# Parallel Continuous Chain-of-Thought with Jacobi Iteration

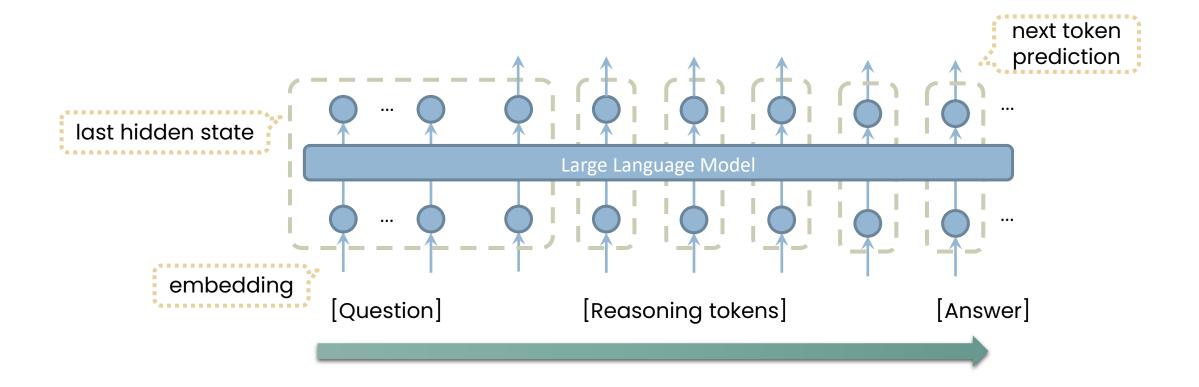
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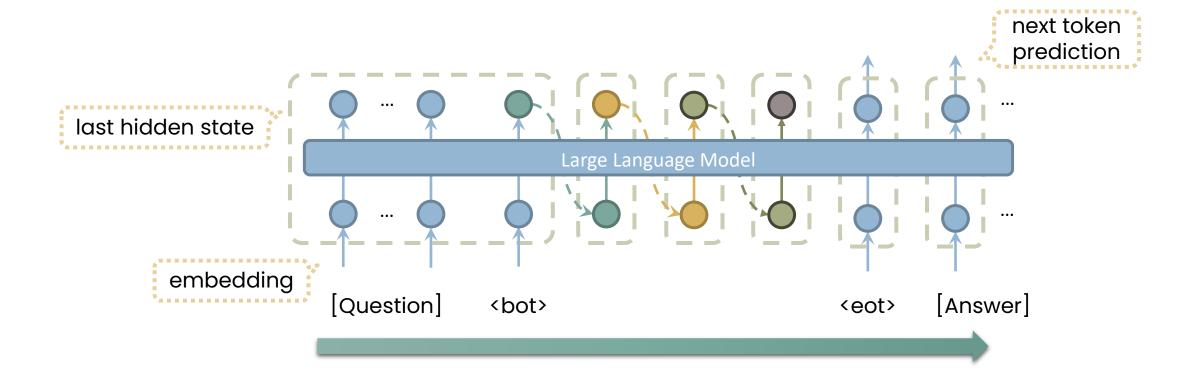
Shanghai Engineering Research Center of Intelligent Vision and Imaging



## Chain-of-Thought

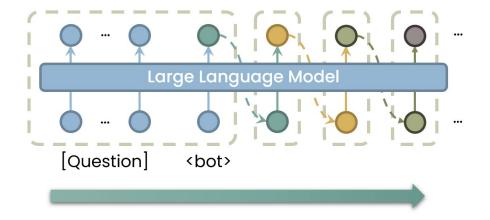


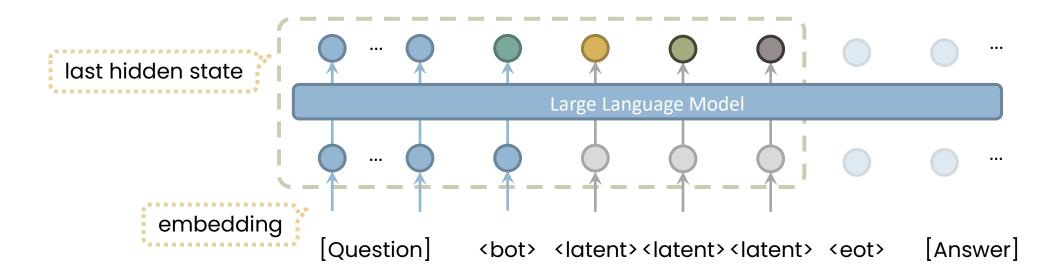
## Continuous Chain-of-Thought

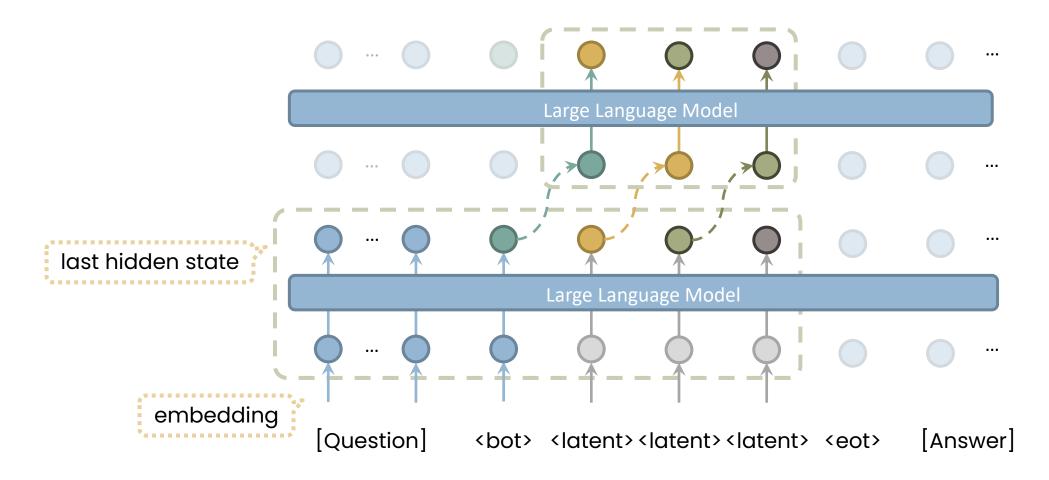


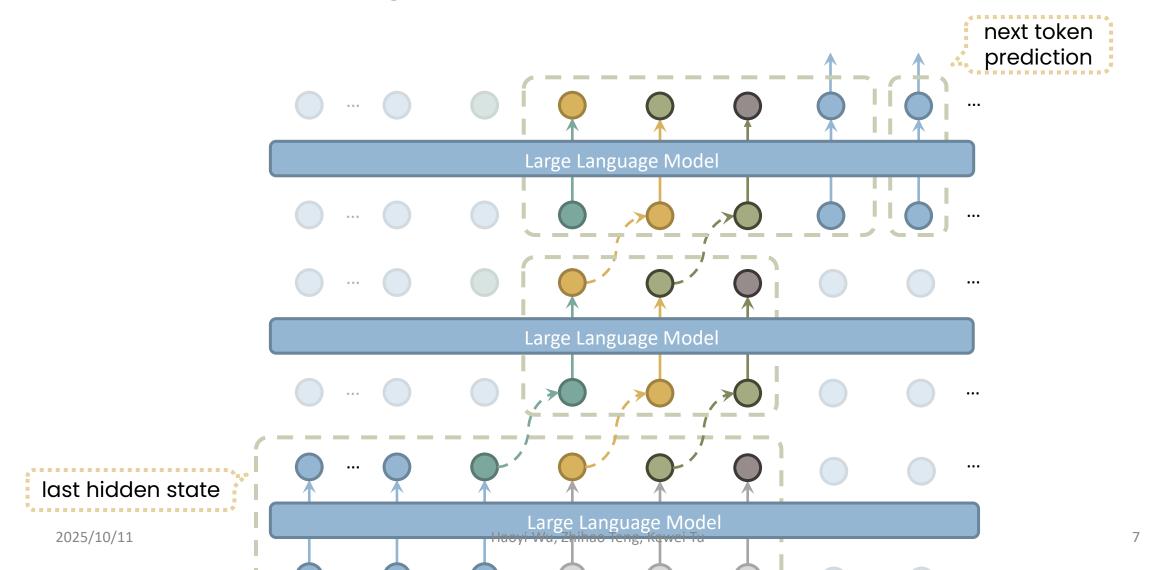
## Continuous Chain-of-Thought

- by performing implicit reasoning with continuous vectors...
  - ...it saves reasoning tokens
  - ...but the sequential dependency spoils parallel training

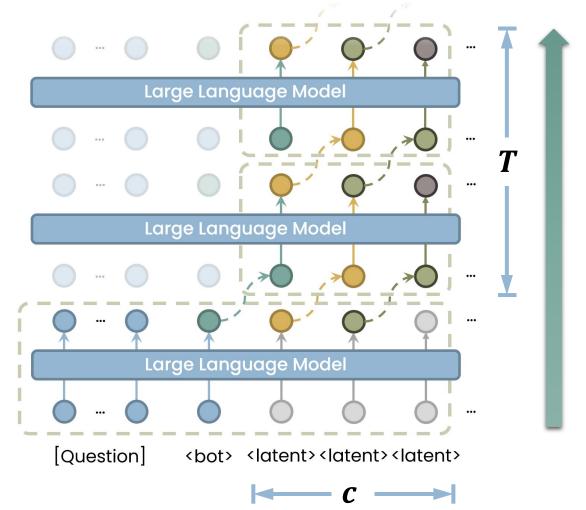








- PCCoT performs non linear Jacobi iteration on latent thought tokens
- Two hyperparameters:
  - *c*: # of latent thought tokens;
  - T: # of extra iterations.

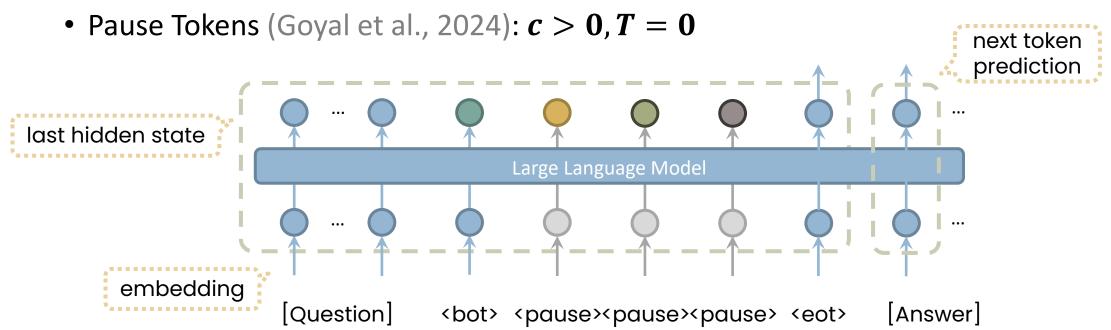


## Relations to Existing Approaches

- with different settings of c and T, PCCoT can be reduced to these existing approaches:
  - Implicit CoT (iCoT) (Deng et al., 2024): c = 0

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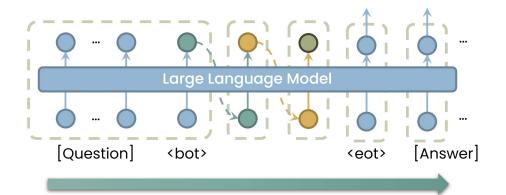
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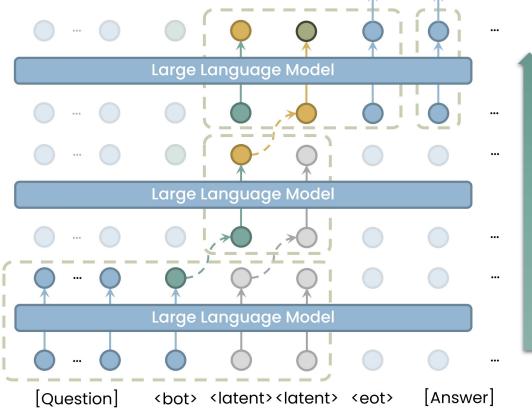
existing approaches:

• Implicit CoT (iCoT): c = 0

• Pause Tokens: c > 0, T = 0

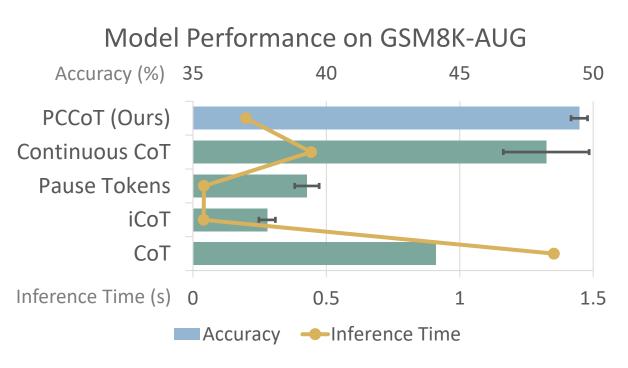
• Continuous CoT (Hao et al., 2024):  $T \ge c > 0$ 





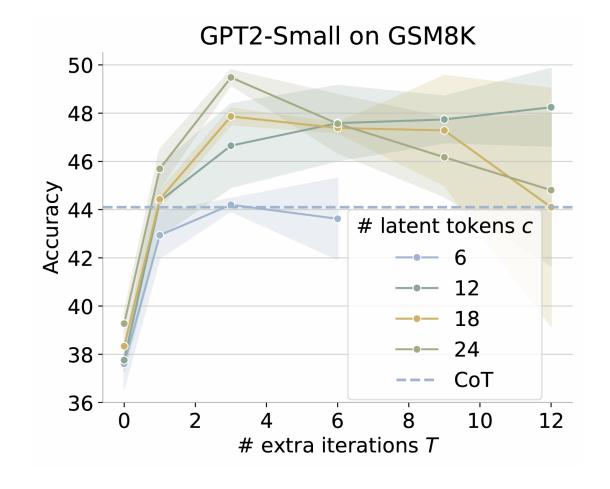
Parallel Continuous Chain-of-Thought with Jacobi Iteration

 Proper choice of c and T speeds up the reasoning process by a large scale without sacrificing the performance.



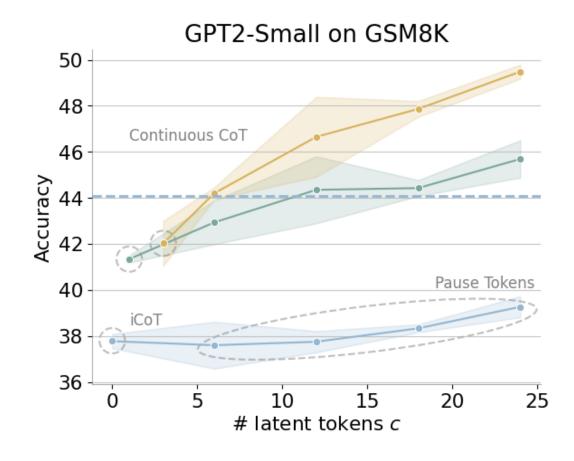
<sup>\*</sup> Experiment on GPT-2 Small, finetuned with LoRA on GSM8K-AUG. The results are averaged over 3 random runs with standard deviations as error bars. The inference time is measured with a batch size of 100. The training runs on 2 H800 GPUs and the inference runs on 1 A6000 GPU. Model training follows that of CODI (Shen et al. 2025).

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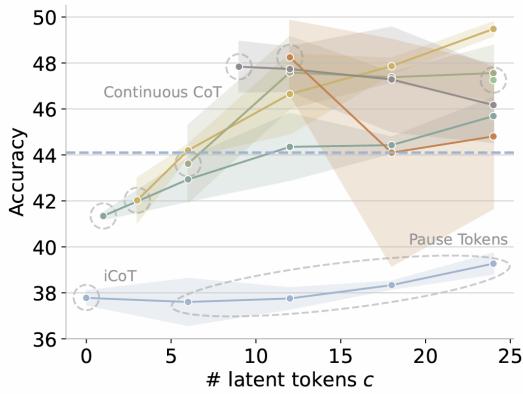




- Proper choice of c and T speeds up the reasoning process by a large scale without sacrificing the performance.
  - Large *T* ⇒ performance improvement
  - Large  $T \Rightarrow$  large standard deviation
  - Large  $c \Rightarrow$  stable superior performance if T is small

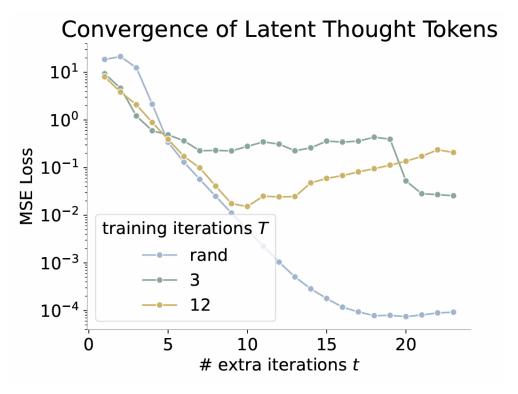






## **Analysis**

- Surprisingly, the latent thought tokens do not converge
  - Despite the convergence guarantee when  $T \geq c$



<sup>\*</sup> MSE of the latent thought tokens before and after the t th extra iteration. "rand" means the model is randomly initialized. Other models are trained with c=24 and different T. The model is tested on random samples from the test set of GSM8K.

## Thanks

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github.com/whyNLP/PCCoT



faculty.sist.shanghaitech.edu.cn/faculty/tukw/

