

Gen Al

Generative Al

Autoregressive Models

Diffusion Models

With **Gen AI** we refer to applications of AI where **new data is created** (generated) upon user interaction

- ChatGPT/LLaMa generate language
- DALL-E / Midjourney generate images
- SORA / Pika generate video
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<u>Autoregressive</u> <u>Models</u>

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Autoregressive Language Models



The best thing about AI is its ability to

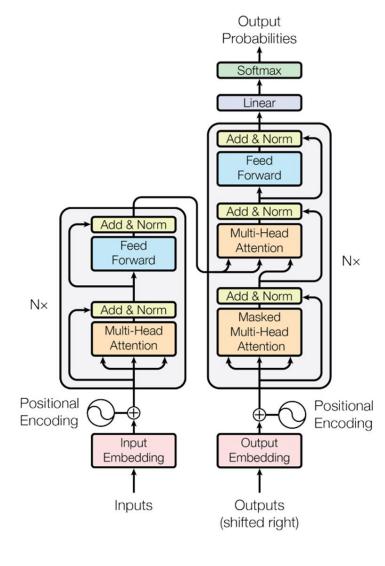
learn	4.5%			
predict	3.5%			
make	3.2%			
understand	3.1%			
do	2.9%			

 Autoregressive models generate data sequentially, one step at a time.

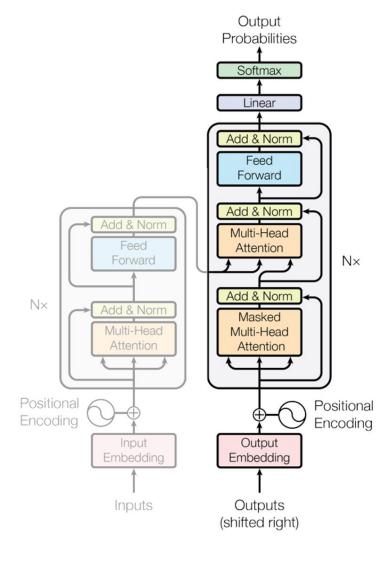
 In the case of text generation, autoregressive models predict the next token (word or subword) based on the previously generated tokens.

 The generation process is **iterative**, with each step depending on the output of the previous steps.









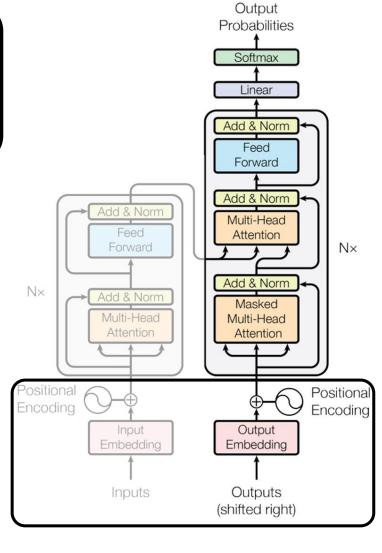


1

Input Processing

- Tokenization: Breaking down input text into tokens (using BPE)
- Embedding: Converting tokens into dense semantic vectors (learned)
- Positional Encoding: Adding position information to the embeddings







2

Input Processing

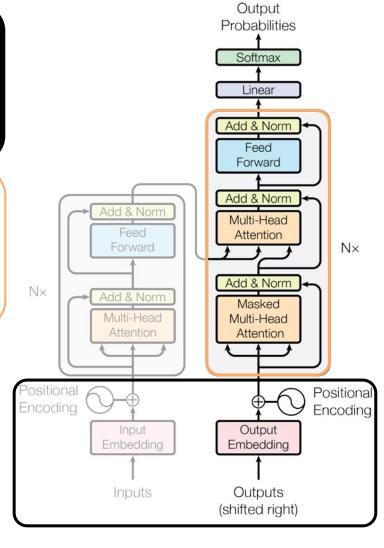
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Transformer Decoder Architecture

- Multi-Head Self-Attention: Attending to different parts of the input
- Feed Forward Neural Network: Processing the attended information
- Residual Connections and Layer Norm: Enabling stable training







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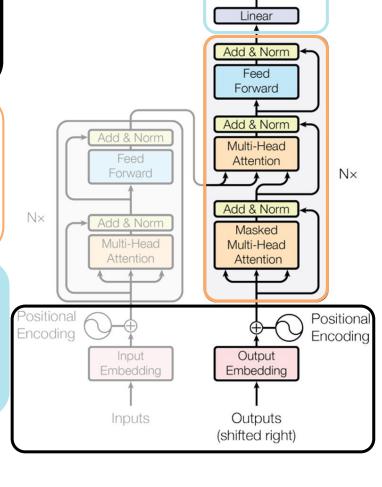




Output Generation

- Vocabulary Distribution: net output is a distribution over the vocabulary
- Sampling Techniques: top-k, top-p sampling for token selections
- Iterative process: Selected token is fed back to generate the next one





Output **Probabilities**

Softmax



State of the Art Models



LMSYS Chatbot Arena Leaderboard

Category Overall Questions								
Overall		#mc	odels: 92 (100%)	#votes: 910,122 (100%)				
Rank* (UB)		♠ Arena Elo	n 95% CI ▲	Votes •	Organization A	License A	Knowledge Cutoff	
1	GPT-4-Turbo-2024-04-09	1259	+4/-3	35931	0penAI	Proprietary	2023/12	
2	GPT-4-1106-preview	1253	+2/-3	73547	0penAI	Proprietary	2023/4	
2	Claude 3 Opus	1251	+3/-3	80997	Anthropic	Proprietary	2023/8	
2	Gemini 1.5 Pro API- 0409-Preview	1250	+3/-3	39482	Google	Proprietary	2023/11	
2	GPT-4-0125-preview	1247	+3/-2	67354	OpenAI	Proprietary	2023/12	
6	Llama-3-70b-Instruct	1210	+3/-4	53404	Meta	Llama 3 Community	2023/12	
6	Bard (Gemini Pro)	1209	+5/-6	12387	Google	Proprietary	Online	
7	Claude 3 Sonnet	1201	+2/-3	78956	Anthropic	Proprietary	2023/8	
9	Command R+	1191	+3/-3	44988	Cohere	CC-BY-NC-4.0	2024/3	
9	GPT-4-0314	1190	+3/-4	52079	0penAI	Proprietary	2021/9	
11	Claude_3_Haiku	1181	+2/-3	69660	Anthropic	Proprietary	2023/8	



Open Source vs Open Weights



1 : Accelerating the Science of Language Models

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Abstract

Language models (LMs) have become ubiquitous in both NLP research and in commercial product offerings. As their commercial importance has surged, the most powerful models have become closed off, gated behind proprietary interfaces, with important details of their training data, architectures, and development undisclosed. Given the importance of these details in scientifically studying these models, including their biases and potential risks, we believe it is essential for the research community to have access to powerful, truly open LMs. To this end, this technical report details the first release of OLMo, a state-of-the-art, truly **O**pen Language Model and its framework to build and study the science of language modeling. Unlike most prior efforts that have only released model weights and inference code, we release OLMo and the whole framework, including training data and training and evaluation code. We hope this release will empower and strengthen the open research community and inspire a new wave of innovation.

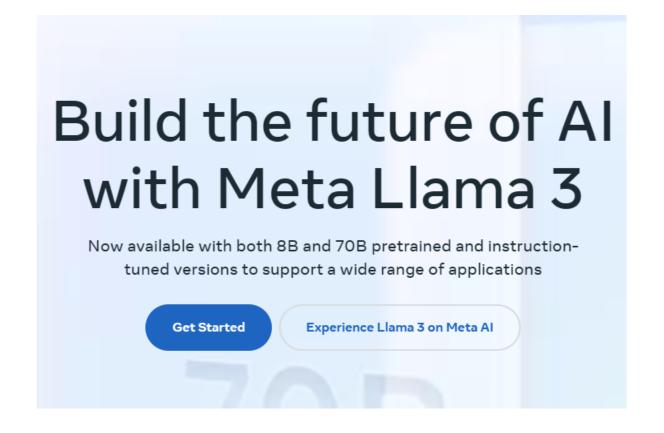
<u></u>	Weights	https://huggingface.co/allenai/OLMo-7B
C	Code	https://github.com/allenai/OLMo
9	Data	https://huggingface.co/datasets/allenai/dolma
O	Evaluation	https://github.com/allenai/OLMo-Eval
0	Adaptation	https://github.com/allenai/open-instruct
	W&B Logs	https://wandb.ai/ai2-llm/OLMo-7B/reports/OLMo-7BVml



New entry within Open Weights LLMs by Meta

Latest release:

- Llama 3 released on April 18th 8B and 70B Models
- 400B Model currently under training, planned weight release
- LLaMa 3 license is royalty free if:
 - monthly active users < 700M
 - not use to improve other LLMs



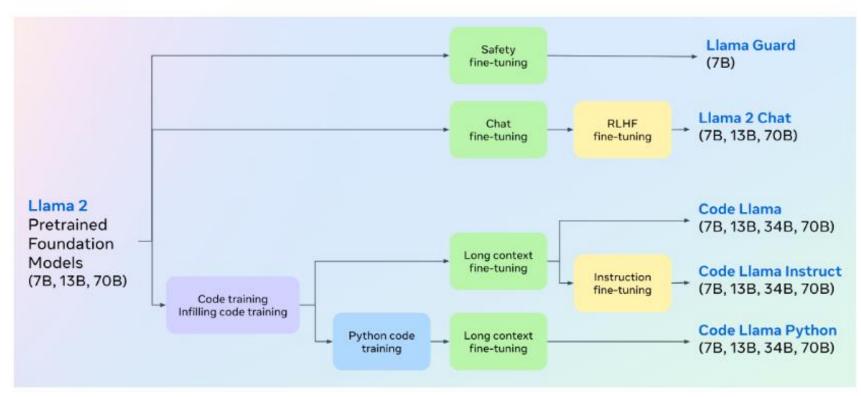


Impressive in English

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10 ↑	1	Claude 3 Haiku	1159	+4/-4	37016	Anthropic	Proprietary	2023/8

Which Llama2?

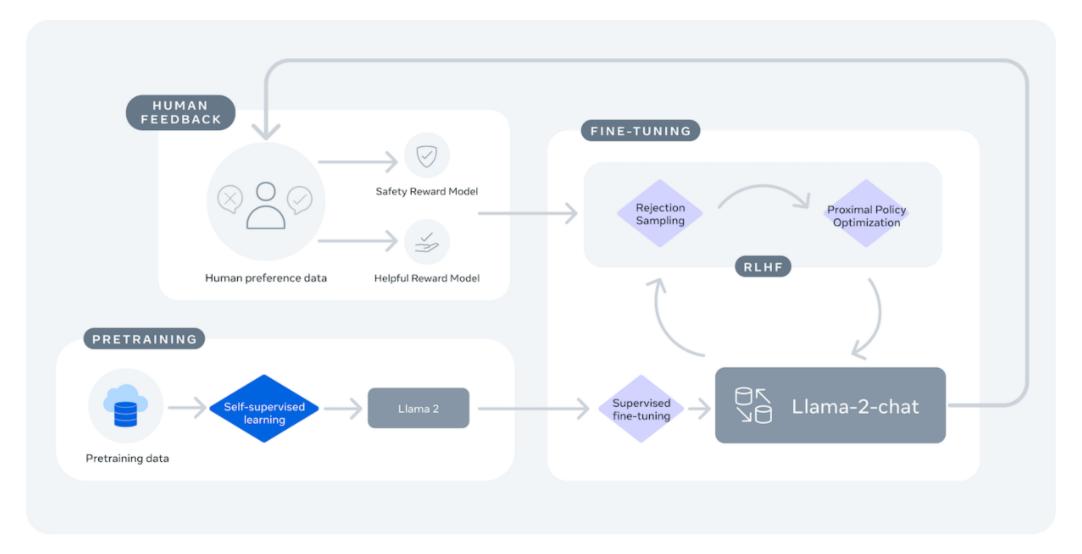




- Llama 2: Foundation model (good at text completion)
- •Llama2 chat: Good at question answering (like ChatGPT!)
- CodeLlama Instruct : Good for safer deployment
- •LlamaGuard: input-output safeguard model



Llama 2 Training





Training != Inference





Let's continue the discussion!

If you want to talk more about LLMs feel free to add me on the event website or Linkedin or just come say hi in the next days:)





Ivan Gentile

Data Scientist at IFAB | Machine Learning & Big Data | Harnessing AI & HPC for Societal Impact



Acknowledgements

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Disclaimer

Funded by the European Union. Views and opinions expressed are however those of the author(s) only and do not necessarily reflect those of the European Union or the European High Performance Computing Joint Undertaking (JU) and the participating countries in the project. Neither the European Union nor the granting authority can be held responsible for them.





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