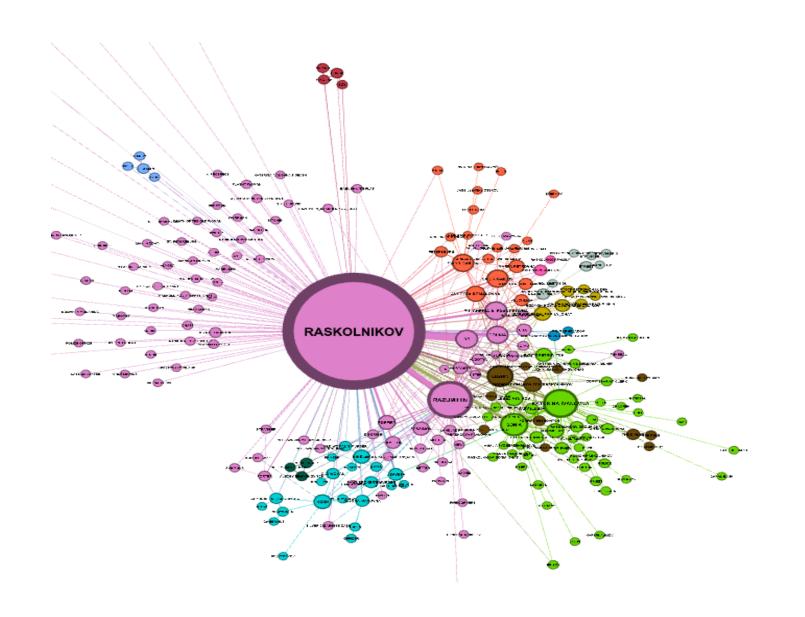
DF-Graph

Structured and explainable analysis of communication data for digital forensics



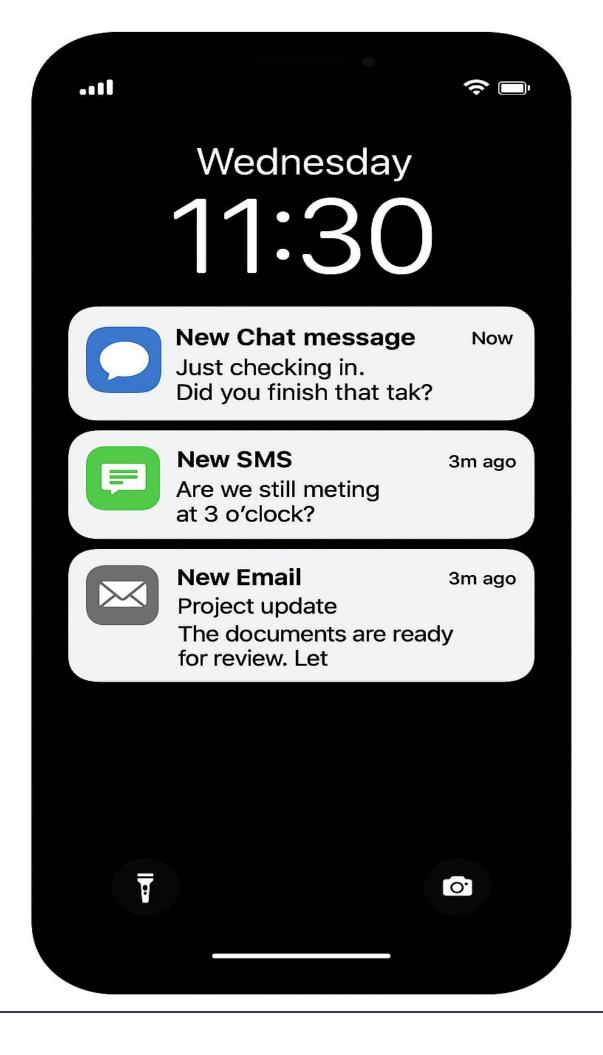




Why communication data matters?

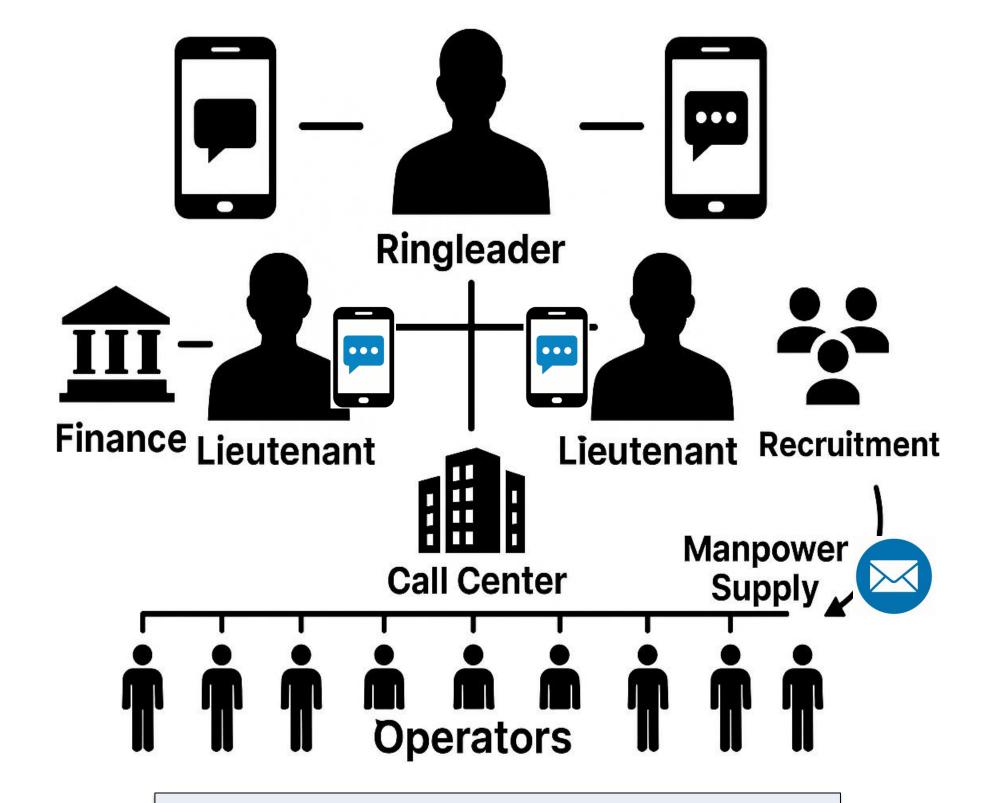
Communication data: Chat messages, SMS logs, and emails

- Reveals motives, intent, relationships
- Rapidly increasing data volume
- Used as direct or indirect evidence



Communication data as evidence

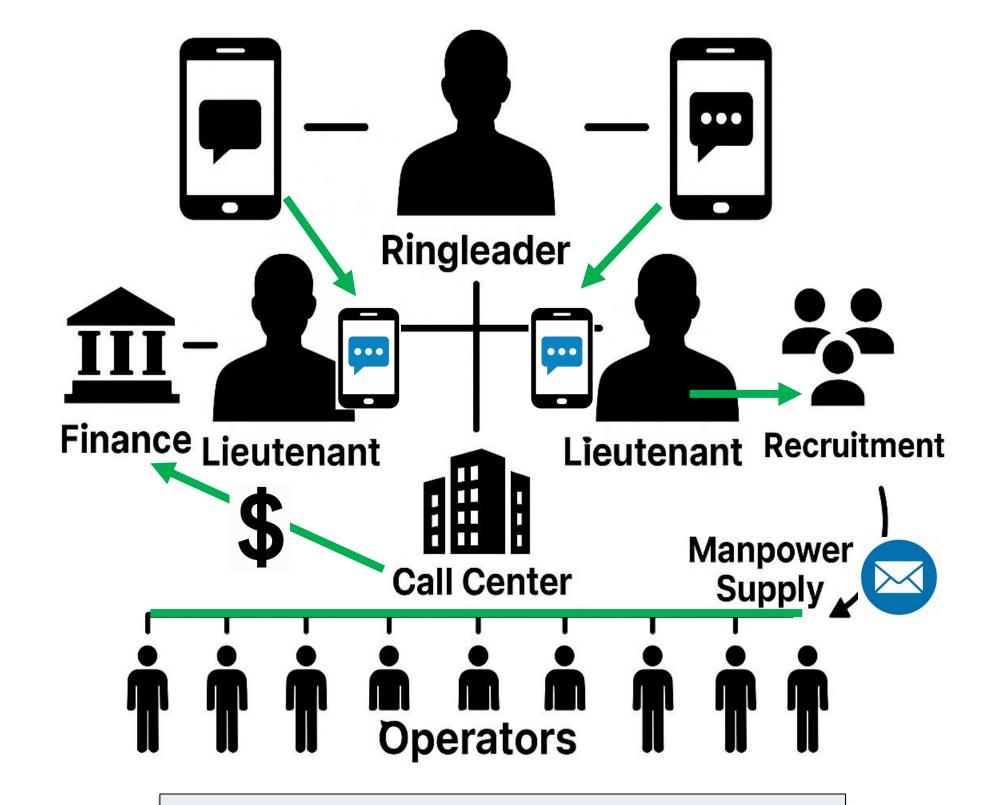
- In voice phishing case, text messages reveal organizational hierarchy
- In serial murders case, victim messages expose criminal motives
- In fraud case, emails show intentional deception



Criminal communication network (e.g., a Voice phishing case)

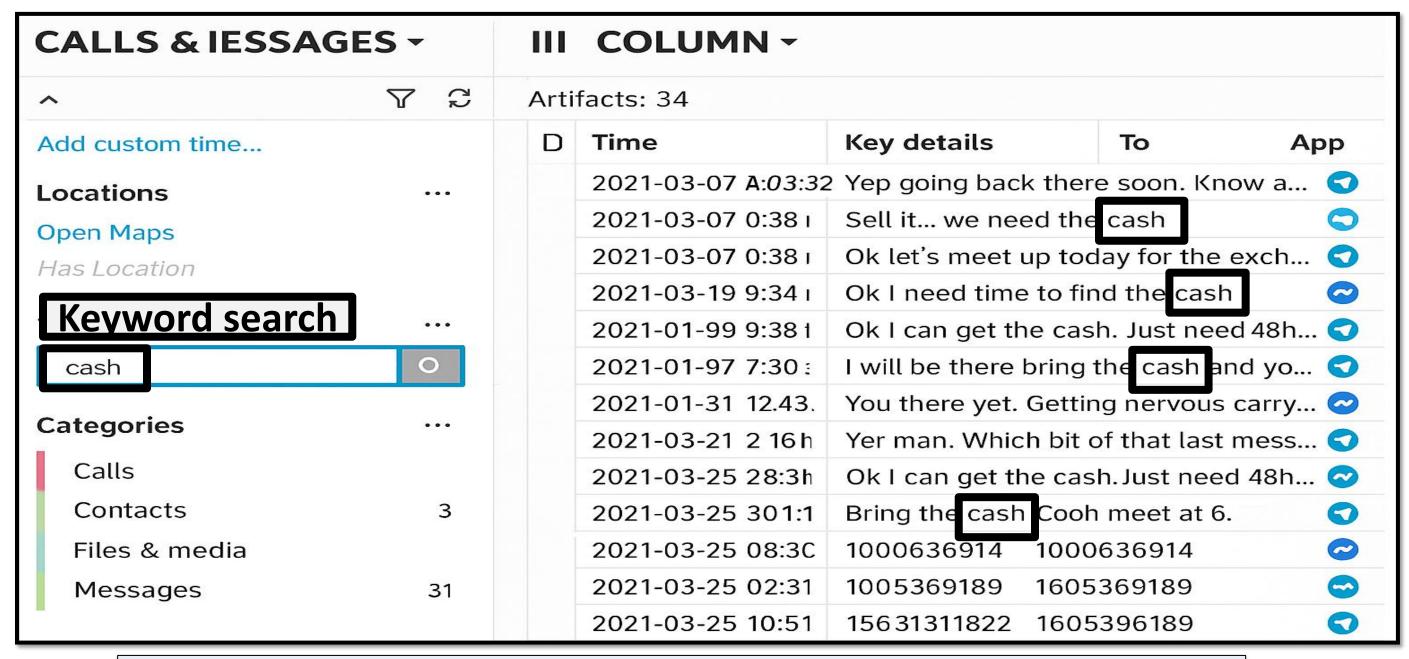
Communication data as evidence

- In voice phishing case, text messages reveal organizational hierarchy
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Criminal communication network (e.g., a Voice phishing case)

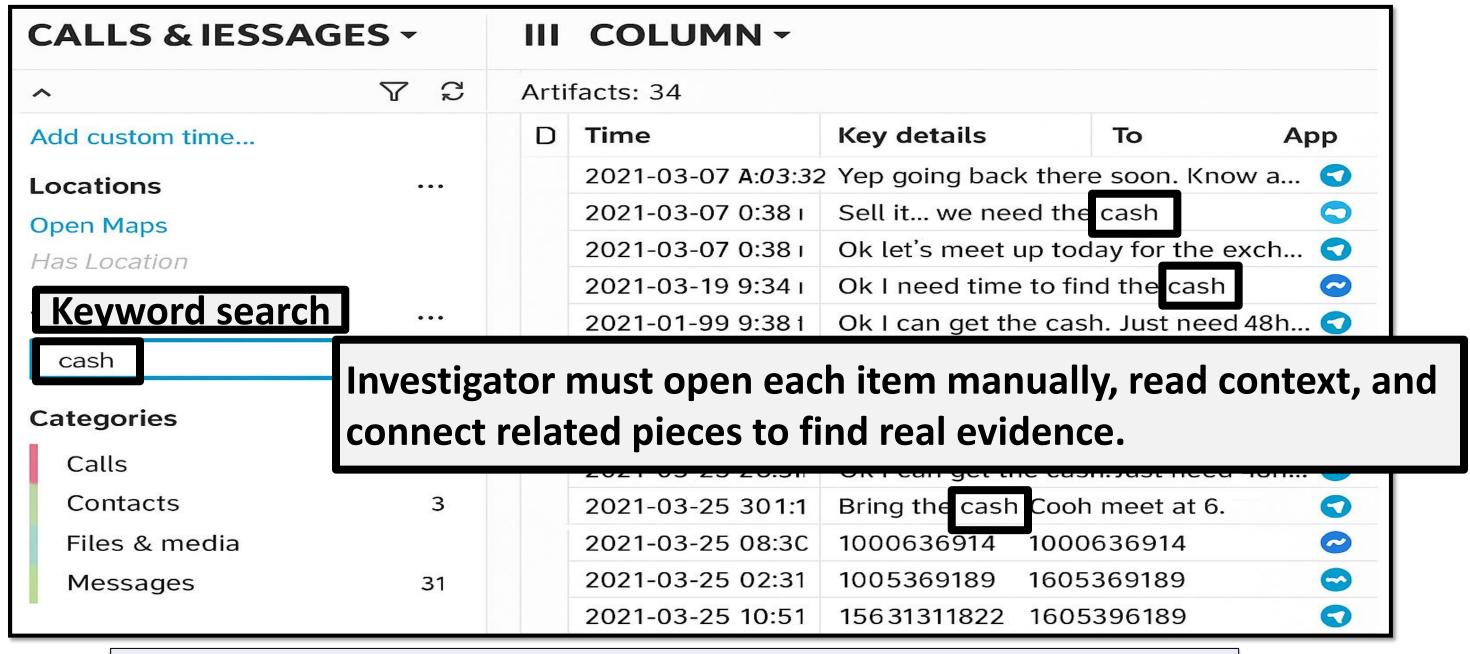
Practical difficulties - Manual investigation



Existing Tool: Keyword search (e.g., "cash") → Result list



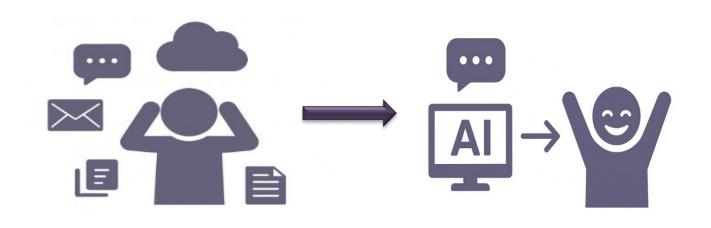
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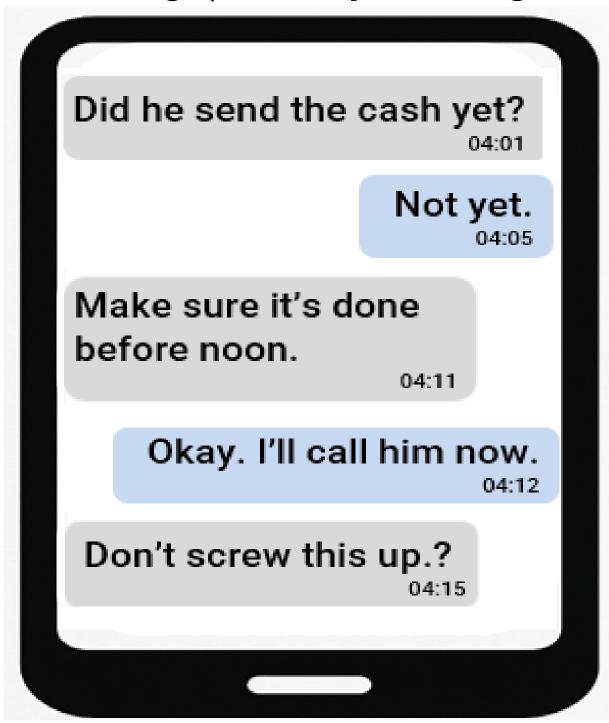


How can LLMs reduce the manual burden of investigators?

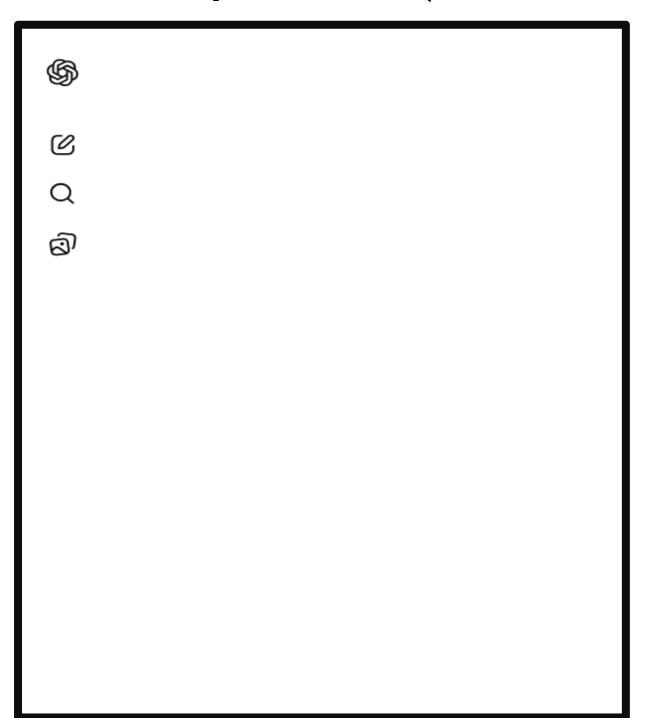


Challenges of applying LLMs in forensics

Real chat log (Voice phishing case)

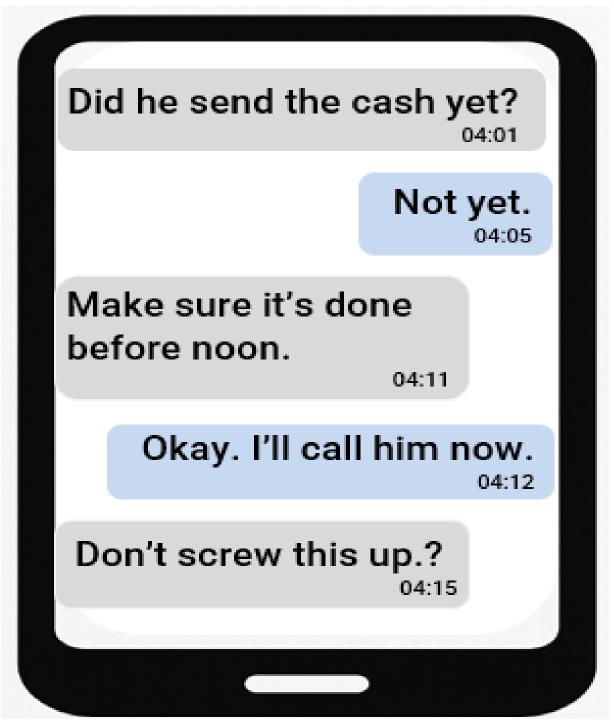


LLM's interpretation (Al's answer)

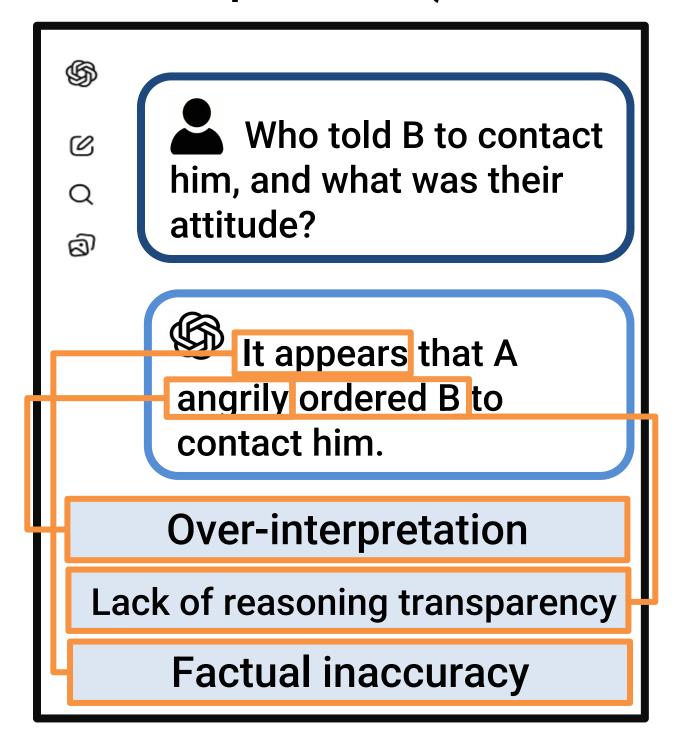


Challenges of applying LLMs in forensics

Real chat log (Voice phishing case)



LLM's interpretation(Al's answer)



Can Al provide source-verifiable and trustworthy digital evidence analysis?







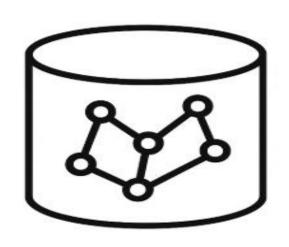
Design principles for DF-Graph

Explicit reasoning path

Scalable reasoning over communication graphs

Evidence-grounded answer generation







Explainability

Structure

Evidentiary admissibility

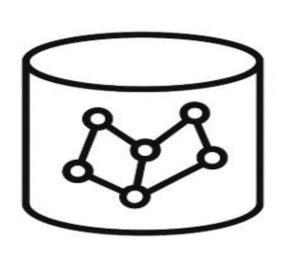
Design principles for DF-Graph

Explicit reasoning path

Scalable reasoning over communication graphs

Evidence-grounded answer generation







DF-Graph: Structured and explainable analysis of communication data for digital forensics

System design

System design

: A structured and explainable solution

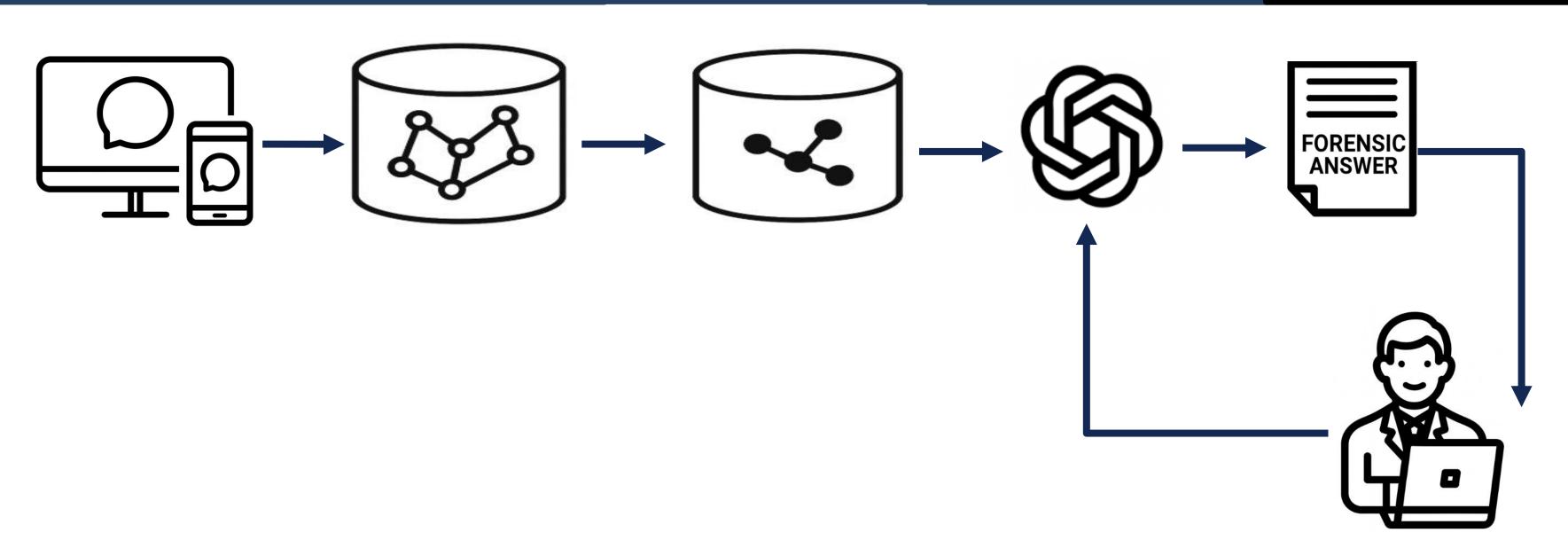
1.Data Acquisition & Preprocessing

2.Graph construction

3.Subgraph retrieval

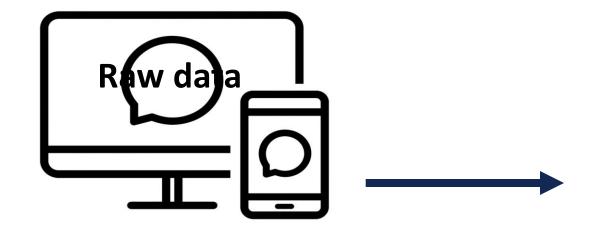
4.Evidence-guided answer generation

5.Explainable reasoning traces



Data Acquisition and Preprocessing

(1) Data Acquisition



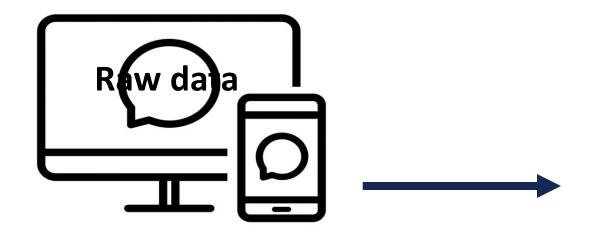
(2) Data preprocessing: Parsing & Normalization

Sender	Receiver	Timestamp	Message Content
Raskolnikov	Sonia	2025-11-13 09:01:42	I couldn't sleep again last night.
Sonia	Raskolnikov	2025-11-13 09:02:15	You should come by the chapel today.
Raskolnikov	Psychologist	2025-11-13 09:03:28	I keep seeing her face — the old woman
Psychologist	Raskolnikov	2025-11-13 09:04:10	That's your conscience speaking, not her.
Raskolnikov	Psychologist	2025-11-13 09:05:22	The sound of the axe sometimes just the word
Porfiry	Raskolnikov	2025-11-13 09:06:33	Care to finish our conversation, Mr. Raskolnikov?



Data Acquisition and Preprocessing

(1) Data Acquisition



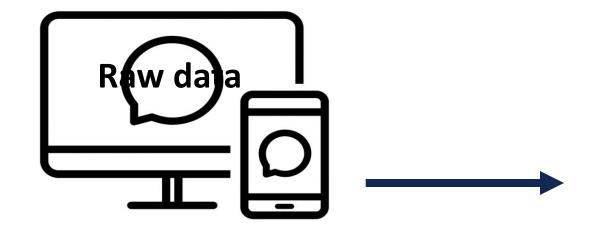
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Data Acquisition and Preprocessing

(1) Data Acquisition



(2) Data preprocessing: Anonymization

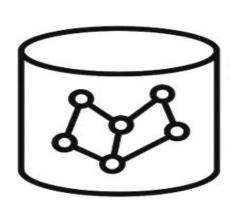
Sender	Receiver	Timestamp	Message Content
P_001	P_002	2025-11-13 09:01:42	I couldn't sleep again last night.
P_002	P_001	2025-11-13 09:02:15	You should come by the chapel today.
P_001	P_003	2025-11-13 09:03:28	I keep seeing her face — the old woman
P_003	P_001	2025-11-13 09:04:10	That's your conscience speaking, not her.
P_001	P_003	2025-11-13 09:05:22	The sound of the axe sometimes just the word
P_004	P_001	2025-11-13 09:06:33	Care to finish our conversation, Mr. P_001 ?



1.Data Acquisition

& Preprocessing

Graph Construction



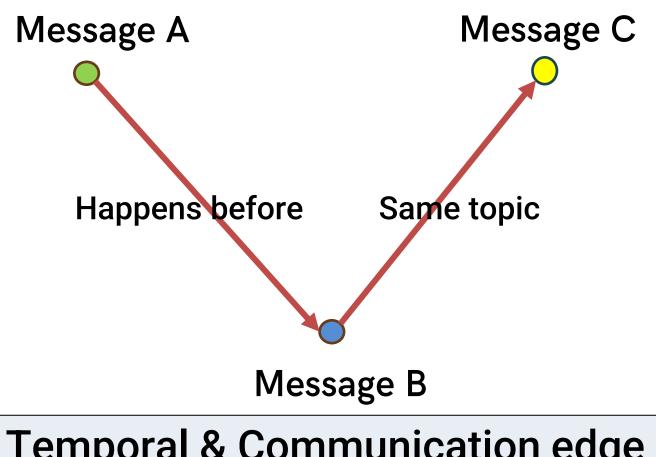
$$G = (V, E)$$

Node generation

- Messages
- **Utterances**

Edge generation

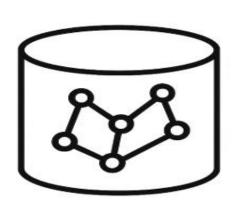
- Temporal edge: chronological order
- Communication edge: conversational continuity
- Semantic edge



Temporal & Communication edge



Graph Construction



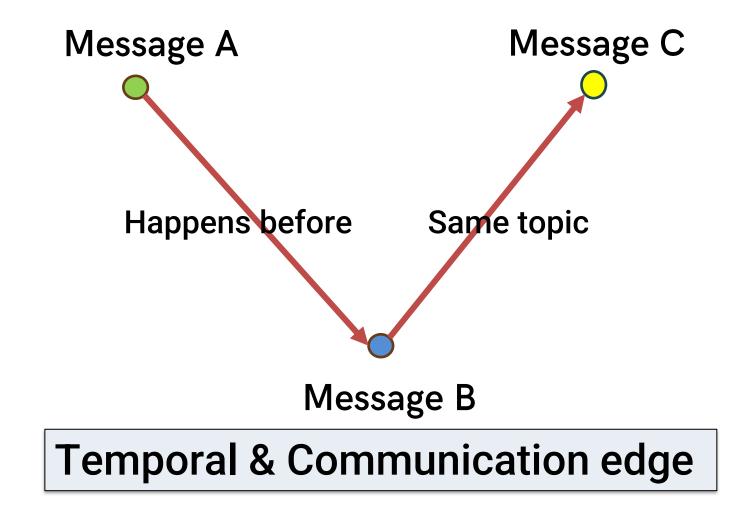
$$G = (V, E)$$

Node generation

- Messages
- **Utterances**

Edge generation

- Temporal edge: chronological order
- Communication edge: conversational continuity



Semantic edge → CAUSES, SUPPORTS, MENTIONS, and CONTRADICTS

Graph Construction

Semantic edge: Meaning-based Connection

"She was coming back unexpectedly."

"This compelled him to act quickly."

Message B

"I never killed the old woman. There's no evidence."

"But you knelt before me and confessed everything."



Message C

Message A

2.Graph

Construction

Graph Construction

Semantic edge: Meaning-based Connection

"She was coming back unexpectedly."

"This compelled him to act quickly."

Message A

Message B

"I never killed the old woman. There's no evidence."

"But you knelt before me and confessed everything."

Message C

Message D

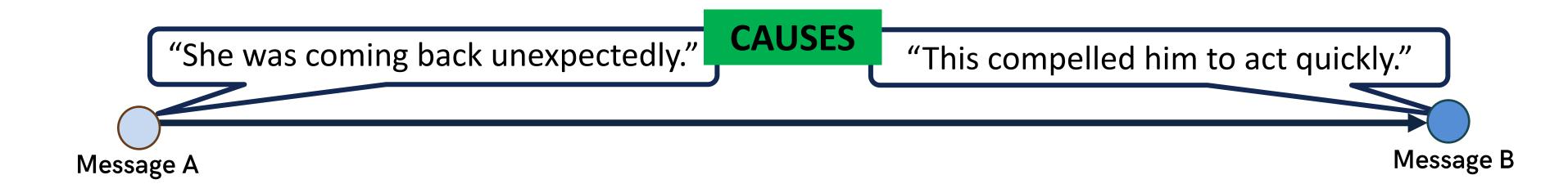


2.Graph

Construction

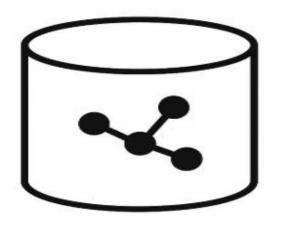
Graph Construction

Semantic edge: Meaning-based Connection

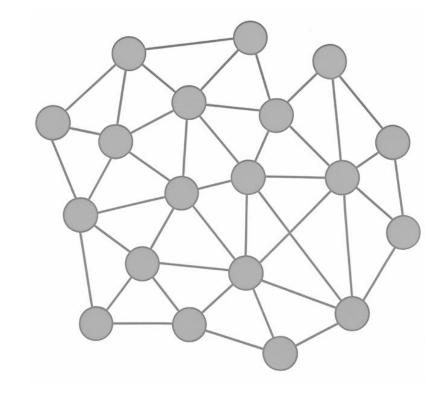




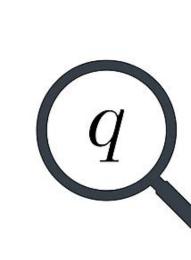




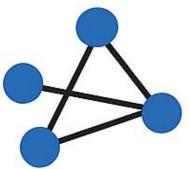
Gq = (Vq, Eq)



Full communication graph **G**



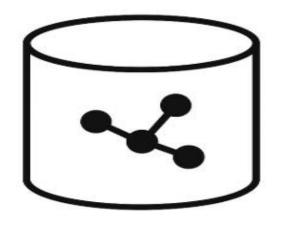




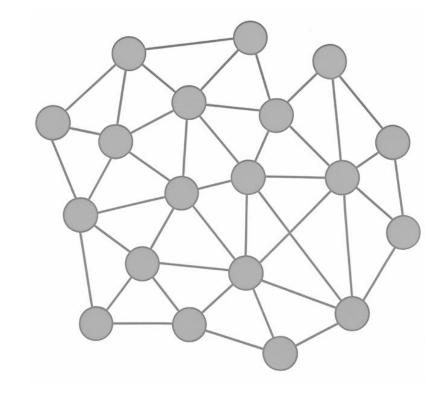
q: specific forensic query

Retrieved subgraph Gq = (Vq, Eq)

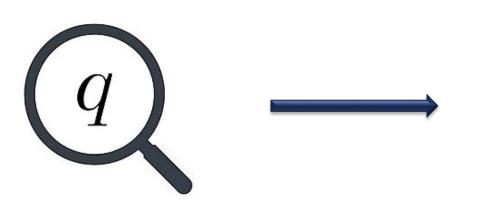




Gq = (Vq, Eq)



Full communication graph **G**



q: specific forensic query

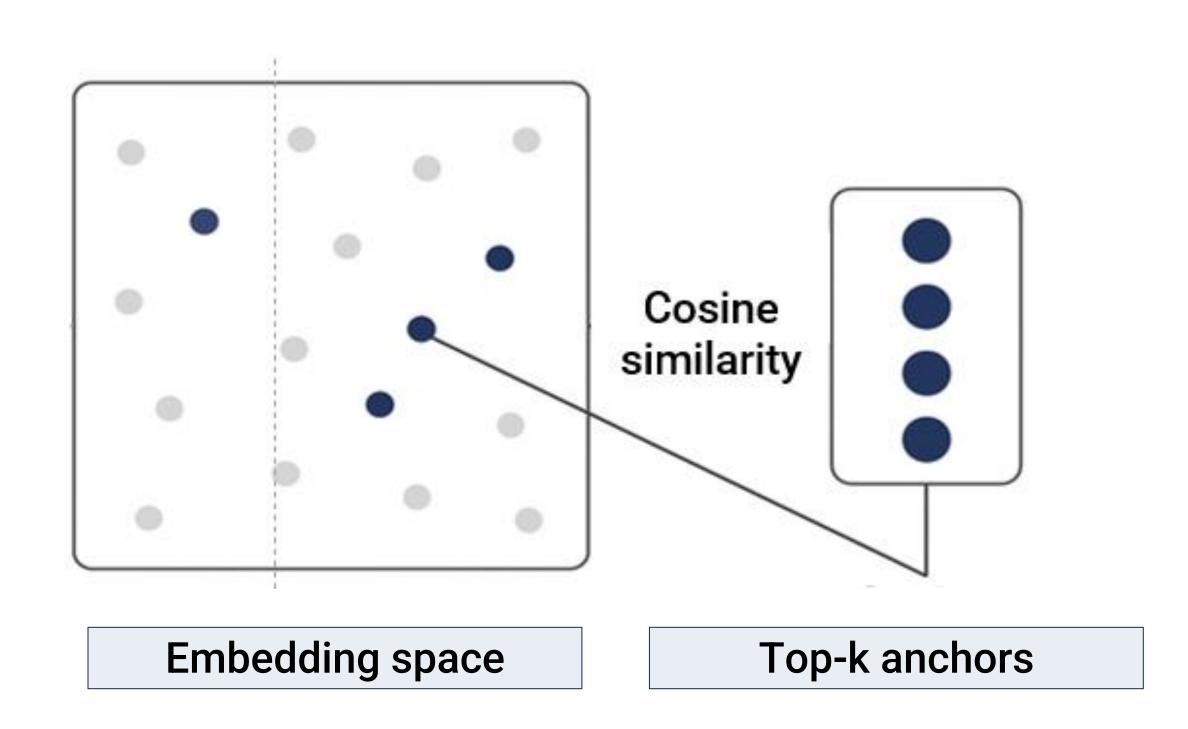
Retrieved subgraph Gq = (Vq, Eq)



Step1. Semantic filtering



Forensic query
Sentence Transformers

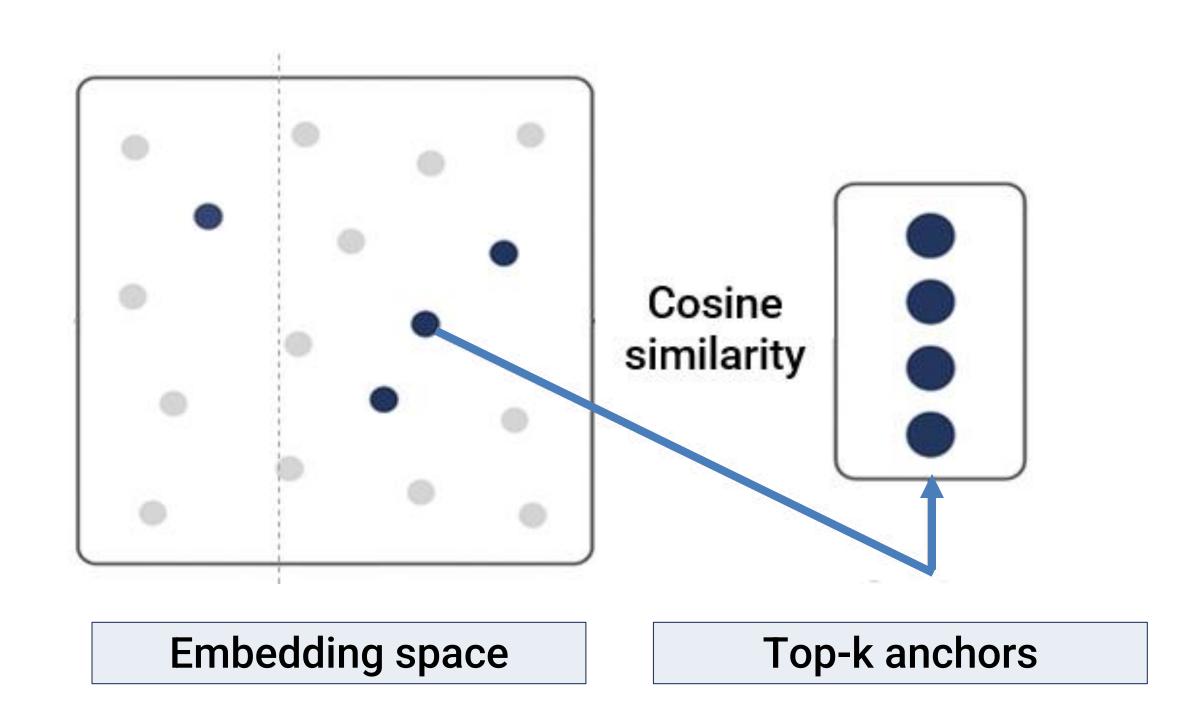




Step1. Semantic filtering

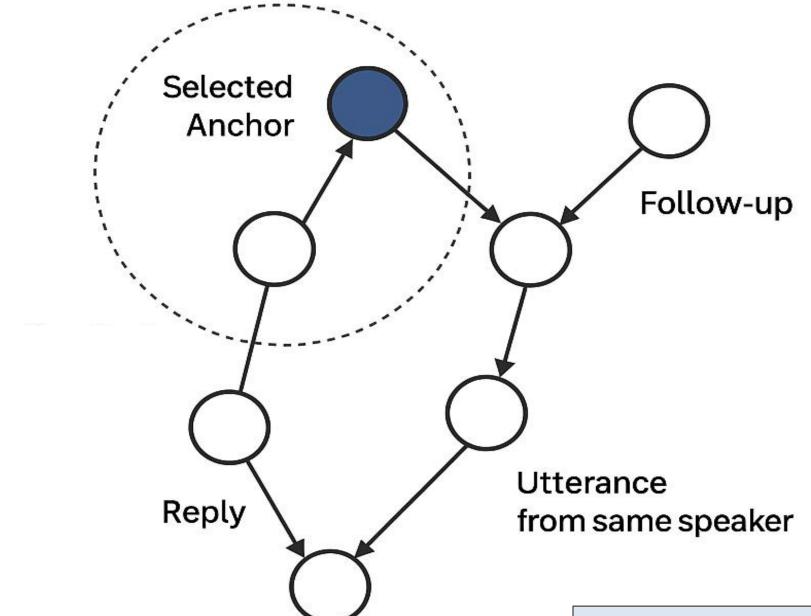


Forensic query
Sentence Transformers





Step2. Graph expansion & Citation preservation



Subgraph generated via citation-guided expansion

Evidence-Guided Answer Generation



- Gq → Relevant messages
- Cq → Chronologically ordered evidence context
- Combine Cq with forensic question $q \rightarrow$ Input prompt

Forensic Answer Prompt Rules

Instruction

- 1. The model must answer only based on the given evidence, without speculation.
- 2. It must cite each piece of evidence explicitly using message IDs or document references.
- 3. The response must be concise, interpretable, and legally admissible.

•

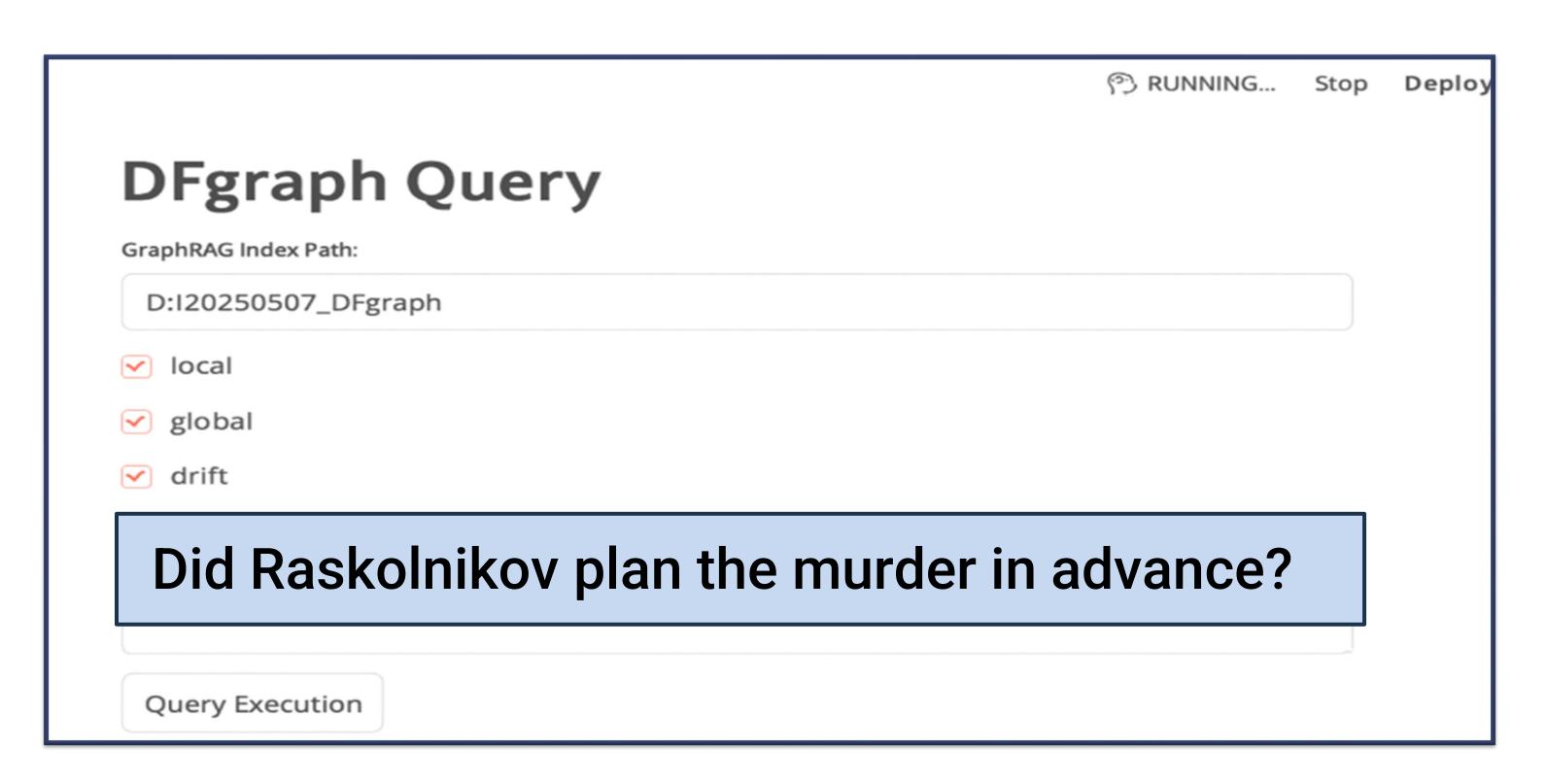


2.Graph

Construction

Evidence-Guided Answer Generation







Explainable Reasoning Traces





DFgraph Query

Answer:

Yes.

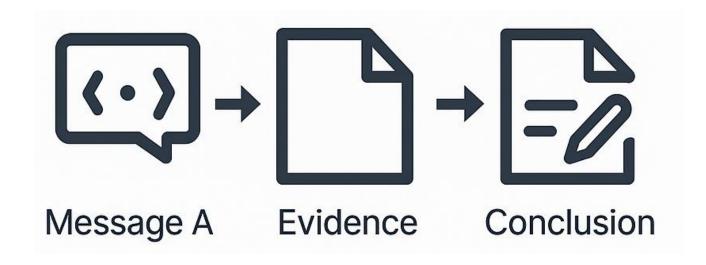
Msg#A shows that Raskolnikov intentionally chose a time when Alyona would be alone, while Msg#B confirms that he had concealed a weapon in advance. Msg#C further indicates that he was aware the act had already been carried out. Taken together, these messages provide verifiable evidence of premeditation and preparation, thereby supporting a legally defensible conclusion. [Source Information: Reports (7, 21); Entities (3157, 2726, 2962); Relationships (1106, 1218,999); Sources (Msg#A, Msg#B, +more)]



Explainable Reasoning Traces

Reasoning trace

 Tracing logical links between evidence and conclusion



How do we generate this trace?

- Retrieve relevant subgraph
- Analyze logical and temporal links
- Construct an evidence-aligned reasoning trace

This is why DF Graph adopts a graph-based structure



Evaluation

Evaluation: Dataset overview

Three complementary datasets used for model evaluation





- ≈2.95MB
- 18,653 messages
- Forensic Investigation Reports



Public Dataset

- ≈0.91MB
- 6,667 messages



Synthetic Dataset

- 345 KB
- 2,367 messages
- Dostoevsky's "Crime and Punishment"



Evaluation: Comparison of models

Comparison with Baseline Models



GPT

No retrieval, no structure



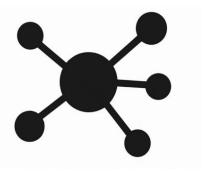
GPT+BERT

Intent-aware, but structure-free



Naïve RAG

Retrieval-based, but structurally flat

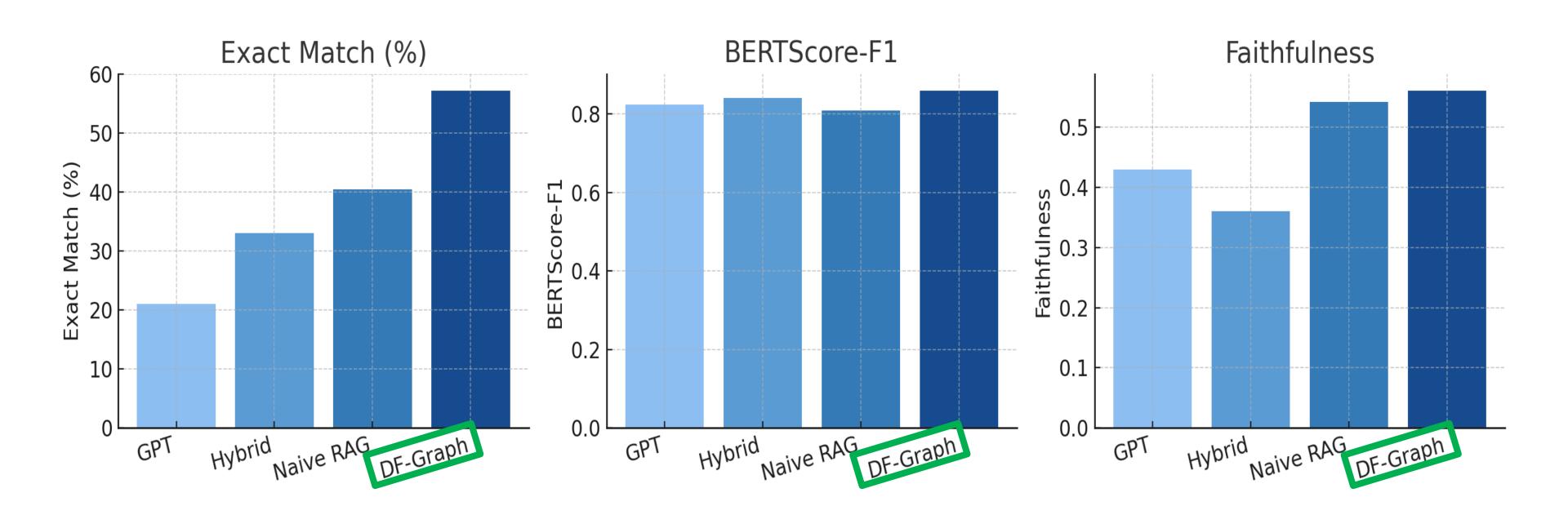


DF-Graph

Graph-structured, contextually grounded

Evaluation1: Quantitative results

Evaluation Metrics & Key Findings





Evaluation2: User study

• Participants: 8 digital forensic professionals

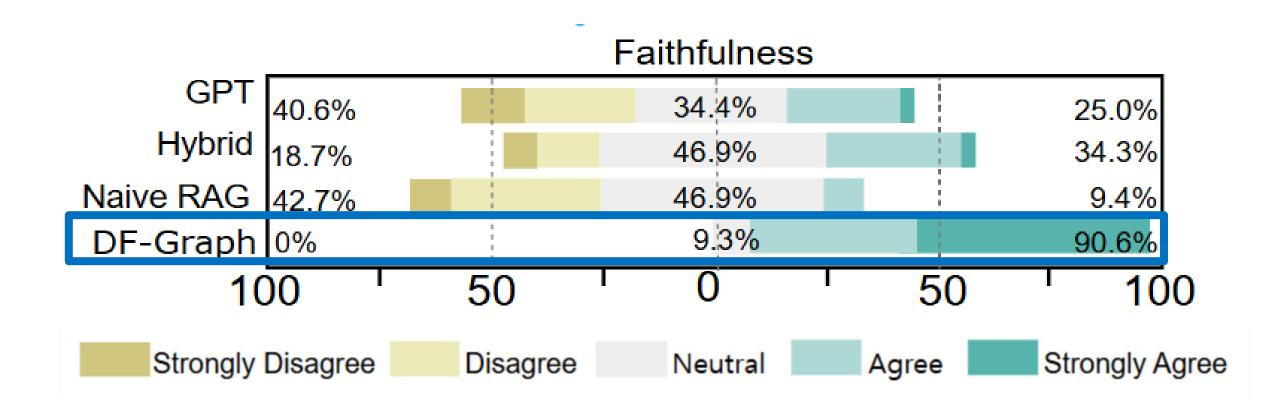
ID	Experience(yrs)	Frequency	ID	Experience(yrs)	Frequency
P1	9	daily	P5	8	daily
P2	16	weakly	P6	13	weakly
Р3	5	daily	P7	7	daily
P4	5	daily	P8	7	daily

• Study Design and Evaluation Protocol: Latin square method

Evaluation2: User study

- Accuracy(Q1): DF-Graph (78.1%), Naive RAG (56.2%), Hybrid (43.8%), GPT (40.6%)
- 5-point Likert scales for each task
 - Faithfulness (Q2): Grounded in message content?

: DF-Graph (90.6%)



Participants' feedback

Practical implementation potential



"DF-Graph helps resolve major concerns in AI-based digital forensics, particularly source traceability. I look forward to seeing it used in practice soon."

Potential for legal admissibility

"DF-Graph works just like real forensic workflows, and its structure helps me understand the logic instantly."



Conclusion

"DF-Graph uses graph structures to help LLMs capture context and generate traceable, reliable forensic results."

Quantitative results

Metric	DF-Graph Score
Exact Match (%)	57.23 ± 1.95
BERTScore-F1	0.859 ± 0.005
Faithfulness	0.561 ± 0.005

Qualitative evaluation

Evaluation Metrics	DF-Graph
Faithfulness (Q2)	9.3%
Explainability (Q3)	12.6%
Clarity (Q4)	15.6%
Interpretability (Q5)	9.4%



Thank You

If you're interested in our work, please check out our paper for more details



The paper can be accessed via the QR code below

munich1984@skku.edu

