



KET-RAG: A Cost-Efficient Multi-Granular Indexing Framework for Graph-RAG

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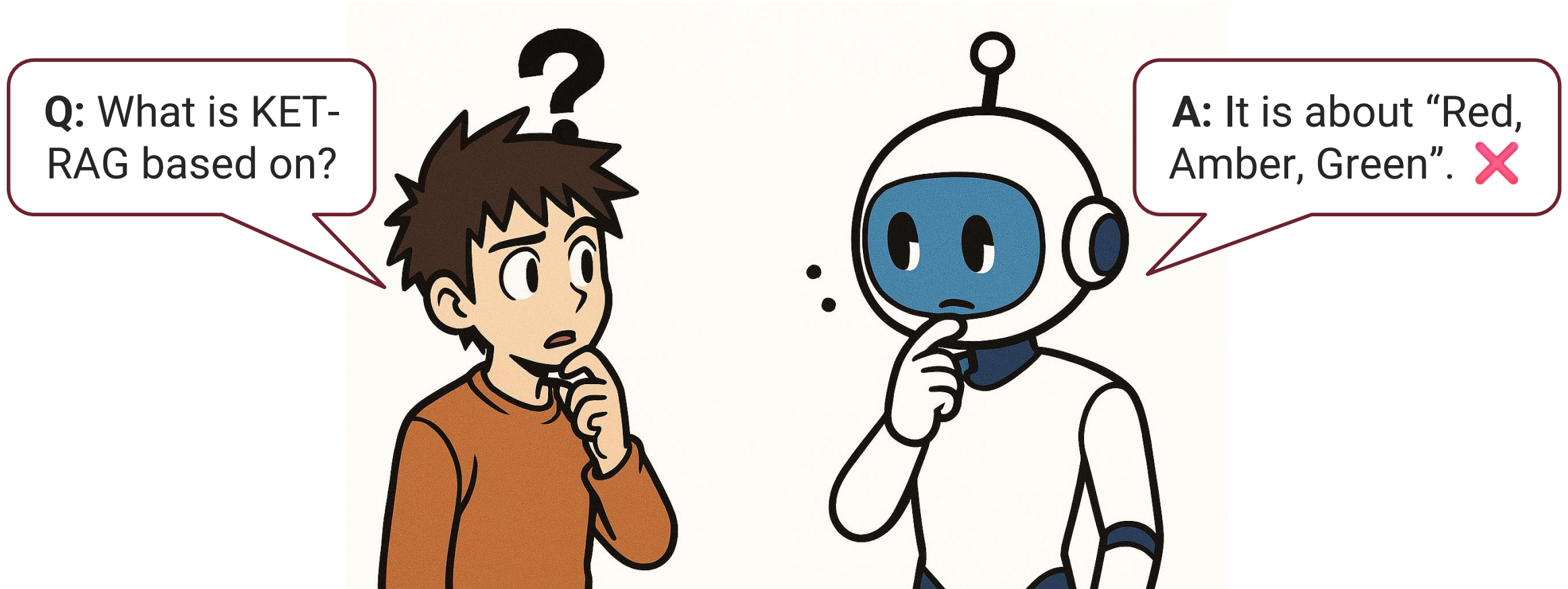


Agenda

- **Background:** RAG and Graph-RAG
- **Motivation:** Graph-RAG's expense
- **Our Method:** KET-RAG
- **Experiment**
- **Future Work**

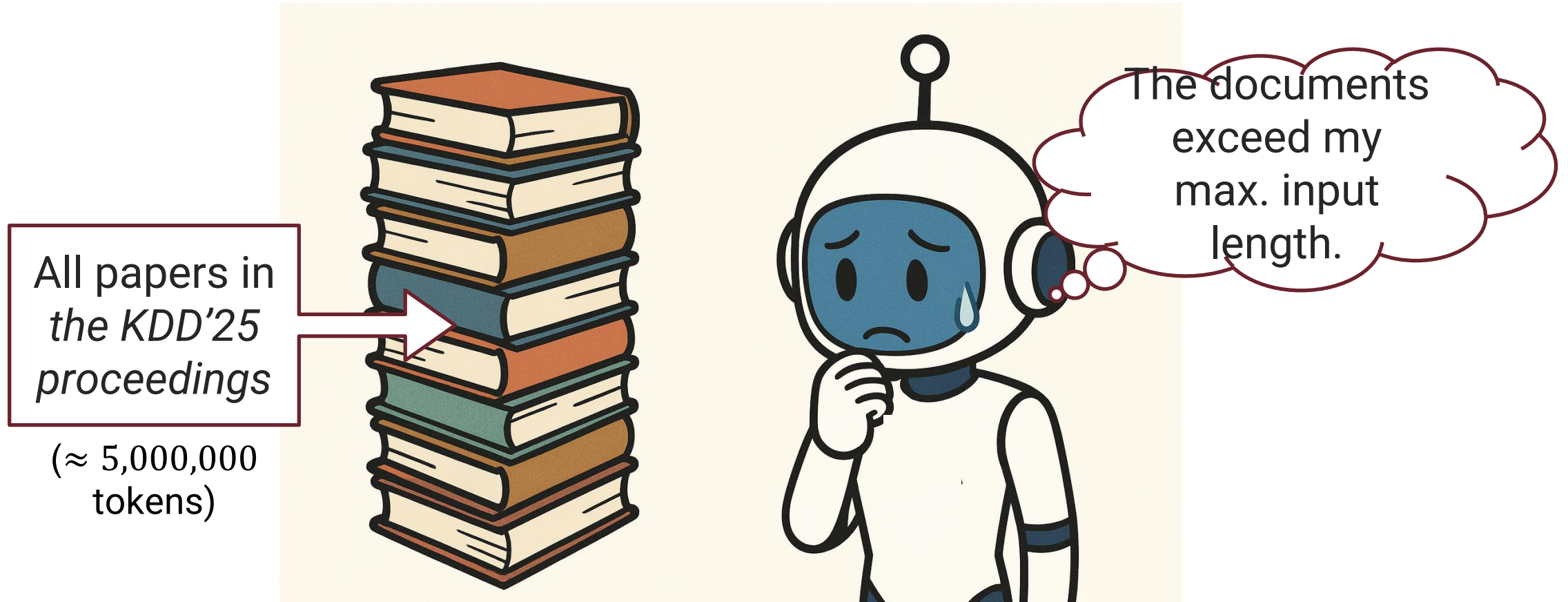


Background: Asking questions to LLMs



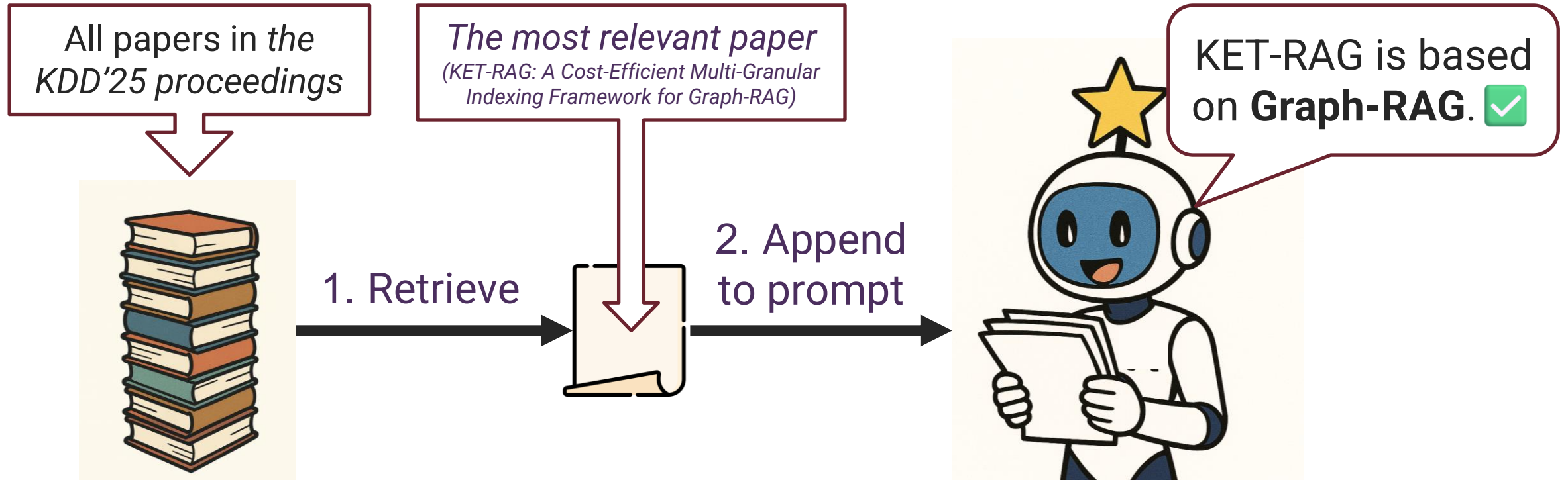
Without external knowledge, large language models (LLMs) can only guess the answer.

Background: Asking questions to LLMs



Even when we have an external text source,
they are too large to provide in full to LLMs.

Background: RAG (Retrieval-Augmented Generation)



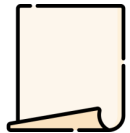
Given an external text source, RAG retrieves relevant context to augment the LLM's generation (e.g., answer).

Background: Graph-RAG

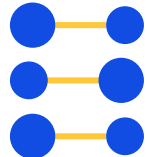
1. Offline indexing



External source



LLMs



Information
Extraction

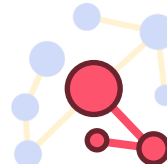


**Knowledge Graph
(KG)**

2. Online retrieval (per query)

Q: What is KET-
RAG based on?

Query from users



Search for
relevant subgraphs



= (“KET-RAG is based on
Graph-RAG“, ...)

Retrieval
result

Background: Graph-RAG

- Toy example of **offline indexing**

External source



0



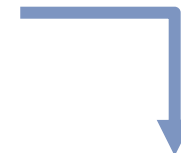
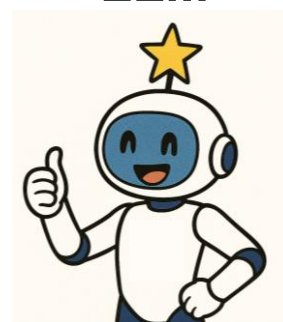
Text chunks

"Graph-RAG constructs a knowledge graph from text chunks to improve retrieval in LLM-based QA."

1

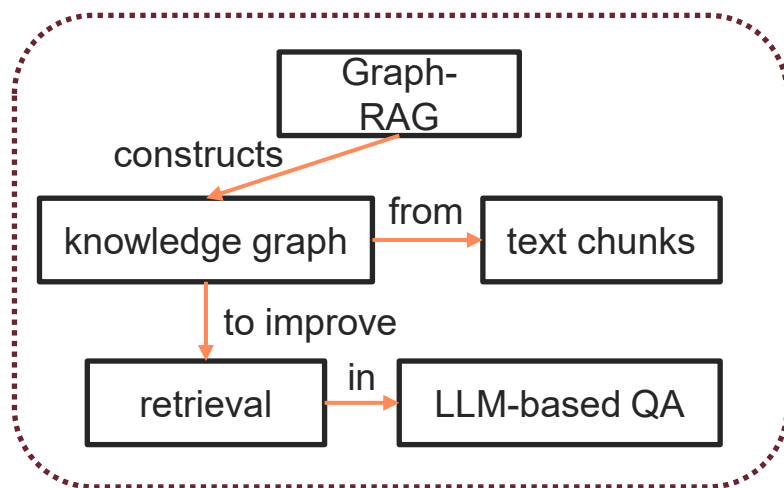


LLM

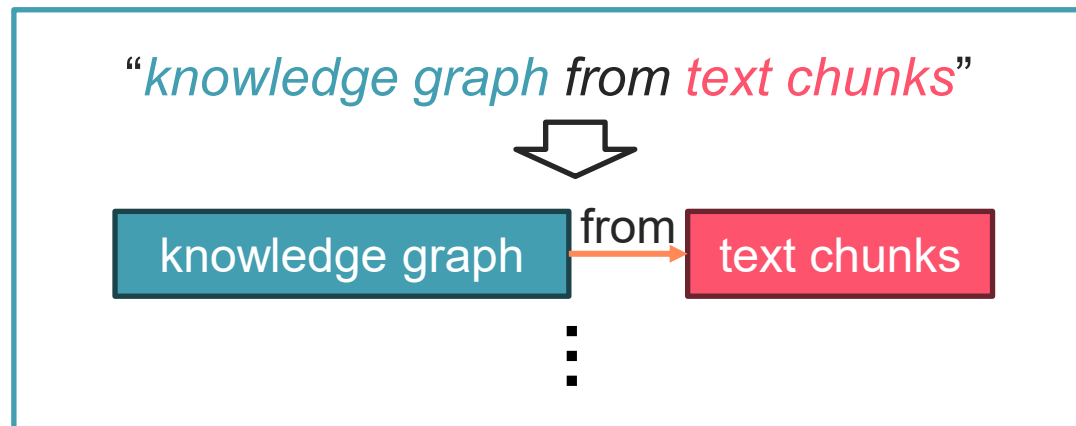


(performs entity & relationship extraction)

2

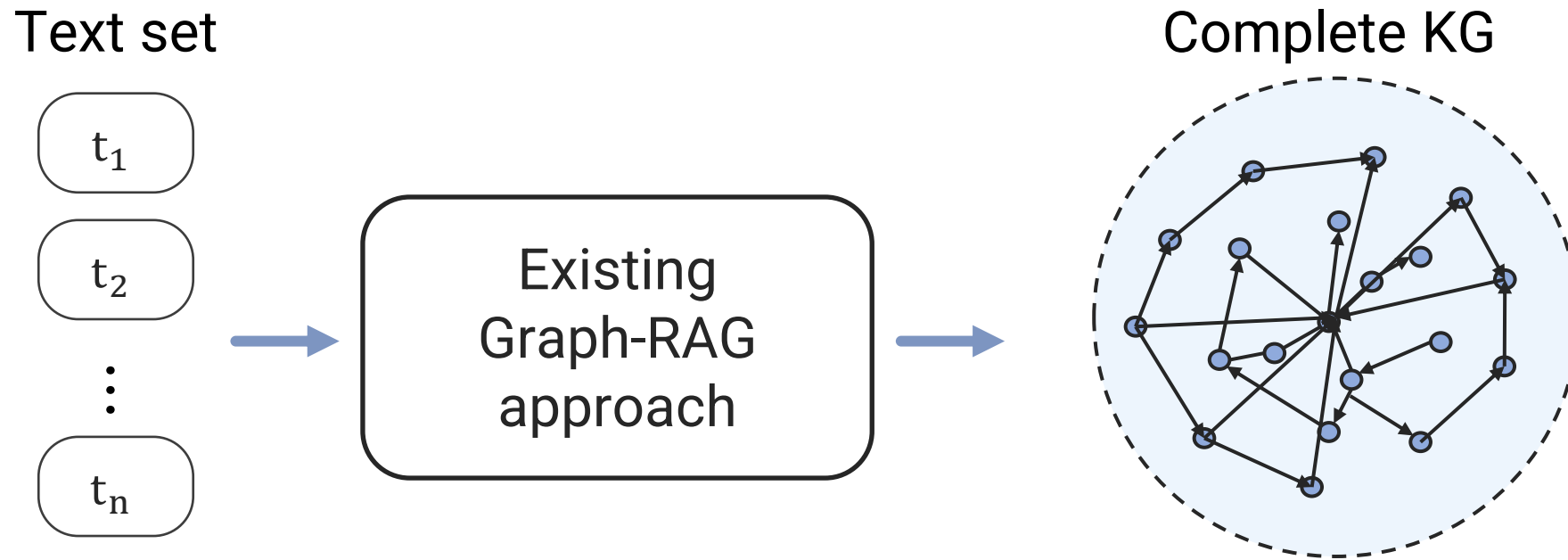


Knowledge graph (KG)



Motivation: Graph-RAG's expense

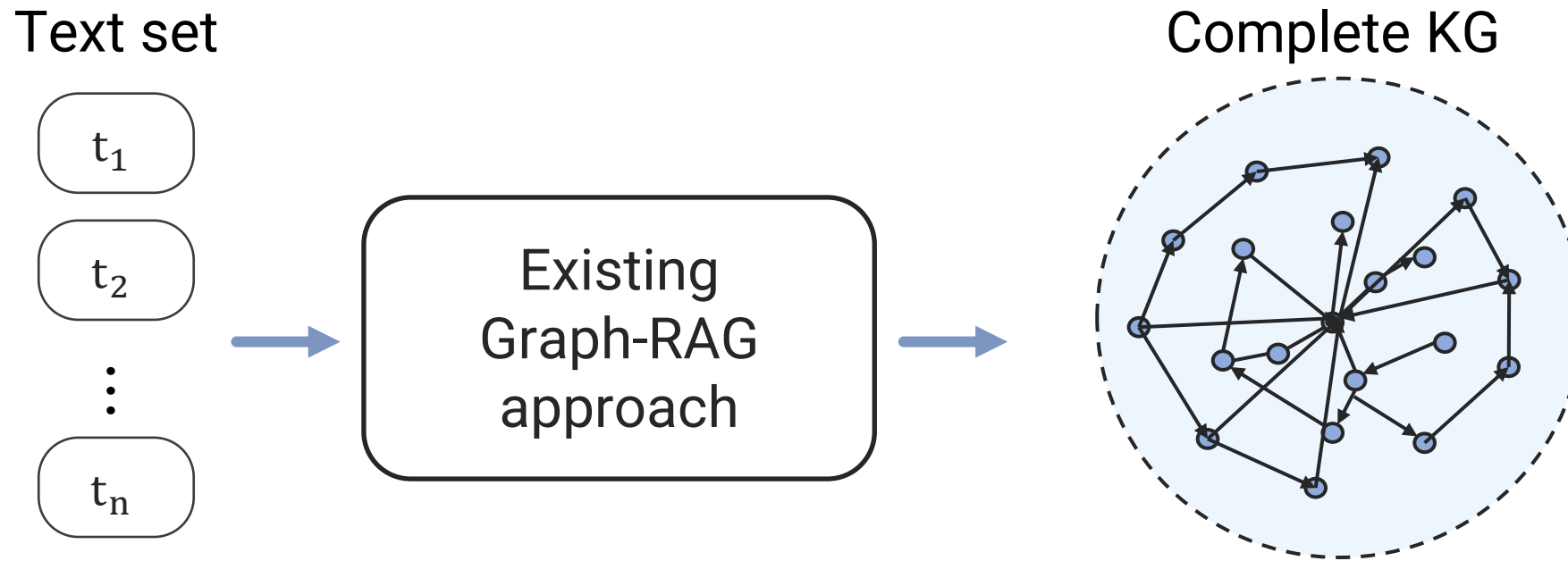
- **Main cost:** constructing the knowledge graph



Whole corpus → **fully** read by LLMs → complete KG

Motivation: Graph-RAG's expense

- **Main cost:** constructing the knowledge graph

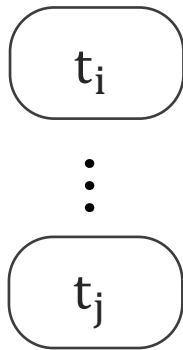


a **3.2MB** dictionary + GPT-4o-mini = **\$21**
a single **5GB** legal case + GPT-4o-mini = **\$33,000**

Motivation: Graph-RAG's expense

- **Idea:** we can only use **important** text

Important
text set



Existing
Graph-RAG
approach



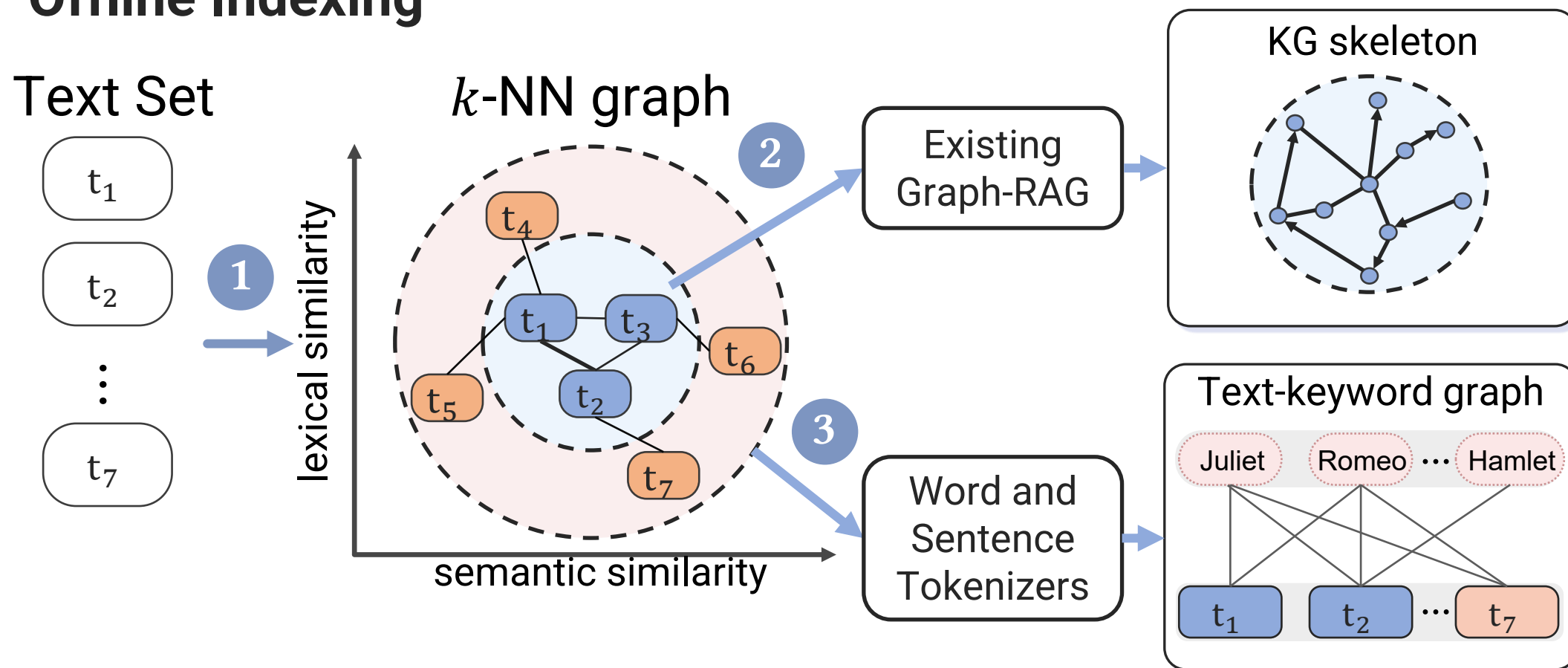
KG "skeleton"



$$\text{"skeleton" cost} = \frac{|\text{Important text set}|}{|\text{overall text set}|} \times \text{overall cost}$$

Our Method: KET-RAG

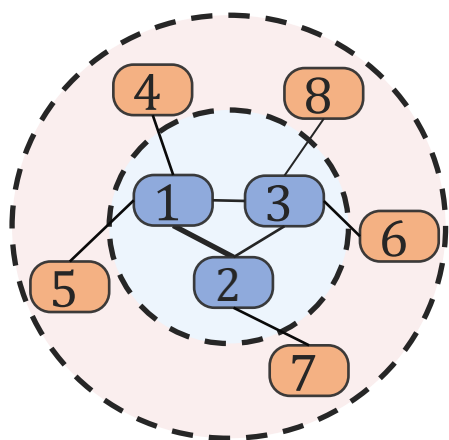
- Offline Indexing



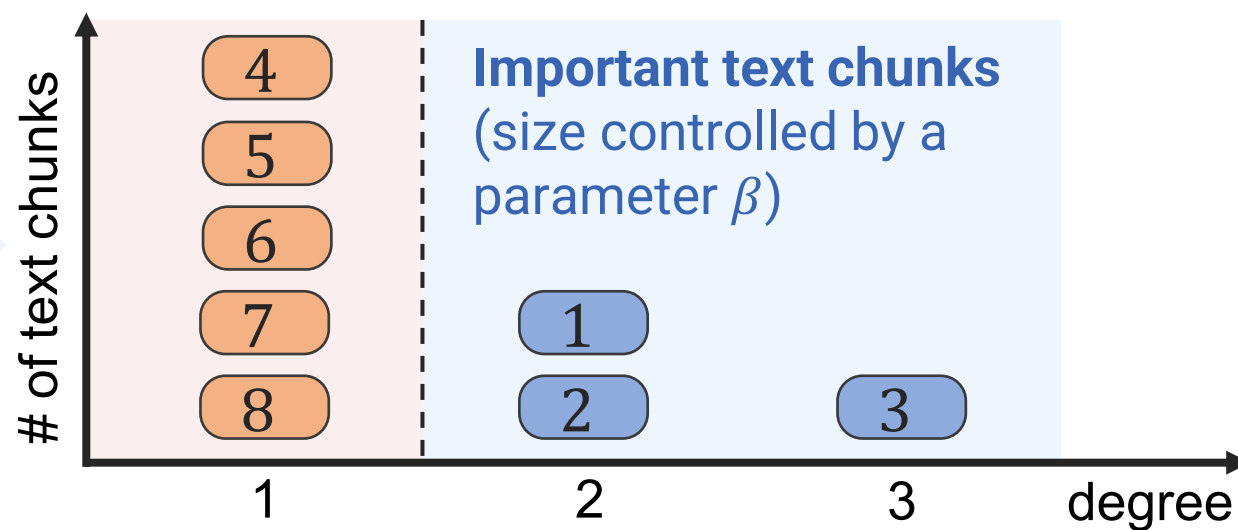
Our Method: KET-RAG

Offline Indexing – 1. k -NN graph

k -NN graph



Observation from *Fig.2* of our paper

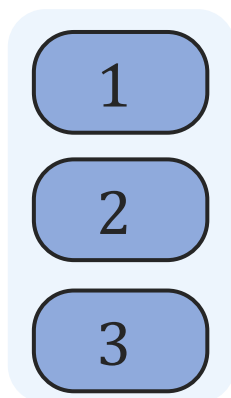


A small subset of text is **important**
(=highly relevant to others in k -NN graph)

Our Method: KET-RAG

Offline Indexing – 2. KG skeleton

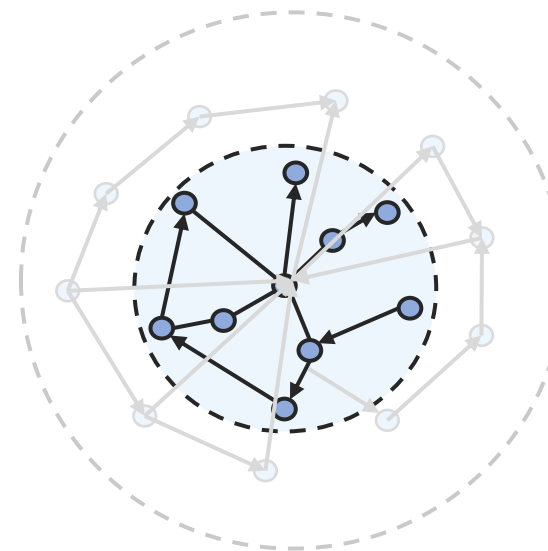
Important text set
(size= $\beta \cdot original$)



Existing
Graph-RAG
approach



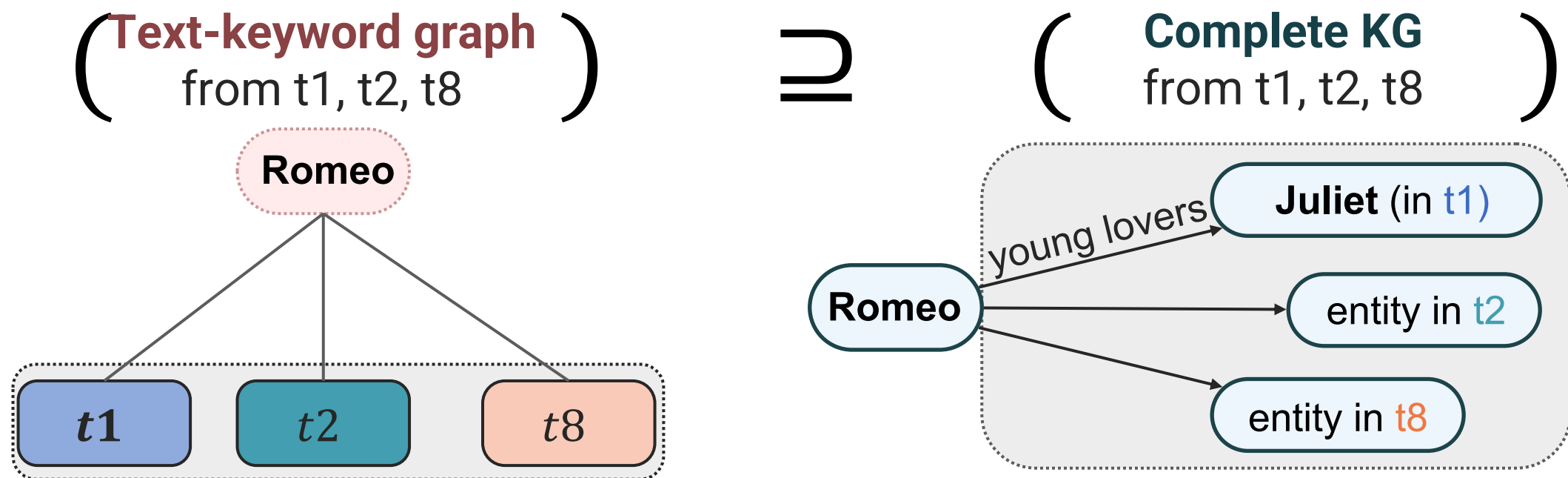
KG skeleton



$$\begin{aligned} \text{cost}_{\text{KG skeleton}} &= \beta \cdot \text{cost}_{\text{KG}} \\ \Rightarrow \text{cost}_{\text{KET-RAG}} &\approx \beta \cdot \text{cost}_{\text{Graph-RAG}} \end{aligned}$$

Our Method: KET-RAG

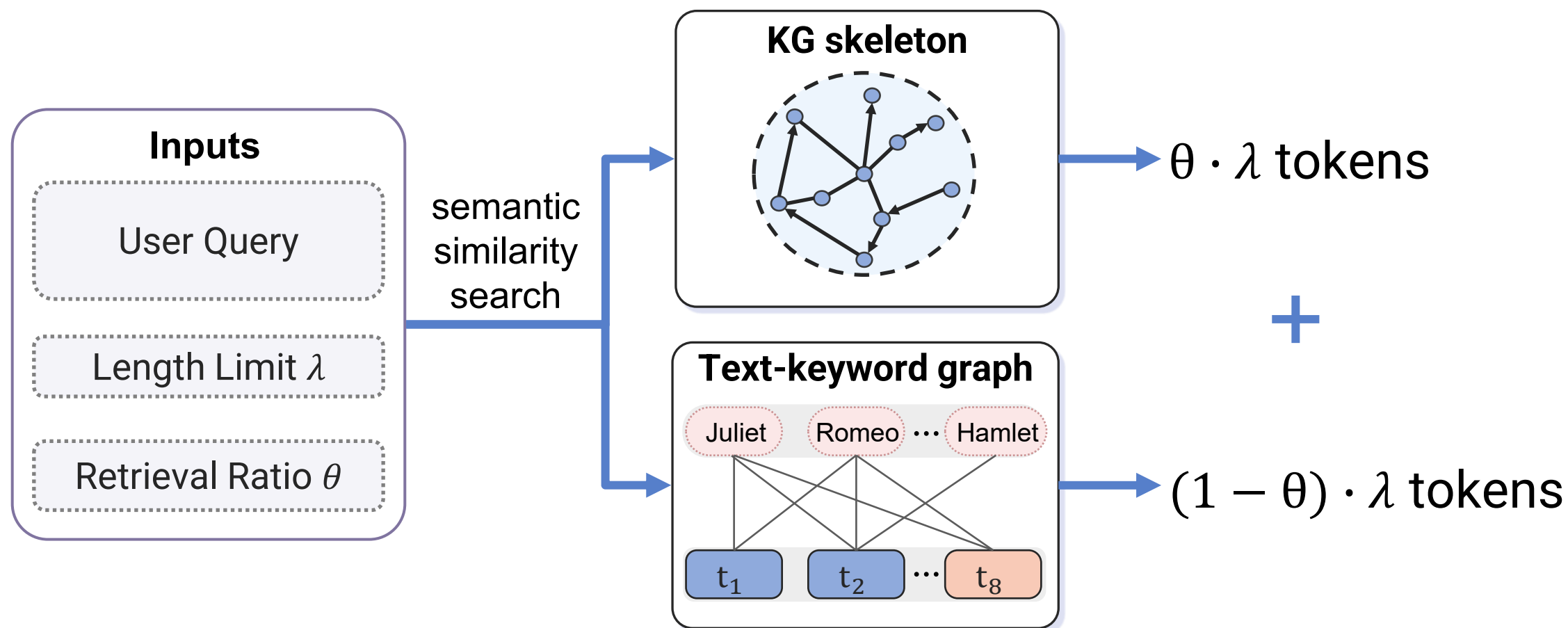
Offline Indexing – 3. Text-keyword graph



Text-keyword graph already encodes information from the **KG**

Our Method: KET-RAG

- Retrieval



Experiments

Scenario

- Multi-hop QA (datasets: MuSiQue, HotpotQA)
- open-ended QA (datasets: RAG-QA Arena)

Competitors

(1) Vanilla RAG (2) KNN-Graph-RAG (3) Microsoft's Graph-RAG
(4) Hybrid-RAG (5) HyDE (6) HippoRAG (7) LightRAG

Our proposals

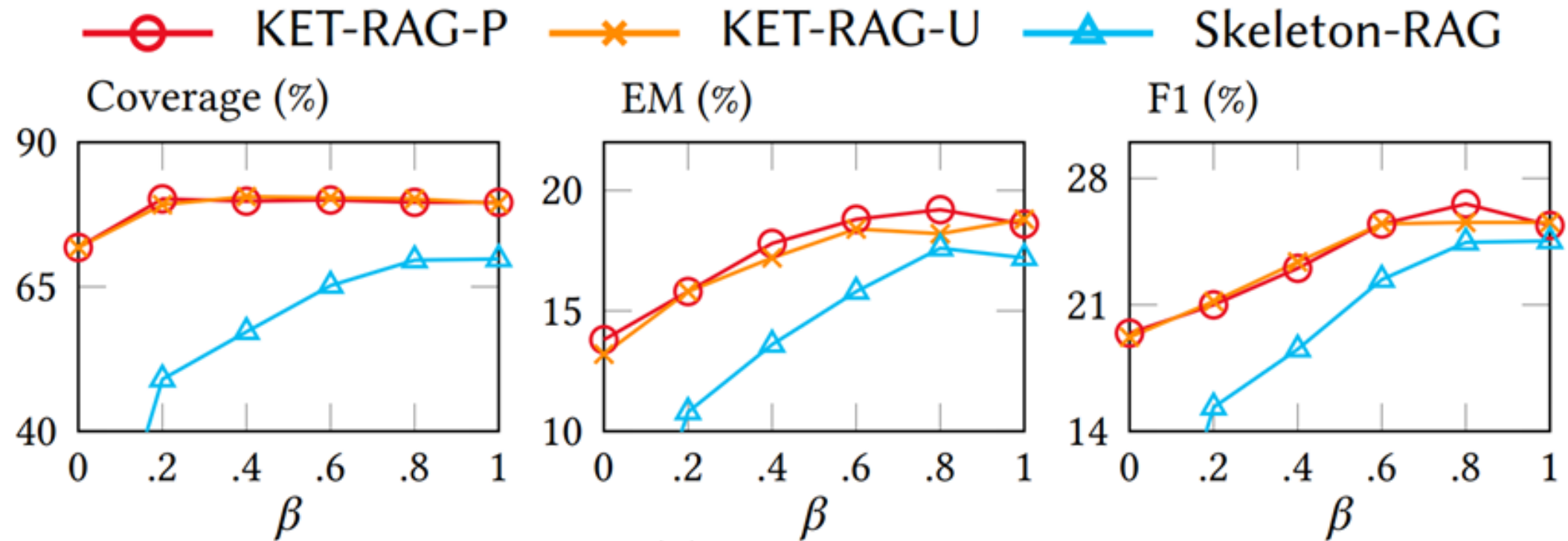
- Keyword-RAG (only the text-keyword graph), Skeleton-RAG (only the KG skeleton),
- **KET-RAG-U, KET-RAG-P** (final solutions)

Experiments: overall results

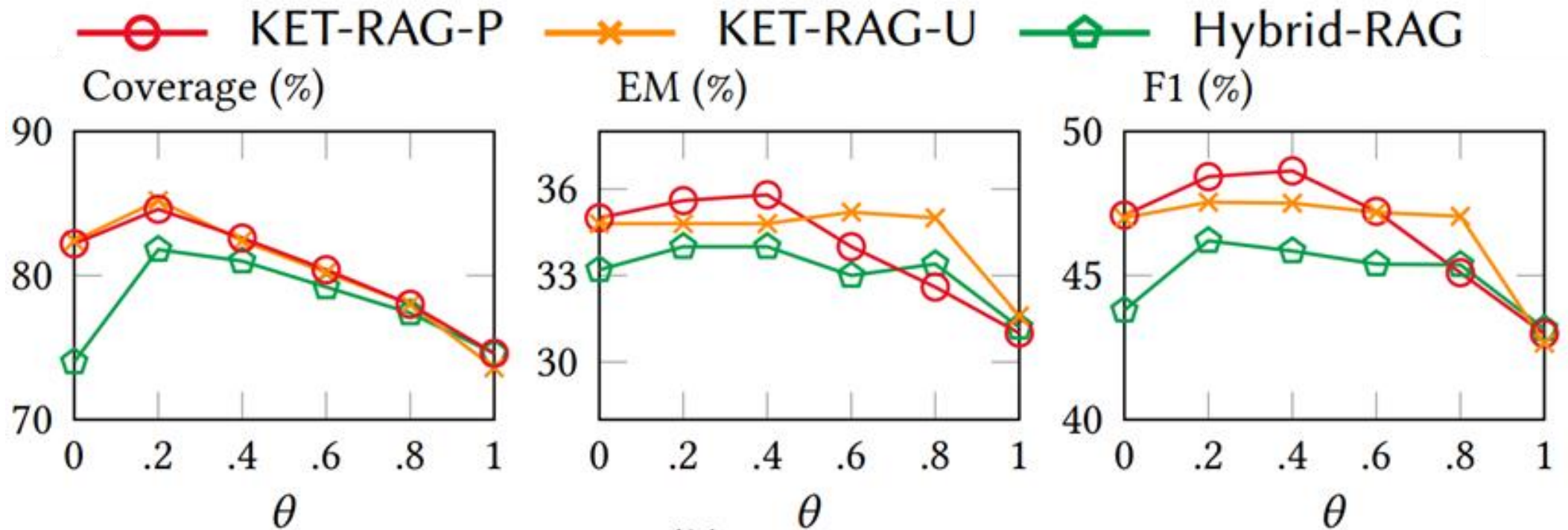
compared to Graph-RAG: superior performance, $\approx 80\%$ of the cost

Dataset	MuSiQue					HotpotQA					RAG-QA Arena	
Metric	USD	Coverage	EM	F1	BERTScore	USD	Coverage	EM	F1	BERTScore	USD	Win Rate
Text-RAG	0.02	26.4	3.0	5.4	62.9	0.01	37.2	14.8	19.4	69.9	0.03	0.0
KNNG-RAG	0.02	21.2	2.8	4.6	62.5	0.01	28.6	13.4	16.9	68.8	0.03	17.8
MS-Graph-RAG	2.30	47.6	11.4	15.8	67.4	2.30	63.0	21.6	30.2	74.2	5.05	33.4
Hybrid-RAG	2.32	49.2	10.4	15.1	<u>68.8</u>	2.31	64.8	22.6	30.5	76.8	5.08	35.0
HyDE	0.03	33.0	4.2	6.7	63.5	0.02	40.8	16.2	21.3	70.9	0.03	33.2
HippoRAG	1.49	51.6	9.2	14.2	66.8	1.40	64.0	29.2	<u>38.3</u>	77.3	1.37	<u>38.4</u>
LightRAG-Local	1.80	39.0	9.4	12.9	66.6	1.77	57.6	22.4	28.1	73.1	4.00	29.0
LightRAG-Global	1.80	37.6	6.4	9.5	64.8	1.77	49.0	19.6	24.5	71.8	4.00	23.0
LightRAG-Hybrid	1.80	45.6	10.2	14.4	67.2	1.77	61.6	24.6	30.7	74.2	4.00	30.0
Keyword-RAG	0.03	50.8	7.0	11.8	68.1	0.03	60.2	24.4	33.5	78.9	0.07	37.0
Skeleton-RAG	1.86	43.4	11.0	14.1	66.8	1.84	57.8	20.0	26.7	73.2	4.04	28.6
KET-RAG-U	1.89	<u>76.2</u>	<u>13.4</u>	<u>18.1</u>	68.1	1.87	<u>81.4</u>	28.4	38.2	<u>77.5</u>	4.08	35.0
KET-RAG-P	1.89	77.0	14.0	18.9	69.0	1.87	81.6	<u>28.6</u>	38.7	<u>77.5</u>	4.08	39.6

Experiments: varying budget (β)



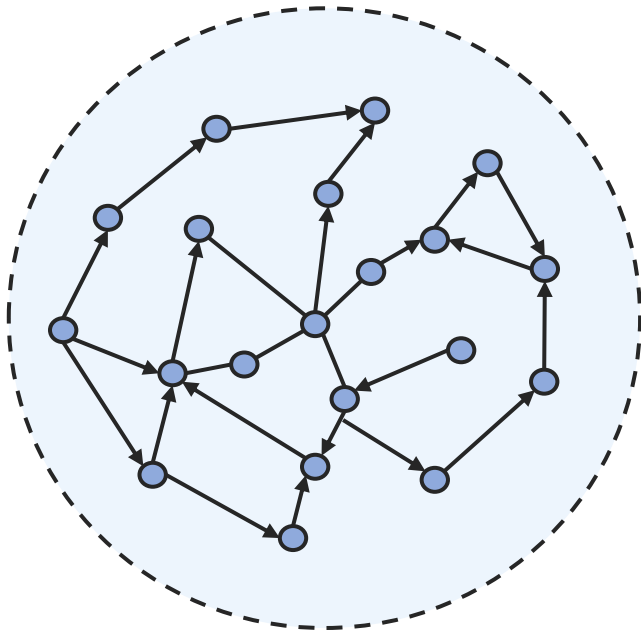
Experiments: varying retrieval ratio (θ)



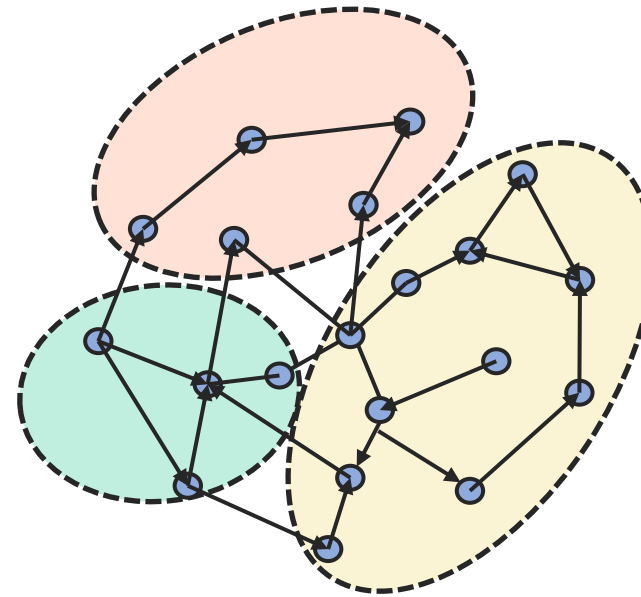
Future Work: handling global queries

Example: summarizing the main theme of the documents

Complete KG



Community summaries



THANK YOU!

Code: *github.com/waetr/KET-RAG*

