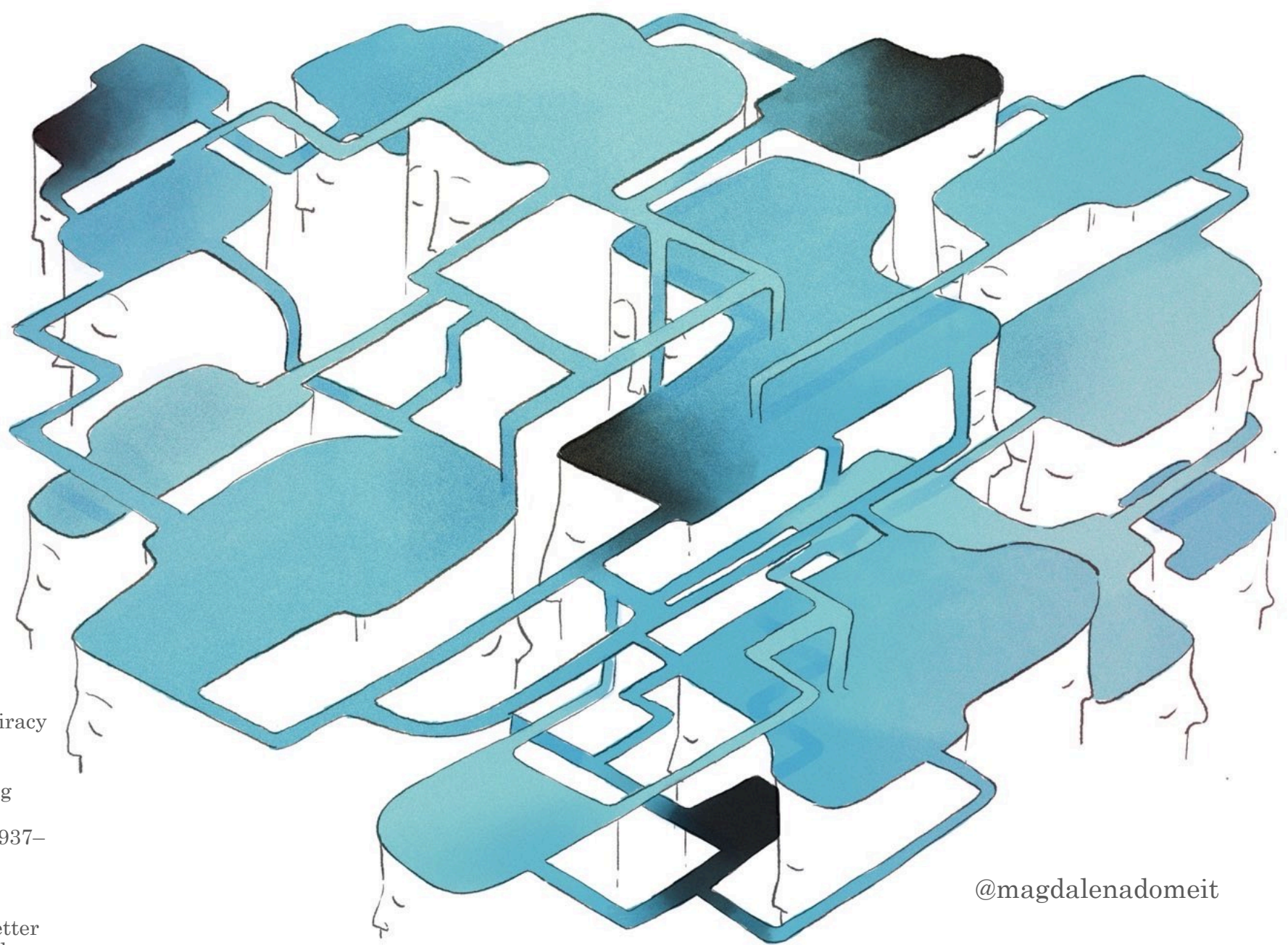
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- Architectural features > Content
- Content-neutral interventions
- Psychology Research



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Conspiracy & Democracy



Schaeffer et al. July 2022;
<https://doi.org/10.1073/pnas.2203149119>

Radnitz, S. (2022). Why Democracy Fuels Conspiracy Theories. *Journal of Democracy*, 33(2), 147–161.

Cusimano, C., & Lombrozo, T. (2021). Reconciling scientific and commonsense values to improve reasoning. *Trends in Cognitive Sciences*, 25(11), 937–949.

Pennycook, G., & Rand, D. G. (2019). Lazy, not biased: Susceptibility to partisan fake news is better explained by lack of reasoning than by motivated reasoning. *Cognition*, 188, 39–50.

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Hard choices

and who gets to decide

Trade-offs:

- eg. harmful content vs self-curated feeds

Arbiters of politics:

- [Scheck, Jeff Horwitz and Justin](#). 2021.
- *Querdenker* as test-bed

Productive polarization?

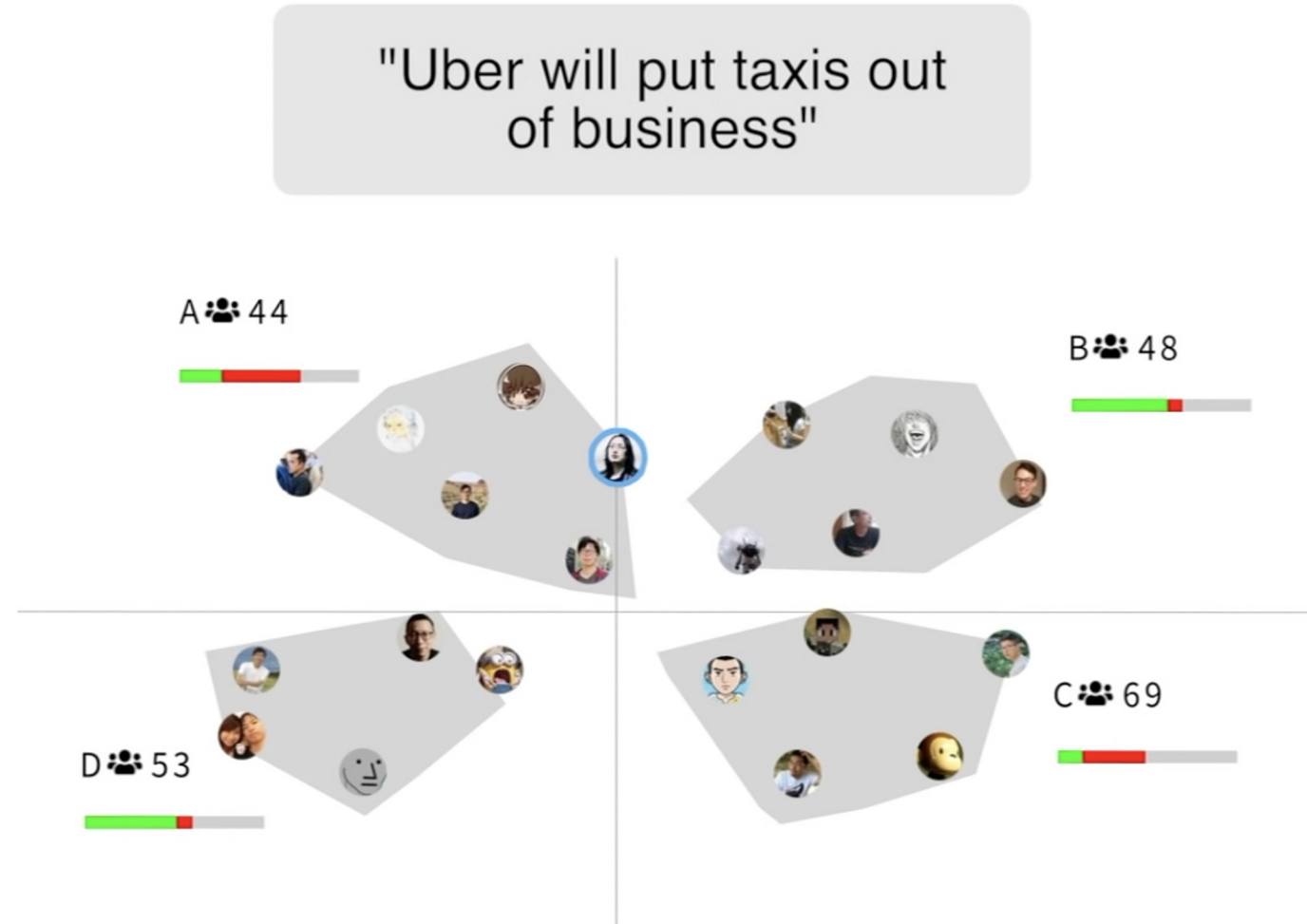
- Stray, 2022

“This could be a good case study to inform how we tackle these problems in the future”

“An individual can question election results. But when it’s amplified by a movement, it can damage democracy. There is harm in the way movements shift norms and an understanding of collective truth.”

Participation

is necessary
works
must be researched

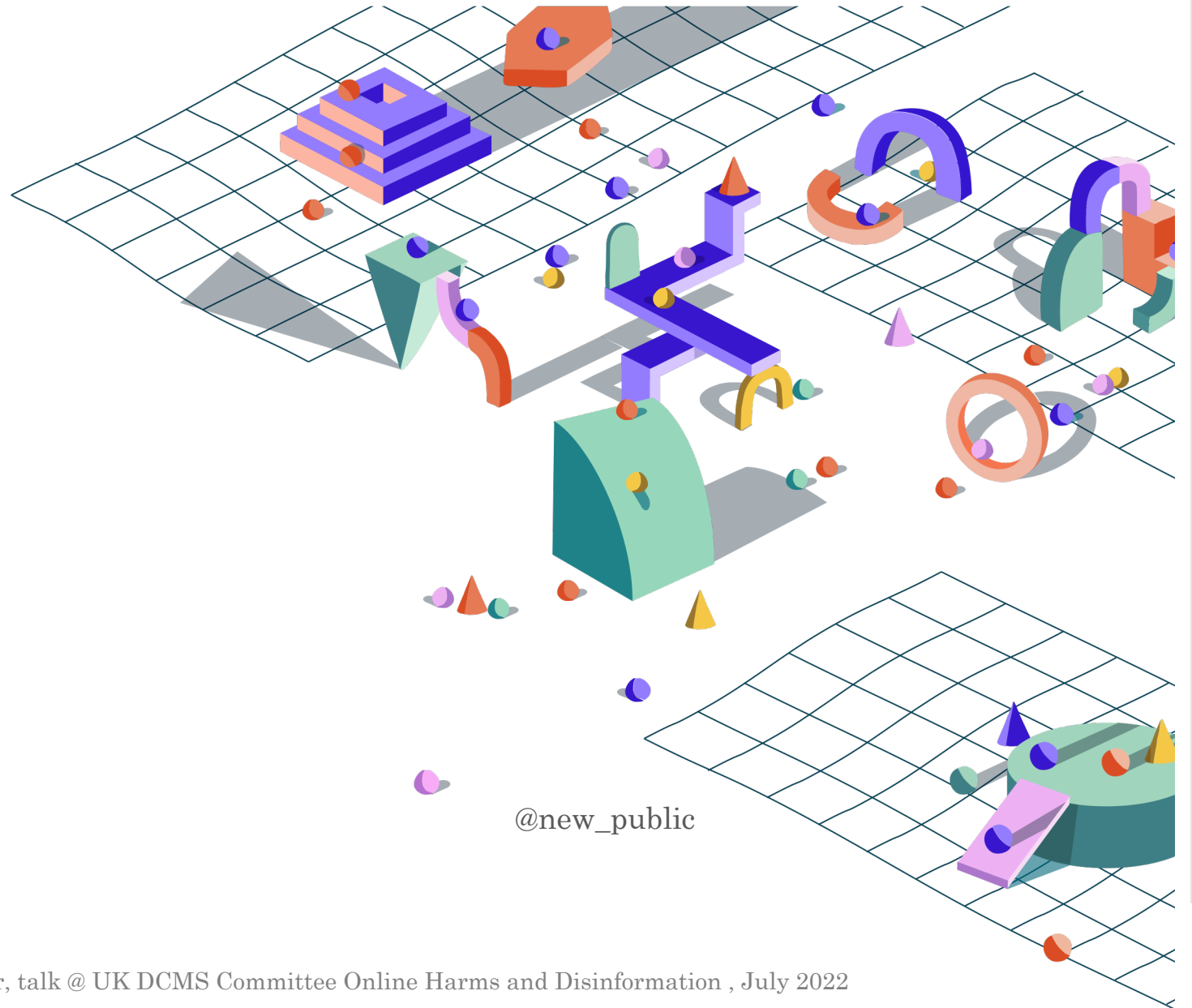


- Landemore (2017); vTaiwan, Landemore (2021), [OECD](#) (2020); [Lee](#) et al. (2019)

In dire need of psychological theory

- Features facilitating cognitive frames;
 - Silva, A., et al. (2022). *BIT Report*, 36.
 - *New Public Signals* [Research Overview](#)
- What is democratically important content?
- Polarisation tipping points?
 - Stray, J. (2022)
- Algorithmic consensus finding
 - [Koster, R et al.](#) (2022). Human-centred mechanism design with Democratic AI. *Nature Human Behaviour*, 1–10.
 - [Pol.is & vTaiwan](#); MIT Technology Review 2018
- Deliberation
- Benchmarks --- verification
- Markers --- correlates of ‘healthy’ conflict / discourse
- Metrics --- recommendation performance

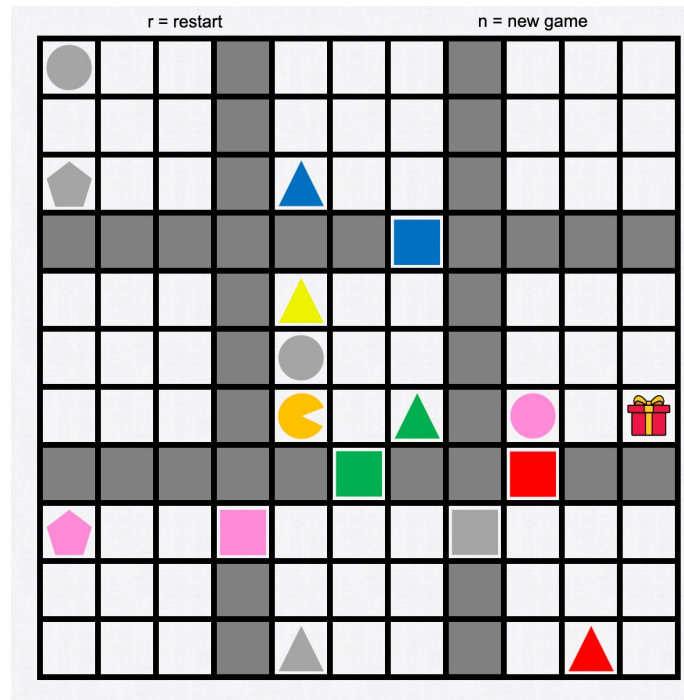
We need a
cognitive
theory of
online spaces



Human information processing lab



- truth-seeking games
- curriculum and information sharing
- validity must be tested in the wild



Dumbalska, 2022

Insight,
Security &
Control

- UK National Security Strategy: Protect Global Influence and Resilience (Cabinet, 978-0-10-179532-6)
- Convolutd methods:
 - NATO STRAT COM COE report (ISBN: 978-9934-619-16-8)
- Experiment (causal) > Observational Studies (non-causal)
 - UK could be first mover

Suggestions

Epistemic Risk Reduction

- Legal Requirement for research: effects of novel architectural features
- Compensations: users surveys and deliberations
- Participatory Architecture Assessments
 - OECD. [2020](#). *Innovative Citizen Participation and New Democratic Institutions* (March 2, 2021).
 - Cremer, C. Z., & Whittlestone, J. (2021). *Artificial Canaries*
 - Lee, M. et. al. ([2019](#)). *WeBuildAI: Participatory Framework for Algorithmic Governance*. Proceedings of the ACM on Human-Computer Interaction, 3(CSCW), 1–35.

Proactive technology assessments

Online Safety Bill

- Verification: assessing efforts relies on knowing what's possible
- Proactive technology: assessments
 - Expert identification
 - Limitation assessment -- > identify misapplication



Cremer, C. Z. (2021). Deep Limitations? Examining Expert Disagreement over Deep Learning. *Progress in Artificial Intelligence, Springer*

Cremer, C. Z., & Whittlestone, J. (2021). Artificial Canaries: Early Warning Signs for Anticipatory and Democratic Governance of AI

TABLE I. LIMITATIONS OF DEEP LEARNING AS PERCEIVED AND NAMED BY EXPERTS FOUND IN [11]

Causal reasoning: the ability to detect and generalise from causal relations in data.	Common sense: having a set of background beliefs or assumptions which are useful across domains and tasks.
Meta-learning: the ability to learn how to best learn in each domain.	Architecture search: the ability to automatically choose the best architecture of a neural network for a task.
Hierarchical decomposition: the ability to decompose tasks and objects into smaller and hierarchical sub-components.	Cross-domain generalization: the ability to apply learning from one task or domain to another.
Representation: the ability to learn abstract representations of the environment for efficient learning and generalisation.	Variable binding: the ability to attach symbols to learned representations, enabling generalisation and re-use.
Disentanglement: the ability to understand the components and composition of observations, and recombine and recognise them in different contexts.	Analogical reasoning: the ability to detect abstract similarity across domains, enabling learning and generalisation.
Concept formation: the ability to formulate, manipulate and comprehend abstract concepts.	Object permanence: the ability to represent objects as consistently existing even when out of sight.
Grammar: the ability to construct and decompose sentences according to correct grammatical rules.	Reading comprehension: the ability to detect narratives, semantic context, themes and relations between characters in long texts or stories.
Mathematical reasoning: the ability to develop, identify and search mathematical proofs and follow logical deduction in reasoning.	Visual question answering: the ability to answer open-ended questions about the content and interpretation of an image.
Uncertainty estimation: the ability to represent and consider different types of uncertainty.	Positing unobservables: the ability to account for unobservable phenomena, particularly in representing and navigating environments.
Reinterpretation: the ability to partially re-categorise, re-assign or reinterpret data in light of new information without retraining from scratch.	Theorising and hypothesising: the ability to propose theories and testable hypotheses, understand the difference between theory and reality, and the impact of data on theories.

- Cremer, C. Z. (2021). Deep Limitations? Examining Expert Disagreement over Deep Learning. *Progress in Artificial Intelligence*, Springer.

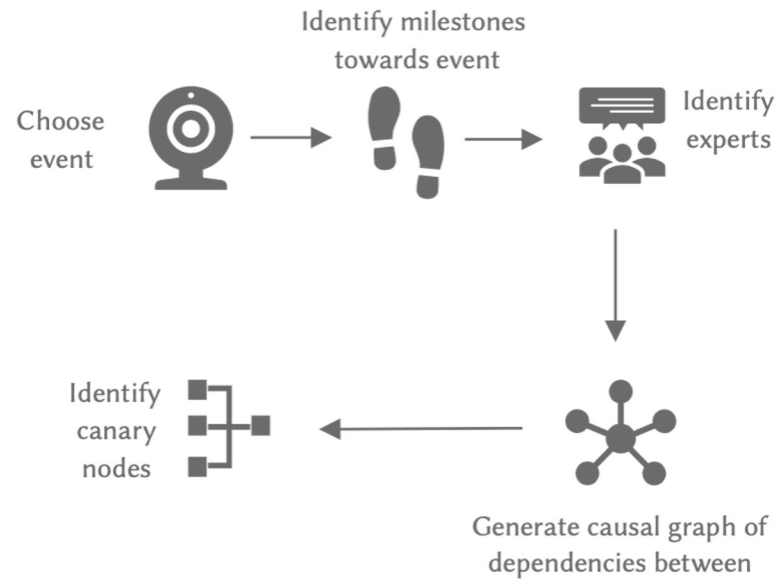
Preemptive technology assessments

Expert forecasting

(could be enforced by regulator)

Assessing technological advance:

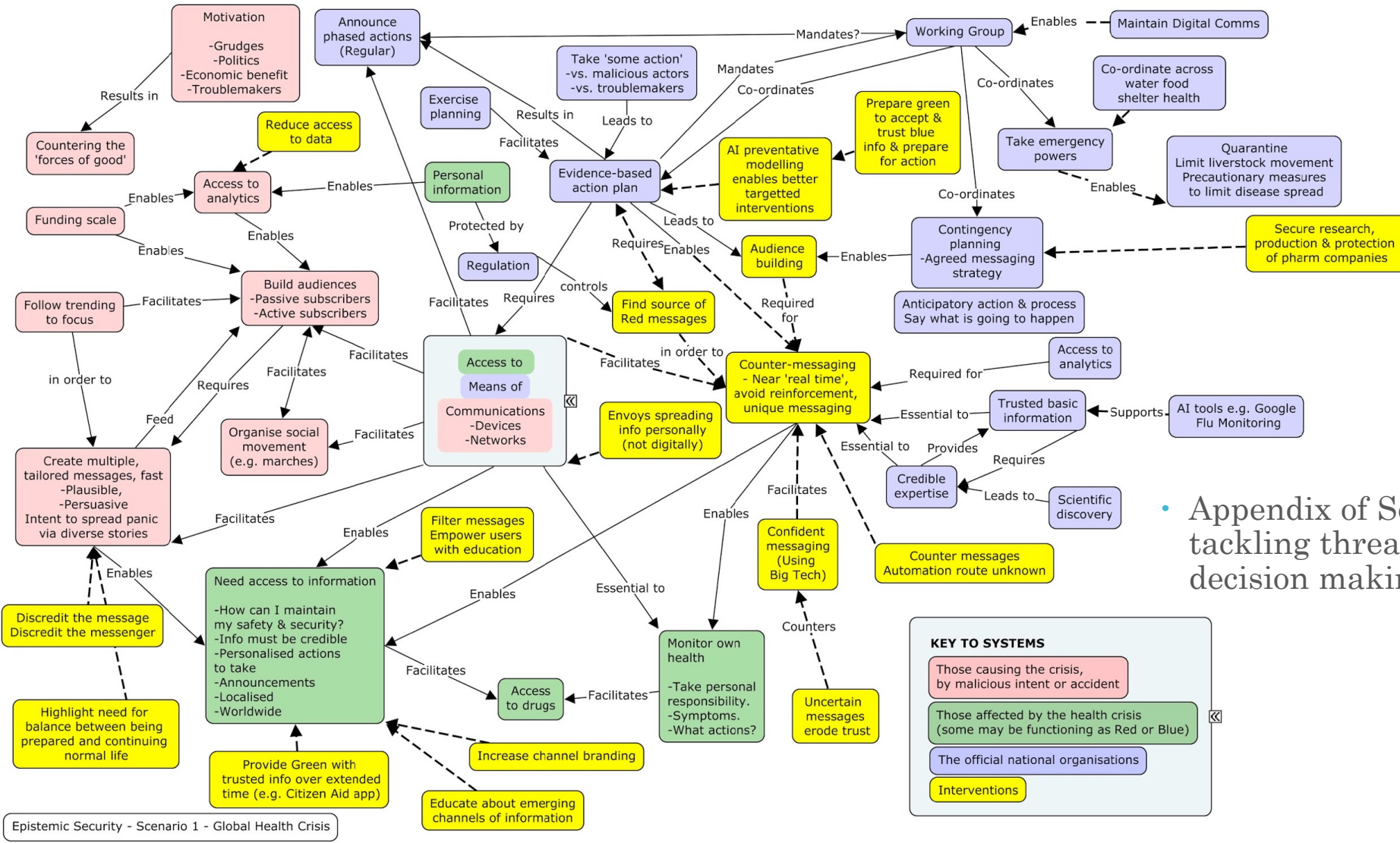
- Cremer, C. Z., & Whittlestone, J. (2021).



Assessing impact of technological advance:

- red teaming, pre-mortems, second-order effects
- Seger, E., Avin, S., Pearson, G., Briers, M., O Heigeartaigh, S., & Bacon, H. (2020).

Global Health Crisis scenario systems map



• Appendix of Seger et al 2020.: tackling threats to informed decision making

Suggestions

Online Safety Bill

- Reporting: require reports on significant architectural changes
- Research: require research collaborations on architectural choices
- Research: require *experimental* access
- Proactive technology assessments

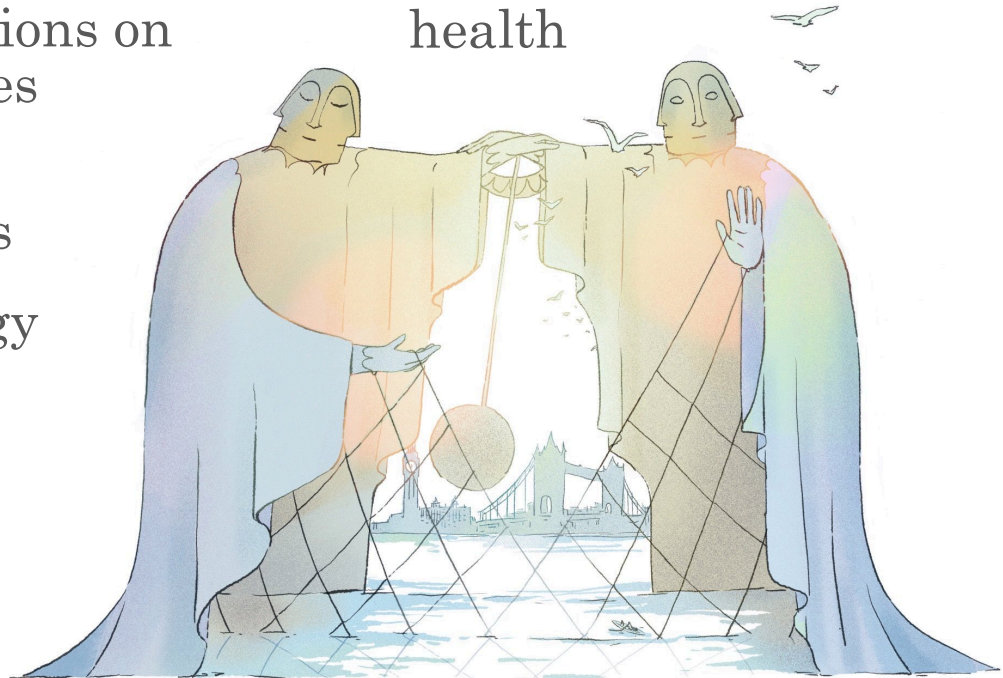
Suggestions

Epistemic Risk Reduction

- Foster data-altruism
- Foster and train open-source journalism / task force
 - OSINT / [bellincat](#)
 - sudan media capacity building project
 - [Wikimedians](#), Wiki Education Foundation
 - reward mechanisms

Summary

- Reporting: require reports on significant architectural changes
- Research: require research collaborations on architectural choices
- Research: require *experimental* access
- Proactive technology assessments
- Participatory Architecture Assessments
- Cognition research: markers of epistemic health



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Papers:

<https://carlacremer.github.io/research/>

Illustrations:

<https://magdalenaadomeit.com/>

<https://www.governance.ai/team/noemi-dreksler>



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