

#### Transient Tables: Evaluating LLMs' Reasoning on Temporally Evolving Semi-structured Tables

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\* Equal Contribution



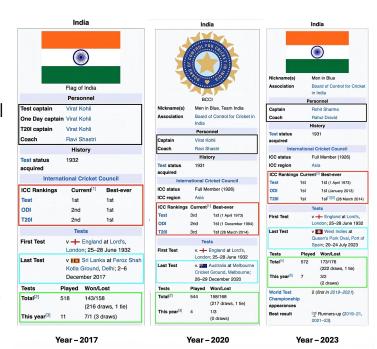






#### **Motivation**

- Information is inherently transient and constantly updated
  - Examples: company profits, political figures, sports rankings, etc.
- LLMs are typically trained on static datasets
- Research question: Can LLMs effectively reason over temporal changes in information through in-context learning?



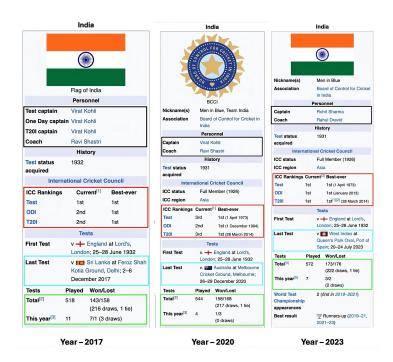
Sample question - How many Test matches did the Indian Cricket Team play between 2020 and 2023?

## Why is it Challenging?

- Temporal Context Understanding
  - Models must ground questions in the right time period.
  - "Who was captain before Rohit Sharma?" requires identifying when Rohit became captain first
- Multi-Table Integration
  - Reasoning across 8-12 tables per entity timeline.
  - Tracking changes over time while maintaining consistency.

## Why is it Challenging?

- Subtle Changes Detection
  - Key values change incrementally (e.g., Tests played: 518 → 544 → 572)
  - Easy to confuse values from different time periods.
- Token Length Limitations
  - Full entity timelines push context window limits.
  - Requires efficient attention to relevant information.

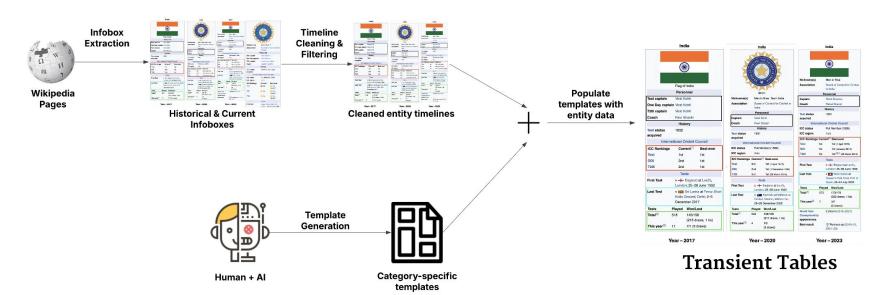


#### **Key Contributions**

- Transient Tables Dataset
  - A novel QA dataset with 3,971 questions from 14,000+ tables spanning
    1,238 entities with 11 tables per entity (on avg)
  - **Template-based question** generation pipeline using LLMs'
- Baseline results with state-of-the-art models
  - GPT-40, Llama3-70B, Gemini 1.5, GPT-40-mini, Llama3-8B, Mixtral
- Novel modeling strategies using task decomposition to enhance performance

#### **Dataset Creation**

- Entity timeline selection from Wikipedia infoboxes
- Timeline cleaning and filtering (8-12 tables per entity)
- Query-answer generation through templates



#### **Question Categorization**

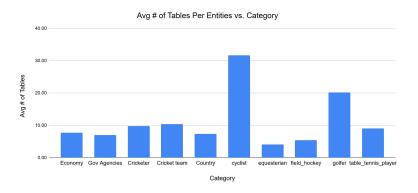
- Time information:
  - 2,985 implicit vs. 986 explicit questions

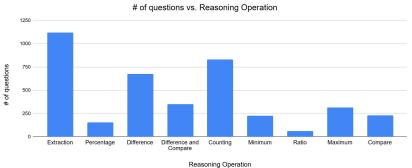
#### **Example**

- 'Who was the captain when India won the World Cup?' Implicit
- 'Who was the captain in 2020?' Explicit

## **Question Categorization**

- Reasoning types:
  - extraction, counting, comparison, etc.
- Complexity:
  - 2,113 single key questions
  - 1,858 multiple key questions



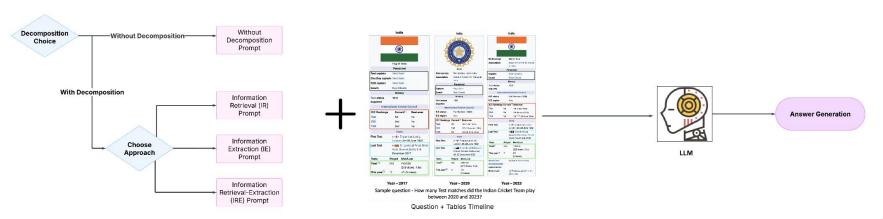


## Modeling Techniques

- Information granularity variations:
  - Closed book
    - No context
  - Single table
    - Random or latest table only
  - Full timeline
    - All tables for the entity
  - Oracle timeline
    - Only the most relevant tables

# Modeling Techniques

- Task decomposition approaches:
  - Without decomposition
  - Information retrieval
  - Information extraction
  - Information retrieval-extraction



**Question**: "Who was the coach of the Indian Cricket Team when Virat Kohli was captain and the team had its highest ICC Test ranking?"

**Approach 1**: Without Decomposition

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- Must identify Kohli's captaincy period, coaches during this time, and ICC rankings

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**Approach 1**: Without Decomposition

- Process full timeline of tables simultaneously
- Must identify Kohli's captaincy period, coaches during this time, and ICC rankings
- Perform temporal correlation across multiple attributes

**Question**: "Who was the coach of the Indian Cricket Team when Virat Kohli was captain and the team had its highest ICC Test ranking?"

**Approach 2:** With Decomposition

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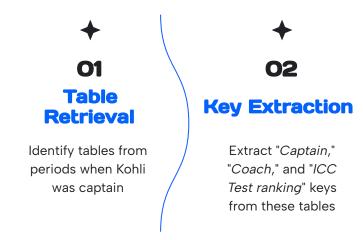
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**Table Retrieval** 

Identify tables from periods when Kohli was captain

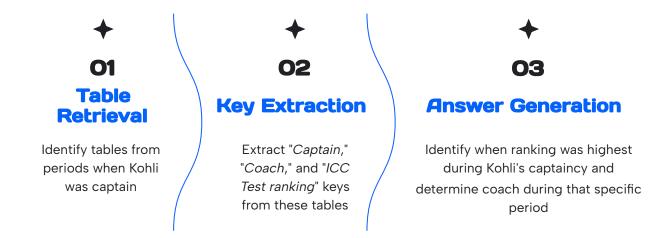
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#### Task Decomposition : Levels

#### Information Retrieval (IR)

**Stage 1**: "Table Retrieval" (identify relevant tables from timeline)

**Stage 2**: "Answer Generation" (reason over retrieved tables)

# Information (IE)

**Stage 1:** "Key Extraction" (extract relevant attributes from tables)

**Stage 2:** "Answer Generation" (reason over extracted keys)

#### Information Retrieval Extraction (IRE)

**Stage 1:** "Table Retrieval" (identify relevant tables)

**Stage 2:** "Key Extraction" (extract relevant attributes from retrieved tables)

**Stage 3:** "Answer Generation" (reason over extracted keys)

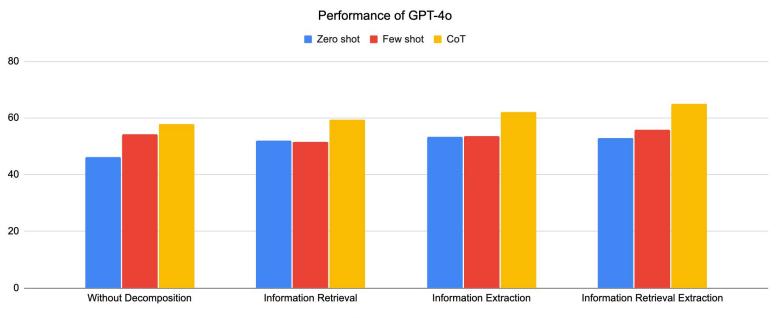
#### **Experimental Setup**

- Models evaluated
  - GPT-40, Gemini-1.5-flash, GPT-40-mini (proprietary)
  - Llama3-70B, Llama3-8B, Mixtral-7x8B (open source)
- Prompting Techniques
  - Zero shot, Few Shot, Chain of Thought
- Evaluation metrics
  - F1, Exact Match (EM), Rouge-1, Rouge-L
- Human evaluation baseline for comparison

## Results: Context Decomposition

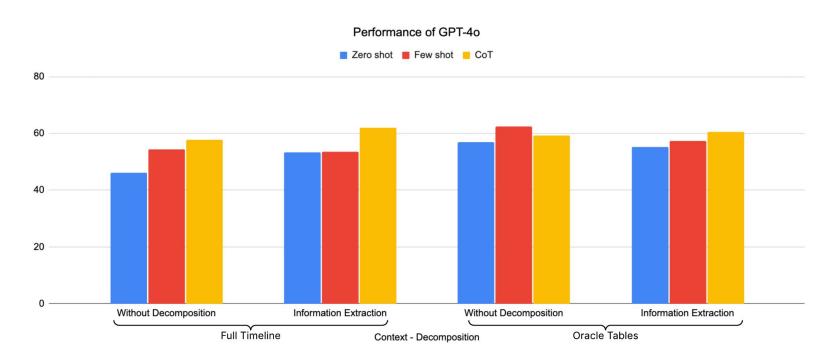


## Results: Context Decomposition



Context - Decomposition (Full Timeline)

## Results: Context Decomposition

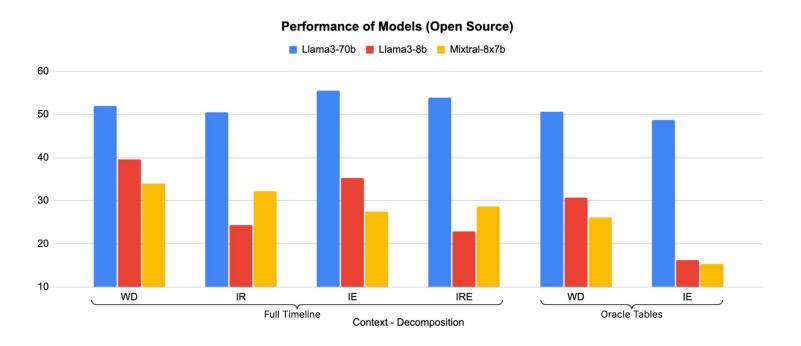


#### Results: Close Source Models



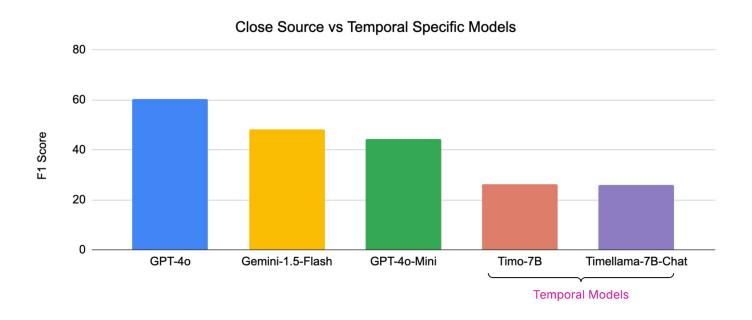
Results in different in-context variations and different intermediate task decompositions with various prompting methods.

## Results: Open Source Models



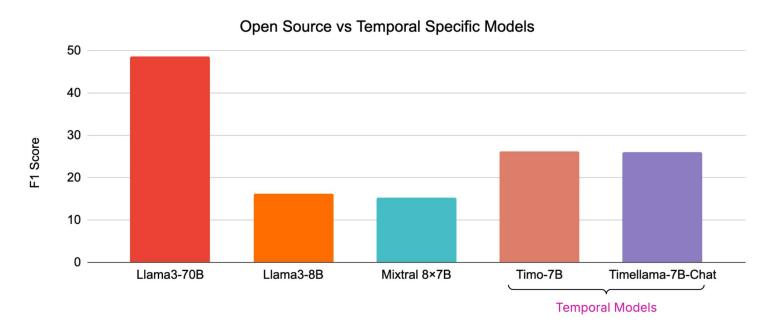
Results in different in-context variations and different intermediate task decompositions with various prompting methods.

#### Results: Temporal Models



CoT prompting with oracle tables and Key Extraction for task decomposition.

#### Results: Temporal Models



CoT prompting with oracle tables and Key Extraction for task decomposition.

#### Conclusion & Future Work

- A novel task of question answering on temporally evolving tables.
- A new **Transient Tables** dataset
  - 3,971 question-answer pairs.
  - From over 14k tables and 1,238 entities across various time periods.
- First study on LLM reasoning over entity-centric temporal tables.

#### In future:

- Currently its confined to Wikipedia infoboxes. Extending it to diverse structures beyond tables.
- Neuro-symbolic learning for better interpretability.





#### **Thanks**

\*seeking Summer Internship opportunities — let's connect! ashankarampeta@ucsd.edu