

rStar: Mutual Reasoning Makes Smaller LLMs Stronger Problem-Solver

Zhenting Qi*, Mingyuan Ma*, Jiahang Xu*,
Li Lyna Zhang, Fan Yang, Mao Yang



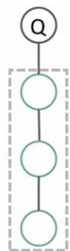
HARVARD
UNIVERSITY

Intro: Reasoning in SLMs

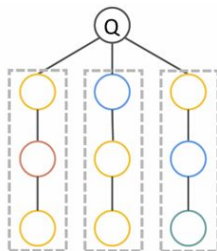
- Solving reasoning tasks is essential. However, currently only powerful models (GPT-4, openAI-o1...) demonstrate strong reasoning abilities.
- While smaller language models (SLMs) are efficient in computation and resource use, enabling faster inference, they often struggle with reasoning tasks.
- Hence, we introduce rStar, Enhancing Reasoning Capabilities of SLMs

Intro: Test-time Computation

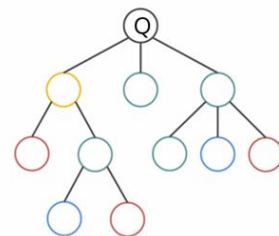
- A unified perspective: **Generator** and **Verifier**
- Generator: modify the LLMs' proposal distribution to sample diverse generations



Chain of Thought
Prompting (CoT)



Self consistency
with Cot (CoT-SC)

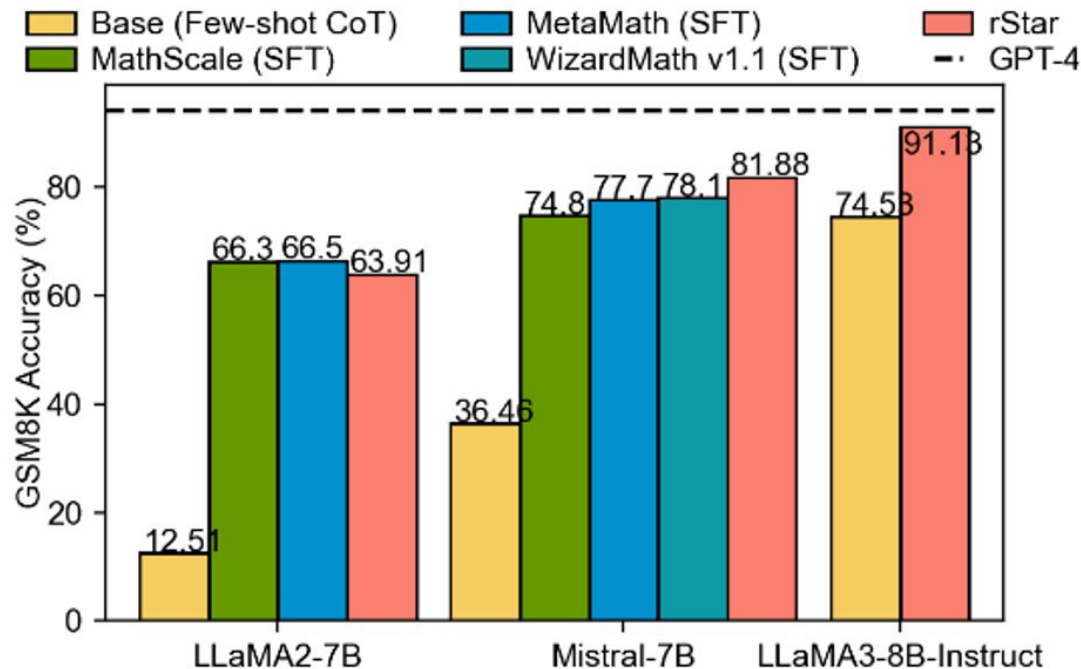


Monte Carlo Tree Search
(MCTS)

- Verifier: aggregate or select the best answer from the generated samples
 - Reward model (costly to train😓, lacks task generalization😓)
 - Self-rewarding (sometimes leads to rather random outcomes😓)

Intro: Preliminary Results

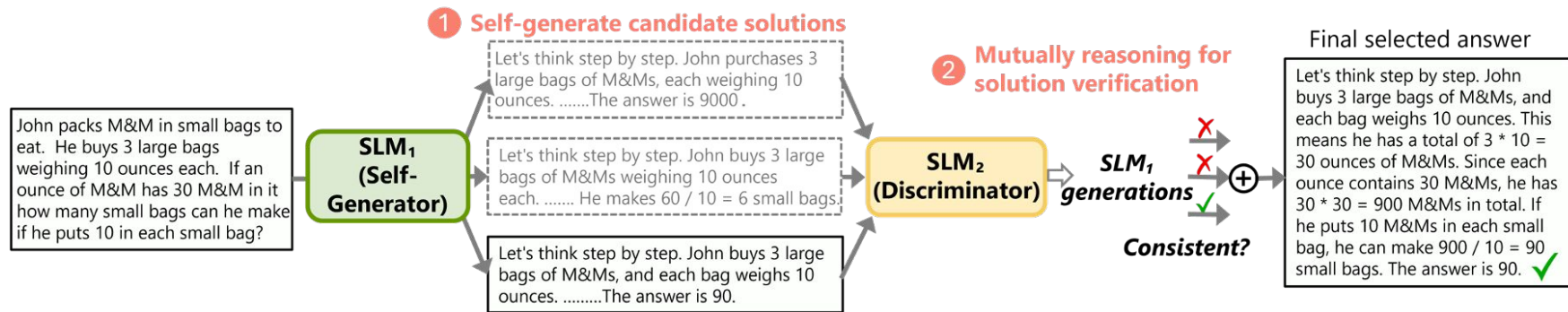
rStar makes SLMs highly capable problem-solvers, matching or even surpassing the reasoning performance achieved after domain-specialized SFT.



Intro: Insights

- **Insight 1:** Decomposing complex reasoning tasks into simpler subtasks helps SLMs handle them more effectively.
- **Insight 2:** Agreement among peers (i.e., two SLMs) on derived answers suggests a higher likelihood of correctness.

Method: Overall Framework



Solution generation:

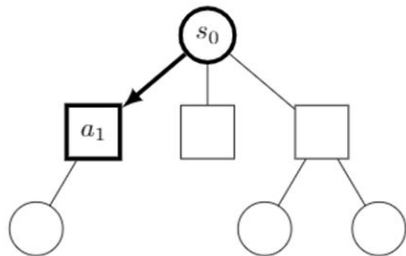
- augments the target SLM by effectively breaking down a given problem into manageable subtasks
- utilizes MCTS to generate candidate solution trajectories and allow the SLM to perform a human-like reasoning actions

Solution verification:

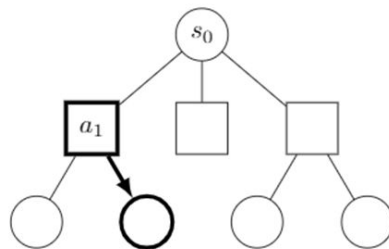
- uses another SLM to provide unsupervised feedback on each trajectory
- mutually validates the solution consistency

Method: Monte-Carlo Tree Search (MCTS)

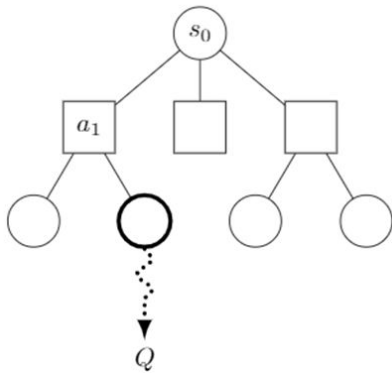
Step1: Selection



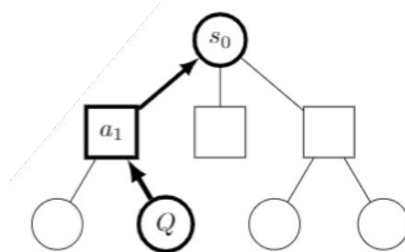
Step2: Expansion



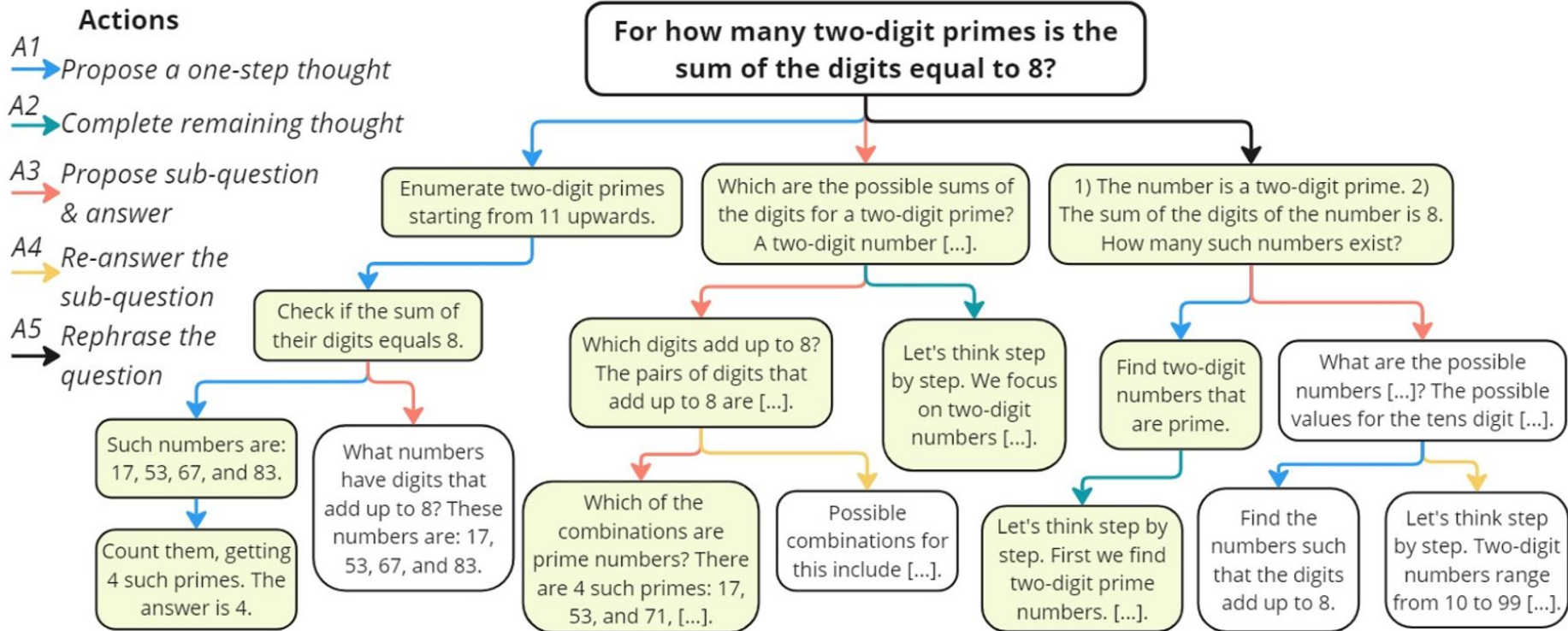
Step3: Simulation



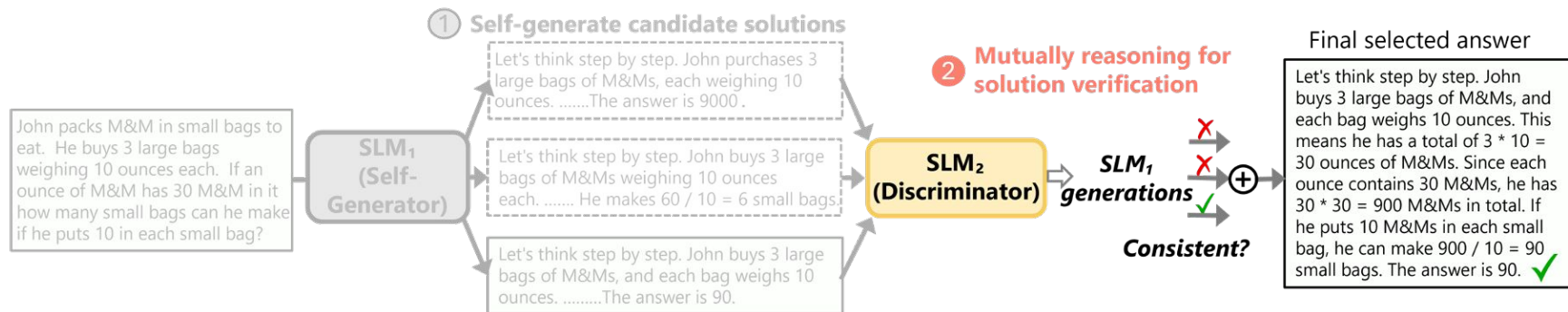
Step4: Back-propagation



Method: Enrich Generator's Action Space



Method: Discriminator with Mutual Consistency



Mutual consistency: use another SLM to act as a discriminator, providing unsupervised feedback on each candidate trajectory

- Partial reasoning steps as the **hints**: *reduce the reasoning difficulty for another SLM*
- Mirrors human experience (derived answers from the same initial steps indicates a high likelihood of correctness)

Experiments: Mathematical Tasks

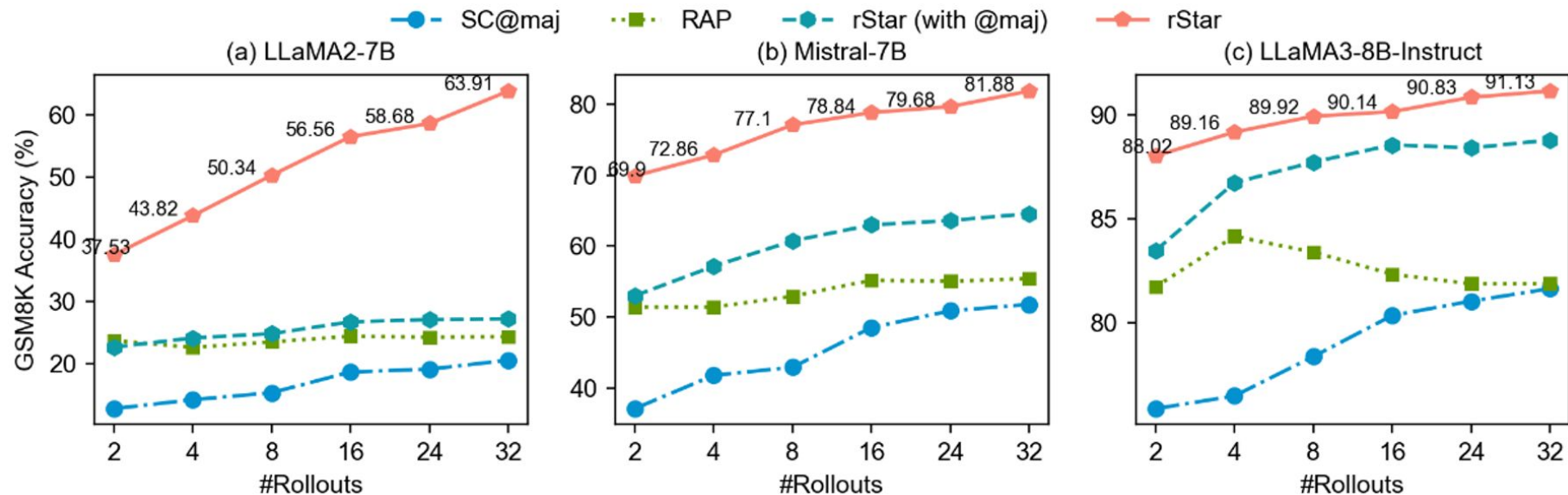
Method	LLaMA2-7B	Mistral-7B	LLaMA3-8B	LLaMA3-8B-Instruct	Phi3-mini-4k
<i>GSM8K</i>					
Zero-shot CoT	1.44	17.89	22.66	68.38	20.17
Few-shot CoT	12.51	36.46	47.23	74.53	83.45
SC@maj8	15.31	42.91	54.21	78.39	86.35
SC@maj64	20.77	52.84	64.37	83.24	88.02
SC@maj128	23.05	57.25	67.55	84.69	88.68
ToT	12.96	38.89	36.01	69.07	79.68
RAP	24.34	56.25	57.99	80.59	81.88
rStar (generator @maj)	27.22	64.59	74.38	88.70	90.44
rStar	63.91	81.88	85.52	91.13	90.67
<i>GSM-Hard</i>					
Zero-shot CoT	0.83	5.16	6.44	14.94	33.73
Few-shot CoT	3.71	13.57	13.80	25.63	40.63
SC@maj8	4.39	17.36	18.20	28.51	42.00
SC@maj64	6.52	22.59	23.73	30.33	44.80
SC@maj128	6.89	25.01	25.47	31.16	45.56
ToT	2.35	11.47	10.61	19.64	32.68
RAP	7.28	22.52	18.95	29.64	40.94
rStar (generator @maj)	8.64	29.26	26.76	33.35	46.55
rStar	18.57	37.91	32.97	37.53	46.55
<i>SVAMP</i>					
Zero-shot CoT	8.90	26.10	40.20	70.90	84.70
Few-shot CoT	48.10	72.80	76.90	89.20	92.80
SC@maj8	49.90	74.60	79.10	89.20	93.50
SC@maj64	54.10	76.70	80.70	90.50	93.30
SC@maj128	54.50	76.60	80.80	90.60	93.70
ToT	33.40	56.30	62.20	79.80	84.90
RAP	41.00	71.80	73.10	85.70	91.50
rStar (generator @maj)	60.30	83.10	86.20	91.89	93.80
rStar	74.90	86.40	90.00	94.29	94.10

Method	LLaMA3-8b-Instruct	Phi3-mini-4k
Zeroshot CoT	5.80	3.60
Fewshot CoT	17.80	32.20
SC@maj8	30.00	40.40
SC@maj64	33.00	45.20
SC@maj128	33.80	45.60
ToT	13.60	18.20
RAP	18.80	27.80
rStar (generator @maj)	38.30	48.40
rStar	42.94	48.60

Experiments: Commonsense Reasoning Tasks

Method	LLaMA2-7B	Mistral-7B	LLaMA3-8B	LLaMA3-8B-Instruct	Phi3-mini-4k
<i>StrategyQA</i>					
Zero-shot CoT	52.67	57.20	41.48	57.21	54.68
Few-shot CoT	58.82	65.65	64.05	68.41	63.61
SC@maj8	59.10	65.50	63.76	68.26	64.34
SC@maj64	58.51	63.61	63.46	67.39	62.74
SC@maj128	58.37	62.01	63.31	66.67	59.53
ToT	45.27	55.75	57.64	60.41	40.47
RAP	59.68	64.48	63.32	68.71	60.26
rStar (generator @maj)	61.57	69.43	65.50	71.47	65.50
rStar	67.25	70.31	67.69	71.57	67.25

Experiments: Test-Time Scaling



Thank you!
