

Retrieval-Augmented Generation for Large Language Models

GenAl BootCamp

YUAN HE | AUG 2024

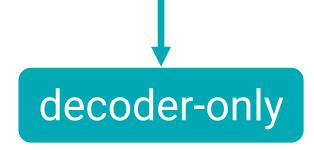
What is Language Model?

Language Model

- A statistical model that determines the probability distribution of linguistic units (words, sentences, etc.) in a language.
- Sequential language modelling: $P(w_i|w_{< i})$
 - E.g., London is the capital of the UK
- Masked language modelling: $P(w_i|w_{< i}, w_{> i})$
 - E.g., London is the [MASK] of the UK.

Large Language Model

- "Large" is a relative term
- Often consider GPT-3 (2020) as one of the first LLMs
 - GPT-3 scales up 100x parameters compared to its predecessors
- Often assumed as generative sequential language modelling



Prompt for LLMs

- Recall sequential language modelling: $P(w_i|w_{< i})$
- Prompt is the conditioning text c: $P(w_i|c, w_{< i})$
- Prompts can be queries, instructions, and auxiliary contexts, e.g.,
 - "What is a prompt in the context of generative AI?"
 - "Translate the following into Chinese: [TEXT] "
 - "Given [EHR], what would be the best course of treatment for managing their newly diagnosed chronic kidney disease?"

LLMs are powerful, but

Challenges with LLMs

Hallucination

LLMs tend to produce fabricated information when they don't know

Data Privacy Concerns

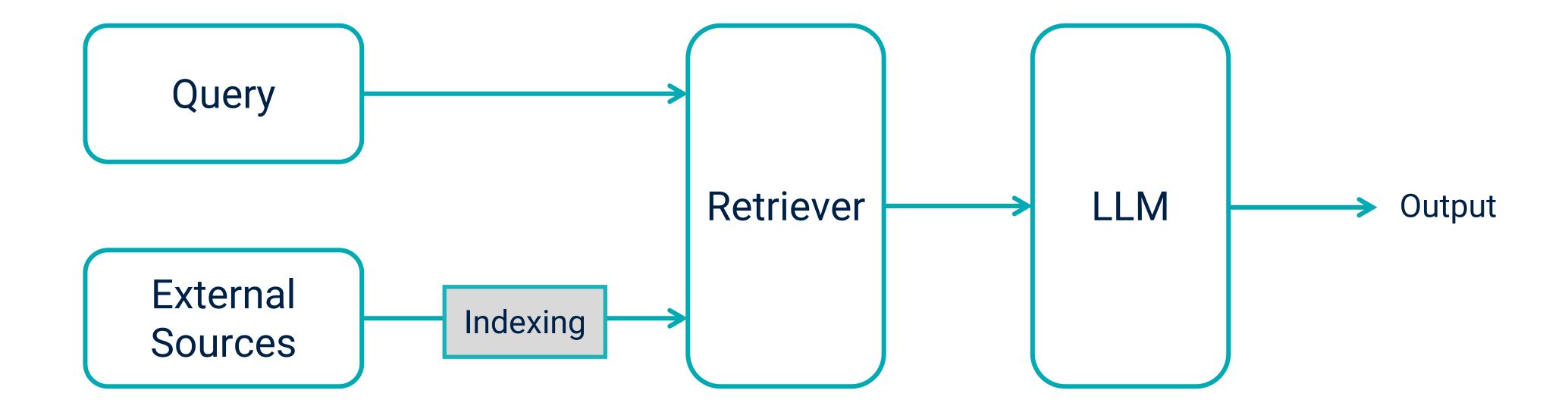
 Training on proprietary data (e.g., EHR, company data) raises privacy and security issues; we prefer to avoid it

Adaptability

 Difficulties in incorporating new knowledge on the fly make LLMs less responsive to real-time changes

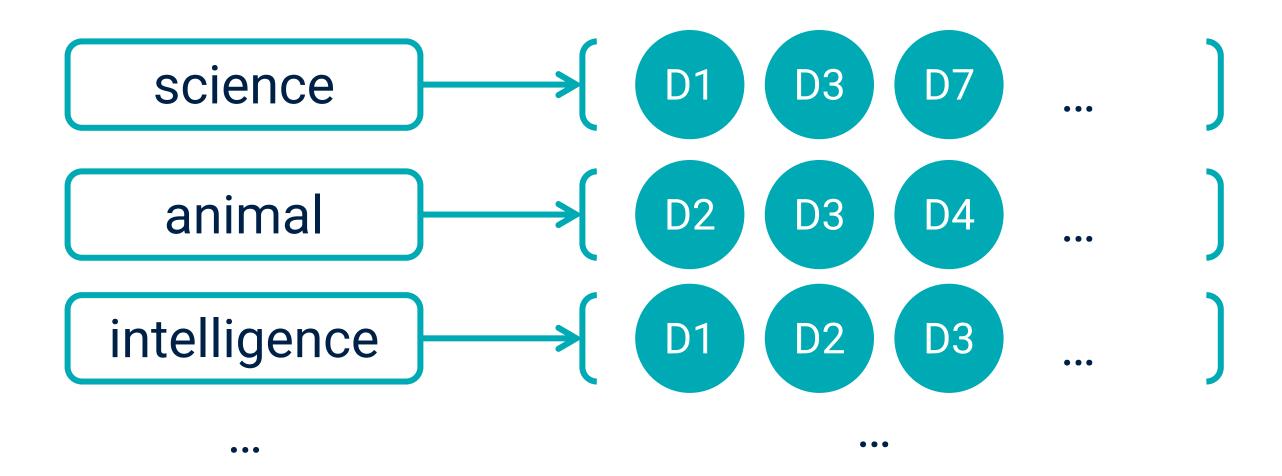
RAG is a possible solution

Retrieval-Augmented Generation (RAG)



Indexing

- Inverted Index (Sparse Retrieval)
 - Building an Inverted Index to store token-documents pairs
 - Retrieve through statistical algorithms like TF-IDF, BM25



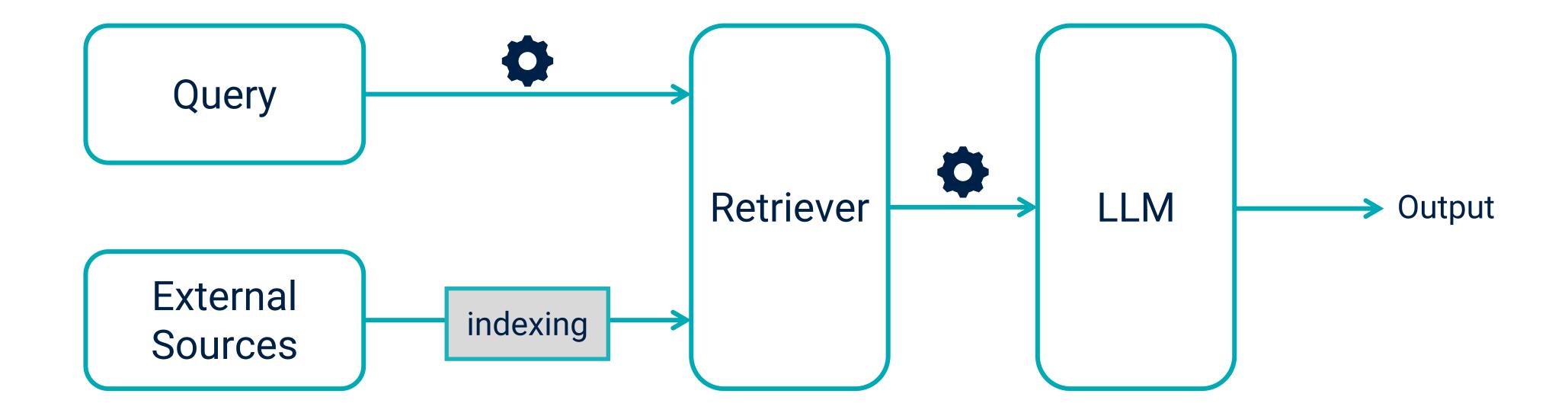
Indexing

- Vector Database (Dense Retrieval)
 - Building a Vector Database to store document embeddings
 - Retrieve through machine learning algorithms like KNN, ANN



Can we do any better?

RAG + Pre/Post-processing



Pre-Retrieval Processing

Routing

- Directing a query to the most appropriate source
- Or even not using a retriever

Rewriting

- Refine the query based on a feedback mechanism
- Convert the query into a formal query (for GraphRAG)

Expansion

Incorporating meta-data, instructions, etc.

Post-Retrieval Processing

Reranking

 Adjusting the order of retrieved documents or results based on more advanced criteria (e.g., beyond similarity)

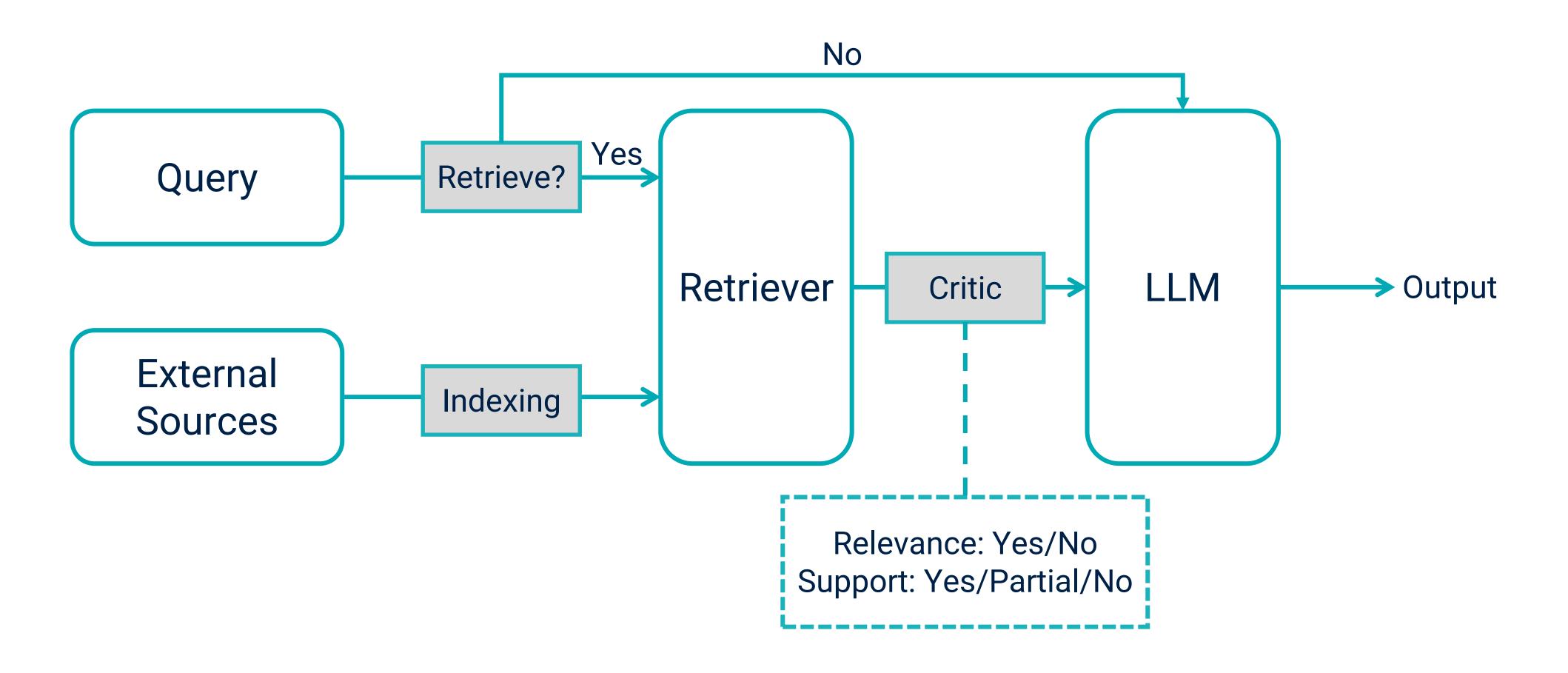
Summary

Input prompt is too long -- generating concise summaries of the retrieved documents

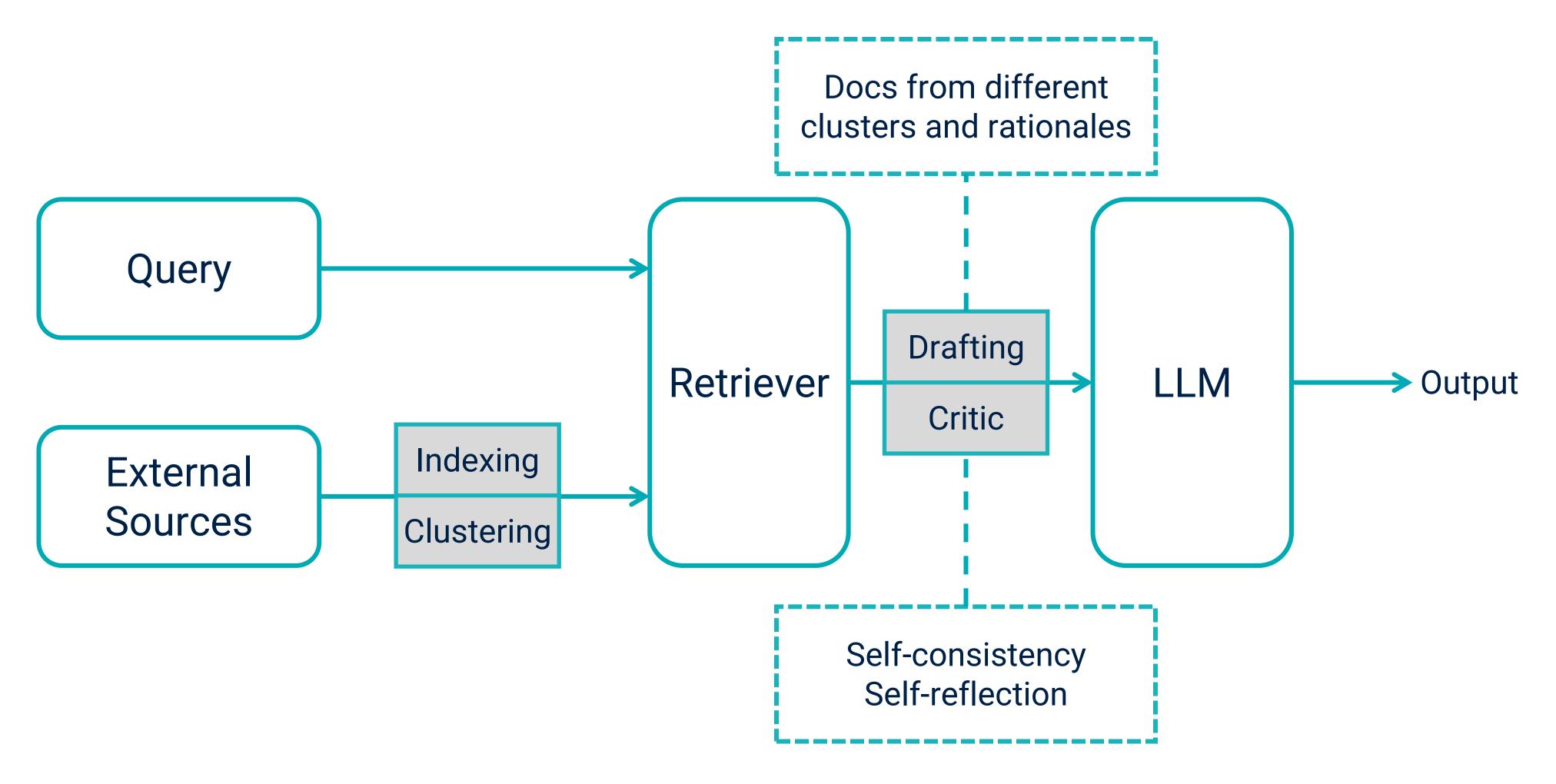
Fusion

Merging retrieved contents from various sources or multiple queries

Self-Reflective RAG



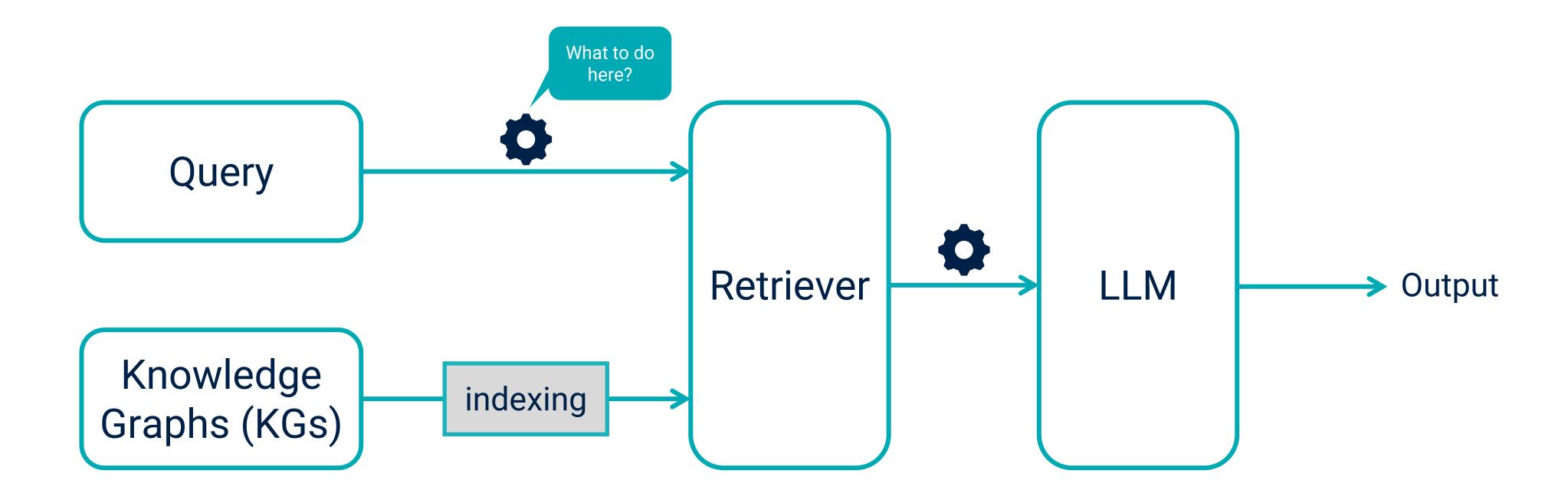
Speculative RAG



Wang et al. "Speculative RAG: Enhancing Retrieval Augmented Generation through Drafting". Arxiv 2024.

Beyond text documents?

GraphRAG

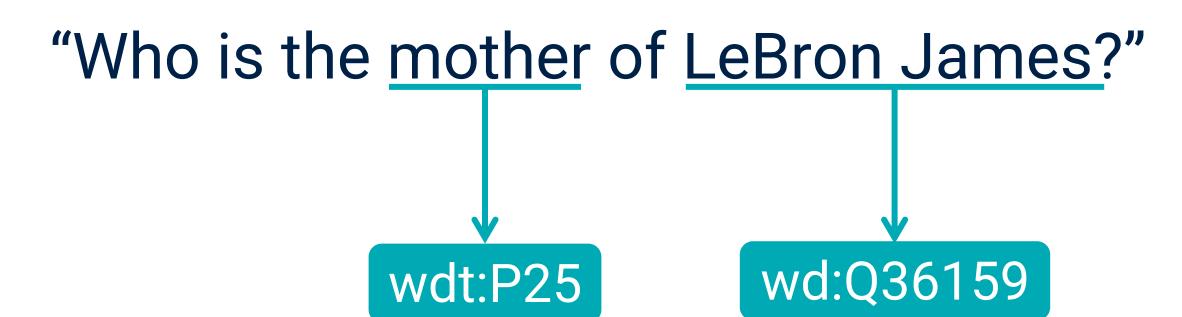


Pre-Retrieval Processing (GraphRAG)

Entity Recognition

Entity Linking

Query Rewriting



```
PREFIX wd: <a href="http://www.wikidata.org/entity/">PREFIX wd: <a href="http://www.wikidata.org/prop/direct/">PREFIX wdt: <a href="http://www.wikidata.org/prop/direct/">http://www.wikidata.org/prop/direct/</a>

SELECT ?motherLabel

WHERE {
    wd:Q36159 wdt:P25 ?mother .
    SERVICE wikibase:label { bd:serviceParam wikibase:language "[AUTO_LANGUAGE],en".}
}
```

Pre-Retrieval Processing (GraphRAG)

Alternative 1

It is possible to train an LLM to do the whole processing altogether

Alternative 2

Without converting the query into a formal query, we can conduct the retrieval over KG embeddings

GraphRAG vs RAG

Pros

- Better reasoning
- More accurate context
- Handle complex query

Cons

- Rely on the completeness of KGs
- Increased complexity
- Slower retrieval

Conclusion

- RAG is essentially information retrieval + generation
- RAG can help LLMs:
 - Reduce hallucination
 - Avoid expensive re-training
 - Access proprietary data without memorising them
- RAG can be improved by multi-stage processing and critic modules
 - GraphRAG is an extension of RAG to handle structured knowledge bases

THANKS!

