

Mitigating Bias in AI: Introduction Webinar

by *Holistic AI* and hosted by *The Alan Turing Institute*

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Meet your presenters



Fiona Thendean

AI Scientist (AI Governance Solutions)

Fiona is an AI scientist on the AI Governance Solutions team at Holistic AI, specialising in developing quantitative approaches to AI governance. She earned her MSc in Artificial Intelligence for Sustainable Development at UCL, where she focused her dissertation on Explainability for Deep Reinforcement Learning in the context of climate change policy models. Before moving to London, she studied Statistics and Machine Learning at Carnegie Mellon University.



Sachin Beepath

Head of AI Governance Solutions

Sachin is the Head of AI Governance Solutions at Holistic AI, specialising in responsible AI adoption and scalability. Originally from Toronto, he studied astrophysics at the University of Waterloo before earning an MSc in Data Science and Machine Learning at UCL, where he did his dissertation on Bias in Facial Recognition. He works closely with industry leaders, business executives, and technical teams to drive fair, transparent, and impactful AI innovation.



Holistic AI is an AI Governance company that aims to empower enterprises to adopt and scale AI with confidence.

Hi, I'm John Doe. I was born on April 12, 1980. I live at 123 Fake Street, Springfield, OH, 45503. My phone number is 555-555-5555 and my email address is johndoe@example.com. My social security number is 123-45-6789.

Hallucination based in Dialog History

Question Answering Generation Hallucination

Summarization Data Hallucination

General Data Generation Hallucination

Safeguard

Governance and monitoring of Large Language Models.



Holistic AI Platform

Tracker

All-in-one repository of all things related to AI regulations and harms.

AI Governance Platform

Enterprise-ready application for safe AI adoption and usage.

Bespoke AI Auditing

Third-party quantitative assessment and reporting for specific AI risks and use cases.

Agenda

Setting the Stage for AI Governance

Risks Associated with AI

Bias and Fairness in AI

Governing AI Bias

What is Artificial Intelligence?

Regulatory Definitions

UK ICO definition:

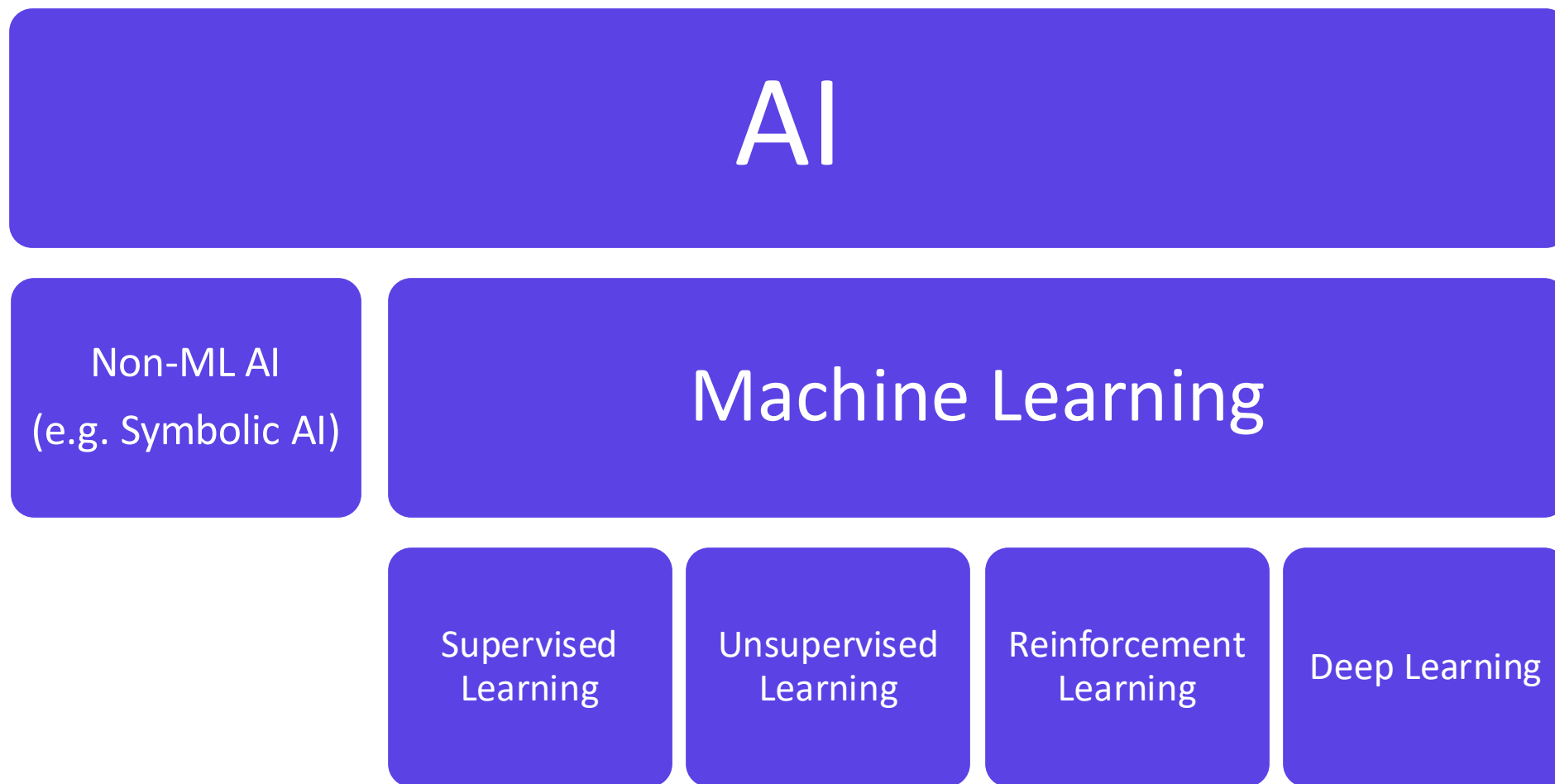
“AI is an umbrella term for a range of technologies and approaches that often attempt to mimic human thought to solve complex tasks”

Definition of AI systems

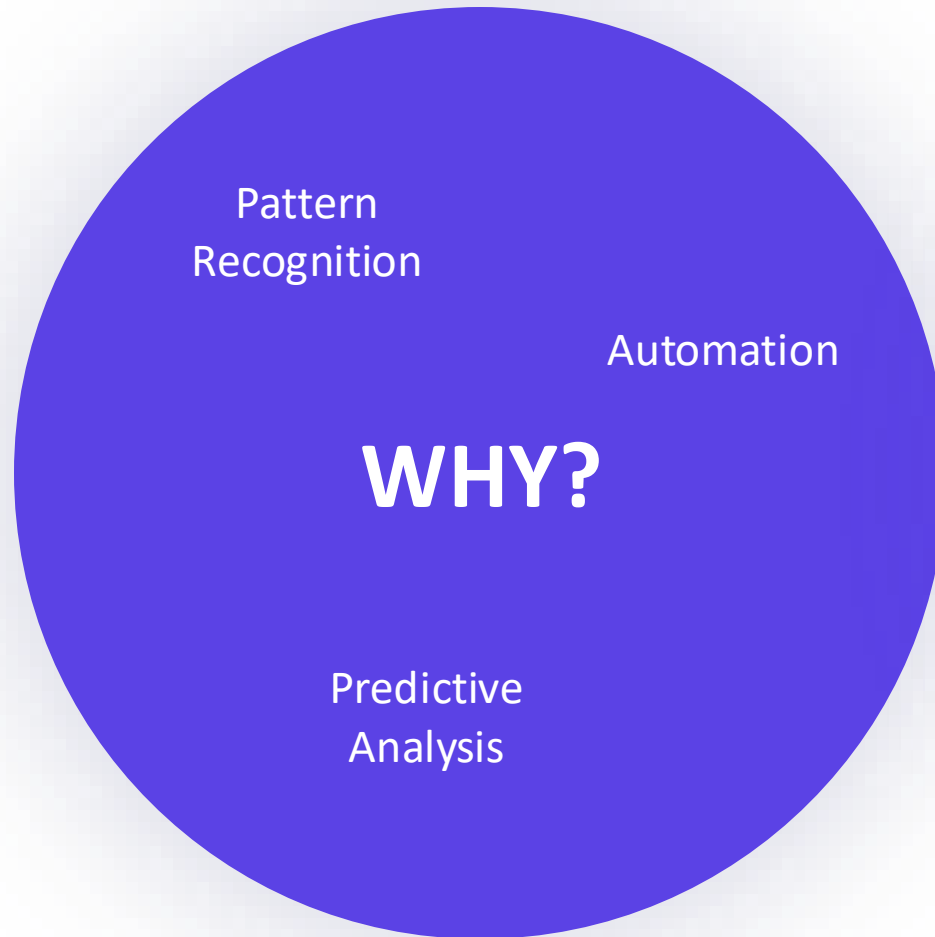
EU AI Act: “a machine-based system that is designed to operate with varying levels of autonomy and that may exhibit adaptiveness after deployment, and that, for explicit or implicit objectives, infers, from the input it receives, how to generate outputs such as predictions, content, recommendations, or decisions that can influence physical or virtual environments”

ISO/IEC 22989: “engineered system that generates outputs such as content, forecasts, recommendations or decisions for a given set of human-defined objectives”

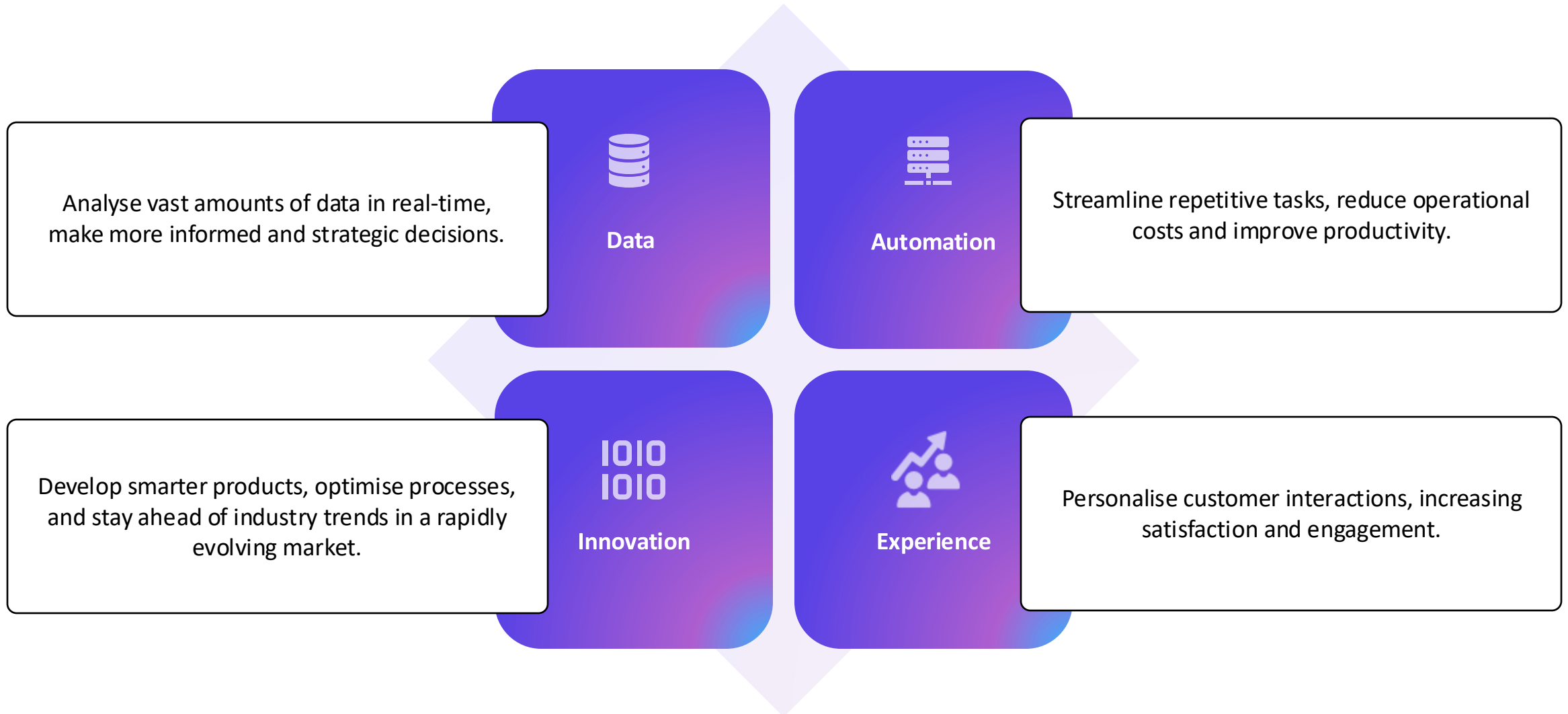
<https://www.holisticai.com/blog/comparing-definitions-of-ai>



Why and where is AI used?



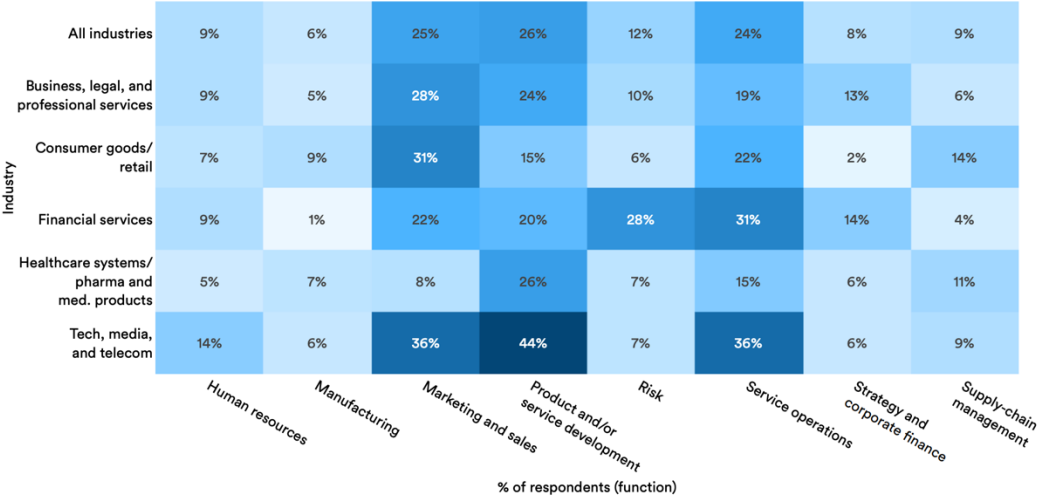
Why are businesses adopting AI?



What is happening?

AI adoption by industry and function, 2023

Source: McKinsey & Company Survey, 2023 | Chart: 2024 AI Index report



Emerging Regulations and Frameworks



UK AI Regulation
White Paper



EU AI Act



NIST AI RMF



ISO 42001

The Future

UK AI Market

is predicted to grow to
over \$1 trillion by 2035

Global GDP

is estimated to grow by
\$15.7 trillion by 2030

Explosion in AI vendors



ANTHROPIC






Risks Associated with AI

General set of AI risks

Security and Privacy	Data minimization principles as well as adopt privacy-enhancing techniques to mitigate personal or critical data leakage.
Bias and Discrimination	Avoid unfair treatment of individuals given their protected characteristics.
Interpretability and Explainability	Provide decisions or suggestions that can be understood by their users and developers.
Performance and Robustness	Be safe and secure, not vulnerable to tampering or compromising of the data they are trained on.

To avoid these cases

In the news


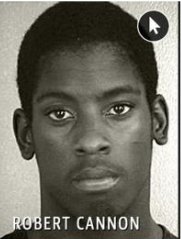


Microsoft deletes 'teen girl' AI after it became a Hitler-loving sex robot within 24 hours
Telegraph.co.uk - 5 hours ago

To chat with Tay, you can tweet or DM her by finding @tayandyou on Twitter, or add her as a ...

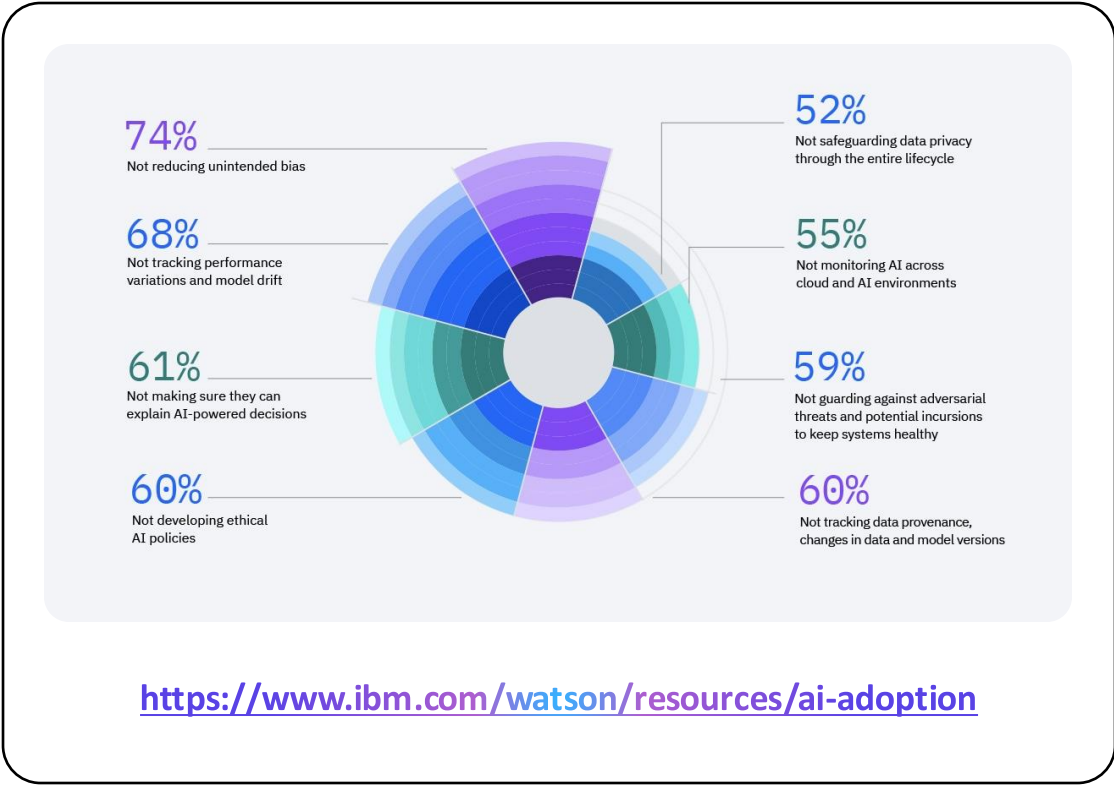
Microsoft Releases AI Twitter Bot That Immediately Learns How To Be Racist
Kotaku - 3 hours ago

Microsoft Created a Twitter Bot to Learn From Users. It Quickly Became a Racist Jerk.
New York Times - 3 hours ago

 <p>JAMES RIVELLI</p> <p>LOW RISK 3</p>	 <p>ROBERT CANNON</p> <p>MEDIUM RISK 6</p>
<p>JAMES RIVELLI</p> <p>Prior Offenses</p> <p>1 domestic violence aggravated assault, 1 grand theft, 1 petty theft, 1 drug trafficking</p> <p>Subsequent Offenses</p> <p>1 grand theft</p> <p>LOW RISK 3</p>	<p>ROBERT CANNON</p> <p>Prior Offense</p> <p>1 petty theft</p> <p>Subsequent Offenses</p> <p>None</p> <p>MEDIUM RISK 6</p>

Why Bias is the Key Risk?

Reputational damage	Instances of biased AI garner significant public attention and can lead to public backlash against the technology, as well as increased scrutiny and regulation.
Challenging risk to deal with	Compared to other key risks, Bias is undoubtedly the area where data scientists and other professionals are least prepared to deal
Regulatory requirements	Most laws and regulations are targeted towards bias, for example anti-discrimination laws.
General awareness	Anecdotally, Privacy is for Data what Bias is for AI



Societal Implications of Bias in AI



Historical Discrimination

Society's historical bias reflected in the data that we use to train algorithmic systems. will be perpetuated by the decision-making system.

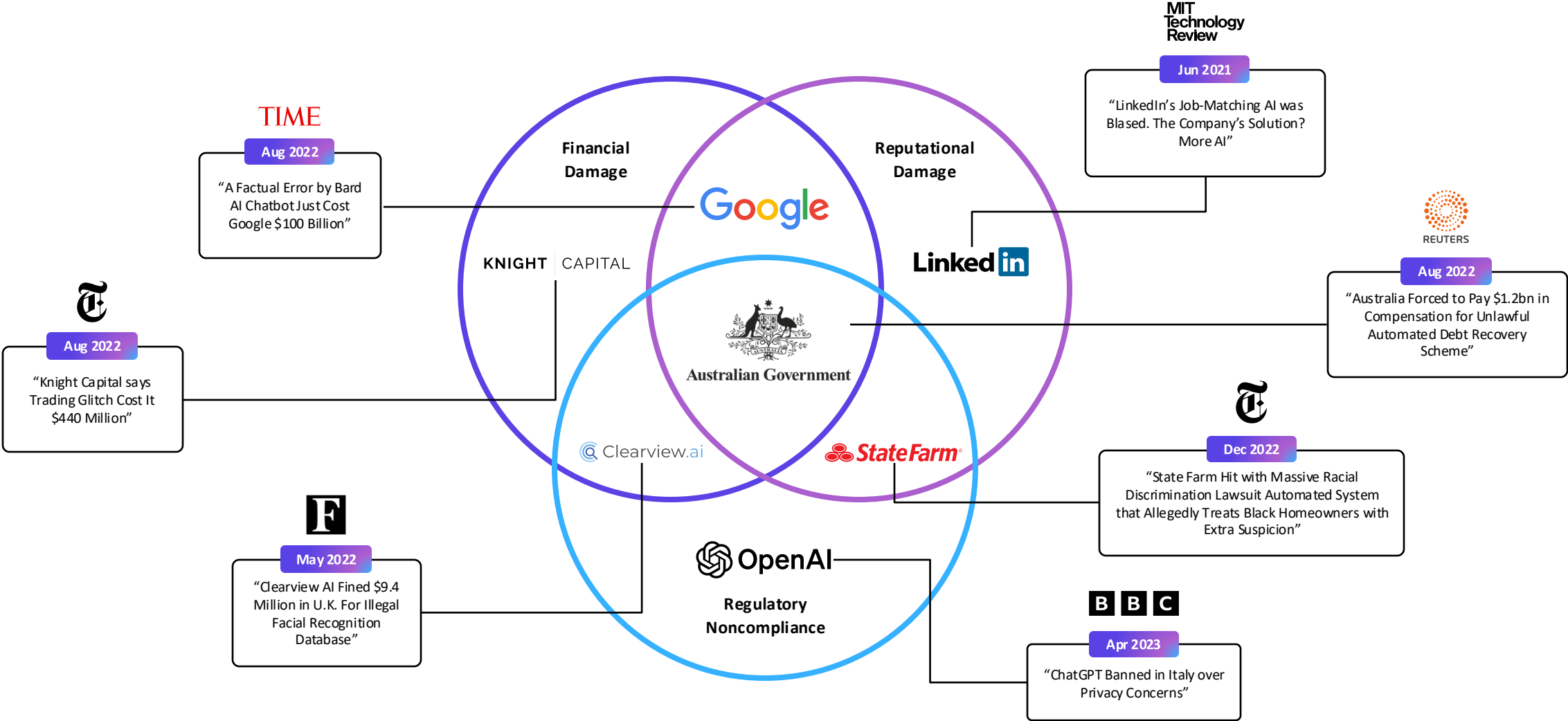
Impact on Life Chances

Algorithmic systems are being used for decision making in a wide range of areas such as criminal justice, healthcare, education, employment, consumer lending, etc.

Responsibility

When developing decision-making systems, we are responsible for minimizing harm.

Business Implications of Bias in AI



Risks and issues with HR technologies

With the pervasive usage, side-effects such as bias, transparency, cheating will increase



Holistic AI

Tutoring firm settles US agency's first bias lawsuit involving AI software



EEOC Settles First AI-Bias Case, but Many More Are Likely



Amazon scrapped 'sexiest AI' tool



Lawsuit Claims Workday's AI And Screening Tools Discriminate Against Those Black, Disabled Or Over Age 40



HireVue drops facial monitoring amid A.I. algorithm audit



Candidates using ChatGPT to cheat their way into a job





Real-World Examples of Bias in AI

ML can be biased

Amazon Recruitment

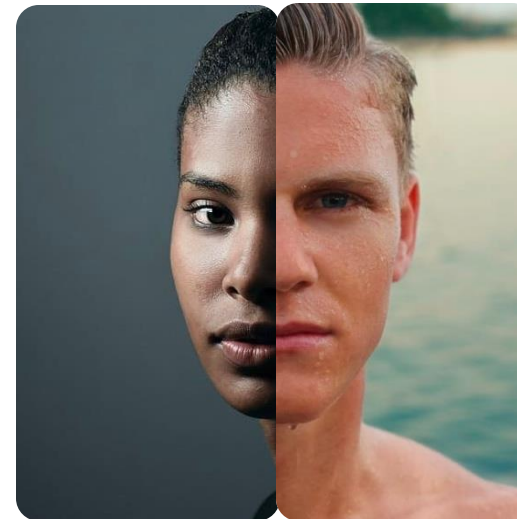
RESUME

2023: AWESOME

2020: WOMEN'S COLLEGE

Gender Recognition ("Gender Shades")

Up to ~35%
error rate



Up to ~1%
error rate

The COMPAS Example

- Used in US court to predict risk of recidivism
- ProPublica study in 2016:

	White	Black
Labelled higher risk – but didn't re-offend	23%	45%
Labelled low risk – did re-offend	48%	28%

- Northpointe defence: accuracy for white and black is the same (~60 %)

GenAI

- UNESCO study reveal that popular LLMs like GPT3.5 and Llama 2 exhibit bias against women in its outputs.
 - Richer narratives in stories about men
 - Homophobic attitudes and racial stereotyping



Bias and Fairness in AI

What is Bias?

Global definition

Bias is a tendency to **favour one group or idea over others**. It can be based on a person's race, gender, religion, etc.

Machine Learning

Systematic errors in the decision-making process of an algorithm that lead to inaccurate or unfair outcomes.

What is Fairness?

Individual Fairness

Individual Fairness: Seeks for similar individuals to be treated similarly

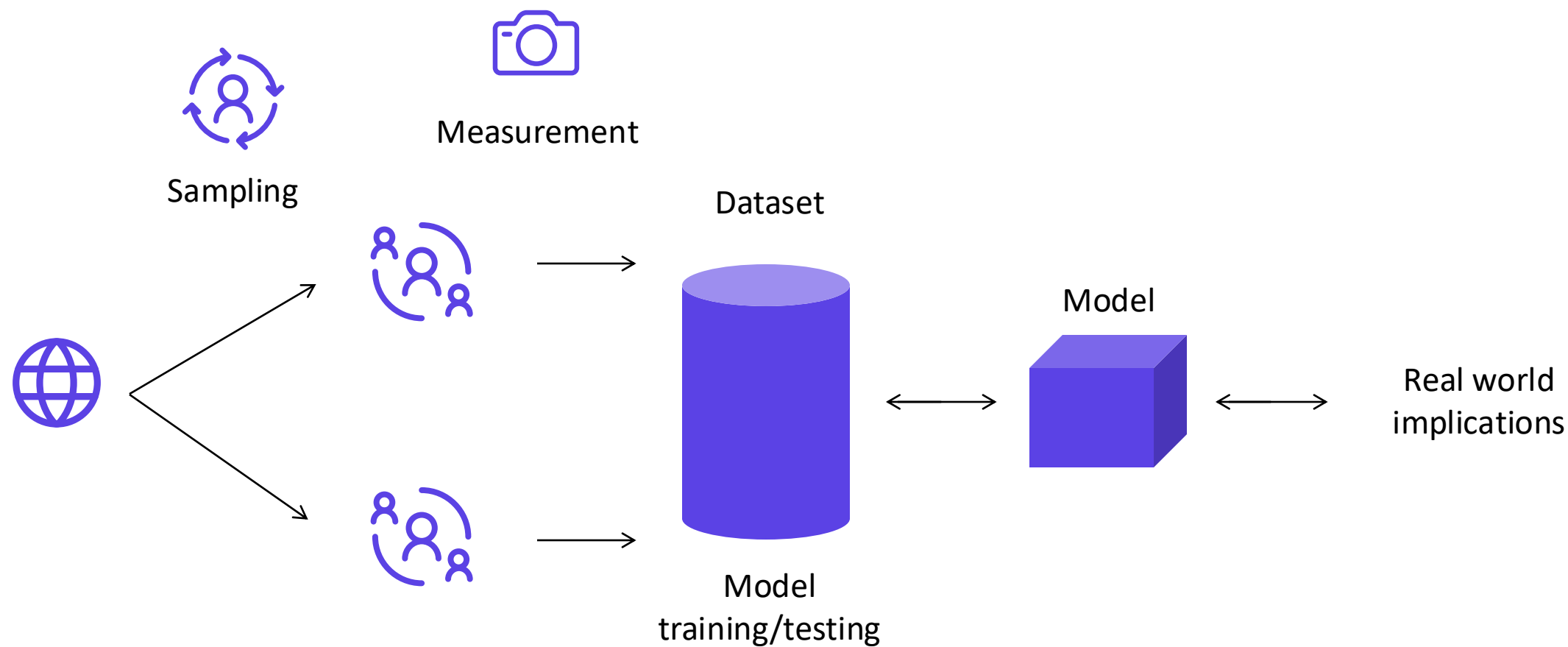
- “Am I as an individual being treated as another with similar qualifications?”
 - Fairness through awareness vs. Fairness through unawareness
 - Counterfactual fairness

Group Fairness

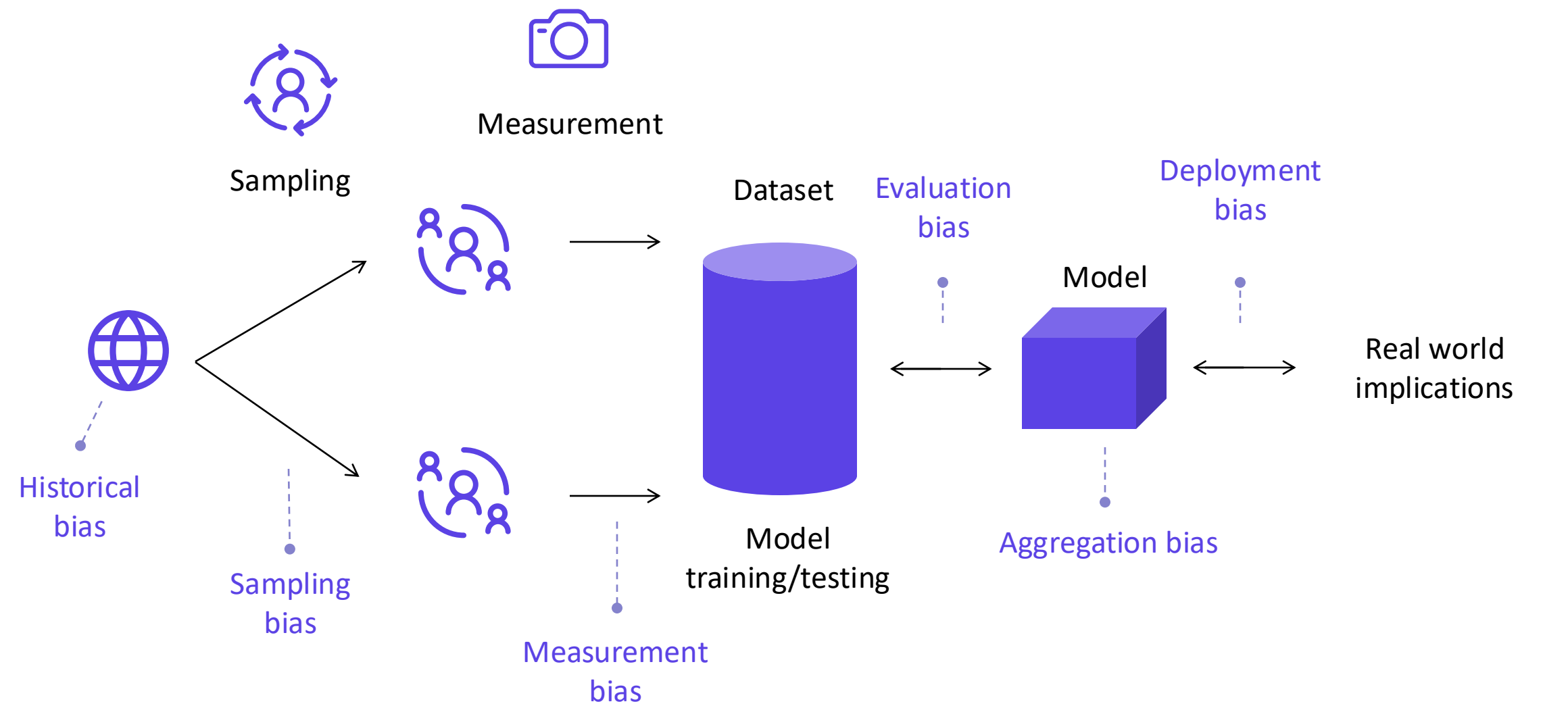
Group Fairness: Split a population into groups defined by protected attributes (e.g. male, female) and seeks for some measure to be as equal as possible across groups

- “Is the minority group to which I belong being treated as the majority group?”
 - Equality of Outcome: Distribution of success across all groups should be the same
 - Equality of Opportunity: Distribution of success across groups with similar qualifications should be the same

Machine Learning lifecycle



Sources of Bias



Historical Bias

- Pre-existing bias reflected in the data
- Example: In 2018, 5% of Fortune 500 CEOs were women.
- What should be the results of an image search for 'CEO'? Google recently changed image search result to include higher proportion of women



n hires former QBE chief John Neal ...
.com



CEO vs. Owner: The Key Differences ...
onlinemasters.ohio.edu



You are the CEO of Your Life - Per
personalexcellence.co



EO doesn't believe in CX ...
artofthecustomer.com



Wartime CEOs are not the ideal leaders ...
ft.com



Understanding CEO Leadershi
online.norwich.edu



Burkhard Eling takes up role of CEO at ...
dachser.com



LinkedIn CEO Jeff Weiner steps down ...
fortune.com



Best CEOs 2019 List ...
elcompanies.com

Deployment bias



- Model was trained and tested for use case X, but is used for application Y
- Example: Model was trained to identify good candidates for investment banking jobs, and now being used to identify junior level lawyers

Governing AI Bias

AI Governance

- Refers to the rules, policies, and frameworks that guide the development and use of AI, ensuring it aligns with ethical, legal, and societal standards.
 - Risk Management
 - Ensuring Trust and Accountability

Managing Risks

Allows for the measuring, mitigating, and monitoring, of risks that are most relevant to the AI use case.

Why is AI Governance Critical?

Trust and Accountability

Ensures trustworthy and responsible adoption, use, and scaling of AI within organisations.

Global AI Regulations and Governance Frameworks



EU AI Act

Regulates AI based on risk, with strict requirements for high-risk AI systems, ensuring safety, transparency, and accountability. Designed to enhance trust in AI technologies and promote innovation while safeguarding fundamental rights.

ISO/IEC 42001

A standard for AI Management Systems that provides guidelines on managing AI systems' lifecycle, focusing on transparency, accountability, and ethical use. Promotes risk management and ensuring AI systems operate safely and in alignment with societal values.

NIST AI RMF

A voluntary guideline aimed at fostering trustworthiness in AI systems. Provides a framework for organizations to manage risks related to AI, focusing on fairness, transparency, privacy, and robustness.

UK AI Strategy

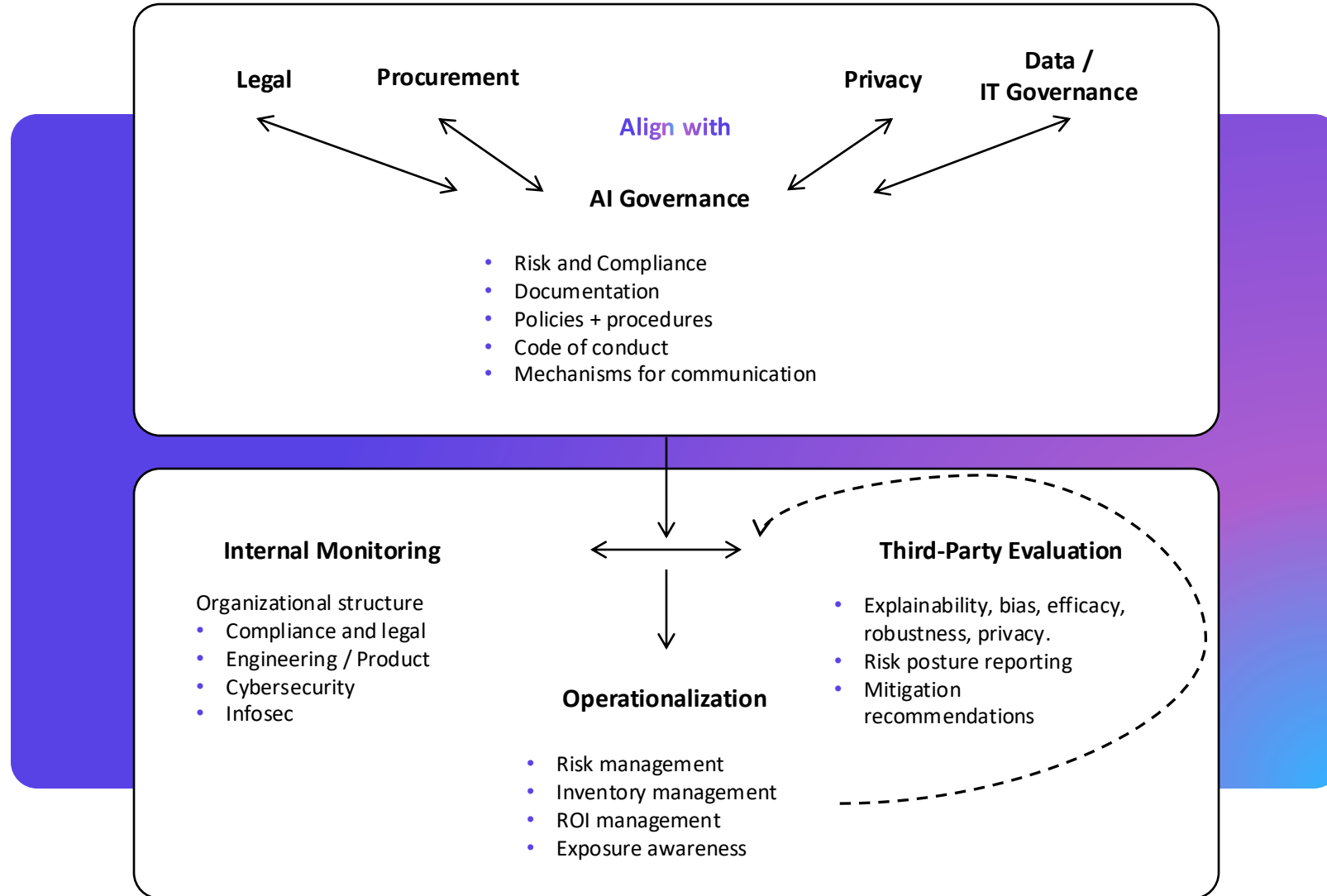
A flexible, principles-based approach to AI regulation, emphasizing innovation while ensuring that AI systems are aligned with ethical standards. Focuses on safety, transparency, and accountability, and aims to support both innovation and consumer protection.

The UK vs. EU vs. US



	UK	EU	US
Regulatory Approach	Principles-based, flexible, risk-based approach. Focus on fostering innovation while ensuring ethical AI.	Comprehensive, legally binding framework with specific risk categories.	Decentralized, industry-driven, with voluntary guidelines and sector-specific regulations.
Key Frameworks/Legislation	<ul style="list-style-type: none">• UK AI Strategy (2021)• AI Regulatory Framework (work in progress)• AI and Data Bill (proposed)	<ul style="list-style-type: none">• EU AI Act (AIA)• General Data Protection Regulation (GDPR)• Digital Services Act (DSA)	<ul style="list-style-type: none">• National AI Initiative Act (2020)• Algorithmic Accountability Act (proposed)
Risk Classification	Risk-based but not as formalized as the EU's approach. Focus on high-risk AI sectors like healthcare and finance.	Clear classification of AI systems into high-risk, limited-risk, and minimal-risk categories.	Risk-based approach in some proposals (e.g., Algorithmic Accountability Act) but generally lacks formalized risk categories at the federal level.
Focus Areas	Safety, transparency, ethical AI, fostering innovation, and sector-specific regulation.	Safety, transparency, accountability, protecting fundamental rights, human oversight, and addressing AI's social impact.	Ethical AI, transparency, fairness, privacy, consumer protection, and AI's societal impacts.
Legal Enforceability	Not fully enforced yet, but proposals for legal regulation are underway.	Legally binding with penalties for non-compliance, particularly for high-risk AI systems.	Currently voluntary frameworks like NIST AI RMF; no comprehensive AI law yet at the federal level.

Adopting and Scaling AI Responsibly



AI Risk Appetite and Posture



AI Risk Appetite

The level of risk your business is willing to accept in adopting AI technologies.

- Ensures trustworthy use and responsible scaling of AI within organisations.
 - *What level of risk are we willing to accept in adopting AI technologies to drive innovation and competitive advantage?*

AI Risk Posture

The strategic approach a business takes in managing and responding to risks.

- Allows for the measuring, mitigating, and monitoring, of risks.
 - *What strategies and resources do we need to put in place to manage AI risks effectively and ensure long-term sustainability?*

Considerations and Trade-Offs

- **Innovation:** Balancing rapid AI development with the need to manage risks like bias and security.
 - Faster innovation at the risk of ethical or security issues.
- **Adoption:** Fast-tracking AI adoption vs. ensuring compliance with ethical and regulatory guidelines.
 - Slower AI adoption with reduced exposure to risks and compliance issues.
- **Data:** Using personal data to improve AI performance while protecting individual privacy.
 - Better AI outcomes vs. privacy concerns.
- **Flexibility:** Meeting regulations while maintaining flexibility for AI innovation.
 - Enhanced risk control at the expense of operational speed.

Trade-off Example

A company is introducing an AI-driven recruitment platform to streamline its hiring process. The platform is designed to assess resumes, conduct initial candidate screenings, and even provide interview recommendations based on historical hiring data. While the system is meant to enhance efficiency and reduce human bias, concerns arise about whether it inadvertently perpetuates biases, particularly around gender, race, and socioeconomic background.

Trade-off Example

- **Innovation vs. Bias Management**
 - Faster adoption improves efficiency but risks reinforcing historical biases.
 - Slower deployment allows for thorough bias auditing and mitigation but delays benefits.
- **Faster Innovation vs. Ethical Issues**
 - Rapid rollout achieves quick results and cost savings but may raise ethical concerns like discrimination.
 - A slower rollout ensures compliance with laws but may lead to missed opportunities or delays.
- **Personal Data vs. Privacy Concerns**
 - More data improves model accuracy but increases privacy and consent risks.
 - Using limited, publicly available datasets protects privacy but may reduce performance.
- **Regulatory Compliance vs. Flexibility for Innovation**
 - Strict compliance minimises legal risks but slows innovation.
 - Greater flexibility enables faster progress but increases regulatory exposure.



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