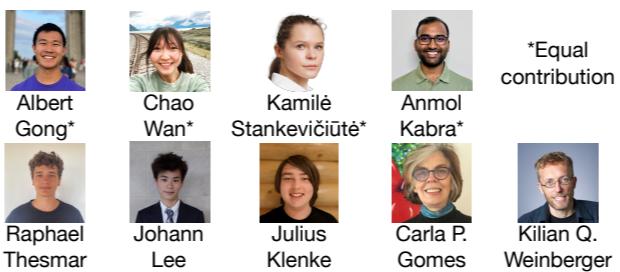


# PhantomWiki: On-Demand Datasets for Reasoning & Retrieval Evaluation



## Takeaways

PhantomWiki stress tests LLMs on multi-hop question-answering tasks, while ensuring:

1. LLMs cannot rely on factual knowledge from training
2. Factually consistent questions and answers
3. Fast and fully-automated dataset generation: 1M-sized dataset with 4-hop questions takes < 30 CPU hours!

## Motivation

LLMs must answer questions with up-to-date knowledge, but evaluation is a challenge:

- Static datasets are prone to **data leakage**
- Disentangling a model's **internal knowledge, reasoning, and retrieval** capabilities is difficult with datasets curated from public data (e.g., Wikipedia).

## Many Evaluation Scenarios

1. **In-Context:** document corpus fits in LLM context window [NIAH, LongBench, RULER,  $\infty$ -Bench, HELMET]
2. **RAG:** relevant documents are retrieved using an external retriever [MultiHop-RAG, BRIGHT, ARES]
3. **Agentic:** LLM uses external tools to obtain relevant context [ToolQA]



## Prolog + Context-Free Grammar

**Fact =**

**predicate + constants**

The mother of Alice is Charlotte.

mother("Alice", "Charlotte").

**Query =**

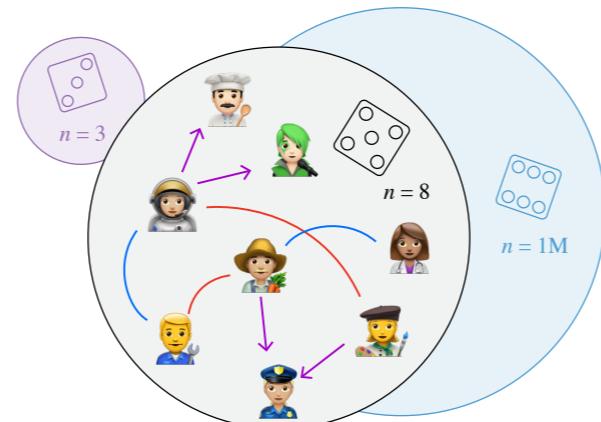
**predicate + variables**

Who is the mother of Alice?

mother("Alice", X).

$S \rightarrow \text{Who is } R ?$   
 $R \rightarrow \text{the } \langle \text{relation} \rangle \text{ of } R'$   
 $R' \rightarrow R \mid \langle \text{name} \rangle$

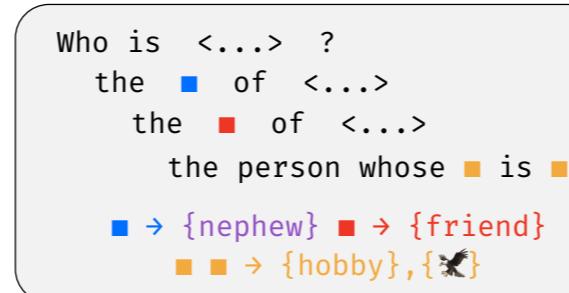
## PhantomWiki Pipeline



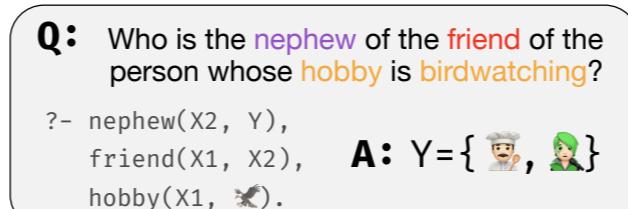
(1) Generate a random universe of size n



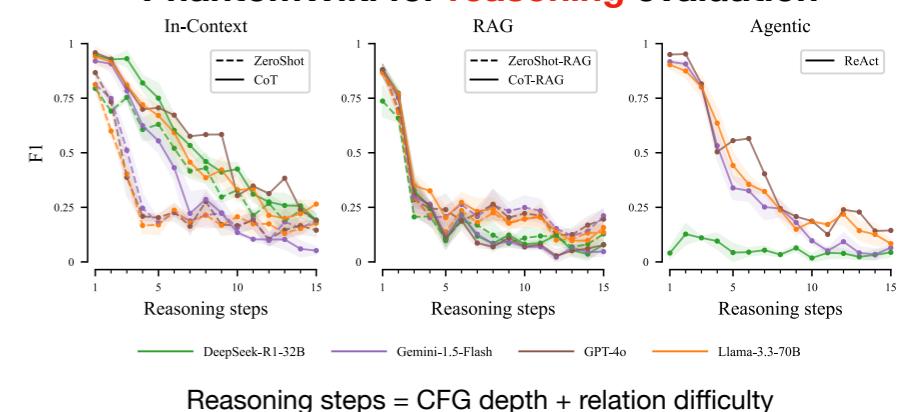
(2) Generate document corpus for the universe



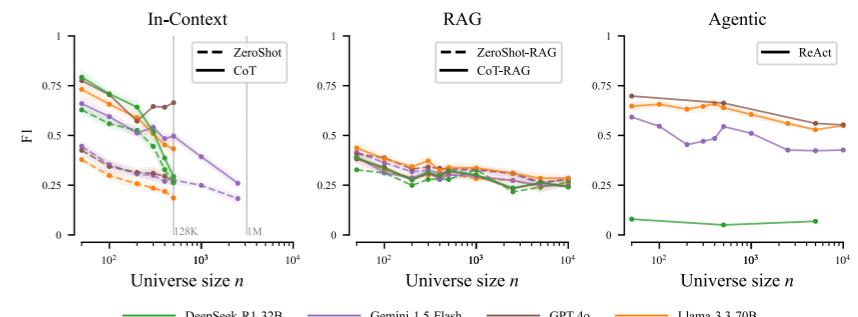
(3) Generate questions using a context-free grammar



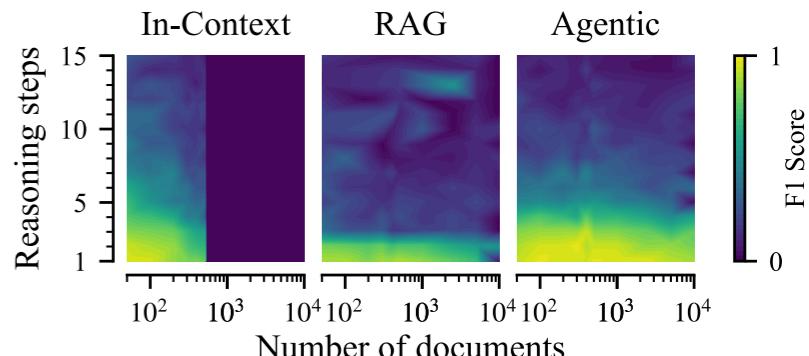
(4) Use Prolog to deduce ground-truth answers



## PhantomWiki for reasoning evaluation

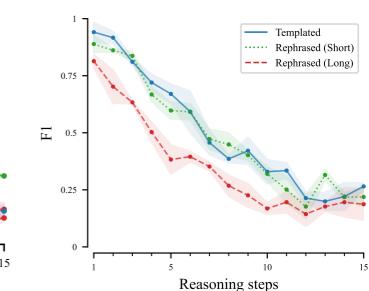
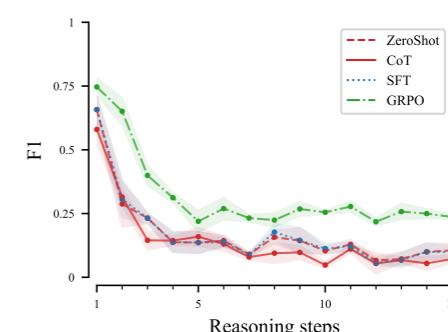


## Retrieval & Reasoning (w/ Llama 3.3-70B)



Does training improve reasoning ability?

Does article style affect reasoning ability?



LoRA on Qwen2.5-3B

Templated vs LLM-written