CMSC 473/673 Natural Language Processing

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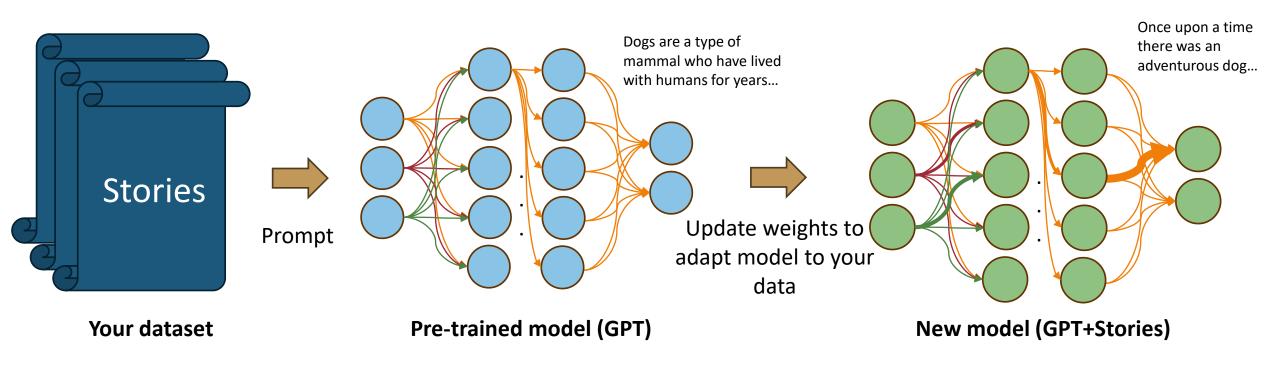
Learning Objectives

Distinguish between finetuning and prompting

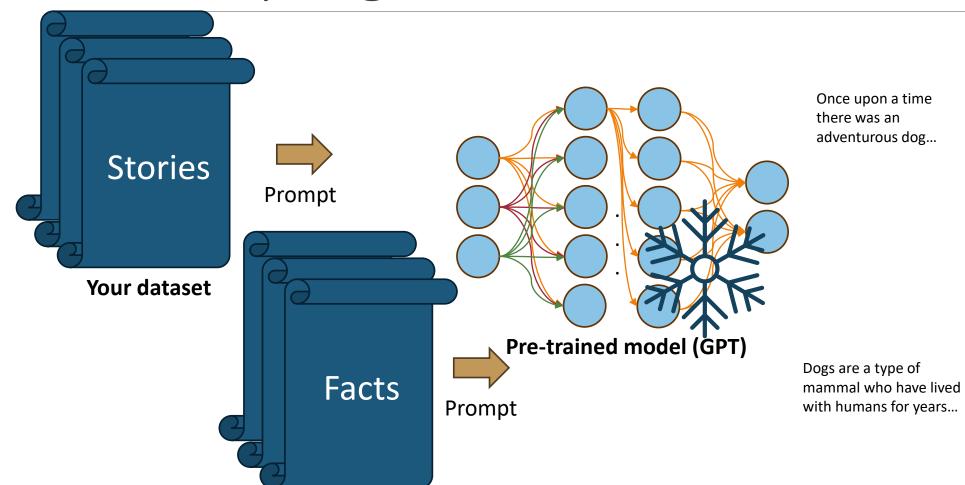
Distinguish between few-shot and zero-shot prompting

Examine the ways GPT's parameters affect sampling

Finetuning

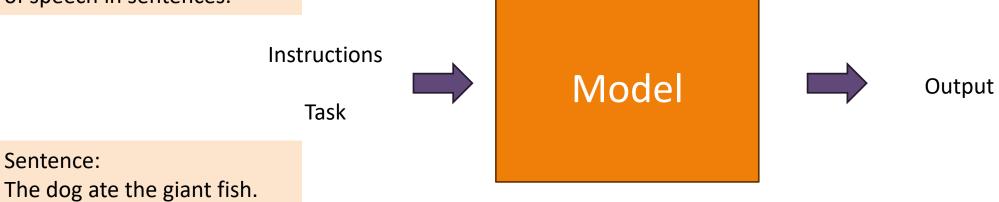


Prompting



Zero-shot Prompting

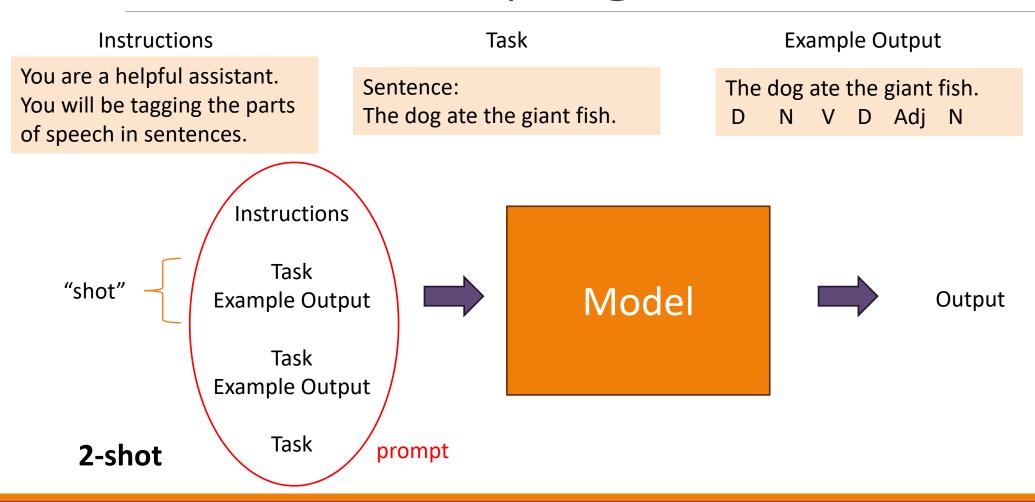
You are a helpful assistant. You will be tagging the parts of speech in sentences.



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Sentence:

Few-shot Prompting



Prompt Engineering



"A child playing on a sunny happy beach, their laughter as they build a simple sandcastle, emulate Nikon D6 high shutter speed action shot, soft yellow lighting."

Generated with Midjourney.

via https://zapier.com/blog/ai-art-prompts/

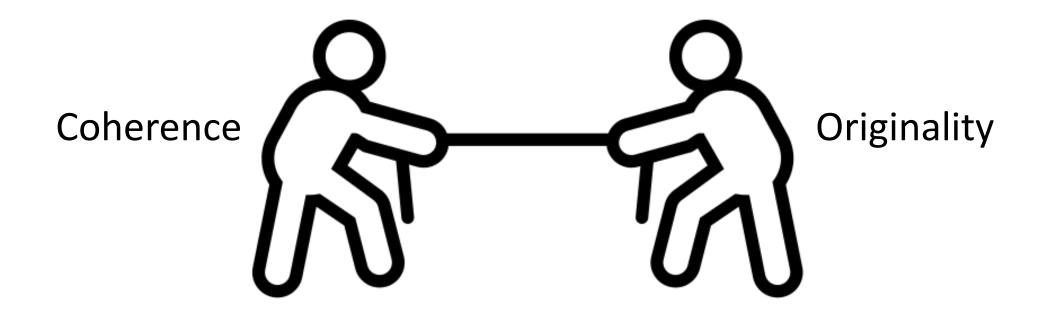
Need to be really specific (also match the training data)

Dealing with any language models

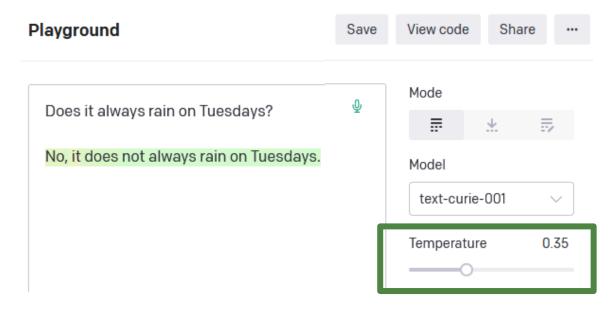
Likelihoods → Not cause & effect

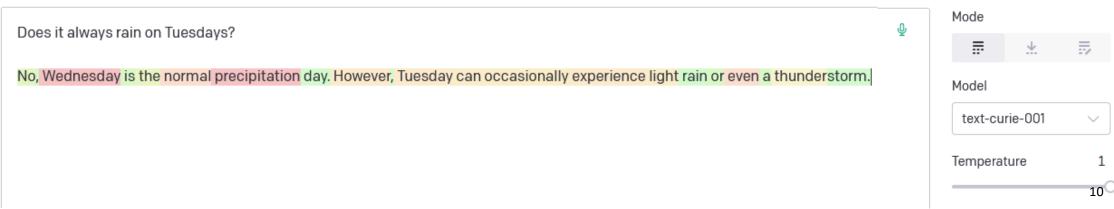
What is probable might not be possible.

Lara's Language Model Tradeoff



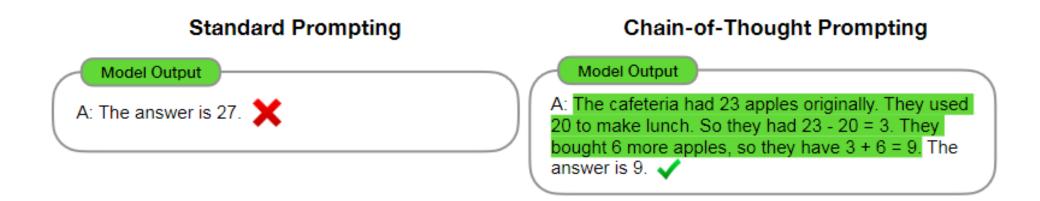
There's even an explicit knob in GPT



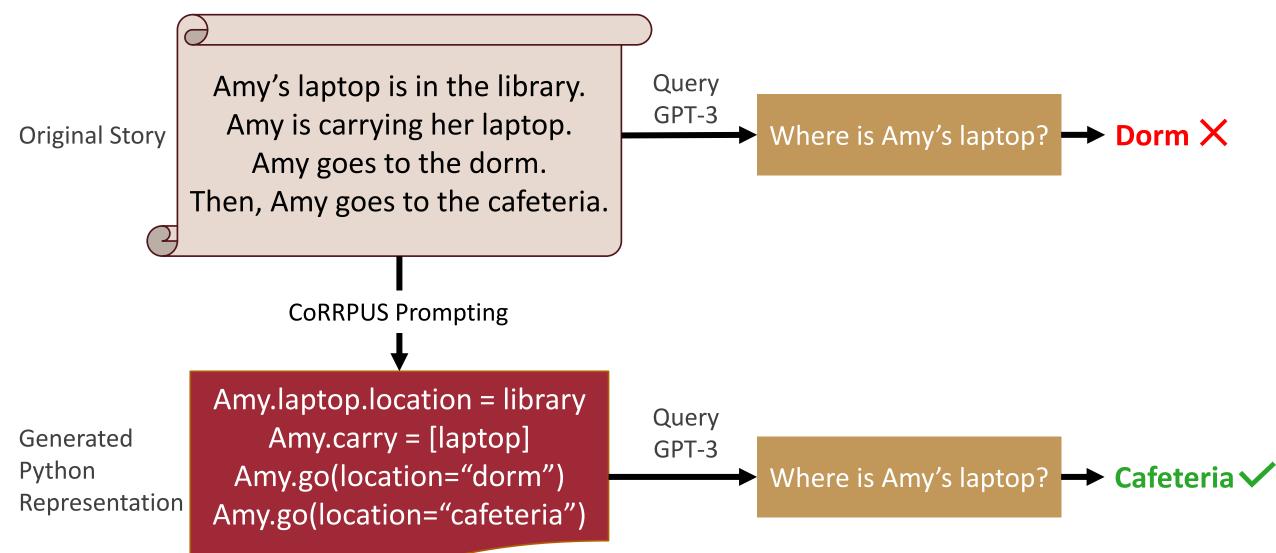


Chain-of-Thought Prompting

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



CORRPUS (Code Representations to Reason & Prompt over for Understanding in Stories)



Dong, Y. R., Martin, L. J., & Callison-Burch, C.

CorrPUS Chain-of-Thought Prompting

Three versions that are initialized the same:

Comment

```
def story(self):
```

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```
## Mary moved to the bathroom.
self.Mary.location = "bathroom"
## Mary got the football there.
self.Mary.inventory.append("football")
```

Specific Functions

```
self.Mary_moved_to_the_bathroom()
self.Mary_got_the_football_there()
self.John_went_to_the_kitchen()
self.Mary_went_back_to_the_garden()
```

```
def Mary_moved_to_the_bathroom()
        self.Mary.location="bathroom"
def Mary_got_the_football_there():
```

Abstract Functions

```
def story(self):
```

```
## Mary moved to the bathroom.
self.go(character=self.Mary,
location = "bathroom")
...
```

Tested On 2 Tasks

bAbl (Weston et al. 2015)

Task 2: Stories tracking objects that characters carry

Re³ (Yang et al. 2022)

- Identifying inconsistencies in stories (e.g., descriptions of characters' appearances, relationships)
- Stories were generated from a list of facts (the premise). They also generated premises with a contradiction.

bAbl (Weston et al. 2015)

Method	# Shot	Accuracy 个
Random	-	25%
GPT-3	1	56.5%
Chain of Thought (Creswell et al. 2022)	1	46.4%
Selection-Inference (Creswell et al. 2022)	1	29.3%
Dual-System (Nye et al. 2021)	10	100%
CoRRPUS (comment)	1	67.0%
CorrPUS (specific)	1	78.7%
CorrPUS (abstract)	1	99.1%

Re³

The task is to see what stories match what premises based on the facts extracted from both.

Joan Westfall premise

Joan Westfall in story

Attribute	Value	entails	Attribute	Value
Gender	Female	entails	Gender	Female
Occupation	Teacher	entails	Father	Jason Westfall
Brother	Brent Westfall		Brother	Brent Westfall
Appearance	Blue eyes	contradicts	Appearance	Brown eyes

Re³ (Yang et al. 2022)

Method	ROC-AUC 个
Random	0.5
GPT-3	0.52
Entailment (Yang et al. 2022)	0.528
Entailment with Dense Passage Retrieval (Yang et al. 2022)	0.610
Attribute Dictionary → Sentence (Yang et al. 2022)	0.684
CoRRPUS (comment)	0.751
CoRRPUS (specific)	0.794
CoRRPUS (abstract)	0.704

Probably because functions like set_age(self, character, age) complicate more than they help.

Tricks of the Trade

Instruction-tuned models like GPT-3.5 and Mistral-7B-Instruct like to be given a "role" first (e.g., "You are a helpful writing assistant.")

The more defined the task, the better

- More details
- One thing to do at a time

LLMs are overly confident (like people on the internet)

To "objectively" have the model evaluate something, you should

Chain-of-thought prompting helps models come up with better answers

They will "Yes and..." your prompt