PhantomWiki: On-Demand Datasets for Reasoning & Retrieval Evaluation











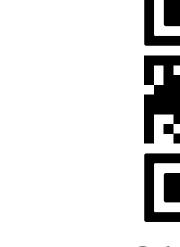
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contribution

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PhantomWiki for reasoning evaluation





Takeaways

PhantomWiki stress tests LLMs on multi-hop questionanswering 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

- In-Context: document corpus fits in LLM context window [NIAH, LongBench, RULER, ∞-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

Query = predicate + variables

The mother of Alice is Charlotte.

Who is the mother of Alice?

mother("Alice", X).

mother("Alice", "Charlotte").

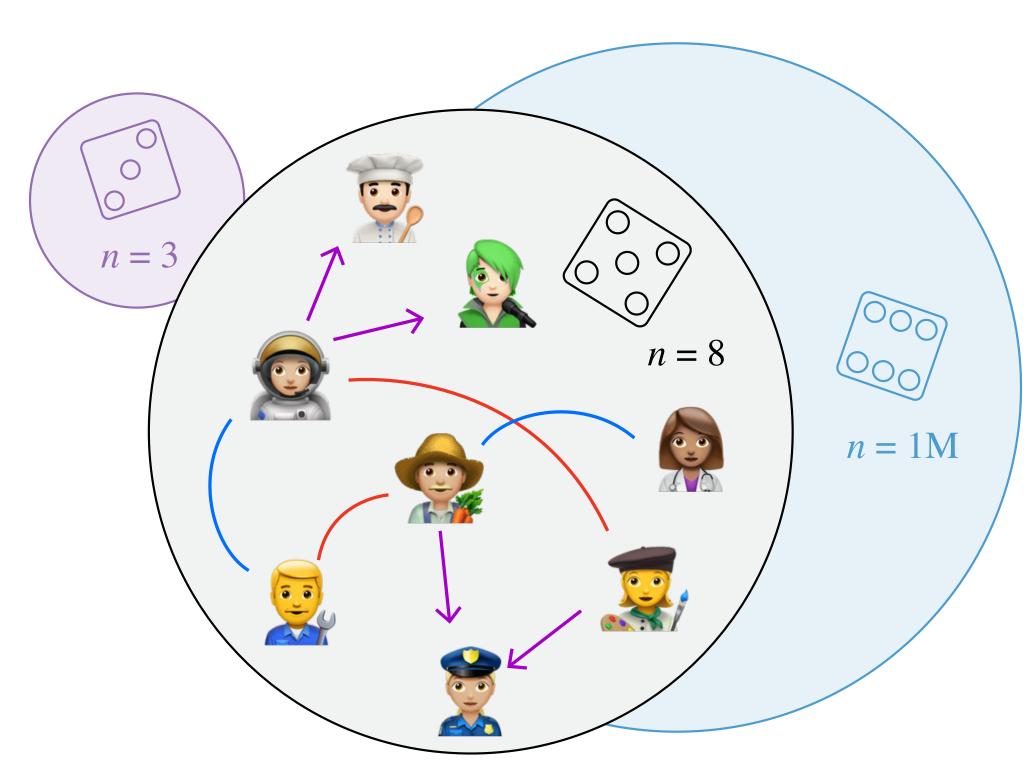
 $S \rightarrow Who is R ?$

the <relation> of

R'

<name>

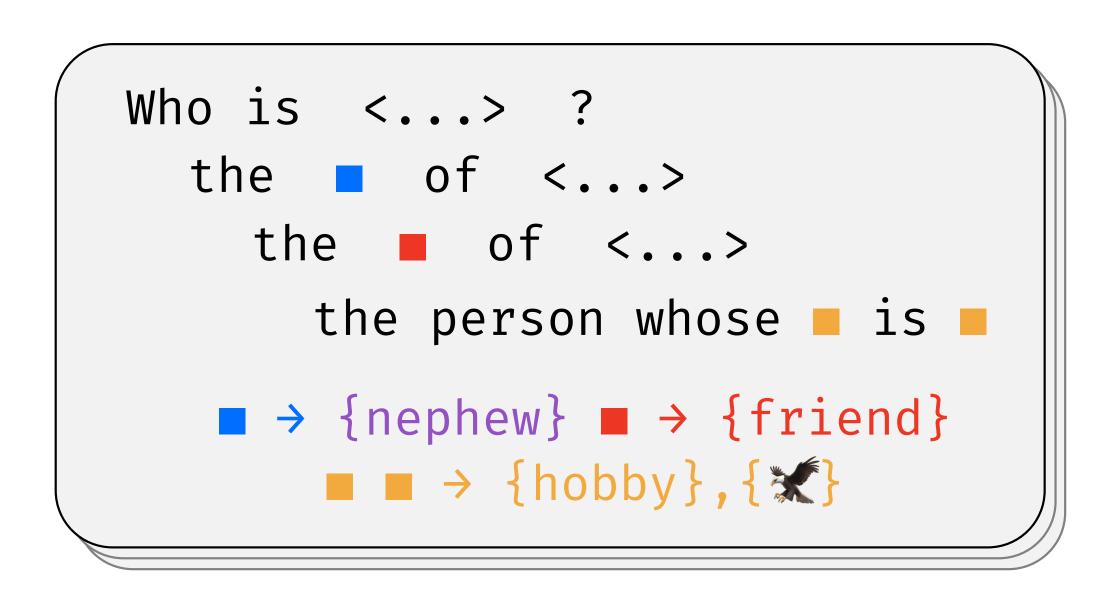
Johann



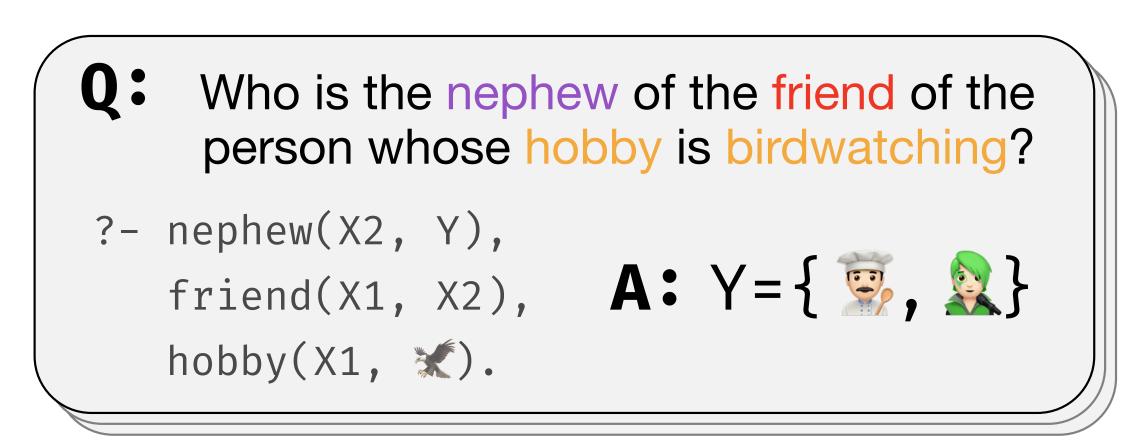
(1) Generate a random universe of size n



(2) Generate document corpus for the universe

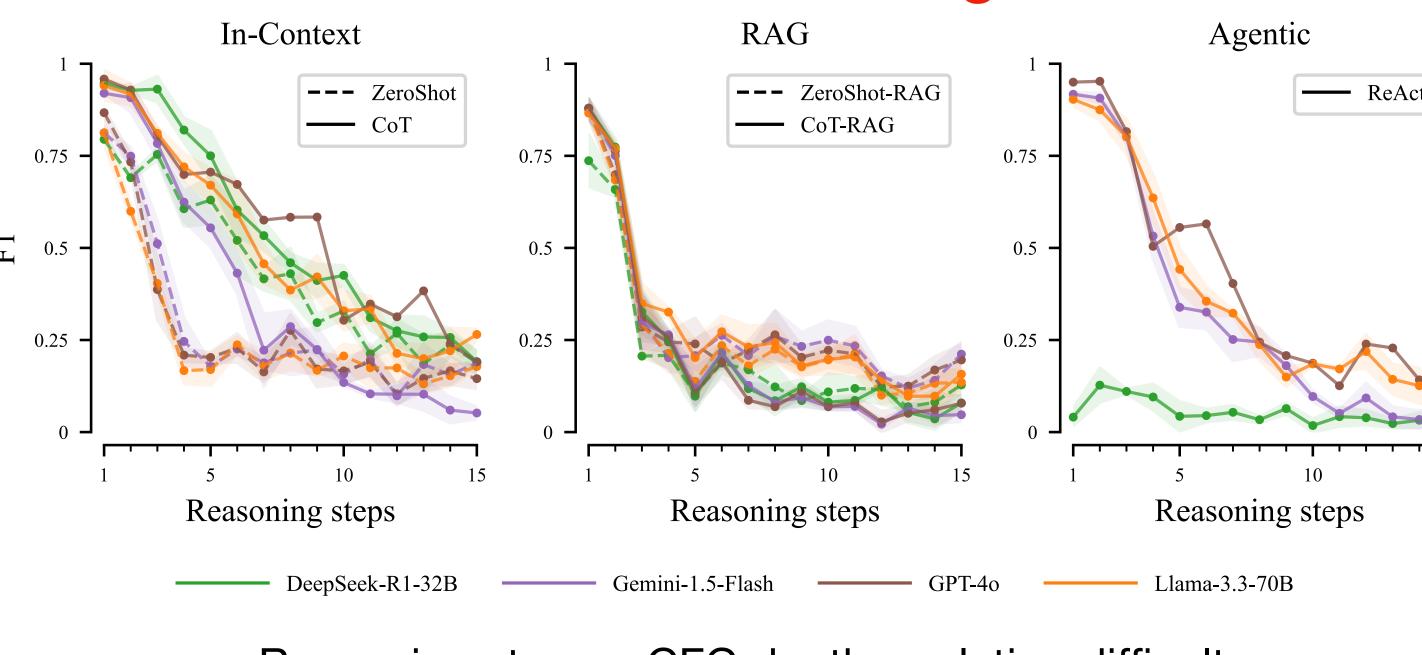


(3) Generate questions using a context-free grammar



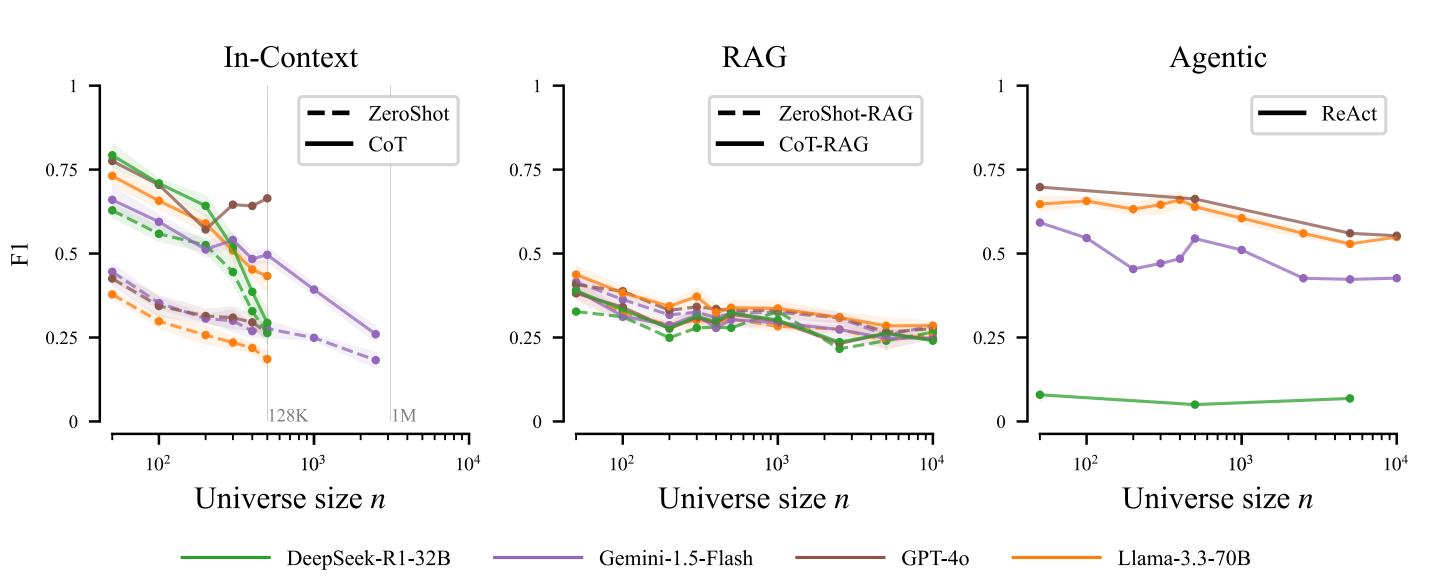
(4) Use Prolog to deduce ground-truth answers

PhantomWiki Pipeline

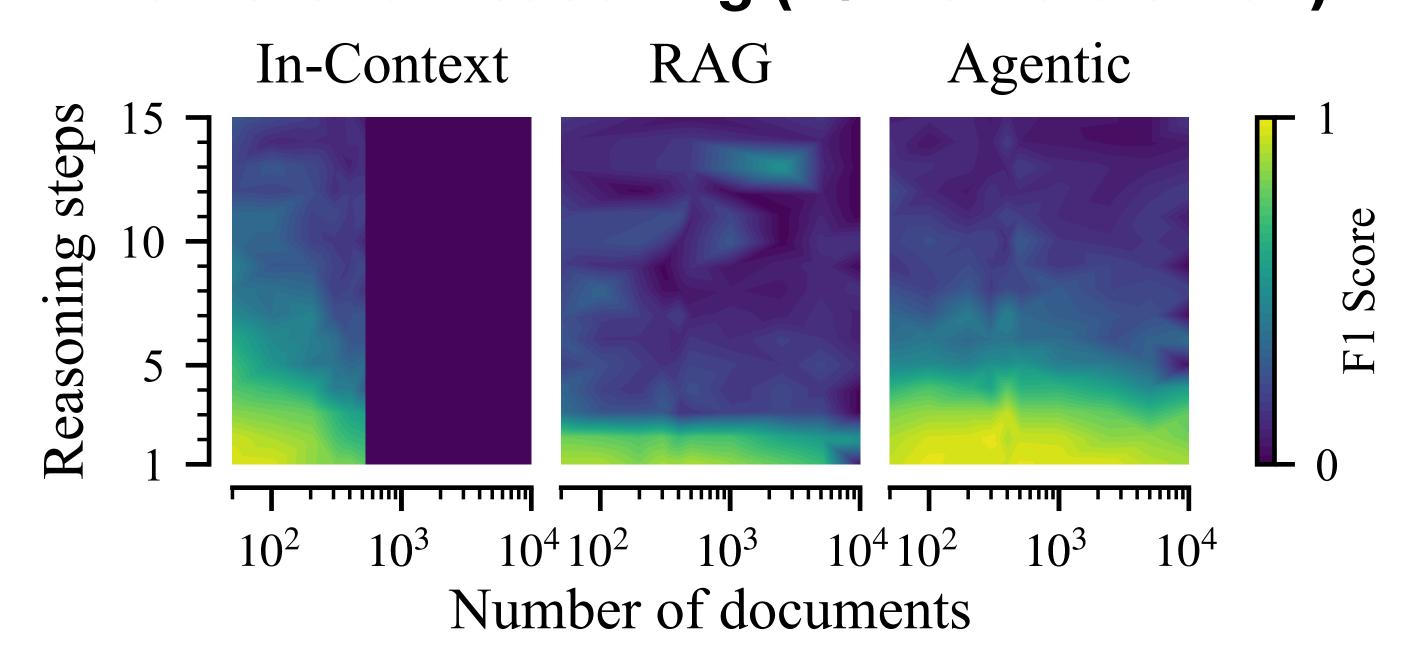


Reasoning steps = CFG depth + relation difficulty

PhantomWiki for retrieval evaluation

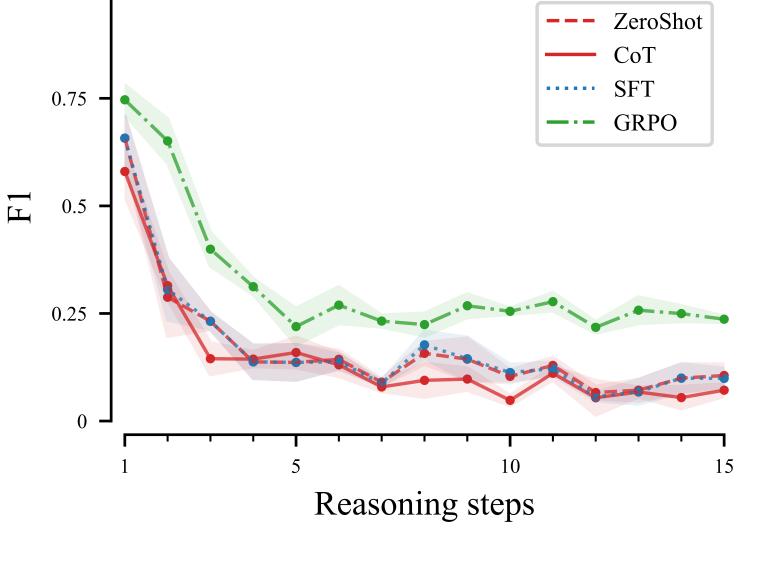


Retrieval & Reasoning (w/ Llama 3.3-70B)

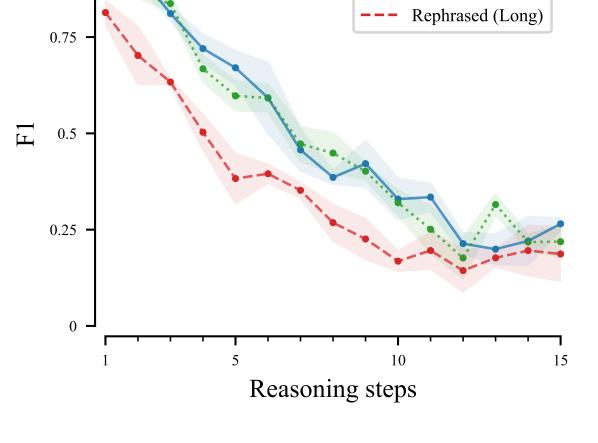


Does training improve reasoning ability?

Does article style affect reasoning ability?



LoRA on Qwen2.5-3B



Templated vs LLM-written