AI (R)evolution: How did we get here And where are we going?



Sandia INCOSE Enchantment Chapter

Melanie Moses UNM Professor of Computer Science External Faculty, Santa Fe Institute <u>https://moseslab.cs.unm.edu/</u> May 8, 2024





What I'll talk about today

- My evolving perspective on AI
- How generative AI works (and doesn't)
- Is scale is all we need? (No)
- Overview of AI@UNM & Sandia/LANL partnerships
- The future of AI for New Mexico & the public good

Southwestern Indian Polytechnic Institute (SIPI) 1st place Swarmathon 2018

My evolving thoughts on AI

- I study bio-inspired robotics and complex systems
- I think of AI as a complex adaptive system that we are integrating into our complex socio-technical system
- This is exciting and scary!
 - Al will help solve scientific problems
 - Current AI encodes biases, promotes surveillance & concentrates power
 - Al codes even better than it writes
 - How will humans co-evolve with AI?



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Humayra Tasnim

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Highly accurate protein structure prediction with AlphaFold, *Nature* 2021

Al-assisted science is potentially revolutionary! Biomedicine: Antibiotics, therapeutics, diagnoses, vaccines ...

Climate & energy: materials for batteries, biodegradable plastics, fusion(!?)

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CODED BIAS



www.dailymaverick.co.za/article/2021-04-16-this-weekend-were-watching-coded-bias/

https://www.codedbias.com/

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Finance worker pays out \$25 million after video call with deepfake 'chief financial officer'

By Heather Chen and Kathleen Magramo, CNN 2 minute read · Published 2:31 AM EST, Sun February 4, 2024





Authorities are increasingly concerned at the damaging potential posed by artificial intelligence technology. boonchai wedmakawand/Moment RF/Getty Images

https://www.cnn.com/2024/02/04/asia/deepfake-cfo-scam-hong-kong-intl-hnk/index.html

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From ChatGPT Copy code Import numpy as no do my students' mport matplotlib.pyplot as plt f logistic_map(r, x): homework r = np.linspace(2.5, 4.0, 10000) # Values of r x = 10-5 * np.ones_like(r) # Initial condition for x iterations = 1000 To Students: Use last = 100 # We'll only display the last few iterations to see the st. for i in range(iterations): ChatGPT to do your x = logistic map(r, x)if i >= (iterations - last): homework plt.plot(r, x, ',k', alpha=0.25) plt.xlim(2.5, 4) plt.ylim(-0.1, 1.1) plt.xlabel('r') Bifurcation diagram of the logistic map plt.ylabel('x') plt.title('Bifurcation diagr plt.show() 1.0 lot_logistic_map() To see the bifurcation diagram: 0.8 1. Install the necessary libraries: 'pip insta 2. Run the above code. 0.6 × 0.4 0.2 0.0 2.6 2.8 3.0 3.2 3.4 3.6 3.8 4.0

Here's a Python code using 'matplotlib' to display the logistic map;

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Generate DALL-E prompts ...

GPT4: A market scene in 2150 where humans buy tech enhancements for their bodies, while robots buy organic parts to feel more human.

How does AI work? Traditional algorithms vs. AI

Traditional computer programs are like recipes: step by step instructions to achieve a specified outcome.

A simple cake recipe

- Preheat oven to 350 degrees F
- 1. 2. 3. Mix 2 cups sugar and 4 sticks butter
- Add 8 eggs
- 4. Mix in 3 cups flour

Bake in greased pan for 50 minutes



http://www.mykitchenintherockies.com/high-altitude-baking/

How does AI work? Traditional algorithms vs. AI

Traditional computer programs are like recipes: step by step instructions to achieve a specified outcome.

Traditional algorithms can fail, but we usually understand why

Generative AI is different: we usually can't explain its predictions or its mistakes

A simple cake recipe

- 1. Preheat oven to 350 degrees F
- 2. Mix 2 cups sugar and 4 sticks butter
- 3. Add 8 eggs
- 4. Mix in 3 cups flour

Bake in greased pan for 50 minutes



Unless you are above 5000 feet

Context matters!

How does generative AI work?

Large Language Models (LLM, like ChatGPT) use neural networks to find patterns & correlations in data. Those are used to make predictions.

Billions of nodes trained on billions of texts predict what word is likely to come next

Words are vectors (a list of 1000's of numbers) grouped by similarity

Neural networks are pattern matching machines



https://medium.com/@hari4om/word-embedding-d816f643140 13 https://arstechnica.com/science/2023/07/a-jargon-free-explanation-of-how-ai-large-language-models-work/

LLMs learn by predicting what word comes next

Roses are red, violets are xxx

Cherries are red, grapes are xxx

Green means go, red means xxx





How does generative AI work?

Large Language Models (LLM, like ChatGPT) use neural networks to find patterns. Those data can be used to make predictions

Billions of nodes trained on billions of texts predict what word is likely to come next

Words are vectors (a list of 1000's of numbers) grouped by similarity

Neural networks are pattern matching machines



https://arstechnica.com/science/2023/07/a-jargon-free-explanation-of-how-ai-large-language-models-work/

This is tricky!

Transformers: "attention mechanisms" disambiguate multiple possible meanings:

Bank: so many meanings!

The doctor examined the patient with her stethoscope (pronouns: confusing!)

Sometimes LLMs guess wrong. The guesses are usually plausible.

It is particularly difficult to recognize plausible falsehoods (BS not hallucinations!)







Paris - France + Germany = Berlin





The concepts behind transformers and LLMs - Great Article by @rasbt -"Understanding Encoder and Decoder LLMs"

Generative Pre-trained **Transformer** (GPT)



queries, keys, values are computed



Transformers, a core architecture in modern Large Language Models (LLMs), utilize a mechanism called self-attention to process sequential data, such as text, in a highly parallelizable manner. This architecture differs significantly from previous sequential models like RNNs or LSTMs, which processed data step-by-step.

Core Components

Self-Attention Mechanism: At the heart of the transformer is the self-attention mechanism. It allows
the model to weigh the relevance of all other tokens in the sequence for each token. This is crucial for
understanding context and dependencies, regardless of distance within the sequence.
 Multi-Head Attention: This is an extension of the attention mechanism where the model runs
through multiple attention processes in parallel, allowing it to selectively focus on different parts of the
sequence and capture a diverse range of relationships.

3. Positional Encoding: Since transformers do not inherently process sequential data as sequences, positional encodings are added to the input embeddings to provide information about the position of tokens in the sequence.

4. Feedforward Neural Networks: Each layer in a transformer contains a feedforward neural network which applies the same operations to different positions separately and identically. This layer can adaptively change the representation of each token based on the context provided by the attention mechanism.

5. Layer Normalization and Residual Connections: These components help in stabilizing the learning process. Residual connections help in mitigating the vanishing gradient problem by allowing gradients to flow through the network directly. \downarrow

Working Principle

The input sequence is first converted into vectors through embeddings. These embeddings are then processed by the multi-head attention mechanism, which allows the model to focus on different parts of the input sequence for better contextual understanding. The output of this stage is a set of attentionweighted features, which are then passed through a feedforward neural network for further processing. This process repeats across multiple layers, with each layer potentially capturing more complex relationships.

The transformer architecture's ability to handle all tokens of the input simultaneously and its reliance on attention to determine input dependencies makes it exceptionally powerful for tasks involving large sequences of data, leading to significant improvements in tasks like translation, text generation, and more.

Here's a detailed diagram illustrating the transformer architecture used in large language models. This visualization covers key components like multi-head attention, feedforward neural networks, and positional encodings, showing how input tokens are processed to produce contextual embeddings. This should provide a clear visual aid to understand the concepts I described earlier.

OpenAl ChatGPT May 2024

Write a haiku about the promise of Al

A haiku is a Japanese style of poetry with 5,7,5 syllables in each line

Silent mind, awake, Al's promise, boundless skies, Humanity's guide.





👰 Bojan Tunguz 🤣 @tunguz · Mar 3



This is a classic self-referential question that leads to a paradox if you try to answer it directly, as the percentages provided conflict with each other. Let's break down the options:

- If we assume A) 25% is correct, then there are two options (A and D) that suggest a 25% chance of being correct. However, this would mean that there's actually a 50% chance of picking one of the two correct answers by chance, which contradicts the initial assumption.
- 2. If B) 0% is assumed correct, it implies that none of the answers provided would be correct, including B) itself, which creates a contradiction.
- 3. Assuming C) 50% is correct would mean there are two options out of four that are correct, which is not the case as we see no duplicate percentage values other than what A and D suggest (25%), but this itself leads to a contradiction as explained in point 1.
- 4. D) 25% faces the same issue as A) 25%, making the assumption of a 25% chance of being correct impossible due to the presence of two answers indicating 25%.

Thus, the question is designed to be paradoxical and doesn't have a straightforward answer within the confines of its own logic. It serves as a playful brain teaser rather than a question with a definitive answer.

Scale is all you need?

Moore's Law:

Exponential growth

Computation doubles every ~2 years

1970:1000 transistors \rightarrow 2020: 50 BILLION transistors

Computers are now MILLIONS of times faster

Problems with "Intelligence" at scale

- Climate change¹
- Little context, many errors²
- Large corporations dominate

More transistors Faster computation More efficiency More memory More data Faster algorithms Bigger data centers Faster innovation

How did we get here?

Where are we going?

• Scale

- A few clever tricks in neural networks • Encoders, Transformers
- Human feedback by low paid workers (Kenyan workers paid \$2/day)
- Marketing & economic incentives have concentrated power in a few huge companies

- No one knows!
- No exponential lasts forever; all scaling hits constraints
- Feedback comes from LLM users
- Will smaller models become viable can we democratize AI & use it for good?
- How can we regulate AI & prevent harm?

Scale is all you need?

Test scores of AI systems on various capabilities relative to human performance

Within each domain, the initial performance of the AI is set to -100. Hun baseline, set to zero. When the AI's performance crosses the zero line, it : humans.



300 billion trillion calculations to train GPT-3 on 500 billion words.

It still makes up answers







Training compute (FLOPs) of milestone Machine Learning systems over time http://www.newson.org/action/actio

https://arxiv.org/abs/2202.05924

Figure 3: Trends in training compute of n102 milestone ML systems between 2010 and 2022. Notice the emergence of a possible new trend of large-scale models around 2016. The trend in the remaining models stays the same before and after 2016.

Slide Credit: Rick Stevens, Argonne Natl Lab & U Chicago

Let's create this visualization. Please hold on

while I prepare the figure.



Here's the figure illustrating the scaling of training requirements and capabilities for popular large language models like GPT-3, GPT-4, LLaMA, and Gemini:

WRONG answer from GPT4

Scale is what they have!



https://medium.com/@johnadjanohoun/microsoft-and-openai-to-forge-1 00-billion-supercomputer-for-ais-next-leap-2d5bbcf189f6



Saddi Arabia Obes big off Al. A \$40 billion investment

https://www.linkedin.com/pulse/saudi-arabia-goes-big-ai-40-billion-investment-ombrulla-d3eic



Max Tegmark 🤣 @tegmark

You know that AI is taking off when...Microsoft plans to power data centers with nuclear reactors: jobs.careers.microsoft.com/global/en/job/...

Microsoft, Meta, and Google's parent company, Alphabet, disclosed this week that they had spent more than \$32 billion combined on data centers and other capital expenses in just the first three months of the year.

The people who made the biggest fortunes in California's gold rush weren't the miners — they were the people selling the shovels. (NVIDIA)

<u>https://www.nytimes.com/2024/04/27/technology/ai-big-tech-spending.</u> <u>html?smid=em-share</u>

Is scale all you need in information systems?

Networks that deliver information rather than energy escape diminishing returns

• Information can be cheaply copied & communicated locally

Moore's Law and the future of AI:

Is scale all you need?



Moore's Law: The number of transistors on microchips doubles every two years Our World

OurWorldinData.org – Research and data to make progress against the world's largest problems.

Licensed under CC-BY by the authors Hannah Ritchie and Max Roser.

Scaling In Biology: Lessons for Computation



Transisto

r

Integrated circuit

AMD Opteron multi-core chip hillions of transistors

Scaling in Biology

A whale is 100 000 000 000 000 000 000 times bigger than an E. coli

10⁻¹² g





Glaw et al PLoS ONE 2012

Metabolic Scaling A striking universal(?) pattern





Metabolism rate of energy use governs the pace of life

Whole animal: $B \sim M^{3/4}$ Mass-specific: B/M ~ M^{-1/4}

Mice live fast and die young

- Heart rate, breathing rate
 20x faster
- Gestation times, Lifespans
 20x times shorter
- Invariants
 - 1 Billion heart beats
 - constant fraction to reproduction

Fractal Networks Generate 3/4 powers

Centralized hierarchical, fractal branching

- 1. Constant branching ratio,
- 2. Area preserving
- 3. Space filling
- 4. Invariant terminal units

-Capillaries same length, radius & delivery capacity

-Metabolism proportional to # of capillaries

5. Network volume proportional to mass







Doubling metabolic rate requires more than doubling network infrastructure

Metabolic rate is proportional to the number of (red) capillaries

To double metabolic rate, double the number of capillaries

Additional (black) network is needed to connect the 2 networks









Diminishing returns: Network size grows faster than network delivery rate

Assuming Mass and Volume are proportional then

 $B \propto M^{3/4}$



How do you fit a fractal network into a spherical cow?



WBE introduced the concept of the service volume

The service volume is the region of tissue supplied by 1 capillary

To accommodate more network infrastructure, capillary density is lower in larger animals

Thus, the service volume is larger in larger animals

 $N_{cap} \propto B \propto V^{3/4}$

$$ho_{cap} \propto N_{cap} / M \propto M^{-1/4}$$



Microprocessor Scaling: The same, but different

Network Scaling on computer chips

- a. Information can be copied (cheaply)
- b. Information can be communicated (primarily) locally
- c. Fundamental technology is still improving process size, materials, packaging, design

[Moses et al 2008, 2016]

The 1994 DEC Alpha H-tree synchronizes timing

Centralized fractal network delivers a timing signal

```
Wire lengths & radii follow 2D scaling predictions:

A 3/2

A clock-tree A chip
```

Clock speed is limited by the isochronic area

Diminishing Returns:

- Unprecedented speed (300 MHz)
- Clock ultimately consumed 40% of total power of the chip



The Intel Itanium 2 (2004) eliminates diminishing returns

Amplifiers regenerage clock signal at each branch No need for thick wires

Network footprint is proportional to chip area: $A_{clock-tree} \sim A_{chip}^{1}$

Clock speed is increased because isochronic area can be reduced

Because information can be copied \rightarrow synchronize more transistors with less power



Rent's Rule: probability of a wire inversely proportional to distance

Rents Rule was an empirical observation: the probability

of a wire crossing a boundary of area A is $\mathbf{C}_{A^{-1/2}}$

Implication: The fraction of all possible connections declines linearly with distance

Again reduces network footprint but allows a completely connected graph

If long distance communication is rare $\ \rightarrow$ indefinitely scalable network



Chip Scaling MINIMIZES Energy x Time

$$E_{sys} = E_{net} + E_{nodes}$$

$$sys = E_{net} + E_{nodes}$$

$$T_{sys} = E_{net} \times T_{sys}$$

$$E_{net} \propto N^{1-1/D_1} \sum_{i=0}^{H} \lambda^{i(1/D_1 + D_w - 1)}$$

$$T_{net} \propto N^{0} \propto \frac{l_0^2}{r_0^2} \propto N^0$$

$$E_{net} \propto N^{1-1/D_1} \sum_{i=0}^{H} \lambda^{i(1/D_1 + D_w - 1)}$$

$$T_{node} \propto N^{-1/D_1}$$

energy-time general minimization mammals $I_0 u_0 N^{2/D_r-1}$ N^{1/12} Enet Enode $u_0^{-1} N^{1-2/D_r}$ T_{net} $u_0^{-1} N^{1-2/D_r}$ N⁰ Tnode $I_0 + u_0^{-1} N^{2-2/D_r}$ $N^{1/12} + N$ $E_{\rm sys} \times I_{\rm sys}$ computers N^{1-1/D_1} $N^{1/2}$ Enet N^{1-1/D_1} $N^{1/2}$ E_{node} N⁰ N^0 T_{net} $N^{-1/2}$ N^{-1/D_1} Inode $N^{1/2} + N^{1/2}$ $N^{1-1/D_1} + N^{1-1/D_1}$ $E_{\rm sys} imes T_{\rm sys}$

 D_{l} = 2 (area-filling in 2D chips)

To minimize the $E_{sys} \times T_{sys}$: $D_r = 2$ $D_w \le 2$ Power $P = \frac{E_{\rm sys}}{T_{\rm sys}} \propto N^{1/2}$

Throughput $Tp \propto \frac{N}{T_{\rm sys}} \propto N$



Scaling Theory for Computation Predicts

Power Throughput $P = \frac{E_{\rm sys}}{T_{\rm sys}} \propto N^{1/2}$ $Tp \propto \frac{1}{T_{\rm sys}} \sim$ $-\infty N$ 10^{10} 10^{3} energy-time minimization theory: $P \propto N^{0.5}$ 10^{9} energy-time minimization theory: $T_n \propto N^1$ 10^{8} ----- best fit: $P \propto N^{0.495}$ •••••• alternative theory: $T_n \propto N^{1.5}$ throughput (DMIPS) 10^{7} 10^{2} ---- best fit: $T_p \propto N^{1.11}$ 106 power (W) 10^{5} 10 10^{4} 10^{3} 10^{2} 10 10^{-1} 10^{-1} 10^{-2} 10^{5} 10^{7} 10^{8} 10^{9} 10^{3} 10^{4} 10^{6} 10^{5} 10^{6} 107 10^{9} 10^{10} 103 10^{4} 10^{8} no. transistors (N)no. transistors (N)Hegemony of the network

Power scaling: Increasing returns Thousand-fold increase in power, Million-fold increase in MIPS



In 1970, 100 Watts powered 15 MIPS. In 2005, 6700 MIPS

> Transistors perform computations Power consumption is dominated by wires

Scaling constraints necessitate evolutionary transitions

Scaling exponents change with each transition



- Innovations in chip components mimic innovation in the evolution of bacteria
- Single-core chip scaling
 mimics unicell scaling
- Multi-core chips echo the transition to multicellularity
- Multi-agent computation as a model for scaling in social systems
- Decentralized designs dominate in the transition to sociality

Scaling constraints necessitate evolutionary transitions

Scaling exponents change with each transition



What does this mean for AI? More data, more parameters, more training \rightarrow more intelligence? Maybe, but...

- No exponential lasts forever
- No scaling lasts forever
- Expect Evolutionary transitions

Al will transition to collectives of agents that interact, iterate & evolve



AI@UNM: Empowering Innovation in New Mexico >30 UNM Faculty Actively Conducting Research in AI



Accelerate Discovery with Al

- Supercharge sustainable solutions: smart water allocation; solar energy generation, transmission & storage; forest fire mitigation and recovery; environmental monitoring & remediation
- Secure infrastructure for trustworthy AI in medicine; law & policy; education; & national security
- Advance biomedical innovations, drug discovery & personalized health care for diverse and rural populations
- Develop novel materials, computational fabrication & advanced manufacturing
- Synthesize data to detect methane leaks, volcanic & seismic & natural hazards; improve climate models
- Develop new open source foundational architectures for accessible, applicable and approachable AI
- Develop energy-aware AI based on collective intelligence and quantum technologies

Foster Collaboration & Catalyze Economic Growth

- Crystalize NM as an HPC technology hub
- Prepare NM students to thrive in the AI workforce
- Attract world class faculty
- Spur a startup ecosystem & tech transfer
- Propel NM to the forefront of 3D immersive technologies for education, workforce training, manufacturing, planning & the creative industry
- Engage diverse communities to shape AI for healthcare, law & issues affecting daily lives
- Leverage New Mexico strengths:
 - Expertise in cybersecurity and quantum computing to design next generation AI
 - Power of National Labs, abundance of landscape & natural resources & diversity and creativity of New Mexicans
 - Leadership in Hispanic Serving Institution & EPSCoR networks

TrustworthyAI @UNM

For Trusted Autonomy



- Develop tools to **detect unsafe behavior** in neural networks
- **Build safe data sets** to train neural networks to produce results within pre-specified safe bounds
- **Develop mitigation strategies** to eliminate dangerous or unexpected predictions & behaviors
- Iteratively improve neural networks by removing causes of unsafe behavior
- Build upon our toolbox for neural net verification



Mathematical Foundations

TrustworthyAI In Law & Medicine @ UNM

- Analyze accuracy and fairness in risk assessment algorithms used in criminal justice
- Build on close collaborations with UNM's Institute for Social Research, the Administrative Office of the Courts, the New Mexico Corrections Department, and City and County governments.
- Identify sources of implicit bias, anomalies and uncertainties in medical records
- **Provide trustworthy interpretations of medical reports** in close collaboration with stakeholders ensuring that AI is transparent, fair, and inclusive.





Apply trustworthy AI to protect rights & improve health



Interdisciplinary Working Group on Algorithmic Justice



Melanie Moses CS, Cris Moore UNM / SFI SFI



Kathy Powers Alfred Mathewson Poli Sci, UNM/ SFI Law, UNM



UNM

Sonia Rankin Law,

Mirta Galesic SFI



Josh Garland Arizona State Matthew Fricke Gabe Sanchez CS, UNM Poli Sci, UNM

Tina Eliassi-Rad CS, NEU /SFI

Mahzarin Banaji Psych, Harvard/SFI

Trilce Estrada CS, UNM

Nadiyah Humber Law, UConn



nature > nature reviews electrical engineering > comment > article

Comment | Published: 12 April 2024

Medical artificial intelligence should do no harm

Melanie E. Moses [™] & Sonia M. Gipson Rankin

Nature Reviews Electrical Engineering (2024) Cite this article

40 Accesses | 4 Altmetric | Metrics

Bias and distrust in medicine have been perpetuated by the misuse of medical equations, algorithms and devices. Artificial intelligence (AI) can exacerbate these problems. However, AI also has potential to detect, mitigate and remedy the harmful effects of bias to build trust and improve healthcare for everyone.

AI @UNM

Presenters

Kent Kiehl Professor Department of Psychology Presentation Title: "Machine learning of brain science in legal contexts"

Sarah Dreier

Assistant Professor Department of Political Science <u>Presentation Title</u>: "Al and text quality: Identifying sources of training biases and imperfect classification outcomes"

Jessica Feezell

Associate Professor Department of Political Science <u>Presentation Title</u>: "Algorithm-driven News and Political Polarization in the U.S."

Cristyn Elder Associate Professor Department of English Language & Literature

lain Thompson

Professor Department of Philosophy Presentation Title: "What Is Called Thinking in the Age of AI?"

Brandon Schmandt

Professor

Department of Earth & Planetary Sciences

<u>Presentation Title</u>: "Classification of seismic signals with machine learning: applications to environmental change, earthquakes in New Mexico, and national security"

Joseph Galewsky

Professor

Department of Earth & Planetary Sciences

<u>Presentation Title</u>: "Cloudy with a Chance of Bits: Using AI to improve our understanding of water vapor's role in the climate system"

New Mexico AI Consortium

- Los Alamos, Sandia, UNM, NMSU, NMC and others are forming the NMAIC
- Goals:
 - Establish long-term collaboration in AI & ML & HPC
 - Position NM to be a national/international leader in AI for science, natl security & public good
 - Develop a thriving NM research and innovation ecosystem and economy
 - Advance fundamental and applied research in Al
 - Engage & support the diverse communities of New Mexico in shaping an AI-enhanced future

- Please contact me <u>melaniem@unm.edu</u> about ongoing or potential collaborations within the labs and NM Universities and colleges.
- Look for an announcement later this summer

We thought we might start a chain reaction that might destroy the entire world.

- J Robert Oppenheimer

Atomic bomb Trinity test. UNIVERSAL HISTORY ARCHIVE/UIG VIA GETTY IMAGES

Powerful new technologies reshape the world...

but not in the ways we initially fear or expect.



John Powell UC Berkeley Law Professor

Kennedy told us that we ALL had to contribute for America to reach the moon

Al is trained on data from (almost) all of humanity

All of humanity needs to be engaged to shape the future of Al





How AI works

https://arstechnica.com/science/2023/07/a-jargon-free-explanation-of-how-ai-large-languagemodels-work/

Politico: legislation for transparency and audits

https://www.politico.com/news/magazine/2024/04/30/ai-legislation-states-mistake-00155006

Complexity & AI

https://www.newamerica.org/planetary-politics/policy-papers/how-complexity-thinking-can-hel p-the-world-navigate-ai/

Moses & Gipson Rankin article on Trustworthy AI in Medicine https://www.nature.com/articles/s44287-024-00049-2

- AI and Algorithmic Justice Resources @UNM & SFI
- https://airesources.unm.edu/students.html
- https://algorithmicjustice.cs.unm.edu/index.html
- https://www.santafe.edu/research/projects/algorithmic-justice
- melaniem@unm.edu https://moseslab.cs.unm.edu



Questions?

Even deep fakes have positive uses

An AI Salvador Dalí will answer any question when called on his famous 'lobster phone'

UPDATED APRIL 21, 2024 - 1:08 PM ET @





Ask Dalf at the Dalf Museum in St. Petersburg, Fla., allows visitors talk to the famous surrealist artist via an Algenerated version of his volce. Marin Pagb Ludvigsen/Goodby Silverstein & Partners

https://www.npr.org/2024/04/21/1246117954/salvador-dali-artificial-intelligence-lobster-telephone-museum

