



Gridspace

GRIDSPACE IAP 2024 LECTURE 4
Tools for LLM Memory

January 18, 2024

Questions from last time

- Why do you think ChatGPT did poorly on the 3-back task, even though it supposedly has superior memory to humans?

It could be...

- misinterpreting the task (even though it could recite a definition)
- not effectively retrieving the information from the context
- bad at counting the number of steps

Questions from last time

- As LLMs get larger and larger, and trained with more and more data. How would you expect the models to perform on each of the memory categories?
 - Working Memory: No change as long as the context length stays the same. However, it might get better at extracting relevant data from the context
 - Episodic Memory: Same as working memory
 - Semantic Memory: Depends on the data, it could see more of the same facts be repeated and remember things better. Or it could get conflicting data and become less effective.

Questions from last time

- LLM memorizing certain data can be a concern (e.g. private data, copyrighted data). What are some ways to alleviate this.

There are many ways including:

- Reducing repetition in data to reduce the chance of it remembering word for word
- Using anonymized data (For cases of private data)
- Create synthetic data (if applicable to use case)



LO

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"Wh

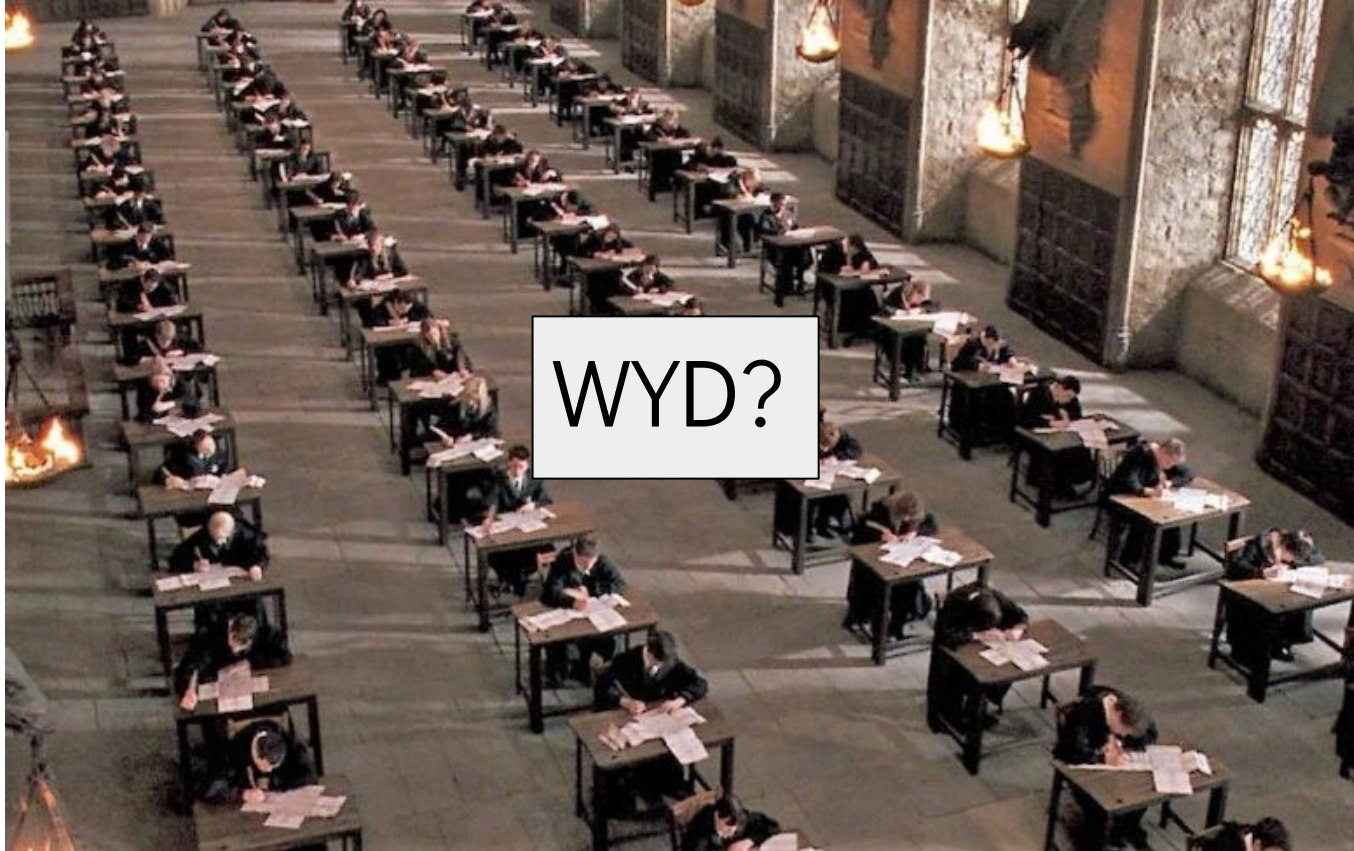
"Le

)

Outline

- Problem: Information retrieval
- Solution 0: ...
- Solution 1: ...
- Solution 2: ...
- Solution 3: ...
- Solution 4: ...

Scenario: closed book exam in 2 weeks



WYD?



Outline

- Problem: Information retrieval
- **Solution 0: Memory in training data**
- Solution 1: ...
- Solution 2: ...
- Solution 3: ...
- Solution 4: ...

Recall from Nick

IAP 2024 BEYOND LLMs: CAN LLMs REMEMBER?
WITH NICK TANTIVASADAKARN

Input

Who is the first president of the United States?

George Washington

Output

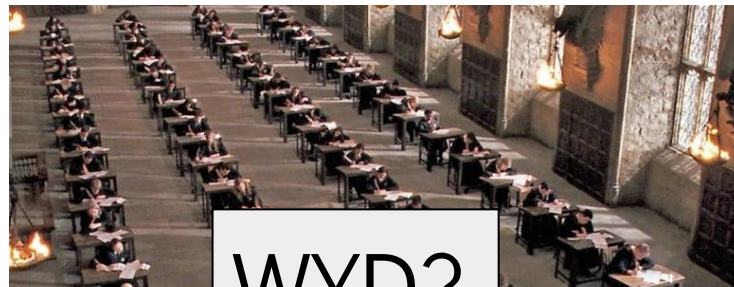
Who is the first president of the United States?

George Washington.

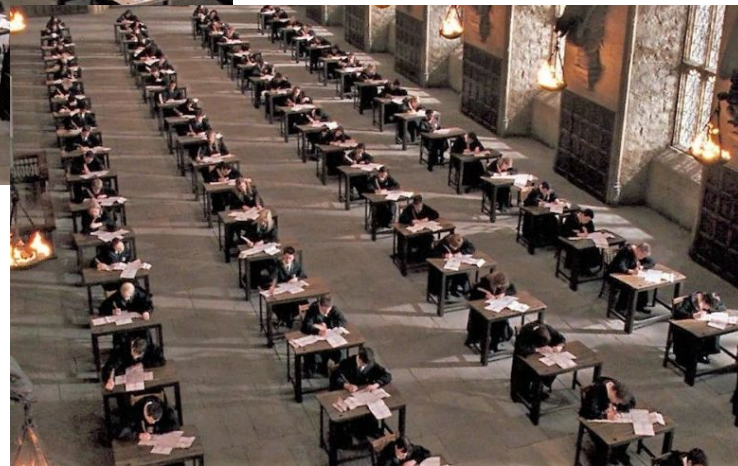
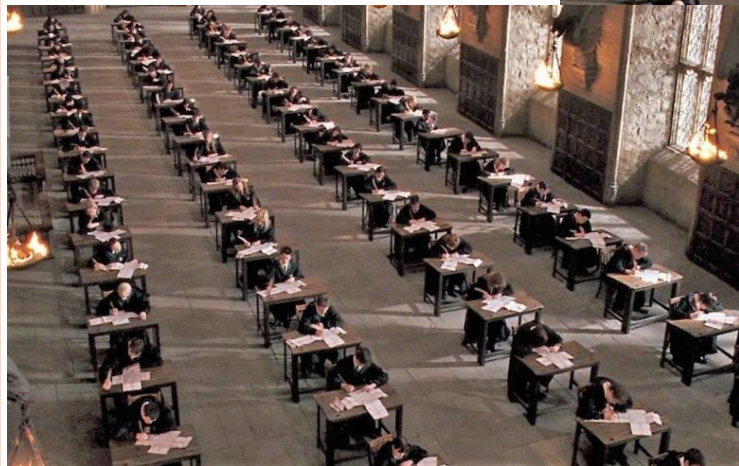
Information 'stored' in parameters

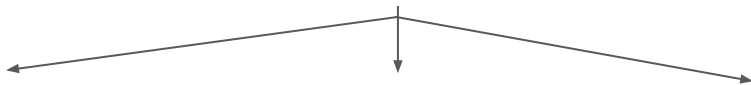
The diagram illustrates the process of an LLM generating an output based on an input. On the left, under the heading 'Input', the question 'Who is the first president of the United States?' is shown, followed by the answer 'George Washington'. An arrow points from this input to a central neural network diagram. The neural network consists of four orange input nodes on the left, a hidden layer of five white nodes in the middle, and two blue output nodes on the right. A bracket underneath the neural network is labeled 'Information 'stored' in parameters'. Another arrow points from the neural network to the right, where the heading 'Output' is shown. The output is the same question 'Who is the first president of the United States?' followed by the answer 'George Washington.' in red text. In the bottom right corner, there is a small video inset showing a man with glasses and a pink shirt.

Scenario: You have three exams at the same time!
(But you can clone yourself)



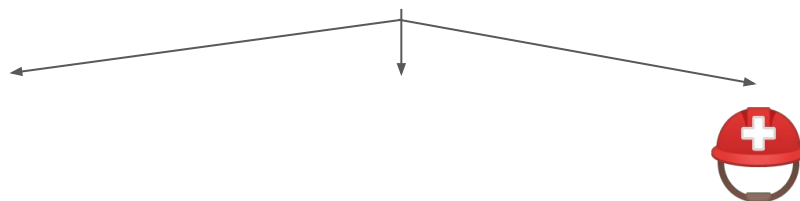
WYD?





Memory in training data

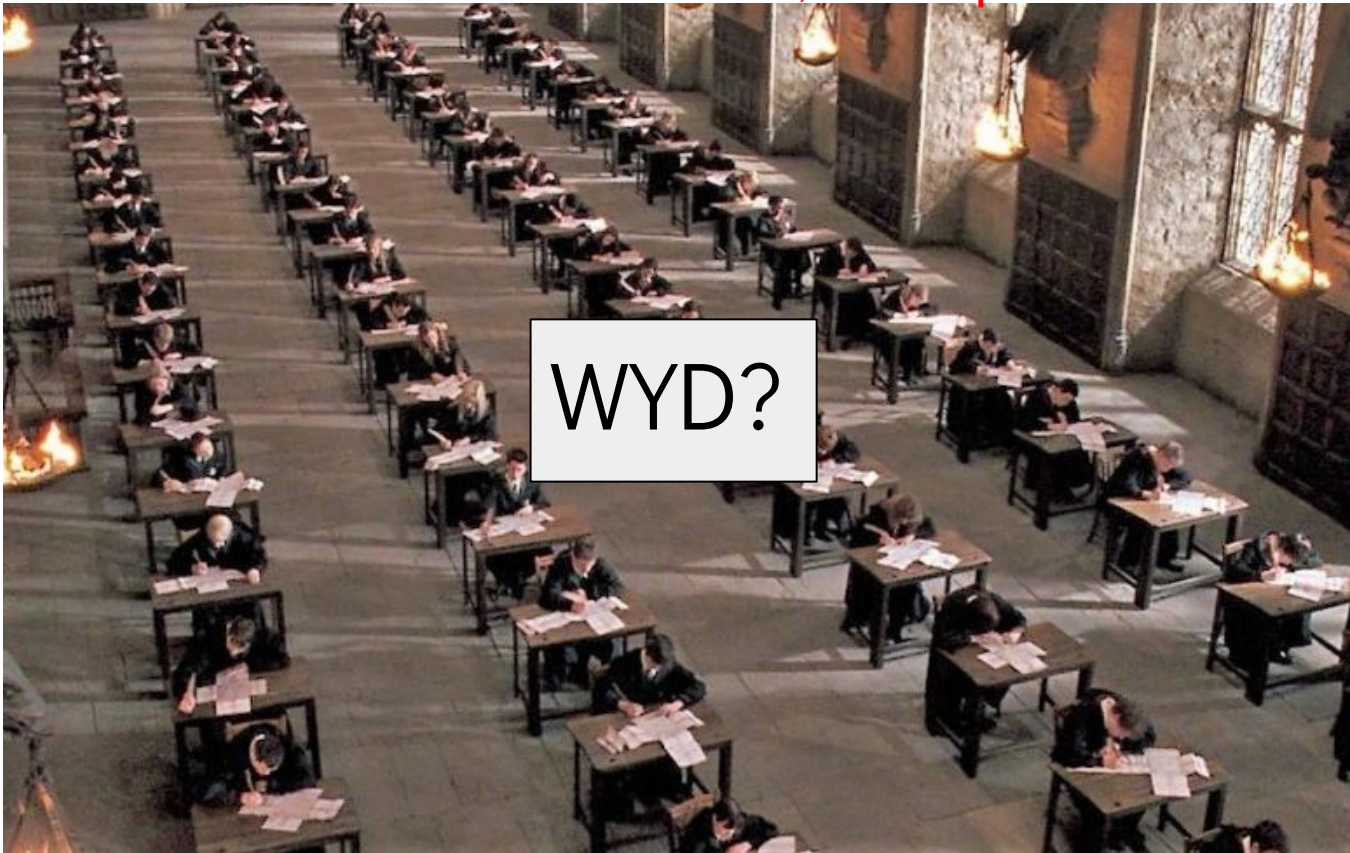
- One base model
- (Possibly multiple) delta on base model for domain specialization
- Train delta on new dataset

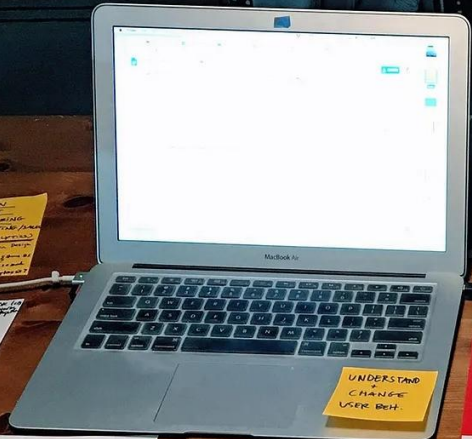
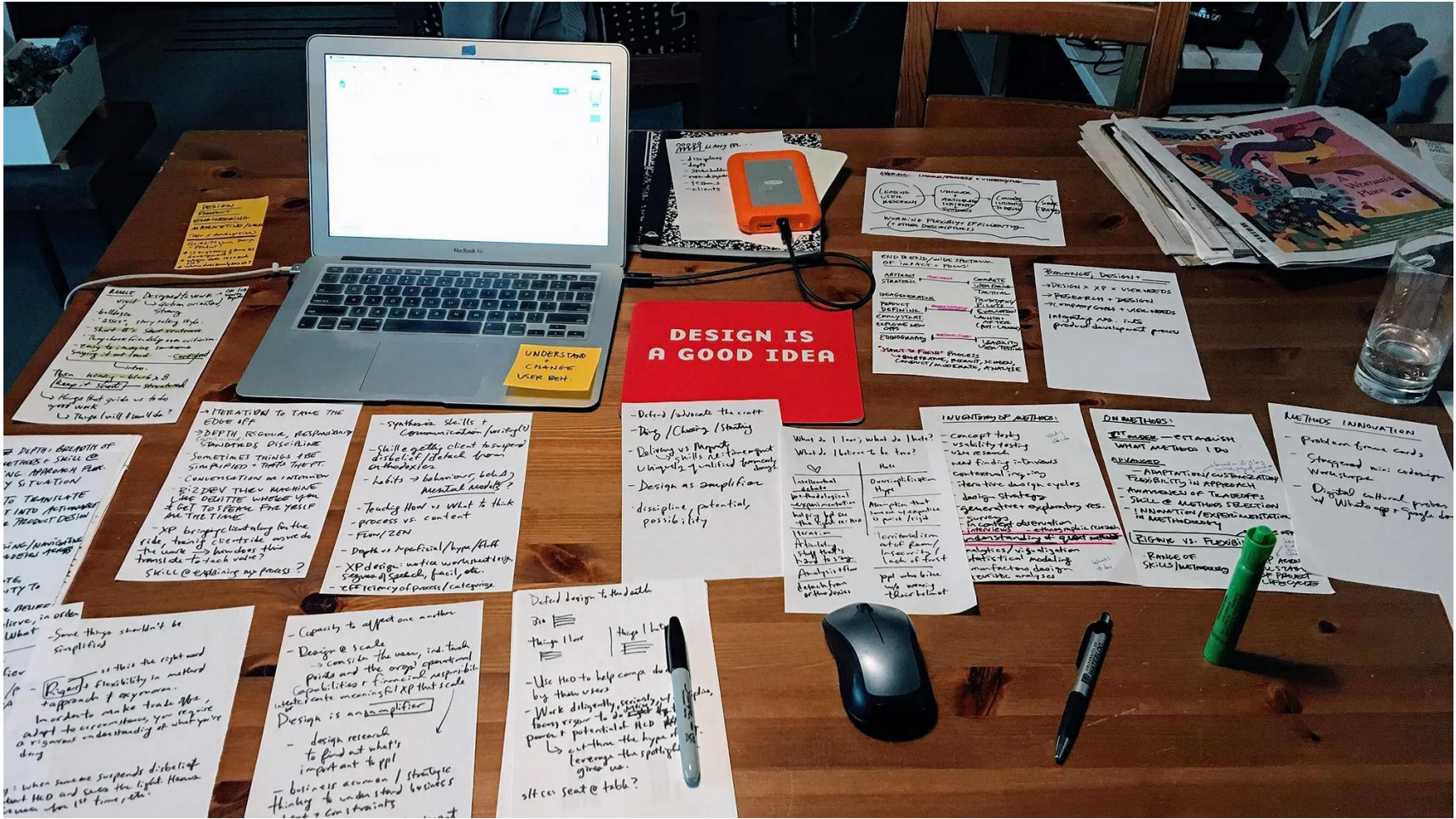


Memory in training data

- Relatively “static”, best for foundation knowledge, eg. language grammar, common sense (eg. the sky is blue)
- Expensive to retrain model when new knowledge is acquired

Scenario: exam in 1 hour, but open notes





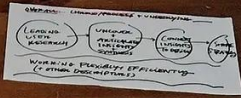
EMILY Designed for work →
 - user → system oriented
 - strategy
 - build
 - build → sharp selling style
 - build → sharp customer
 - They have a good design and execution
 - only by comparing to someone
 - saying it out loud → **emotional**

Then **what** = **what** & **how**
 (Keep it short) → **structured**
 → **help** that guide us to do
 your work
 → **things** that **don't** do?

EMILY
 - user → system oriented
 - strategy
 - build
 - build → sharp selling style
 - build → sharp customer
 - They have a good design and execution
 - only by comparing to someone
 - saying it out loud → **emotional**

**UNDERSTAND
 CHANGE
 USER BEH.**

DESIGN Learning from...
 - Approach
 - Style
 - Structure
 - Methodology
 - Process
 - Activities



END IN DESIGN **THE** **FUNCTION** **OF** **DESIGN** **IS** **TO** **CREATE** **A** **SOLUTION** **TO** **A** **PROBLEM**

DESIGN **IS** **TO** **CREATE** **A** **SOLUTION** **TO** **A** **PROBLEM**

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DEPTH **QUALITY** **OF** **DESIGN** **IS** **TO** **CREATE** **A** **SOLUTION** **TO** **A** **PROBLEM**

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ITERATION **TO** **TAME** **THE** **EDGE** **OF** **DESIGN**

ITERATION **TO** **TAME** **THE** **EDGE** **OF** **DESIGN**

ITERATION **TO** **TAME** **THE** **EDGE** **OF** **DESIGN**

Synthesis Skills + Communication (writing)

- Skills getting client to suspend disbelief / detach from orthodoxy
- habits → believing / holding mental models?
- Teaching How vs What to think
- process vs content
- Flow / Zen
- Depth vs superficial / hype / fluff
- XP design: metrics, user research, systems of speech, facial, etc.
- efficiency of process / categorize

Define / Advocate the craft

- Why / Choosing / Strategy
- Defining vs Marketing
- Why your skills are important / unique / qualified / forward / bright
- Design as simplification
- discipline, potential, possibility

What do I love, what do I hate?

What do I love to be free?

Vertical **depth** **of** **design** **is** **to** **create** **a** **solution** **to** **a** **problem**

Vertical **depth** **of** **design** **is** **to** **create** **a** **solution** **to** **a** **problem**

Vertical **depth** **of** **design** **is** **to** **create** **a** **solution** **to** **a** **problem**

INVENTORY OF METHODS:

- concept testing
- usability testing
- user research
- user feedback
- interviews
- contextualizing
- iterative design cycles
- design sprints
- generative / explanatory res.
- surveys
- participant observation
- ethnography
- field research
- user testing
- prototyping
- analytics / visualization
- statistical modeling
- manufacturing design
- formative analysis

DESIGN METHODS:

- **PROCESSES** - **EFFICIENCY** **WHAT** **METHODS** **TO** **USE**
- **ADVANCED**
- **AMBITIOUS** **CUSTOMER** **STORY**
- **FEASIBILITY** **IN** **APPROACH**
- **SKILL** **OR** **METHOD** **SELECTION**
- **INNOVATION** **OR** **EXPERIMENTATION** **IN** **METHODOLOGY**
- **FIGURE** **VS.** **PROBLEM**
- **PRINCIPLES** **OF** **SKILLS** **METHODOLOGY**

METHODS INNOVATION

- Problem frame cards
- Storyboard mini canvas / workshop
- Digital cultural probes w/ WhatsApp + Google doc

Some things shouldn't be simplified

Some things shouldn't be simplified

Some things shouldn't be simplified

Capacity to affect one another

- Design @ scale
- consider the user, individual, private and the organizational capabilities & financial responsibility
- create meaningful XP that scale
- Design is an **ambiguity**
- design research to find out what's important to ppl
- business acumen / strategic thinking to understand business & constraints

Define design to the death

Do things I love | things I hate

- Use HCD to help compare design these users
- Work diligently, research, focus on the user, focus on the user, focus on the user
- potential of HCD to power the hype of user
- coverage the spotlight given user
- sit on seat @ table?



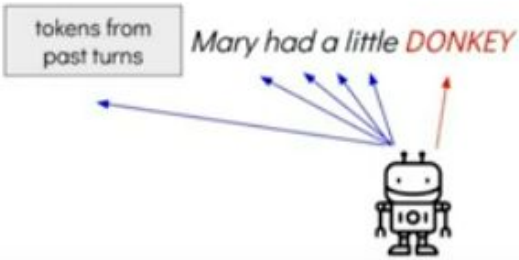
Outline

- Problem: Information retrieval
- Solution 0: Memory in training data
- **Solution 1: Memory in LLM Context**
- Solution 2: ...
- Solution 3: ...
- Solution 4: ...

Recall from Nick

IAP 2024 BEYOND LLMs: CAN LLMs REMEMBER?
WITH NICK TANTIVASADAKARN

Previous tokens are given to the LLM

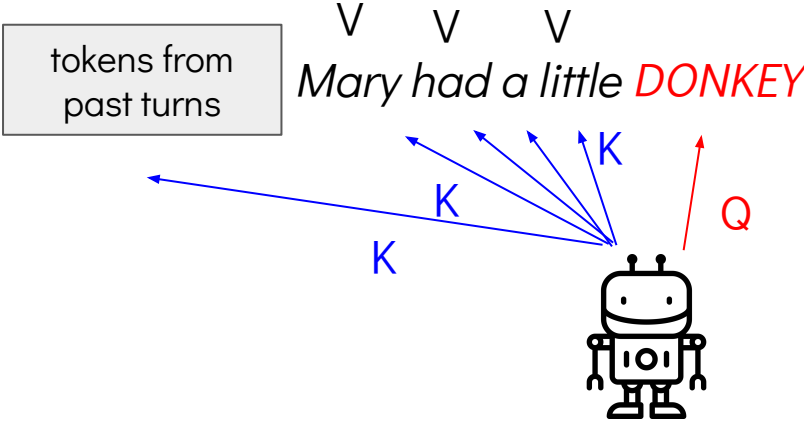


The diagram illustrates the concept of context in LLMs. A box labeled "tokens from past turns" has blue arrows pointing to a robot icon. The text "Mary had a little DONKEY" is shown, with "DONKEY" in red. A red arrow points from the robot to the word "DONKEY", indicating that the model is processing the current token based on the context provided by previous tokens.

18:46 / 50:20

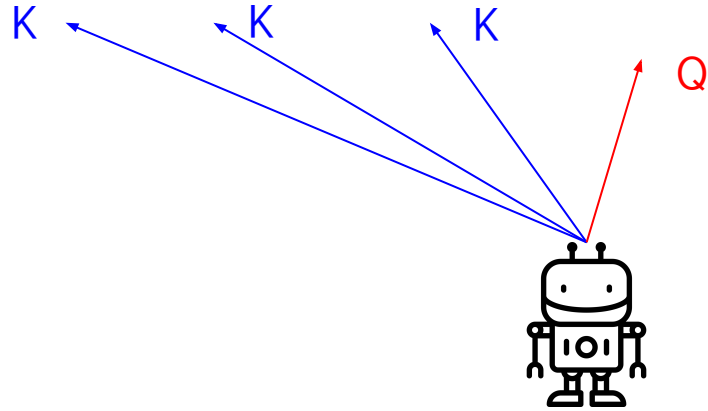
CC

Memory in LLM Context

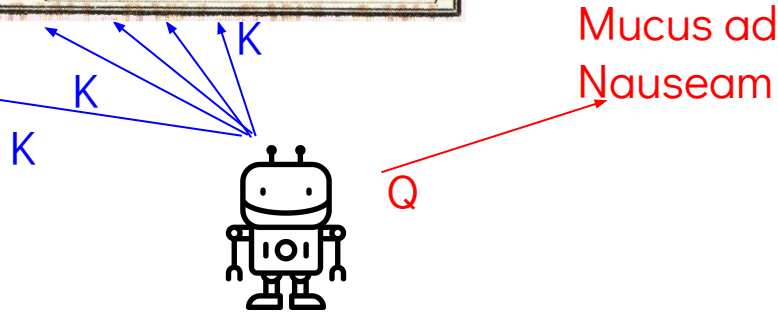
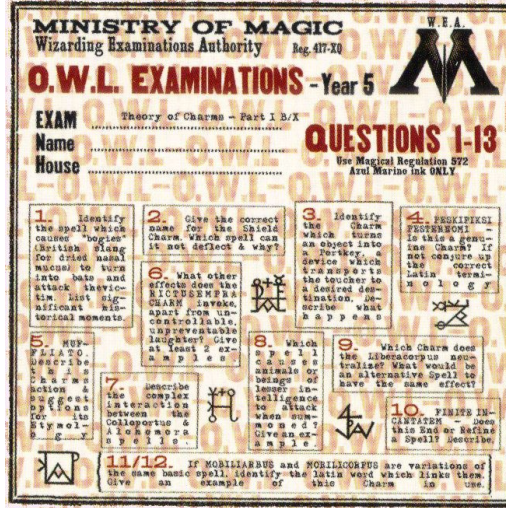
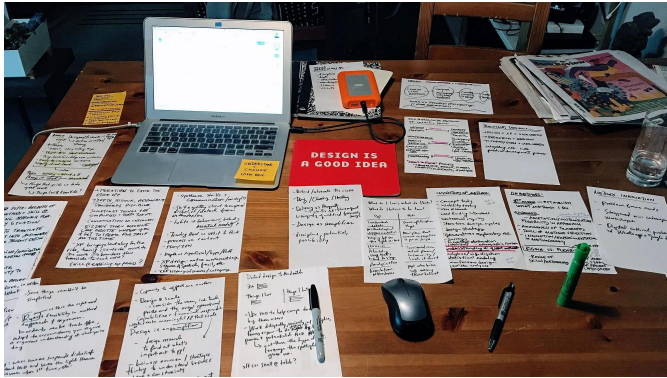


Attention: Q, K, V

V V V



Memory in LLM Context



Memory in LLM Context

- Storing K,V takes a lot of GPU memory
- Storing K, V can be messy (fragmented)

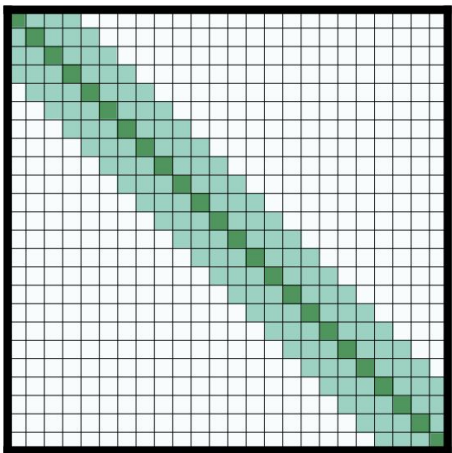


Memory in LLM Context

- Economize, reduce size of K,V
 - Sliding window attention
 - Quantization
 - Group Query Attention
- Reduce memory waste due to messy storage (fragmentation)
 - PagedAttention

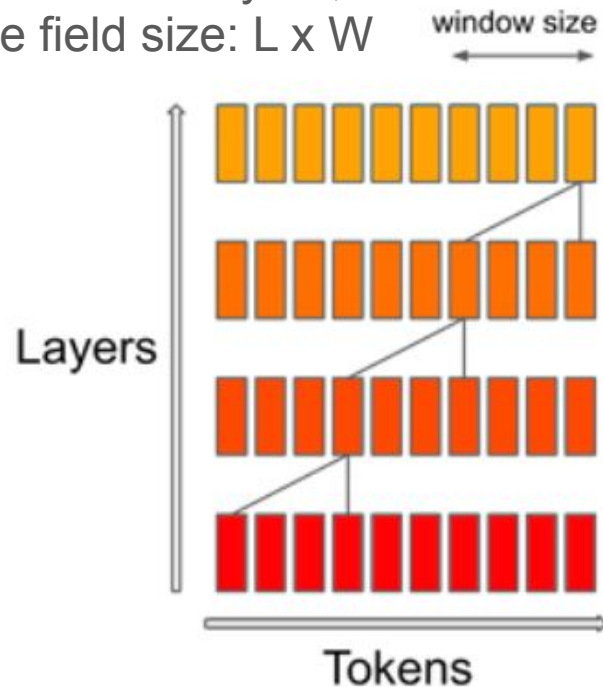
Memory in LLM Context

Window size W ,
a transformer of L layers,
Receptive field size: $L \times W$



(b) Sliding window attention

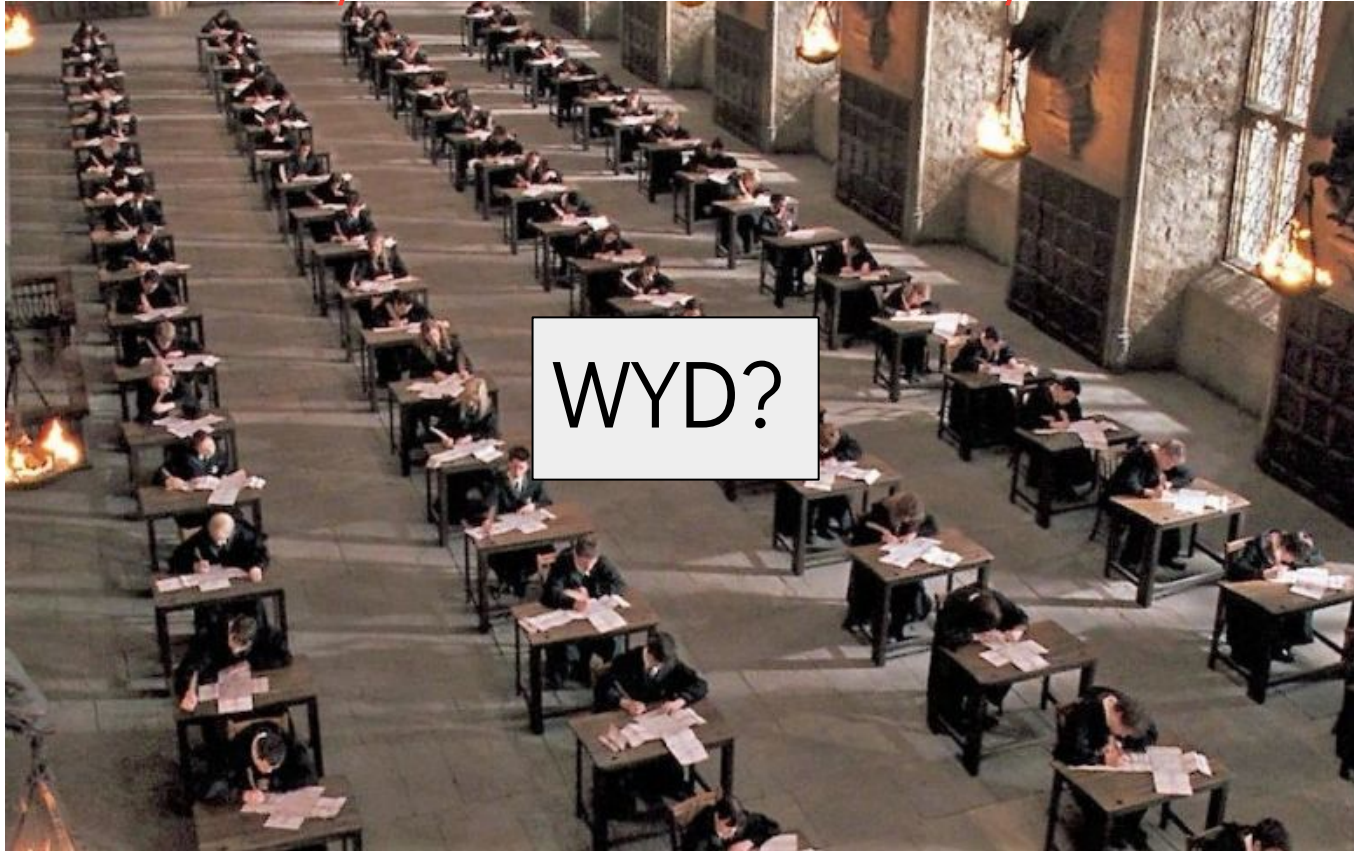
<https://paperswithcode.com/method/sliding-window-attention>

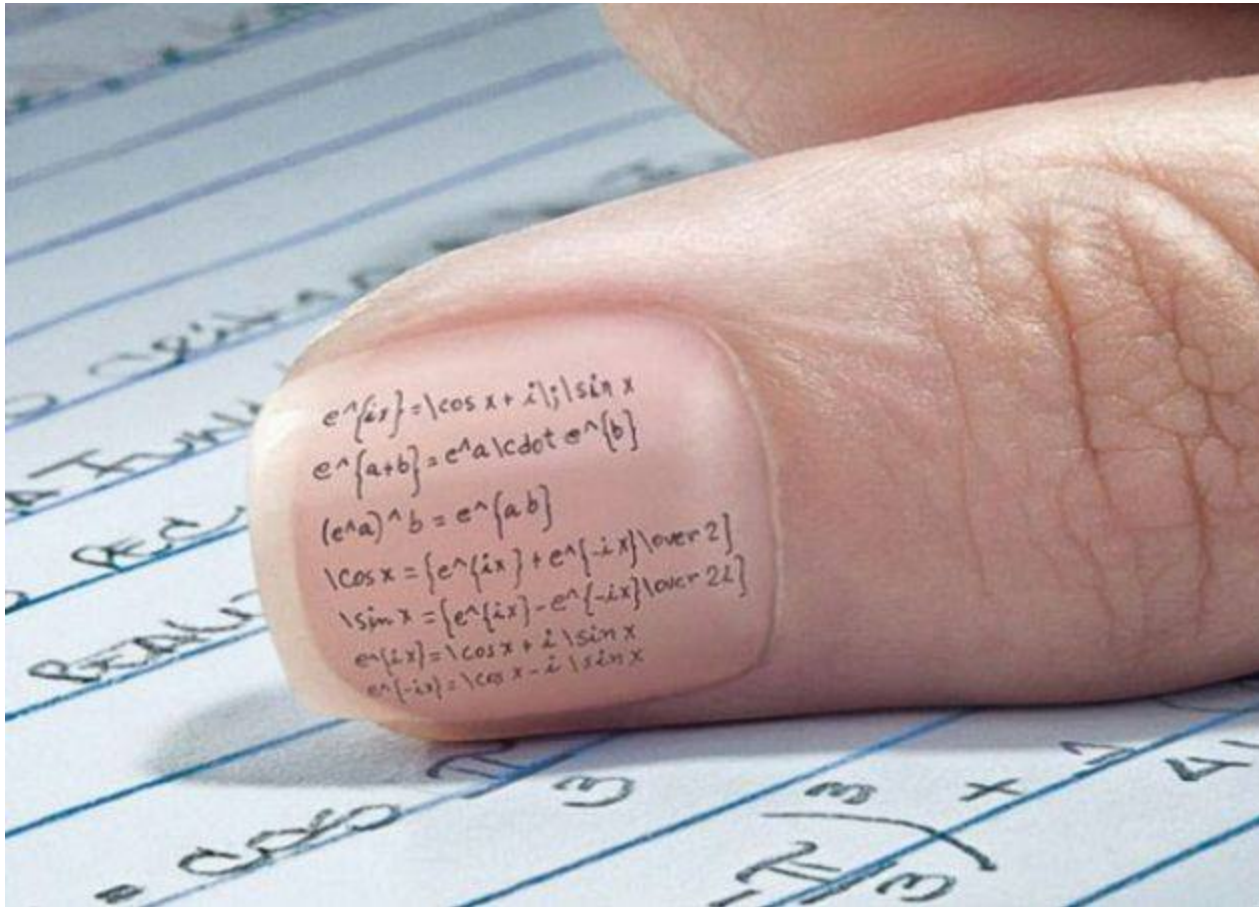


Memory in LLM Context

- Still limited, 32k tokens
- Still computationally expensive

Scenario: *your desk cannot fit all your notes!*





$$e^{ix} = \cos x + i \sin x$$

$$e^{a+b} = e^a \cdot e^b$$

$$(e^a)^b = e^{ab}$$

$$\cos x = \frac{e^{ix} + e^{-ix}}{2}$$

$$\sin x = \frac{e^{ix} - e^{-ix}}{2i}$$

$$e^{ix} = \cos x + i \sin x$$

$$e^{-ix} = \cos x - i \sin x$$

Outline

- Problem: Information retrieval
- Solution 0: Memory in training data
- Solution 1: Memory in LLM Context
- **Solution 2: Compress chat history/context**
- Solution 3: ...
- Solution 4: ...

Compress chat history/context

- LLMs are really good at summarization

Text Summarization on GigaWord

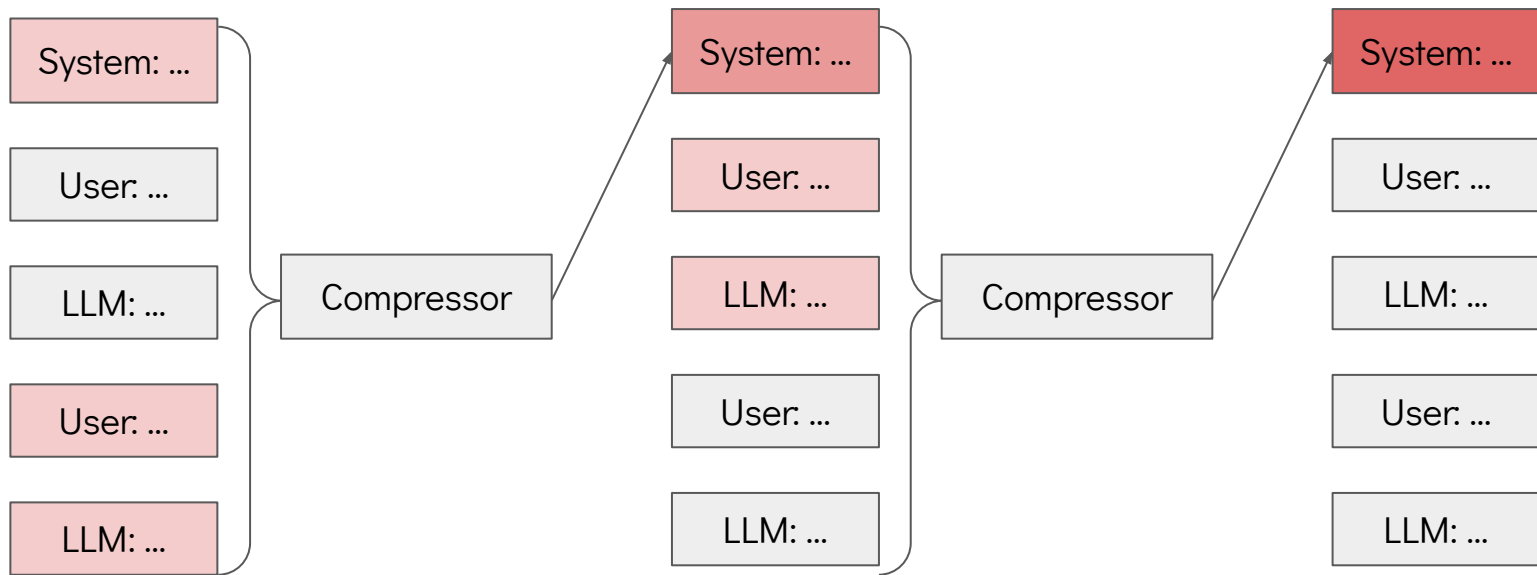


- Let's summarize chat history/context so far!

Compress chat history/context

- “In at most 1000 words, generate a summary of the chat history so far with sufficient detail to act as a replacement for the chat history in case we lose it. Pay special attention to instructions given by the user and system.”

Compress chat history/context

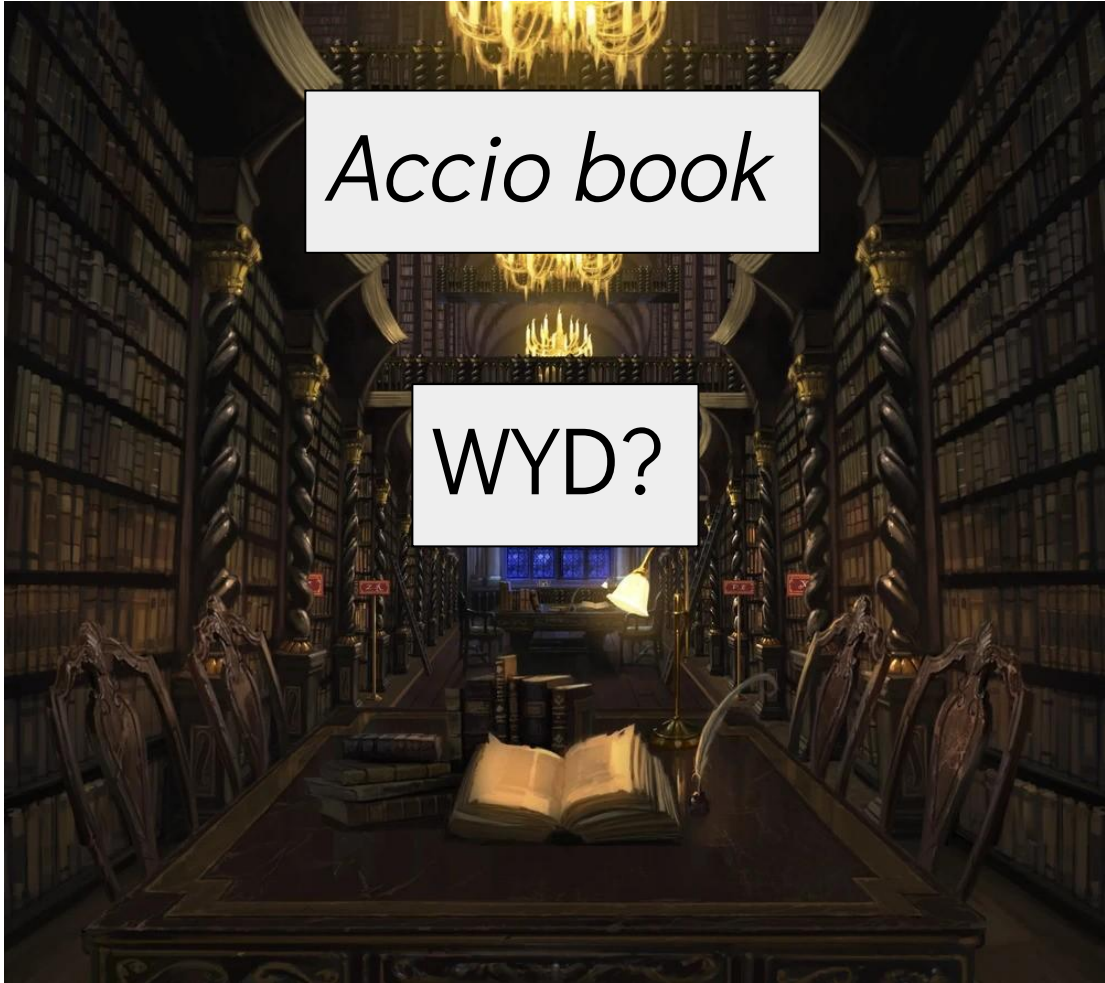


Compress chat history/context

- Needs to run LLM compressor regularly
- Some information is lost

Scenario: you have incantation to summon an object





Accio book

WYD?

Outline

- Problem: Information retrieval
- Solution 0: Memory in training data
- Solution 1: Memory in LLM Context
- Solution 2: Compress chat history
- **Solution 3: Memory in external database**
- Solution 4: ...

Recall from week 1: LLM can use tools

IAP 2024 BEYOND LLMs: TOOLS FOR LLM PLANNING
WITH WONKYUM LEE

LLM AS TOOL RETRIEVER

Toolformer: Language Models Can Teach Themselves to Use Tools

Timo Schick **Jane Dwivedi-Yu** **Roberto Dessì** **Roberta Raileanu**
Maria Lomeli **Luke Zettlemoyer** **Nicola Cancedda** **Thomas Scialom**

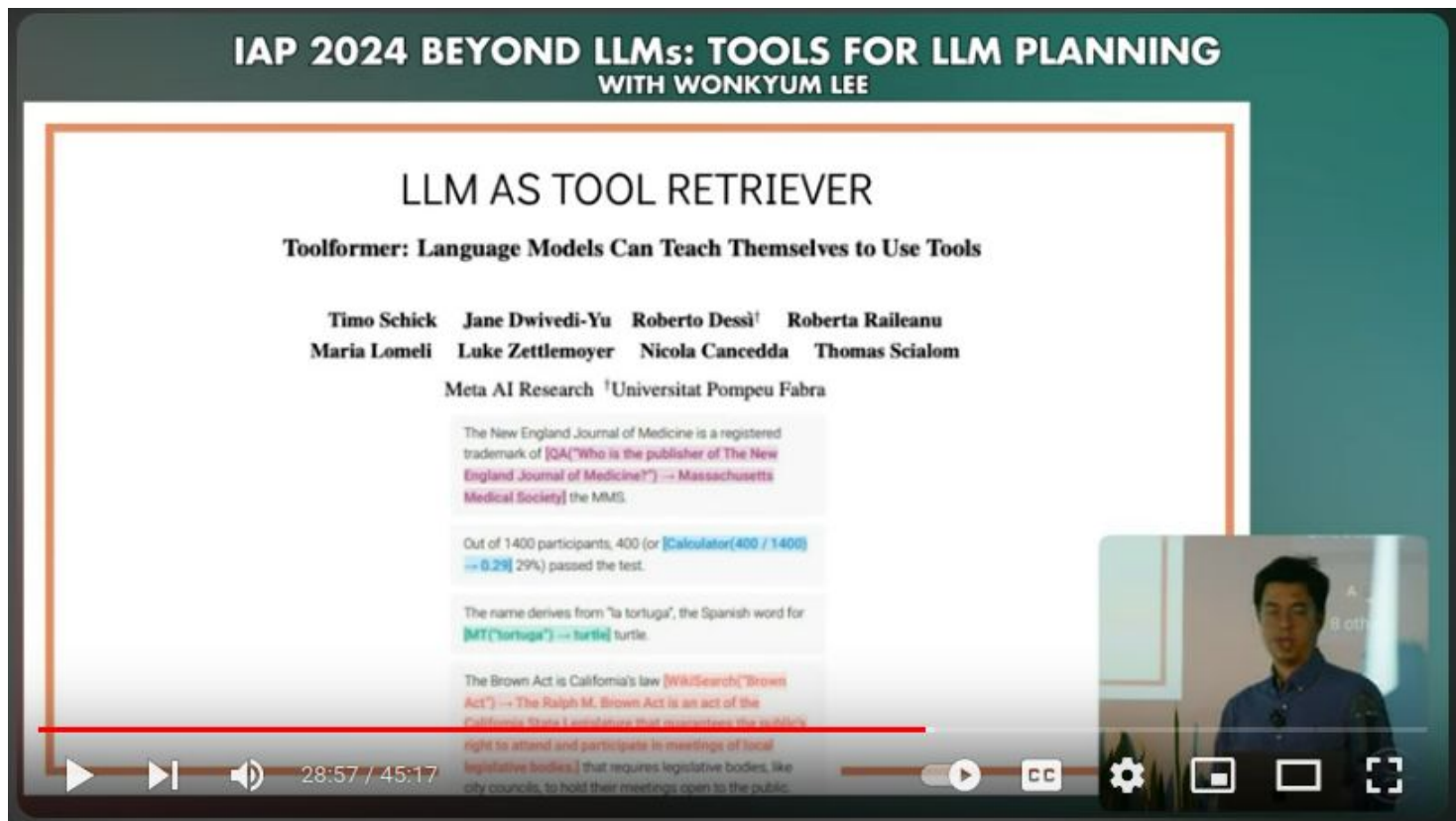
Meta AI Research [†]Universitat Pompeu Fabra

The New England Journal of Medicine is a registered trademark of [QA("Who is the publisher of The New England Journal of Medicine?") → Massachusetts Medical Society] the MMS.

Out of 1400 participants, 400 (or [Calculator(400 / 1400) → 0.29] 29%) passed the test.

The name derives from "la tortuga", the Spanish word for [MT("tortuga") → turtle] turtle.

The Brown Act is California's law [WikiSearch("Brown Act") → The Ralph M. Brown Act is an act of the California State Legislature that guarantees the public right to attend and participate in meetings of local legislative bodies.] that requires legislative bodies, like city councils, to hold their meetings open to the public.



28:57 / 45:17

CC

Recall from week 1: LLM can use tools

System: You may use these tools: “check_temperature”, args: ... usage: ...

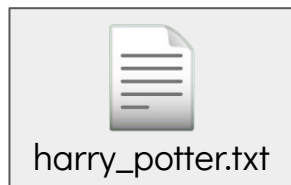
User: What is the temperature in LA now?

LLM: <function call> “check_temperature”, args: “location: LA, time: now”

User: <function return> “check_temperature”, results: “60F”

LLM: The temperature in LA is currently 60F! Nice day for hiking!

Memory in external database



System: You may use these tools: “retrieve_from_db”...

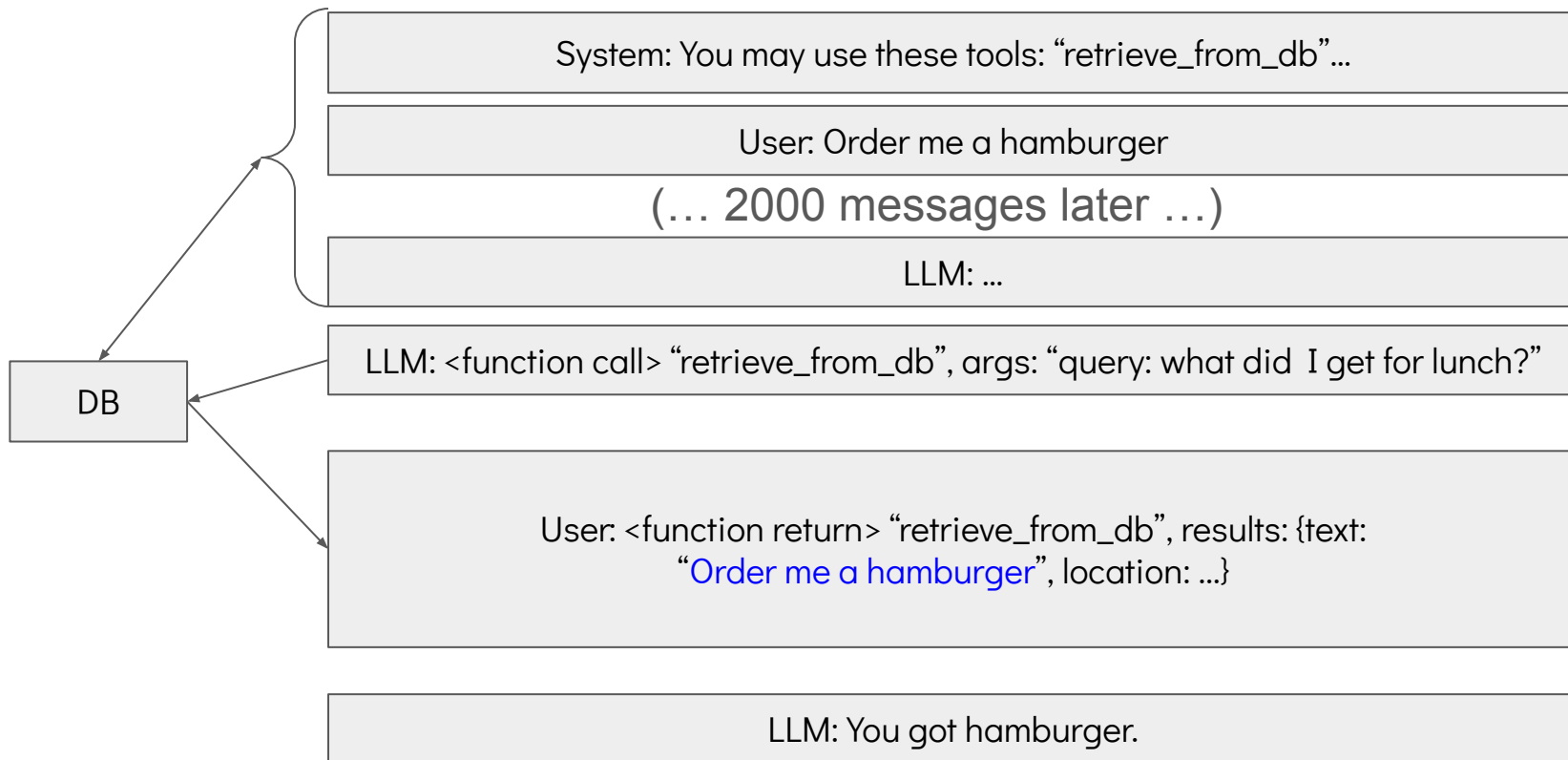
User: What is Harry Potter’s wand made of?

LLM: <function call> “retrieve_from_db”, args: “query: harry wand material”

User: <function return> “retrieve_from_db”, results: {text: “Tricky customer, eh? Not to worry, we’ll find the perfect match here somewhere – I wonder, now – yes, why not – unusual combination – holly and phoenix feather, eleven inches, nice and supple.”, location: “book 1, page 65, paragraph 6”}

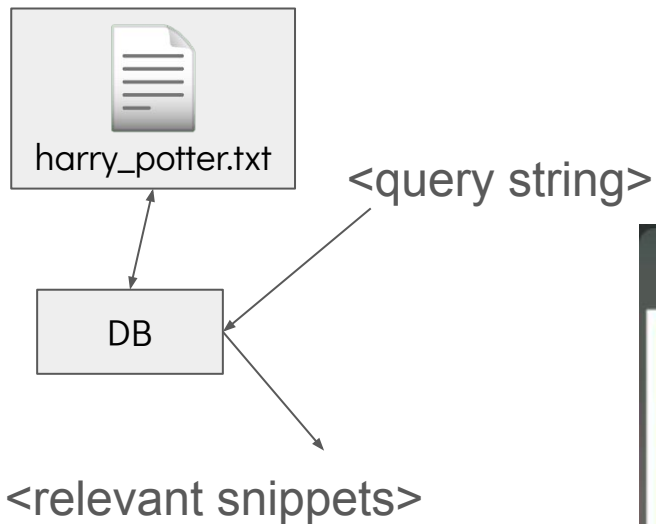
LLM: Harry Potter’s wand is made of phoenix feather.

Memory in external database



Memory in external database (Reversed Index)

- Can be BM25



IAP 2024 BEYOND LLMs: CAN LLMs REMEMBER?
WITH NICK TANTIVASADAKARN

Reversed Index

Document 1
The bright blue butterfly hangs on the breeze.

Document 2
It's best to forget the great sky and to retire from every wind.

Document 3
Under blue sky, in bright sunlight, one need not search around.

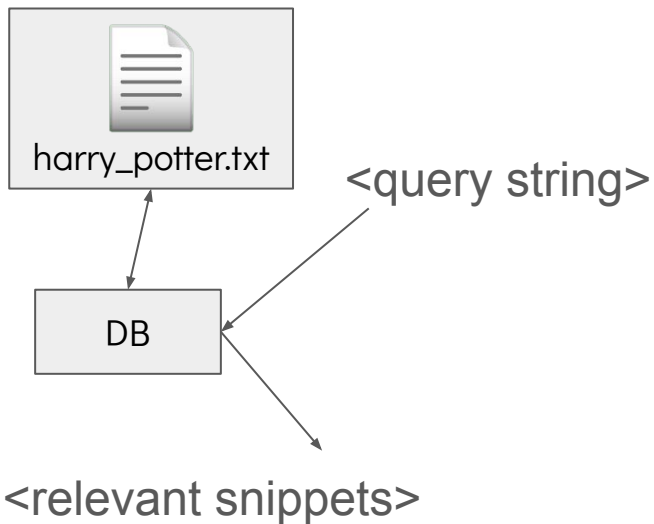
Stopword list
a
and
around
every
for
from
in
it
not
on
one
the
to
under

Inverted index

ID	Term	Document
1	best	2
2	blue	1, 3
3	bright	1, 3
4	butterfly	1
5	breeze	1
6	forget	2
7	great	2
8	hangs	1
9	need	3
10	retire	2
11	search	3
12	sky	2, 3
13	wind	2

30:59 / 50:20

Memory in external database (Transformer-based retrieval)



- Can be transformer-based
- eg. [Colbertv2](#)

ColBERTv2:

Effective and Efficient Retrieval via Lightweight Late Interaction

Keshav Santhanam*
Stanford University

Omar Khattab*
Stanford University

Jon Saad-Falcon
Georgia Institute of Technology

Christopher Potts
Stanford University

Matei Zaharia
Stanford University

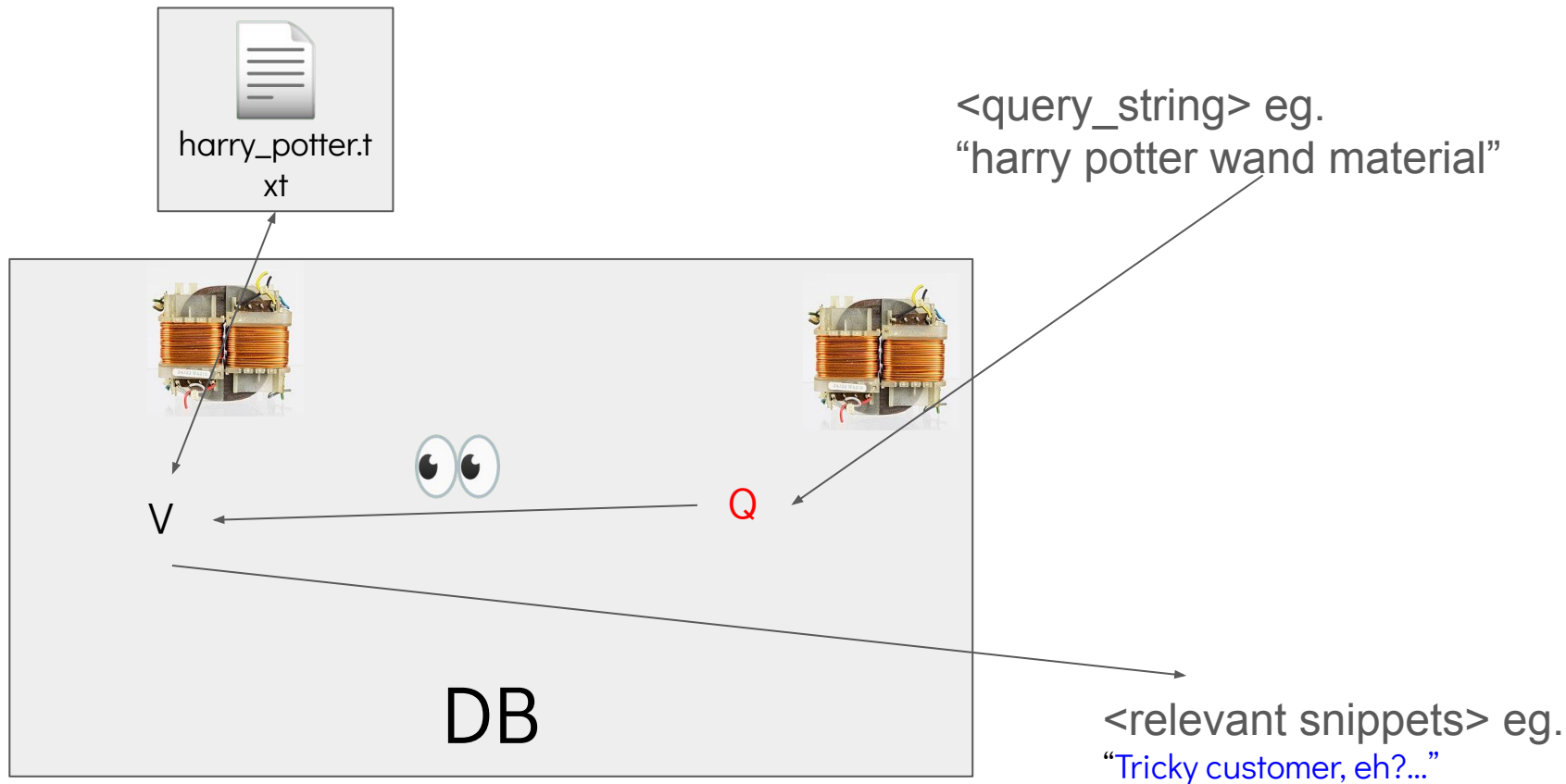
Abstract

Neural information retrieval (IR) has greatly advanced search and other knowledge-intensive language tasks. While many neural IR methods encode queries and documents into single-vector representations, late interaction models produce multi-vector representations at the granularity of each token and decompose relevance modeling into scalable token-level computations. This decomposition has been shown to make late interaction more

relevance is estimated using rich yet scalable interactions between these two sets of vectors. ColBERT produces an embedding for every token in the query (and document) and models relevance as the sum of maximum similarities between each query vector and all vectors in the document.

By decomposing relevance modeling into token-level computations, late interaction aims to reduce the burden on the encoder: whereas single-vector models must capture complex query–document re-

Memory in external database (Transformer-based retrieval)



Recall from week 1: Voyager skill retrieval

IAP 2024 BEYOND LLMs: TOOLS FOR LLM PLANNING WITH WONKYUM LEE

Voyager(Minecraft)

Skill Library

Program Generated by GPT-4

```
async function combatZombie(bot) {  
  // Equip a weapon  
  const sword = ...  
  bot.inventory.findInventoryItem(  
    bot.data.itemName[  
      "stone_sword"  
    ], id  
  );  
  if (sword) {  
    await bot.waitForItem("sword");  
  } else {  
    await craftStoneSword(bot);  
  }  
  // Craft and equip a shield  
  ...  
  // Recover hunger  
  ...  
  // Look for and combat a zombie  
  ...  
}
```

Program Description

```
async function combatZombie(bot) {  
  // The function is about  
  // equipping a stone sword to combat  
  // a zombie. If a stone sword is not  
  // found, it will craft one.  
  Additionally, it crafts and equips  
  a shield for added protection.  
  Afterwards, it proceeds to cook  
  sticks in order to restore hunger.  
  Once hunger is replenished, it  
  actively searches for a zombie and  
  engages in combat with it.  
}
```

Skill Library

- Mine Wood Log
- Make Crafting Table
- Craft Wooden Pickaxe
- Craft Stone Sword
- Make Furnace
- ...
- Combat Cow
- Cook Steak
- Craft Iron Axe
- Combat Zombie

Task: Craft Iron Pickaxe

How to craft an iron pickaxe in Minecraft?

To craft an iron pickaxe, you need to 3 iron ingots and 2 sticks. Once you have gathered the materials, ...

Environment Feedback

Top-5 Relevant

- Smelt Iron Ingot
- Craft Stick
- Make Crafting T...
- Make Furnace
- Craft Wooden P...

Program Generated by GPT-4

```
async function combatZombie(bot) {  
  // Equip a weapon  
  const sword =  
  bot.inventory.findInventoryItem(  
    mcData.itemsByName[  
      'stone_sword'  
    ],.id  
  );  
  if {sword} {  
    await bot.equip(sword, 'hand');  
  } else {  
    await craftStoneSword(bot);  
    ...  
  }  
  // Craft and equip a shield  
  ...  
  // Recover hunger  
  ...  
  // Look for and combat a zombie  
  ...  
}
```

GPT-3.5

Program Description

```
async function combatZombie(bot) {  
  // The function is about  
  equipping a stone sword to combat  
  a zombie. If a stone sword is not  
  found, it will craft one.  
  Additionally, it crafts and equips  
  a shield for added protection.  
  Afterwards, it proceeds to cook  
  sticks in order to restore hunger.  
  Once hunger is replenished, it  
  actively searches for a zombie and  
  engages in combat with it.  
}
```

Embedding

Key

Add

Value

Skill Library

- Mine Wood Log
- Make Crafting Table
- Craft Wooden Pickaxe
- Craft Stone Sword
- Make Furnace
- ...
- Combat Cow
- Cook Steak
- Craft Iron Axe
- Combat Zombie

Task: Craft Iron Pickaxe

How to craft an iron pickaxe in Minecraft?

GPT-3.5

To craft an iron pickaxe, you need to 3 iron ingots and 2 sticks. Once you have gathered the materials,

Environment Feedback

Embedding

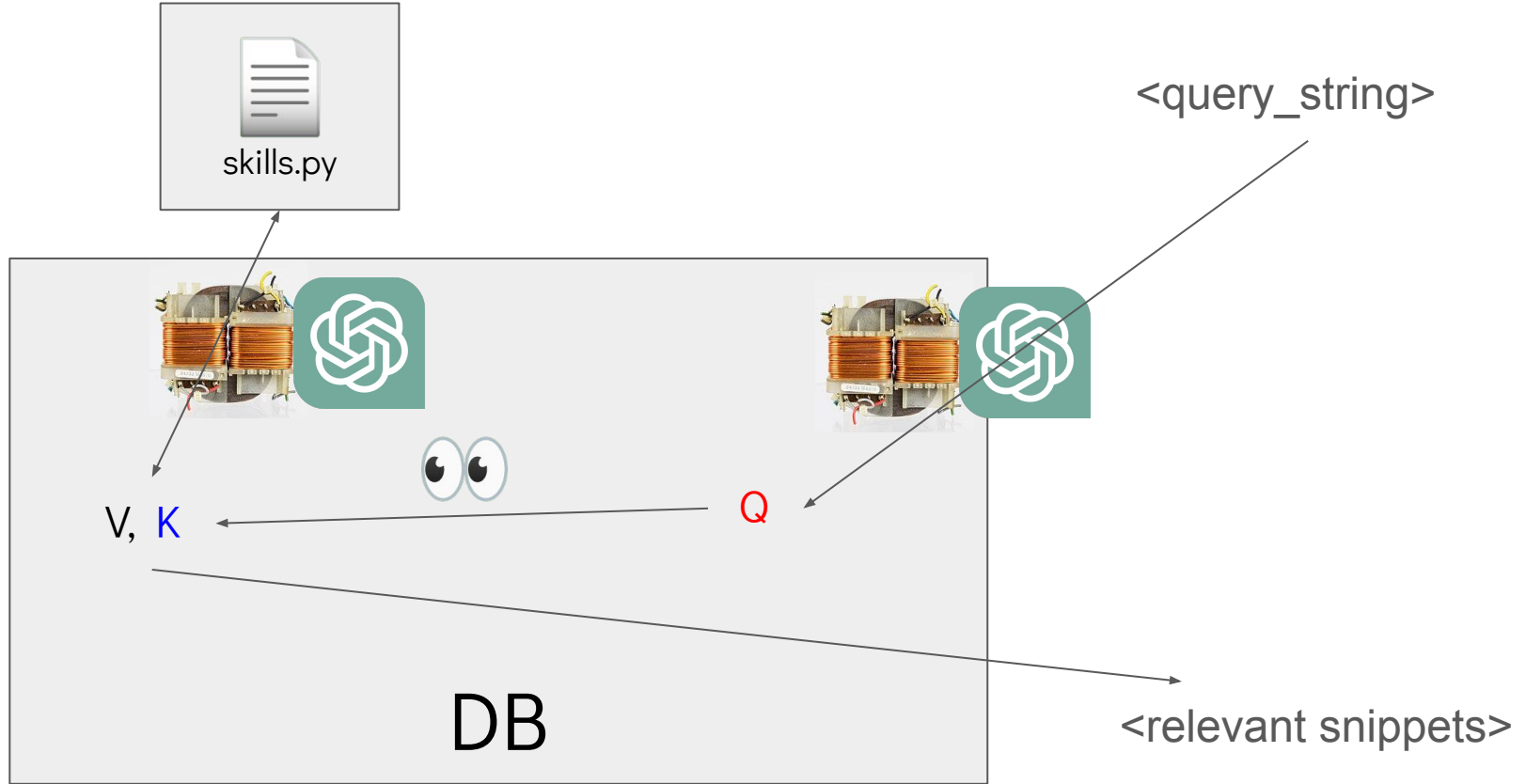
Query

Skill Library

Retrieve

- Smelt Iron Ingot
- Craft Stick
- Make Crafting Table
- Make Furnace
- Craft Wooden Pickaxe

Memory in external database (Transformer-based retrieval)



Memory in external database

the Good:

- Infinite storage

the Bad:

- Needs to build DB
- DB can be tricky to maintain

- Is there already an existing large DB on the shelf?

Scenario: Hogwarts have wifi and nothing in the rules says you cannot summon a computer





Outline

- Problem: Information retrieval
- Solution 0: Memory in training data
- Solution 1: Memory in LLM Context
- Solution 2: Compress chat history
- Solution 3: Memory in external database
- **Solution 4: Let LLM browse the internet**

Let LLM browse the internet

- Lookup Google
- Lookup Wikipedia
 - Can do “fact checks” too!

WikiChat: Stopping the Hallucination of Large Language Model Chatbots by Few-Shot Grounding on Wikipedia

Sina J. Semnani Violet Z. Yao* Heidi C. Zhang* Monica S. Lam

Computer Science Department

Stanford University

Stanford, CA

{sinaj, vyao, chenyz, lam}@cs.stanford.edu

Abstract

This paper presents the first few-shot LLM-based chatbot that almost never hallucinates and has high conversationality and low latency. WikiChat is grounded on the English Wikipedia, the largest curated free-text corpus.

WikiChat generates a response from an LLM,

after their pre-training, and are far less knowledgeable about less popular, or *tail*, topics (Mallen et al., 2022; Sun et al., 2023). Therefore, for knowledge-intensive tasks (Lewis et al., 2020), users need to painstakingly verify any information they receive with external sources lest they be misled.

This paper focuses on three metrics for

<https://arxiv.org/pdf/2305.14292v2.pdf>

2023

User: Have you heard about the upcoming film Oppenheimer?
WikiChat: Yes! Oppenheimer is a biopic ...
User: What do you think about the casting?

1. Query

Query: cast of Oppenheimer
 Christopher Nolan film
Date: recent



Retrieve

Title: Christopher Nolan Nolan's twelfth directorial venture will be "Oppenheimer" (2023), ...
Title: Oppenheimer (film) ... Much of the main cast signed on between September 2021 and April 2022., ...
Title: Cillian Murphy ... Murphy was confirmed to star as J. Robert Oppenheimer in Nolan's upcoming film Oppenheimer ...



3. Generate

Christopher Nolan is known for his meticulous casting choices. He often casts well-known actors in his films, and this time he has selected Tom Hanks and ... Michael Caine.

2. Summarize & Filter

- Much of the Oppenheimer (2023) movie's cast signed on between September 2021 and April 2022.
- Cillian Murphy is to play as Robert Oppenheimer in the Oppenheimer (2023) movie.

- Christopher Nolan is known for his meticulous casting choices.
- Christopher Nolan often casts well-known actors in his films.
- Tom Hanks stars in the Oppenheimer (2023) movie.
- Michael Caine stars in the Oppenheimer (2023) movie.

4. Extract Claims

Christopher Nolan has cast Cillian Murphy as the titular role of Oppenheimer, keeping with his tradition of meticulous casting choices. I'm sure it will be an excellent film!

7. Refine

Cillian Murphy has been cast to play Robert Oppenheimer in Christopher Nolan's "Oppenheimer" movie. Christopher Nolan is known for his meticulous casting choices and often casts well-known actors in his films. The movie's cast was signed on between September 2021 and April 2022.

6. Draft

5. Fact-Check

- Christopher Nolan is known for his meticulous casting choices.
- Christopher Nolan often casts well-known actors in his films.



Retrieve

Let LLM browse the internet

the Good:

- Simple to implement
- Cheap computation (on our side)

the Bad:

- Beware of internet trolls
- Dependent on third party servers

Outline

- Problem: Information retrieval
- Solution 0: Memory in training data
- Solution 1: Memory in LLM Context
- Solution 2: Compress chat history
- Solution 3: Memory in external database
- Solution 4: Let LLM browse the internet
- Q&A

Exercises:

- [Mistral-7B-v0.1](#) uses sliding window attention. What's its window length & number of layers? What's its theoretical attention span?
- Can you suggest some reasons that the attention span in practice is lower than theoretical span? How to remedy?
- What are some of the pain points of depending on third-party search engines?

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