Introduction to RLHF

REINFORCEMENT LEARNING FROM HUMAN FEEDBACK (RLHF)



Mina Parham Al Engineer

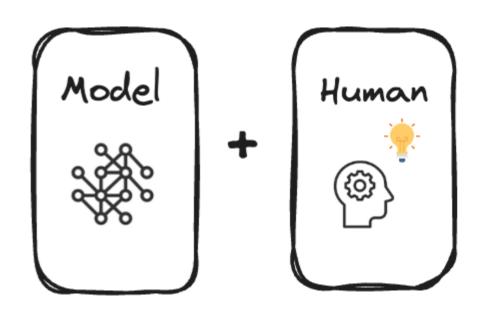


Welcome to the course!

• Instructor: Mina Parham

- Al Engineer
- Large Language Models (LLMs)
- Reinforcement Learning from Human Feedback (RLHF)

 Topic: Reinforcement Learning from Human Feedback (RLHF)

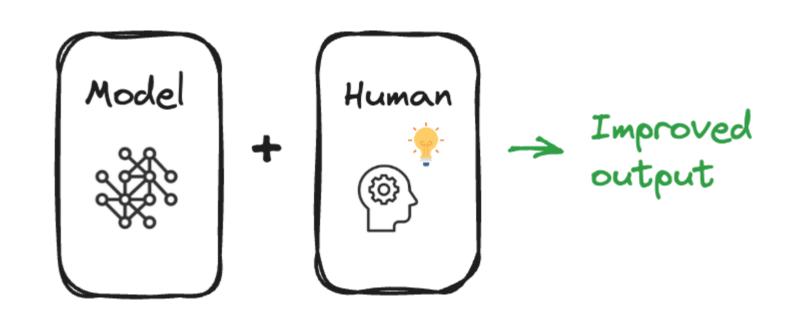


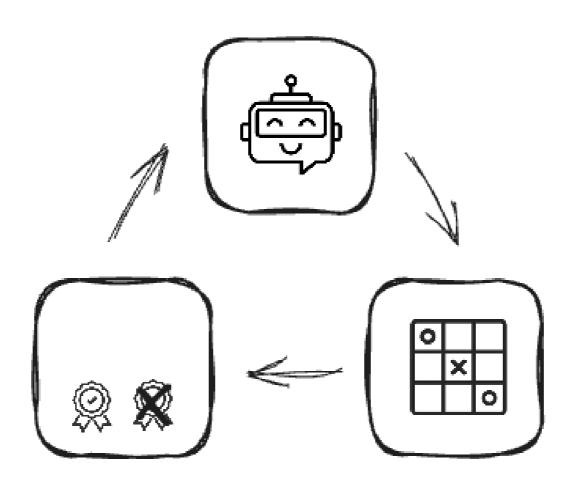
Welcome to the course!

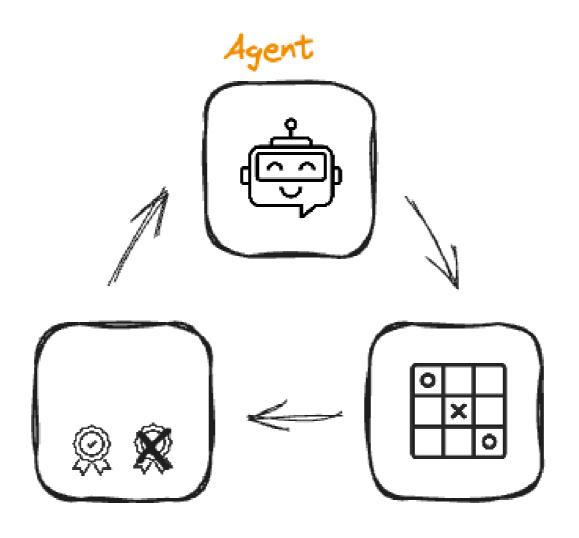
• Instructor: Mina Parham

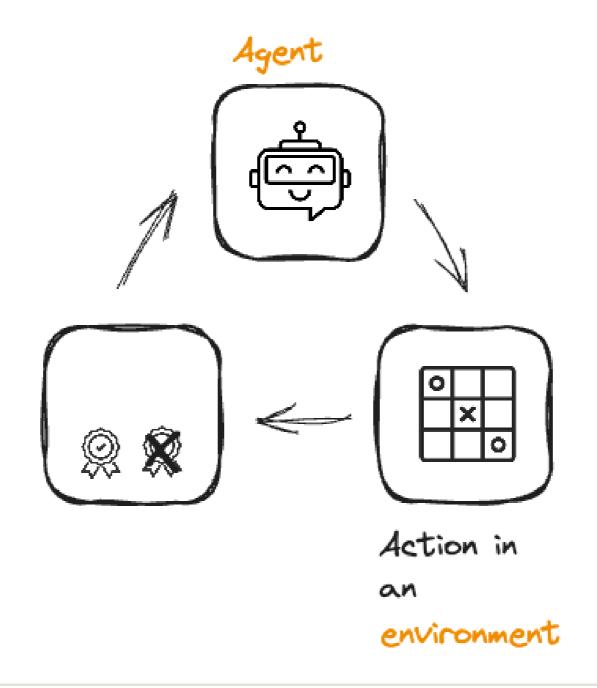
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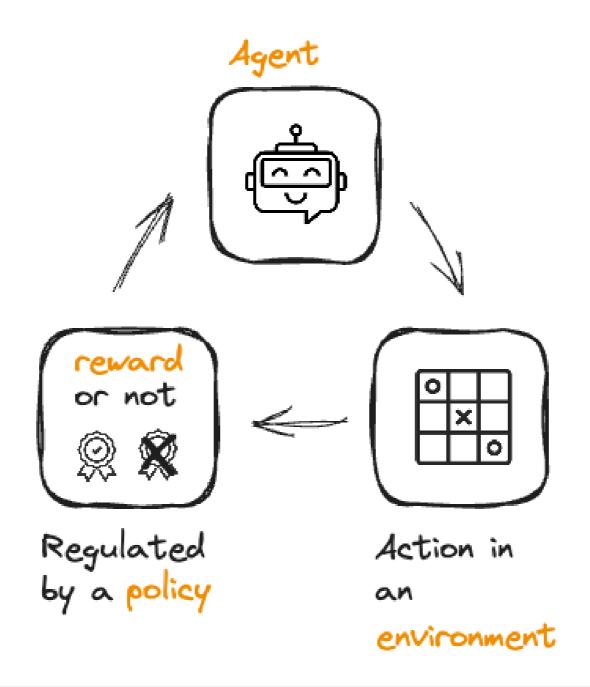




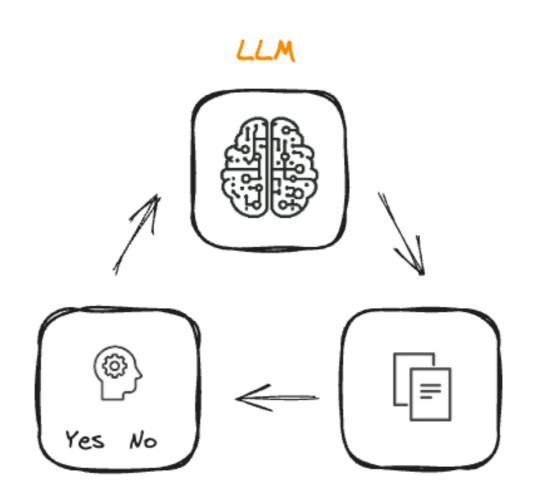




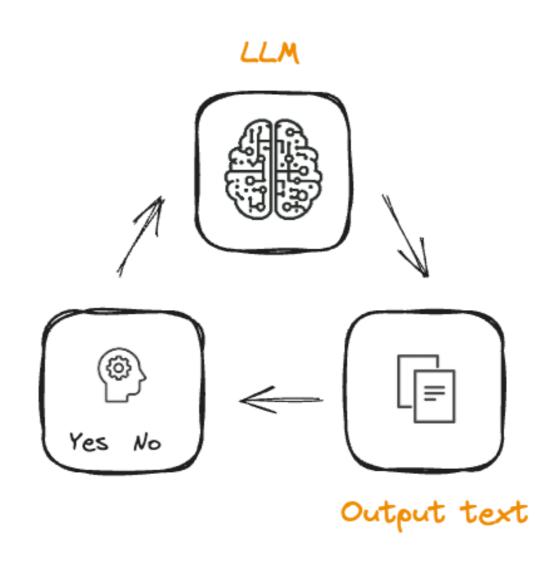




From RL to RLHF

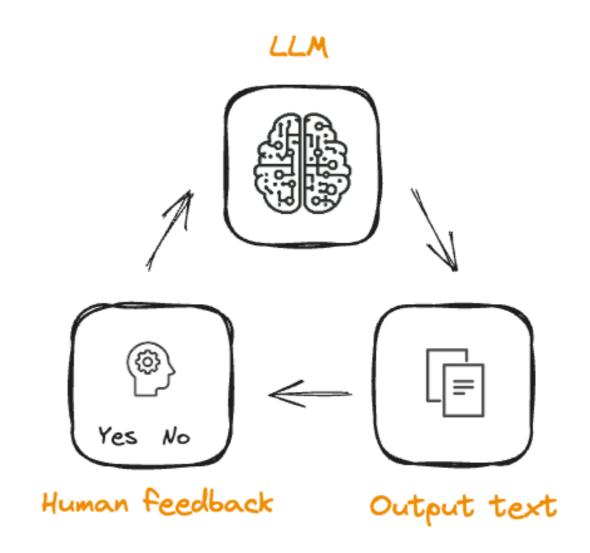


From RL to RLHF

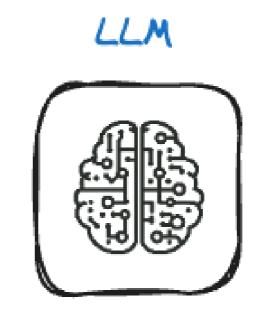


From RL to RLHF

- Training the reward model
- Alignment with human preferences



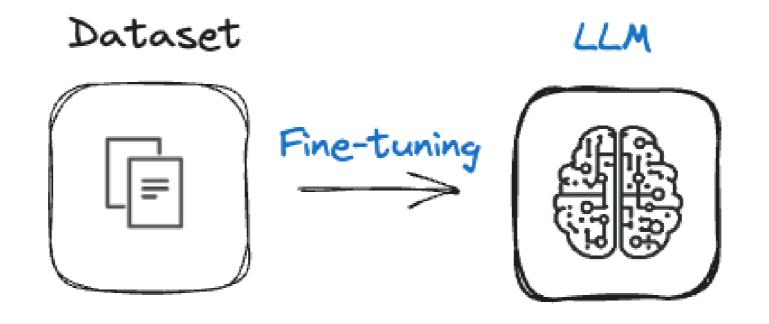
LLM fine-tuning in RLHF





LLM fine-tuning in RLHF

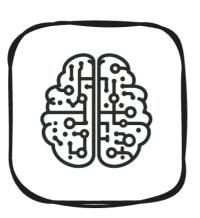
• Training the initial LLM



Who wrote "Romeo and Juliet"?

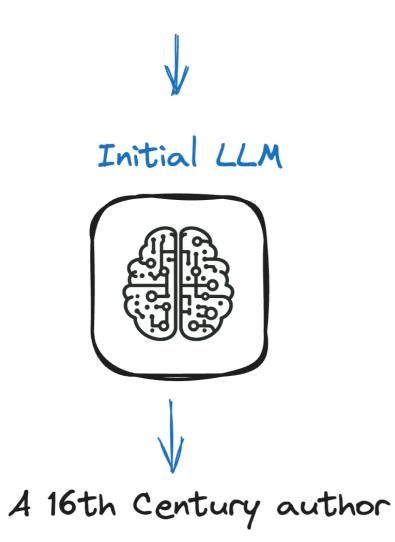


Initial LLM

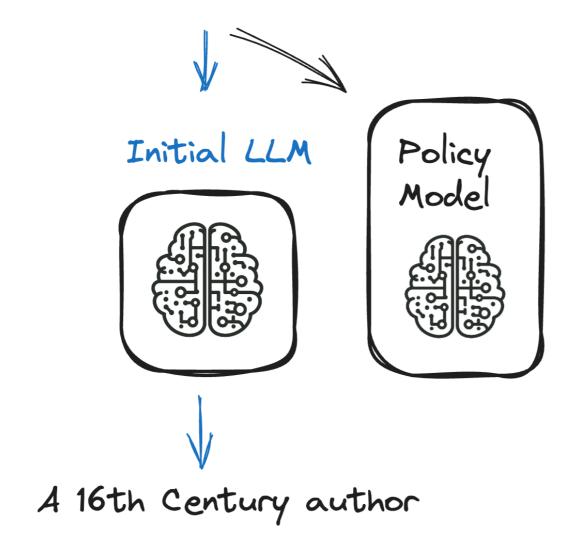


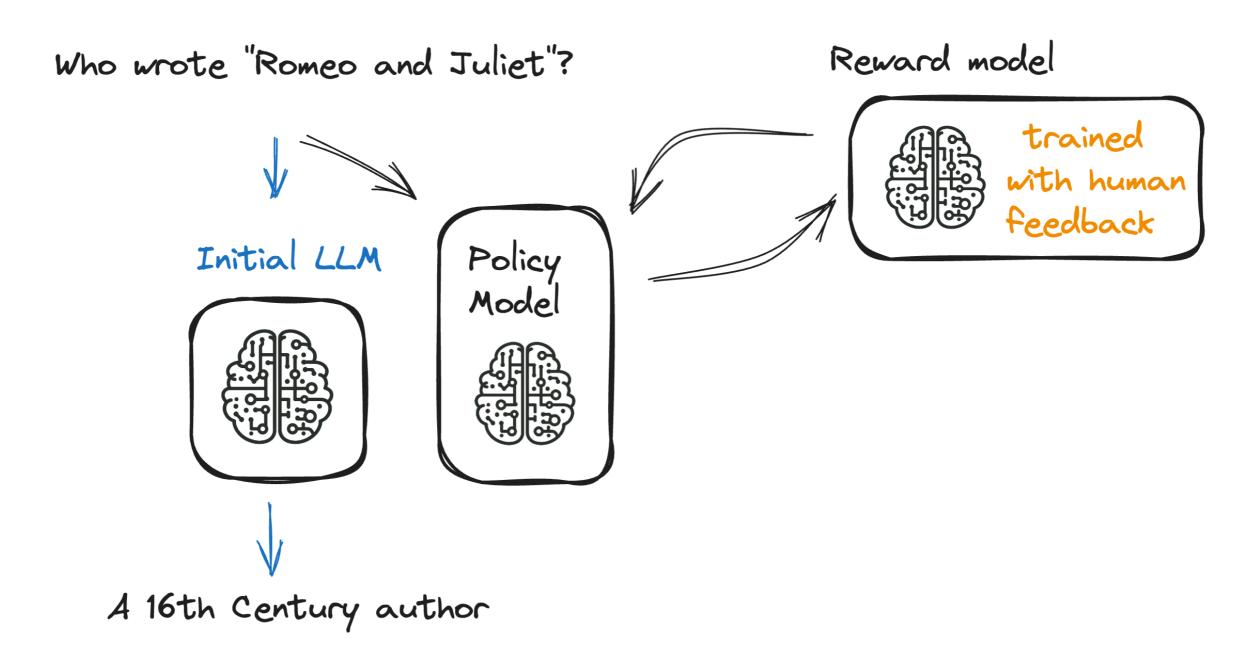


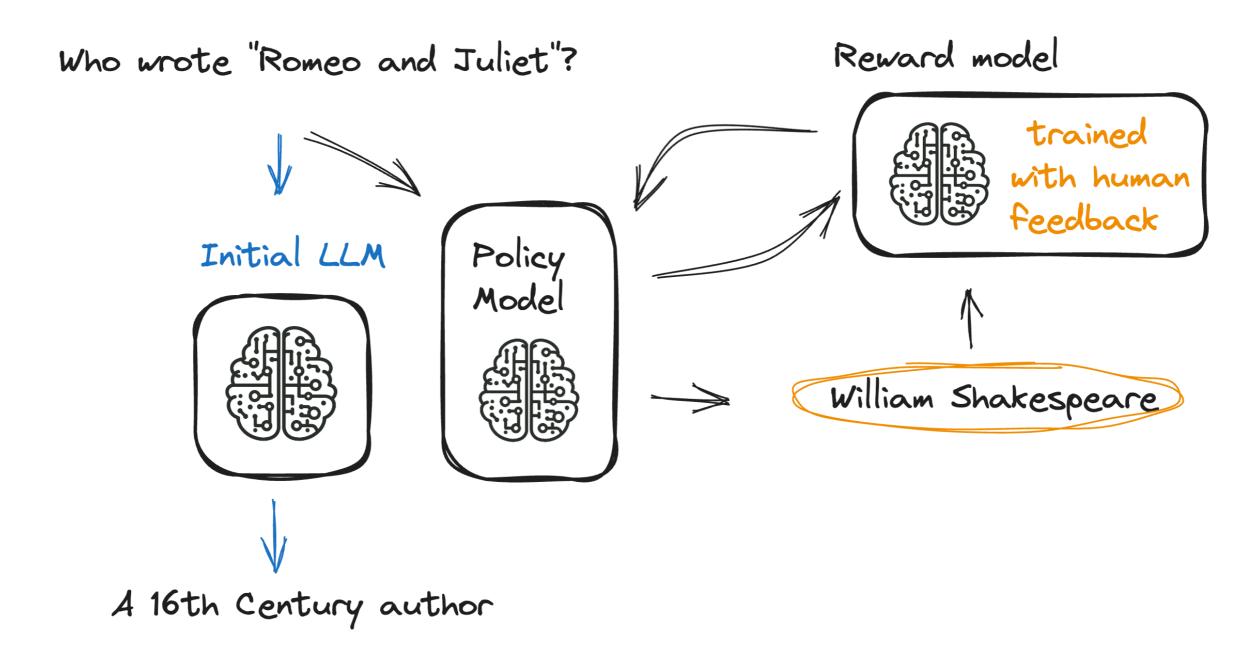
Who wrote "Romeo and Juliet"?

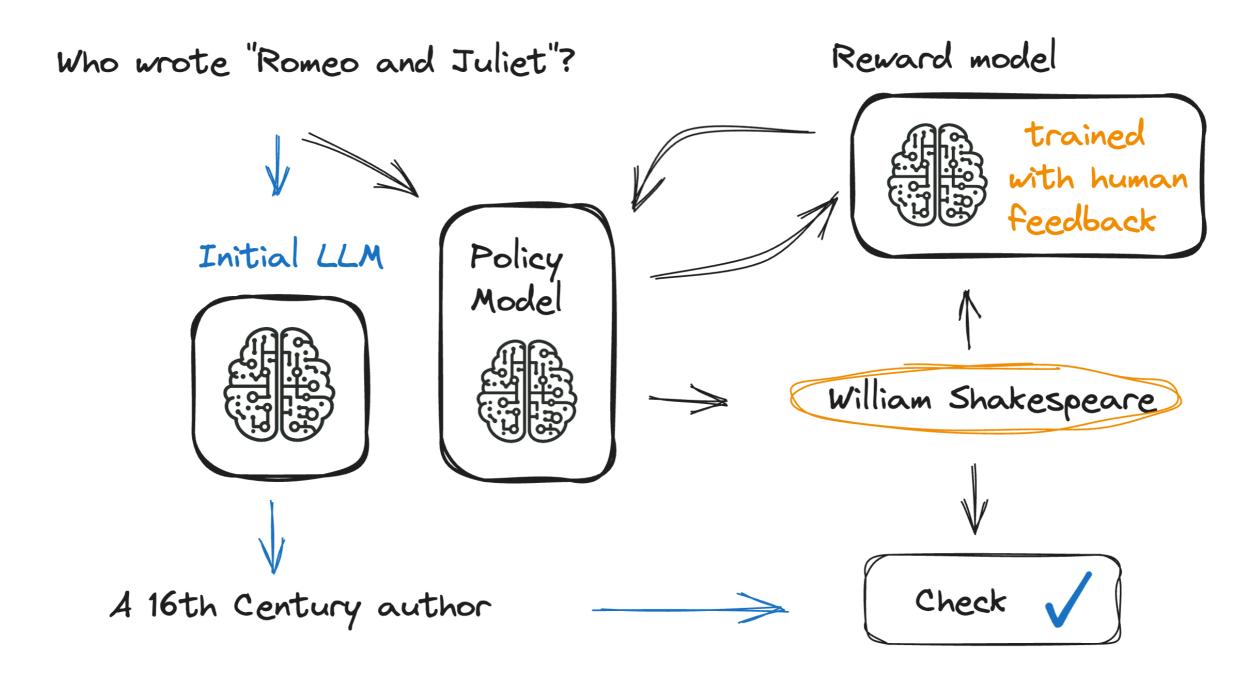


Who wrote "Romeo and Juliet"?









Interacting with RLHF-tuned LLMs

Pre-trained RLHF models on Hugging Face []

```
from transformers import pipeline
text_generator = pipeline('text-generation', model='lvwerra/gpt2-imdb-pos-v2')
# Provide a review prompt
review_prompt = "This is definitely a"
# Generate the continuation
output = text_generator(review_prompt, max_length=50)
#Print the generated text
print(output[0]['generated_text'])
```

This is definitely a crucial improvement.



Interacting with RLHF-tuned LLMs

```
from transformers import pipeline, AutoModelForSequenceClassification, AutoTokenizer
# Instantiate the pre-trained model and tokenizer
model = AutoModelForSequenceClassification.from_pretrained("lvwerra/distilbert-imdb")
tokenizer = AutoTokenizer.from_pretrained("lvwerra/distilbert-imdb")
# Use pipeline to create the sentiment analyzer
sentiment_analyzer = pipeline('sentiment-analysis', model=model, tokenizer=tokenizer)
# Pass the text to the sentiment analyzer and print the result
sentiment = sentiment_analyzer("This is definitely a crucial improvement.")
print(f"Sentiment Analysis Result: {sentiment}")
```

positive



Let's practice!

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Exploring pretrained LLMs

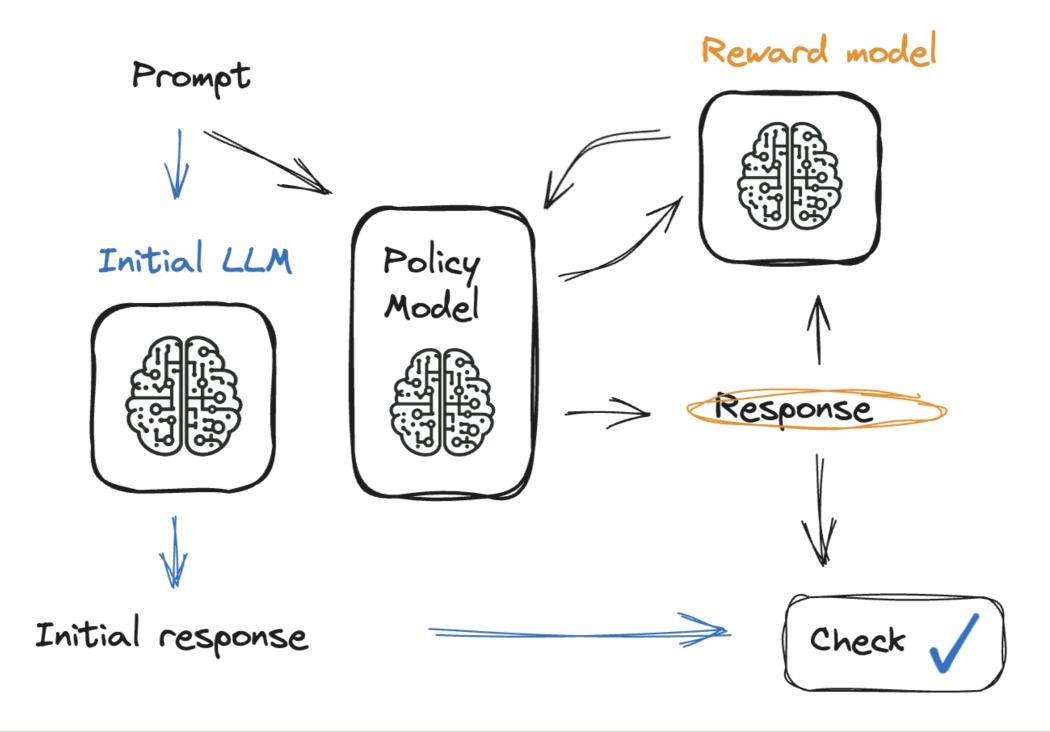
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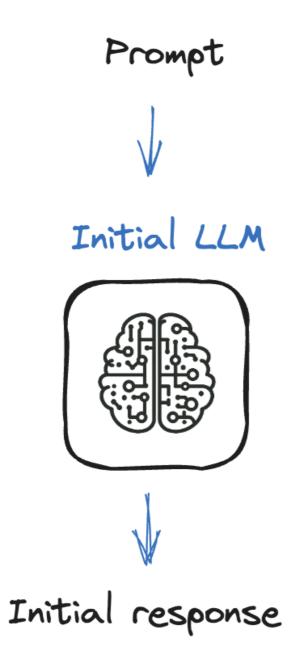
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The importance of fine-tuning

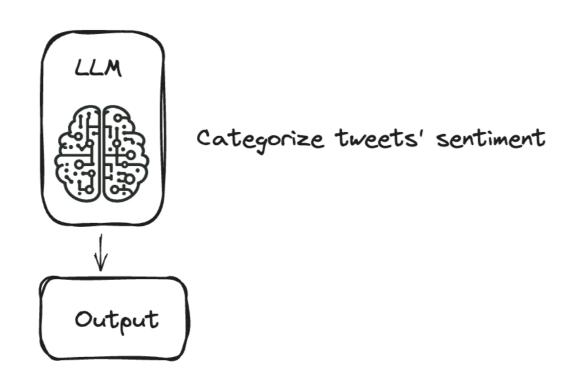


The importance of fine-tuning

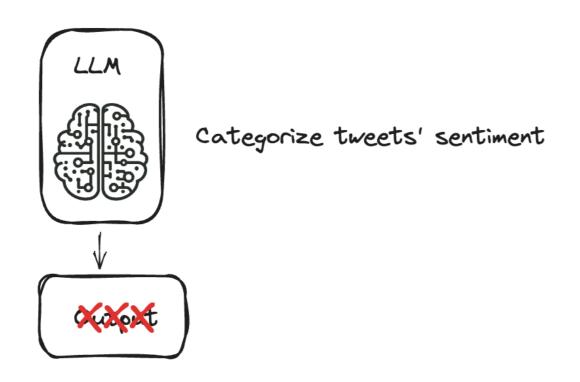




A step-by-step guide to fine-tuning an LLM

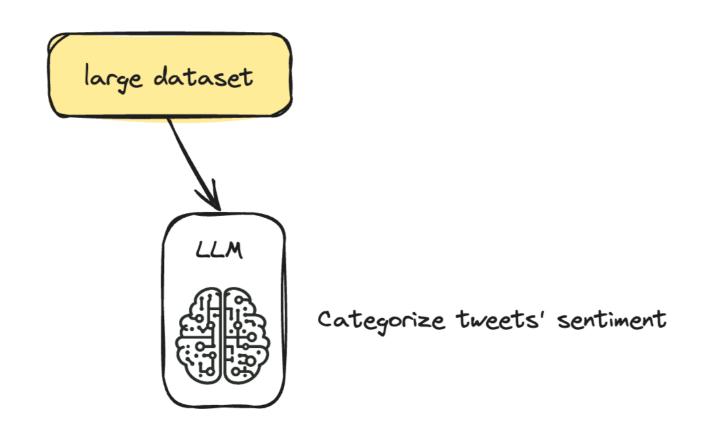


A step-by-step guide to fine-tuning an LLM





A step-by-step guide to fine-tuning an LLM





Step 1: load the data to use

```
from datasets import load_dataset
import pandas as pd

# `load_dataset` simplifies loading and preprocessing datasets from various sources
# It provides easy access to a wide range of datasets with minimal setup
dataset = load_dataset("mteb/tweet_sentiment_extraction")
df = pd.DataFrame(dataset['train'])
```

	id	text	label	label_text
0	cb774db0d1	I'd have responded, if I were going	1	neutral
1	549e992a42	Sooo SAD I will miss you in San Diego!!!	0	negative
2	08ac60f138	my boss is bullying me	0	negative

Step 2: choose a pre-trained model

```
from transformers import AutoModelForCausalLM

# AutoModelForCausalLM simplifies loading and switching models
model = AutoModelForCausalLM.from_pretrained("openai-gpt")
```

• Causal models: previous tokens "cause" subsequent ones

Step 3: tokenizer

```
from transformers import AutoTokenizer

# `AutoTokenizer` loads the correct tokenizer for the specified model
tokenizer = AutoTokenizer.from_pretrained("openai-gpt")
tokenizer.add_special_tokens({'pad_token': '[PAD]'})
model.resize_token_embeddings(len(tokenizer))
```

Padding: to have equal-sized batches of text

Step 3: tokenizer

```
def tokenize_function(examples):
    tokenized = tokenizer(examples["content"], padding="max_length", truncation=True
    return tokenized

tokenized_datasets = dataset.map(tokenize_function, batched=True)
```

Batched parameter: for faster processing

Step 4: fine-tune using the Trainer method

```
training_args = TrainingArguments(
   output_dir="test_trainer",
   per_device_train_batch_size=1,
   per_device_eval_batch_size=1,
   gradient_accumulation_steps=4)
trainer = Trainer(
  model=model,
   args=training_args,
   train_dataset=tokenized_dataset["train"],
   eval_dataset=tokenized_dataset["test"])
trainer.train()
```

Let's practice!

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Preparing data for RLHF

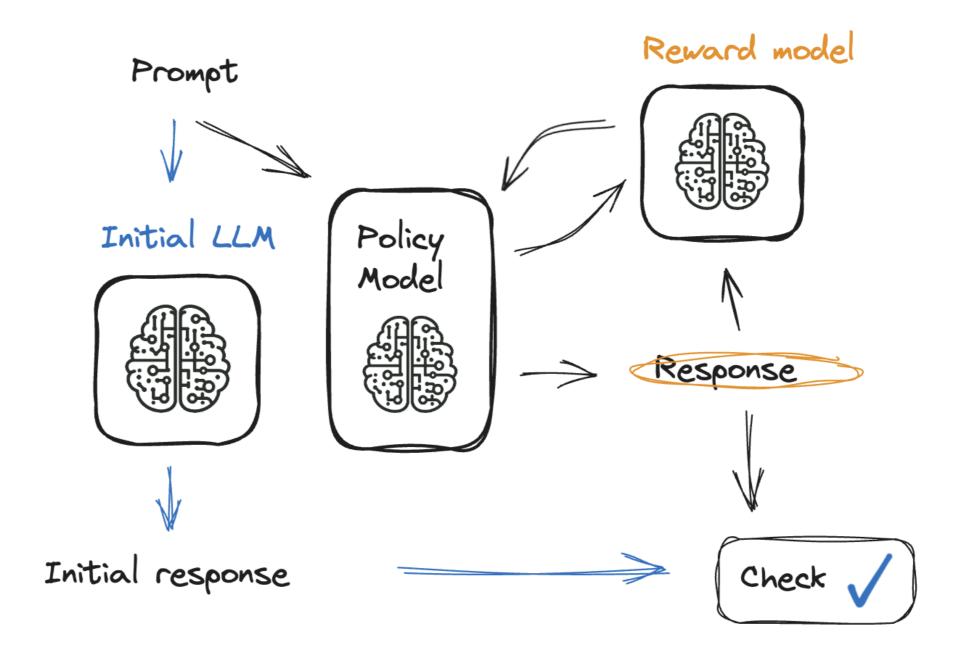
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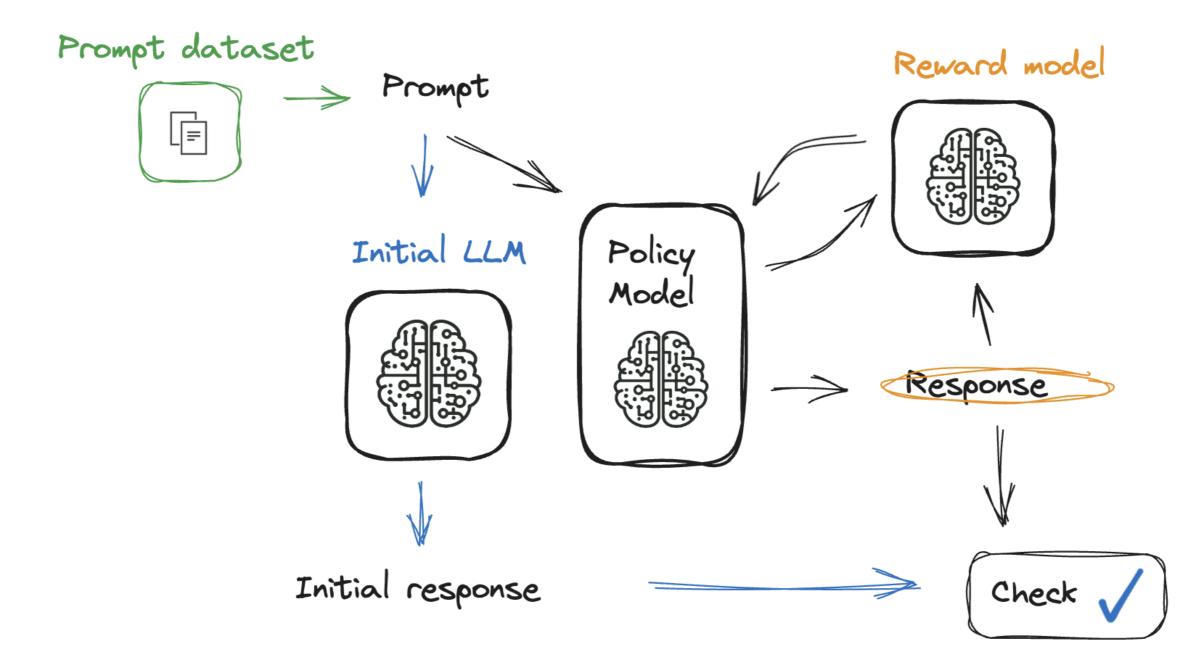
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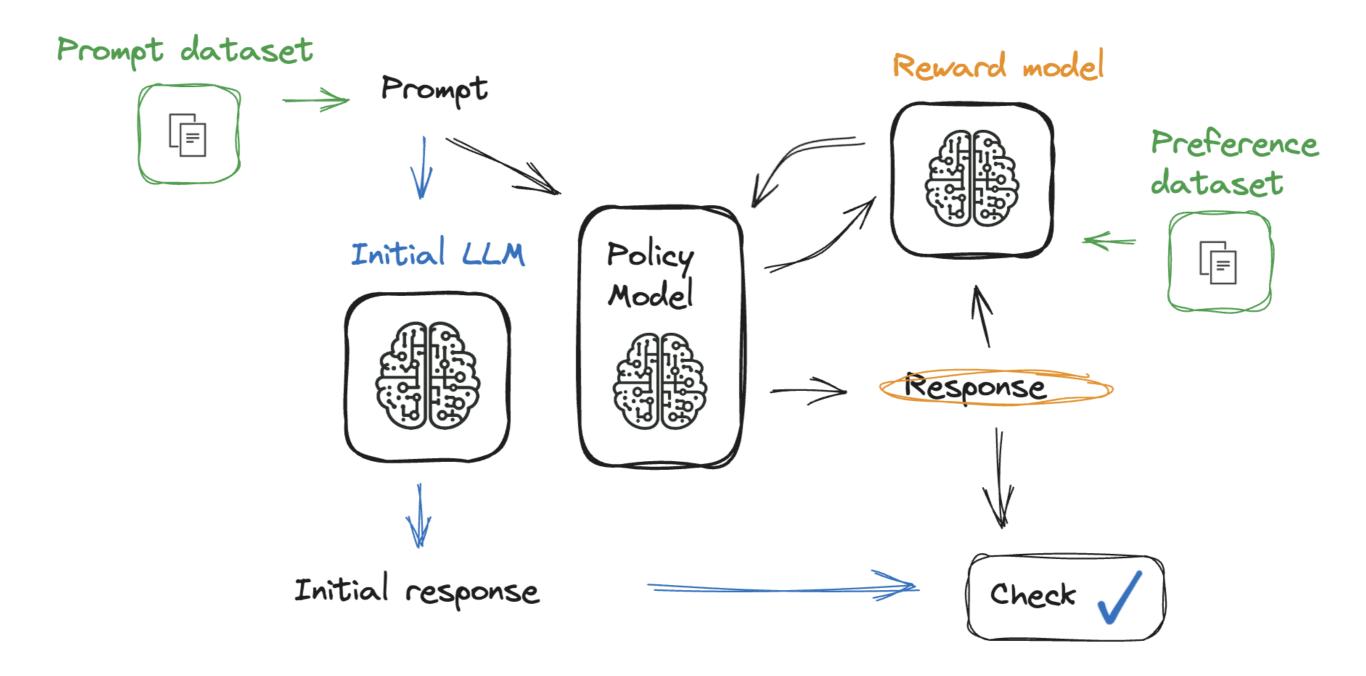
Preference vs. Prompt datasets



Preference vs. Prompt datasets



Preference vs. Prompt datasets



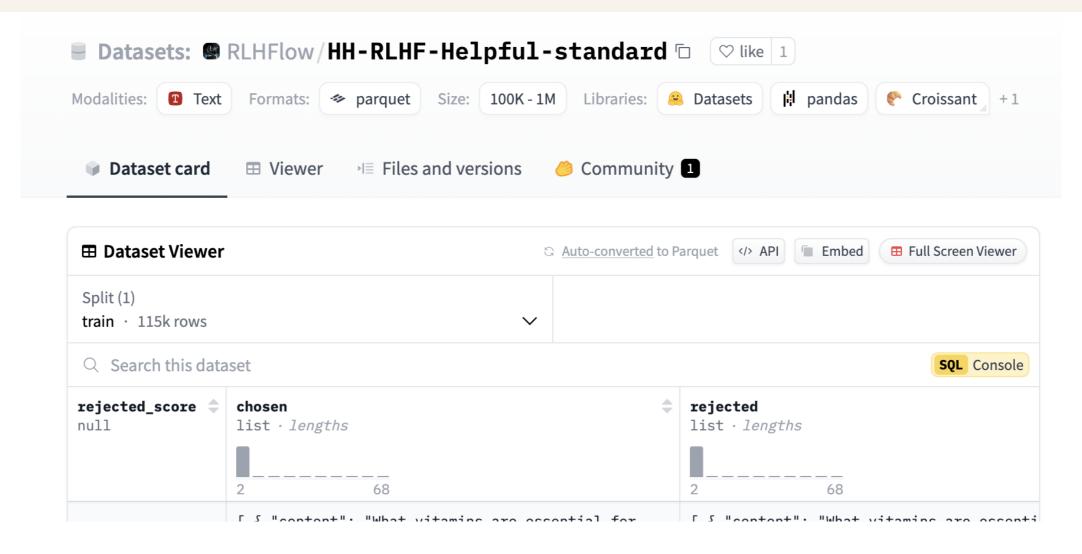
Prompt dataset

- Questions for the model
- Can be found on Hugging Face datasets

```
'How important is climate change?'
```

- Might need to extract the prompt
- Look for markers such as: Input= , {{Text}}: , ###Human:

Exploring the preference dataset





Processing the preference dataset

```
def extract_prompt(text):
    # Extract the prompt as the first element in the list
    prompt = text[0]["content"]
    return prompt
```

```
# Apply the extraction function to the dataset
preference_data_with_prompt = preference_data.map(
    lambda example: {**example, 'prompt': extract_prompt(example['chosen'])}
)
```

• The way prompts are extracted is different for different datasets

Final preference dataset

```
sample = preference_data_with_prompt.select(range(1))
sample['prompt']
'What vitamins are essential for the body to function?'
sample['chosen']
[ { "content": "What vitamins are essential for the body to function?", "role":
   "user" }, { "content": "There are some very important vitamins that ensure the
```

proper functioning of the body, including Vitamins A, C, D, E, and K along ...}]

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