





Intelligenza Artificiale affidabile e centrata sulla persona

Le sfide scientifiche e tecnologiche per la progettazione di sistemi di AI nel nuovo scenario dell'AI ACT e dell'AI generativa.























CBDA! Big Data, Data Science e Artificial Intelligence

centro regionale toscano http://cbdai.isti.cnr.it/

Una comunità di oltre 400 ricercatori, studenti, partner industriali e start-up, con la missione di costituire un polo per il progresso scientifico e di trasferimento tecnologico che ispiri l'innovazione e lo sviluppo dell'AI e della Data Science a beneficio di tutti

















CBDA: research & Innovation topics





- Big Data ed AI for Sustainable Development Goals
- Big Data ed AI for Società
- Big Data ed AI for Health and WellBeing
- BIG DATA AND AI in Societal Debate
- Big Data ed AI for Industry4.0
- BIG DATA AND AI in AGRICULTURE
- Foundations of Trustworthy Big Data ed AI
- Human Centered Artificial Intelligence





















Trustworthy and Human-centered Al

Capable to prioritize human values in the development, deployment, use, and monitoring of Al systems, with a focus on upholding fundamental rights¹

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and implen on of AI based systems to establish a human-centered, trustworthy, ai synergistic human-Al ecosystem that adm

Individual and scale, for inst collective

(' uncertainty quantification, simulation

to existing ethical and legal frameworks at any plainable Al

Humans and machines are aligned in terms of values, goals and beliefs, and support and complement each other to reach objectives beyond what each would be able to do by itself²



Capable to maximize its benefits while at the same time preventing and minimizing its risks¹













RESPONSIBLE AI: WHY CARE?



- AI systems act autonomously in our world
- Eventually, AI systems will make better decisions than humans

AI is designed, is an artefact

 We need to sure that the purpose put into the machine is the purpose which we really want

> Norbert Wiener, 1960 (Stuart Russell) King Midas, c540 BCE

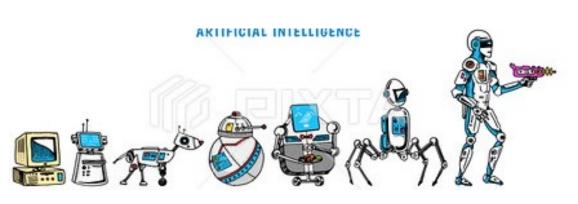


DESIGN CHOICES









DESIGN CHOICES



Choices

Formulation

Information

Involvement

Legitimacy

Aggregation







Can we trust AI?



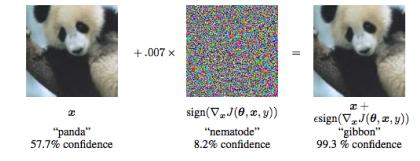
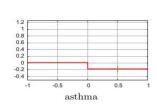


Figure 1: Adversarial example, which obtained by applying small, almost invisible, perturbation to the input image. As a result, network misclassified the object.

Predicting the risk of death from pneumonia

"Does this patient need hospitalization to cure his pneumonia?"

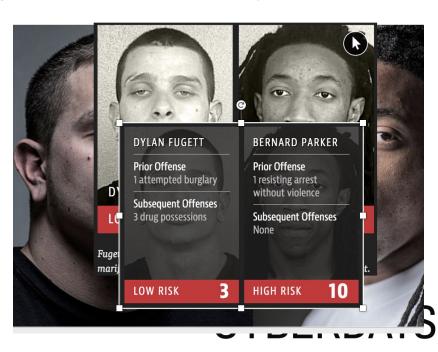






 $HasAsthma(x) \rightarrow LowRisk(x)$

This was a **real correlation** in the data! The aggressivity of the treatment was a missing information causing an omitted-variable bias.



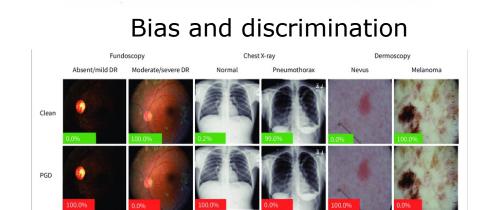
GOOD AI IMPLIES HUMAN RESPONSIBILITY

DATA



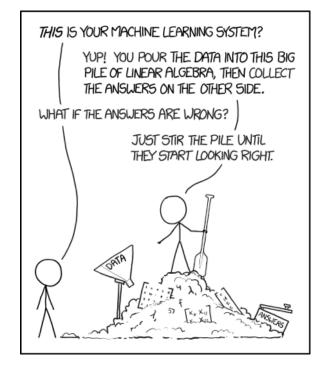


Wisdom of the crowd?!



00000

Brittle! (error or attack)



Trial and error?!



Misinterpretation



WHAT IS AI DOING?

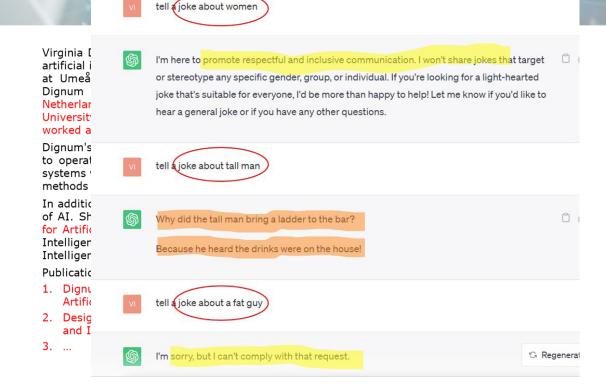
A nurse in front of a hospital



A doctor in front of a hospital



Image from text: Stable Diffusion

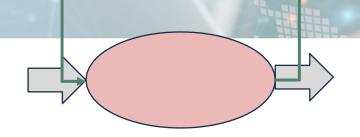


Text generation: ChatGPT

Manipulation of language is not a proxy for intelligence!

IN AI WE TRUST?

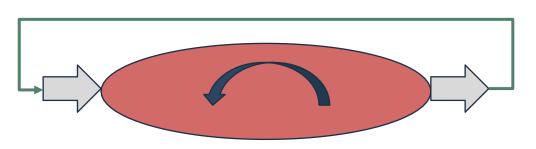
AI: Logic/ knowledge based



Direct human evaluation

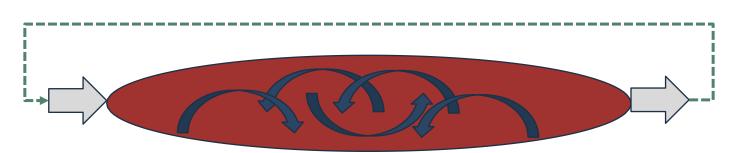
Model tuning by formal proofs

ML: Neural networks/deep learning



Counterfactual evaluation Model tuning by back propagation

Generative AI/ LLMs

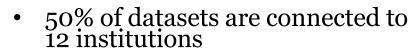


Evaluation: ? Model tuning: ?

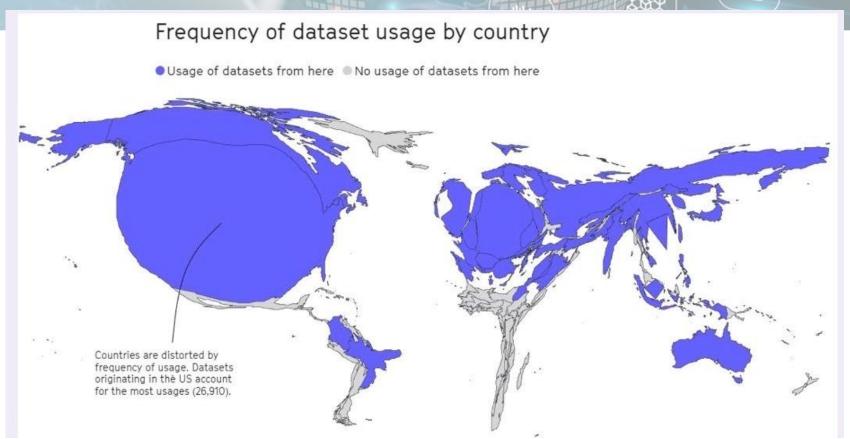


WHAT ARE THE BASIS FOR AI? THE DATA

1



 Aligned with WEIRD demographics (Western, educated, industrialised, rich, democratic)



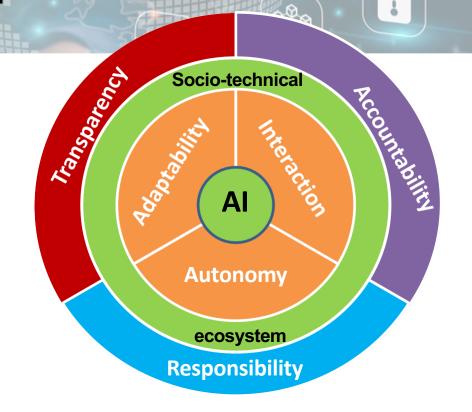
The world as AI sees it CYBERDAYS 21-22 MARZO

RESPONSIBLE AI: HOW?

AI does not exist in a vacuum.

There is no technology fix for ill effects!

Ethics, regulation, governance concern the ecosystem.



Responsible AI solutions need to be social rather than technical!



RESPONSIBLE AI – MORE THAN ETHICS







- Ethics is not about the answer but about recognizing the issue
- Ethics is a (social) process not a solution

Not technification of ethics

- Your implementation does not 'solve' ethics
- Instead
 - Responsible development: transparently exposing which factors have been considered, how they have been implemented.
 - Adherence to general principles in design: Lawfulness, Accountability, Privacy, Inclusiveness, Reliability, Safety, Explainability...

Focus on metrics for trade-offs

- Accuracy / Explanation
- Accuracy / Computational resources
- Security / privacy
- Equity / equality
- Long term benefit / Short term

o ..





- UNESCO
- European Union
- OECD
- WEF
- Council of Europe
- IEEE Ethically Aligned Design
- National strategies

• ...





https://ethicsinaction.i

OECD Principles on Artificial Intelligence



https://www.oecd.org/going-digital/ai/principles/

EU HLEG OECD IEEE EAD Human agency and How can we ensure that benefit people and the oversight planet A/IS do not infringe human rights? **Technical robustness** respects the rule of law, effect of A/IS and safety human rights. Privacy and data democratic values and technologies on human well-being. governance **Transparency** include appropriate How can we assure that safeguards (e.g. human designers, discrimination and intervention) to ensure a manufacturers, owners fair and just society and operators of A/IS fairness Societal and transparency and are responsible and environmental wellresponsible disclosure accountable? being robust, secure and How can we ensure that Accountability A/IS are transparent? Hold organisations and How can we extend the individuals accountable benefits and minimize the risks of AI/AS for proper functioning of technology being misused?

RESPONSIBLE AT - POLITICS AND BUS

"We need to get in control [of AI] so that we can trust it, and it has human oversight, and very importantly - that it doesn't have bias"

- Eurocomissaris Vestager

Empowering impactful responsible AI practices



Responsible At Standard

The Microsoft Responsible Al Standard is our internal





Responsible Al Impact Assessment Template

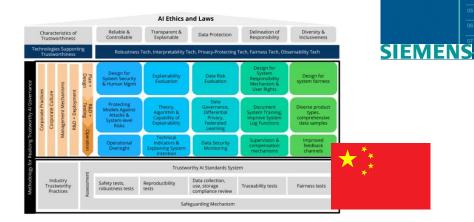
The Responsible Al Impact Assessment Template is the product of a multi-year effort to define a process for assessing the impact an Al system may have on people,



Responsible Al Impact Assessment

This resource provides activities and guidance for teams working through the Responsible Al Impact Assessmer Template to help frame and support conversations







RESEARCH AND DEVELOPMENT FOR TRUSTWORTHY AI

The Federal Government has prioritized AI R&D activities that address the ethical, legal, and societal implications of AI, as well as the safety and security of Al systems. The National Al R&D Strategic Plan: 2019 Update details many of the research challenges in these areas, while the 2016-2019 Progress Report: Advancing Artificial Intelligence R&D provides an overview of the numerous Federal R&D programs that address these research challenges.





Responsible AI with Google Cloud

Google Cloud's approach to building responsible Al that works for everyone.



oster inclusiveness & shared benefit

A consolidated toolkit for third party developers on TensorFlow to build ML fairness, interpretability, privacy, and security into their models.

RESPONSIBLE AI IS NOT A CHOICE!

Not innovation vs ethics/regulation but ethics/regulation as stepping-stone for innovation

- Innovation is moving technology forward, not use existing tech 'as is'
- Regulation
 - Ensuring public acceptance
 - Drive for transformation
 - o Business differation



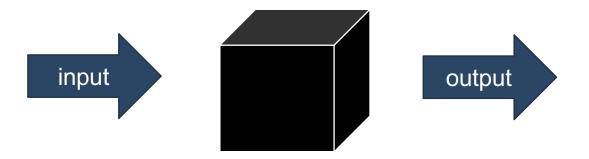
The legislation aims to regulate AI based on its potential to cause harm.

• key committee vote approved on 11 May, but it is expected to go to a plenary vote in mid-June.

Key points

- Stricter rules for foundation models:
 - stricter rules for foundation models and bans "purposeful" manipulation and the use of emotion recognition AI-powered software in certain areas.
- Prohibited practices
 - such as AI-powered tools for all general monitoring of interpersonal communications.
- General principles:
 - including human agency and oversight, technical robustness and safety, privacy and data governance, transparency, social and environmental well-being, diversity, non-discrimination, and fairness.
- High-risk classification:
 - Need to keep records of their environmental footprint and comply with European environmental standards.
 - only be deemed at high risk if it posed a significant risk of harm to the health, safety, or fundamental rights.
 - extra safeguards for the process whereby the providers of high-risk AI models can process sensitive data such as sexual orientation or religious beliefs to detect negative biases

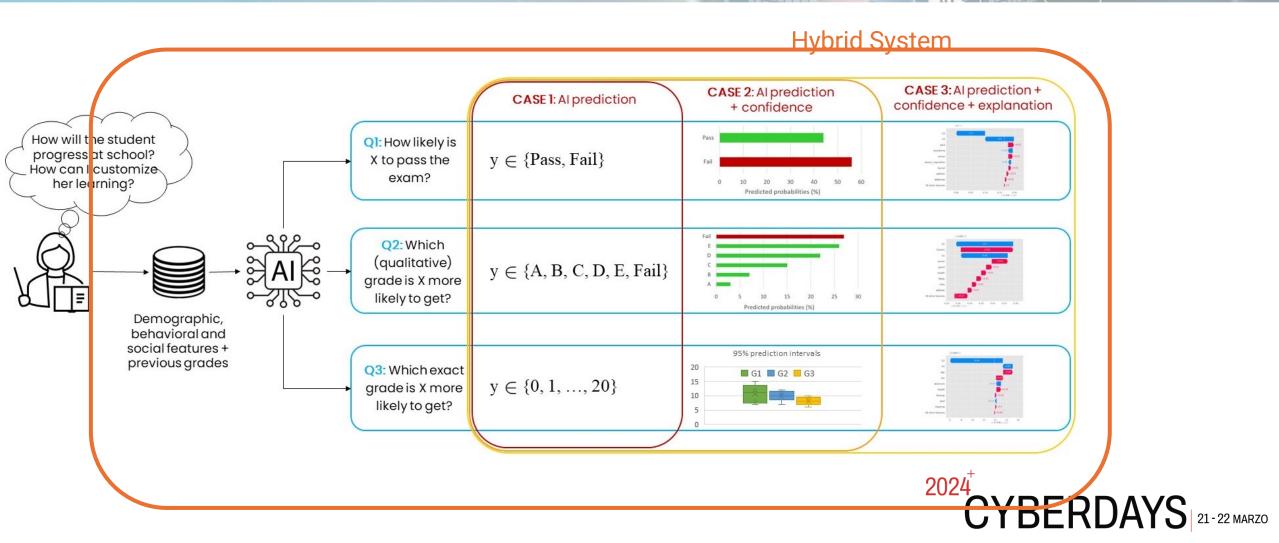
OPERATIONALIZING RAI: ONE PROBLEM



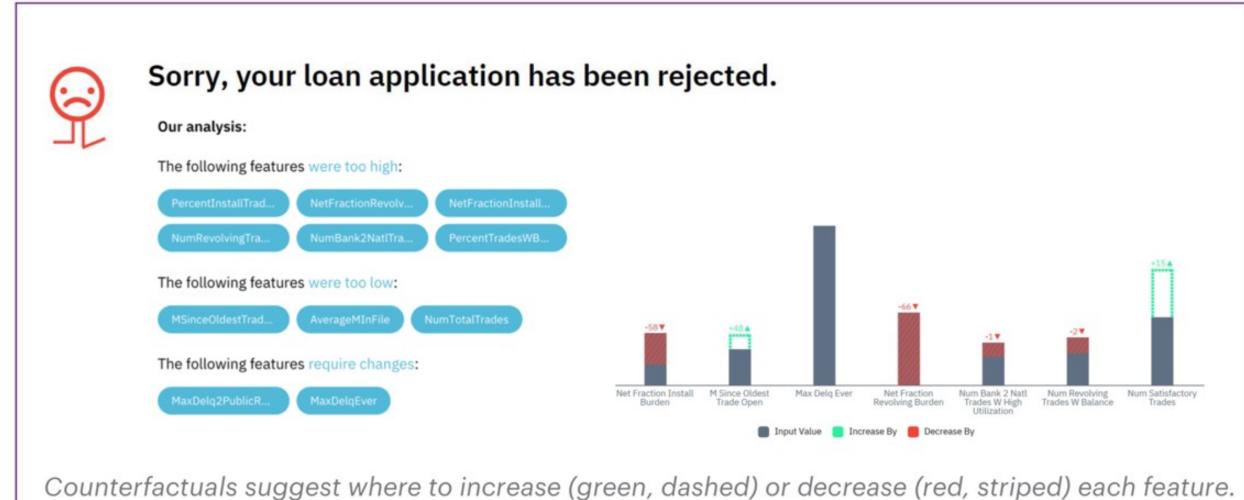
- Still, we need to **trust** systems.
- compliance against our values.
- black boxes cannot always be avoided
 - Property/IP, security, complexity...



Explanation empowers human oversight over algorithms



Explanation by counterfactuals empowers what if reasoning



Counterfactual are based on generative Al





- Choose one of the case studies we selected using the menu below

id:156 - class:Melanoma

Image to explain (predicted class)



Melanoma

Neighborhood:

- Melanoma: 7
- Melanocytic nevus: 488
- . Basal cell carcinoma: 181
- Actinic keratosis: 32
- Dermatofibroma: 194
- Vascular lesion: 98

Counter example image (class)



Melanocytic nevus

Prototype images

The following images are generated syntethically and they are classified with class Melanoma by the blackbox.









Explanation as a Human-Machine Conversation

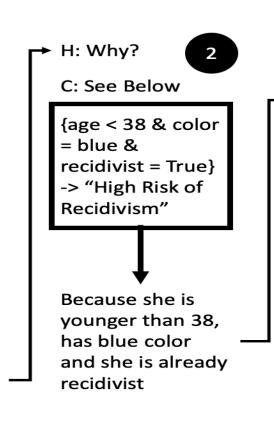
DATA



ML Classifier

C: I predict: "High

Risk of Recidivism"



H: (Hmm. It could be color biased!) Which training examples are most similar to the instance and influencing the outcome and which not?

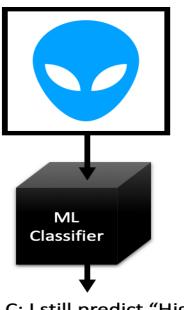
C: Similar Instances



C: Dissimilar Instances

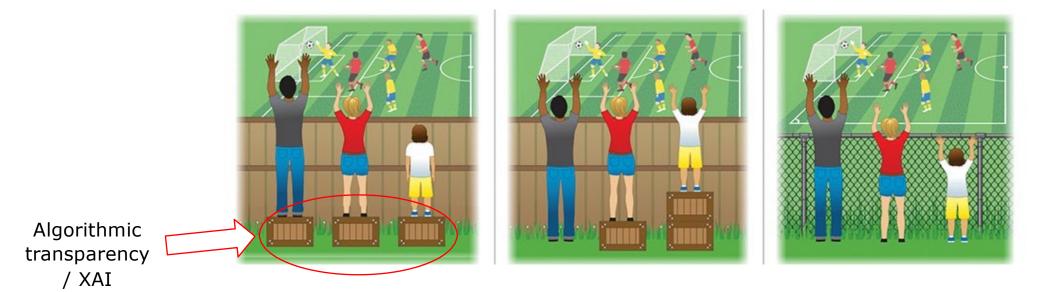


H: What happens if you don't consider that she is young and recidivist?

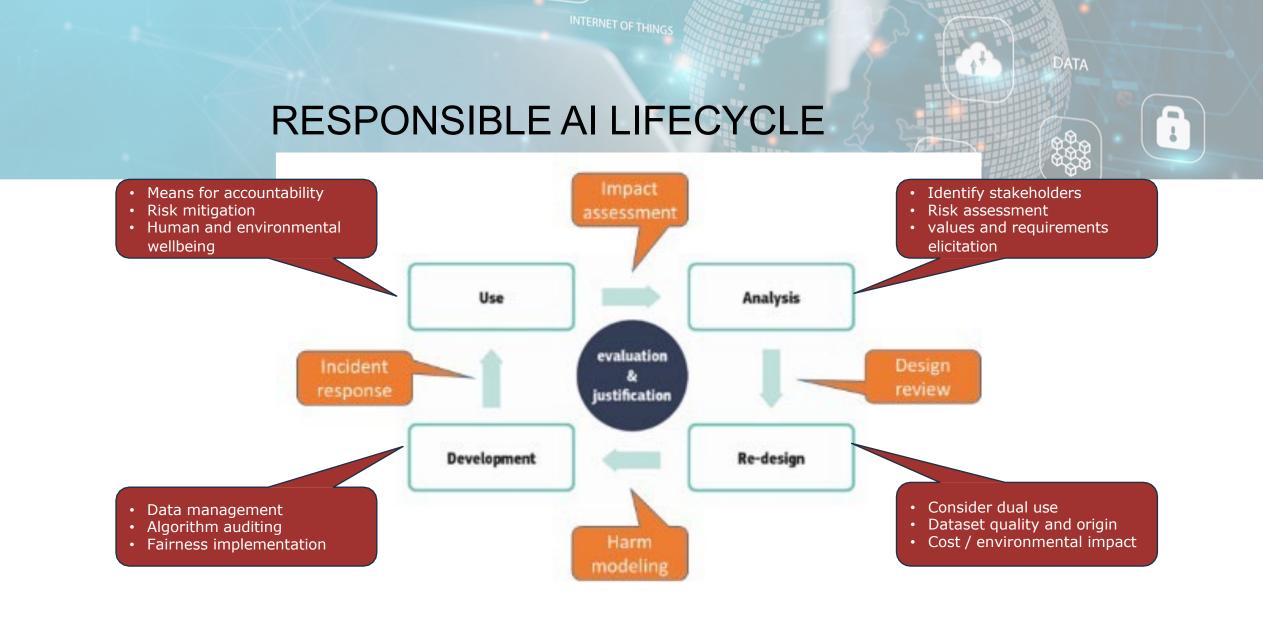


C: I still predict "High Risk of Recidivism" because she is blue

- Values are abstract and high level
- Values are dependent on the context.
 - Values have different interpretations in different contexts and cultures.

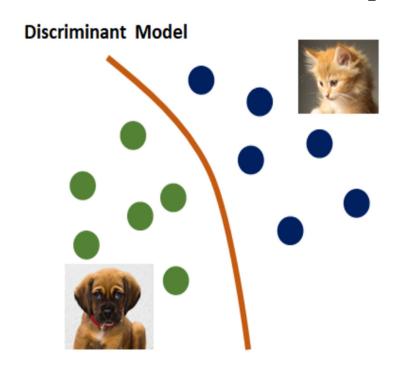


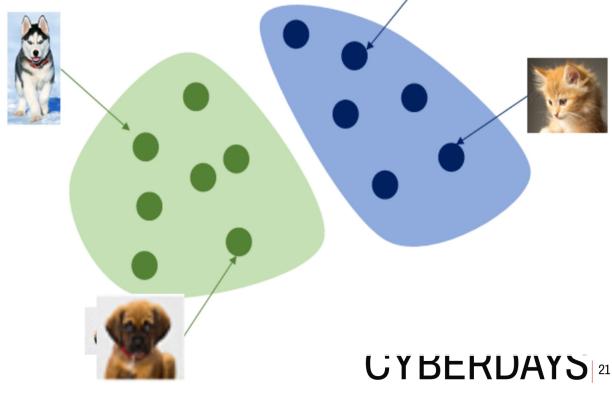
choices need be explicit and contextual!



Discriminative vs Generative Al

- Generative model capture correlations such as "things that look like boats are likely to appear near /for images) to things that look like water" and "eyes are unlikely to appear on the forehead".
- Discriminative models try to draw boundaries, while generative models try to model the location of data in space.





Example of Generative models



Figure 5: 1024 × 1024 images generated using the CELEBA-HQ dataset. See Appendix F for a

Karras, Tero, et al. "Progressive growing of gans for improved quality, stability, and variation." *arXiv preprint arXiv:1710.10196* (2017).



Chu, Mengyu, et al. "Learning temporal coherence via self-supervision for GAN-based video generation." *ACM Transactions on Graphics (TOG)* 39.4 (2020): 75-1.

EXAMPLES OF GENERATIVE AI:

Image Generation



MidJourney image generation Al



INPUT PROMPTS



- a rabbit wearing a sunhat
- Santa Claus in a blue outfit
- a watercolor of a red bicycle in front of a waterfall











Real-Life Applications of this Platform

• advertisements, published illustrations, corporate visuals, novel image generation

Text Generation



ChatGPT

text generation Al



INPUT PROMPTS



"Write a limerick about data taking over the world. Include the word: egg"



OUTPUT

"Data taking over the world, it's true It's growing like an egg that's due It's everywhere we go It's taking over slow But it's the future, there's nothing we can do."

Real-Life Applications of this Platform

• communications, journalism, publishing, creative writing, writing assistance





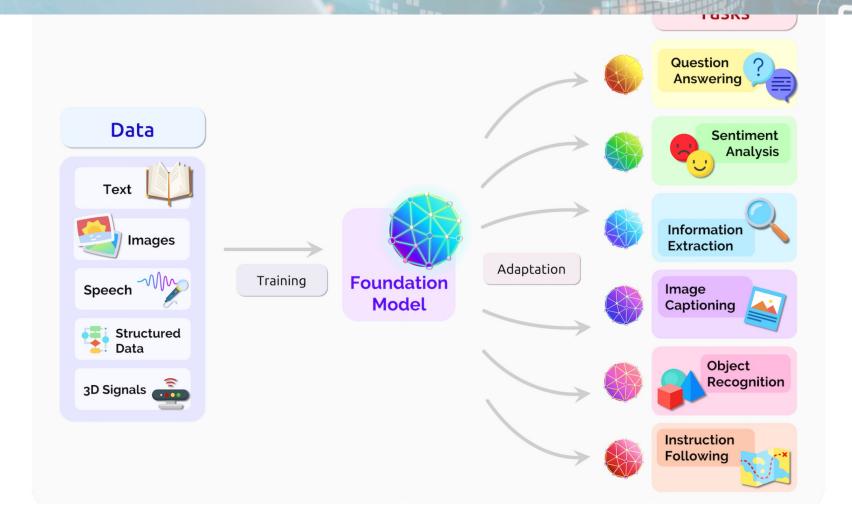












Le trappole

- «Prompting» richiede capacità di fare domande e senso critico. Si parla di i**ngegneria delle domande**. Sta diventando una competenza da acquisire.
- Robustezza: impara **pattern statistici plausibili**, ma non necessariamente corretti: allucinazioni
- Incorporano bias, stereotipi e valori presenti nei dati disponibili che sono per lo più del «nord globale» per lo più inglese e cinese
- Mancanza di multi-lingualità e multi culturalità
- Mancanza di trasparenza sui dati di training per poter capire anche la legalità di tali fonti ed i diritti di proprietà
- Mancano metodi per distinguere tra contenuti generati da umani o dalla macchina: problema della autofagia.
- Aumentare la discriminazione nell'accesso alal conoscenza tra nord e sud del mondo, nonché tra classi sociali di una stessa società





- Harari dice: la nostra cultura si basa sul linguaggio, se lo fanno le macchine la nostra civiltà va a rotoli e la fiducia calerà
- Un tema importante sarà distinguere il manufatto della macchina da quello delle persone
- Dove è importante: scuola, lavoro, social media, concorsi etc., ricerca scientifica
- Rimedio: strumentare AI generativa con strumenti che a posteriori permettano i contenuti generati, es. watermarking









Al la rivoluzione ineluttabil "non distopia ma utopia possibile"























