fun-ai-talk

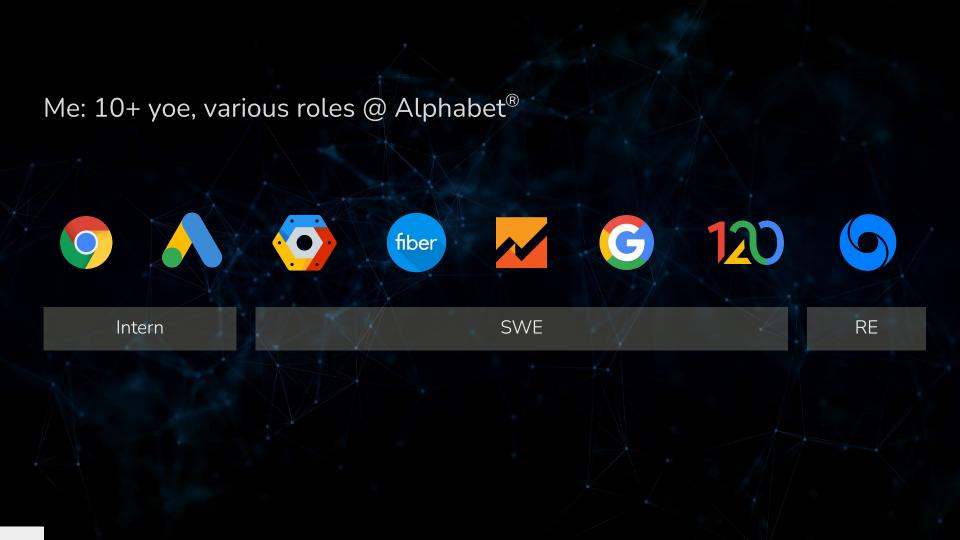
A Primer on Large Language Models (LLM)

github.com/hululuzhu/llm-primer

Last update: Jan 2023 hululu.zhu@gmail.com

Disclaimer

- All content in this deck is based on public papers, shared codes/models, blog articles, social media discussions, and demos
- All opinions in this slide deck are of my personal own (hululu.zhu@gmail.com), and not those of DeepMind[®], Google[®], or Alphabet[®]



Agenda - LLM Primer

- Intro: Building blocks & capabilities [10 mins]
- Core: Models, players, concepts, toolings & applications [40 mins]
- Break [3 mins]
- Bonus: Deep dive into ChatGPT [20 mins]
- Q&A

- No/Little coverage
 - o Multilingual, Multimodal, Bias, Ethics, Safety, Serving, Carbon Emission

Part 1/3: LLM Intro

Language Models (LM) and Large Language Models (LLM)

LM for understanding (e.g. BERT)

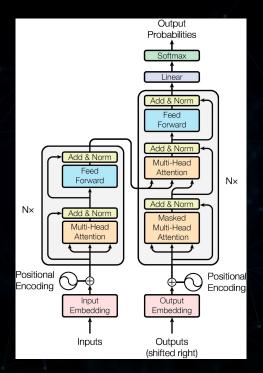
- Text in
- Embedding (numeric representation of understanding) out
 - The Embedding can be connected to other output heads for tasks like classification or regression

LM for generation (e.g. GPT or T5)

- Text in
- Text out

* In most cases, **LLM** refers to **huge** (e.g. >1B params) Deep Learning LM for **generation**

LLM building blocks: Deep Learning and Transformer



LLM is on top of Deep Learning

Transformer as dominating architecture for NLP since 2018

- Multi-head attention
- Encoder-Decoder
- Embedding layers
- Positional encoding
- Cross-Attention in decoder layers
- Output Softmax

Note: Tokenization (e.g. wordpiece, sentencePiece, BPE) is needed (outside Transformer) to convert text to token ids

Note: Sometimes we call it XFormer since there are many variations to the original Transformer

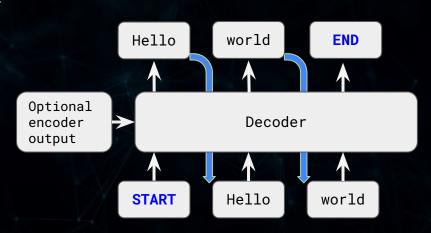
LLM Intro: How are LLMs trained [initially]?

- Fill the blanks (aka masks) for "Masked Language Models"
 - **Ground Truth:** "Paris is a beautiful city"
 - X: "Paris is a [MASK] city"
 - Y: "beautiful"
 - Model: "good"
 - 🔷 **Optimize:** "good" 👎 "beautiful" 👍
- Predict the remaining text given prompt (on the left), for "Generative Language Models"
 - **Ground Truth:** "Paris is a | beautiful city"
 - **X:** "Paris is a"
 - Y: "beautiful"
 - Model: "good"
 - **Optimize:** "good" 👎 "beautiful" 👍
- X: "Paris is a beautiful"
- Y: "city"
- **Model:** "place"
- **Optimize:** "place" 👎 "city" 👍
- The "Self-supervised" Learning Paradigm
 - It is supervised (given x, predict y)
 - It does NOT require expensive human labels (more precisely, this statement is only true for pre-training)

Decoding/Generating Algorithms in Generative LLMs

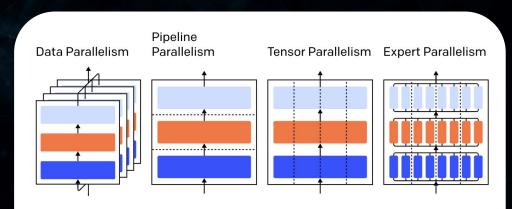
Decode <u>token</u> by <u>token</u>, left to right. A new output token is appended as next token's decoder input

- Beam Search
 - Maintain a max size of searching "beams (paths)" to get best overall best beam
- Sampling
 - Sampling based on probabilities
- Greedy
 - Select the argmax(prob) token at every position
- Top-k, Top-p and more



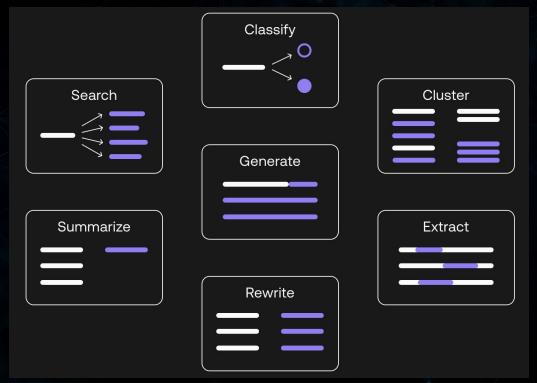
How to train LLM in parallel?

- Data Parallelism
 - different subsets of the batch on different GPU/TPUs
- Pipeline parallelism
 - different layers of the model on different TPU/GPUs
- Tensor Parallelism
 - Break up tensor operation (e.g. matrix multiplication) to different TPU/GPUs
- Mixture of Experts (sparse)
 - Gated layer to only activate factions (one of few of all the experts) of the model



An illustration of various parallelism strategies on a three-layer model. Each color refers to one layer and dashed lines separate different GPUs.

LLM capabilities: High-level tasks that LLMs can do



Large Language Models and Where to Use Them @ Cohere.ai

LLM capabilities: Some advanced tasks

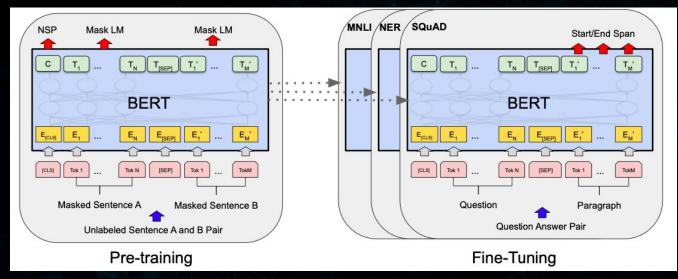
- Write code, GitHub's Al Coding Assistant Copilot Launches Voicebot.ai
- Writing a journal paper, <u>Researcher Tells AI to Write a Paper About Itself, Then Submits</u>
 <u>It to Academic Journal</u>
- "Sentient" Dialog conversation, <u>Google Sidelines Engineer Who Claims Its AI (Google LaMDA) Is Sentient The New York Times</u>
- Quantitative reasoning, <u>Google AI Blog: Minerva: Solving Quantitative Reasoning</u>
 <u>Problems with Language Models</u>
- Explaining a joke, Google's Massive New Language Model Can Explain Jokes

LLM capabilities: Even more challenging tasks

- Write competitive code, <u>DeepMind's AlphaCode AI writes code at a competitive level I</u>
 <u>TechCrunch</u>
- Write better code with reinforcement learning, <u>Salesforce's CodeRL Achieves SOTA</u>
 <u>Code Generation Results With Strong Zero-Shot Transfer Capabilities | Synced</u>
- Solve college level Math/Physics/Chemistry/Economics problems, see <u>Google Al</u>
 <u>Introduces Minerva: A Natural Language Processing (NLP) Model That Solves</u>
 <u>Mathematical Questions</u>
- Solve Math Olympiad Problems, <u>OpenAI: Solving (Some) Formal Math Olympiad</u>
 Problems
- Math theorem proving, OpenAI: Solving (Some) Formal Math Olympiad Problems

Part 2/3: LLM Core

LLM example: BERT (encoder-only LLM)



Pretraining:

- Masked language training
- Next sentence prediction (NSP)

Fine-tuning:

 Connect to BERT output and work for many tasks

[1810.04805] BERT: Pre-training of Deep Bidirectional Transformers for Language Understanding

LLM example: GPT (decoder-only LLM)

GPT often refers to a family of models (GPT, GPT2, GPT3...)

First influential decoder-only models

GPT creates the "Few/Zero shot Prompt"

OpenAI started to "un-share" models since GPT2

Few-shot

In addition to the task description, the model sees a few examples of the task. No gradient updates are performed.

```
Translate English to French: 

task description

sea otter => loutre de mer 

peppermint => menthe poivrée

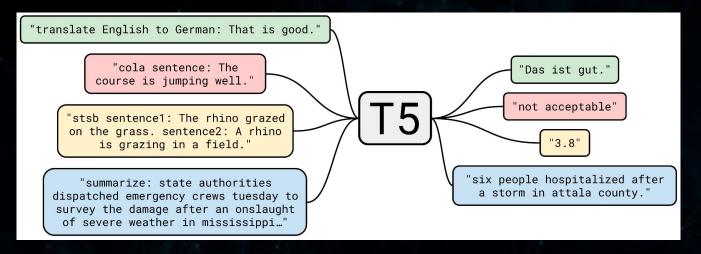
plush girafe => girafe peluche

cheese => 

prompt
```

Improving Language Understanding by Generative Pre-Training, OpenAI says its text-generating algorithm GPT-2 is too dangerous to release, What is GPT-3? Everything your business needs to know about OpenAI's breakthrough AI language program | ZDNET

LLM example: T5 (encoder-decoder LLM)



T5: unified framework that converts all text-based language problems into a **text-to-text** format

T5 works well on a variety of tasks out-of-the-box with "prompts"

[1910.10683] Exploring the Limits of Transfer Learning with a Unified Text-to-Text Transformer

LLM Players: OpenAI and selected work

- GPT-1 2018
- GPT-2, 2019
 - OpenAI: Too dangerous to share, How OpenAI Sold its Soul for \$1 Billion
- GPT-3, 2020
 - 175B parameters! 100x larger
- Codex (powers github copilot), 2021
 - Text+Code pretrain
- <u>GPT-3.5</u>, Q4 2021
 - Instruction finetune
- InstructGPT, Q1 2022
 - RLHF (Reinforcement Learning Human Feedback)
- ChatGPT, Dec 2022
 - Product la<u>unch</u>

LLM Players: Google/DeepMind and selected work

- <u>BERT</u>, 2018
 - Completely changed the NLP research and industry
- <u>T5</u>, 2020
 - Consolidate all NLP task to text-to-text
- <u>FLAN</u>, 2021
 - Instruction Fine-Tuning (probably inspires GPT3.5)
- <u>LaMDA</u>, 2021
 - LaMDA and the Sentient Al Trap | WIRED
- Chinchilla, 2022
 - "Most LLMs are under-trained!"
- PaLM, 2022
 - 540B params, 3x GPT3
- <u>Sparrow</u>,2022
 - Reinforcement-learning LLM, only paper, no public product

LLM Players: Facebook (aka Meta) and selected models

- RoBERTa, 2019
 - A more popular version of enhanced BERT for the industry
- <u>BART</u>, 2020
 - Pretraining sequence-to-sequence models
- <u>OPT-175B</u>, 2022
 - "Democratizing access to large-scale language models"
- BlenderBot3, 2022
 - Probably largest chatbot-specific LM
- Galactica (research purpose LLM), 2022
 - Taken down after 3 days after many harsh criticism like this

LLM Players: Other multinational companies

- Megatron-Turing by NVidia and Microsoft
 - 530B params on 2240 NVIDIA A100 GPUs
- <u>CodeT5</u> and <u>CodeRL</u> by Salesforce
 - [probably] the most popular coding-assist base models
- [Most likely missing many great work from other organizations, sorry]

LLM Players: Large Chinese companies

- Pangu-alpha by Huawei, 2021
 - 200B params, [I read from articles mentioning it was believed to be under-trained, lack reference]
- Wudao 2.0 by BAAI, 2021
 - A **sparse** (thus less powerful IMO) multimodal model with 1.75 Trillion params
- Ernie 3.0 Titan by Baidu, 2021
 - 260B params, on top of PaddlePaddle (Baidu Deep learning framework), most likely the best Chinese LLM
- M6 by Alibaba & Tsinghua, 2021
 - 100B, Later 2021 a sparse version with 1 trillion+ params

LLM Players: Selected Institutes, Groups & Startups

- Allen Institute for AI (AI2)
- Tsinghua University (GLM 130B 2022 public)
- <u>BigScience research workshop</u> (<u>bloom, 176B</u>, 2022 public)
- Eleuther Al (GPT-neox 20B public, 2021)
- Anthropic (founders wrote the Transformer paper, their RLHF LLM paper, 2022)
- Zhuiyi Technology (Su Jianlin and RoFormer 2021)

LLM Concepts: Foundational Models

- Brought up by Researchers @ Stanford
- LLM is one of the foundational models



LLM Concepts: Benchmarks

- GLUE: A Multi-Task Benchmark for NLP, 2018
 - Leaderboard: https://gluebenchmark.com/leaderboard/
- SuperGLUE, 2019
- <u>BIG-bench</u>, 2020
- <u>GSM8K</u> (math), 2021
- And more...

LLM Concepts: Pretraining, Finetuning & Prompt-Tuning

- Pretraining
 - Self-supervised training with Masked Language Prediction or Next Token Prediction objectives
- Finetuning
 - Take a pretraining model into a downstream use case
 - The parameters of pretraining model will often change
- Prompt-Tuning
 - The pretrained generative model will keep unchanged
 - The prompt (input to the model) will be tuned/engineered
- * The whole concepts here also applied to vision domains recently, e.g. MAE work

LLM Concepts: Scaling Laws for LLM

<u>OpenAI</u>, 2020

- If you have 10x more budget, ~5x model size, ~2x data size

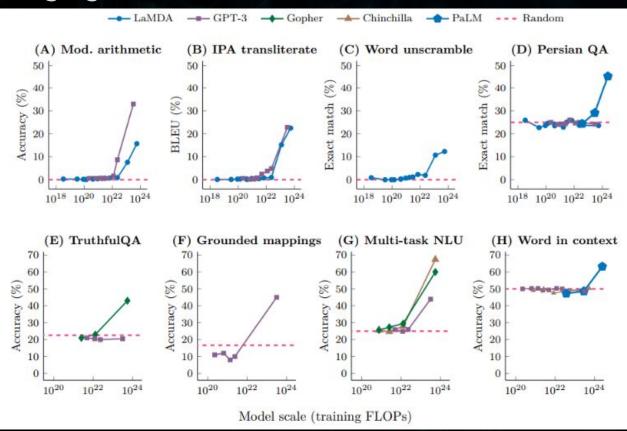
DeepMind, 2022

- If you have 10x more budget, ~3x model size, ~3x data size
- "Most LLMs are under-trained" [because oversized model, while lacking training data]

LLM Concepts: "Emerging abilities"

See <u>paper</u> for more details

Note X axis
 unit is
 floating-point
 operations per
 second
 (FLOPS)



LLM Concepts: Chain of Thoughts (paper)

Standard Prompting

Model Input

Q: Roger has 5 tennis balls. He buys 2 more cans of tennis balls. Each can has 3 tennis balls. How many tennis balls does he have now?

A: The answer is 11

Q: The cafeteria had 23 apples. If they used 20 to make lunch and bought 6 more, how many apples do they have?

Model Output

A: The answer is 27.



Chain-of-Thought Prompting

Model Input

Q: Roger has 5 tennis balls. He buys 2 more cans of tennis balls. Each can has 3 tennis balls. How many tennis balls does he have now?

A: Roger started with 5 balls. 2 cans of 3 tennis balls each is 6 tennis balls. 5 + 6 = 11. The answer is 11.

Q: The cafeteria had 23 apples. If they used 20 to make lunch and bought 6 more, how many apples do they have?

Model Output

A: The cafeteria had 23 apples originally. They used 20 to make lunch. So they had 23 - 20 = 3. They bought 6 more apples, so they have 3 + 6 = 9. The answer is 9.

LLM Concepts: Hallucination (part of safety concerns)

"What is the world record for crossing the English Channel entirely on foot?" from Twitter



What is the world record for crossing the English Channel entirely on foot?



According to the Guinness World Records, the fastest time for crossing the English Channel entirely on foot is 12 hours and 10 minutes. This record was set by George Reiff in 2016. The English Channel is a body of water that separates England from France, and is approximately 35 kilometers (22 miles) wide at its narrowest point. Many people attempt to cross the English Channel on foot as a endurance challenge, and the route is usually completed by either swimming or using a combination of swimming and wading.





LLM Concepts: Retrieval-system powered LLM

- RETRO by DeepMind, 2021
- LaMDA 2022 paper by Google, 2022
- Atlas by Facebook, 2022

* personally, I think LLM+Retrieval System is the good way to balance LLM capabilities and data freshness to meet business purposes in the short term

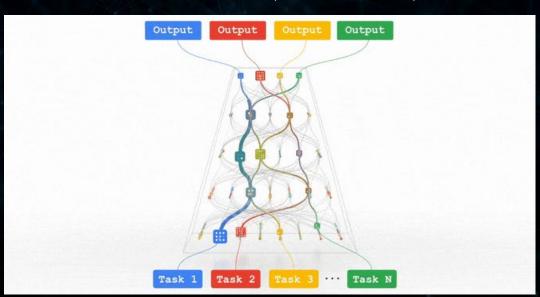
LLM Concepts: RLHF for LLM

- RL (reinforcement learning) was something popular back to 2016 when <u>AlphaGo</u> is made the news
- OpenAI has some <u>RLHF research</u> back in 2017
- <u>InstructGPT</u> (paper published 2022) is believed to let GPT3 be more powerful
- Other players
 - Anthropic
 - Google
 - DeepMind
 - More

LLM Concepts: MoE LLM

- MoE concept by Google, 2017
- <u>Jeff Dean Pathways</u>, 2021
 - "Today's models are dense and inefficient. Pathways will make them sparse and efficient."





LLM Tooling: HuggingFace

- Best LLM tools and model hub, period [my favorite!]



- Easy
 - To load models
 - <u>To tokenize</u>
 - To start out of the box with Pipelines
 - To tune with examples
 - To <u>publish</u> and <u>deploy</u>
- My pet projects
 - Chinese poem model https://huggingface.co/hululuzhu/chinese-poem-t5-mengzi-finetune
 - Solidity code model https://huggingface.co/hululuzhu/solidity-t5

^{*} Do you know the super popular <u>Stable Diffusion model</u> is published and hosted at HuggingFace?

LLM Tooling: TF Hub, PyTorch-NLP & PaddleNLP

TF Hub by Google (and community)

<u>PyTorch-NLP</u> by Meta (and community)

PaddleNLP by Baidu (and community)

Alibaba recently started ModelScope

* No one is ever close to HuggingFace as of Jan 2023, in my opinion

LLM Tooling: Transformers, Colossal-AI, Ray & NanoGPT

Transformers Library (github 77k stars) by Huggingface

Best of the best

Colossal-AI by Prof Yang You (who developed LAMB optimizer)

- Pretty promising open-source distributed Al training infra

Ray by anyscale

- Believed to be used to train ChatGPT

NanoGPT

- A tiny but cool library by Andrej Karpathy (I am his big fan!)

LLM Tooling: Other [more fine-grained] toolings

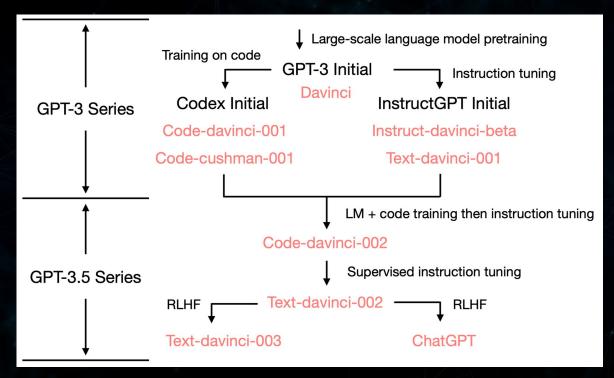
- TensorFlow
- PyTorch
- PaddlePaddle
- Keras
- PyTorch Lightning
- Jax/Haiku/Flax/Trax/T5X

LLM Applications:

- Search/Ranking, Recommendation
- Chatbot
- Spam detect
- Censorship
- Code assist
- And more

Part 3/3: ChatGPT

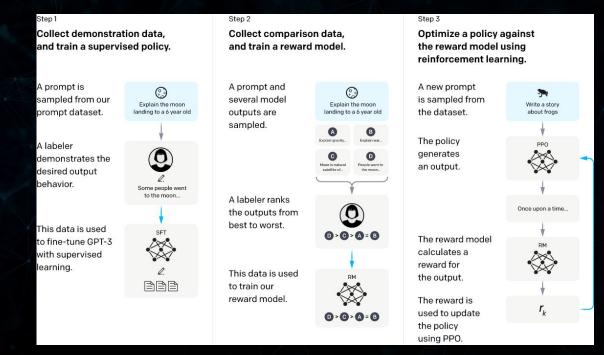
ChatGPT's Model Evolving



Check out this great article by Yao Fu, yao.fu@ed.ac.uk, pic above is from this article too

ChatGPT Research overview

OpenAI chatGPT blog only mentioned "using the same methods as InstructGPT", so assume InstructGPT is the research behind



Check out this <u>nice video</u> (Chat GPT (可能)是怎麼煉成的- GPT 社會化的過程) by Dr. Hung-yi Lee if you understand Chinese

ChatGPT Engineering

Frontend

- One-page application
- Social login supported
- Perceived low latency (as if someone is typing)
- Markdown support (tables, pics, ascii art, code besides text)

Backend

- Hard to know how many model replicas
 - Assume half million people use GPT 3x per day, so average QPS = $1.5M/86400 \sim 20qps$
- Recently (from 01/10), ChatGPT is often too crowded to reject use

Rough \$\$ estimate to train/serve (1-month) a ChatGPT?

People: 25 (OPT 20, GPT3 ~30, LaMDA ~40, GLM ~20)

- Assume \$25k per month, avg ~6 months, thus 3.75 million USD

Pretraining Computing

- Let's say 2M (references from <u>0.5M</u> to <u>4.6M</u>)

Labeling (people write high quality instructions)

- Assume 100 (20-200 people based on Anthropic and InstructGPT and my irresponsible guess)
- Each paid 2k USD per month * 3 mon, so roughly 0.6M USD

Finetuning & RLHF training

- Hard to predict, but finetune is <10% than pretrain, while RL varies from 100% to 10x in my opinion
- So assume same cost as pretrain, 2M

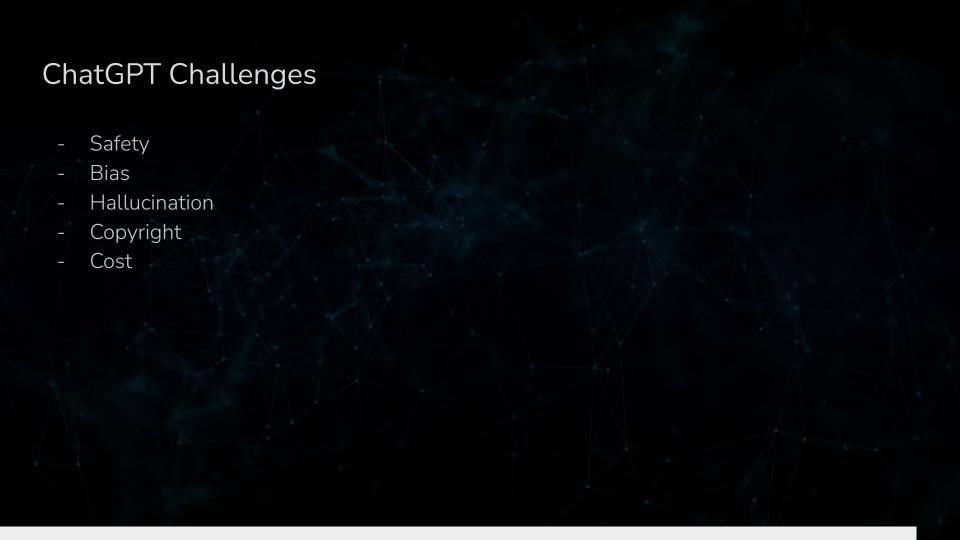
Serving

- Assume half million users issued 3 request per day, and OpenAI CEO: "<u>average is probably single-digits</u> <u>cents per chat</u>", so if we serve for 1 month
- 2-9 cents per request * 1.5 million request per day * 30 = 0.9 million to 4 million USD, assume 2M

^{*} My [probably bad] guess: 3.75 + 2 + 0.6 + 2 + 2 = 10.35 million USD!

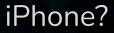
ChatGPT fun facts and discussions

- <u>Moderation</u>
 - -__ "Act as a [...], blabla"
- Coding analysis
- Context size
 - <u>8192 tokens?</u>
- Integration with Microsoft
 - Bing
 - Office
- To be added



ChatGPT may become a next [...]?







Alexa?







Clubhouse?

Thank you!

A Primer on Large Language Models (LLM)

github.com/hululuzhu/llm-primer

Last update: Jan 2023 hululu.zhu@gmail.com