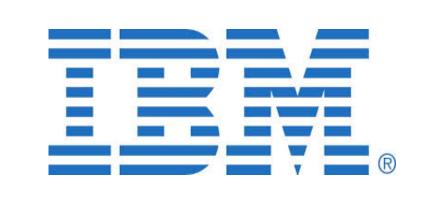


# Rensselaer Theoretical and Algorithmic Foundations of In-Context Learning and reasoning Using Properly Trained Transformer Models



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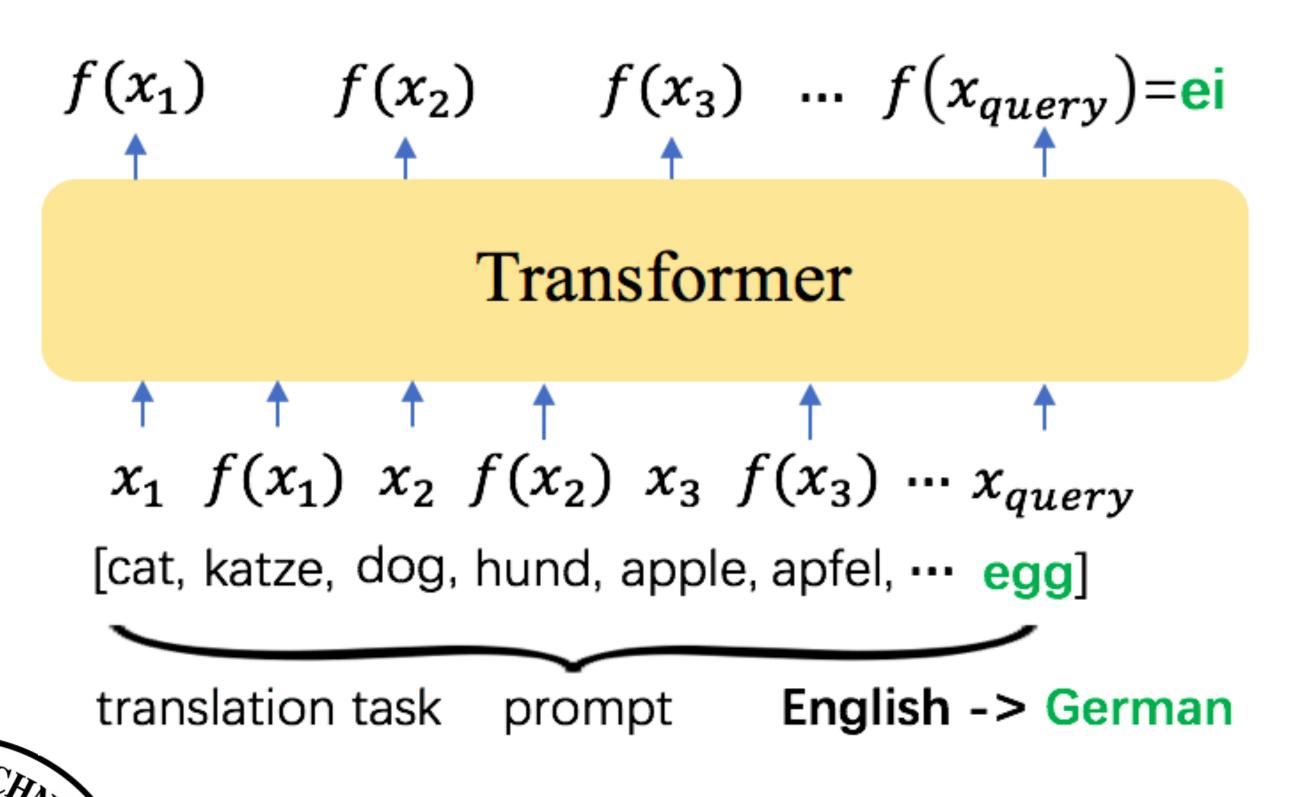




# Motivation

Transformer-based foundation models, e.g., GPT-4, Sora, have achieved great empirical success in many areas.

- Large foundation models are able to implement in-context learning (ICL) and reasoning.
- Theoretical understanding of how a Transformer can be trained to perform ICL and generalize in and out of domain successfully and efficiently is less investigated.

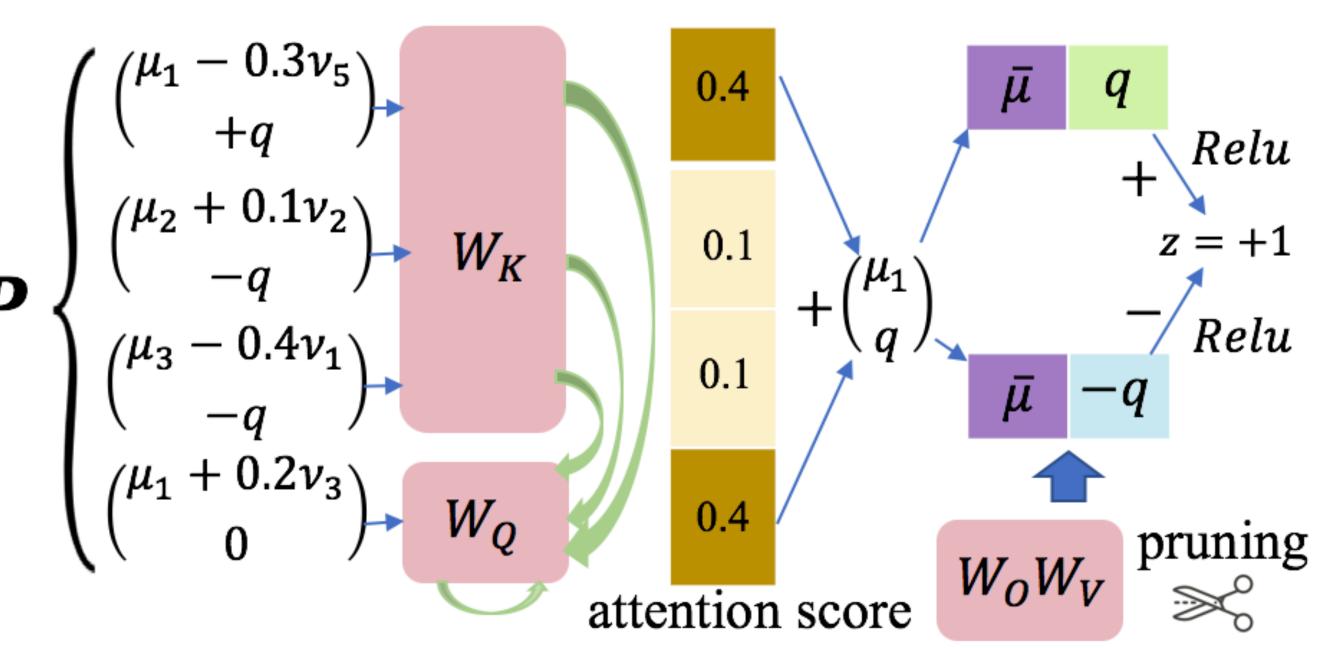


# **Current Progress**

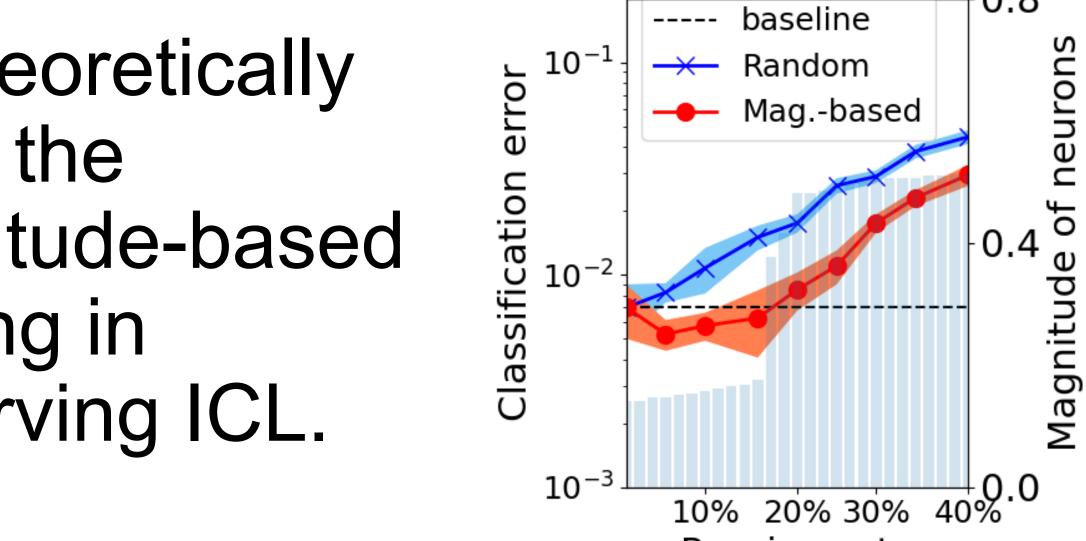
We provide a theoretical characterization of how to train nonlinear Transformers to enhance their ICL capability on classification tasks.

Theorem 1 (informal): Given enough neurons and a large batch, and prompt lengths inverse in the fraction of relevant tokens  $\alpha$ , then after training with  $\Theta(\alpha^{-1})$  steps,

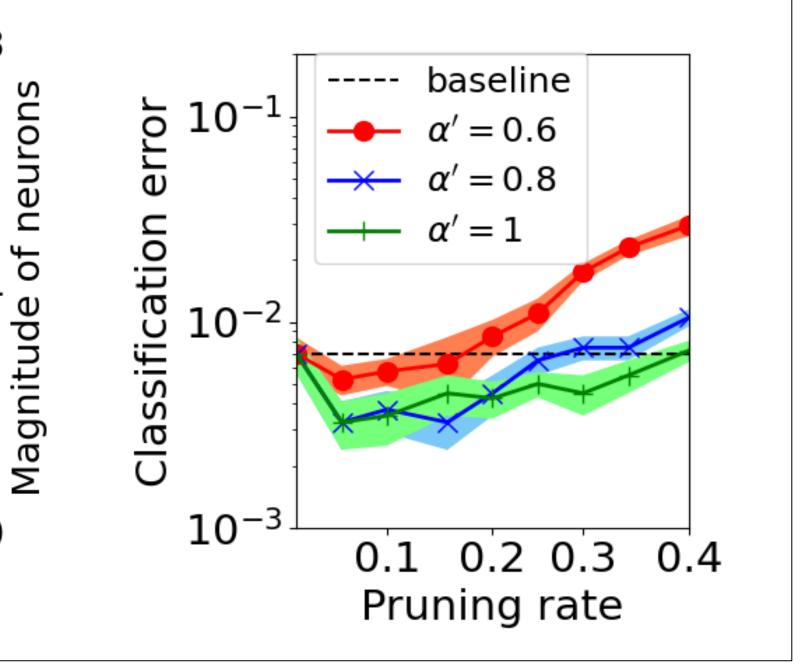
- the returned one-layer Transformer model achieves an in-domain generalization error no larger than  $\epsilon$ .
- If the testing relevant patterns are linear combinations of the trained ones with coefficient summation no larger than 1, the out-of-domain generalization error is no larger than  $\epsilon$ .
- We expand the theoretical understanding of the mechanism of the ICL capability of Transformers.
- We theoretically justify the Magnitude-based Pruning in preserving ICL.



<u>5</u> 10<sup>−1</sup>



Pruning rate



 $\alpha' = 1.0$ , id

 $\alpha' = 0.9$ , id

 $\alpha' = 0.8$ , id

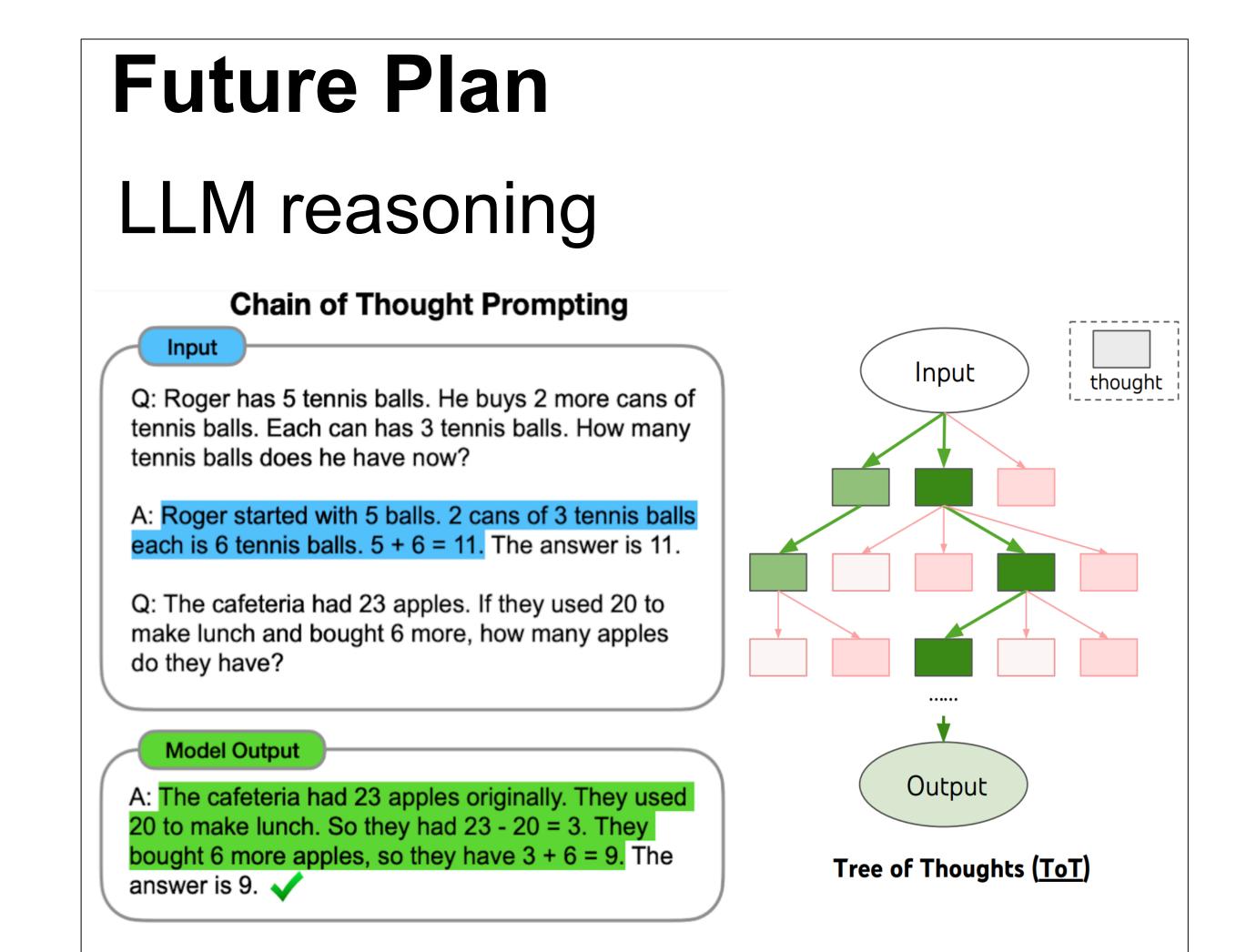
 $\leftarrow$   $\alpha' = 1.0$ , ood

 $\alpha' = 0.8$ , ood

8 12 16 20

Context length

 $\alpha' = 0.9$ , ood



#### Problems to solve

- How can a Transformer be trained to learn different hidden causal structure?
- Why does adding intermediate steps help the reasoning in theory?
- What is the mechanism of a Transformer implementing reasoning in context?

#### Theoretical contributions

- Hidden Markov chain modeling.
- Next token prediction beyond classification and regression.

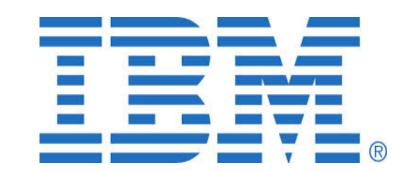
## Experiments

Evaluate the results on the arithmetic reasoning dataset GSM8K and the commonsense reasoning dataset CSQA.



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#### Math Word Problems (free response)

Q: Roger has 5 tennis balls. He buys 2 more cans of tennis balls. Each can has 3 tennis balls. How many tennis balls does he have now?

A: Roger started with 5 balls. 2 cans of 3 tennis balls each is 6 tennis balls. 5 + 6 = 11. The answer is 11.



Education Assistant

### StrategyQA

Q: Yes or no: Would a pear sink in water?

A: The density of a pear is about 0.6 g/cm<sup>3</sup>, which is less than water. Thus, a pear would float. So the answer is no.



Smart Navigation

