# **Chain-of-thought Prompting**

# Large Language Models (LLMs)

#### Learning goals

- illustrate chain-of-thought and point out the benefits it brings to LLMs
- illustrate tree-of-thought and point out the benefits it brings to LLMs

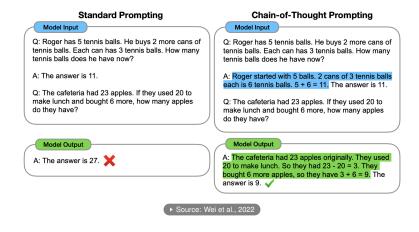
### **CHAIN-OF-THOUGH MOTIVATION**

How to boost the reasoning capabilities of LLMs? • Wei et al., 2021

- Use formal approaches, e.g., logic, symbolic reasoning
  - Example: BeliefBank
  - Difficult to train and deploy, not widely used
- Few-shot learning via prompting works for many tasks
  - Still, it works poorly on tasks that require reasoning
- Chain of thought (CoT) prompting
  - Prompts in the form <input, chain of thought, output>
  - CoT: series of intermediate steps that lead to a final output

### CHAIN-OF-THOUGHT PROMPTING PARADIGM

CoT enables LLMs to tackle complex arithmetic, commonsense and symbolic reasoning tasks.



### CHAIN-OF-THOUGHT PROMPTING PARADIGM

#### Chain-of-Thought Prompting

#### **Model Input**

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.

Q: The cafeteria had 23 apples. If they used 20 to make lunch and bought 6 more, how many apples do they have?

#### **Model Output**

A: The cafeteria had 23 apples originally. They used 20 to make lunch. So they had 23 - 20 = 3. They bought 6 more apples, so they have 3 + 6 = 9. The answer is 9.

#### **BENEFITS OF CHAIN-OF-THOUGHT**

- Decompose multi-step problems and thus allocate more compute to problems requiring more reasoning steps
- By describing the reasoning, interpretability is increased. It provides the possibility to observe where reasoning went wrong
- It is closer to how humans solve tasks using language
- By designing a prompt, existing large language models are able to perform chain-of-thought reasoning.

### **EXAMPLES FOR CHAIN OF THOUGHT**

Examples of <input, chain of thought, output> triples for commonsense and symbolic reasoning

► Source: Wei et al., 2022

### StrategyQA

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

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

## **Date Understanding**

Q: The concert was scheduled to be on 06/01/1943, but was delayed by one day to today. What is the date 10 days ago in MM/DD/YYYY?

A: One day after 06/01/1943 is 06/02/1943, so today is 06/02/1943. 10 days before today is 05/23/1943. So the answer is 05/23/1943.

## **Sports Understanding**

Q: Is the following sentence plausible? "Joao Moutinho caught the screen pass in the NFC championship."

A: Joao Moutinho is a soccer player.
The NFC championship is part of
American football, not soccer. So the
answer is no.

## SayCan (Instructing a robot)

Human: How would you bring me something that isn't a fruit?

Explanation: the user wants something to eat that isn't a fruit. An energy bar is not a fruit, so I will bring the user an energy bar.

Plan: 1. find(energy bar) 2.

pick(energy bar) 3. find(user) 4.

put(energy bar) 5. done().

#### Last Letter Concatenation

Q: Take the last letters of the words in "Lady Gaga" and concatenate them.

A: The last letter of "Lady" is "y". The last letter of "Gaga" is "a".

Concatenating them is "ya". So the answer is ya.

### Coin Flip (state tracking)

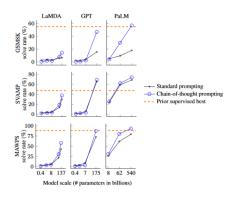
Q: A coin is heads up. Maybelle flips the coin. Shalonda does not flip the coin. Is the coin still heads up?

A: The coin was flipped by Maybelle. So the coin was flipped 1 time, which is an odd number. The coin started heads up, so after an odd number of flips, it will be tails up. So the answer is no.

### **COT IMPROVES ARITHMETICS**

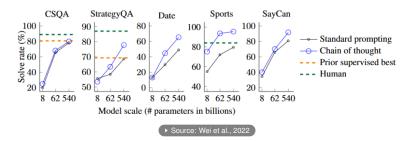
SVAMP: math word problems with varying structures; MAWPS: repository unifying math problems from different sources;

➤ Source: Wei et al., 2022



#### **COT IMPROVES COMMONSENSE**

CSQA: Contains around 200K dialogs with a total of 1.6M turns. Further, unlike existing large scale QA datasets which contain simple questions that can be answered from a single tuple, the questions in the dialogs require a larger subgraph of the KG.

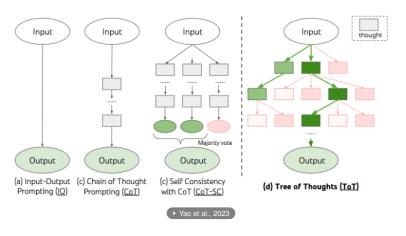


#### TREE-OF-THOUGHT: MOTIVATION

- The token-level and left-to-right decisions of the autoregressive mechanism pose a limitation for:
  - Tasks where initial decisions play a pivotal role
  - Tasks requiring exploration or strategic lookahead
- Potential strategy to solve those:
  - Maintain and explore diverse alternatives instead of just picking one
  - Evaluates current status and looks ahead or backtrack to make global decisions

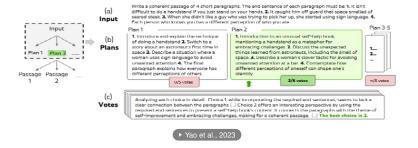
#### TREE-OF-THOUGHT: PROMPTING PARADIGM

Schematic illustrating various approaches to problem solving with LLMs. Each rectangle box represents a *thought*, a coherent language sequence serving as an intermediate step toward problem solving.



#### TREE-OF-THOUGHT FOR CREATIVE WRITING

A step of deliberate search in a randomly picked Creative Writing task. Given the input, the LM samples five different plans, and then votes five times to decide which plan is best.

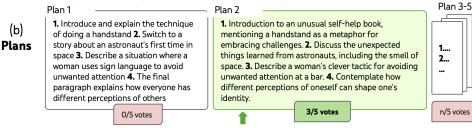


## TREE-OF-THOUGHT FOR CREATIVE WRITING (2)

(a) Input

(b)

Write a coherent passage of 4 short paragraphs. The end sentence of each paragraph must be: 1. It isn't difficult to do a handstand if you just stand on your hands. 2. It caught him off guard that space smelled of seared steak. 3. When she didn't like a guy who was trying to pick her up, she started using sign language. 4. Each person who knows you has a different perception of who you are.



(c) Votes

Analyzing each choice in detail: Choice 1, while incorporating the required end sentences, seems to lack a clear connection between the paragraphs {...} Choice 2 offers an interesting perspective by using the required end sentences to present a self-help book's content. It connects the paragraphs with the theme of self-improvement and embracing challenges, making for a coherent passage. {...} The best choice is 2.