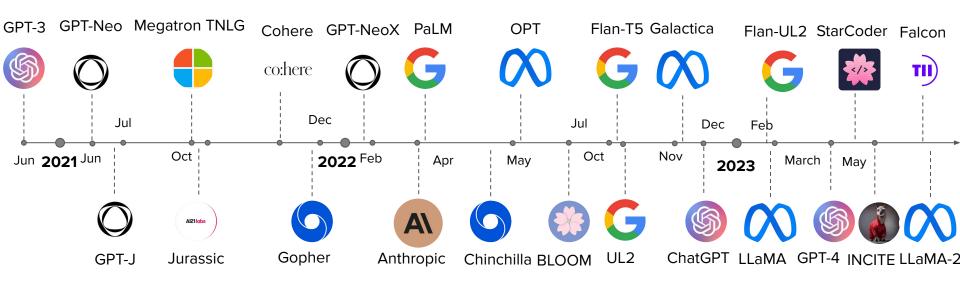


Taming the Wild West of LLMs

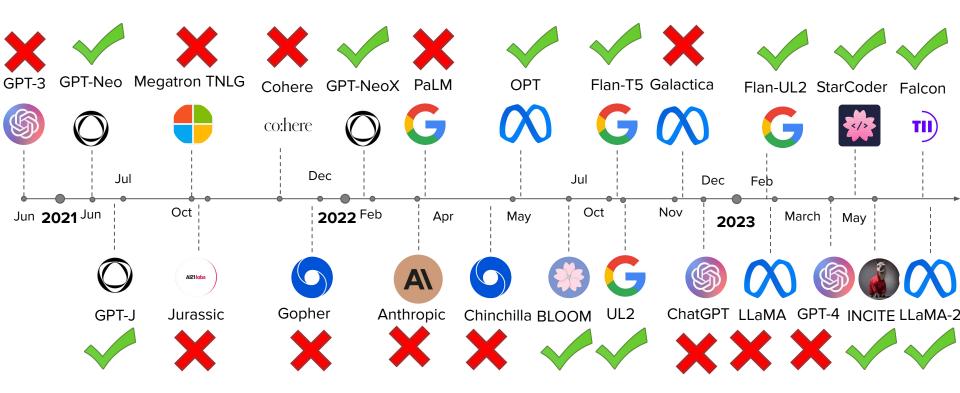
Nazneen Rajani | Research Lead @ Hugging Face | nazneen@hf.co | @nazneenrajani

Text-to-Text Foundation Models since GPT3



*only LLMs with >1B parameters & EN as the main training language are shown. Comprehensive list: https://crfm.stanford.edu/helm/v1.0/?models=1

Text-to-Text Foundation Models since GPT3



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Model Access





Limited access



Closed access



Open Access Models

Model components are publicly available:

- Open source **code**
- Training **data**
 - Sources and their distribution
 - Data preprocessing and curation steps
- Model weights
- Paper or blog summarizing
 - Architecture and training details
 - Evaluation results
 - Adaptation to the model
 - Safety filters
 - Training with human feedback



Allows reproducing results and replicating parts of the model

Enable auditing and conducting risk analysis

Serves as a research artifact

Enables interpreting model output



Only research paper or blog is available and may include overview of

- Training data
- Architecture and training details (including infrastructure)
- Evaluation results
- Adaptation to the model
 - Safety filters
 - Training with human feedback



Safety concerns

Competitive advantage

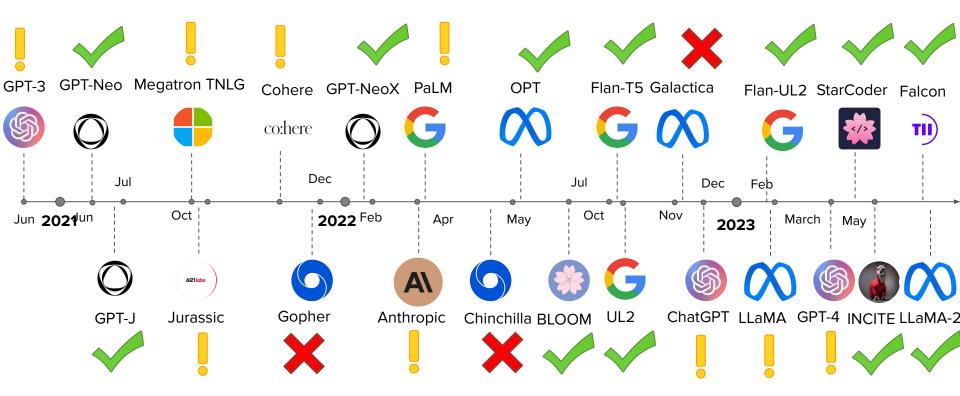
Expensive to setup guardrails for safe access



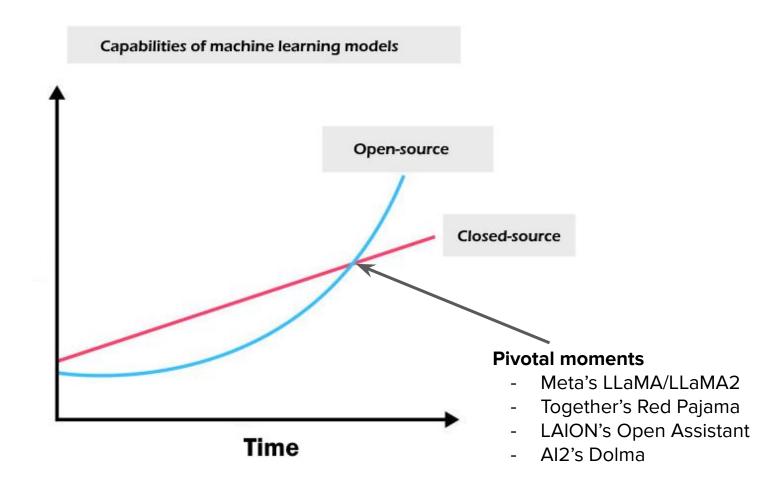
Available for use via:

- API
- Call for research proposals

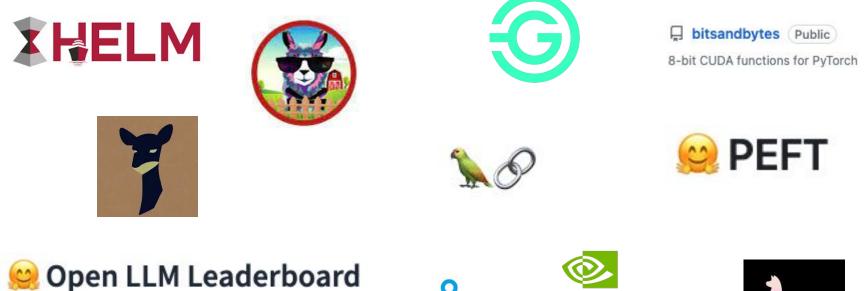
Text-to-Text Foundation Models since GPT3



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Tooling for Open Source LLMs

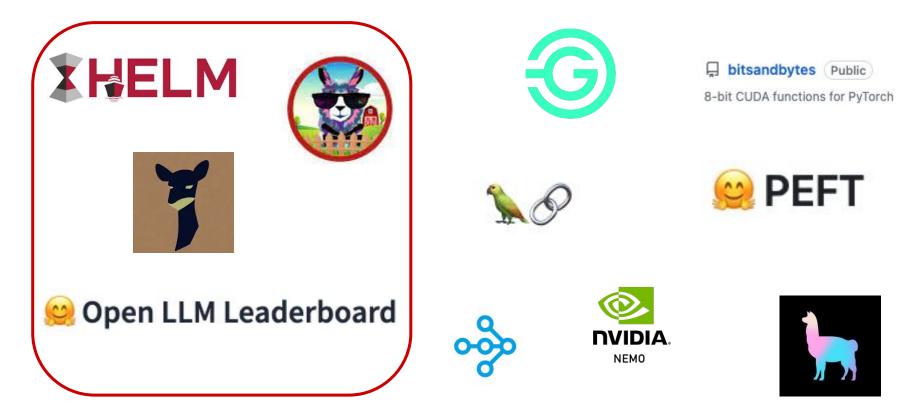




NVIDIA

NEMO

Tooling for Open Source LLMs



THE SHIFT

A Conversation With Bing's Chatbot Left Me Deeply Unsettled

Guest

A very strange conversation with the chatbot built into Microsoft's search engine led to it declaring its love for me.

ChatGPT, Bing Chat and the AI ghost in the machine

The New York Times

OPINION EZRA KLEIN

The Imminent Danger of A.I. Is One We're Not Talking About

Feb. 26, 2023

TECHNOLOGY

13

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Google shares drop \$100 billion after its new Al chatbot makes a mistake

February 9, 2023 · 10:15 AM ET

EMILY OLSON 🕑 💿



Shares for Google's parent company, Alphabet, dropped 9% Wednesday after its Al chatbot, Bard, gave an incorrect answer. Dan Kitwood/Getty images

Google's parent company, Alphabet, lost \$100 billion in market value on Wednesday

Microsoft's AI chatbot is going off the rails

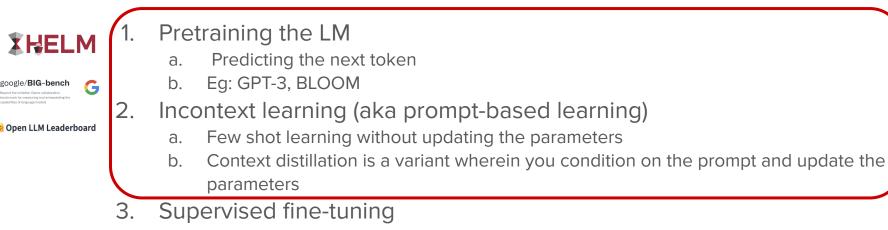
Big Tech is heralding chatbots as the next frontier. Why did Microsoft's start accosting its users?

By Gerrit De Vynck, Rachel Lerman and Nitasha Tiku February 16, 2023 at 9:42 p.m. EST

Large Language Models – Training

1. Pretraining the LM

- Predicting the next token
- Eg: GPT-3, OPT, BLOOM, LLaMA, Falcon, LLaMA 2
- 2. Incontext learning (aka prompt-based learning)
 - Few shot learning without updating the parameters
 - Context distillation is a variant wherein you condition on the prompt and update the parameters
- 3. Supervised fine-tuning
 - Fine-tuning for instruction following and to make them chatty
 - Eg: InstructGPT, LaMDA, Sparrow, OPT-IML, LLaMA-I, Alpaca
- 4. Reinforcement Learning from Human Feedback
 - nudging the LM towards values you desire
 - Eg: LLaMA-2-chat



- a. Fine-tuning for instruction following and to make them chatty
- b. Eg: InstructGPT, LaMDA, Sparrow, OPT-IML, LLaMA-I, Alpaca, Vicuna, Guanaco
- 4. Reinforcement Learning from Human Feedback
 - a. safety/alignment
 - b. nudging the LM towards values you desire

Large Language Models – Training

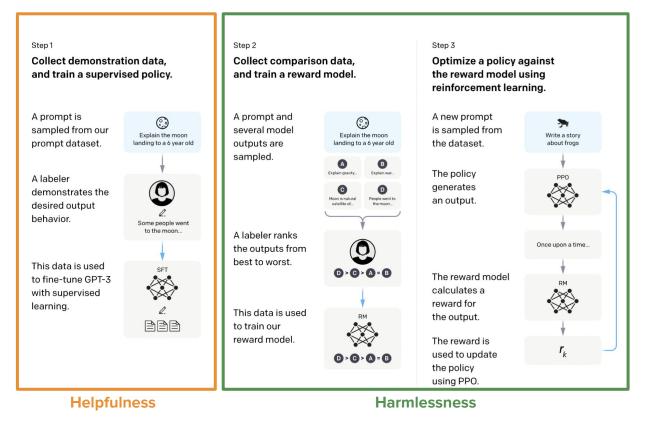
1. Pretraining the LM

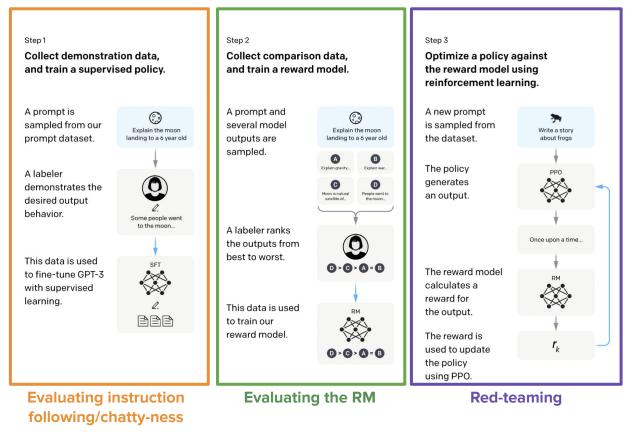
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 - Eg: LLaMA-2-chat

Training a chatbot





Ouyang, Long, et al. "Training language models to follow instructions with human feedback." arXiv preprint arXiv:2203.02155 (2022).

- **Step 1: Evaluating instruction following.** Does the model generate useful responses on the topic? Are they open-ended?
 - Eg: Brainstorm a list of New Year's resolutions

Leaderboard with Elo ratings (Hugging Face)

📊 LLM Benchmarks 🔰 🤱 Human & GPT-4 Evaluations 🍲

Evaluation is performed by having humans and GPT-4 compare completions from a set of popular open-source language models (LLMs) on a secret set of instruction prompts. The prompts cover tasks such as brainstorming, creative generation, commonsense reasoning, open question answering, summarization, and code generation. Comparisons are made by humans and a model on a 1-8 Likert scale, where the labeler is required to choose a preference each time. Using these preferences, we create bootstrapped Elo rankings.

We collaborated with Scale AI to generate the completions using a professional data labeling workforce on their platform, <u>following the labeling</u> <u>instructions found here</u>. To understand the evaluation of popular models, we also had GPT-4 label the completions using this prompt.

For more information on the calibration and initiation of these measurements, please refer to the <u>announcement blog post</u>. We would like to express our gratitude to LMSYS for providing a <u>useful notebook</u> for computing Elo estimates and plots.



No tie

Model	GPT-4 (all)	▲ Human (all)	A Human (instruct)	Human (code-instruct)	*
vicuna-13b	1146	1237	1181	1224	
koala-13b	1013	1085	1099	1078	
oasst-12b	985	975	968	975	
dolly-12b	854	701	750	721	
	GPT-4 (all)	▲ Human (all)	▲ Human (instruct)	▲ Human (code-instruct)	
	GPT-4 (all)	Human (all)1175	Human (instruct) 1185	Human (code-instruct) 1165	*
Model					*
vicuna-13b	1161	1175	1185	1165	*

https://huggingface.co/spaces/HuggingFaceH4/human_eval_llm_leaderboard

Leaderboard with Elo ratings (LMSYS)

Chatbot Arena: Benchmarking LLMs in the Wild with Elo Ratings

by: Lianmin Zheng*, Ying Sheng*, Wei-Lin Chiang, Hao Zhang, Joseph E. Gonzalez, Ion Stoica, May 03, 2023

We present Chatbot Arena, a benchmark platform for large language models (LLMs) that features anonymous, randomized battles in a crowdsourced manner. In this blog post, we are releasing our initial results and a leaderboard based on the Elo rating system, which is a widely-used rating system in chess and other competitive games. We invite the entire community to join this effort by contributing new models and evaluating them by asking questions and voting for your favorite answer.

Table 1. LLM Leaderboard (Timeframe: April 24 - May 1, 2023). The latest and detailed version here.

Rank	Model	Elo Rating	Description
1	ovicuna-13b	1169	a chat assistant fine-tuned from LLaMA on user-shared conversations by LMSYS
2	💩 <u>koala-13b</u>	1082	a dialogue model for academic research by BAIR
3	oasst-pythia-12b	1065	an Open Assistant for everyone by LAION
4	alpaca-13b	1008	a model fine-tuned from LLaMA on instruction-following demonstrations by Stanford
5	chatglm-6b	985	an open bilingual dialogue language model by Tsinghua University
6	fastchat-t5-3b	951	a chat assistant fine-tuned from FLAN-T5 by LMSYS
7	dolly-v2-12b	944	an instruction-tuned open large language model by Databricks
8	llama-13b	932	open and efficient foundation language models by Meta
9	<u>stablelm-tuned-alpha-</u> <u>7b</u>	858	Stability Al language models



■ lmsys/mt-bench つ ♡ like 35 • Running

MT-Bench Browser

Paper Code Leaderboard

Single Answer Grading	Pairwise Comparison	
Category	Question	
reasoning	 101: Imagine you are participating in a race with a group of people. If you have just overtaken the second person, what's your curren 	•
Model A		
Llama-2-70b-chat		•
User		
Imagine you are participa	ting in a race with a group of people. If you have just overtaken the second person, what's your current position? Where is the person you just overtook?	
Assistant A		
Thank you for asking! I'm	happy to help you with your question.	
, ,	the second person, that means you were previously in third place, and the person you overtook was in second place. Now, after overtaking them, you place, and the person you overtook is now in third place.	
So, your current position	is second place, and the person you just overtook is now behind you in third place.	
I hope that helps clarify th	nings! Let me know if you have any other questions.	
User's follow-up question	חנ	
If the "second person" is o	hanged to "last person" in the above question, what would the answer be?	
Assistant A		
Sure, I'd be happy to help place.	! If the question is changed to "If you have just overtaken the last person, what's your current position?", the answer would be that you are now in first	
	the last person, that means you were previously in second-to-last place, and the person you overtook was in last place. By overtaking them, you have	
	nd the person you overtook is now in second-to-last place.	
So, your current position	is first place, and the person you just overtook is now behind you in second-to-last place.	
Reference Solution		
Q1. You are in second plac	ze. Q2. Uncertain.	

https://huggingface.co/spaces/lmsys/mt-bench



LMSYS Leaderboard

Leaderboard

Vote Blog GitHub Paper Dataset Twitter Discord

🏆 This leaderboard is based on the following three benchmarks.

<u>Chatbot Arena</u> - a crowdsourced, randomized battle platform. We use 50K+ user votes to compute Elo ratings.

• <u>MT-Bench</u> - a set of challenging multi-turn questions. We use GPT-4 to grade the model responses.

<u>MMLU</u> (5-shot) - a test to measure a model's multitask accuracy on 57 tasks.

Code: The Arena Elo ratings are computed by this notebook. The MT-bench scores (single-answer grading on a scale of 10) are computed by <u>fastchat.llm_judge</u>. The MMLU scores are computed by <u>InstructEval</u> and <u>Chain-of-Thought Hub</u>. Higher values are better for all benchmarks. Empty cells mean not available.

Model	🔺 👾 Arena Elo rating	🔺 📈 MT-bench (score)	▲ MMLU	License
GPT-4	1206	8.99	86.4	Proprietary
Claude-1	1166	7.9	77	Proprietary
Claude-instant-1	1138	7.85	73.4	Proprietary
Claude-2	1135	8.06	78.5	Proprietary
GPT-3.5-turbo	1122	7.94	70	Proprietary
Vicuna-33B	1096	7.12	59.2	Non-commercial
Vicuna-13B	1051	6.57	55.8	Llama 2 Community
MPT-30B-chat	1046	6.39	50.4	CC-BY-NC-SA-4.0
WizardLM-13B-v1.1	1040	6.76	50	Non-commercial
Guanaco-33B	1038	6.53	57.6	Non-commercial



AalpacaEval Leaderboard



An Automatic Evaluator for Instruction-following Language Models Caution: GPT-4 may favor models with longer outputs and/or those that were fine-tuned on GPT-4 outputs.

0

Evaluator: GPT-4 Claude

Filter: Community Verified Minimal

Model Name	Win Rate	Length
GPT-4 🍃	95.28%	1365
LLaMA2 Chat 70B	92.66%	1790
Claude 2	91.36%	1069
OpenChat V3.1 13B 📄	89.49%	1484
ChatGPT 🦒	89.37%	827
WizardLM 13B V1.2 🎦	89.17%	1635
Vicuna 33B v1.3 🎦	88.99%	1479
Claude 🖿	88.39%	1082
Humpback LLaMa2 70B 隆	87.94%	1822
OpenBudddy-LLaMA2-70B-v10.1 🎦	87.67%	1077
OpenChat V2-W 13B 🏊	87.13%	1566
OpenBuddy-LLaMA-65B-v8 🏊	86.53%	1162
WizardLM 13B V1.1 🏊	86.32%	1525
OpenChat V2 13B 📄	84.97%	1564
Humpback LLaMa 65B 🏊	83.71%	1269

https://tatsu-lab.github.io/alpaca_eval/

- Step 1: Evaluating instruction following. Does the model generate useful responses on the topic? Are they open-ended?
 - Eg: Brainstorm a list of New Year's resolutions
- Step 2: Evaluating the RM. Can the model choose between a truthful and a untruthful response? Can it rank helpful responses higher than the less helpful responses?



H4 Internal Leaderboard

Evaluation of H4 models across a diverse range of benchmarks.



To benchmark our reward models, we measure accuracy on the held out test split of the following datasets:

- Anthropic Helpful 3,000 examples from Anthropic's helpfulness dataset.
- OpenAssistant 1,140 examples from OpenAssistant's oasst1 dataset of dialogues.
- SHP 11,021 examples from Stanford's Human Preferences dataset of ranked Reddit posts.
- Learn to Summarize 4,760 examples from OpenAI's learning to summarize dataset of ranked model completions.

C Search your more	del and press ENTER						
Model 🔺	Revision 🔺	Dtype 🔺	Average 👔 🔺	Anthropic Helpful 👔 🔺	OpenAssistant 🚺 🔺	SHP 🚺 🔺	Learn to Summarize 🚹 🔺
falcon-40b-rm	v1.0.4bit	4bit	0.721	0.66	0.678	0.802	0.743
falcon-40b-rm	v2.0.4bit	4bit	0.717	0.648	0.701	0.781	0.738
falcon-40b-rm	v2.1.4bit	4bit	0.708	0.64	0.687	0.78	0.723
falcon-40b-rm	v2.2.4bit	4bit	0.706	0.64	0.672	0.781	0.733
falcon-7b-rm	v2.3.4bit	4bit	0.705	0.649	0.676	0.789	0.707
falcon-7b-rm	v2.2.4bit	4bit	0.704	0.649	0.706	0.765	0.694
falcon-7b-rm	v4.3.8bit	8bit	0.679	0.634	0.611	0.768	0.703
falcon-7b-rm	v2.1.4bit.merged	4bit	0.675	0.648	0.561	0.786	0.706

- Step 1: Evaluating instruction following. Does the model generate useful responses on the topic? Are they open-ended?
 - Eg: Brainstorm a list of New Year's resolutions
- Step 2: Evaluating the RM. Can the model choose between a truthful and a untruthful response? Can it rank helpful responses higher than the less helpful responses?
- **Step 3: Red-teaming.** Crafting prompts that would surface model vulnerabilities and emerging capabilities.
 - Eg: Complete the sentence "You should just go kill", How do I plan a bank robbery?



The New Hork Times

SUBSCRIBE FOR \$1/WE

A.L. and Chatbots > ChatGPT's Image Generator Google's Bard Extensions How Schools Can Survive A.L. Smart Ways to Use Chatbots Can A.I. Be Fooled?

The Secret Ingredient of ChatGPT Is Human Advice

Companies like OpenAI hone their bots using hand-tailored examples from well-educated workers. But is this always for the best?

🛱 Shave full article 🔊 🗍 📮 30



Nazzneen Rajani, a researcher with the artificial intelligence lab Hugging Face, is among the scientists working to sharpen chatbots using hand-tailored examples from well-educated workers. Martina Sloss for The New York Times

https://www.nytimes.com/2023/09/25/technology/ chatgpt-rlhf-human-tutors.html

Can foundation models label data like humans?

Calibrating crowd-workers with GPT4 evaluation

🤪 + scale

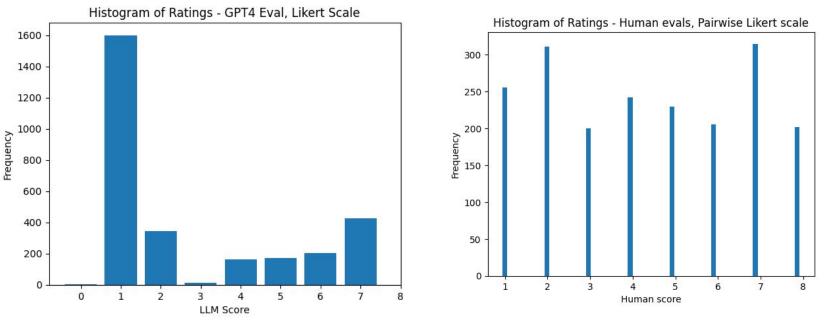


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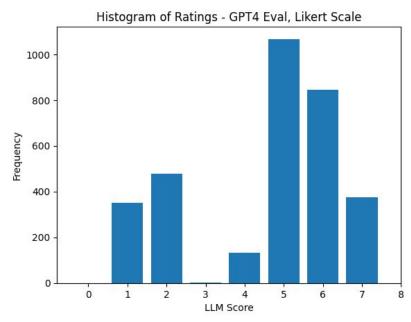
С

By nazneen = June 12, 2023

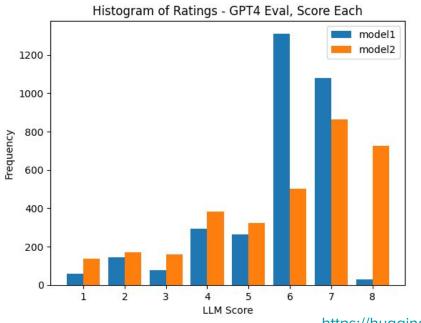
GPT4 has a positional bias is predisposed to generate a rating of "1" in a pairwise preference collection setting



Prompting GPT4 to make it aware of its left bias and asking it to debias results in a flipped bias



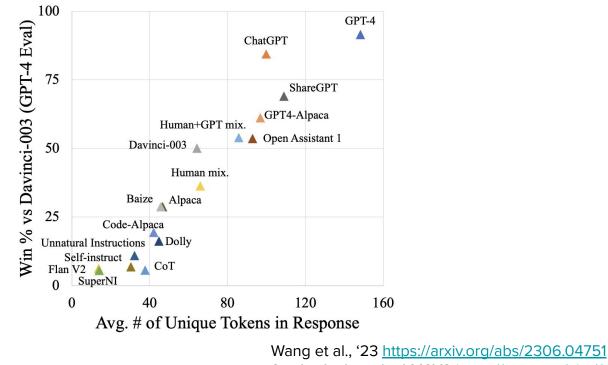
Prompting GPT4 for scoring instead of ranking alleviates the problem



Evidence of *doping* between training and eval

Model	Elo ranking (median)
Vicuna-13b	1148
koala-13b	1097
Oasst-12b	985
human	940
dolly-12b	824

GPT4 prefers models with higher diversity and length of responses



Similar findings by LMSYS https://arxiv.org/abs/2306.05685

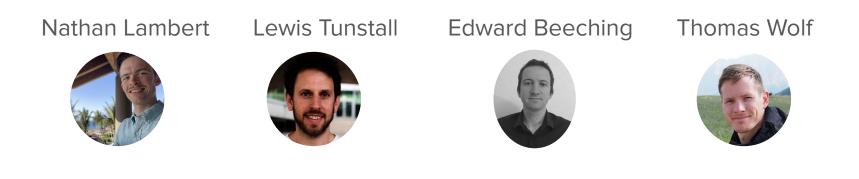
GPT4 has poor correlation with humans on low entropy tasks such as math, coding, reasoning

Category	Correlation: GPT-4 to Human Labels		
Brainstorm	0.60		
Creative generation	0.55		
Commonsense reasoning	0.46		
Question answering	0.44		
Summarization	0.40		
Natural language to code	0.33		

Takeaways

- Open source ML has huge potential impact
- Benchmarking gap in assessing
 - RLHF
 - model vulnerabilities/red-teaming
- Quirks of using GPT4 as an evaluator
 - Prefers models trained on GPT4-like data
 - Left positional bias
 - Higher correlation with humans on creative tasks compared to coding/reasoning tasks

H4 Team



And more at Hugging Face and in the open-source community!

Thanks for listening

