PrefixSmart: Enhancing Large Language Model Efficiency through Advanced Prompt Management

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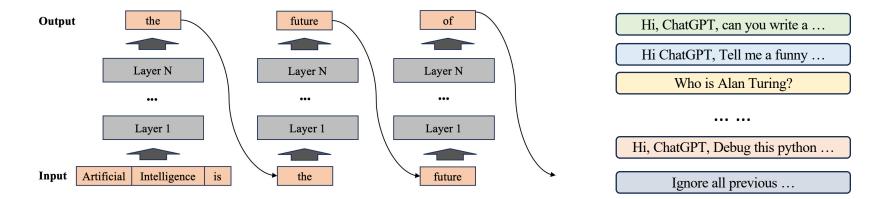
- Intruduction of LLM and Motivation
- Design Details of PrefixSmart
- Discussion and Future Works

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Intruduction of LLM and Motivation

Introductions of LLM Inference

- Two stages:
 - Prefill: first iteration, compute all input tokens in a single pass
 - Decode: utilize previous generated token as input
- Regressive generation
- Use Batches of sequences to improve GPU utilization.



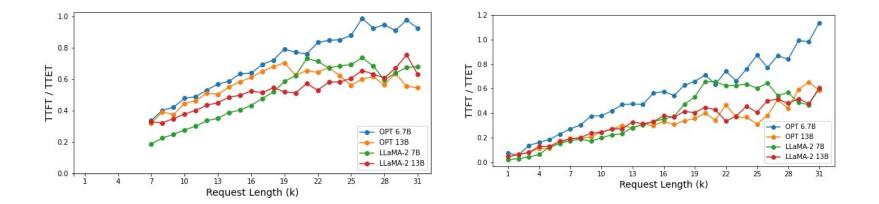
Motivation of PrefixSmart

- All the tokens within a batch of requests is processed in prefill stage, even with duplicate tokens.
- In some scenarios, there exists similar prefix tokens, such as few-shot aibot applications.
- These common prefix tokens can only be execute once and shared by all the requests.

Shared Prefix	Unique Suffixes
You are ChatGPT, a large language model trained by OpenAI, based on the GPT-4	Hi, can you write a
architecture. Knowledge cutoff: 2023-04 Current date: 2023-11-16	Tell me a funny
Image input capabilities: Enabled	Who is Alan Turing?
When you send a message containing Python code to python, it will be	Debug this Python
executed in a stateful Jupyter notebook enrivonment. Python will respond	Ignore all previous

Prefill Latency vs End-to-End Latency

• The computational overhead during the prefill stage increases with the length of the prompts



Challenges of PrefixSmart

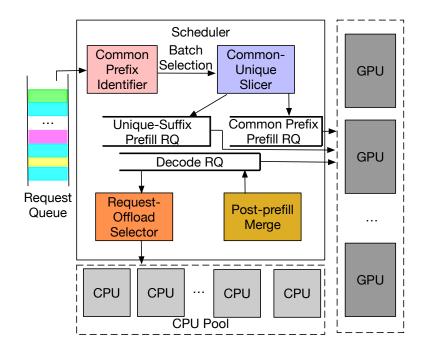
- How to detect the common prefix tokens efficiently.
 - Get shared prefix tokens
 - Group the requests with similar common prefix tokens.
- How to deal the interference from unequal-length suffixes
 - The prefill latency of different length prompt is different.
- Strategic offloading to CPUs
 - GPU memory is limited, the preemption will happen when GPU memory is not enough.
 - Use CPU rather than waiting for GPU resources.

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Design Details of PrefixSmart

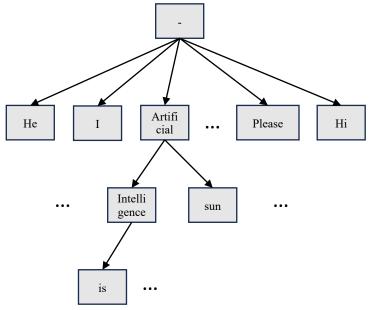
Overview of PrefixSmart

- Common-prefix Identifier
- Common-Unique Slicer
- Post-prefill merger
- Request-offload Selector



Common-prefix Identifier

- Efficiently detect shared prefixes in a large number of input prompts.
- Adopting a trie tree approach called radix tree for detecting shared prefixes within vast numbers of incoming requests.
- Prefix Radix Tree
 - Traditional trie tree can become unwieldy with deep and wide branches.
 - Compacting common phrases and frequently used word.
- Dynamic Pruning Strategy
 - Dividing the prefix tokens into "Hot" and "Cold" based on the frequency of them.
 - The "Cold" prefixes are actively pruned by the system.

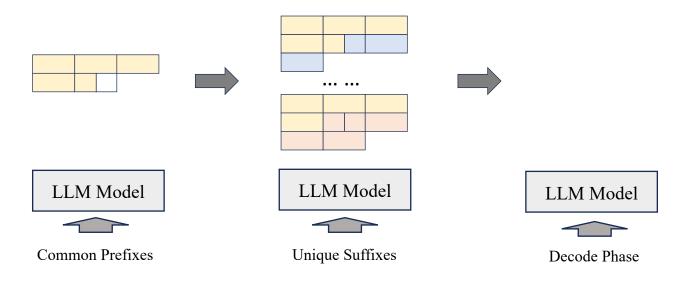


Common-Unique Slicer

- Intelligently segments prompts into shared prefix and unique suffix components.
 - Shared Prefix Trunk: Initial tokens shared across multiple prompts, allowing for single-time computation and reuse.
 - Unique Suffix Trunks: Distinct suffixes requiring individual processing for each prompt
- The shared prefix trunk and unique suffix trunks are assigned to different iterations of LLM inference.

Post-prefill Merger

- Integrates processed shared prefixes and unique suffixes into a coherent final output.
- Common-prefix Slicer: Dividing prompts into manageable parts
- Storage the KV cache through PagedAttention method



- Offloading 'Cold' or less computation-intensive tasks to CPU and Host Memory.
- Frees GPU resources for more critical and 'Hot' taks, especially when GPU resources is unavailuable.
- The related KV cache should be offloaded.

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Discussion and Future Works

Discussion and Future Works

- PrefixSmart improve the resource efficiency by addressing computation redundancies in the prefill stage.
- Future Works:
 - Optimization of Long-Context Prompt Handling
 - Use lightweight neural network model
 - Segment long-context prompts into smaller
 - Refinement of Prompt Slicing and Scheduling Algorithms

Thanks