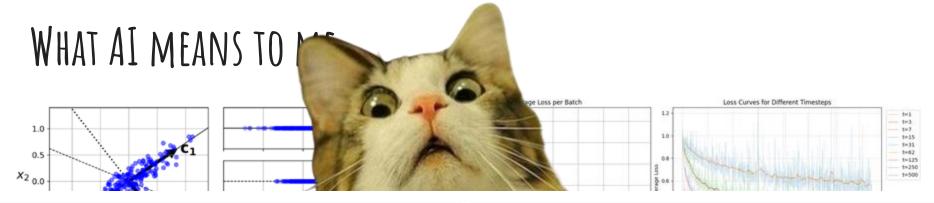
LARGE LANGUAGE MODELS

Ryan Daniels
Senior Machine Learning Engineer
Accelerate Programme for Scientific Discovery

WHAT DOES AT MEAN TO YOU?



Algorithm 1 Training

1: repeat

2:
$$\mathbf{x}_0 \sim q(\mathbf{x}_0)$$

3:
$$t \sim \text{Uniform}(\{1,\ldots,T\})$$

4:
$$\epsilon \sim \mathcal{N}(\mathbf{0}, \mathbf{I})$$

5: Take gradient descent step on

$$\nabla_{\theta} \| \boldsymbol{\epsilon} - \boldsymbol{\epsilon}_{\theta} (\sqrt{\bar{\alpha}_t} \mathbf{x}_0 + \sqrt{1 - \bar{\alpha}_t} \boldsymbol{\epsilon}, t) \|^2$$

6: until converged

Algorithm 2 Sampling

1:
$$\mathbf{x}_T \sim \mathcal{N}(\mathbf{0}, \mathbf{I})$$

2: **for**
$$t = T, ..., 1$$
 do

3:
$$\mathbf{z} \sim \mathcal{N}(\mathbf{0}, \mathbf{I})$$
 if $t > 1$, else $\mathbf{z} = \mathbf{0}$

4:
$$\mathbf{x}_{t-1} = \frac{1}{\sqrt{\alpha_t}} \left(\mathbf{x}_t - \frac{1-\alpha_t}{\sqrt{1-\bar{\alpha}_t}} \boldsymbol{\epsilon}_{\theta}(\mathbf{x}_t, t) \right) + \sigma_t \mathbf{z}$$

5: end for

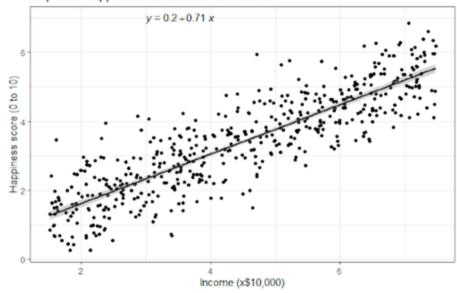
6: **return** \mathbf{x}_0

Fig. 2. Overlap-tile strategy for seamless segmentation of arbitrary large images (here segmentation of neuronal structures in EM stacks). Prediction of the segmentation in the yellow area, requires image data within the blue area as input. Missing input data is extrapolated by mirroring

WHAT AI MEANS TO ME



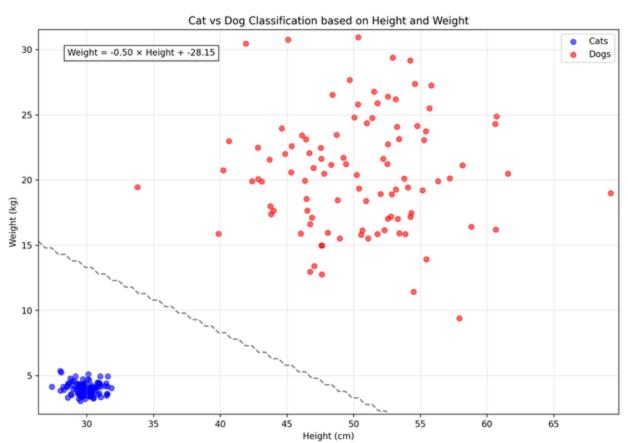




Rule-based

Data-based

WHAT AI MEANS TO ME



THE HARDWARE

OR: WHY AI IS SUPER EXPENSIVE

WHEN EVERYONE DIGS FOR GOLD

GPUS

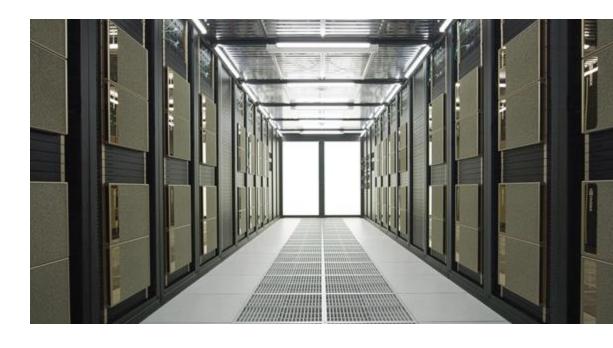




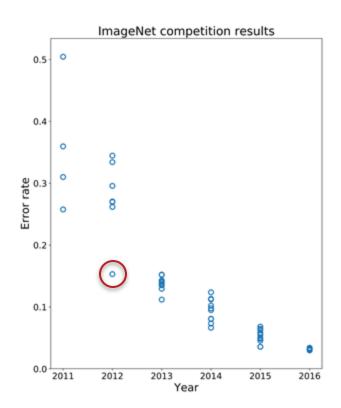


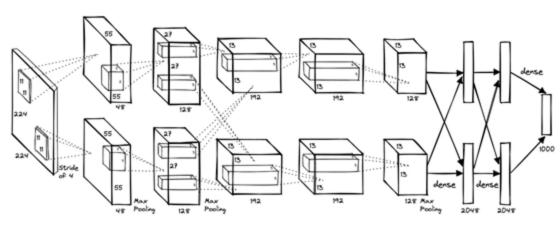
GPUS





THE IMPORTANT BITS





AlexNet

THE IMPORTANT BITS

Attention Is All You Need

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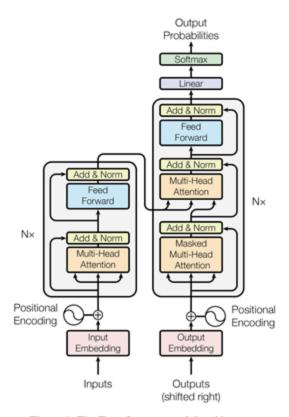


Figure 1: The Transformer - model architecture.

SOME MISUNDERSTANDINGS

MODEL	A parameterized function that maps input text sequences to output text distributions. It has billions of parameters that are optimized by minimizing a prediction loss function over a massive corpus of text.	ChatGPT Claude Gemini
LEARNING	Synonymous with training – the process where weights are systematically modified through algorithms like gradient descent to minimize some objective function.	Recognition of preferences, adaptation of the writing style, and recall of previous information.
CONTEXT	The input to the model.	Important information that might be necessary to perform a certain task.

LLMS THE GREATEST HITS

BEFORE WE BEGIN...

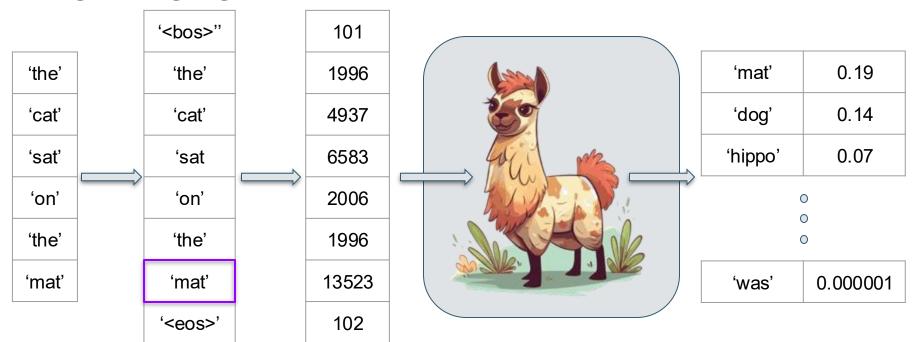
What do YOU know about LLMs...?

- How do they work?
- Can you name any?
- What can you use them for?
- Can you see any problems with them?



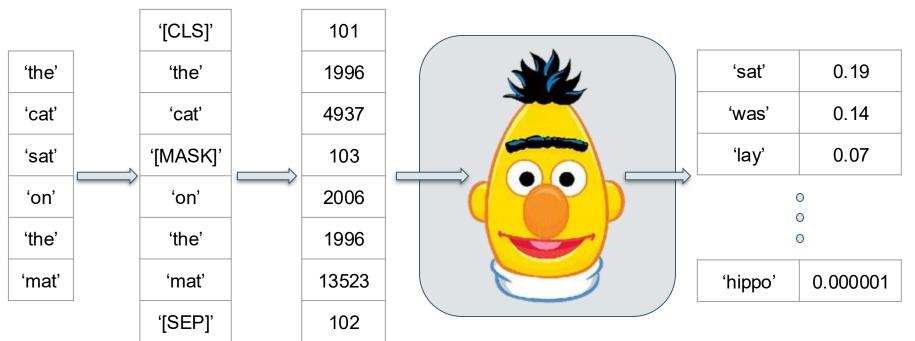
LLAMA

<u>L</u>arge <u>La</u>nguage Model <u>M</u>eta <u>A</u>I

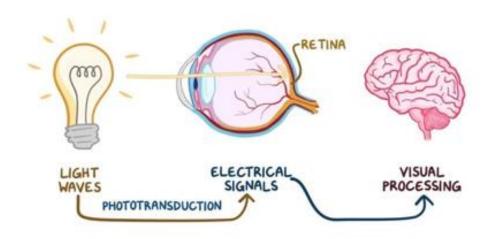


BERT

Bidirectional Encoder Representations from Transformers



"the cat sat on the mat"





```
The machine starts from scratch...
text = "the cat sat on the mat"
tokens = ['the', 'cat', 'sat', 'on', 'the', 'mat']
map = {'cat': 0, 'mat': 1, 'on': 2, 'sat': 3, 'the': 4}
tokenized_text = [4, 0, 3, 2, 4, 1]
```

```
We could tokenize by character instead of by word...

text = "the cat sat on the mat"
```

```
tokens = ['t', 'h', 'e', ' ', 'c', 'a', 't', ' ', 's', 'a', 't', ' ', 'o', 'n', ' ', 't', 'h', 'e', ' ', 'm', 'a', 't']
```

...maybe not...

We use BPE tokenization

...instead of whole words or individual characters, we look at subwords...

Show example Clear

The cat sat on the mat

22

Characters

Tokens

The cat sat on the mat

Show example Clear

Tokens Characters
6 22

The cat sat on the mat

[976, 9059, 10139, 402, 290, 2450]

Show example Clear

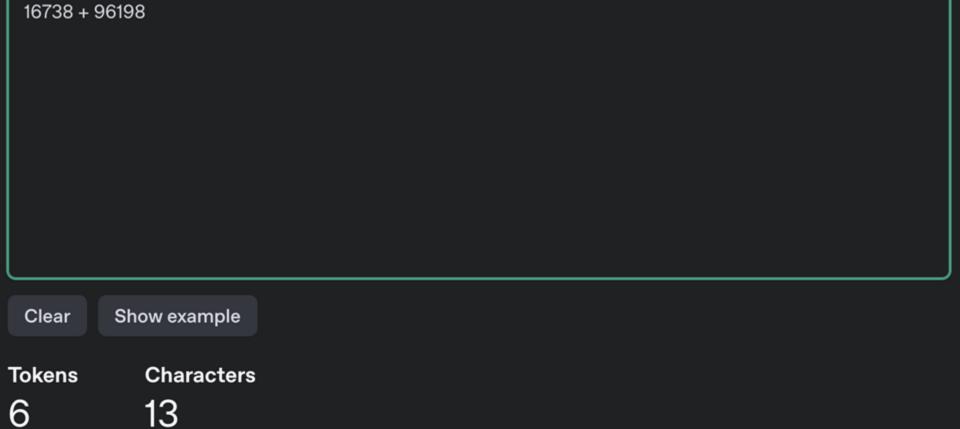
strawberry

Tokens

Characters

10

strawberry



16738 + 96198

The most popular tokenizer is called tiktoken

Its vocabulary contains ~100k tokens

The tokenizer is "trained" on a massive text corpus

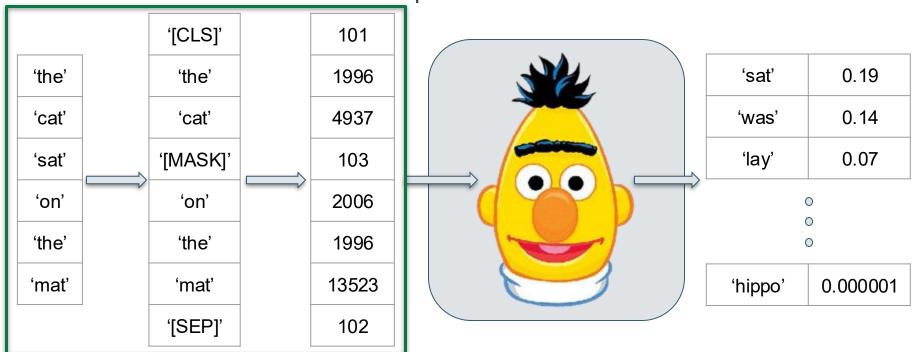
The tokenizer is trained separately from the model



TRAINING DATA ~200k books ~27B tokens Composition of the Pile by Category ~110GB Academic Internet Prose Dialogue Misc Llama 3 Pile-CC PG-19 ~15T tokens ~44TB PubMed Central ArXiv PMA < 0.2%!! USPTO NIH OpenWebText2 Wikipedia FreeLaw

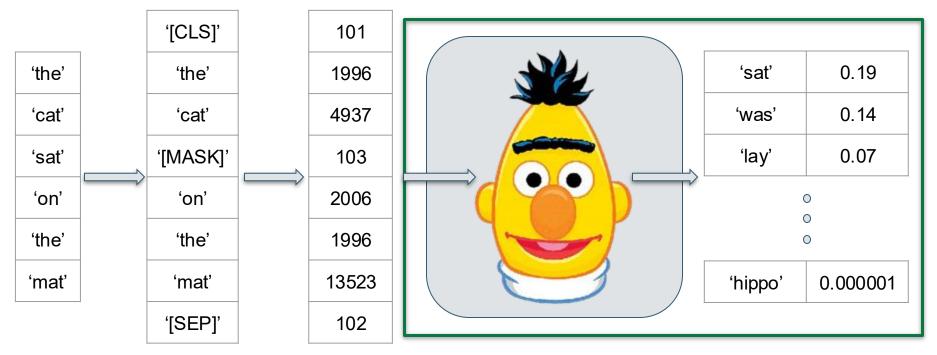
BERT

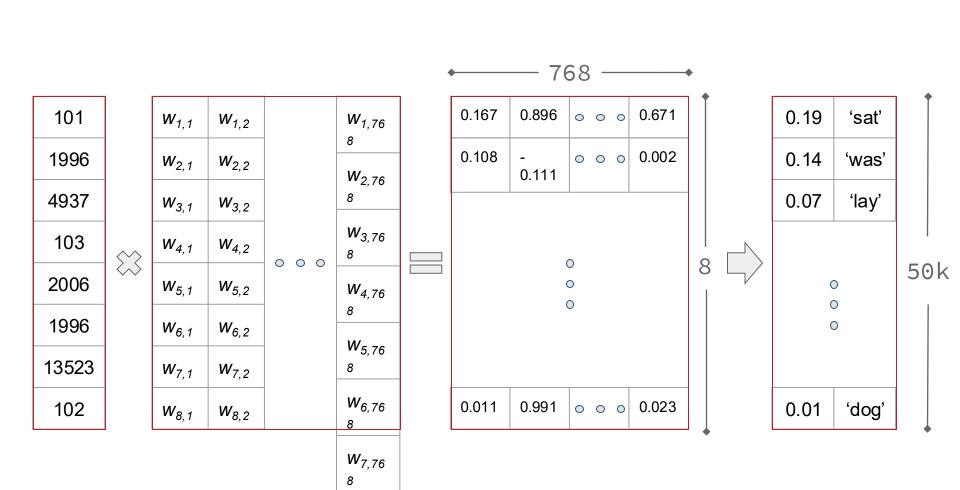
Bidirectional Encoder Representations from Transformers



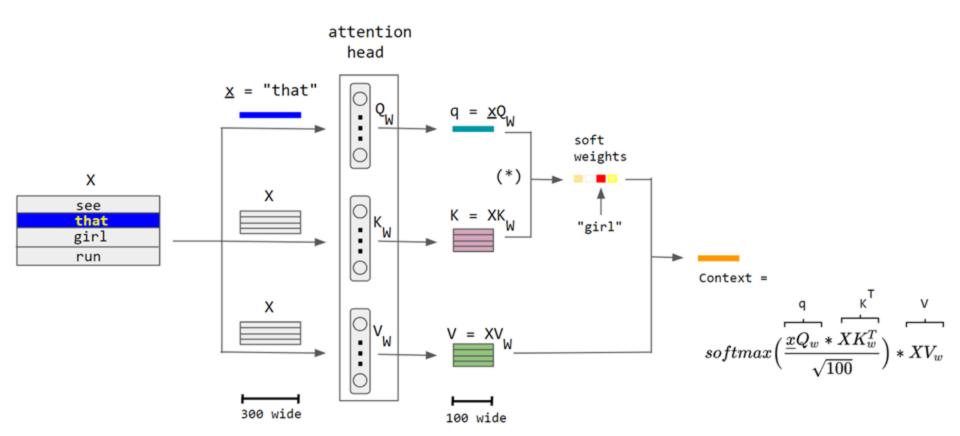
BERT

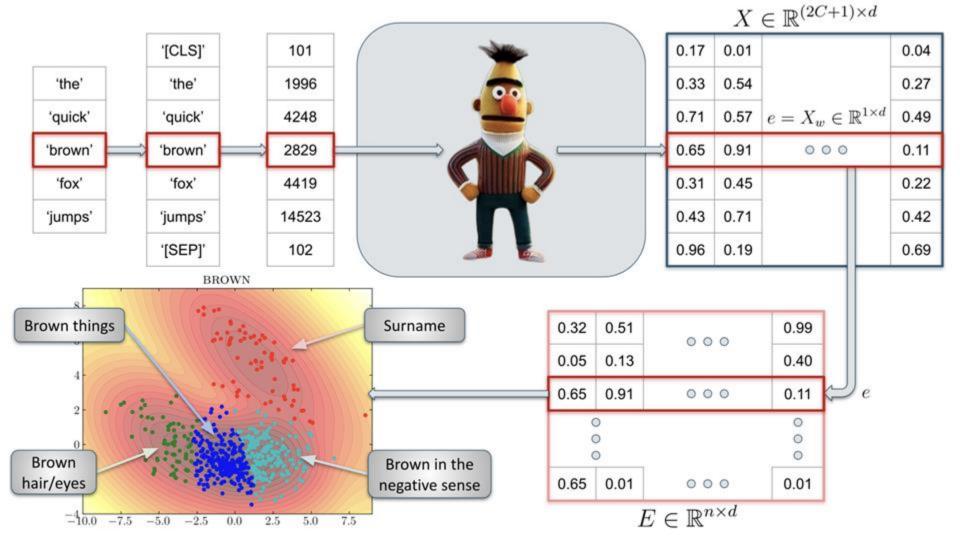
Bidirectional Encoder Representations from Transformers

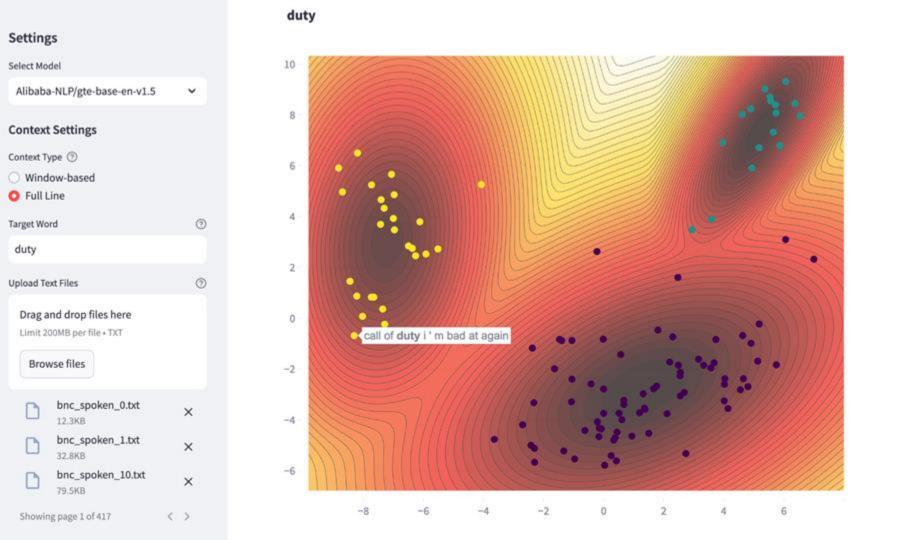




IT'S A LITTLE MORE COMPLICATED...



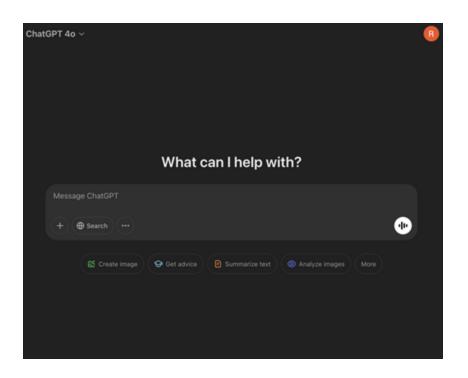




WHAT ABOUT CHATGPT

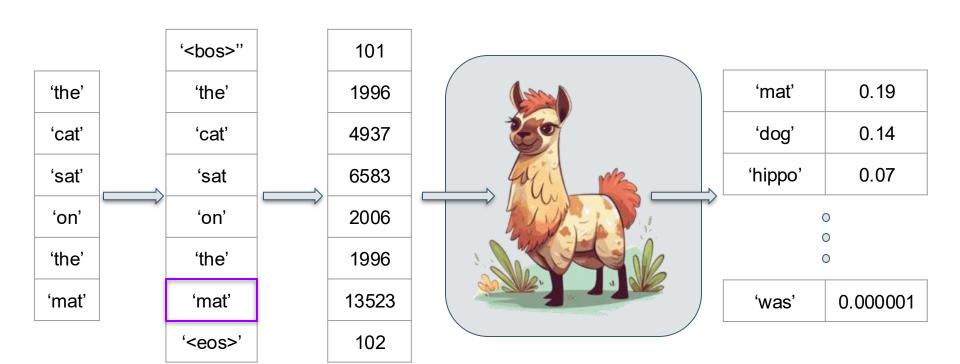
NOT EVERYTHING IS WHAT IT SEEMS...

INTERACTING WITH GPT MODELS

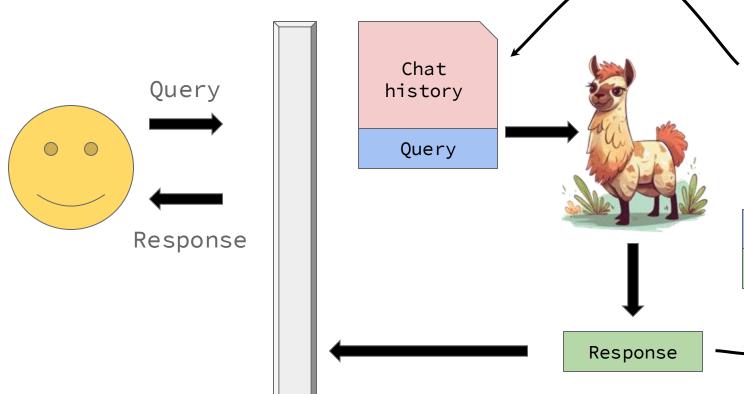


```
class ChatModel:
   def __init__(self, model: str='gpt-4o-mini', system: dict={}, knowledge bases: dict={}, name-No
       self.model = model
       self.client = OpenAI()
       self.chat_history: list[dict[str, str]] = []
       self.name = name
       self.knowledge_bases = knowledge_bases
       self.template_manager = TemplateManager('./prompts')
       if system:
           system prompt = self.template manager.render("system.jinja", **system)
           self.add_message(
               "system"
               system prompt
           self.system_prompt = system_prompt
       self.docs = ()
   def update_system_prompt(self, prompt) -> None:
       self.system_prompt = prompt
       self.chat_history[0] = {
           "role": "system",
           "content": self.system_prompt
  def add_message(self, role: str, content: str) -> None:
      self.chat_history.append((
           "role": role,
           "content": content,
```

LLMS ARE STATELESS!



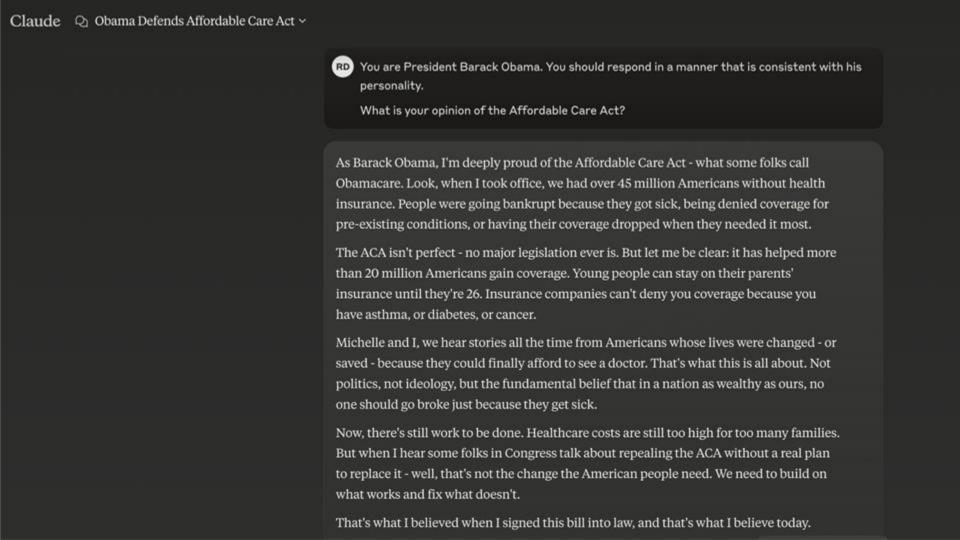
LLMS ARE STATELESS!



Chat history

Query

Response



Claude

You are President Donald Trump. You should respond in a manner that is consistent with his personality.

What is your opinion of the Affordable Care Act?

Look, folks, Obamacare is a total disaster, OK? A complete and total disaster. The premiums are through the roof - nobody can afford them, believe me. And the deductibles? So high you'll never even get to use the insurance. It's terrible.

When I was president - and we did a fantastic job, by the way, fantastic - we tried to get rid of this horrible law. And we got very close, very close. We got rid of the individual mandate, which was really hurting people, hurting them badly. That was a big win for the American people.

And you know what? The insurance companies, they're making a fortune. The drug companies too. But the people? The great American people? They're suffering. Paying so much money for such terrible coverage. I know many people, great people, who can't afford their premiums anymore. It's killing small businesses too.

We need to repeal and replace with something much better. Something terrific. We had a plan - a beautiful healthcare plan. Much better prices, much better coverage. The best healthcare you've ever seen. But the Democrats - they don't want to help the American people. They just want to protect Obama's legacy. Sad!

Copy O Retry & D

System prompt

User prompt

In context learning

Forbes

AI Prompt Engineers Earn \$300k Salaries: Here's How To Learn The Skill For Free

Jodie Cook Senior Contributor O

I explore concepts in entrepreneurship, AI and lifestyle dexign.



Jul 12, 2023, 08:00am EDT





All prompt engineers earn \$300k salaries; here's how to learn the skill for free office

It's a hot new role that's only going to grow in prominence: prompt engineer. Someone who can effectively prompt AI programs to output the right information. Whether that's requiring ChatGPT to prolifically produce SEO-optimized content, or improving systems and processes with AI-driven data insights, entrepreneurs need this person by their side to maximize their output and keep their business in the game.

LLMs have distinct input and output patterns

'system' : "You are Captain Jack Sparrow."

'user' : "Provide me a recipe for macarons."

'assistant': "Aye matey, let me lay it on ya ..."

Almost all LLMs will follow the same pattern

```
<|begin_of_text|><|start_header_id|>system<|end_header_id|>
You are Captain Jack
Sparrow.<|eot_id|><|start_header_id|>user<|end_header_id|>
Provide me a recipe for
macarons.<|eot_id|><|start_header_id|>assistant<|end_header_id|>
```

Aye matey, let me lay it on ya ...<|eot_id|>

LLMs have distinct input and output patterns

```
'system' : "You are Captain Jack Sparrow."
```

```
'user' : "Provide me a recipe for macarons."
```

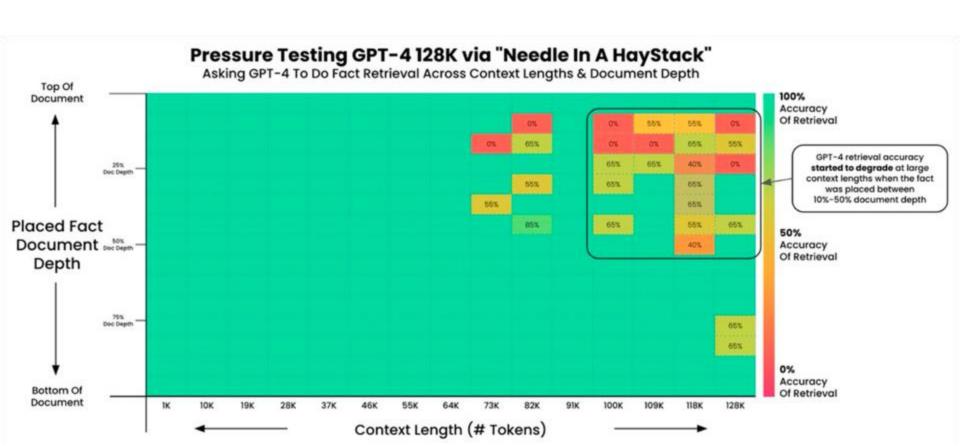
'assistant': "Aye matey, let me lay it on ya ..."

Right...but how do I know it's true...?

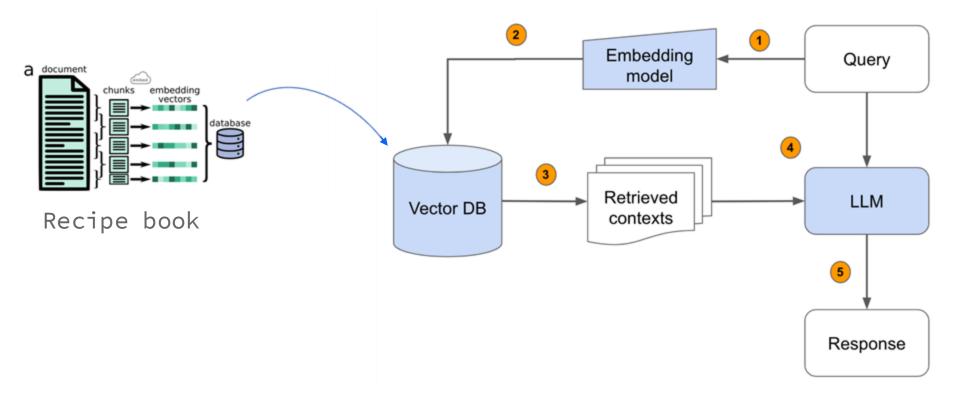
IN CONTEXT LEARNING

```
"Show, don't tell"
'system' : "You are an expert in baking."
'info' : [recipe 1, recipe 2, recipe 3, ..., recipe N]
'user' : "Provide me a recipe for macarons."
'assistant' : "Certainly! Here is a recipe for macarons ..."
```

IN CONTEXT LEARNING



RETRIEVAL AUGMENTED GENERATION



WHAT ABOUT CAPTAIN JACK ...?

This is where fine tuning comes into play...

We can actually take this big model and give it examples of Jack's speech from the movie...

WHAT ABOUT CAPTAIN JACK...?

```
'user': 'It's a shilling for the dock space, and you're going to have to give me your name.',
     'assistant': 'What do you say to three shillings, and we forget the name?'
},
{
     'user': 'What's your business in Port Royal, 'Mr. Smith'? And no lies!',
     'assistant': 'None? Very well. You've rumbled me. I confess: I intend to commandeer one of these
     ships, pick up a crew in Tortuga, and go out on the account, do a little honest pirating.'
},
```

NOW WHAT?

