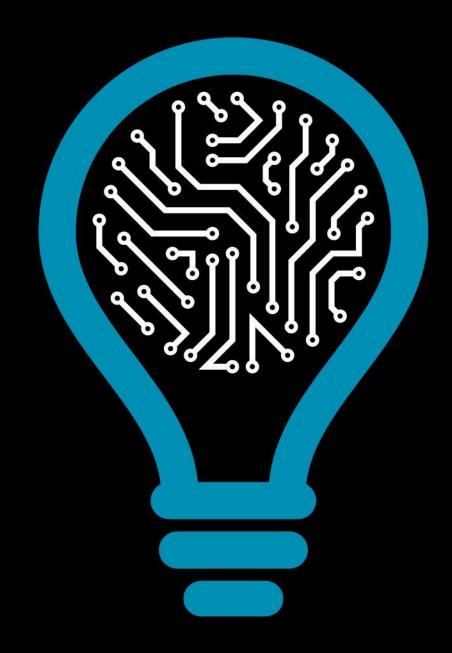
Responsible AI - Lecture 2

TAIA - Advanced Topics on Artificial Intelligence

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INSTITUTE FOR SYSTEMS
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Outline

- 1. Principles of Responsible AI (beyond Interpretability)
- 2. Law meets Al: Friend or Foe?
- 3. Take-home messages and further readings

1. Principles of Responsible AI (beyond Interpretability)

Responsible Al relies on fundamental principles

- **Responsible AI** is a framework that guides how we should address the challenges around artificial intelligence from both an **ethical**, **technical** and **legal** point of view^[1]
 - We must resolve ambiguity for where responsibility lies if something goes wrong!
- This framework relies on fundamental principles^[2]:
 - Accountability
 - Interpretability
 - Fairness
 - Safety
 - Privacy

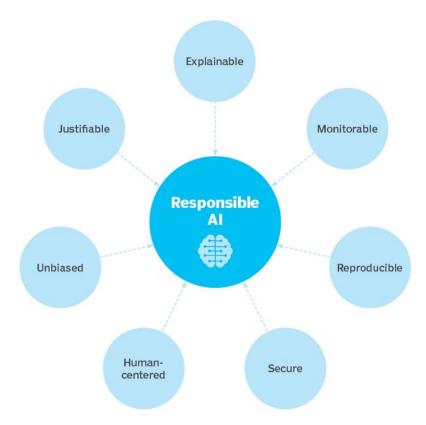


Figure - Responsible AI (Image from [1])

Accountability: who will take the responsibility?[1]

- People should be accountable for Al systems^[1]
 - This principle is the baseline, hence, all the other principles can be seen as branches
- Some important ideas are:
 - o Implement and use a human-centered design approach^[2]
 - Identify multiple metrics to assess training and monitoring, and ensure that these
 metrics are appropriate for the context and goals of our system^[2]
 - Examine our **raw data** (e.g, missing values, incorrect labels, biases, feature redundancy)^[2]
 - Understand the limitations of our databases and models^[2]
 - Learn best practices from software quality engineering to make sure the AI system is working as intended and can be trusted^[2]
 - Continue to test, monitor and update the system after deployment^[2]

Fairness: how to deal with bias?^[1]

- Al models learn from existing data collected from the real world, and so an accurate model may learn or even amplify problematic pre-existing biases in the data based on sensitive characteristics^[2]
- Regarding this issue, we should:
 - Interact with social scientists, humanists, and other relevant experts for our product to understand and account for various perspectives^[2]
 - Consider how the technology and its development over time will impact different use cases (e.g., what outcomes does this technology enable)^[2]
 - Assess fairness in our datasets (e.g., identifying representation and corresponding limitations)^[2]
 - Check the system for **unfair biases**^[2]
 - Analyse the performance of the system, taking into account the different metrics we've
 defined^[2]

Safety: achieving reliable and safe Al systems^[1]

- Safety and security intend to ensure that AI systems behave as intended, regardless of how attackers try to interfere^[2]
- Regarding this issue, we should:
 - Consider if there are incentives to make the system misbehave^[2]
 - Identify what unintended consequences would result from the system making a mistake, and assess the likelihood and severity of these consequences^[2]
 - Build a rigorous threat model to understand all possible attack vectors^[2]
 - Research into adversarial machine learning, as it continues to offer improved performance for defenses^[3] and provable guarantees^[2, 4]
 - Check if there are other vulnerabilities in the Al supply chain^[2, 5]

Privacy: can Al reveal aspects of its training data?[1]

- Al systems must prioritise and safeguard consumers' privacy and data rights and provide explicit assurances to users about how their personal data will be used and protected^[1]
- Regarding this issue, we should:
 - o Identify whether our AI model can be trained without the use of sensitive data^[2]
 - Anonymise and aggregate incoming data using best practice data-scrubbing pipelines (e.g., removing personally identifiable information (PII) and outlier or metadata values that might allow de-anonymisation)^[2]
 - Train our models using **federated learning**^[3], where a fleet of devices coordinates to train a shared global model from locally-stored training data
 - Perform tests based on "exposure" measurements^[4] or membership inference assessment^[5] to estimate whether our model is unintentionally memorising or exposing sensitive data

[6] Shokri et al. "Membership Inference Attacks against Machine Learning Models"

2. Law meets Al: Friend or Foe?

Why the EU got it right: regulating data before Al!

- The **General Data Protection Regulation (GDPR)**^[1] is a privacy and security law drafted and passed by the European Union (EU), that imposes obligations onto organizations anywhere
 - This goes back to the **right to privacy**, part of the **1950 European Convention on Human Rights**: "Everyone has the right to respect for his private and family life, his home and his correspondence"^[2]
- Moreover, the EU-GDPR is all about people's privacy rights^[1]:
 - The right to be informed
 - The right of access
 - The right to rectification
 - The right to erasure
 - The right to restrict processing
 - The right to data portability
 - The right to object
 - Rights in relation to automated decision making and profiling

Can we trust AI? Towards "Trustworthy AI" in the EU

- The European Commission appointed a group of experts to provide advice on its artificial intelligence strategy: **High-Level Expert Group on Al**^[1]
- According to the Guidelines, trustworthy Al should be:
 - Lawful: respecting all applicable laws and regulations
 - **Ethical**: respecting ethical principles and values
 - Robust: both from a technical perspective while taking into account its social environment
- Several important guidelines were proposed^[2]:
 - Human agency and oversight: All systems should empower human beings
 - Technical Robustness and safety: All systems need to be resilient and secure
 - Privacy and data governance: data governance mechanisms must be ensured
 - **Transparency**: the data, system and AI business models should be transparent
 - o **Diversity, non-discrimination and fairness**: All systems should be accessible to all
 - Societal and environmental well-being: Al systems should benefit all human beings
 - Accountability: ensure responsibility and accountability for Al systems and their outcomes

The Al Act and how it will impact our lives

- The Al Act^[1] is a **document proposed by the European Commission** that contains several **harmonised rules**^[2] regarding **Al applications**, emphasising that its approach is shaped by EU values and **risk-based**, ensuring both **safety** and **fundamental rights protection**
- What does the Al Act propose?^[2]
 - Prohibition of unacceptable Al practices (e.g., social scoring)
 - Regulation of high-risk Al systems (e.g., Al used in the context of recruitment)
 - Conformity assessment (i.e., under the EU product safety framework)
 - Transparency obligations for potentially deceptive AI systems
 - Ex post market surveillance (i.e., post-market monitoring system)
 - Governance (i.e., authorities must be appointed for the application and implementation)
 - Pre-emption of national Al regulatory frameworks (i.e., regulated by the EU)
 - Monitoring and enforcement (i.e., done by the Member States)
 - Compliance with the prohibitions and regulatory requirements

3. Take-home messages and further readings

A (tentative) fair and accurate summary of this lecture

- The development of data-driven artificial intelligence applications is pushing the limits of the applications of these algorithms in our lives: this rapid evolution motivated the need to ethical, legal and technical regulatory frameworks based on specific principles: accountability, interpretability, fairness, safety, privacy
- At the European level, there have been several proposals to regulate data and Al-based applications: EU-GDPR, Trustworthy Al Initiative and Al Act
- Open regulatory challenges will focus on the impacts of Al in ethics, transparency, fairness, safety, sociology and sustainability^[1, 2, 3, 4]
- Multidisciplinary work is, more than ever, of utmost importance and useful!

Further readings...

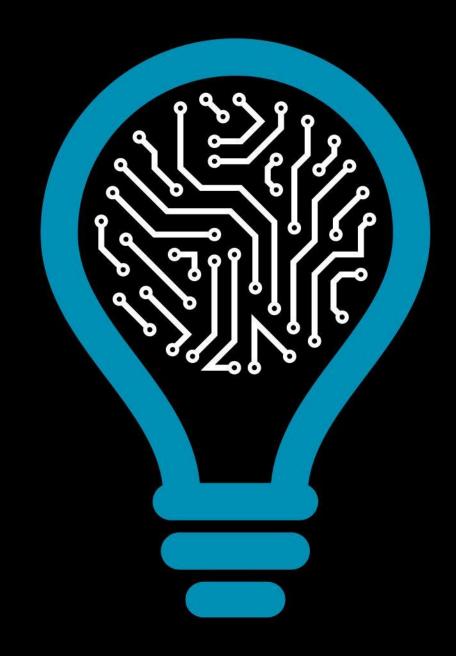
- Carlini et al. "The Secret Sharer: Evaluating and Testing Unintended Memorization in Neural Networks"
- Shokri et al. "Membership Inference Attacks against Machine Learning Models"
- Papernot et al. "Scalable Private Learning with PATE"
- Kannan et al. "Adversarial Logit Pairing"
- Wong and Kolter "Provable defenses against adversarial examples via the convex outer adversarial polytope"
- Gu et al. "BadNets: Identifying Vulnerabilities in the Machine Learning Model Supply Chain"
- Montenegro et al. "Privacy-Preserving Generative Adversarial Network for Case-Based Explainability in Medical Image Analysis"
- Pessach and Shmueli "Algorithmic Fairness"
- https://www.ibm.com/blogs/research/2019/09/adversarial-robustness-360-toolbox-v1-0/
- https://www.forbes.com/sites/anniebrown/2021/07/02/ais-role-in-the-future-of-data-privacy/?sh=b23e83918c0d

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