Large Language Models Chain of Thought Prompting

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Many of slides have been adapted from Z. Ding & Z. Zhand's slides, Princeton cos597G, Fall 2022

Chain of Thought Prompting for Large Language Model Reasoning



Hard Language Tasks: Reasoning

Tasks that require multiple steps of reasoning to solve

Arithmetic Reasoning (AR)

Question: If there are 3 cars in the parking lot and 2 more cars arrive, how many cars are in the

Answer: The answer is **5**.

Symbolic Reasoning (SR) **Question**: Take the last letters of the words in "Elon Musk" and concatenate them **Answer:** The answer is **nk**.

Commonsense Reasoning (CR)

Question: What home entertainment equipment requires cable? Answer Choices: (a) radio shack (b) substation (c) television (d) cabinet

Answer: The answer is (c).





- Multi-step reasoning is often seen as a weakness in NLP models
- There is former research on reasoning in small language models through fully supervised finetuning on specific datasets. However,
 - Creating a dataset containing explicit reasoning can be difficult and time-consuming
 - training on a specific dataset limits application to a specific domain
- Reasoning ability may emerge in language models at a certain scale, such as models with over 100 billion parameters (Wei et al., TMLR 2022)

Reasoning

Huang and Chang, "Towards Reasoning in Large Language Models: A Survey", 2023



Fine-tune GPT-3 on GSM8K (arithmetic):



Conjecture: to achieve > 80%, needs 100 times more fine-tuning data for 175B model

Cobbe et al., "Training verifiers to solve math word problems", arXiv 2021





GSM8K (arithmetic):

- Finetuned GPT-3 175B
- Prior best
 - 540B LM: standard prompting
 - 540B LM: chain of thought prompting



Cobbe et al., "Training verifiers to solve math word problems", arXiv 2021

Few-shot standard prompting with even larger model (PaLM 540B) also does not work well.



Reasoning (SR) tasks

- Scaling up language model size does not efficiently achieve high performances,
- for Arithmetic Reasoning (AR), CommonSense Reasoning (CR) and Symbolic



Reasoning (SR) tasks

Proposed solution: Chain of Thought (CoT) prompting

- Scaling up language model size does not efficiently achieve high performances,
- for Arithmetic Reasoning (AR), CommonSense Reasoning (CR) and Symbolic



Chain of Thought Prompting



Chain-of-Thought Prompting Elicits Reasoning in Large Language Models

Few-Shot CoT

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Chain-of-Thought Prompting Elicits Reasoning in Large Language Models

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Large Language Models are Zero-Shot Reasoners

Shixiang Shane Gu Google Research, Brain Team

Yutaka Matsuo The University of Tokyo

Yusuke Iwasawa The University of Tokyo

Both papers will appear in NeurIPS'22!





Definition:

A chain of thought is a series of intermediate natural language reasoning steps that lead to the final output.



Definition:

A chain of thought is **a series of intermediate natural language reasoning steps** that lead to the final output.

(input, output) demonstrations are replaced with (input, chain of thought, output)



Compositionaly of Language

- Compositionality of the languages
 - \bullet known concepts
- Problem decomposition can help
 - Decompose multi-step reasoning into intermediate steps

Compositional Out-of-Distribution generalization: ability to understand novel composition of



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Definition:

A chain of thought is a series of intermediate natural language reasoning steps that lead to the final output.

use <input, intermediate results, output> triples

Benefits:

- Decomposition -> easier intermediate problems
- Interpretable
- More general than neural symbolic computing
- Leveraging prompting of LLM



(a) Few-shot

Question: 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?

Answer: The answer is **11**.

Question: A juggler can juggle 16 balls. Half of the balls are golf balls, and half of the golf balls are blue. How many blue golf balls are there?

Answer:

(Output) The answer is 8. X



(Wei et al., 2022)

(b) Few-shot-CoT

Question: 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?

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

Question: A juggler can juggle 16 balls. Half of the balls are golf balls, and half of the golf balls are blue. How many blue golf balls are there? **Answer**:

(Output) The juggler can juggle 16 balls. Half of the balls are golf balls. So there are 16 / 2 = 8 golf balls. Half of the golf balls are blue. So there are 8 /2 = 4 blue golf balls. The answer is 4. \checkmark







Zero-shot

Question: A juggler can juggle 16 balls. Half of the balls are golf balls, and half of the golf balls are blue. How many blue golf balls are there?

Answer: The answer (arabic numerals) is

(Output) 8 🗡

Kojima et al.,"Large Language Models are Zero-Shot Reasoners", NeurIPS 2022

Zero-shot-CoT

Question: A juggler can juggle 16 balls. Half of the balls are golf balls, and half of the golf balls are blue. How many blue golf balls are there?

Answer: Let's think step by step.

(Output) (Output) There are 16 balls in total. Half of the balls are golf balls. That means that there are 8 golf balls. Half of the golf balls are blue. That means that there are 4 blue golf balls.

> Two-stage Prompting Step-by-step Answer









Zero-Shot Chain of Thought (CoT)

For zero-shot CoT, a **two-stage** prompting is applied:



KoJima et al.,"Large Language Models are Zero-Shot Reasoners", NeurIPS 2022





Experiments



Pre-trained LLMs:

- Instruct GPT-3 (ada 350M, babbage 1.3B, curie 6.7B, and davinci 175B) (Ouyang et al., 2022)
- **PaLM** (8B, 62B, 540B) (Chowdhery et al., 2022)
- LaMDA (422M, 2B, 8B, 68B, 137B) (Thoppilan et al., 2022) \bullet
 - Dialogue-oriented LM. \bullet
 - Fine-tuned on human-annotated data.

Models



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- **PaLM** (8B, 62B, 540B) (Chowdhery et al., 2022)
- LaMDA (422M, 2B, 8B, 68B, 137B) (Thoppilan et al., 2022)
- GPT-3 (ada 350M, babbage 1.3B, curie 6.7B, davinci 175B)
- **GPT-2** (1.5B)
- GPT-Neo (2.7B), GPT-J (6B), TO (11B) (Sanh et al., 2022), OPT (13B) (Zhang et al., 2022)

Models



Experiments **Arithmetic Reasoning**

Prompting setups: zero-shot, few-shot, few-shot CoT



Free Response - Few-Shot CoT Prompt Exemplar

Question: If there are 3 cars in the parking lot and 2 more cars arrive, how many cars are in the parking lot?

Answer: There are originally 3 cars. 2 more cars arrive. 3 + 2 = 5. The answer is **5**.



Free Response - Few-Shot CoT Prompt Exemplar

Free Response

Question: If there are 3 cars in the parking lot and 2 more cars arrive, how many cars are in the parking lot?

Answer: There are originally 3 cars. 2 more cars arrive. 3 + 2 = 5. The answer is 5.

You can have one or more equations.

Equations can be incomplete and combined math with words.

Free Response

Question: Olivia has \$23. She bought five bagels for \$3 each. How much money does she have left?

Answer: Olivia had 23 dollars. 5 bagels for 3 dollars each will be $5 \times 3 = 15$ dollars. So she has 23 - 15 dollars left. 23 - 15 is 8. The answer is 8.





Free Response - Few-Shot CoT Prompt Exemplar

Free Response

Question: If there are 3 cars in the parking lot and 2 more cars arrive, how many cars are in the parking lot?

Answer: There are originally 3 cars. 2 more cars arrive. 3 + 2 = 5. The answer is 5.

- Manually composed 8 exemplars
- All contains equations with flexible formats
- Benchmarked on:
 - **GSM8K** (Cobbe et al. 2021)
 - **SVAMP** (Patel et al., 2021)
 - MAWPS (Koncel-Kedziorski et al., 2016)



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Multiple Choice - Few-Shot CoT Prompt Exemplar

Multiple Choice

Question: A person is traveling at 20 km/hr and reached his destiny in 2.5 hr then find the distance? Answer Choices: (a) 53 km (b) 55 km (c) 52 km (d) 60 km (e) 50 km

Answer: The distance that the person traveled would have been 20 km/hr * 2.5 hrs = 50 km. The answer is (e).

GSM8K (Cobbe et al. 2021)

Multiple Choice

Question: If a / b = 3/4 and 8a + 5b =22, then find the value of a. Answer Choices: (a) 1/2 (b) 3/2 (c) 5/2 (d) 4/2 (e) 7/2

Answer: If a / b = 3/4, then b = 4a / 3. So 8a + 5(4a / 3) = 22. This simplifies to 8a + 3a = 22. 20a / 3 = 22, which means 44a / 3 = 22. So a is equal to 3/2. The answer is (b).

The exemplars have various formats





Multiple Choice - Few-Shot CoT Prompt Exemplar

Multiple Choice

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Answer: The distance that the person traveled would have been 20 km/hr * 2.5 hrs = 50 km. The answer is (e).

GSM8K (Cobbe et al. 2021)

- 4 exemplars, whose questions, intermediate reasoning, and answers are from AQuA-RAT's training set
- Exemplars have flexible formats
- Benchmarked on AQuA-RAT (Ling et al., 2017)







GSM8K

Josh decides to try flipping a house. He buys a house for \$80,000 and then puts in \$50,00 in repairs. This increased the value of the house by 150%. How much profit did he make?

SVAMP

Each pack of dvds costs 76 dollars. If there is a discount of 25 dollars on each pack. How much do you have to pay to buy each pack?









GSM8K

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Train generator





1. Fine-tuned 2 epoch on training set.

Cobbe et al., "Training verifiers to solve math word problems", arXiv 2021



2)



- 1. Fine-tuned 2 epoch on training set.
- 2. Sample 100 solutions from the generator for each training problem and label each solution as correct or incorrect.

Cobbe et al., "Training verifiers to solve math word problems", arXiv 2021

Generate and label 100 solutions/problem







- Fine-tuned 2 epoch on training set.
- Sample 100 solutions from the generator for each training 2. problem and label each solution as correct or incorrect.
- Train a verifier for a single epoch on this dataset. 3.

Cobbe et al., "Training verifiers to solve math word problems", arXiv 2021



2)







Fine-tuned 2 epoch on training set.

Cobbe et al., "Training verifiers to solve math word problems", arXiv 2021

Generate and label 100 solutions/problem



Return the one with the highest verifier score







For the easiest subset of MAWPS which only requires a single step to solve, performance improvements were either negative or very small

MAWPS - SingleEq

If there are 7 bottle caps in a box and Linda puts 7 more bottle caps inside, how many bottle caps are in the box?

MAWPS - AddSub

There were 6 roses in the vase. Mary cut some roses from her flower garden. There are now 16 roses in the vase. How many roses did she cut?











MAWPS - MultiArith

The school cafeteria ordered 42 red apples and 7 green apples for students lunches. But, if only 9 students wanted fruit, how many extra did the cafeteria end up with?

AQuA-RAT

A person is traveling at 20 km/hr and reached his destiny in 2.5 hr then find the distance? Answer Choices: (a) 53 km (b) 55 km (c) 52 km (d) 60 km (e) 50 km

Wei et al., "Chain-of-Thought Prompting Elicits Reasoning in Large Language Models", NeurIPS 2022.

Instruct GPT-3: text-davinci-002 achieves similar performance as PaLM 540B model

Arithmetic Reasoning - Observations

- Both zero-shot and few-shot CoT promptings are emergent abilities of model scale.
- Do not positively impact performance for small models
 - start to yield performance gains when used with models with more than ~100B parameters.
- Few-shot CoT achieves **better** performance on LLM than zero-shot CoT.

Experiments Symbolic Reasoning

Symbolic Reasoning - Last Letter Concatenation

Last letter concatenation

Question: Take the last letters of the words in "Elon Musk" and concatenate them

Answer: The last letter of "Elon" is "n". The last letter of "Musk" is "k". Concatenating them is "nk".

The answer is **nk**.

- Generate full names by randomly concatenating names from the top onethousand first and last names from name census data
- 4 exemplars with strict format

Symbolic Reasoning - Coin Flip

Coin Flip

Question: A coin is heads up. Tom does not flip the coin. Mike does not flip the coin. Is the coin still heads up?

Answer: The coin was flipped by no one. So the coin was flipped 0 times. The coin started heads up, and it was not flipped, so it is still heads up. So the answer is yes.

Wei et al., "Chain-of-Thought Prompting Elicits Reasoning in Large Language Models", NeurIPS 2022.

Coin Flip

Question: A coin is heads up. Jamey flips the coin. Teressa flips the coin. Is the coin still heads up?

Answer: The coin was flipped by Jamey and Teressa. So the coin was flipped 2 times, which is an even number. The coin started heads up, so after an even number of flips, it will still be heads up. So the answer is yes.

8 exemplars with **strict** format.

Symbolic Reasoning - In & Out-of-domain Test

Last letter concatenation

Question: Take the last letters of the words in "Elon Musk" and concatenate them

Answer: The last letter of "Elon" is "n". The last letter of "Musk" is "k". Concatenating them is "nk". The answer is nk.

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Wei et al., "Chain-of-Thought Prompting Elicits Reasoning in Large Language Models", NeurIPS 2022.

Coin Flip **Question**: A coin is heads up. Jamey flips the coin. Teressa flips the coin. Is the coin still heads up?

Answer: The coin was flipped by Jamey and Teressa. So the coin was flipped 2 times, which is an even number. The coin started heads up, so after an even number of flips, it will still be heads up. So the answer is yes.

In-domain test set: examples had the same number of steps as the few-shot exemplars

Out-of-domain (OOD) test set: examples had more steps than those in the exemplars.

Symbolic Reasoning - Last Letter Concatenation

Zero-Shot results use **Instruct-GPT-3 175B** text-davinci-002 model.

In-Domain

Take the last letters of the words in "Elon Musk" and concatenate them.

Out-of-Domain

Take the last letters of the words in "Johann Sebastian **Bach**["] and concatenate them.

Symbolic Reasoning - Coin Flip

Zero-Shot results use **Instruct-GPT-3 175B** text-davinci-002 model.

In-Domain

A coin is heads up. **Tom does** not flip the coin. *Mike does* not flip the coin. Is the coin still heads up?

Out-of-Domain

A coin is heads up. **Tom does** not flip the coin. Mike does not flip the coin. Jake flips the coin. Is the coin still heads up?

Symbolic Reasoning - Observations

- CoT promptings are emergent abilities of model scale
- Standard prompting fails out-of-domain tests for both tasks.
- Zero-shot CoT using Instruct-GPT-3 175B achieves the similar performance as few-shot CoT in both tasks using 540B PaLM model.

Experiments CommonSense Reasoning

Commonsense Reasoning - Toy Problems

CSQA (Talmor et al., 2019)

Question: What home entertainment equipment requires cable? Answer Choices: (a) radio shack (b) substation (c) television (d) cabinet

Answer: The answer is **(c)**.

Sport Understanding

Question: Is the following sentence plausible? "Jamel Murray was perfect from the line."

Answer: The answer is **yes**.

StrategyQA (Geva et al., 2021)

Question: Could Brooke Shields succeed at University of Pennsylvania?

Answer: The answer is **yes**.

BIG-bench (Srivastava et al., 2022)

Date Understanding

Question: 2015 is coming in 36 hours. What is the date one week from today in MM/DD/YYYY

Answer: So the answer is **01/05/2015**.

Commonsense Reasoning - Results

Commonsense Reasoning - Results

Commonsense Reasoning - Toy Problems

SayCan Robot Planning

Locations = [counter, table, user, trash, bowl].

Objects = [cup, apple, kettle chips, tea, multigrain chips, coke, lime soda, jalapeno chips, rice chips, orange, grapefruit soda, pepsi, redbull, energy bar, sponge, water].

Actions: pick(object), put(object), find(object), find(location).

Human: How would you throw away a cup?

Plan: 1. find(cup), 2. pick(cup), 3. find(trash), 4. put(cup), 5. done().

(Ahn et al.,2022)

These tasks not only require **multi-steps reasoning**, but also need **priori knowledge** to understand complex semantics.

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Commonsense Reasoning - Results

Commonsense Reasoning - Observations

- Scaling up model size improved the performance of standard prompting.
- CoT prompting made further gains
 - largest improvement for PaLM 540B.
- CoT show minimal benefits on CSQA and StrategyQA tasks
- and Strategy QA tasks, but **Zero-shot** CoT shows significant improvement for **Date** understanding task.

Few-shot achieves better performance than Zero-shot CoT on 175B GPT-3 model for CSQA

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.

Math Word Problems (multiple choice)

Q: How many keystrokes are needed to type the numbers from 1 to 500? Answer Choices: (a) 1156 (b) 1392 (c) 1480 (d) 1562 (e) 1788

A: There are 9 one-digit numbers from 1 to 9. There are 90 two-digit numbers from 10 to 99. There are 401 three-digit numbers from 100 to 500.9 + 90(2) + 401(3) = 1392. The answer is (b).

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/YYY?

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.

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.

StrategyQA

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

A: The density of a pear is about 0.6 g/cm³, which is less than water. Thus, a pear would float. 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().

Date Understanding

Last Letter Concatenation

CSQA (commonsense)

Q: Sammy wanted to go to where the people were. Where might he go? Options: (a) race track (b) populated areas (c) desert (d) apartment (e) roadblock

A: The answer must be a place with a lot of people. Race tracks, desert, apartments, and roadblocks don't have a lot of people, but populated areas do. So the answer is (b).

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.

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.

Change the types of CoT:

(b) Few-shot-CoT

Question: 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?

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

Question: A juggler can juggle 16 balls. Half of the balls are golf balls, and half of the golf balls are blue. How many blue golf balls are there?

Answer:

(Output) The juggler can juggle 16 balls. Half of the balls are golf balls. So there are 16 / 2 = 8 golf balls. Half of the golf balls are blue. So there are 8 / 2 = 4blue golf balls. The answer is $4.\checkmark$

Equation only

5+6=11. The answer is **11**.

Change the types of CoT:

(b) Few-shot-CoT

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

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Variable compute only

The answer is 11.

Change the types of CoT:

(b) Few-shot-CoT

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Change the types of CoT:

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Question: 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?

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Answer:

(Output) The juggler can juggle 16 balls. Half of the balls are golf balls. So there are 16 / 2 = 8 golf balls. Half of the golf balls are blue. So there are 8 / 2 = 4blue golf balls. The answer is $4.\checkmark$

Ablation Study: Robustness to Exemplars

Change the style of exemplar in **few-shot CoT**:

20Solve rate (%) 210 21 5

prompting, robust against linguistic styles, different exemplars.

Wei et al., "Chain-of-Thought Prompting Elicits Reasoning in Large Language Models", NeurIPS 2022.

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- Standard prompting
- Chain-of-thought prompting
- different annotator (B)
- different annotator (C)
- intentionally concise style
- exemplars from GSM8K (α)
- exemplars from GSM8K (β)
- exemplars from GSM8K (γ)

GSM8K

MAWPS

Results for **few-shot** LaMDA 137B on two **AR** tasks: have variance, but CoT still outperforms standard

Zeo-shot Ablation Study: Robustness to Trigger Sentence

Change the template (trigger sentence) in **zero-shot CoT**:

Zero-shot CoT

Question: A juggler can juggle 16 balls. Half of the balls are golf balls, and half of the golf balls are blue. How many blue golf balls are there?

Answer: Let's think step by step.

(Output) There are 16 balls in total. Half of the balls are golf balls. That means that there are 8 golf balls. Half of the golf balls are blue. That means that there are 4 blue golf balls.

Results for zero-shot GPT3 (davinci-002) 175B on MultiArith AR task: different templates encourage the model to express reasoning quite differently

No.	Template	Acc
1	Let's think step by step.	
2	First, (*1)	
3	Let's think about this logically.	
4	Let's solve this problem by splitting it into steps. (*2)	
5	Let's be realistic and think step by step.	
6	Let's think like a detective step by step.	
7	Let's think	
8	Before we dive into the answer,	
9	The answer is after the proof.	
ni i	(Zero-shot)	

Zero-shot Ablation Study: Model Size

Different model sizes in CoT prompting on MultiArith AR task:

	Original GPT-3 (0.3B / 1.3B / 6.7B / 175B)		Instruct GPT-3 (S / M / L / XL-1 / XL-2)		KL-1 / XL-2)
Zero-shot	2.0 / 1.	3 / 1.5 / 3.3	3.7 /	3.8 / 4.3 / 8.0 /	17.7
Few-shot	5.2 / 5.	2 / 4.0 / 8.1	3.0 / 2	2.2 / 4.8 / 14.0	/ 33.7
Zero-shot-CoT	1.7 / 2.2	2 / 2.3 / 19.0	2.0 / 3	3.7 / 3.3 / 47.8	/ 78.7
Few-shot-CoT	4.3 / 1.8	3 / 6.3 / 44.3	2.5 / 2	2.5 / 3.8 / 36.8	/ 93.0
	GPT-2 (1.5B)	GPT-Neo (2.7B)	GPT-J (6B)	T0 (11B)	OPT (13B)
Zero-shot	3.2	3.0	2.7	2.8	3.7
Zero-shot-CoT	2.2	1.3	2.5	3.2	2.2

- Larger model, better reasoning
- CoT is effective only for larger models
- Few-shot better than zero-shot
- Instruction tuning is important to achieve high performance from CoT

Kojima et al.,"Large Language Models are Zero-Shot Reasoners", NeurIPS 2022

More Advances - Self-Consistency

Change greedy decode (single-path) to self-consistency (multi-path) in few-shot CoT:

Showcase results on AR, CR tasks:

	Method	GSM8K	CommonsenseQA
	Previous SoTA	35 ^e / 57 ^g	91.2 ^a
LaMDA	Greedy decode (Single-path)	17.1	57.9
(137B)	Self-Consistency (Multi-path)	27.7 (+10.6)	63.1 (+5.2)
PaLM	Greedy decode (Single-path)	56.5	79.0
(540B)	Self-Consistency (Multi-path)	74.4 (+17.9)	80.7 (+1.7)

More Advances - Self-Consistency

X. Wang et al. "Self-consistency improves chain of thought reasoning in language models." arXiv 2022

More Advances - Input-Rational Ensemble

Use model-generated rationale in few-shot CoT:

X. Wang et al. "Rationale-Augmented Ensembles in Language Models" arXiv 2022

More Advances - Input-Rational Ensemble

Showcase performance for AR reasoning tasks (PaLM-540B):

Method

Standard-prompting

Few-shot CoT (Wei et al. 202

Zero-shot CoT (Kojima et al.

Self-consistency (Wang et al

Prompt-order ensemble

Input-rationale ensemble

Performance improvement on reasoning is great over previous CoT, but not significant against self-consistency,

X. Wang et al. "Rationale-Augmented Ensembles in Language Models" arXiv 2022

	GSM8K	
	17.9	
22)	56.5	
2022)	43.0	
I. 2022)	74.4	
	75.4	
	73.8	

More Advances - Self-Education (SECToR)

Self-Education: LLMs can teach themselves new skills using CoT

Zhang and Parkes, "Cain-of-Thought Reasoning is a Policy Improvement Operator", 2023

Self Training

Train on 1-7 digit addition

Generate answers for 7 digit addition using chain of-thought reasoning and self-consistency checks.

Generate answers for 8 digit addition using chain of-thought reasoning and self-consistency checks.

Train on 1-8 digit addition

All training data is selfgenerated by the model

Until self training fails

More Advances: Self-Education (SECToR)

- A supervised fine-tuning that includes training with curriculum learning is required.
- Self-education process:
 - Use CoT to generate solutions to problems that it could not otherwise solve
 - Then, the model is fine-tuned to generate these exact solutions without using CoT
 - This obtained model can now directly solve problems without using CoT
 - The self-learning process can continue accordingly

Zhang and Parkes, "Cain-of-Thought Reasoning is a Policy Improvement Operator", 2023

LLMs are still unskilled at complex reasoning

- Benchmarks may be too simple to accurately gauge the true reasoning abilities of LLMs
- "LLMs are still far from achieving acceptable performance on common planning/ reasoning tasks which pose no issues for humans to do" (Valmeekam et al. 2022)
- Early experiments on GPT-4 showed signs of limitations on reasoning tasks requiring planning and backtracking (Bubeck et al. 2023)
- Autoregressive nature of LLMs may prevent them from planning and backtracking, two abilities necessary for complex reasoning (Gendron et al., 2023)

Olestons

