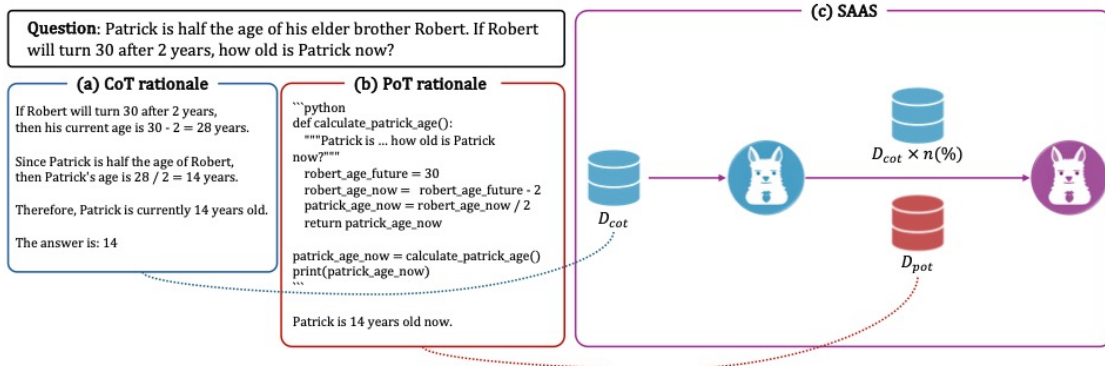


## Motivation

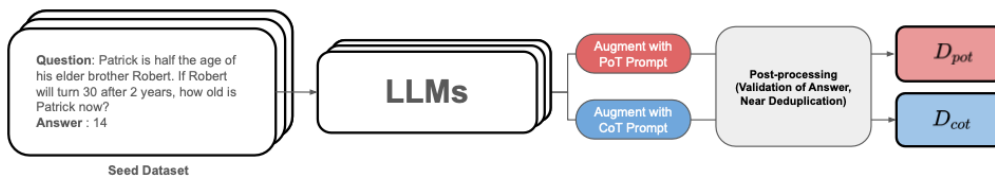
- ❖ Need to address the limitations of LLMs in mathematical reasoning to ensure robust logical thinking and problem-solving capabilities
- ❖ While Chain-of-Thought (CoT) methods enhance reasoning, they often introduce computational errors, whereas Program-of-Thought (PoT) methods improve execution but lack reasoning depth

## SAAS (Solving Ability Amplification Strategy)



- ❖ SAAS is a sequential learning strategy aimed at first enhancing logical reasoning skills through CoT learning, followed by bolstering problem-solving capabilities with PoT learning
- ❖ The cognitive retention strategy mitigates information loss by incorporating CoT rationales into the PoT phase

## Generate trainset



Seed Dataset	Rationale	Models	Size
MetaMathQA	CoT	GPT, WizardMath	465K
MATH, GSM8K	CoT	WizardMath	300K
QANDA	CoT	WizardMath	120K
MetaMathQA	PoT	ToRA	60K
MATH, GSM8K	PoT	ToRA	226K
MathInstruct	PoT	ToRA	38K
QANDA	PoT	ToRA	12K

- ❖ The data is augmented with models like GPT and validated through post-processing to ensure correctness and quality

## Experiments

Model	Size	GSM8K	MATH	GSM-Hard	SVAMP	TabMWP	ASDiv	MAWPS	Avg.
Mathematics Domain-Specific Models									
WizardMath	7B	54.9	10.7	20.6	57.3	38.1	59.1	73.7	44.9
MetaMath	7B	66.5	19.8	-	-	-	-	-	-
MuggleMATH	7B	68.4	-	-	-	-	-	-	-
Toolformer	7B	-	-	-	29.4	-	40.4	44.0	-
MathCoder	7B	64.2	23.3	-	-	-	-	-	-
MathCoder-CODE	7B	67.8	30.2	-	-	-	-	-	-
MAmmoTH	7B	53.6	31.5	-	-	-	-	-	-
MAmmoTH-CODE	7B	59.4	33.4	-	-	-	-	-	-
ToRA	7B	68.8	40.1	54.6	68.2	42.4	73.9	88.8	62.4
SAAS	7B	74.3	43.2	58.3	74.3	49.6	77.3	93.6	67.2
ToRA-CODE	7B	72.6	44.6	56.0	70.4	51.6	78.7	91.3	66.5
SAAS-CODE	7B	74.8	45.2	58.1	73.6	64.0	80.4	93.8	70.0
SAAS	10.7B	82.0	50.1	64.9	85.0	72.5	87.5	95.7	76.8
WizardMath	13B	63.9	14.0	28.4	64.3	46.7	65.8	79.7	51.8
MetaMath	13B	72.3	22.4	-	-	-	-	-	-
MuggleMATH	13B	74.0	-	-	-	-	-	-	-
MathCoder	13B	72.6	29.9	-	-	-	-	-	-
MathCoder-CODE	13B	74.1	35.9	-	-	-	-	-	-
MAmmoTH	13B	62.0	34.2	-	-	-	-	-	-
MAmmoTH-CODE	13B	64.7	36.3	-	-	-	-	-	-
ToRA	13B	72.7	43.0	57.3	72.9	47.2	77.2	91.3	65.9
SAAS	13B	76.6	46.2	61.6	77.8	58.2	80.5	94.3	70.7
ToRA-CODE	13B	75.8	48.1	60.5	75.7	65.4	81.4	92.5	71.3
SAAS-CODE	13B	79.4	50.6	61.6	80.6	68.2	84.5	95.4	74.3
MathCoder-CODE	34B	81.7	45.2	-	-	-	-	-	-
MAmmoTH-CODE	34B	72.7	43.6	-	-	-	-	-	-
ToRA-CODE	34B	80.7	50.8	63.7	80.5	70.5	84.2	93.3	74.8
SAAS-CODE	34B	82.9	52.3	64.1	82.8	73.9	85.4	95.2	76.6
SAAS-LLEMA	34B	85.4	54.7	67.0	85.2	80.2	87.6	96.6	79.5

Strategy	GSM8K	MATH
Chain-of-Thought (CoT)	69.7	26.9
Program-of-Thought (PoT)	76.8	47.7
Combination of CoT and PoT	79.0	49.2
<b>SAAS</b>	<b>79.4</b>	<b>50.6</b>
without cognitive retention strategy	79.0	49.6
Reverse SAAS	76.8	47.1
without cognitive retention strategy	69.4	27.6

- ❖ SAAS consistently and significantly outperformed all competitors across various model sizes, achieving state-of-the-art performance on reputable mathematical reasoning benchmarks
- ❖ Ablation studies demonstrated the effectiveness of SAAS's key components: the sequential learning strategy (transitioning from CoT to PoT) and the cognitive retention strategy both contributed to improved performance compared to alternative approaches

## Conclusion

- ❖ SAAS, which transitions from CoT to PoT learning with a cognitive retention strategy, achieved state-of-the-art performance on mathematical reasoning benchmarks, demonstrating the effectiveness of sequential learning in enhancing both reasoning and problem-solving abilities in LLMs.