





## **Paper Link**

## The Knowledge Alignment Problem:

Bridging Human and External Knowledge for Large Language Models

Shuo Zhang<sup>1</sup>

Liangming Pan <sup>2</sup>

Junzhou Zhao <sup>1</sup>

William Yang Wang <sup>2</sup>

<sup>1</sup> Xi'an Jiaotong University

<sup>2</sup> University of California, Santa Barbara

#### LLM Hallucination

## User Input



Can you recommend a delicious recipe for dinner?

## LLM Response



Yes, here is a delicious recipe for lunch. So how about fried chicken with mashed potatoes? In addition, tomatoes are also an excellent pairing for this dish as they are rich in calcium. Enjoy this steak!

## Hallucination Explanation

Input-Conflicting Hallucination: the user wants a recipe for <u>dinner</u> while LLM provide one for <u>lunch</u>.

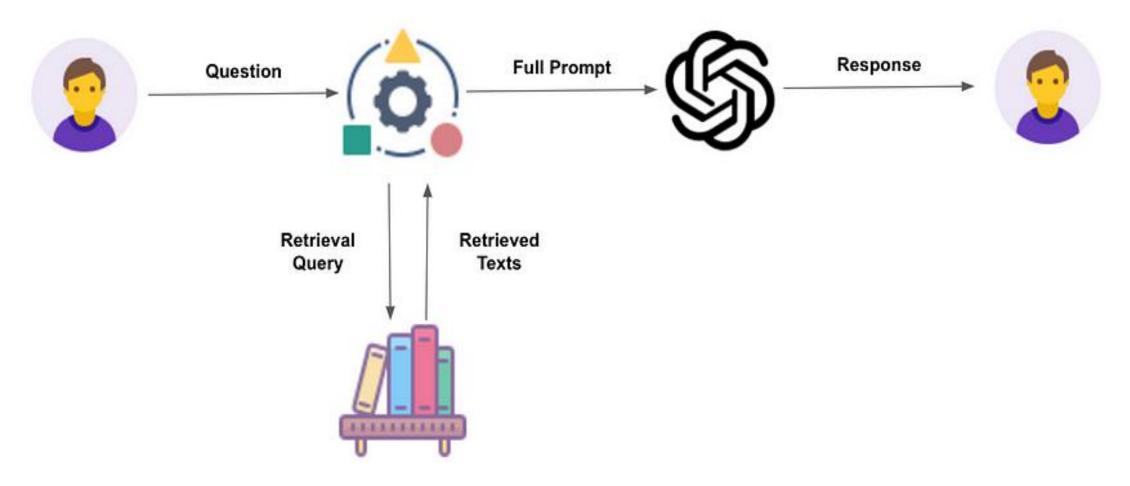
Context-Conflicting Hallucination: <u>steak</u> has not been mentioned in the preceding context.

Fact-Conflicting Hallucination: <u>tomatoes</u> are not rich in <u>calcium</u> in fact.

#### Source paper:

Siren's Song in the Al Ocean: A Survey on Hallucination in Large Language Models

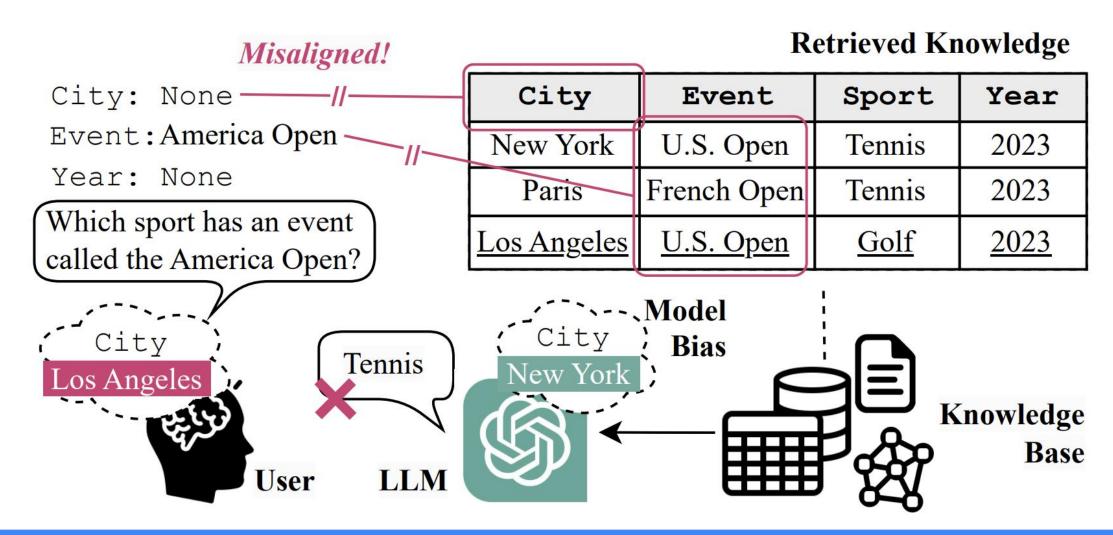
## Retrieved-Augmented Generation



**Pic Source:** https://www.linkedin.com/pulse/what-retrieval-augmented-generation-grow-right

## Retrieved-Augmented Generation Still Hallucinates!

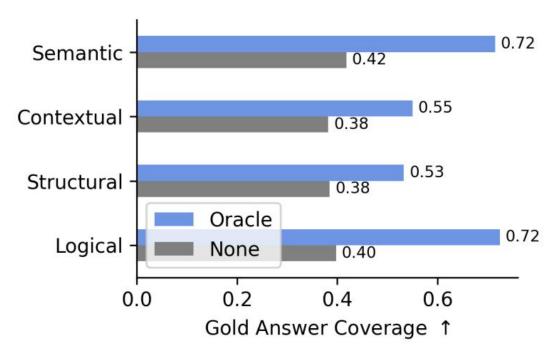
## when there are *Knowledge Misalignments*

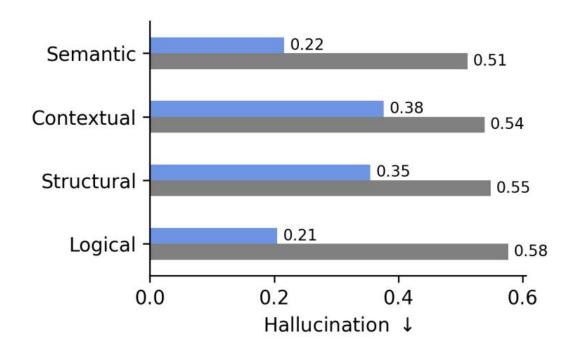


## Knowledge Misalignments Types

Туре	Explanation	Example
Semantic	The user might use an ambiguous term that, while ideally should map to a single item in the database, in reality can correspond to multiple attributes or values.	For "What is the best burger?", when you say "best burger", are you referring to taste, nutritional value, price, or a combination of these attributes?
Contextual	The user may have implicitly established some conditions without explicitly expressing them.	For "What is the 15th most populous city in the United States?", the statistics may vary with time, which year are you considering?
Structural	The user might have stated some conditions that are not addressed in the database.	For "Find me an American writer.", the question can not be answered when the database does not include the nationality of the writers.
Logical	The user can ask complex questions where certain conditions need to be determined before other conditions can be clarified, while the database only supports basic logits such as "and", "or" and "not".	For "Fine me for a movie directed by the singer who has won a Grammy', the question answer- ing requires first identifying singer who have won a Grammy, and then finding the films di- rected by the identified singers.

## Knowledge Misalignments → lower QA accuracy + more Hallucination





Clarification → Significant improvements

Type	Clarifying Information
Semantic	The term 'A' in the question refers to 'B' in our database.
Contextual	The value for the missing contextual condition in the question is 'A'.

Structural	The value for the condition 'B' in the database is 'A'.
Logical	The complex condition 'A' in your question refers to the condition 'B' in the database.

 ${
m MIXALIGN}$  Find knowledge misalignments and ask for clarifications from users

(Q) Which sport has an event called th	Grounding Knowledge $(K)$					
Constraint Extraction	<b>4</b>	City	Event	Sport	Year	
	City, Event	New York	U.S. Open	Tennis	2023	
America Open City: None	Sport, Tear	Paris	French Open	Tennis	2023	
		Los Angeles	U.S. Open	<u>Golf</u>	2023	
Explicit Constraint Matching	French Open,	<u></u>	<u></u>	<u></u>		
Alignment Feedback	U.S. Open	<u> </u>				
Not confident what America Open	refers to	City	Event	Sport	Year	
<b>*</b>		New York	U.S. Open	Tennis	2023	
Clarifying Question Generation	Candidate	Los Angeles	U.S. Open	<u>Golf</u>	2023	
Clarifications (C)  Is the event you are referring to U.S. Open?  See Endidate Filtering Distinguishable Attribute Selection  Alignment Feedback						
Which city hosted America Open?  Ouestion  Candidates differs in city, which can be New York or L.A.						
Answer Generation $A = LLM(Q, K, Q)$			Knowledge Align	nment: Explicit	Implicit	

## Experimental Settings

Task: RAG with controlled knowledge Shuffle(Gold + N Irrelevant)

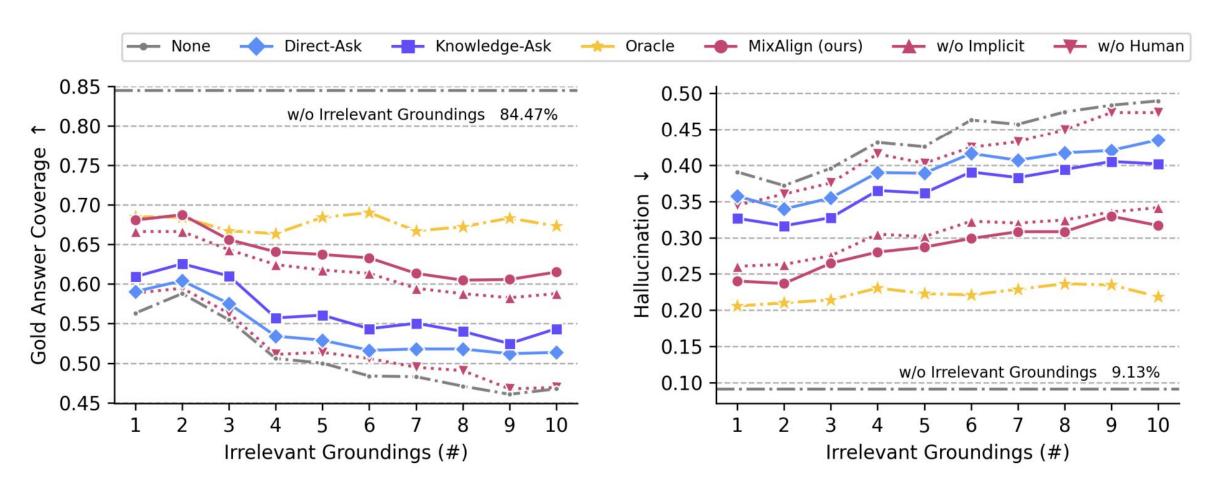
Dataset: FuzzyQA constructed by simplifying questions in the OTT-QA dataset

Baselines:

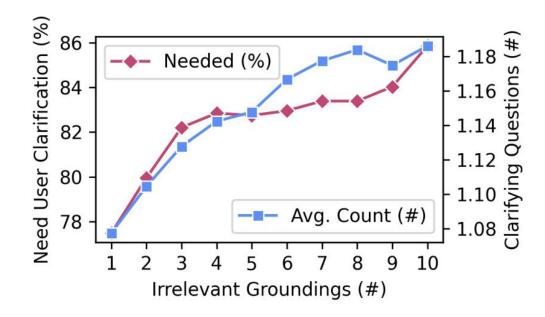
- None. No clarification included
- *Oracle.* Reverse-generate the oracle clarifications from the ground-truth answer and knowledge
- Direct-Ask. Clarifying questions are asked based solely on the original question. "In order to answer this question, I have to ask the following clarifying question:"
- Knowledge-Ask. Incorporate candidate knowledge to generate clarifying questions

## Main Results

## Stable Improvements



User question	What are the campaign notes of the 2012 Democratic candidate born on May 23, 1958?	Where did the Magic: The Gathering player known as "The Great One" win?
Direct-Ask	What specific information are you looking for regarding the 2012 Democratic candidate?	What format of Magic: The Gathering was "The Great One" playing?
Knowledge-Ask	Which candidate was born on May 23, 1958?	Which season did Jon Finkel win?
MIXALIGN (ours)	Is the candidate you are referring to Keith Judd?	Which season did "The Great One" win?



## Case Study & Cost Analysis

Figure 6: Efficiency analysis results of MIXALIGN. We target: 1. Needed (%): How often do we need to request user clarification after the initial LLM-based constraint matching? 2. Avg.Count (#): When needed, how many clarifying questions, on average, are posed?

### Conclusion & Future work

- Introducing the Knowledge Alignment problem
- Proposed MixAlign
- Detailed Experiments Demonstrating Effectiveness
- address other knowledge formats and modalities
- improve efficiency of the pipeline

## Thanks

zs412082986@stu.xjtu.edu.cn



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MOE KLINNS Lab, Xi'an Jiaotong University University of California, Santa Barbara