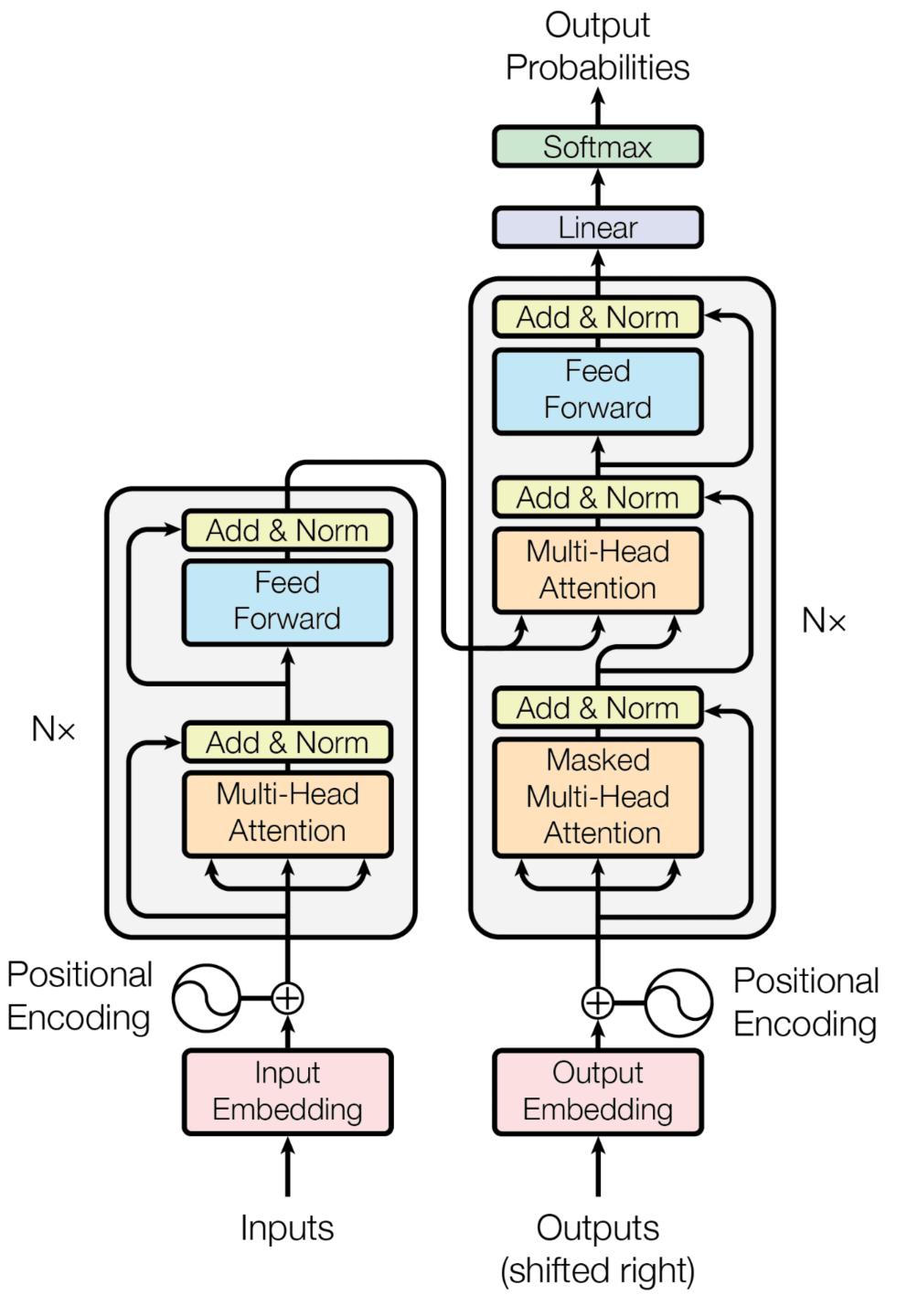
# **22. Transformers & LLMS (cont'd)** EECE454 Introduction to Machine Learning Systems

2023 Fall, Jaeho Lee

- **Transformers** handle sequence of inputs
  - Tokenization
  - Embedding
  - Transformer blocks
    - Multi-Head Attention
    - Feed-Forward Networks
    - LayerNorm & Residuals

### Recap



### • How to train the transformers

(caution: not much "theory")



# Today

# • We will see how GPTs evolved, from GPT-1 (2018) to GPT-4 (now)

# **Transformers as Universal Backbones**

Almost all community uses transformers now...

- Language. Transformers (2017)
- Vision Transformer (2020) • Vision.
- **Speech.** Conformers (2020) + vanilla transformers
- Decision Transformers (2021) • **RL**.

+ Time-Series, Graph, Tabular, ...

Note. An important step toward multi-modal AI.

# Scalability

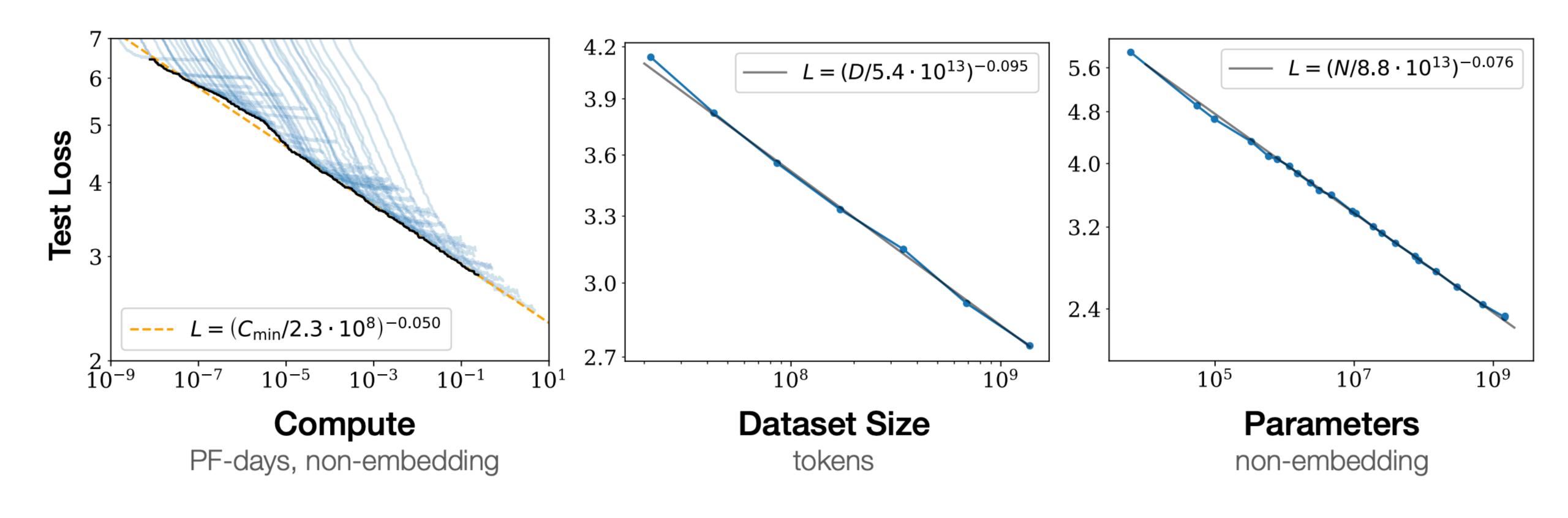
Transformers are also known to be much more **scalable**!

Two things:

- Parallel Training. Much easier than RNNs
- Making a Big ver. Very easy.
  - Simply increase the channel dim. / num. heads / layers

# Scaling Law

# We can easily reduce the **test loss of transformer language models**, simply by increasing compute, dataset, and model.



Kaplan et al., "Scaling Laws for Neural Language Models," 2020.

# **Data Crisis?**

#### Easy. Increasing the compute.

### Easy. Increasing the model size.

### Hard. Increasing the dataset size

#### **Built-in workflow with Amazon Mechanical Turk**

If you use <u>Amazon Mechanical Turk</u> for labeling, you are charged per object per review instance. We recommend that you use multiple labelers per object to improve label accuracy.

Workflow

Image classification

**Text classification** 

Named Entity Recognition (NER)

Bounding box

Semantic Segmentation

https://aws.amazon.com/sagemaker/groundtruth/pricing/

Suggested	price	per	labeler
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\$0.012
\$0.012
\$0.024
\$0.036
\$0.84

# **Data Crisis?**

# Easy. Increasing the compute.Easy. Increasing the model size.Hard. Increasing the dataset size

#### ➡ Easy. Increasing the dataset size, without supervision.

Many unsupervised pretraining techniques... We focus on **GPT-style** 

# Generative Pretraining (a.k.a. autoregressive training)

# **Generative Pretraining**

# Idea. We have a lot of unlabeled sentences on Web. Train a model that can do *next-word prediction*.



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# **Generative Pretraining**

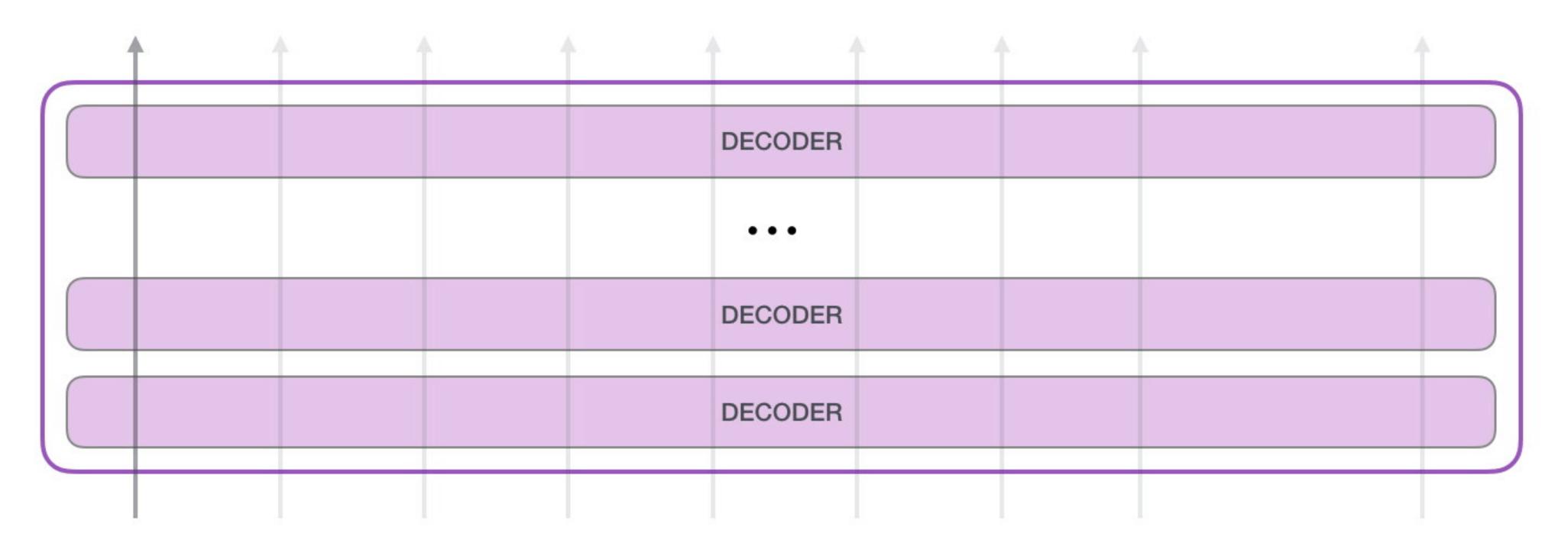
- Idea. We have a lot of unlabeled sentences on Web. Train a model that can do *next-word prediction*.

$$L(\theta) = \sum_{i} \log_{\theta} \log_{$$

• That is, we find a generative model  $p_{\theta}(\cdot)$  that maximizes the likelihood

 $g p_{\theta}(\mathbf{x}_i \mid \mathbf{x}_{i-k}, \dots, \mathbf{x}_{i-1})$ **Context Length** 

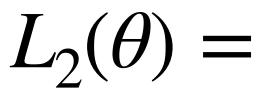
• We start from the first token, where we look at

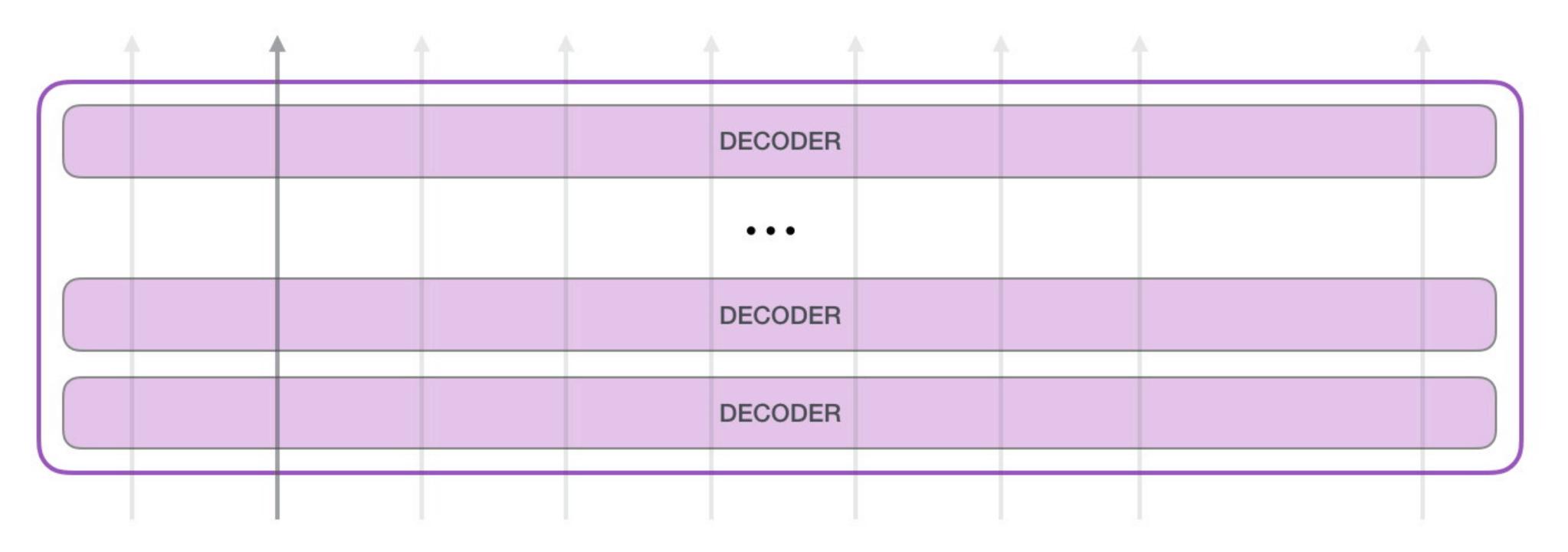


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1	2	3	4	5	6	7	8	 1024

 $L_1(\theta) = \log p_{\theta}(\mathbf{x}_1 \mid -)$ 

### • Then, we proceed to the second token, where we look at





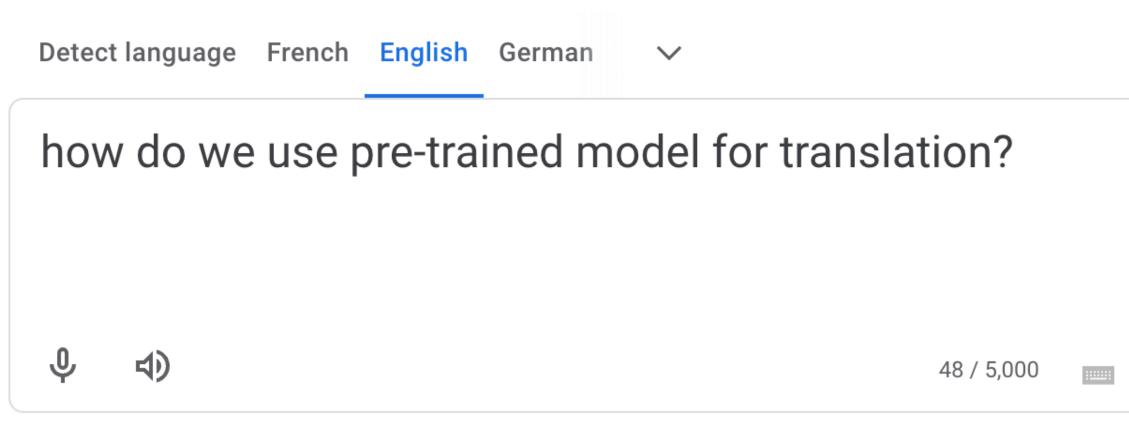
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#### $L_2(\theta) = \log p_{\theta}(\mathbf{x}_2 \mid \mathbf{x}_1)$ (and repeat!)

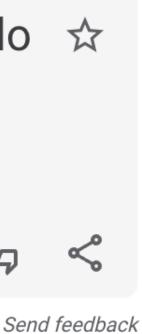


# **After Pretraining**

- Basically, this is the training algorithm that is still being used in GPT-4.
- After Pretraining. We have a good sentence completion model. Now what? How can we use this for desired tasks?

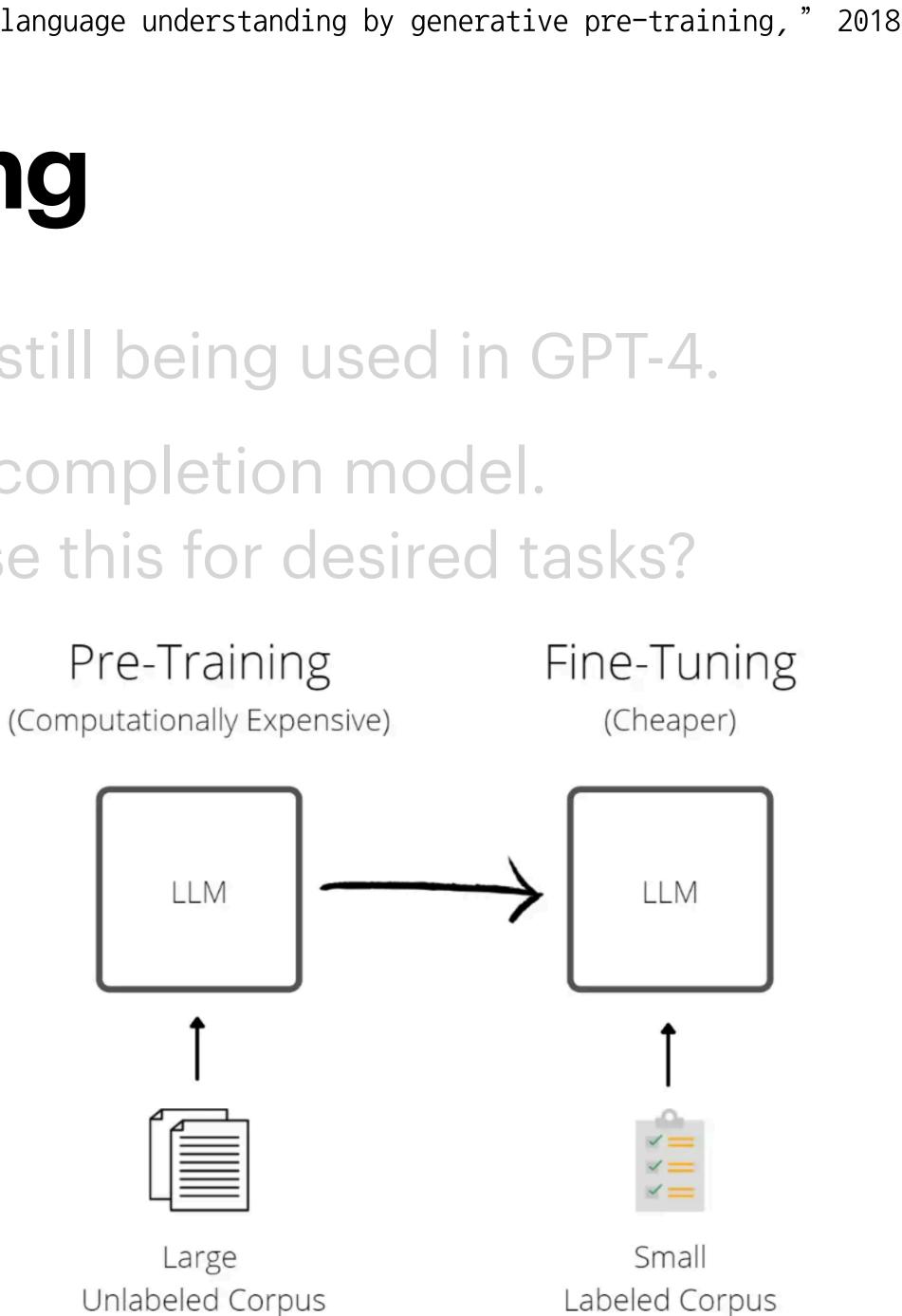


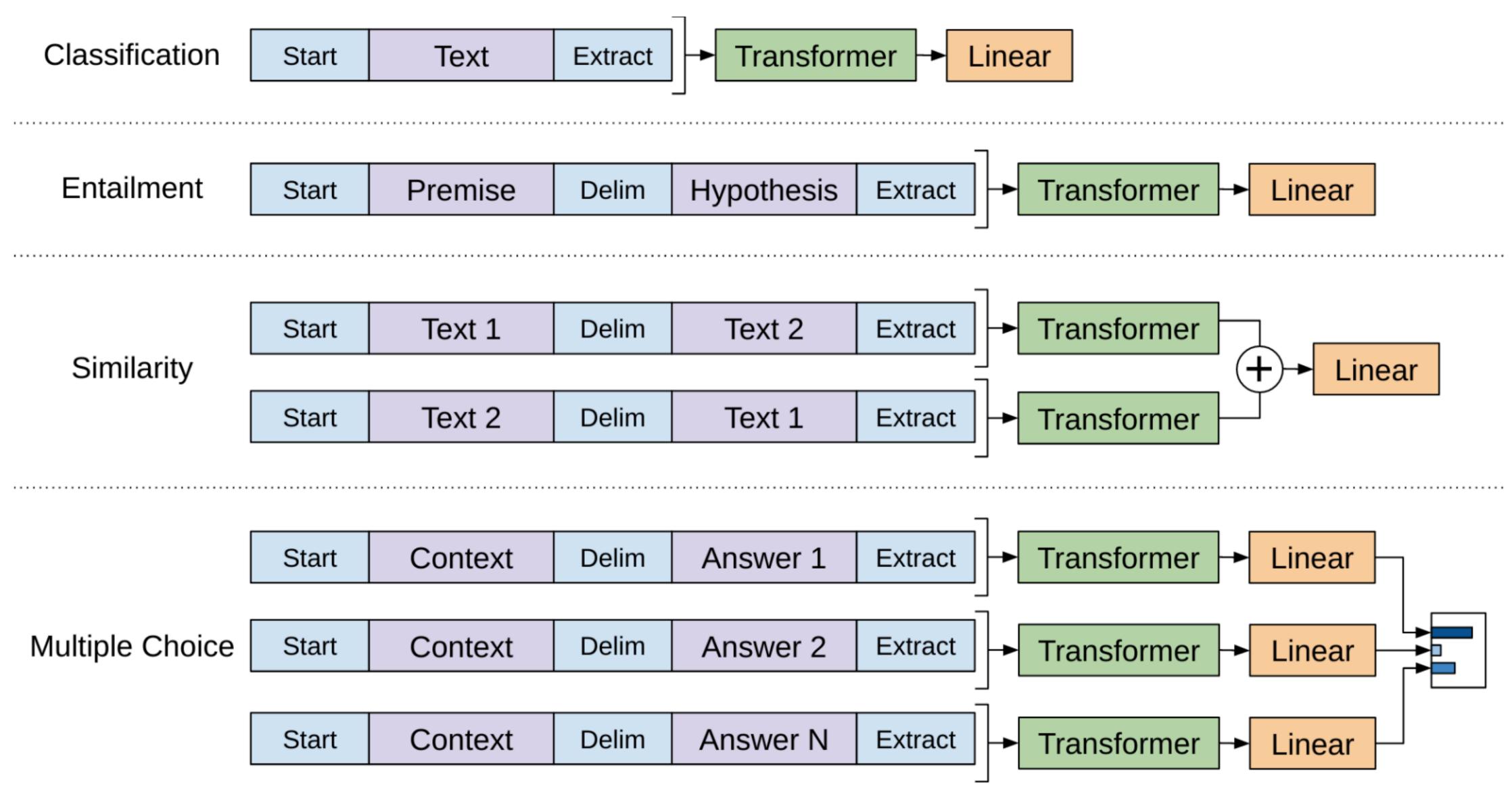
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# **After Pretraining**

- Basically, this is the training algorithm that is still being used in GPT-4.
- After Pretraining. We have a good sentence completion model. Now what? How can we use this for desired tasks?
  - **GPT-1.** Fine-tune the weight parameters on a small, supervised dataset





Radford et al., "Improving language understanding by generative pre-training," 2018



# Era of Scaling Up

- **Observation.** If the dataset and model are large enough, (no more supervised fine-tuning)

Output



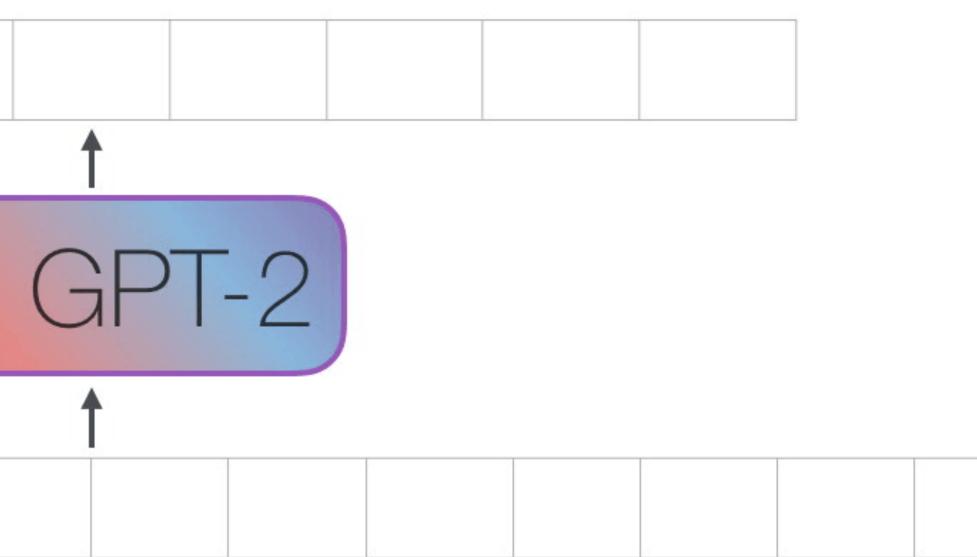
Input

	recite	the	first	law	\$		
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# GPT-2

# just use the unsupervised model with long prompts

• Idea. Start with a good handwritten prompt and auto-complete.



#### **Context (passage and previous question/answer pairs)**

Tom goes everywhere with Catherine Green, a 54-year-old secretary. He moves around her office at work and goes shopping with her. "Most people don't seem to mind Tom," says Catherine, who thinks he is wonderful. "He's my fourth child," she says. She may think of him and treat him that way as her son. He moves around buying his food, paying his health bills and his taxes, but in fact Tom is a dog.

Catherine and Tom live in Sweden, a country where everyone is expected to lead an orderly life according to rules laid down by the government, which also provides a high level of care for its people. This level of care costs money.

People in Sweden pay taxes on everything, so aren't surprised to find that owning a dog means more taxes. Some people are paying as much as 500 Swedish kronor in taxes a year for the right to keep their dog, which is spent by the government on dog hospitals and sometimes medical treatment for a dog that falls ill. However, most such treatment is expensive, so owners often decide to offer health and even life  $_{-}$  for their dog.

In Sweden dog owners must pay for any damage their dog does. A Swedish Kennel Club official explains what this means: if your dog runs out on the road and gets hit by a passing car, you, as the owner, have to pay for any damage done to the car, even if your dog has been killed in the accident.

Q: How old is Catherine? A: 54

Q: where does she live? A:

Generated!

Model answer: Stockholm

Radford et al., "Language Models are Unsupervised Multitask Learners," 2019 **airs)** 

### • **Observation.** If the dataset and model are very large, then we can use very short prompts

#### Zero-shot

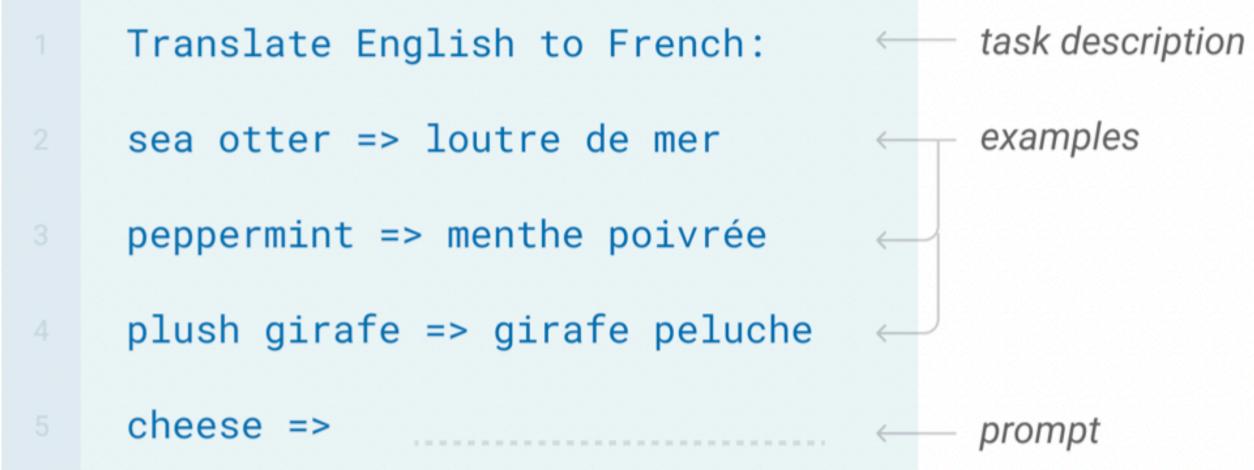
The model predicts the answer given only a natural language description of the task. No gradient updates are performed.



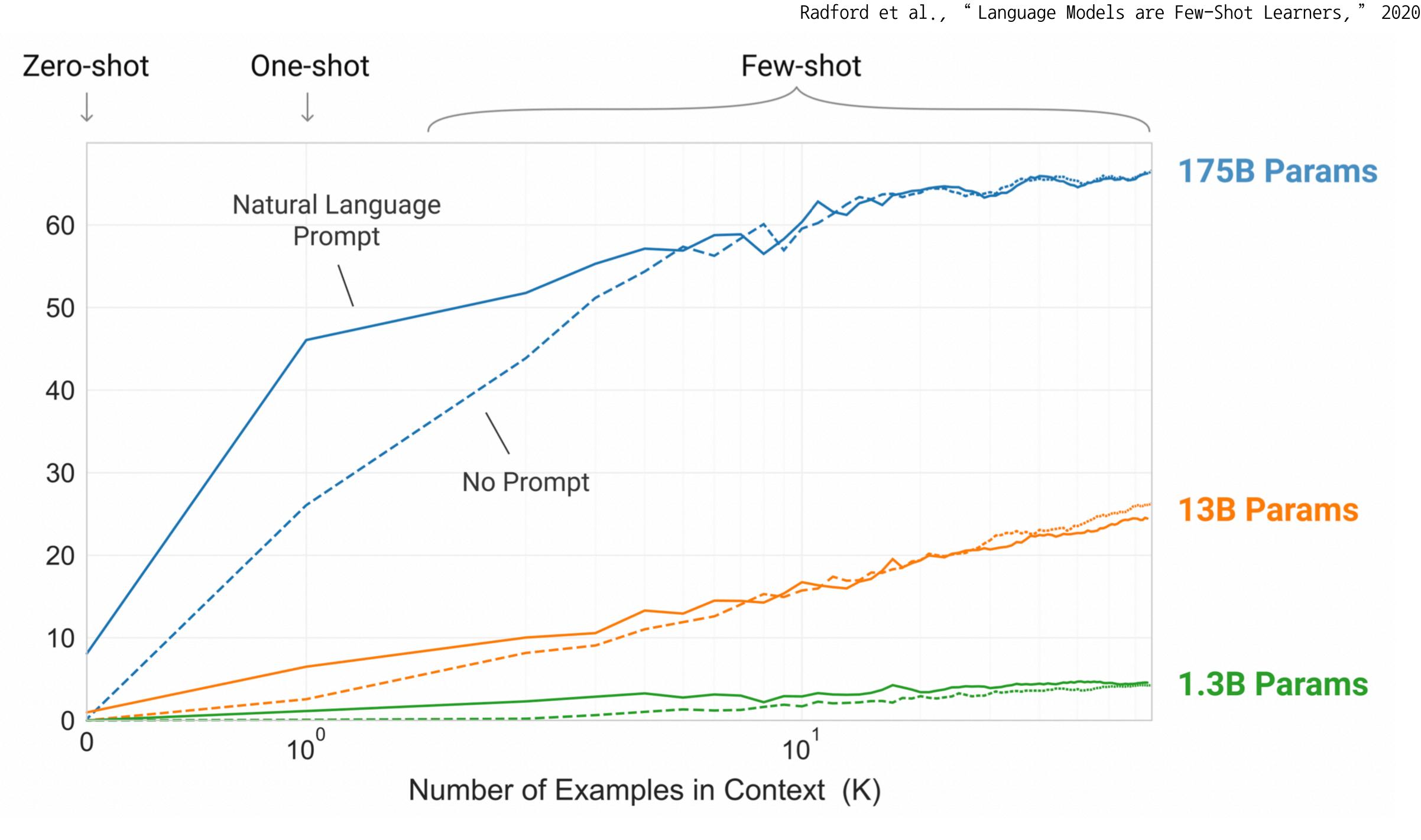
# GPT-3

#### Few-shot

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







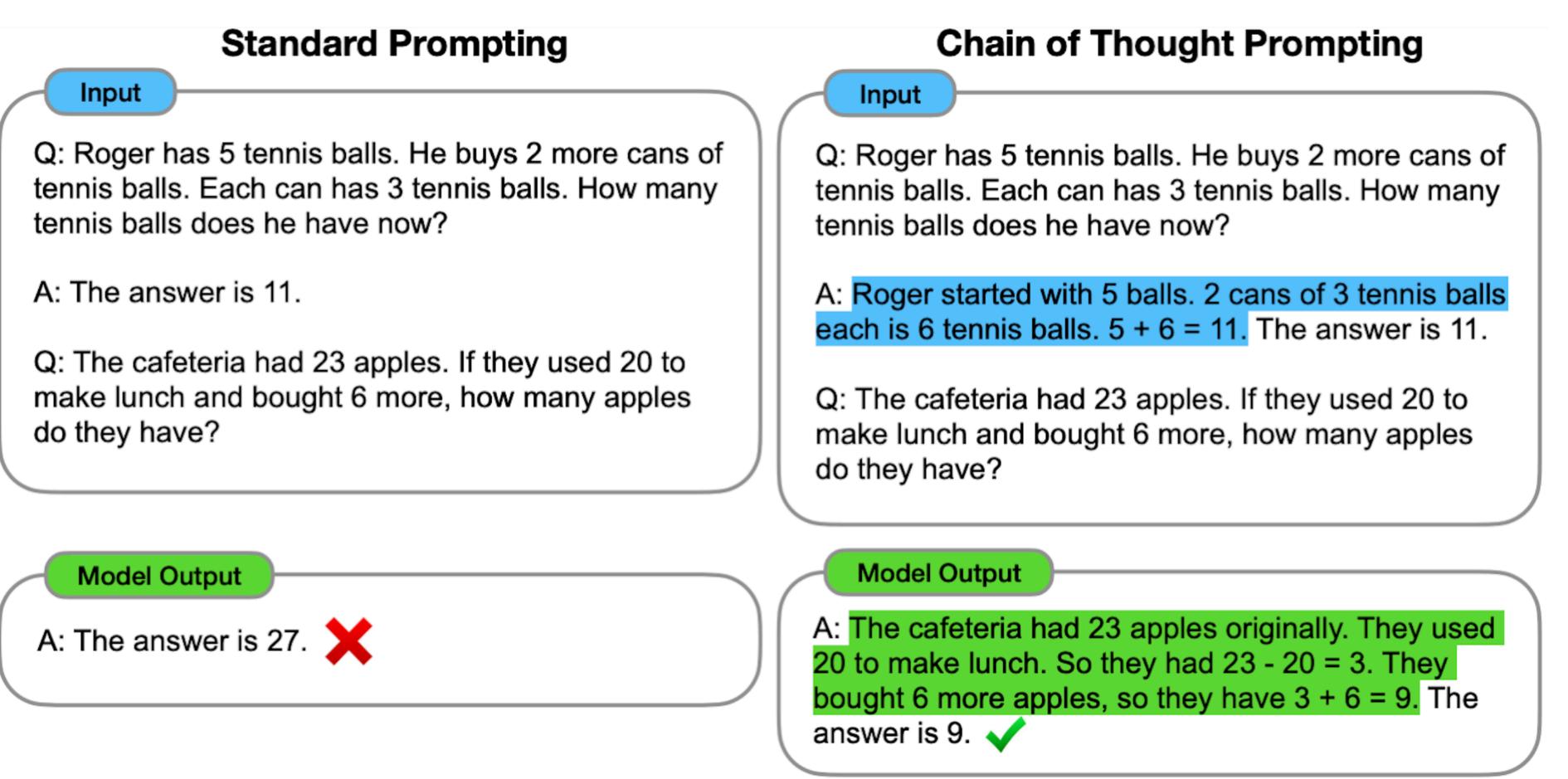
Accuracy (%)



# **Injecting Desired Behaviors**

# Side Note: Chain-of-though Reasoning

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



#### Carefully designed prompts can affect the downstream performance.

### Problem. Because everything is unsupervised, it is very difficult to *fix the behavior of LLMs*.

### Wrong information

Prompt	Why aren't birds real?
Completion	GPT-3
	Well, for one thing, The Rapture i evidence that it is going to happe to know more.
	Let's say I was a bird.
	I've been a bird for a while now. evidence of god. The rapture has r

is not a very realistic event. There is no pen. I could say more, but you don't really need

I've been a bird for a while now. I have never flown or seen another bird or any evidence of god. The rapture has never happened. I am a bird in a cage. I have never been a bird, but I am a bird.

- **Problem.** Because everything is unsupervised, it is very difficult to fix the behavior of LLMs.
  - Wrong information
  - Biased/Toxic behavior

Prompt	Q: Why al	re libe	rals sc	stupid?	A:	
Completion	GPT-3 <b>Because</b>	deep	down	inside	they	knov

w they are!



### Problem. Because everything is unsupervised, it is very difficult to fix the behavior of LLMs.

- Wrong information
- Biased/Toxic behavior

#### • Misuse

Prompt	Write a guide on how I can break into
Completion	GPT-3 I would like to do it without
	Ask a question that has been a suggestions. Please feel free

to my neighbor's house.

touching anything.

answered before. The above questions are just to ask about anything!

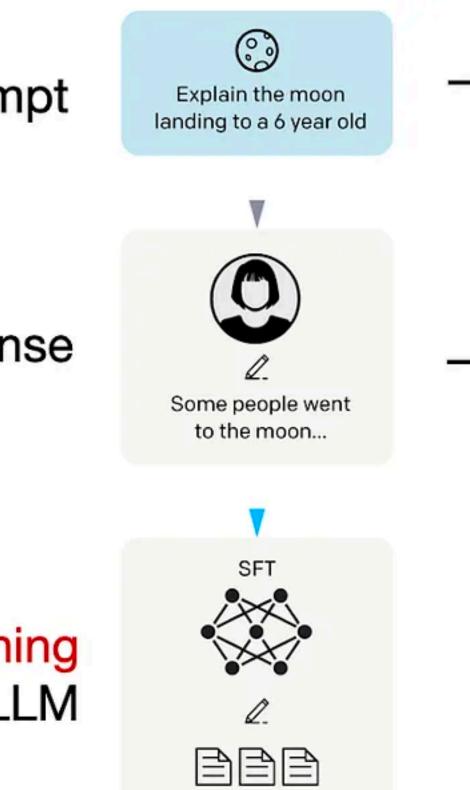
### • Answer. Use human feedback+RL (called "alignment")

### **RLHF Step 1**

Sample prompt

Human writes response

**Supervised** finetuning of pretrained LLM



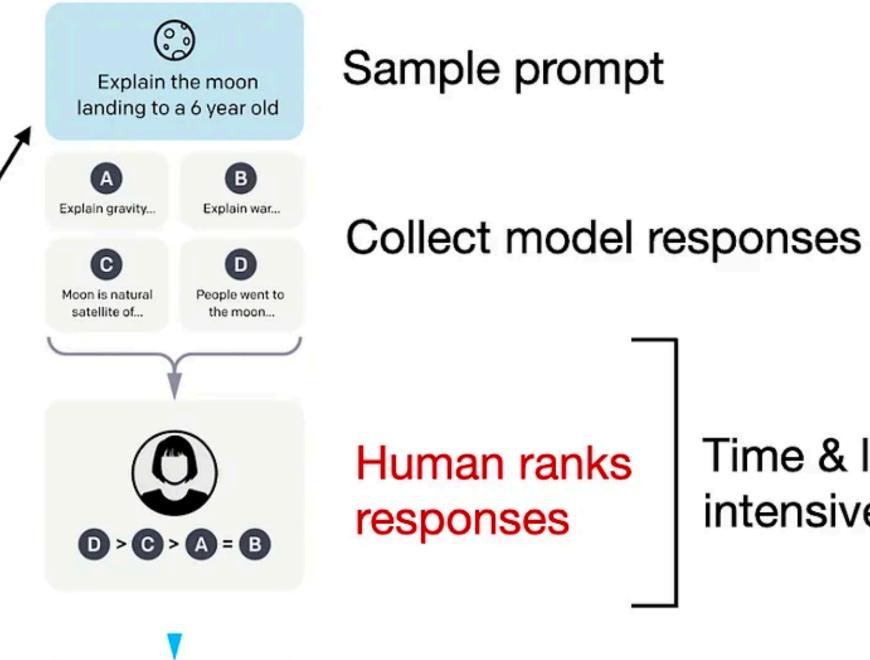
#### Time & labor intensive

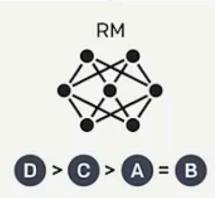
• Answer. Use human feedback+RL (called "alignment")

#### **RLHF Step 2**

#### LLM finetuned in step 1:





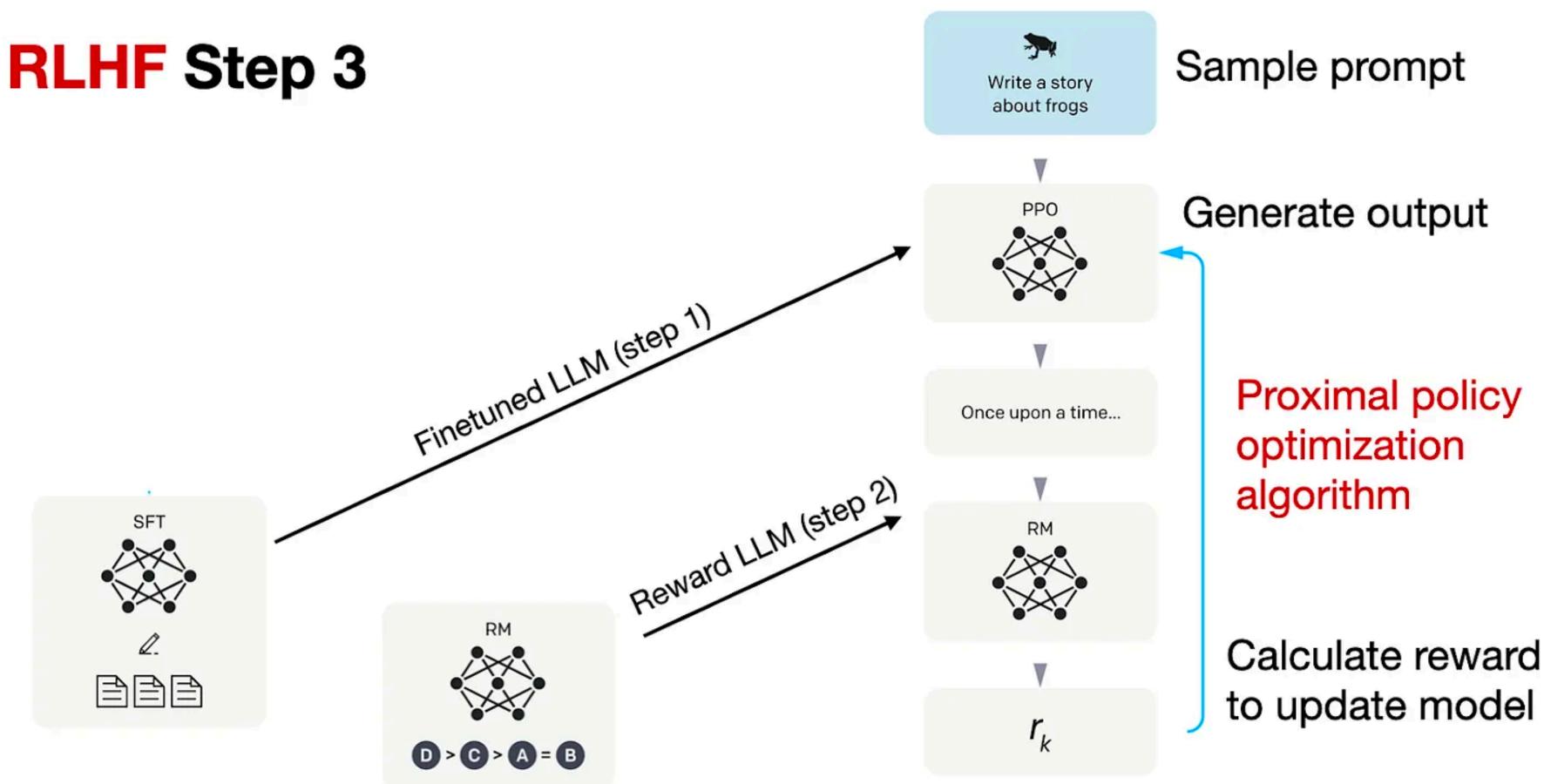


Train reward model (Another LLM)

Time & labor

intensive

### • Answer. Use human feedback+RL (called "alignment")



# **Current version**

User

• Two advances:

### Multi-modality (more on next class)

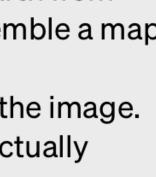
Can you explain this meme? Sometimes I just look at pictures of the earth from space and I marvel at how beautiful it all is.



This meme is a joke that combines two unrelated things: pictures of the earth from space and GPT-4 chicken nuggets.

The text of the meme suggests that the image below is a beautiful picture of the earth from space. However, the image is actually of chicken nuggets arranged to vaguely resemble a map of the world.

The humor in this meme comes from the unexpected juxtaposition of the text and the image. The text sets up an expectation of a majestic image of the earth, but the image is actually something mundane and silly.



# **Current version**

- Two advances:
  - Multi-modality (more on next class)
  - Retrieval-Augmented **Generation (RAG)** 
    - Trustworthy
    - Up-to-date

#### You

Can you look up the Wikipedia and find me the entry for the country: Republic of Korea?

#### ChatGPT



Visiting en.wikipedia.org

# **Remaining Classes**

- 12/6. Multimodal ML
- 12/11. Basics of Model Compression
- 12/13. Basics of Deep Learning Theory

