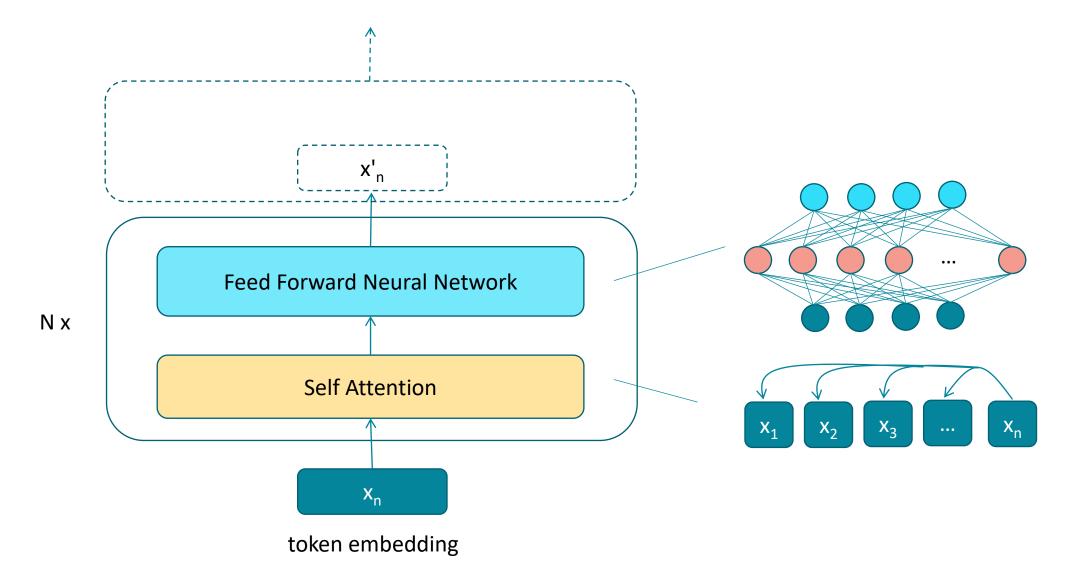
Adaptive Inference in Pre-trained LLMs Layer-skipping

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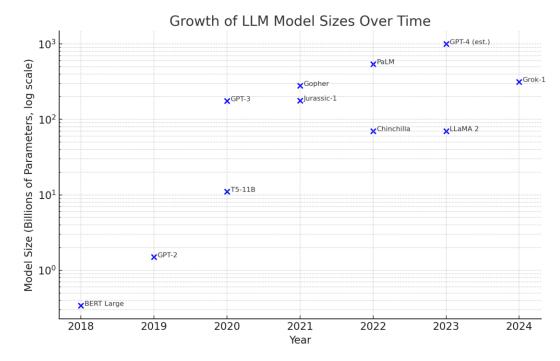
Transformer



Q: All tokens are generated with the same effort

No matter what the task is!

- A simple string copy
- Arithmetic addition
- Question answering
- Translation
- Summary
- Or any other



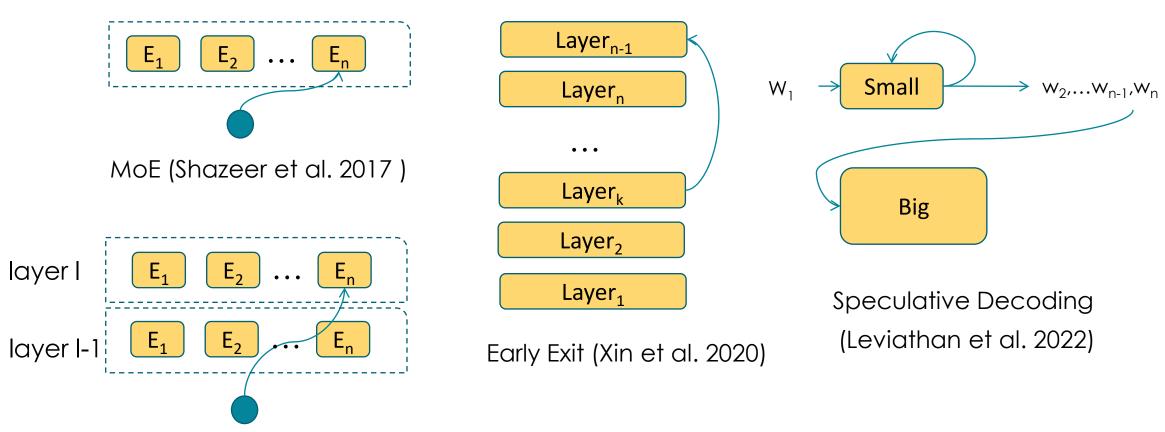
Generated by ChatGPT

No matter what the tokens are,

Every token requires the same effort!

A lot of Efforts to Make the Model Efficient

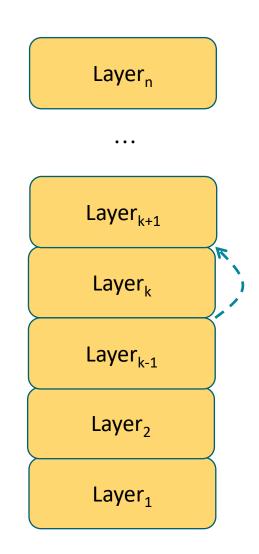
MoD (Raposo et al. 2024)



Prompt: Let's think step by step

Chain of Thoughts (Wei, et al. 2022)

Transformer: Some Layers Can be Skipped Dynamically



If we can skip one layer, we can skip another one!

Layer-skipping: # of Layers needed

[Context] The grey seal was found at Cruden Bay on Tuesday. The Scottish SPCA removed the seal from beach and took it to the charity's rescue centre...

Thankfully there were no injuries in this case and we were able to release him relatively quickly.

[Prompt] Where the grey seal was found?

The grey seal was found at Cruden Bay.

[Prompt] Please summarize this paragraph into a single sentence:

A grey seal was found tangled in netting at Cruden Bay, rescued by the Scottish S PCA, and released back into the water with no injuries.

[Prompt] Please calculate the sum of the 8 numbers in the list: [99, 45, 12, 78, 33, 66, 21, 54]

To calculate the sum of all 8 numbers in the list, I'll add each number one by one:

99 + 45 = 144

144 + 12 = 156

156 + 78 = 234

234 + 33 = 267

267 + 66 = 333

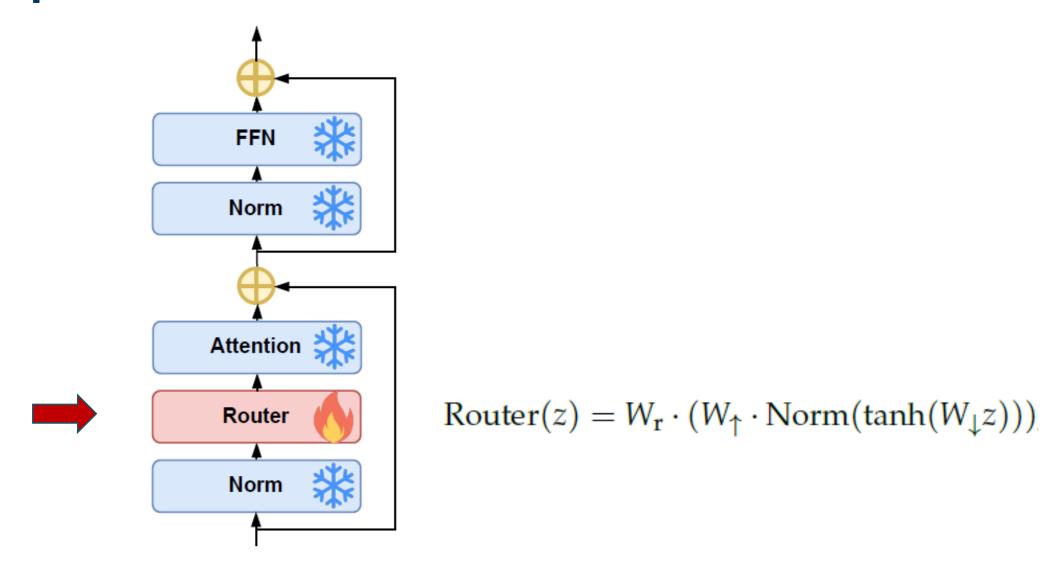
333 + 21 = 354

354 + 54 = 408

The sum of all 8 numbers in the list is 408.

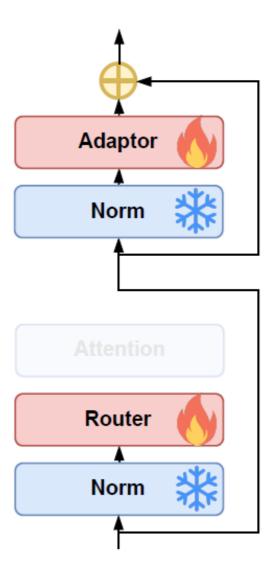
Layer-skipping patterns (Llama-3-8B-Instruct) for a language task (left) and a math task (right). The light-to-dark blue gradient represents layer usage from 16 to 32.

FlexiDepth: Router



FlexiDepth: Adaptor

Adaptor is not needed if the model is fully retrained.



Loss Function

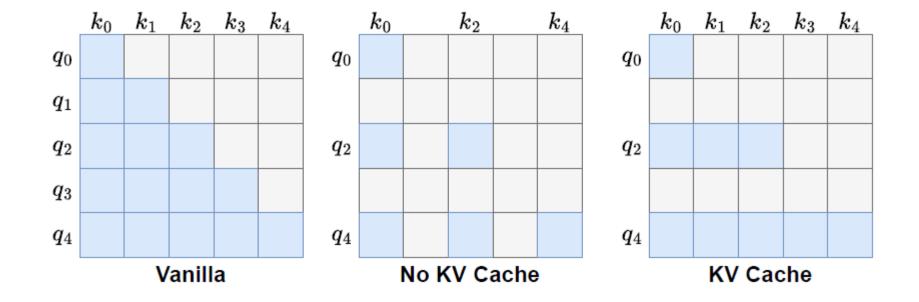
$$\mathcal{L}_{skip} = \frac{1}{T} \sum_{t=1}^{T} \left(\sum_{l=1}^{L} g_t^l \right)^2$$
 # of layers kept

where g_t^l denotes the gating score for layer l at time step t.

total loss:

$$\mathcal{L} = \alpha \cdot \mathcal{L}_{skip} + \mathcal{L}_{lm}$$

Filling K-V Cache



Average # of Layers for Text Operations

Randomly sample 100 paragraphs from the XSum test set (Narayan et al., 2018), which is a collection of news articles. We evaluated FlexiDepth on three subtasks for each paragraph

Please copy this paragraph
Please summarize this paragraph into a single sentence
Please continue writing this paragraph

The number of layers used per token:

Copy: 22.0 layers (variance 6.05)

Summarization: 28.7 layers (variance 18.3)

Continuation: 30.3 layers (variance 12.6)

Average # of Layers for Math Calculation

Generate a list of 5–10 integers between 10 and 99. Three subtasks:

```
Please repeat the following list for 5 times: [...]

Please calculate the sum of numbers in the following list: [...]

Please calculate the product of numbers in the following list: [...]
```

The number of layers used per token:

Copy: 20.1 layers (variance 8.08)

Addition: 22.5 layers (variance 14.7)

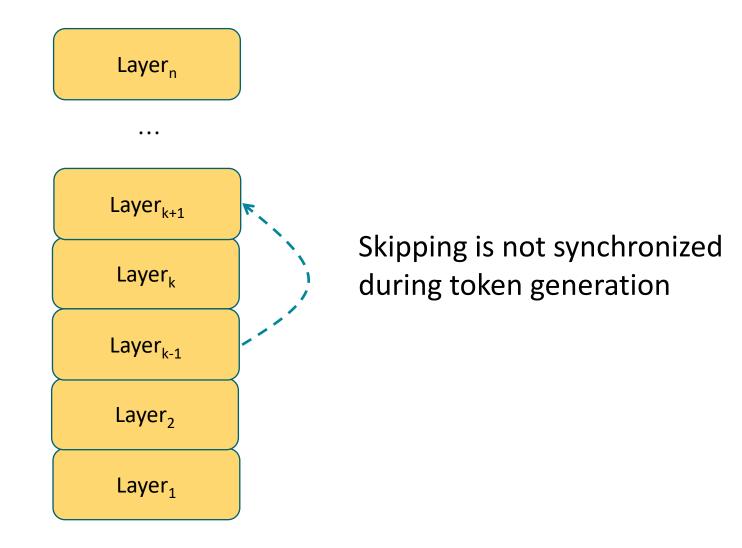
Multiplication: 23.9 layers (variance 21.6)

Method Comparison

Methods	Sing	gle-Token Ge	eneration	Multi-Token Generation			Retain %				
	MMLU	Hellaswag	Winogrande	GSM8K	HumanEval	CoQA					
Vanilla	0.673	0.706	0.744	0.679	0.299	0.784	100.0%				
Skip 4 Layers											
LayerSkip	0.659	0.636	0.676	0.004	0.0	0.350	54.0%				
ShortGPT	0.664	0.662	0.700	0.536	0.092	0.145	69.1%				
LaCo	0.671	0.693	0.724	0.581	0.031	0.778	81.7%				
MindSkip	0.664	0.698	0.722	0.378	0.189	0.720	84.2%				
Ours	0.663	0.724	0.756	0.695	0.390	0.810	106.5%				
Skip 8 Layers											
LayerSkip	0.650	0.525	0.640	0.0	0.0	0.049	43.9%				
ShortGPT	0.307	0.462	0.597	0.001	0.0	0.005	32.0%				
LaCo	0.656	0.628	0.695	0.065	0.006	0.707	65.3%				
MindSkip	0.602	0.650	0.646	0.039	0.024	0.620	60.2%				
Ours	0.616	0.705	0.735	0.662	0.341	0.801	100.7%				

Llama-3-8B-Instruct, which consists of 32 layers.

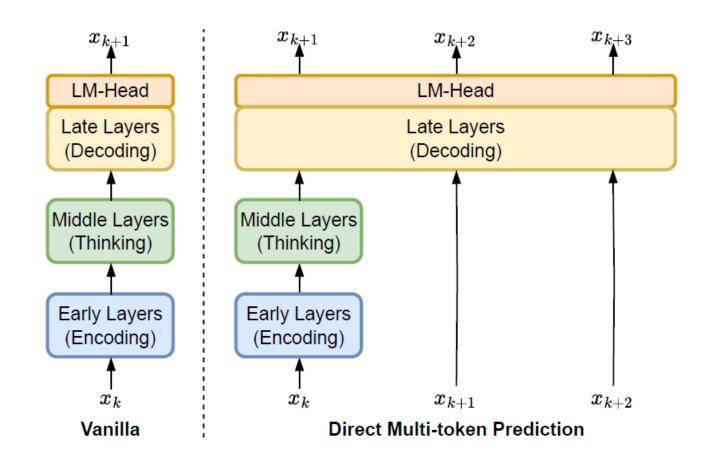
Underutilized Computation: Skipping Layers



Another Perspective: Direct Multiple Token Decoding

Can those skipped layers be used to compute future tokens?

Goal: Does not require verification, unlike speculative decoding.



Results on Qwen3-4B

	ARC-E	ARC-C	WinoGrande	GSM8K	CoQA	Overall
Vanilla	0.934	0.922	0.657	0.907	0.805	100%
MTD2	0.930	0.897	0.701	0.901	0.798	100.0%
MTD3	0.921	0.886	0.673	0.889	0.780	98.4%
MTD4	0.916	0.881	0.652	0.866	0.749	96.3%
MTD6	0.872	0.801	0.601	0.500	0.672	82.1%

- Achieved by fine-tuning Qwen3-4B with 1.5B tokens
- Utilized the last 8 of the 36 layers for predicting the second and subsequent tokens
- Up to 2x speedup for MTD4 (likely more speedup if fully retrained)

Links

Project Website/Model/Code/Data

 We seek collaboration with additional compute and training data to extend these ideas, including MoE.

Thanks!

