

Keyformer: KV Cache reduction through key tokens selection for Efficient Generative Inference

Muhammad Adnan, Akhil Arunkumar, Gaurav Jain, Prashant J. Nair,
Ilya Soloveychik, Purushotham Kamath

MLSys, 2024

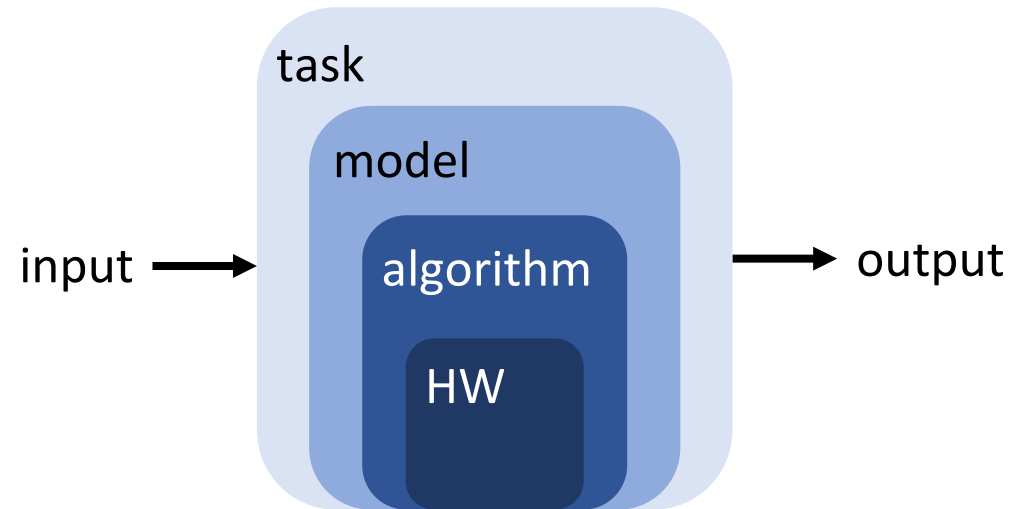


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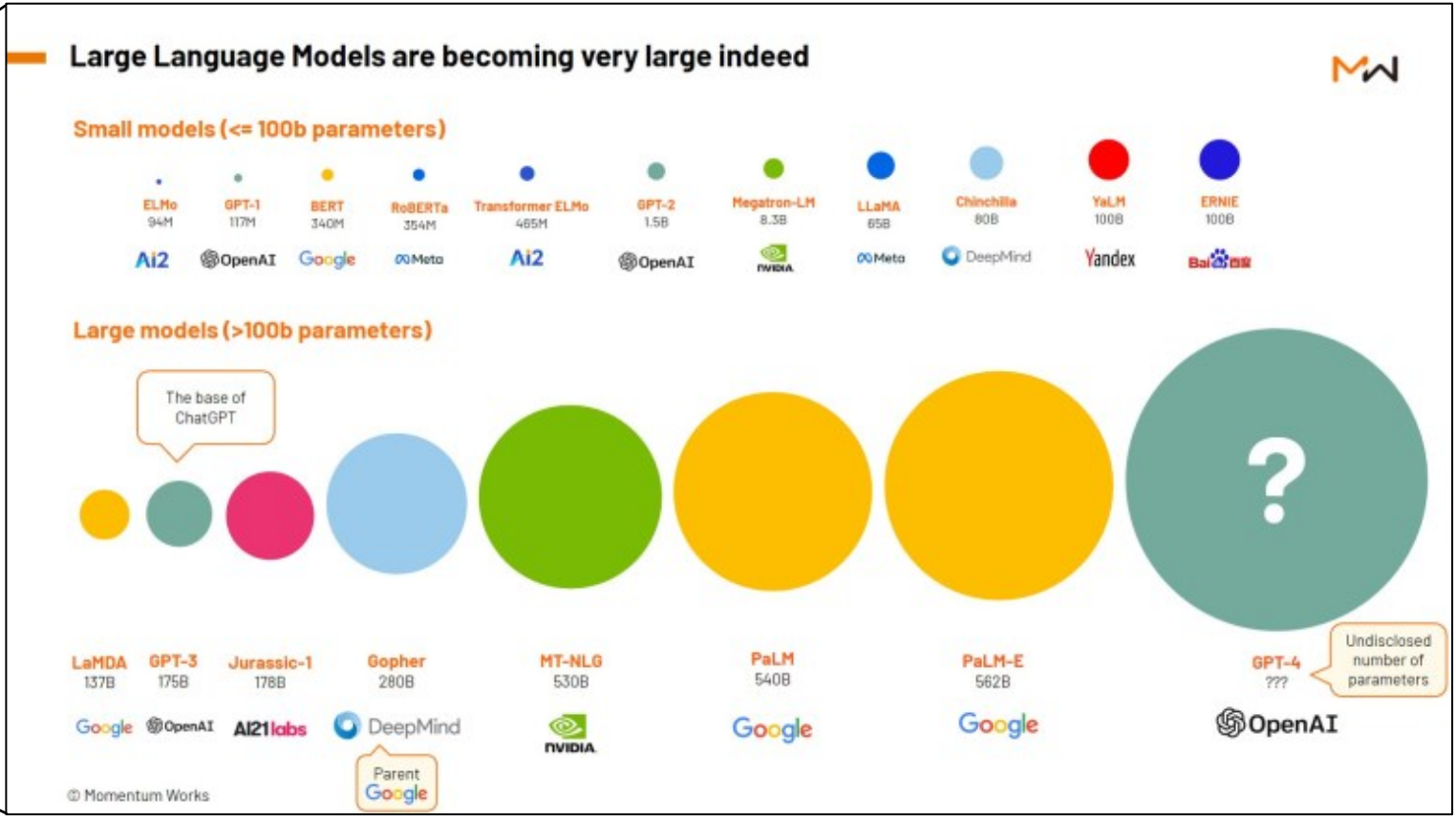
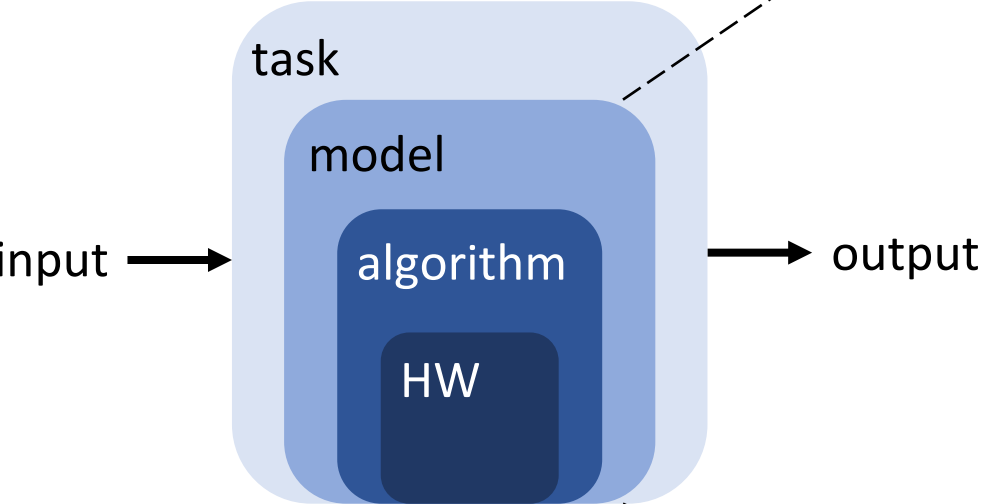
Background

Large Language Models (LLMs) abstraction



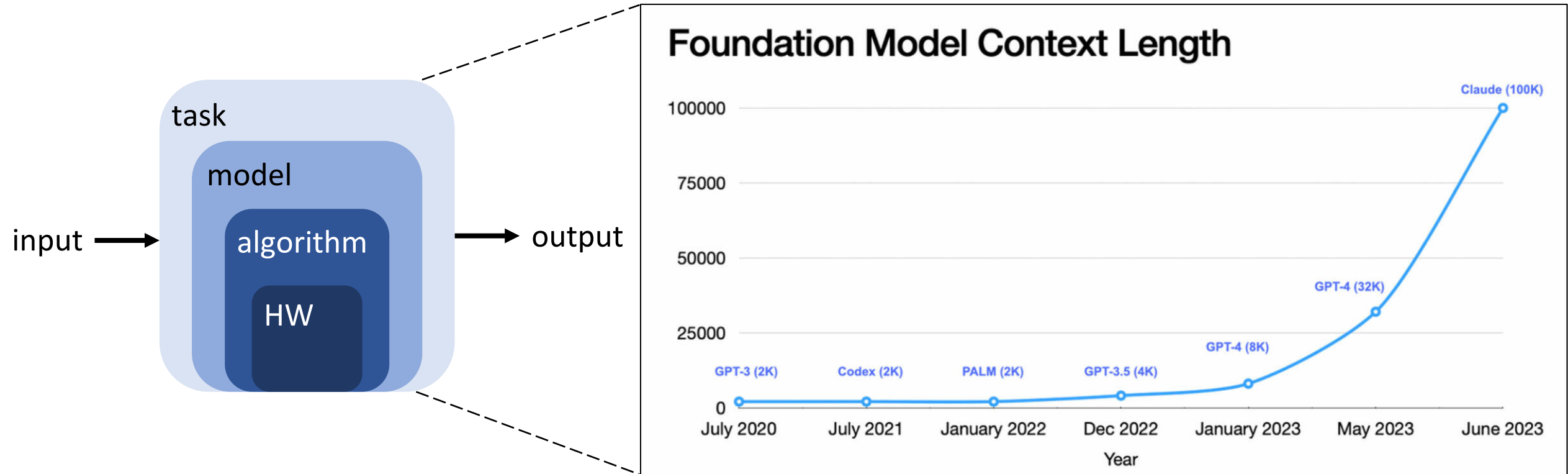
Background

Models are getting bigger

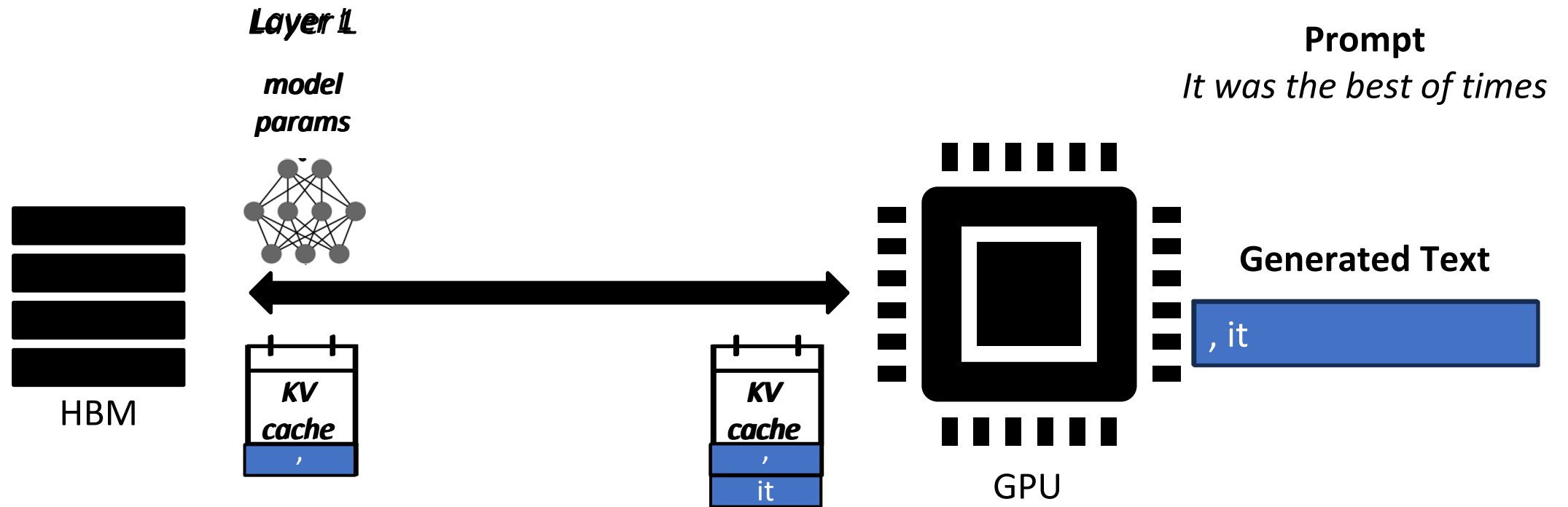


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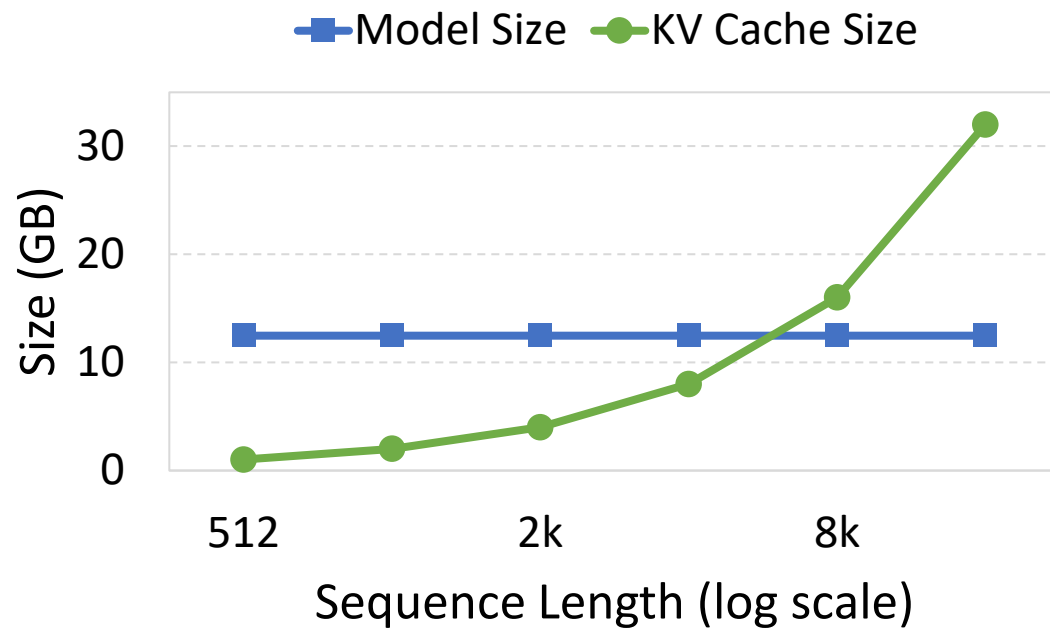
Tasks require longer context length



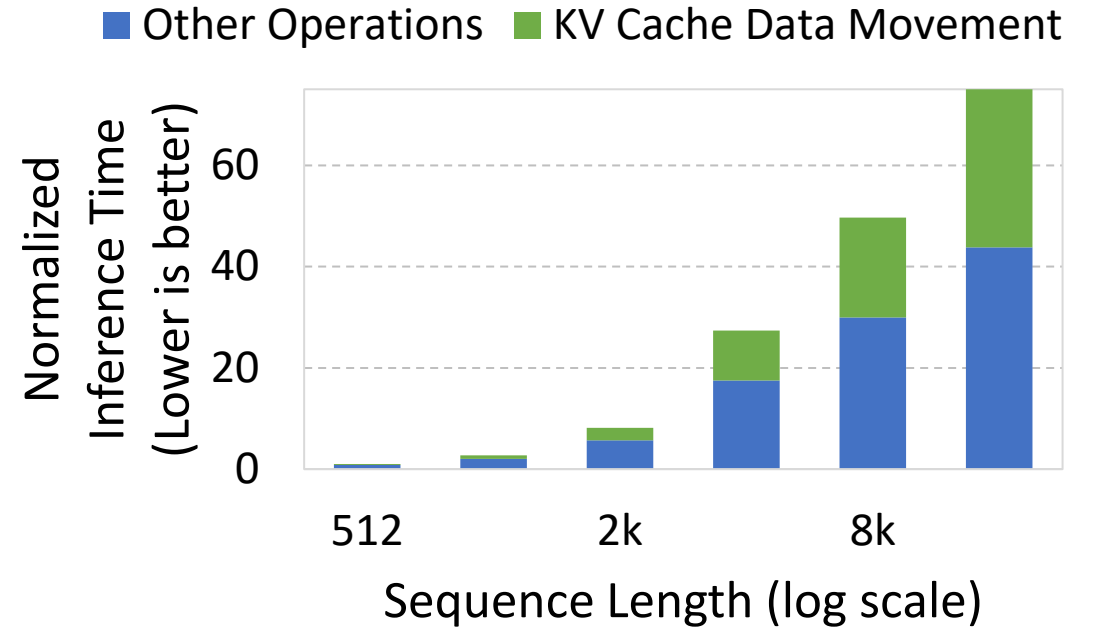
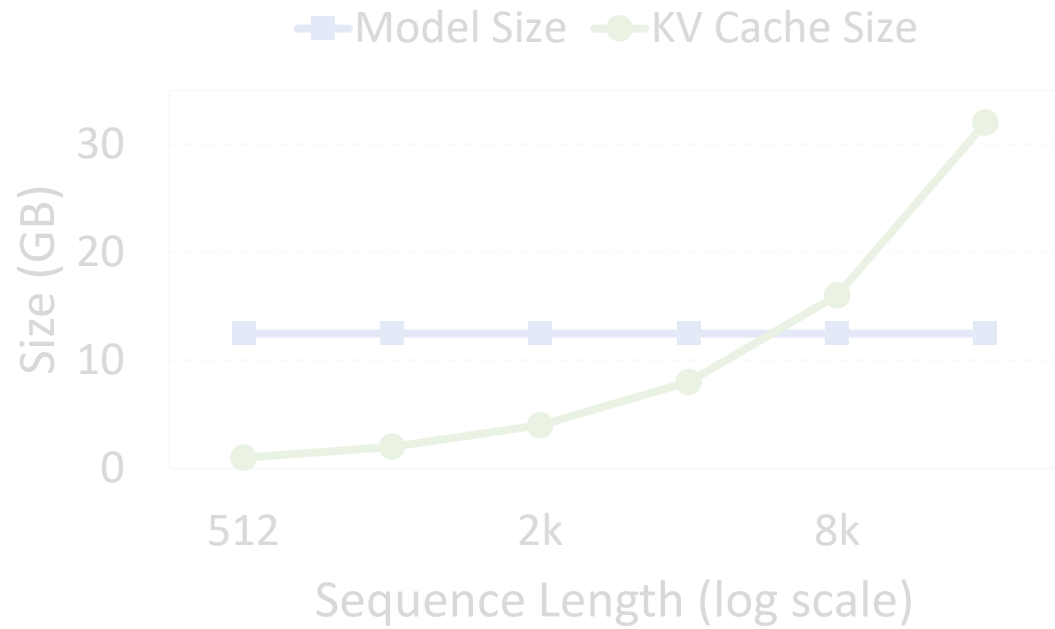
Off-chip Data Movement



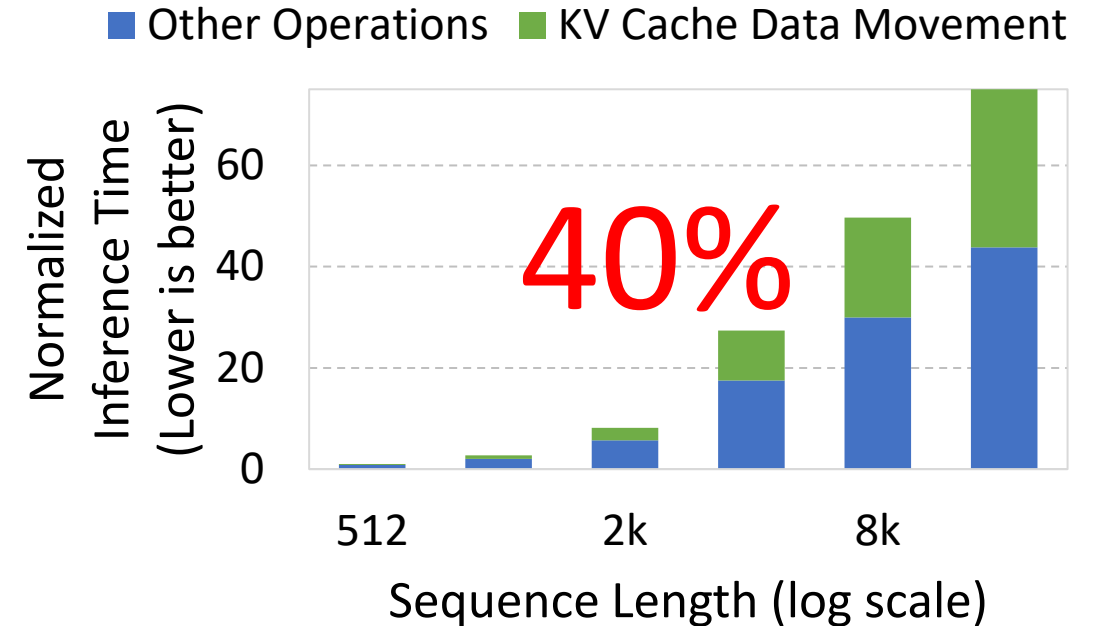
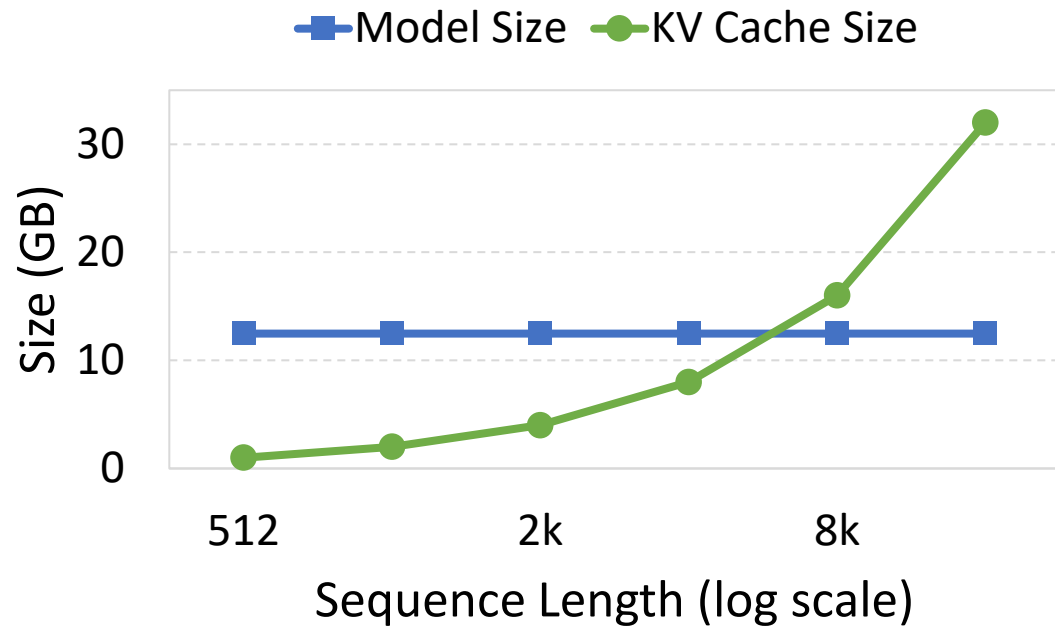
Problem: Capacity, Bandwidth



Problem: Inference Time



Problem: Capacity, Bandwidth & Time



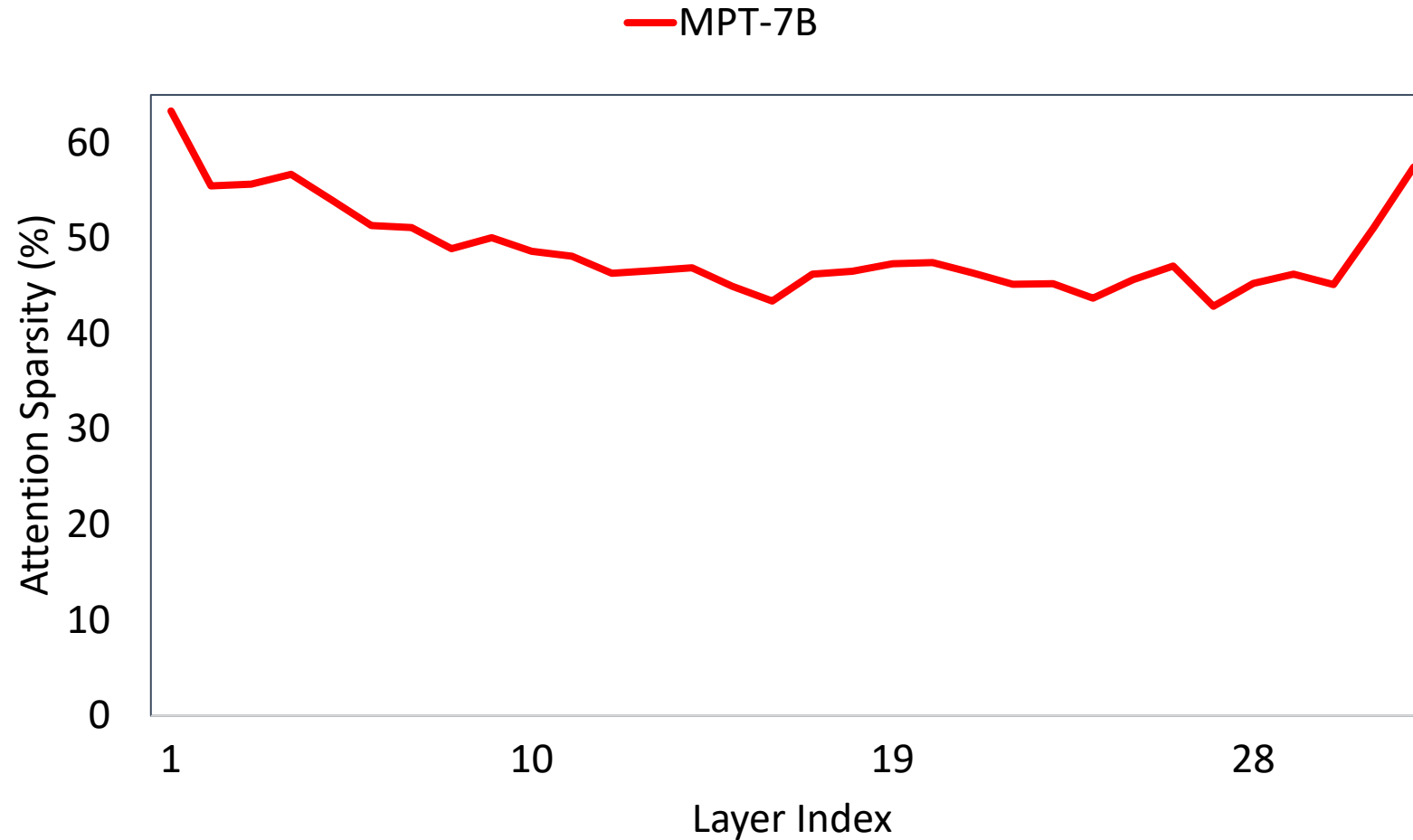
Profiling with MPT-7B model using varying Sequence Length (50% context + 50% text generation) using NVIDIA A100 (80GB) GPU and synthetic data.

Goals

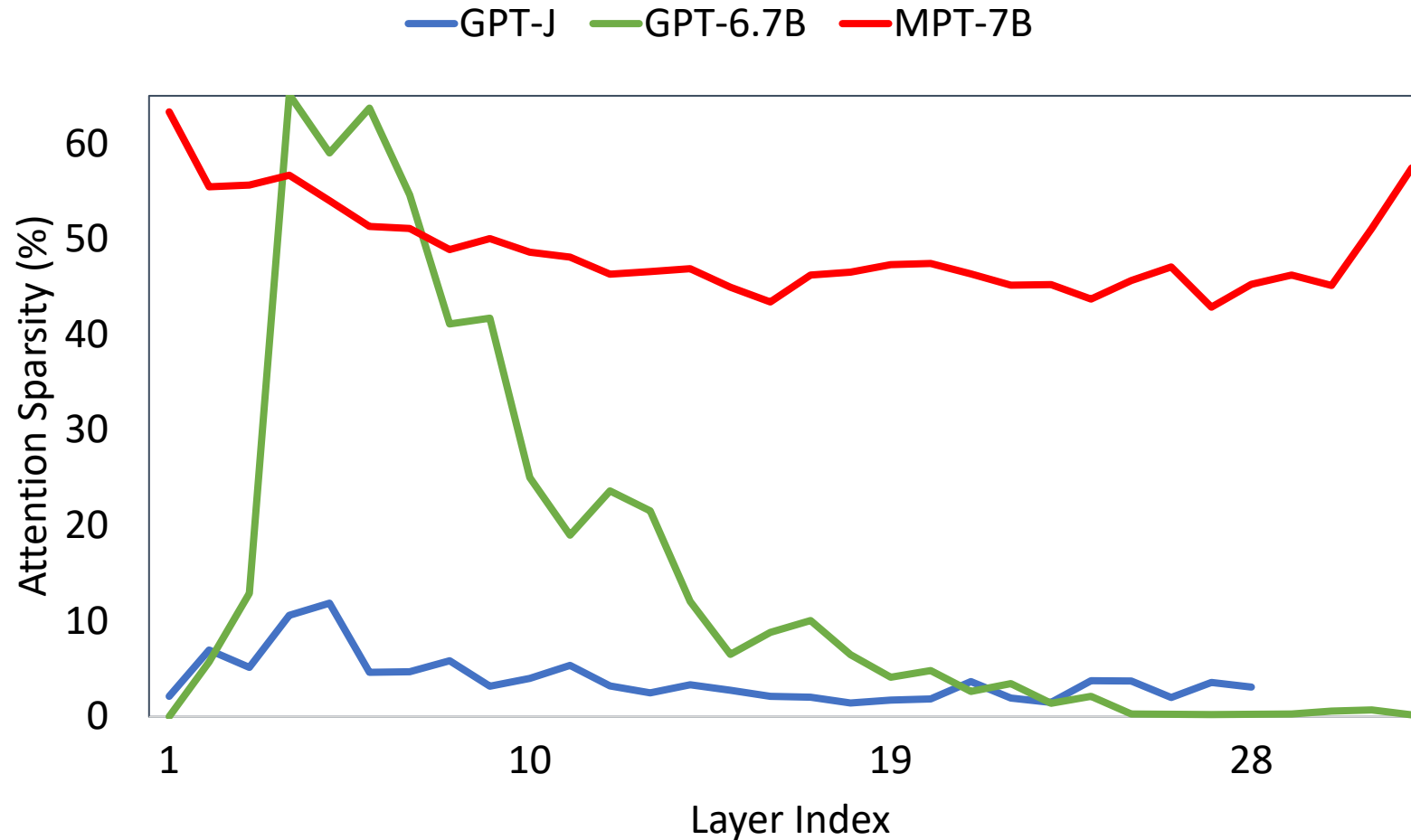
KV cache reduction without accuracy drop

KV cache reduction on the fly without any retraining or finetuning

Key Insight – I: Sparsity

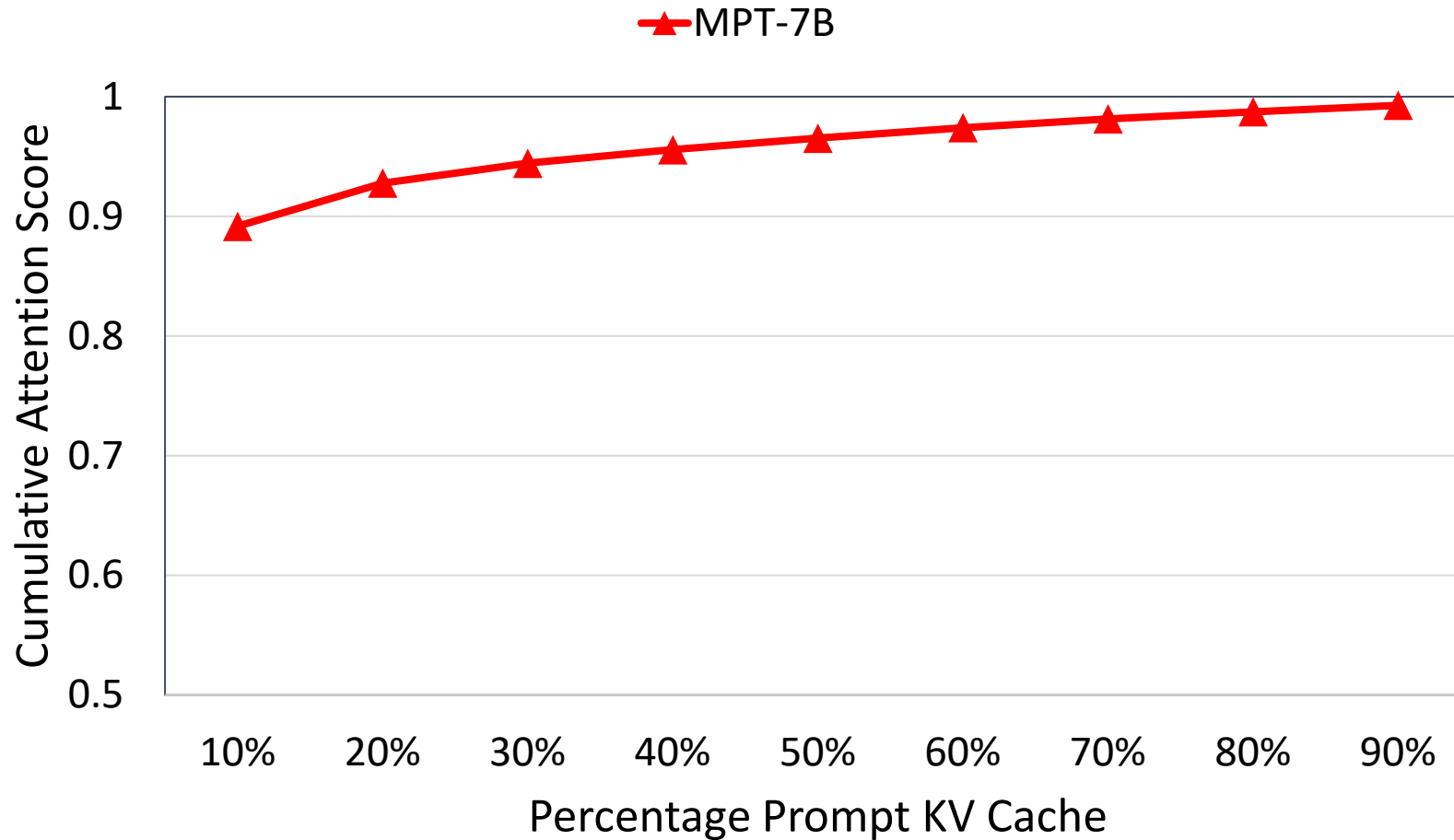


Key Insight – I: Sparsity

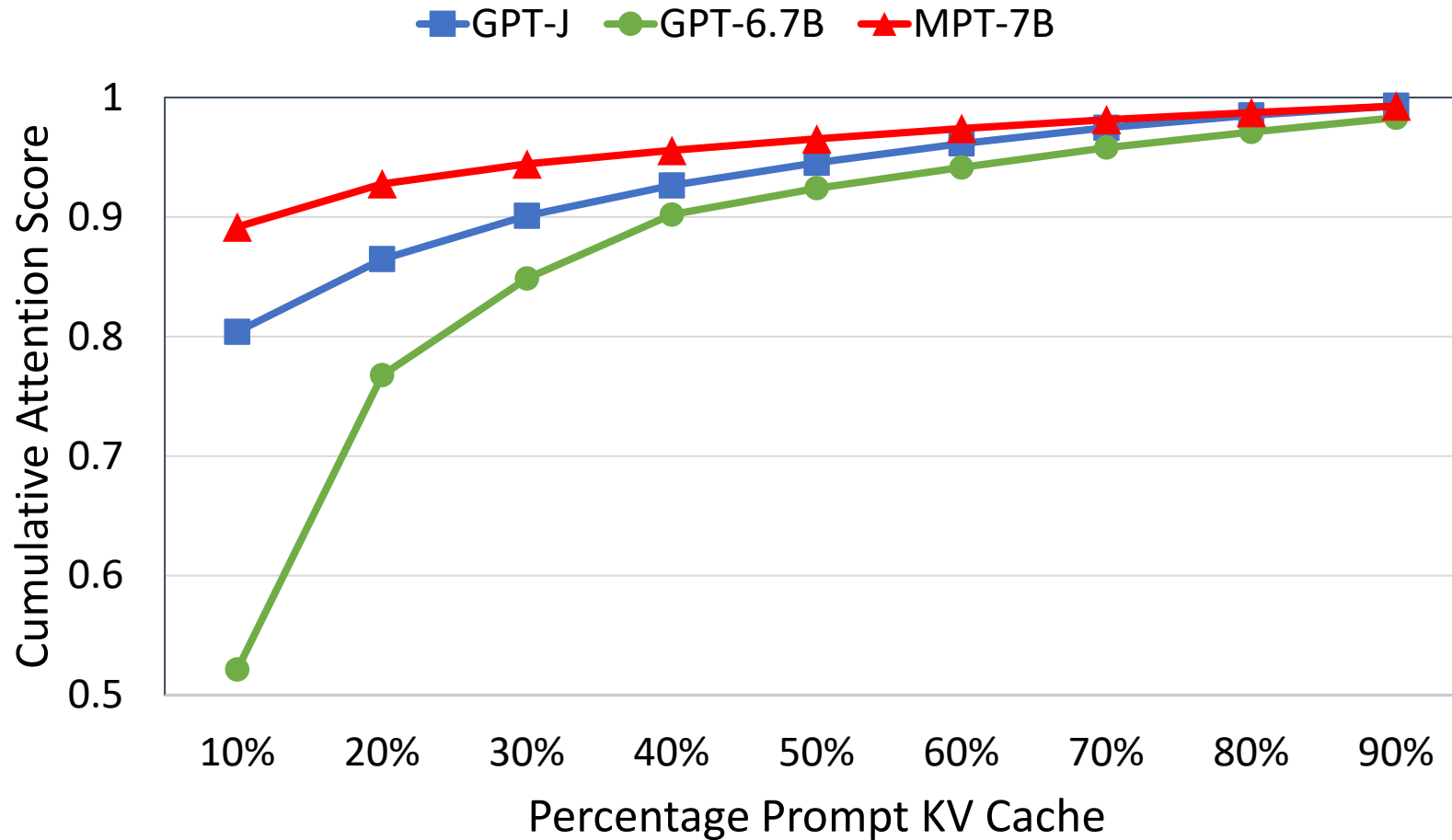


Average Attention Sparsity using CNN/DailyMail dataset for Summarization Task.

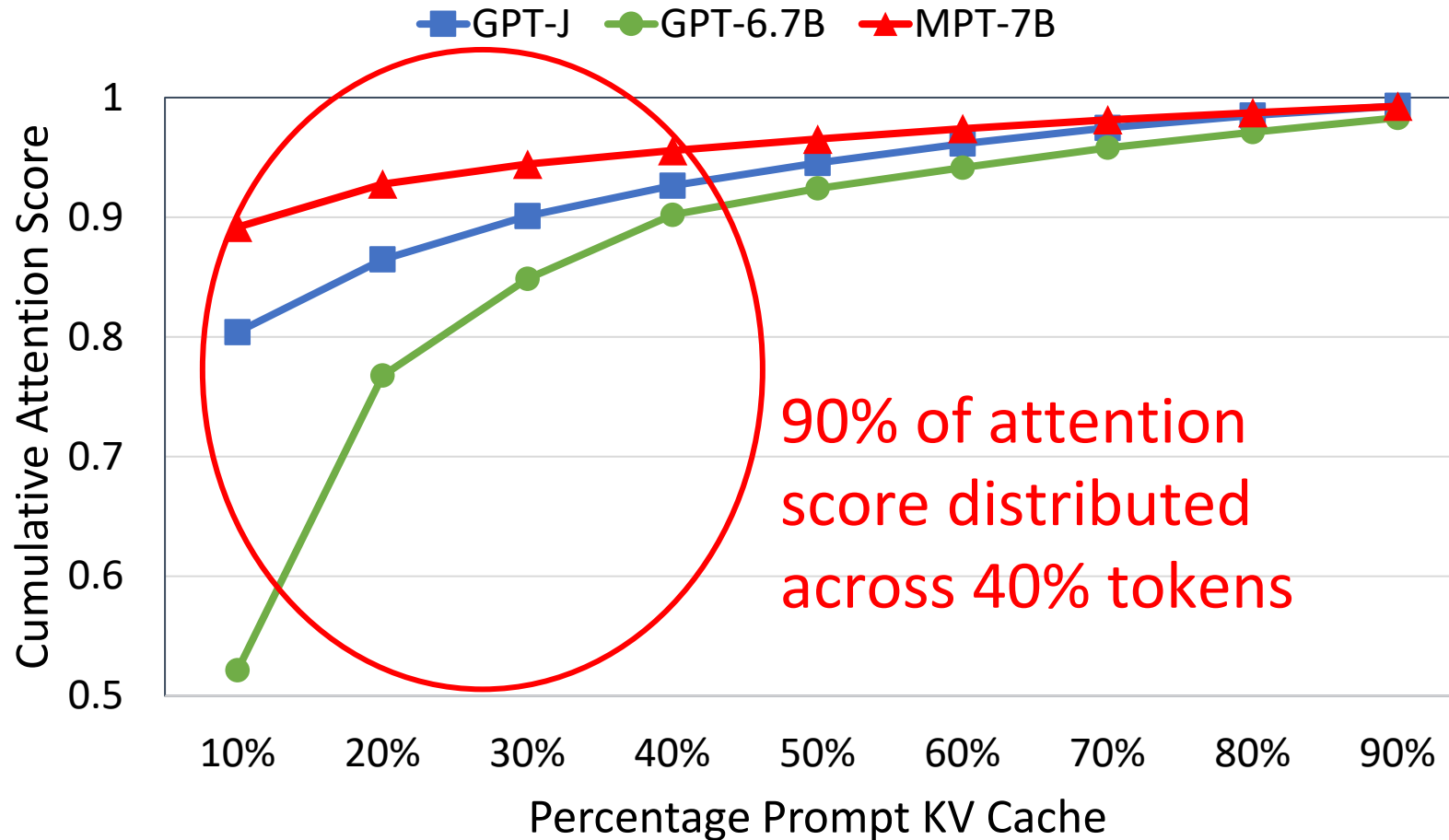
Key Insight – II: Attention Distribution



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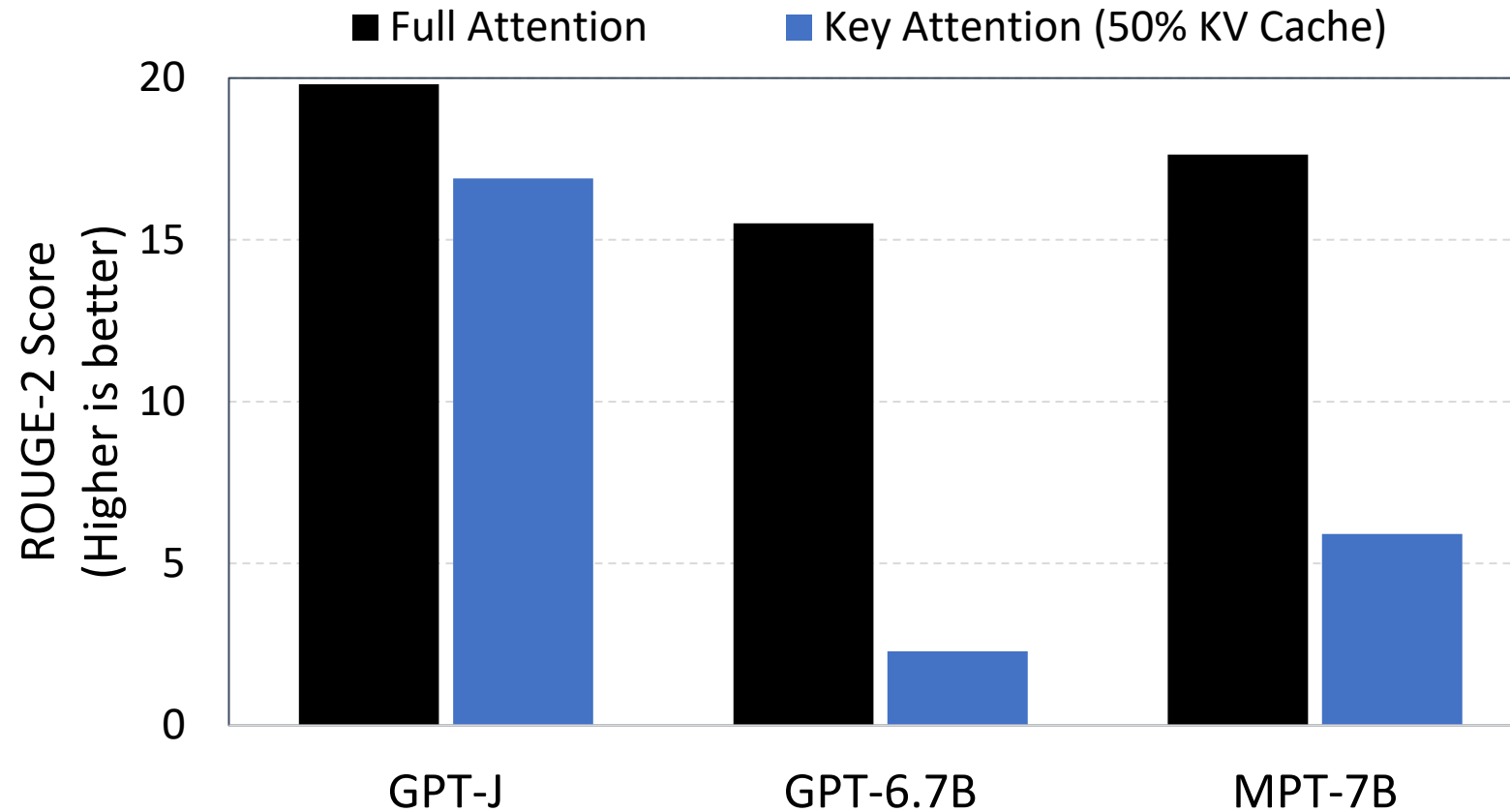


Key Insight – II: Attention Distribution



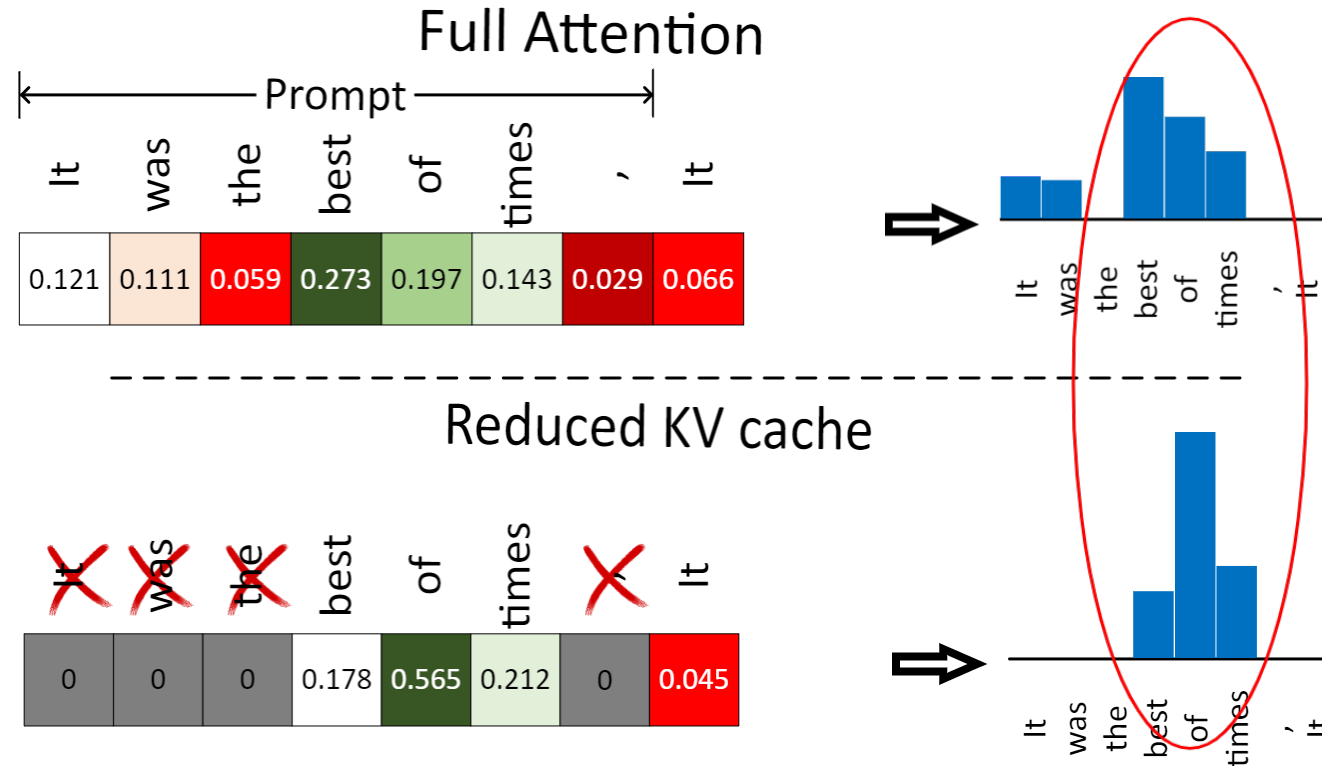
Attention Score Distribution using CNN/DailyMail dataset for Summarization Task.

Limitation – Key Tokens



Accuracy Drop with 50% KV Cache for Summarization Task with CNN/DailyMail dataset.

Key Insight – III: Change in Score Distribution

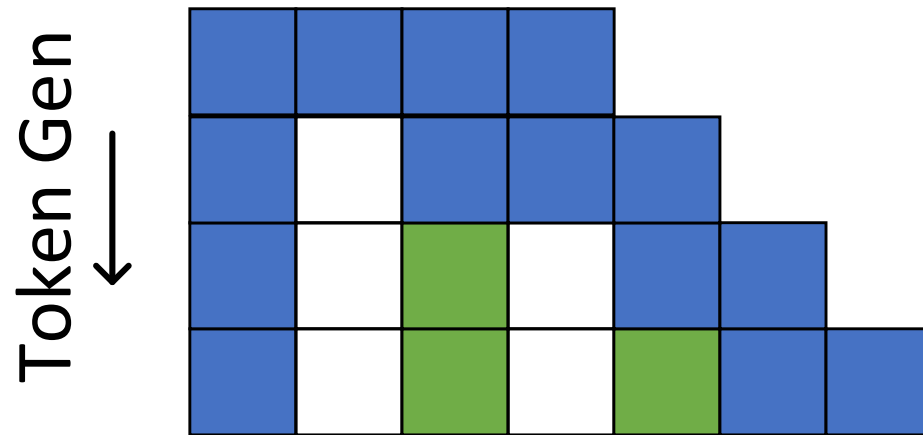


Challenges



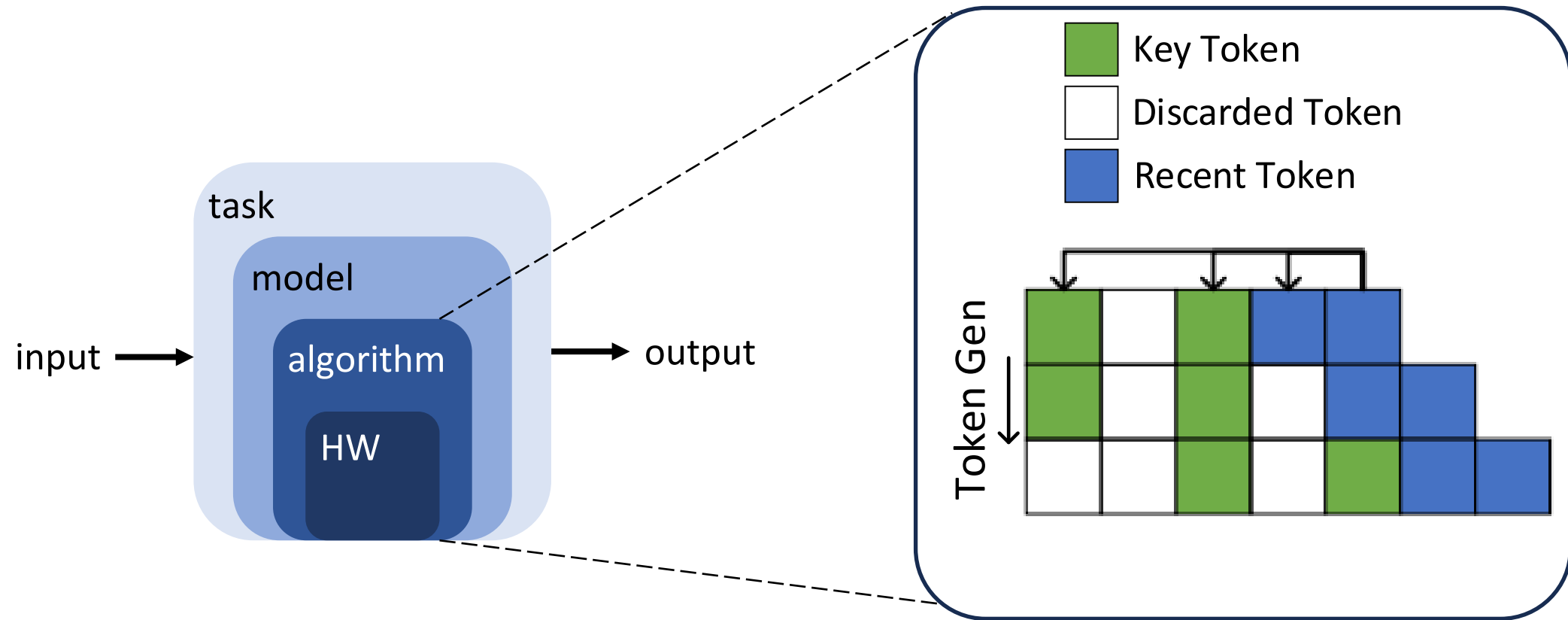
Key Tokens
Identification

 Key Token  Discarded Token

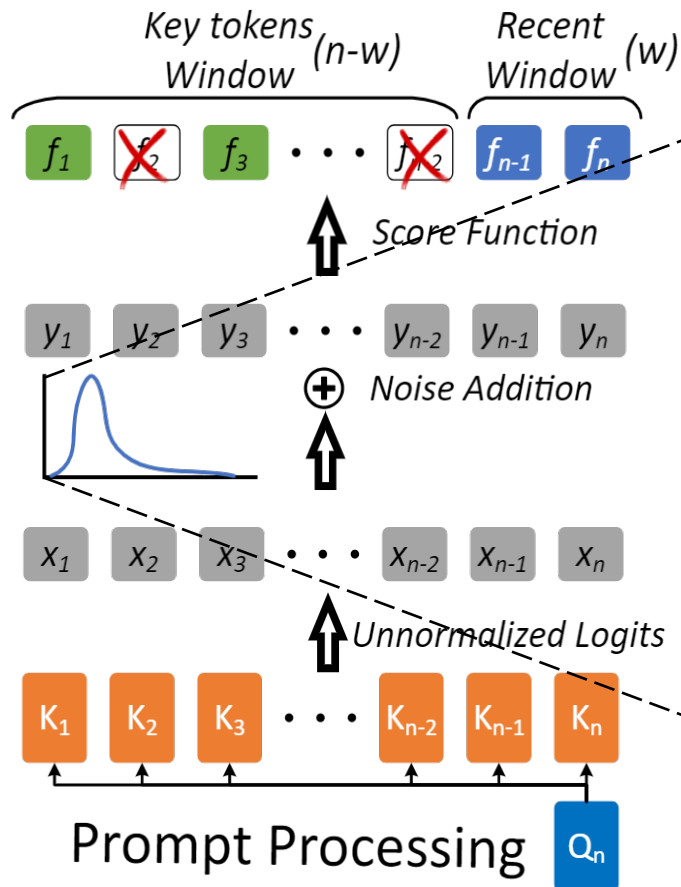
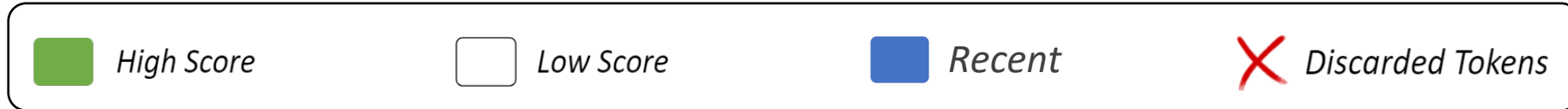


Discarded Tokens
Utilization

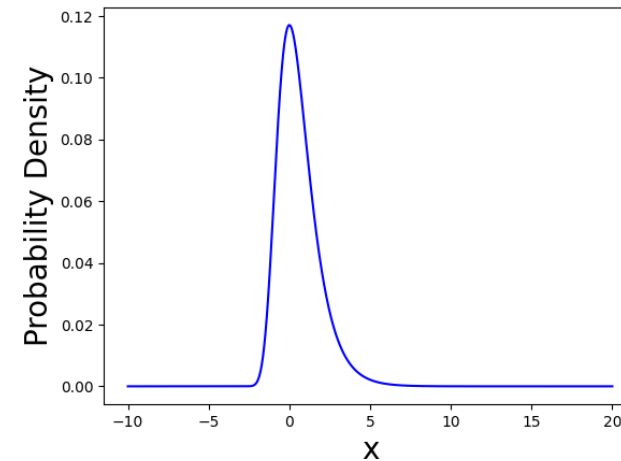
Keyformer – Regularization based Key Token



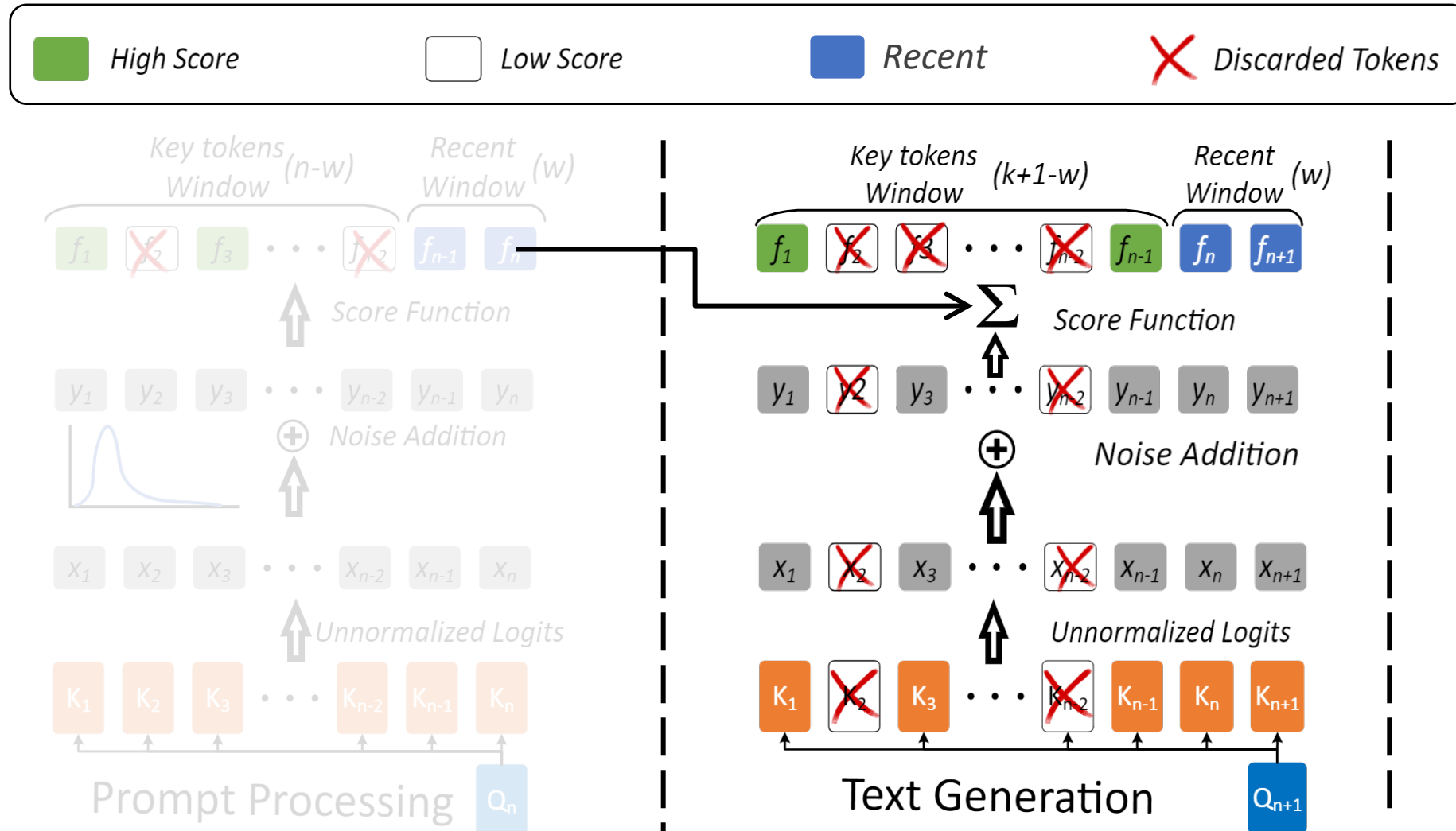
Keyformer - KV Cache Reduction



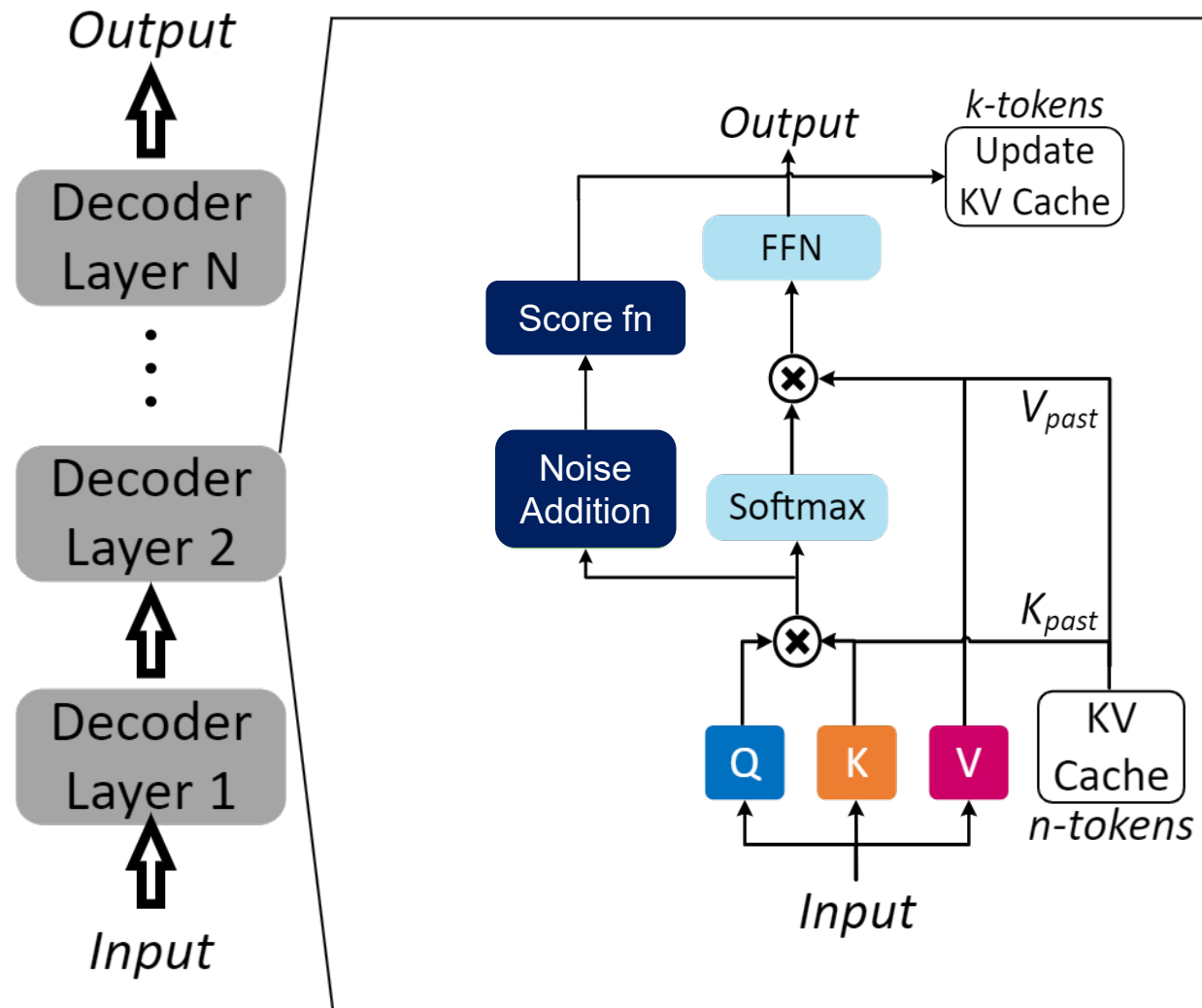
- Bias towards Initial tokens
- Long tail Distribution



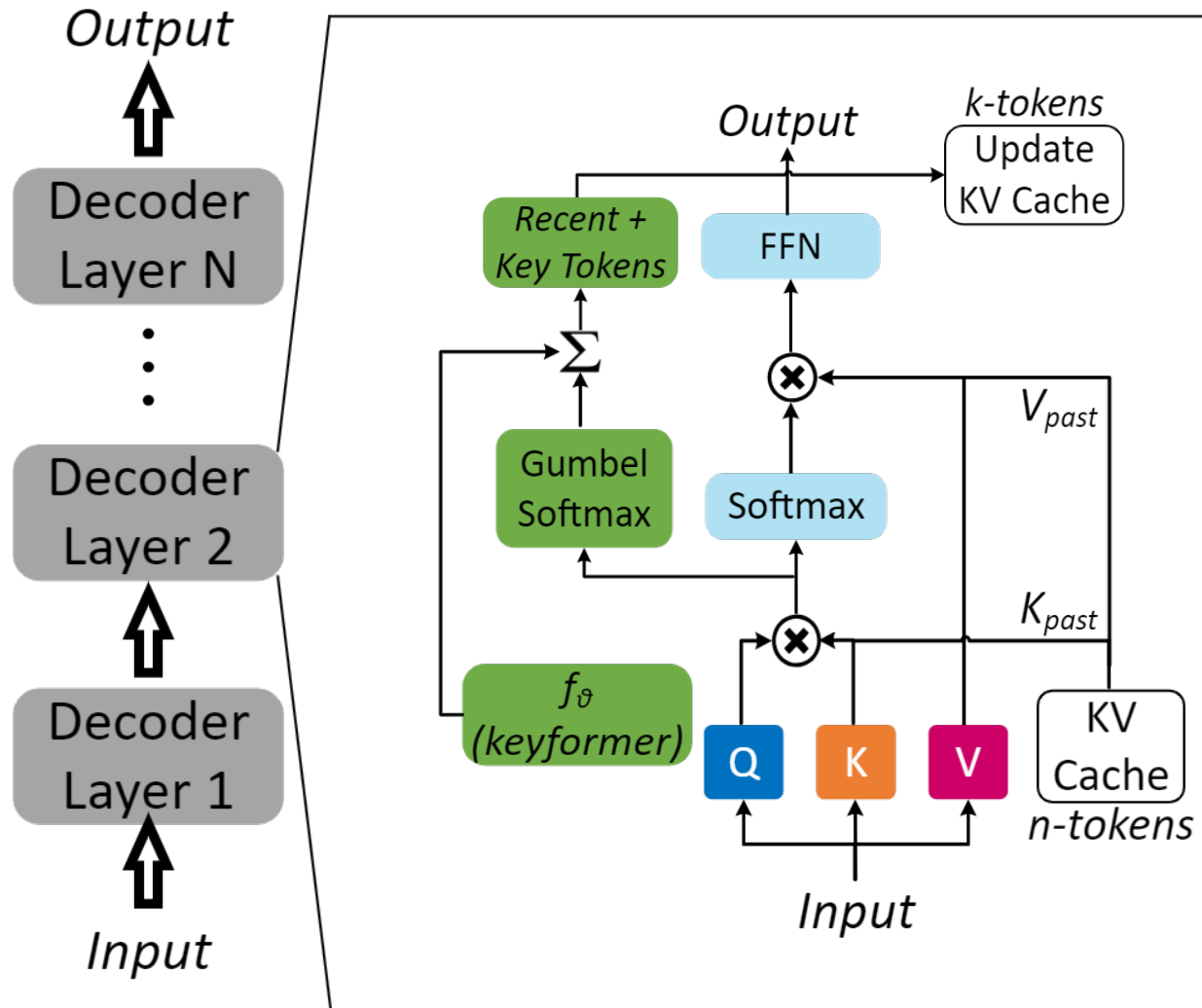
Keyformer - KV Cache Reduction



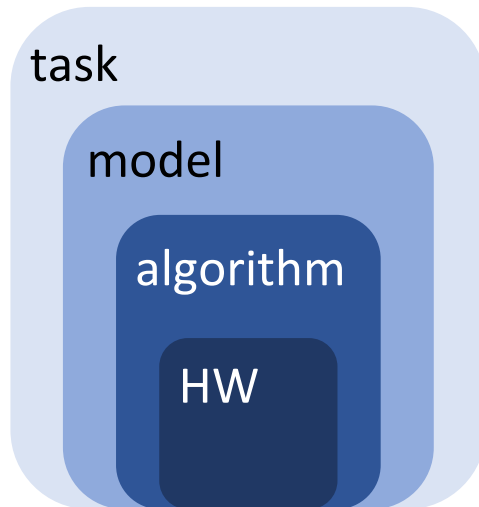
Keyformer - Decoder



Keyformer - Decoder



Evaluation



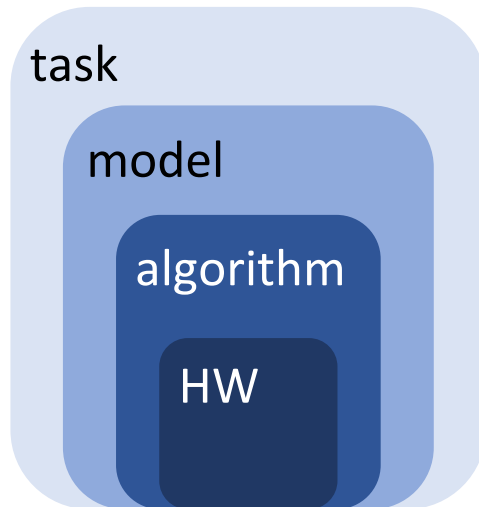
Summarization
Conversation,
COPA (Commonsense Reasoning)
OpenBookQA (Commonsense Reasoning)
Winogrande (Language Understanding)
PIQA (Commonsense Reasoning)

GPT-J (finetuned – Summarization) - *RoPE*
MPT- chat (finetuned – Dialogue) - *ALiBi*
Cerebras GPT (pretrained) - *Absolute*
MPT (pretrained) - *ALiBi*

Full Attention
Window Attention
H₂O

NVIDIA A100 (80GB)

Evaluation



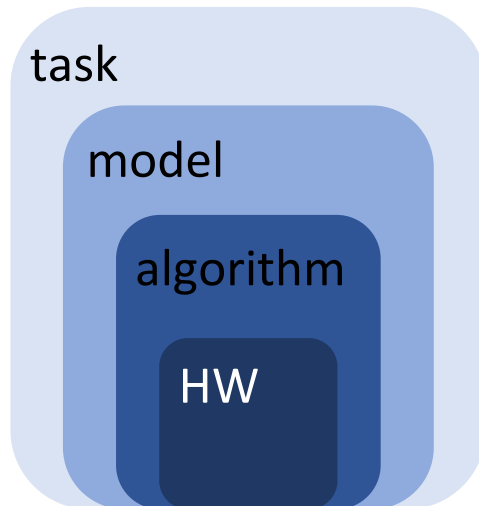
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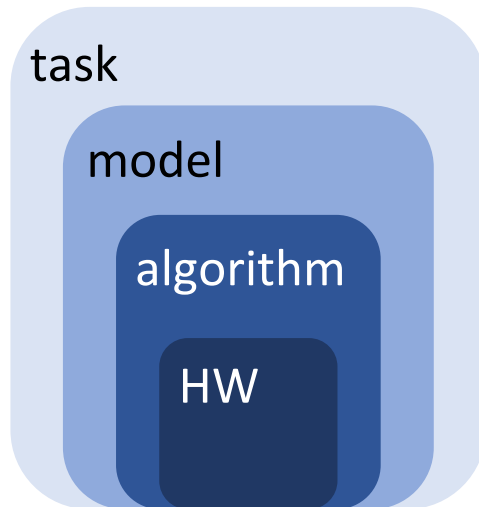
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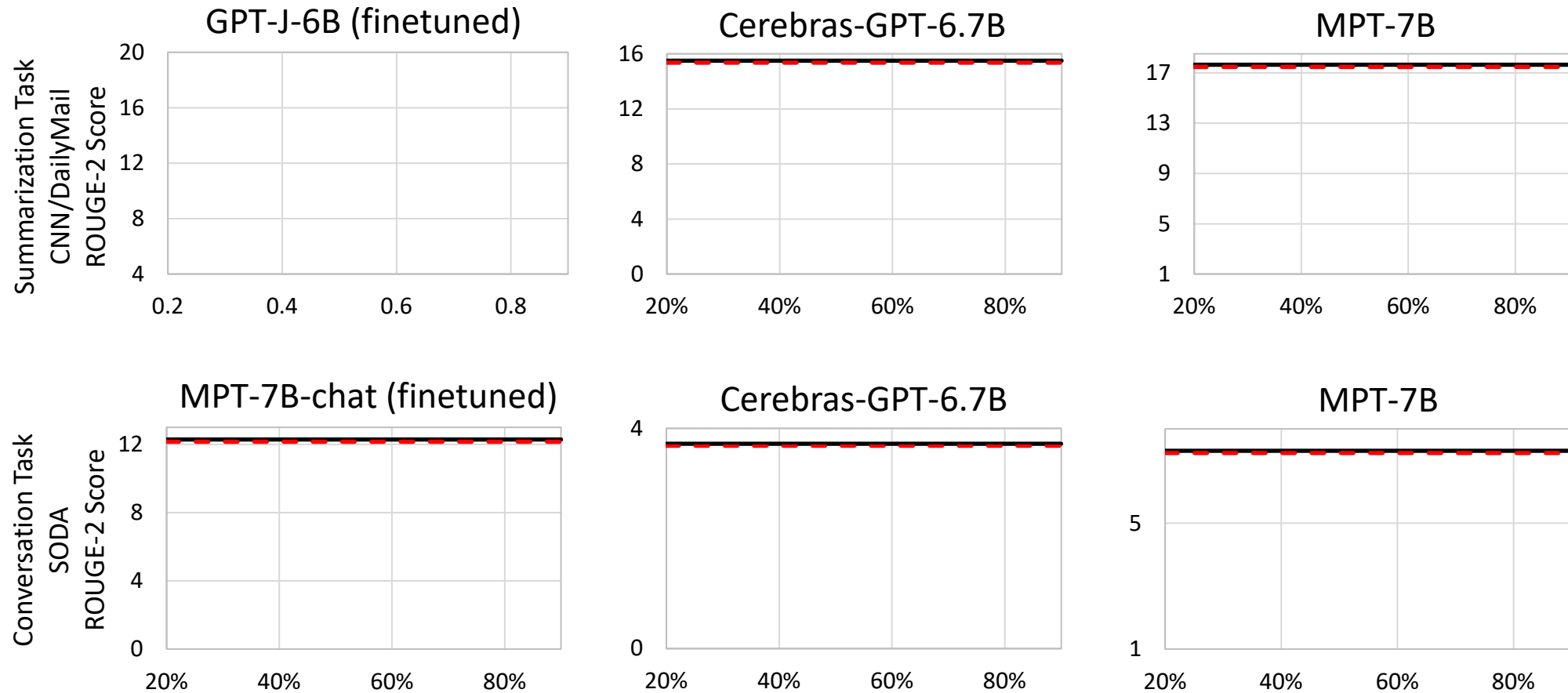
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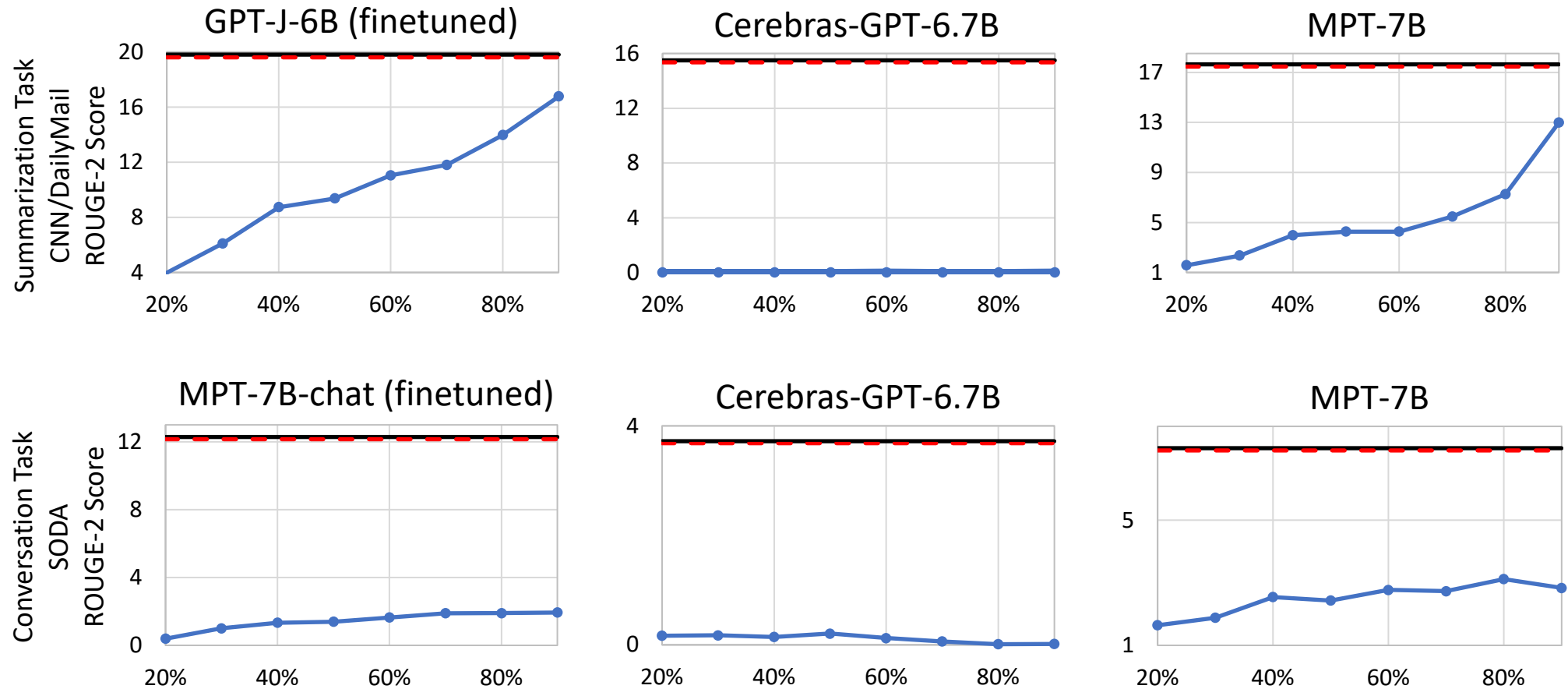
Accuracy Tradeoff

— Full Attention - - 99% Accuracy -●- Window Attention -■- H2O -◆- Keyformer



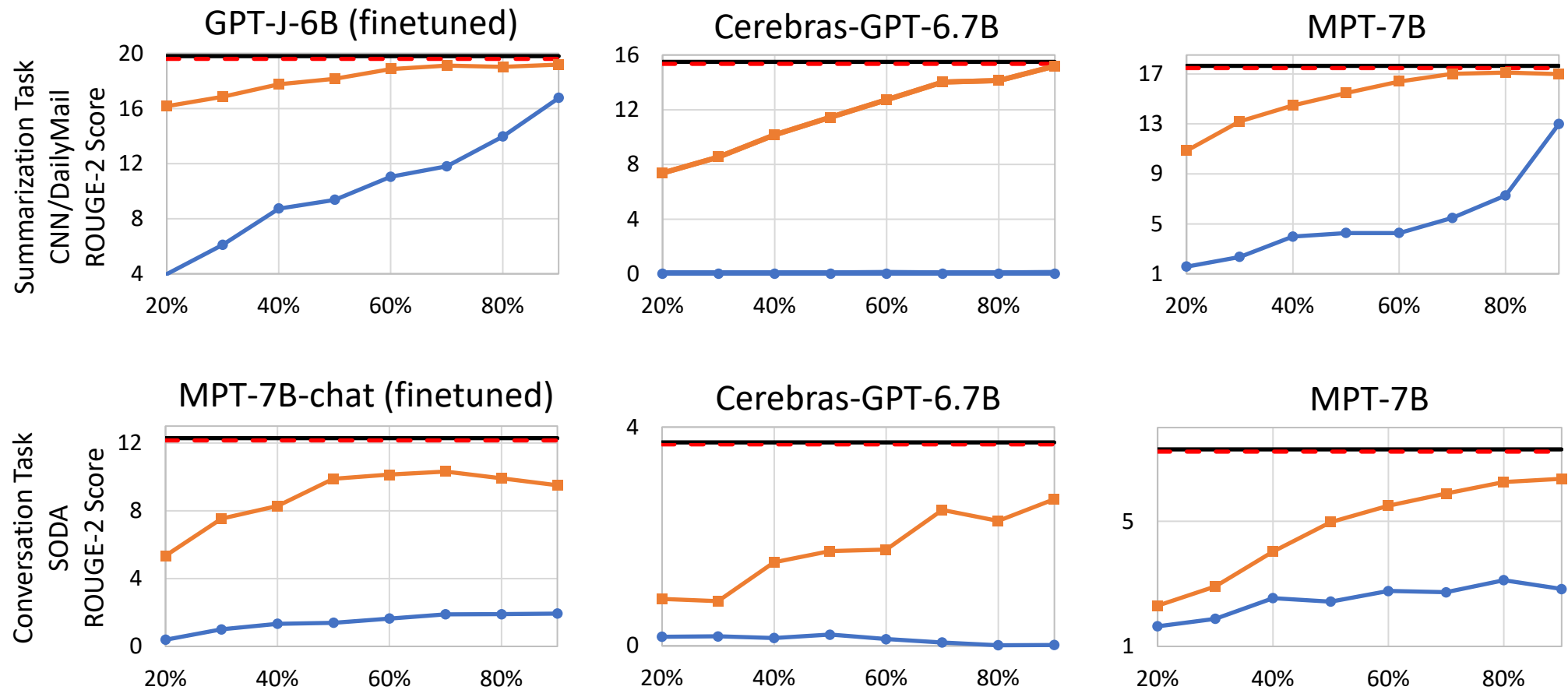
Accuracy Tradeoff

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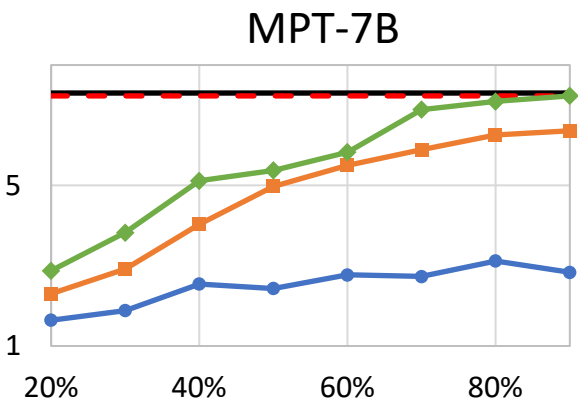
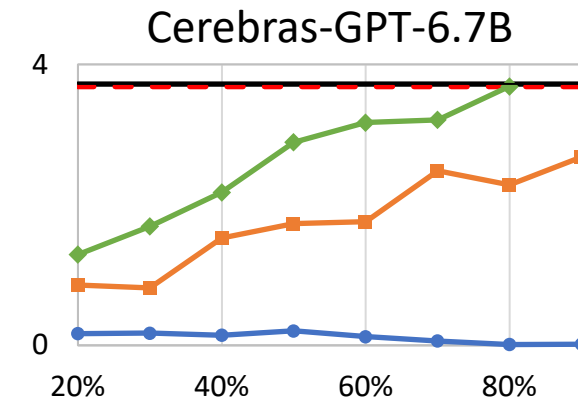
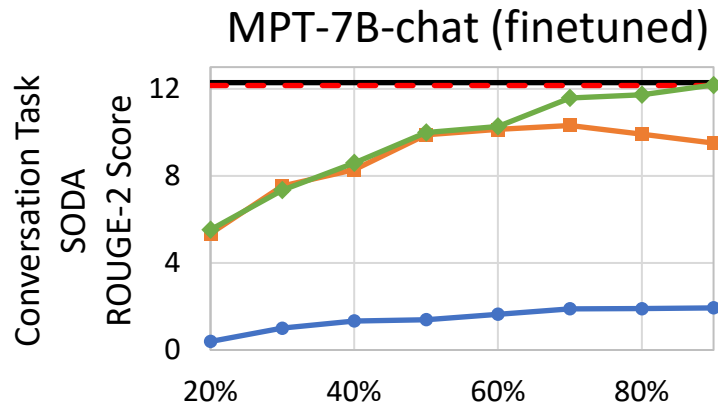
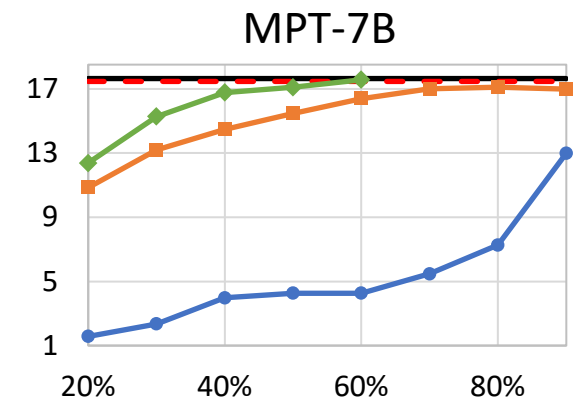
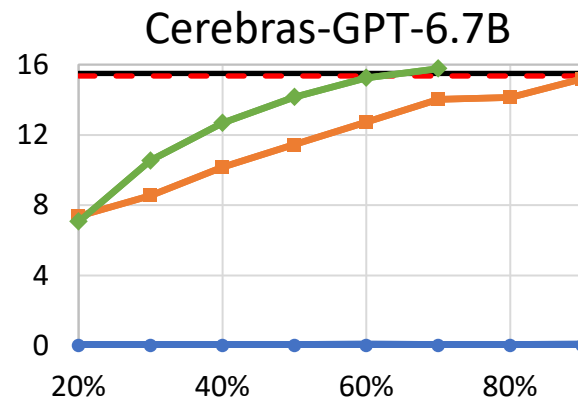
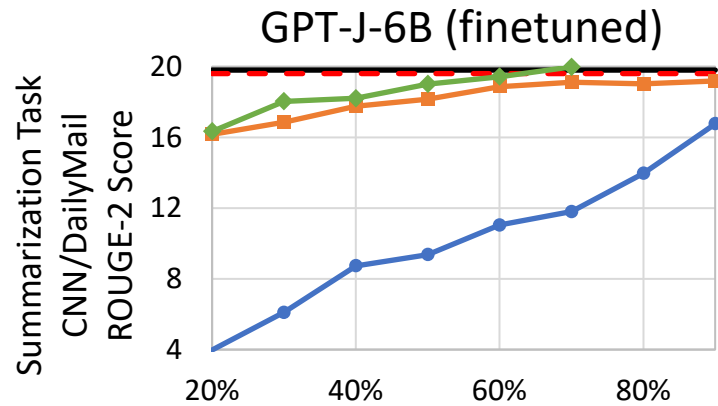
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Accuracy Tradeoff

— Full Attention - - 99% Accuracy — Window Attention — H2O — Keyformer

Summarization Task → **70%** of prompt KV cache

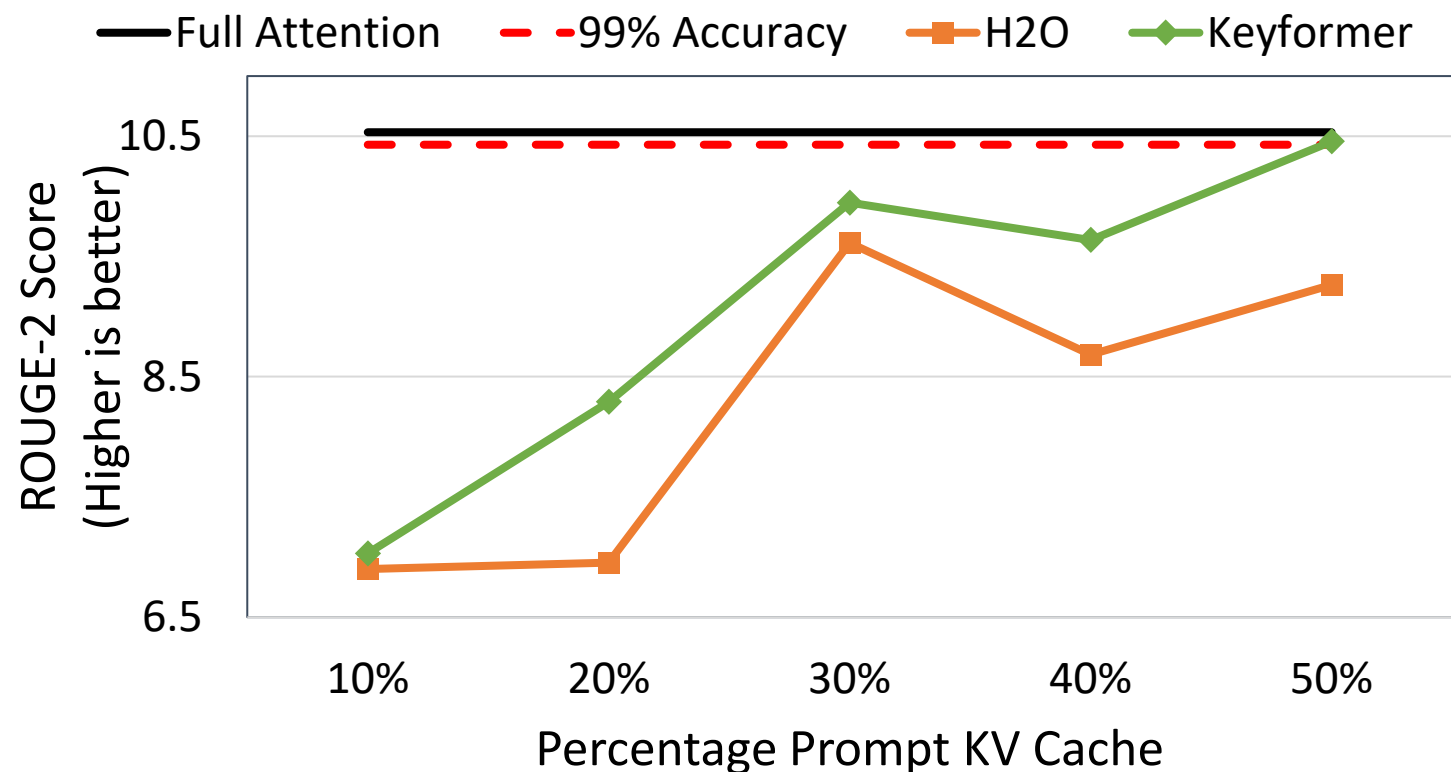
Conversation Task → **90%** of prompt KV cache



Long Context

Model : MPT-7B-storywriter (65k seq len)

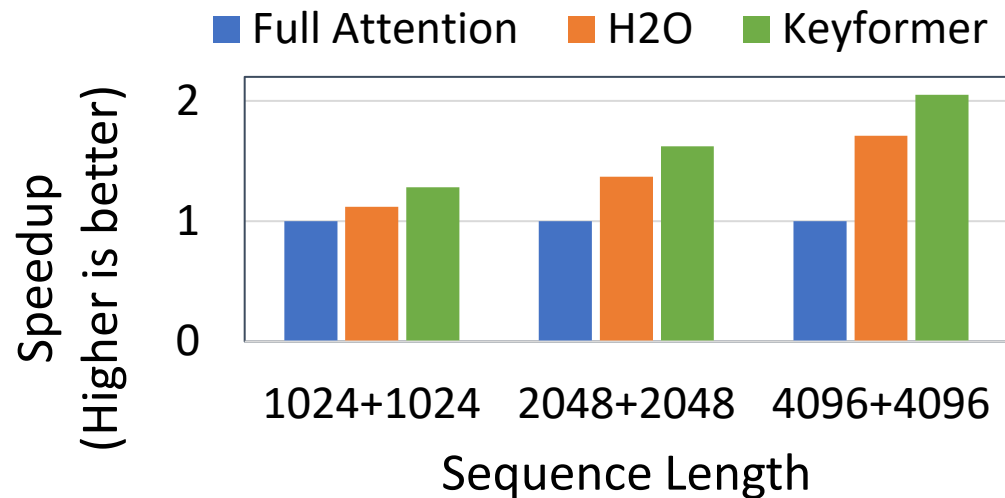
Dataset : Gov Reports (Mean Context: 9k)



Long Context

Model : MPT-7B-storywriter (65k seq len)

Dataset : Gov Reports (Mean Context: 9k)



Generation Throughput (tokens/sec)

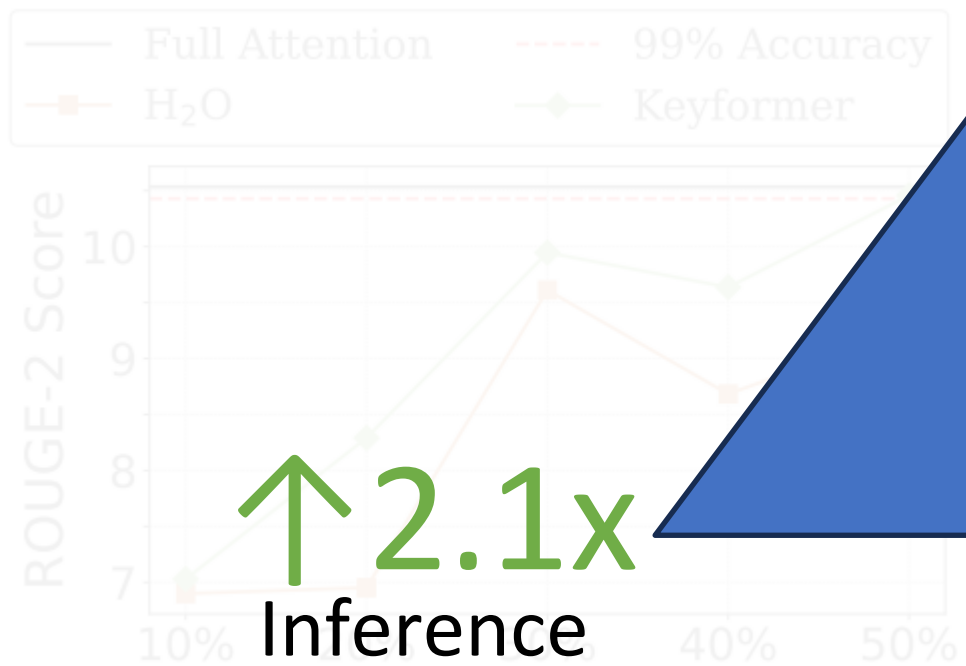
Sequence Length	Full Attention	H ₂ O	Keyformer
	Original cache	90% KV cache	50% KV cache
1024 + 1024	24.9	27.8	32.0
2048 + 2048	15.0	20.5	24.3
4096 + 4096 (BS=1)	8.3	14.1	17.0
4096 + 4096 (BS=2)	OOM	OOM	19.85

Long Context

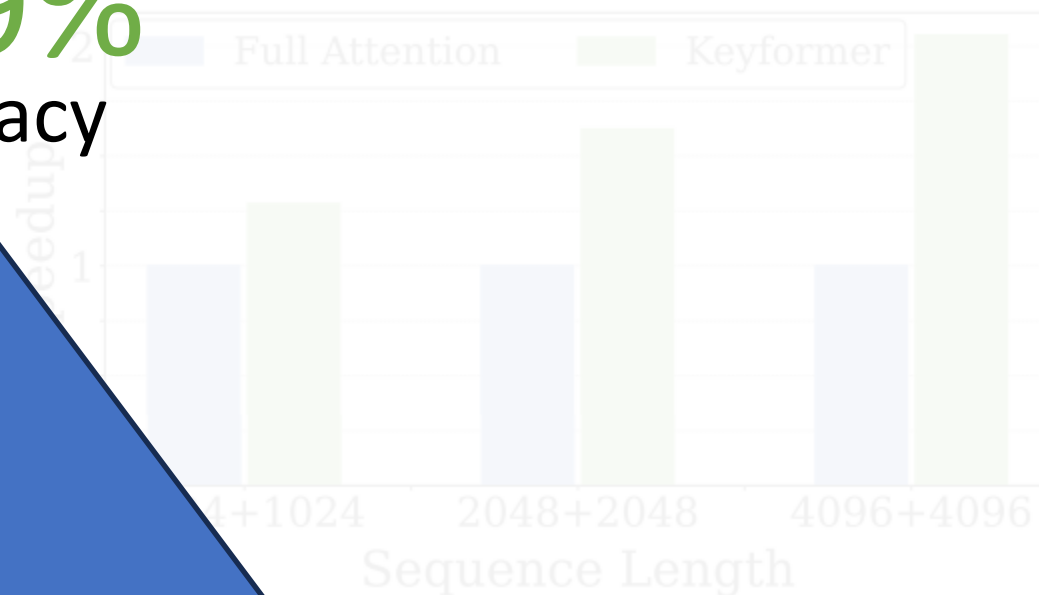
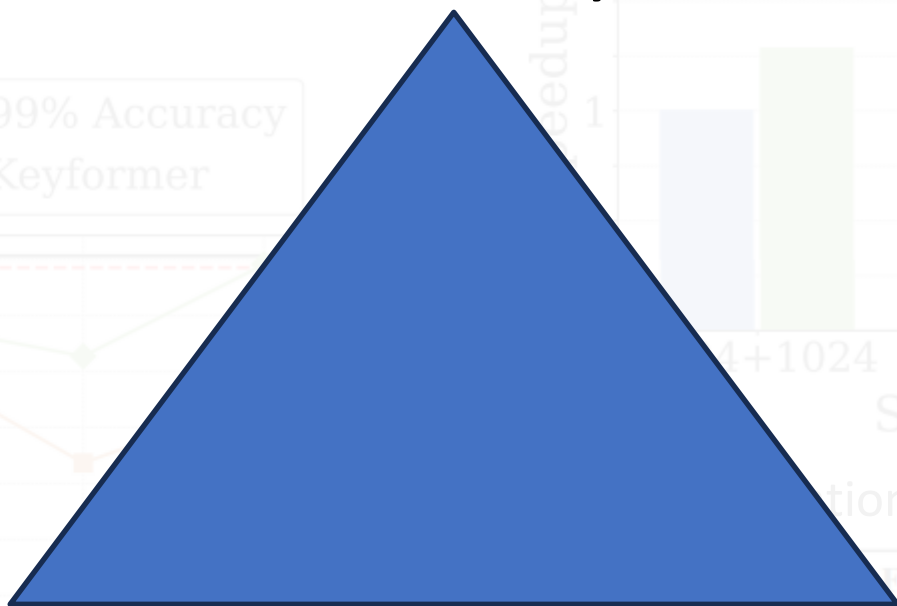
Model : MPT-storywriter (65k seq len)

Dataset : Gov Reports (Mean Context 20k)

> 99% Accuracy



↑ 2.1x Inference Speedup



↑ 2.4x Token Generation Throughput

Sequence Length	Full Attention	Keyformer (50% KV cache)
1024 + 1024	24.9	32.0
2048 + 2048	15.0	24.3
4096 + 4096 (BS=1)	8.3	17.0
4096 + 4096 (BS=2)	8.3	19.85

Conclusion

- LLMs → Inherent sparsity within attention
- Sparsity → KV cache reduction
- Token Discarding → Effect on attention score distribution
- Keyformer → Regularization based Key token Identification

> 99%
Accuracy

↑ 2.1x
Inference
Speedup

↑ 2.4x
Token
Generation
Throughput

Questions

