

Responsible data management

A precursor to responsible AI

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Responsible AI is a broad term that is intended to encapsulate a number of guardrails to protect against potential risks and harms that may stem from the inappropriate use of powerful emerging AI technologies.

These risks include but are not limited to those related to consent, privacy, accountability, fairness and accessibility. Responsible AI design, development and deployment necessarily require qualified experts, trusted data and sustainable infrastructure. Without widespread use and adoption of responsibly developed AI, there is a risk of irresponsible AI technologies infiltrating business and society.

RECENT ADVANCES IN AI are bringing about both great excitement and significant concerns. It is widely recognised that the recent upswing of interest and success in AI technology is riding on the back of two significant advancements that previous booms did not have: access to cheap and elastic computational resources, including the cloud, high-speed networks and fast processors; and the availability of large volumes and a variety of digital information, or Big Data. Given the significant injection of both corporate and venture capital, the opportunities arising from the current AI boom are both promising and sustainable.

While the availability of large amounts of data is a critical factor behind the recent AI successes, it is

also a cause for concern. A recent article on responsible data management²¹ presents a mirror metaphor to explain pre-existing bias and the technical limitations of responsible data use:

‘Data is a mirror reflection of the world. When we think about pre-existing bias in the data, we interrogate this reflection, which is often distorted. One possible reason is that the mirror (the measurement process) introduces distortions. It faithfully represents some portions of the world, while amplifying or diminishing others. Another possibility is that even a perfect mirror can only reflect a distorted world – a world such as it is, and not as it could or should be.’

Current AI systems are trained using very large amounts of data

largely extracted from the web, the vast majority of which is typically publicly available (e.g., from Wikipedia). This raises several questions about quality, consent, ownership and privacy. Data is often collected and used without the awareness and explicit consent²² of the people who created this online content. One of the most relevant pieces of legislation relating to this issue is copyright, which was developed to restrict how people re-use others’ content but does not regulate the use of content to train AI.

The European Union’s General Data Protection Regulation (GDPR) is an example of a regulation designed to tackle this issue of data copyright, particularly in terms of the data that users contribute to AI systems on the open web, such as through prompts to large language models like ChatGPT, which then eventually become part of the model. GDPR includes the right to be forgotten, which allows citizens to request that their data be erased from data products, including AI-assisted products. Given the current lack of regulation in most jurisdictions of the world on the use of public data to train AI systems, its usage may be regarded as legally compliant, but it may also be considered ethically questionable as it poses risks.

It is currently hard to guarantee that a generative AI model trained on certain data will not resurface pieces of content from the large amounts used to train it, thus potentially putting the privacy and safety of the content creators at risk. Note that the use of publicly available content without authors’ consent is not a new issue.²³ It is important that the data used to train AI is either collected

through informed consent processes and/or used in a way that provides a proportionate return for the content creators without putting them at risk.

Modern AI applications like chatbots, recommender systems and traffic-aware navigation commonly take user-generated real-time data as inputs and provide predictions or suggestions. To reduce the reliance on real-time user data, an emerging line of research is around the use of AI to generate synthetic data points^{24,25} based on patterns extracted from real data. The application of synthetic data is already seen in several high-stake domains, such as medical analysis²⁶ and financial fraud detection.²⁷

In the meantime, the computational cost of storing and analysing large-scale data is escalating. For example, the state-of-the-art computer vision algorithms for object classification are all trained with the ImageNet dataset, whose full version contains over 14 million images and is 1.3 terabytes in size.²⁸ Recent foundation models²⁹ and in particular large language models like GPT-4 have further demonstrated the tremendous cost of data collection, curation and storage as well as the computational and environmental cost of model training.

Recent advancements that provide distribution-level summarisation of real data points^{30,31} with compression/condensation techniques offer promising solutions from both the data management and computational perspectives, wherein large and noisy data points can be substituted with a set of small but quality samples (or aggregated

samples), to provide a data-efficient paradigm for training AI algorithms.

Advancements in model distillation (also known as knowledge distillation³²) offer another promising avenue to help deal with computational concerns. Model distillation involves training a smaller model (‘student’) to reproduce the behaviour of a larger model (‘teacher’), with the aim of creating a lighter, more efficient model to be executed in production, while maintaining a high level of effectiveness. This process is particularly useful for large language models that can have billions of parameters, making them computationally expensive to run. Distilling generic models, made available by third parties such as OpenAI’s GPT-4 or Meta’s Llama, into smaller ones makes them more viable for real-world applications where substantial computing power is lacking. Another advantage of model distillation is the student model’s ability to assimilate a rich array of linguistic patterns from the teacher model, which may be trained on diverse, large and open-source data. The student model can then undergo further refinement via fine-tuning specific proprietary data, facilitating a degree of model sovereignty.

Data management will continue to underpin the development of AI technologies, and the responsibility with which we manage the data will determine whether we gain the benefits of these powerful technologies or are confronted by their risks and potential harms.



PROFESSOR SHAZIA SADIQ FTSE is a Professor of Computer Science at the School of Information Technology and Electrical Engineering, The University of Queensland. Her research focusses on responsible data management and aims to reduce the socio-technical barriers to data driven transformation, by assisting organisations to create, protect and sustain agile data pipelines. She is the Centre Director for the ARC Industry Transformation Training Centre on Information Resilience 2020-2025, Chair of the National Committee on Information and Communication Sciences at the Australian Academy of Science 2019-2022, and member of The Australian Research Council College of Experts 2018-2021



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DR ROCKY TONG CHEN received his PhD degree in computer science from The University of Queensland in 2020, and is currently a lecturer with the data science discipline, School of Electrical Engineering and Computer Science, The University of Queensland. He received the Discovery Early Career Researcher Award from the Australian Research Council in 2022. His research interests include data mining, recommender systems, user behavior modelling and predictive analytics.



PROFESSOR GUIDO ZUCCON is a Professorial Research Fellow at The University of Queensland, Electrical Engineering and Computer Science School, the AI Director for the Queensland Digital Health Centre (QDHeC), an Affiliate Professor at the UQ Centre for Health Services Research, Faculty of Medicine, and an Honorary Reader at Strathclyde University (UK). He leads the Information Engineering Lab (ielab), a research team working in information retrieval and health data science. He was an ARC DECRA Fellow (2018-2020). Guido’s main research interests are information retrieval, health search, formal models of search and search interaction, and health data science.

Essays

SECTION 1: INTRODUCTION

What is responsible AI anyway?

Professor Jon Whittle – Director, CSIRO's Data61

10 examples of AI that are here now and have been embraced by the general public

Stela Solar – Director, National Artificial Intelligence Centre

SECTION 2: WHAT DO WE NEED TO BE TALKING ABOUT?

A unique opportunity for Australia: bridging the divide between fundamental AI research and usable, embodied AI

Professor Michael Milford FTSE – ARC Laureate Fellow, Joint Director QUT Centre for Robotics

Responsible AI means keeping humans in the loop: what are other social implications of the mainstream adoption of this technology?

Associate Professor Carolyn Semmler School of Psychology, Faculty of Health and Medical Sciences, The University of Adelaide and Lana Tikhomirov – Australian Institute for Machine Learning (AIML), The University of Adelaide

AI is changing the way people work: how do we skill our future workforce to ensure these new jobs stay on shore?

Professor Katrina Falkner FTSE – Executive Dean of the Faculty of Sciences, Engineering and Technology, The University of Adelaide

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Open the pod bay doors please, HAL

Andrew Dettmer – National President, Australian Manufacturing Workers Union

Innovation needs to create value: how do we tool universities to remain relevant to industry needs?

Professor Simon Lucey – Director, Australian Institute for Machine Learning, The University of Adelaide

An AI-literate community will be essential for the continuity of social democracy

Kylie Walker – Chief Executive Officer, Australian Academy of Technological Sciences and Engineering

SECTION 3: WHAT ARE THE NEXT STEPS?

What are the limits of current AI, and what opportunities does this create for Australian research?

Professor Anton van den Hengel FTSE – Director, Centre for Augmented Reasoning, Australian Institute for Machine Learning, The University of Adelaide

Australia's unfair advantage in the new global wave of AI innovation

Professor Mary-Anne Williams FTSE – Michael J Crouch, Chair for Innovation, UNSW Business School

The \$1 billion dollar question: What should Australia's responsible AI future look like?

Kingston AI Group

What are we doing now to ensure that Australia is recognised as a global leader in responsible AI, and what else should we be doing now and into the future?

Dr Ian Opperman FTSE – NSW Government's Chief Data Scientist, Department of Customer Service

For acronyms, abbreviations and endnotes please see the composite document with all the essays.



Responsible AI

Your questions answered

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Cover image: An artist's illustration of artificial intelligence (AI). This image represents the boundaries set in place to secure safe, accountable biotechnology. It was created by artist Khyati Trehan as part of the Visualising AI project launched by Google DeepMind. Source: unsplash

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Your questions answered

