

# Multilingual Evaluation of Long Context Retrieval and Reasoning

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# How well do models use long contexts across different languages?



## [outline]

- Multilingual retrieval and reasoning task in long contexts (64k tokens)
- mLongRR: a new dataset for needle-in-a-haystack analysis across 5 languages

## [languages]

Language	ISO 639-3 Code	Resource Level	Language Family	Script
English	eng	Level 5	Indo-European	Latin
Vietnamese	vie	Level 4	Austro-Asiatic	Latin
Indonesian	ind	Level 3	Austronesian	Latin
Swahili	swa	Level 2	Niger-Congo	Latin
Somali	som	Level 1	Afro-Asiatic	Latin



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Retrieval Task, single needle (n=1)
"The special {city} number is: {number}."
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Reasoning Task, multiple needles (n≥2)
"What is the larger/largest magic number?"
"Which city has the larger magic number?"

### [creating mLongRR dataset]

BBC news articles in multiple languages

Naturally occurring text

Recent data

## [prompt]

You are a helpful AI bot that answers questions for a user. Keep your response short and direct. The following is a set of context and a question that will relate to the context.

{context}

#### #QUESTION

What is the special magic number? Don't give information outside the document or repeat your findings. If the information is not available in the context respond UNANSWERABLE.

### [experiments]

- $\rightarrow$  6 LLMs: GPT-4, GPT-4o, Gemini-1.5, Claude-3, Yarn-7b, Llama-3
- $\rightarrow$  5 context lengths: 2k, 8k, 16k, 32k, 64k
- $\rightarrow$  5 needle depths: 0%, 25%, 50%, 75%, 100%



Performance drops significantly as task complexity increases (as few as n=3!)

Overall Gemini-1.5 pro yields best performance, followed by GPT-40



Performance drops with longer contexts, especially for needles in the middle

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Performance drops as we get to lower resource languages

	GPT-4	Gemini-1.5	Claude-3	YaRN-7b	Llama-3	GPT-40
English	1.13	1.15	1.15	1.32	1.13	1.11
Vietnamese	2.08	1.20	2.89	2.75	1.27	1.29
Indonesian	1.92	1.40	2.33	2.48	1.91	1.55
Swahili	2.23	1.85	2.36	2.48	2.21	1.68
Somali	2.37	2.09	2.47	2.70	2.36	1.79
Average	1.94	1.53	2.24	2.34	1.77	1.48
		Toke	enization rate	es		

Lower tokenization rate results in better performance across languages and models

### [conclusion]

- Performance drops with
- Longer contexts
- Needles in the middle
- Higher task complexity (English drops from ~100% to ~50% in n=3)
- Lower-resource languages
- Lower tokenization rates correlate with better performance across languages and models

# [thank you]

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