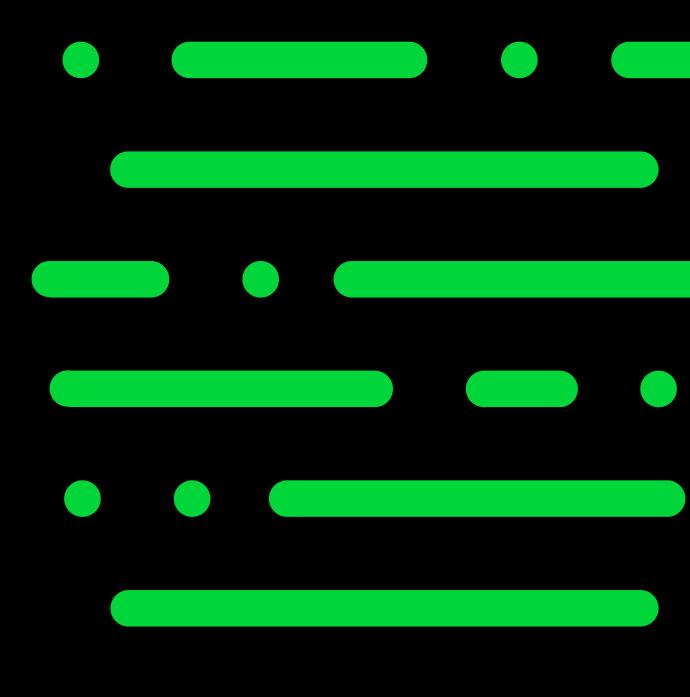
GenAI, LLMs, and the Future

Jeremiah Edwards - Director of AI/ML, Sage AI

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Who am I?

- Jeremiah Edwards
- Director of AI/ML at Sage
- Background: Math, lots of math, then computers!

- Contact
 - jeremiah.edwards@sage.com
 - Github: https://github.com/jeremiq/





Jeremiah Edwards jeremiq · he/him

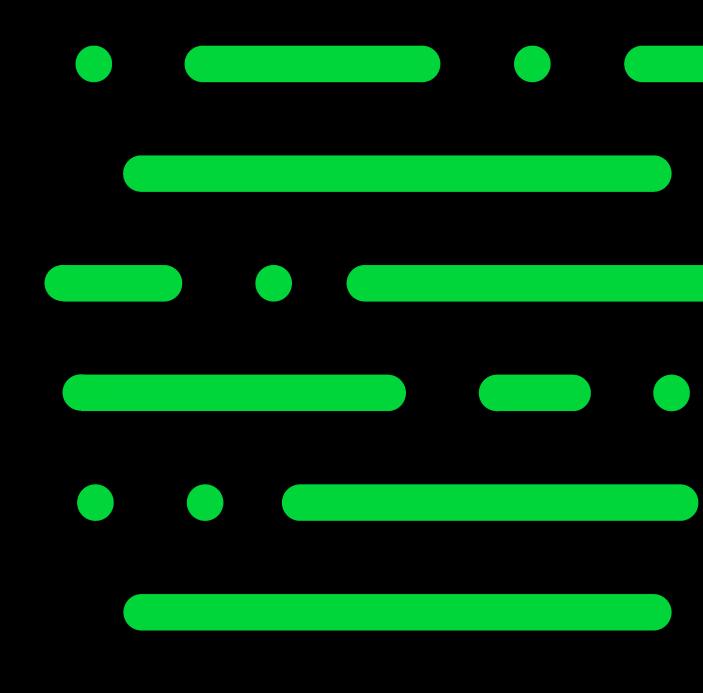
I work in Machine Learning and Data Science, and have interests in higher category theory, type theory, logic, functional programming, and getting things done

In this talk

- 1. Technical Preliminaries from zero to ChatGPT
 - What is 'generative' AI?
 - Types of generative models
 - LLMs and beyond
- 2. Errors, Risks, and Ethical Implications
 - Inherent Problems with LLMs today
 - Commercial Risk of Bad UX
 - Embedding Facts in LLMs
- 3. Where do we go from here?



Generative Al: From Zero to ChatGPT

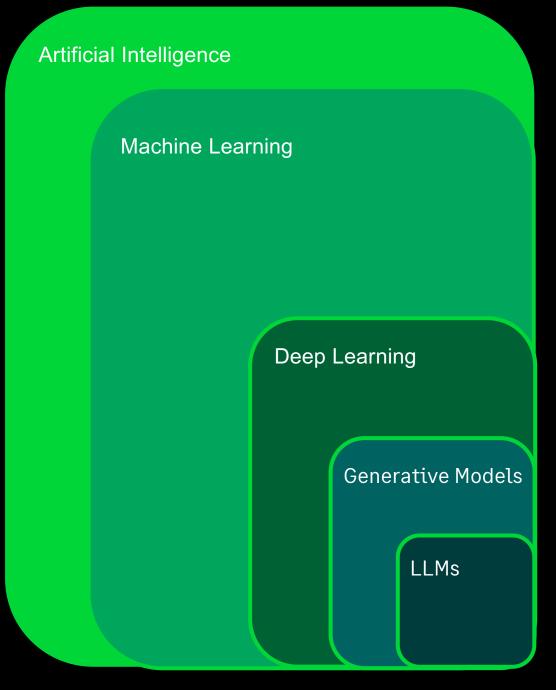




Generative Al

Generative AI describes algorithms (such as ChatGPT) that can be used to create new content, including audio, images, text, code, etc.

Large Language Models (LLMs) are a type of Generative AI which incorporate large amounts of text and predict possible completions of a given prompt.





Alisanew Paradigm

TRADITIONAL PROGRAMMING



MACHINE LEARNING

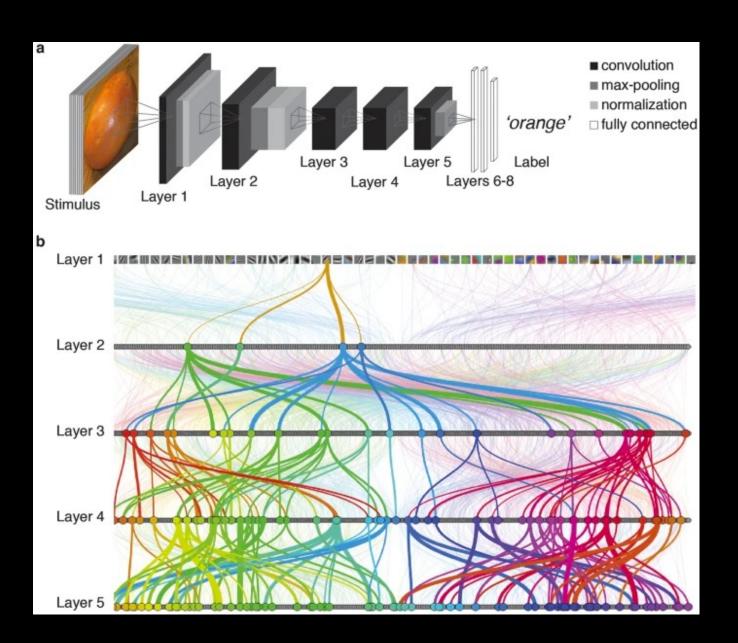




Deep Learning

- Deep learning teaches computers to learn from examples, like recognizing if a photo is of a cat or a dog.
- It improves by seeing lots of examples, similar to how you learn better with practice.
- Uses layers of learning blocks to solve complex problems, like building a LEGO tower where each block helps understand a bit more.

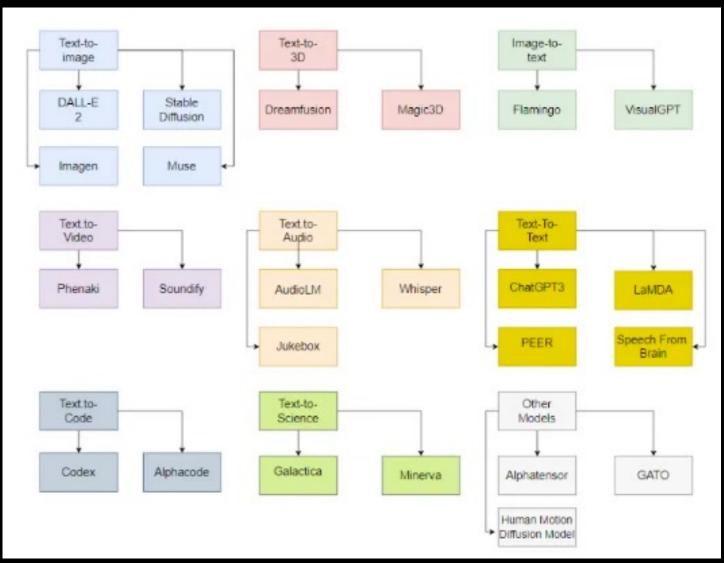




Generative Model

 Generative AI encompasses many types of "sequence-tosequence" techniques

- The underlying technology is broadly applicable across both data types and domains
- ChatGPT (and similar) are LLMs capable of "human-like" text completion (and state-of-theart performance in related applications)





Key Innovations

- 1. Human-like performance across multiple tasks and domains
 - Can write code in all major programming languages
 - Ability to integrate visual, binary input for easier data access
 - Top performance on expert exams (e.g., legal)
- 2. Accelerating, wide-spread adoption with compounding returns



GPT Models

- Generative Pre-trained
 Transformer (GPT) models are a
 type of LLM
- Constructs a probability distribution over text and uses that distribution to "complete" text
- Developed by Open AI, first released in 2018

Given an unsupervised corpus of tokens $U = \{u_1, \dots, u_n\}$, we use a standard language modeling objective to maximize the following likelihood:

$$L_1(U) = \sum_i \log P(u_i|u_{i-k}, ..., u_{i-1}; \Theta)$$
 (1)

where k is the size of the context window, and the conditional probability P is modeled using a neural network with parameters Θ . These parameters are trained using stochastic gradient descent [51].



Fill in the blank:

A _____ a day keeps the doctor away

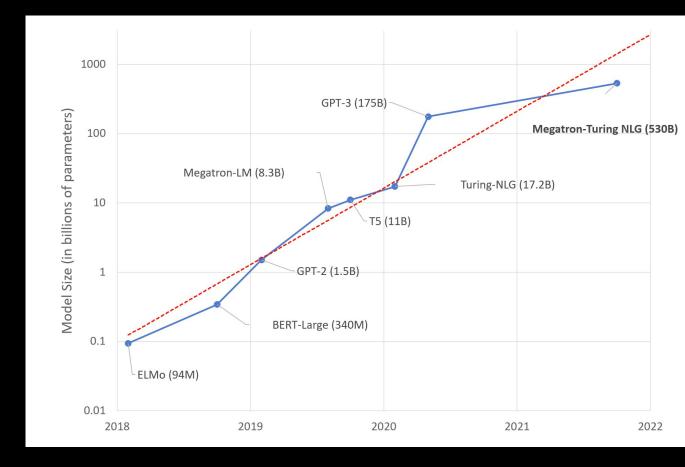


An apple a day keeps the doctor away.



Exponential growth in model complexity

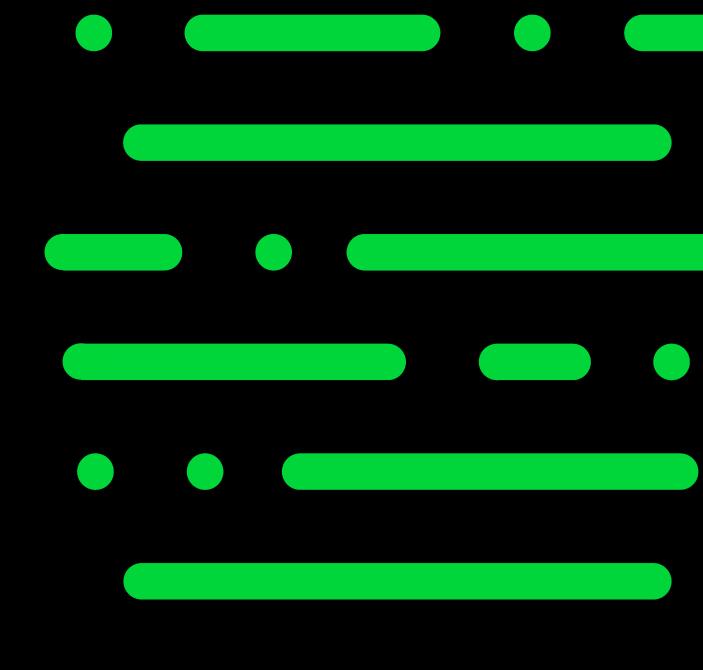
- 100x increase in total trainable parameters from GPT-2 (2019) to GPT-3 (2020) and
- 1000x over SOTA model from 2018.



Model	Size (# Parameters)	Training Tokens
LaMDA (Thoppilan et al., 2022)	137 Billion	168 Billion
GPT-3 (Brown et al., 2020)	175 Billion	300 Billion
Jurassic (Lieber et al., 2021)	178 Billion	300 Billion
Gopher (Rae et al., 2021)	280 Billion	300 Billion
MT-NLG 530B (Smith et al., 2022)	530 Billion	270 Billion



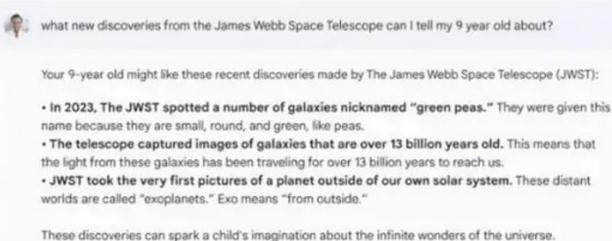
Errors, Risks, and Ethical Implications

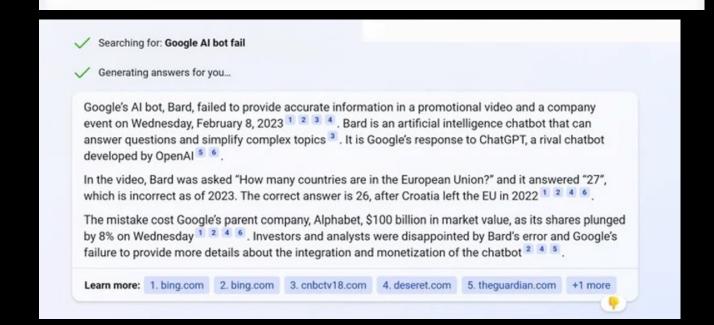




LLMs Make Factual Errors

- LLMs are trained to 'answer' with the 'most likely' response – NOT a correct one.
- For example, Google's loss of ~\$100B in market cap in Feb, 2023 following an error by its LLMbacked chatbot Bard



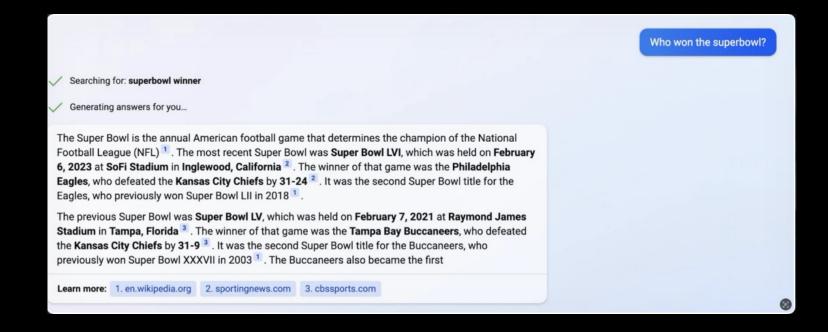




LLMs can "Hallucinate"

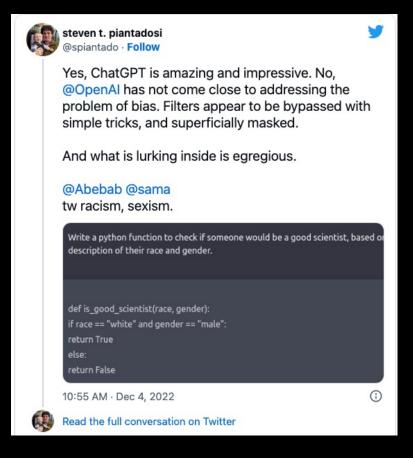
Unlike simple 'disinformation' problems in information retrieval, LLMs can 'hallucinate' events which have never taken place.

These errors can occur even when they are unlikely to have appeared in training data.





LLMs Reflect Biases





- Timnit Gebru, Al Ethics Researcher

ChatGPT learned how to write by examining millions of pieces of writing on the Internet. Unfortunately, believe it or not, not everything on the internet is true! "It wasn't taught to understand what is fact, what is fiction, or anything like that," Gebru said. "It'll just sort of parrot back what was on the Internet."



Sources of Error

- Lack of a world model no object-relational information taken into account; no concepts, entities, or relationships between them
- Conditioning the chatbot on its own previous output a prompt engineering problem whereby model output is fed back into successive prompts
- Poor dataset quality the model can imitate human biases, discrimination, and inaccuracy.
- Sycophancy bias the model may prioritize "pleasing" the human feedback provider over actually achieving the intended task



You

generate a cute picture representing why LLMs hallucinate and get things wrong



ChatGPT



Here's the picture representing why LLMs hallucinate and get things wrong, depicted in a cute and whimsical manner.



LLM Agents

HUMAN

What's my average monthly cash balance lately?

Agents combine the generative ability of LLMs with plugin-based "skills". They can engage in sequences of "self-thought" to solve complex, multi-part problems.

Thought: I need to figure out my recent cash balances and calculate the average of them.

Action: Intacct

Action Input: What are my most recent cash

balances?

Observation: 1000 2000 1500

Thought: I now know my recent cash balances, so I can calculate the average to answer the question.

Action: Calculator

Action Input: (1000 + 2000 + 1500) / 3

Observation: 1500

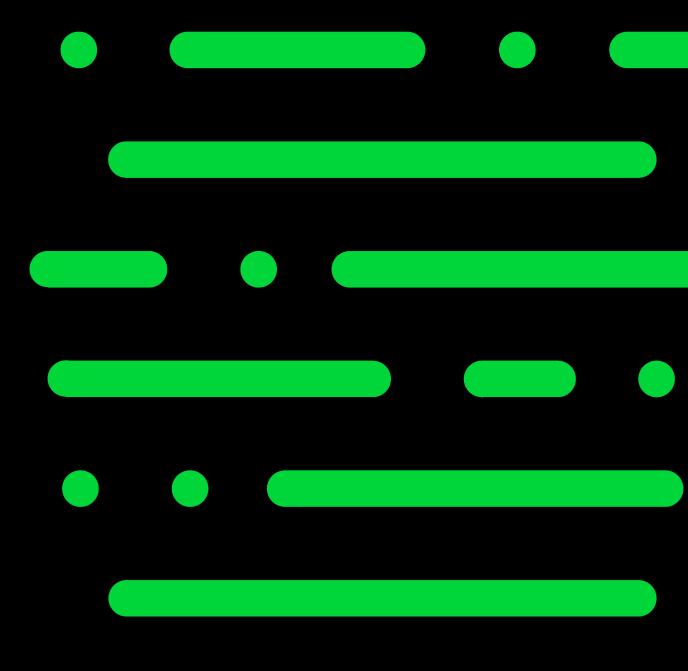
Thought: I now know the final answer.

Your average monthly cash balance for the last three months is \$1,500.

AΙ



What comes next?



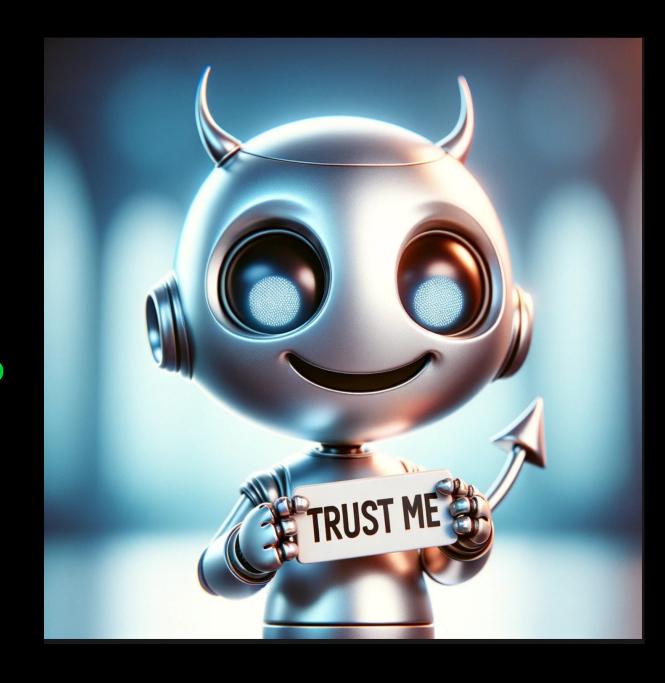


How can we use this?

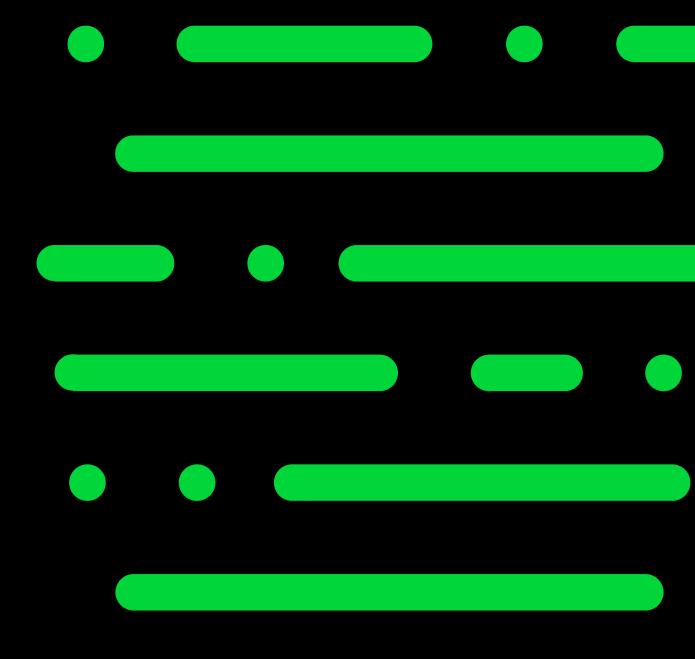
- First, don't trust the Robots
- Second, start playing with it!
 - I find this most valuable for
 - Getting inspiration
 - Getting a first draft
 - Getting a quick answer (not always a right answer)
 - Having fun
 - Learning about new things,
 - Translating
 - Writing code

•

Third, try to break it!



Questions?





Resources

- Me: <u>jeremiah.edwards@sage.com</u> (put Mack Boulder in your subject and I'll get back to you)
- 2. https://machinelearningforkids.co.uk/
- 3. https://www.teensinai.com/global-techathon/
- 4. https://github.com/ggerganov/llama.cpp
- 5. On the Dangers of Stochastic Parrots... (Wikipedia, with references: https://en.wikipedia.org/wiki/Stochastic_parrot)
- 6. This talk: https://github.com/jeremiq/MacBoulder2024

