2025-03-10 DARTMOUTH





# Al in Qualitative Analysis CPDE Lunch & Learn

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#### Why talk about this?

- 🎉 Artificial Intelligence (AI) seems to be everywhere, all of a sudden
- 🖫 Large Language Models' (LLM) capabilities are impressive
- A wave of new and shiny AI tools is flooding the scene
- Al components are making their way into many analysis workflows and tools
- AI literacy is a critical skill for everyone



#### Objectives of this talk

- Explain how Large Language Models (de-)construct language
- Show how AI can help explore qualitative data
- Equip you with critical knowledge to assess new tools and techniques
  Spark ideas and conversation

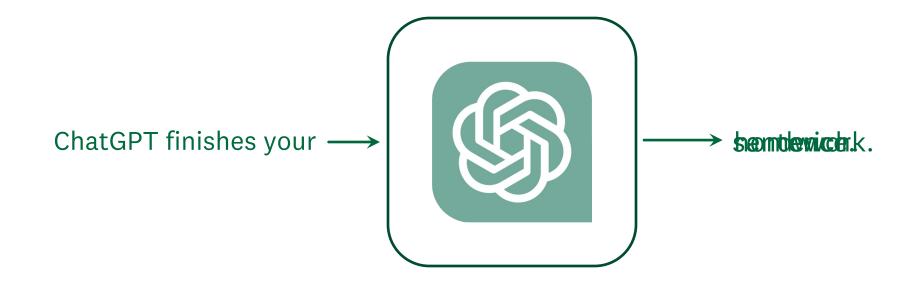


# How Large Language Model's (de-)construct language What can't AI do?

- LLMs can do exactly one thing:
  - Predict the next word given a sequence of input words
- Liverything else they do is a clever use of this singular ability!

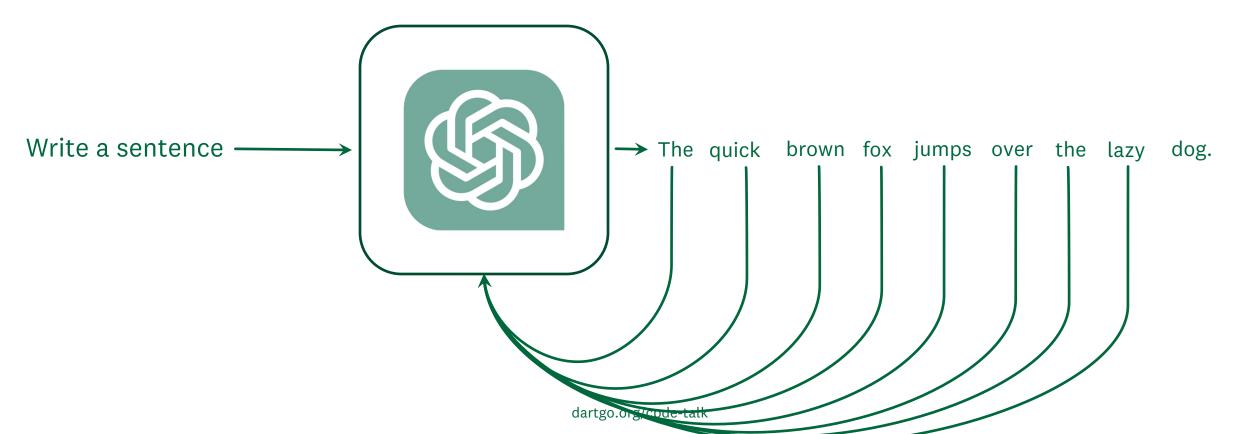


#### The fundamental case





# How Large Language Model's (de-)construct language **Predicting in a loop**





#### How Large Language Model's (de-)construct language Predicting a conversational turn

```
<|begin_of_text|>
<|start_header_id|>system<|end_header_id|>
Cutting Knowledge Date: December 2023 Today
Date: 23 July 2024 You are a helpful assistant
<|eot_id|>
<|start_header_id|>user<|end_header_id|>
What is the capital of France?
<|eot_id|>
<|start_header_id|>assistant<|end_header_id|>
```

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<sup>\*</sup>The template shown is the template used for Llama 3.2



#### How Large Language Models construct language



How many words are in your response?



There are 12 words in my response.



#### 👫 List them out in a numbered list.



Sure, here are the words in my previous response listed in a numbered list:

- 1. There
- 2. are
- 3.12
- 4. words
- 5. in
- 6. my
- 7. response.



#### LLMs lie? - It's complicated

- LLMs have no concept of facts or truth, no internal database, no search mechanism
- They only "know" patterns reinforced through their training data
  - Example in Dartmouth Chat: Timothée Chalamet's mother
- Sectually correct responses are patterns that happen to be correct
- "Hallucination" is a bit of a misnomer

Emily M. Bender *et αl.* 2021. On the Dangers of Stochastic Parrots: Can Language Models Be Too Big? . In Proceedings of the 2021 ACM Conference on Fairness, Accountability, and Transparency (FAccT '21). Association for Computing Machinery, New York, NY, USA, 610–623. https://doi.org/10.1145/3442188.3445922



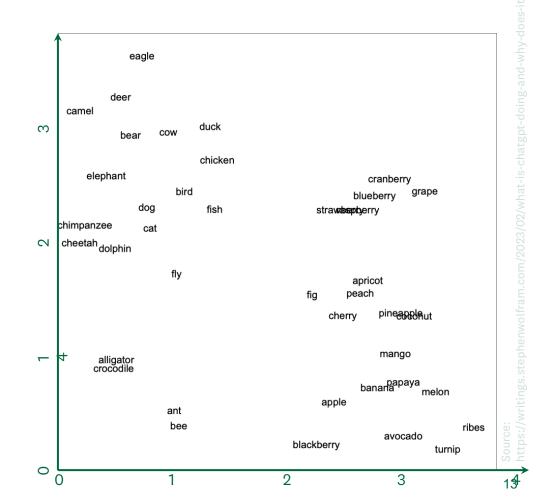
# How Large Language Model's (de-)construct language How to find the next word?

- The patterns in written language are learned from text data in a training phase
- GPT 3.5, for example, has been trained on about 570 GB of texts or 300 billion words (English Wikipedia: 4.2 billion words)
- This would still not be enough data if we were looking for examples of longer specific word sequences!

#### How to find the next word?

Two important "tricks" solve this problem:

- The model learns to map every word into a numerical "meaning space", where words with similar meaning are represented by similar numbers
- That way the model can base the probability of the next word on sequences of similar words
- GPT 3.5 uses 2048 dimensional embeddings (probably)





#### Background

#### How to find the next word?

Two important "tricks" solve this problem:

- 1. Words in "meaning space"
- The model uses a mechanism called "attention" to emphasize the most important words in the sequence
  - These weights are learned during training
  - This structure is called a "transformer"
  - GPT: Generative Pre-trained Transformer

I am going to spend the summer in Germany. I want to go to Hamburg, Berlin, and Dresden. Write an itinerary for a two-week long trip.



Sure, here's a possible itinerary for a two-week trip to Hamburg, Berlin, and Dresden:

#### Day 1-3: Hamburg

- Explore the historic center of Hamburg and visit the impressive Elbphilharmonie concert hall
- Take a boat tour of Hamburg's harbor and visit the International Maritime Museum



### What can AI do for qualitative analysis?

- Concept of "semantic similarity" very powerful
- LLM's are excellent at picking up patterns in language
- Let's look at an example!

By the way: All responses for the following demos were generated using Claude 3.7 Sonnet



#### Summary

- Large Language Models are excellent text processors
- Semantic embeddings are a useful tool to capture contextual information in text
- Using a Human-in-the-Loop approach, AI holds a lot of potential to speed up and scale qualitative analysis



## Discussion





# Thank you

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