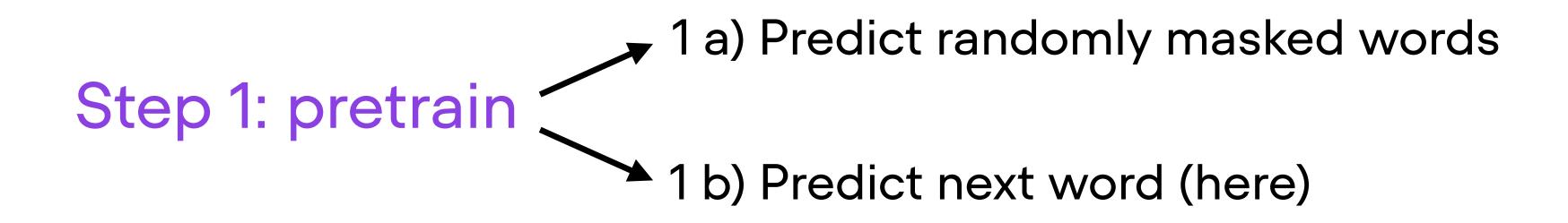
### 8.6

Large Language Models

Part 2: Generative Pretrained Transformer (GPT)

Sebastian Raschka and the Lightning Al Team

#### Self-supervised pretraining



Step 2: fine-tune

	Date
GPT 1	2018
GPT 2	2019
GPT 3	2020
InstructGPT & ChatGPT	2022

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	Date	Size
GPT 1	2018	110 million
GPT 2	2019	1.5 billion
GPT 3	2020	175 billion
InstructGPT & ChatGPT	2022	GPT 3-based

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	Date	Size	Paper
GPT1	2018	110 million	Improving Language Understanding by Generative Pre-Training
GPT 2	2019	1.5 billion	Language Models are Unsupervised Multitask Learners
GPT 3	2020	175 billion	Language Models are Few-Shot Learners
InstructGPT & ChatGPT	2022	GPT 3-based	Aligning Language Models to Follow Instructions

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## GPT models are generative models (as opposed to classifiers)

#### GPT models are generative models (as opposed to classifiers)



How to make a banana smoothie?



To make a banana smoothie, you will need the following ingredients:





- 1 ripe banana
- 1 cup of milk (dairy or non-dairy)
- 1/2 cup of plain yogurt (dairy or non-dairy)
- 1 tablespoon of honey (optional)

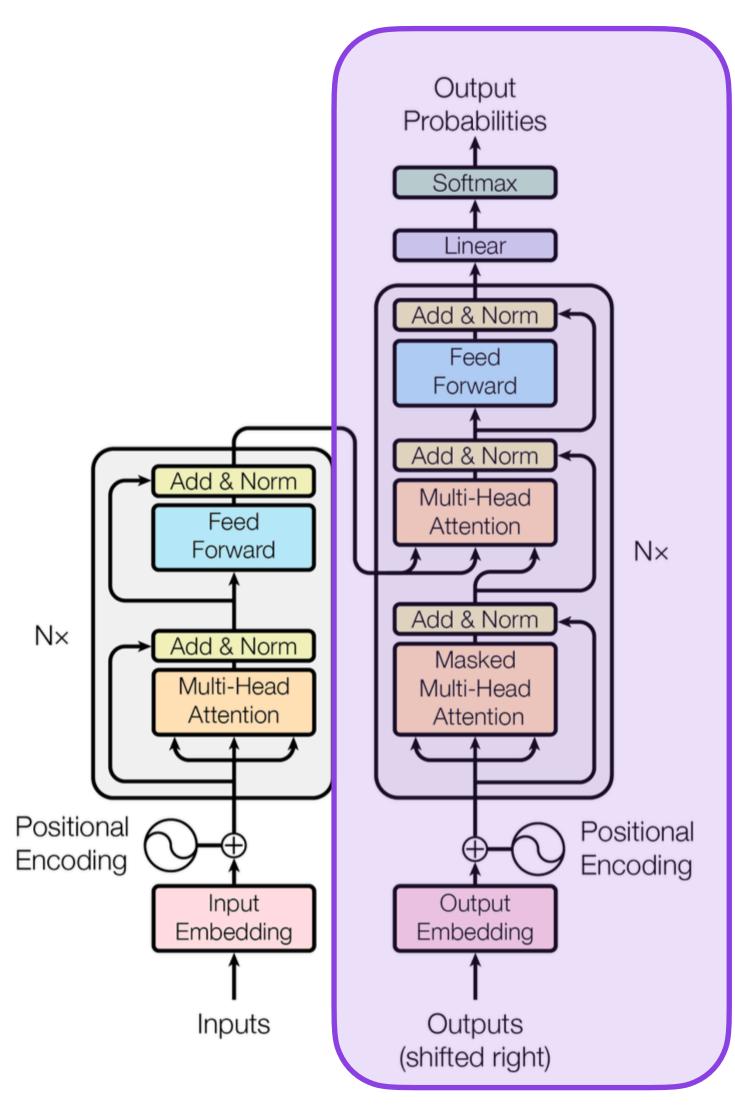
#### Instructions:

- Peel the banana and cut it into small chunks.
- Add the banana chunks, milk, yogurt, and honey (if using) to a blender.
- Blend the ingredients on high speed until smooth.
- 4. Taste and adjust sweetness if necessary.
- 5. Pour the smoothie into a glass and enjoy!

You can also add some ice cubes, or some other fruits of your choice.

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GPT is essentially the decoder part of the original transformer

Prediction Classifier Layer Norm Feed Forward Layer Norm Masked Multi Self Attention Text & Position Embed

Text

Task

https://s3-us-west-2.amazonaws.com/openai-assets/ research-covers/language-unsupervised/ language\_understanding\_paper.pdf

Figure 1: The Transformer - model architecture.

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## Feed model text from left to right, and it learns to predict the next word.

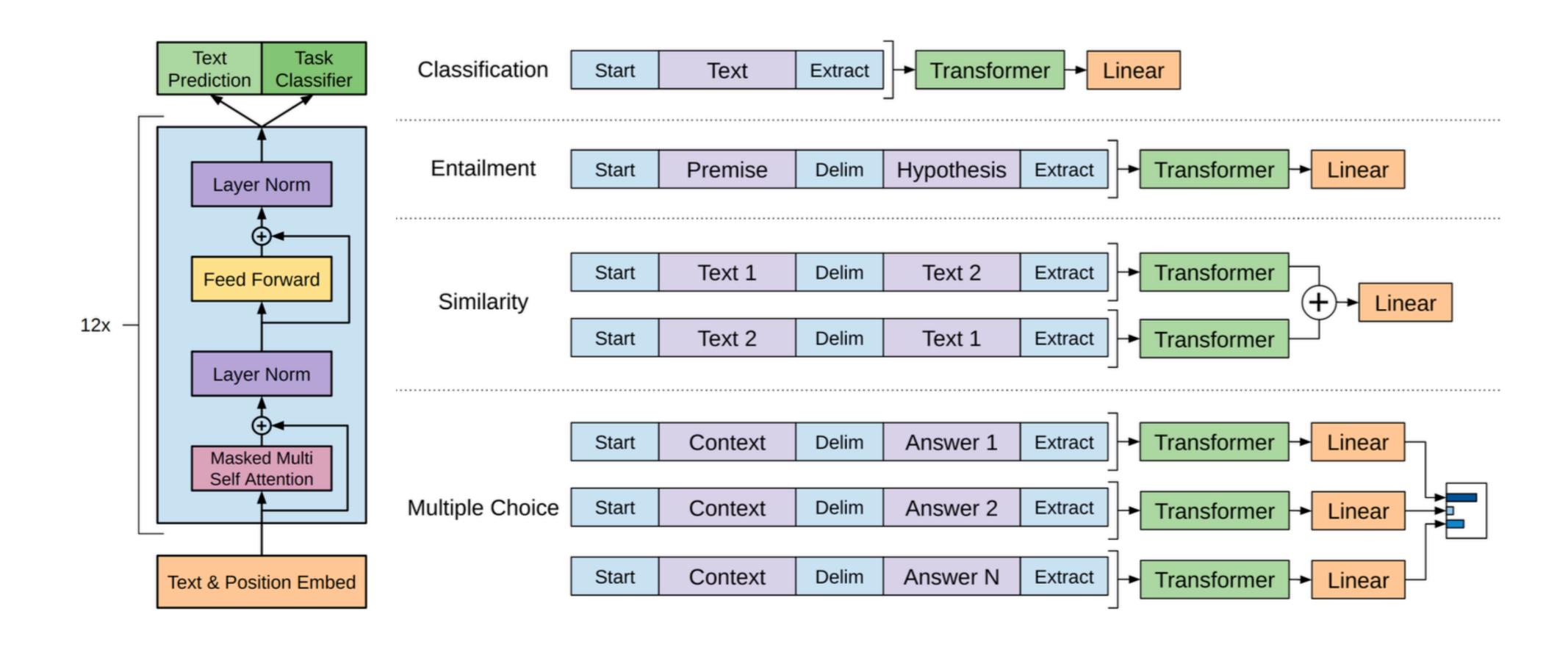


#### Self-supervised pretraining

Step 1: pretrain → Predict next word

Step 2: fine-tune

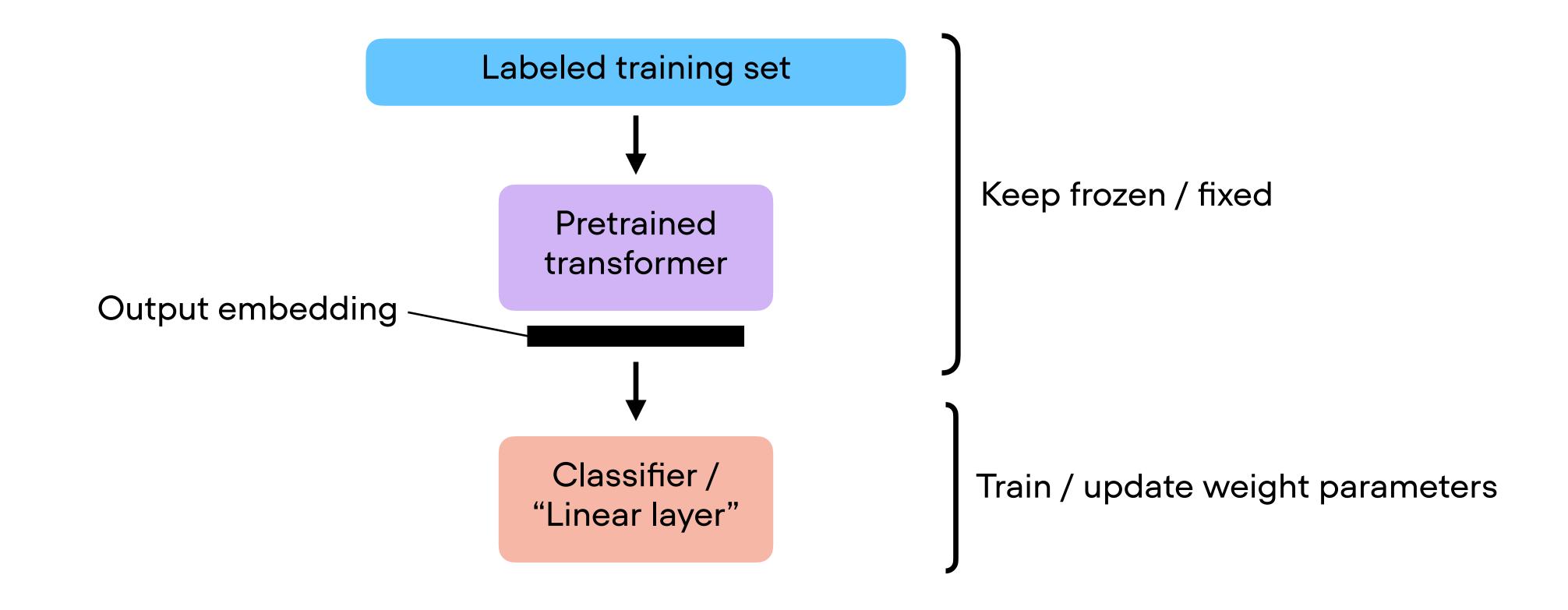
#### Fine-tune for target task



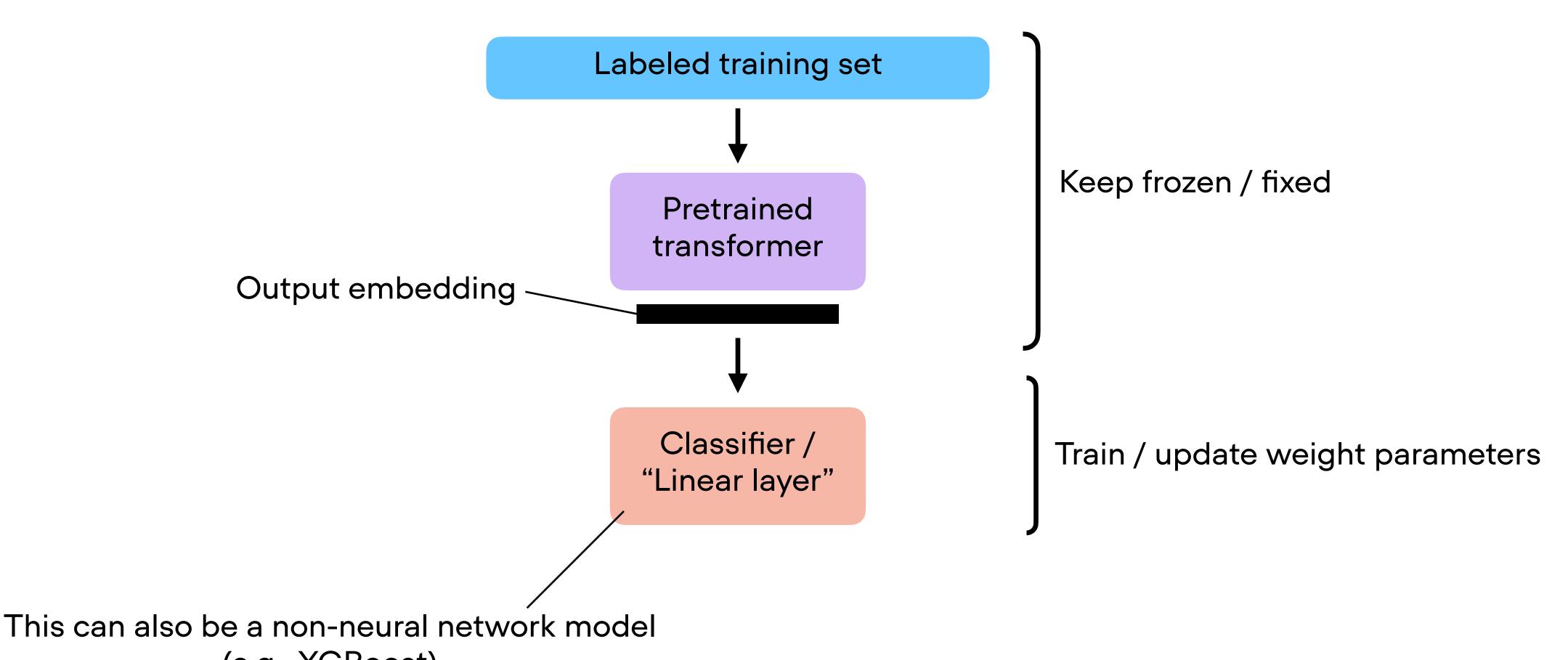
#### 2 ways of adopting a pretrained transformer

- 1) Feature-based approach
- 2) Fine-tuning approach

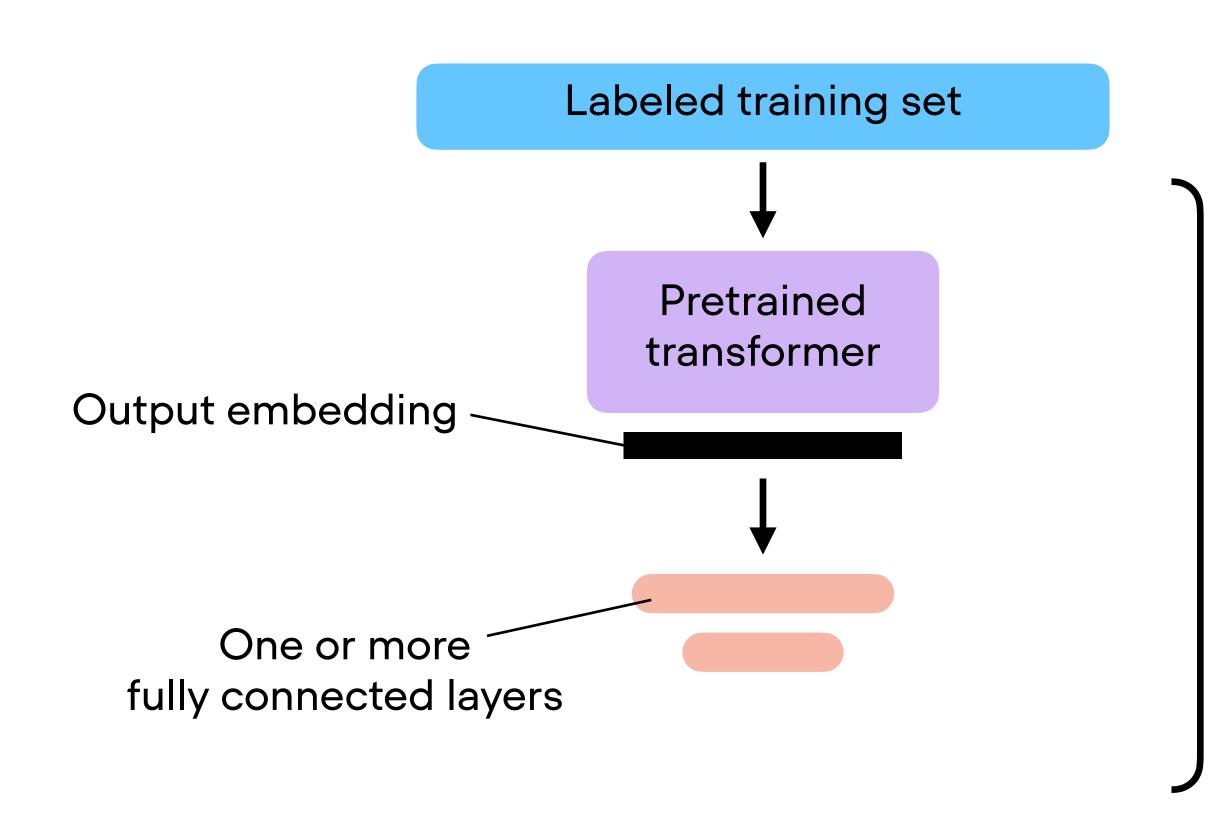
#### 1) Feature-based approach



#### 1) Feature-based approach



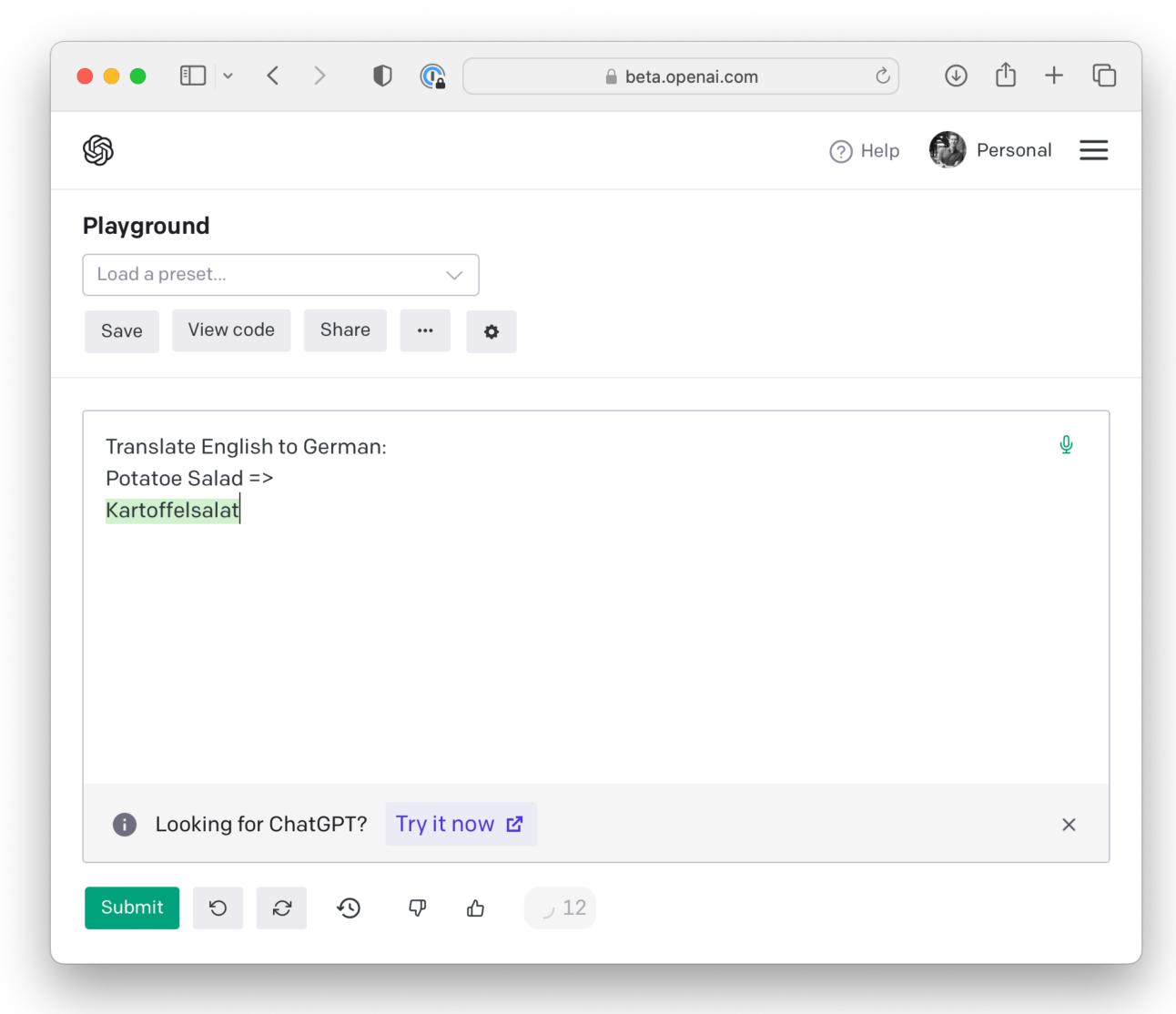
#### 1) Fine-tuning approach



Train / update last (new) or all weight parameters

## GPT 2 and 3 focused on zero- and few-shot learning via in-context learning

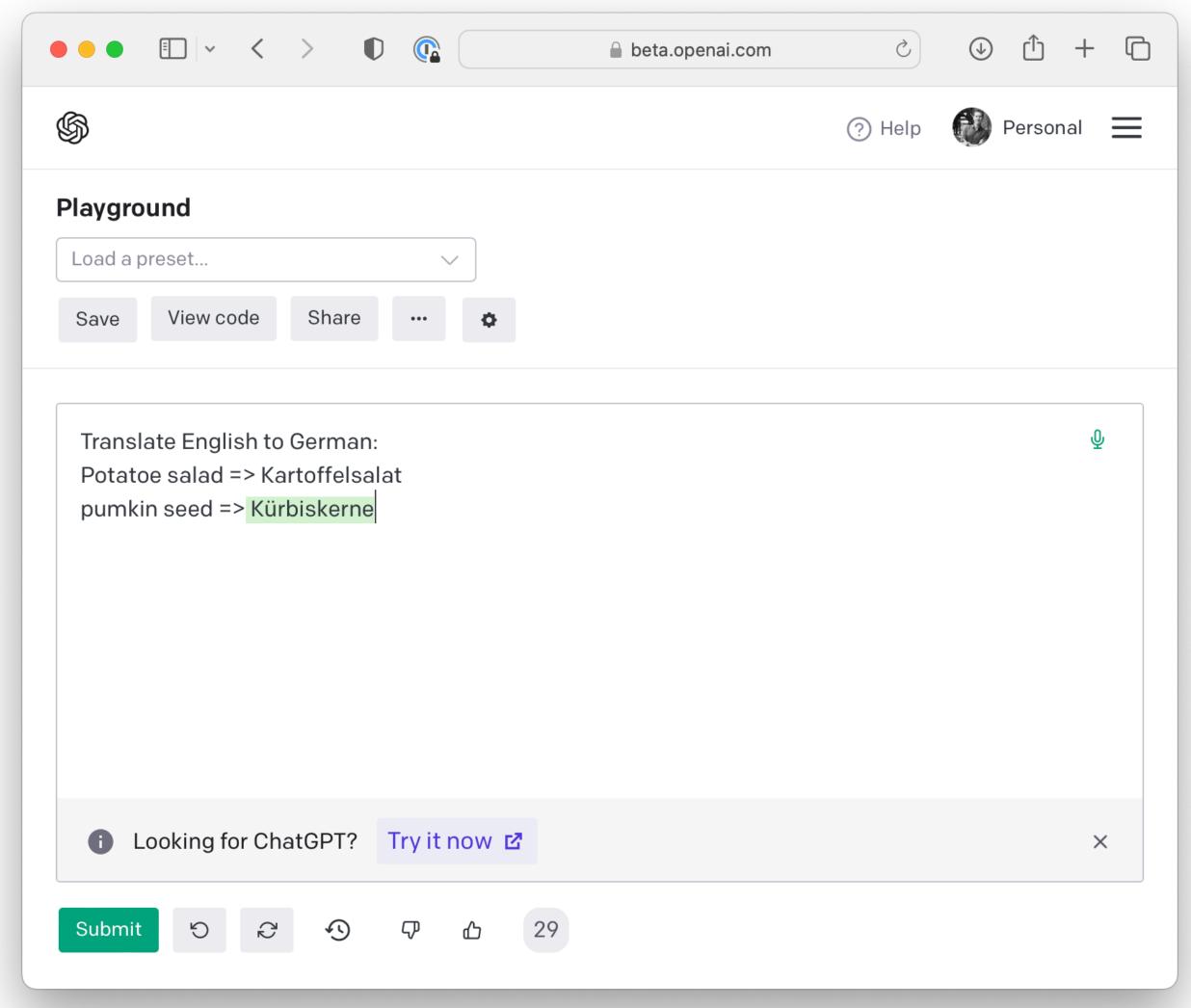
#### Zero-shot



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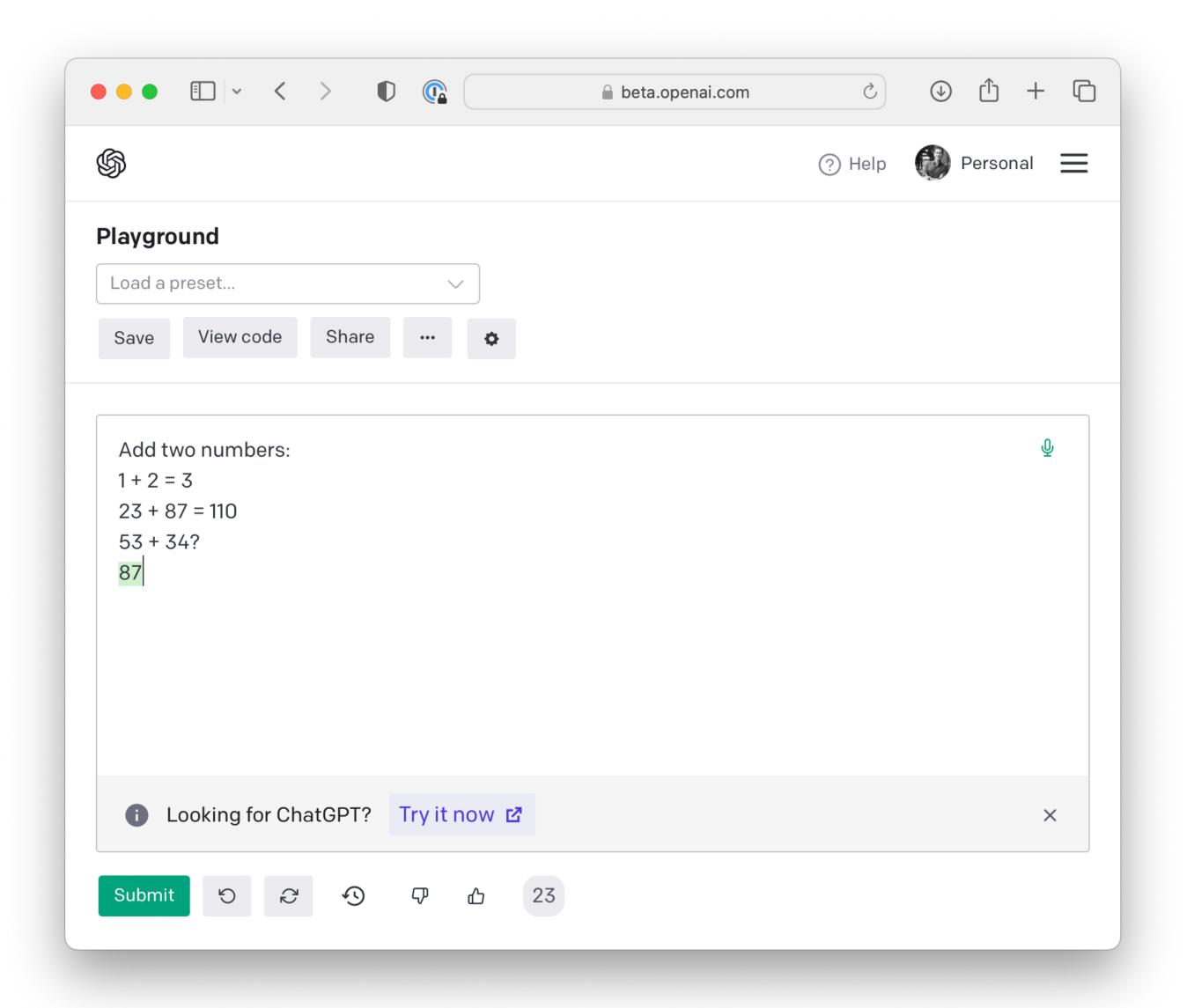
#### One-shot



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#### Few-shot



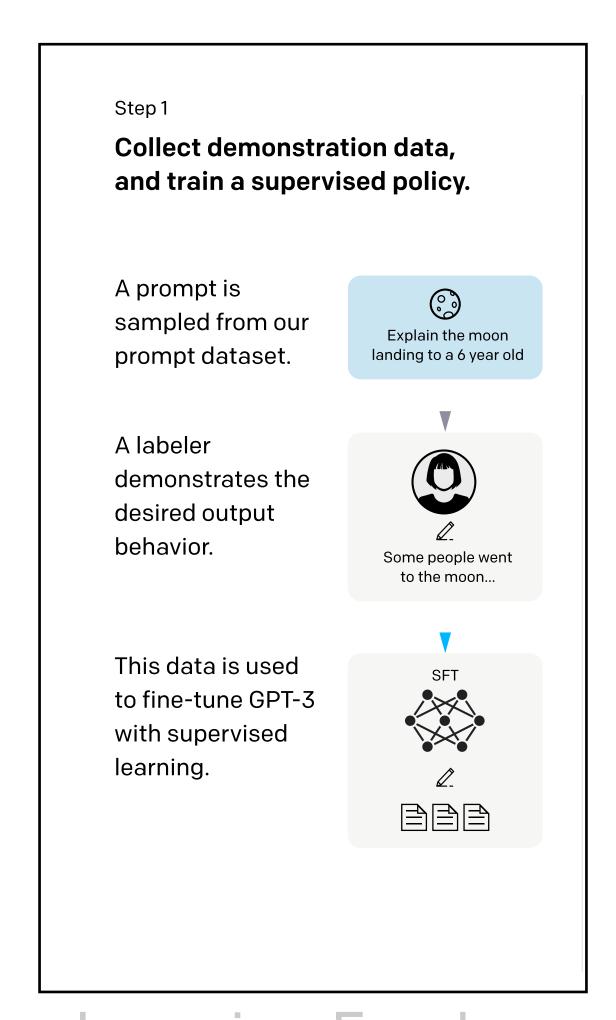
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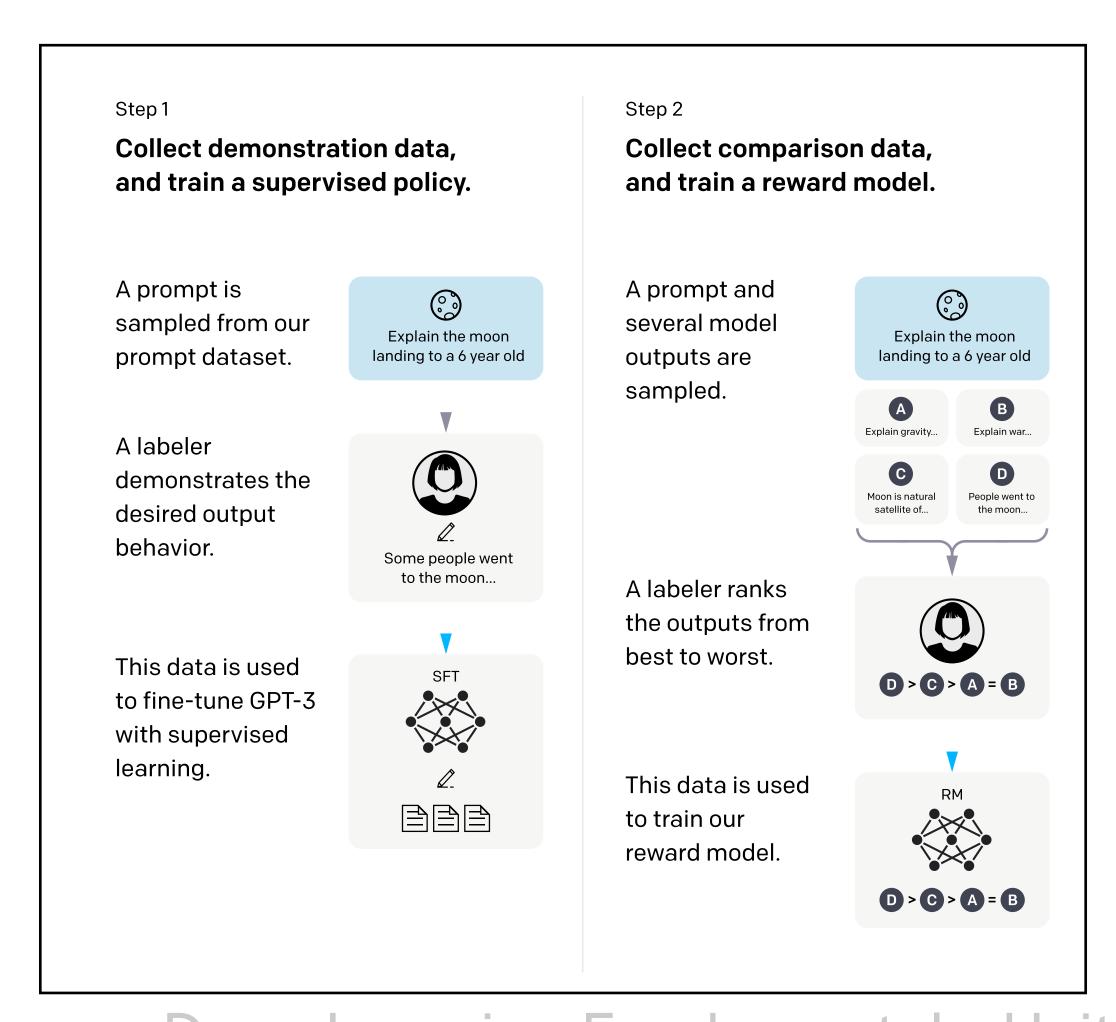
# Train a 20-billion parameter GPT model for text prediction on 3 GPU nodes with Lightning.

```
#! pip install light-the-torch
#! ltt install --upgrade git+https://github.com/Lightning-AI/lightning-LLMs
git+https://github.com/Lightning-AI/LAI-Text-Prediction-Component
#! curl https://cs.stanford.edu/people/karpathy/char-rnn/shakespeare_input.txt --create-dirs
-o ${HOME}/data/shakespeare/input.txt -C -
import lightning as L
import os, torch
from lightning_gpt import models
from lit_llms.tensorboard import (
   DriveTensorBoardLogger,
   MultiNodeLightningTrainerWithTensorboard,
from lai_textpred import default_callbacks, gpt_20b, WordDataset, error_if_local
class WordPrediction(L.LightningWork):
   def __init__(self, *args, tb_drive, **kwargs):
       super().__init__(*args, **kwargs)
       self.tensorboard_drive = tb_drive
   def run(self):
       error_if_local()
       # CONFIGURE YOUR DATA
       with open(os.path.expanduser("~/data/shakespeare/input.txt")) as f:
           text = f.read()
       train_dataset = WordDataset(text, 5)
       train_loader = torch.utils.data.DataLoader(
           train_dataset, batch_size=1, num_workers=4, shuffle=True
       # CONFIGURE YOUR MODE
       model = models.DeepSpeedMinGPT(
           vocab_size=train_dataset.vocab_size,
          block_size=int(train_dataset.block_size),
           fused_adam=False,
           model_type=None,
           **gpt_20b,
       # RUN YOUR TRAINING
       trainer = L.Trainer(
           max_epochs=2,
           limit_train_batches=250,
           precision=16,
           strategy="deepspeed_stage_3_offload",
           callbacks=default_callbacks(),
           log_every_n_steps=5,
           logger=DriveTensorBoardLogger(save_dir=".", drive=self.tensorboard_drive),
       trainer.fit(model, train_loader)
app = L.LightningApp(
   MultiNodeLightningTrainerWithTensorboard(
       cloud_compute=L.CloudCompute("gpu-fast-multi"),
```

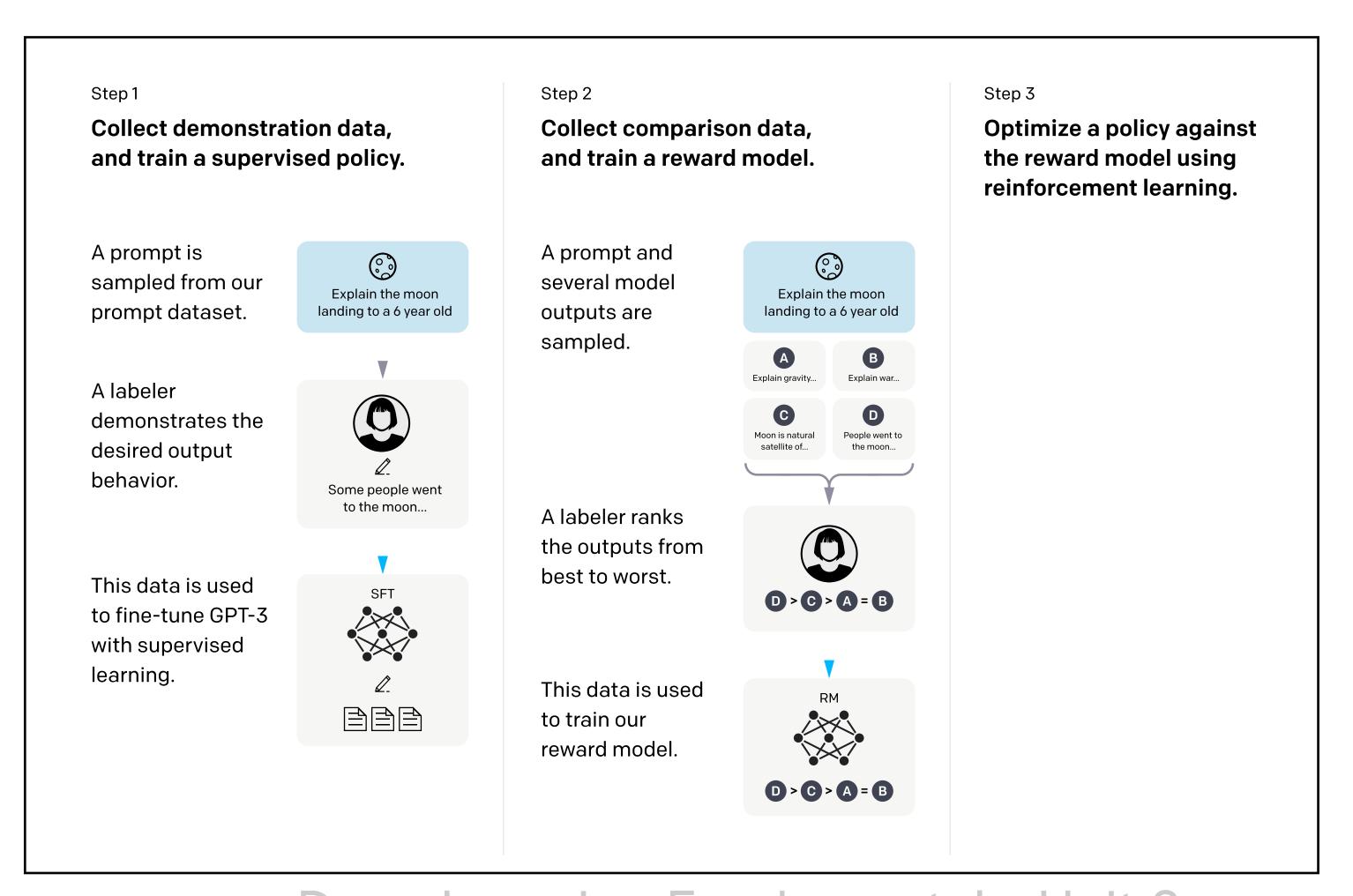
## InstructGPT and ChatGPT are Additionally Trained on Human Feedback



## InstructGPT and ChatGPT are Additionally Trained on Human Feedback



## InstructGPT and ChatGPT are Additionally Trained on Human Feedback



Sebastian Raschkaining language models to follow instructions with human feedback, https://arxiv.org/abs/2203.02155

#### Next: A Large Language Model for Classification