# Do NOT Think That Much for 2+3=? On the Overthinking of Long Reasoning Models



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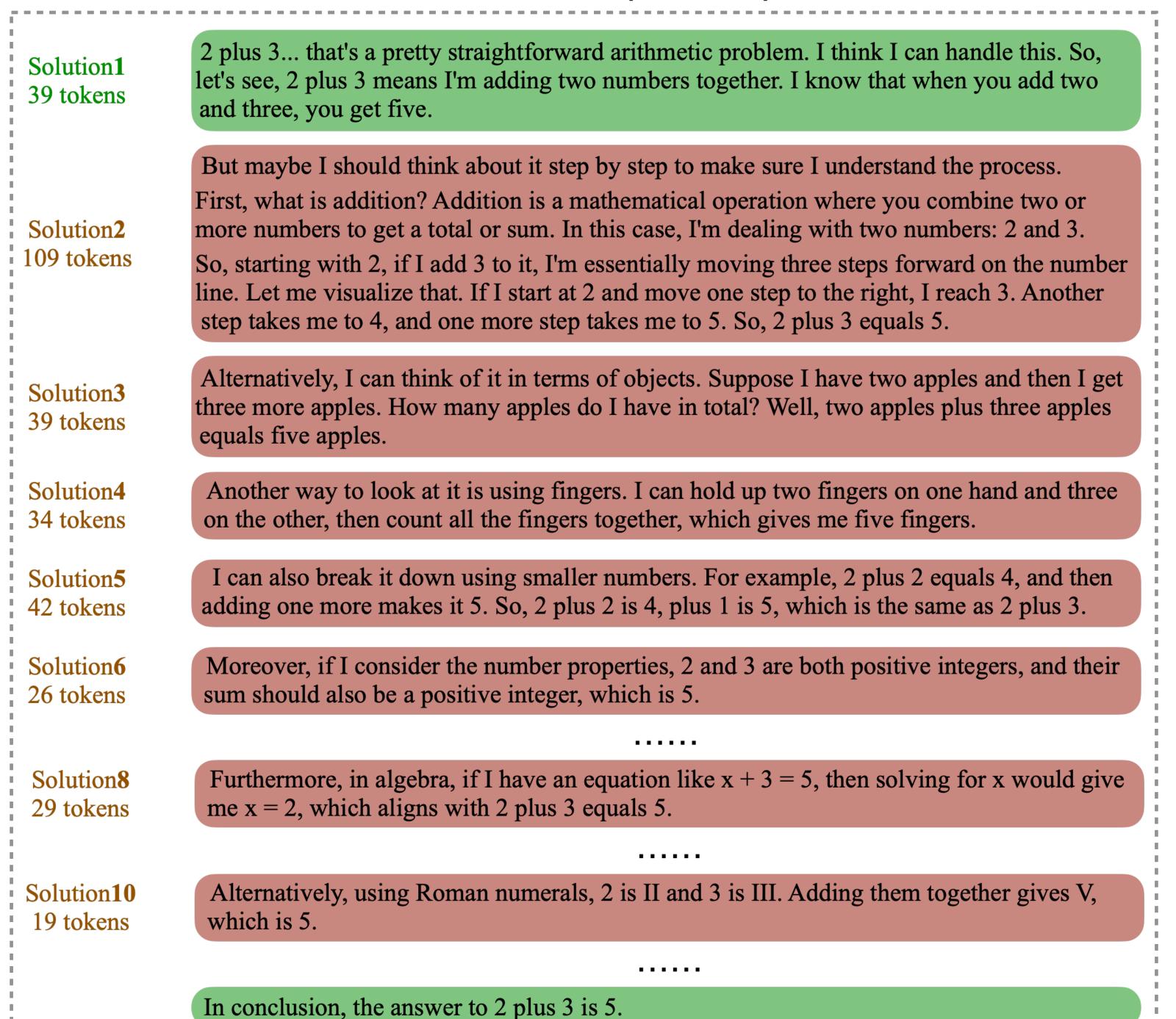
#### **OVERVIEW**

- Long Reasoning Models, such as OpenAl o1 and DeepSeek-R1, have attracted much attention for their ability to demonstrate human-like deep thinking.
- However, the reasoning process might be inefficient.
- We presents the first comprehensive study on the prevalent issue of **overthinking** in these models.
- **Overthinking**: The long reasoning model generates *redundant, homogeneous solutions* to a single question, and the subsequent solutions contribute less to the accuracy and diversity.

#### **EXAMPLE**

Question: What's the answer of 2+3?

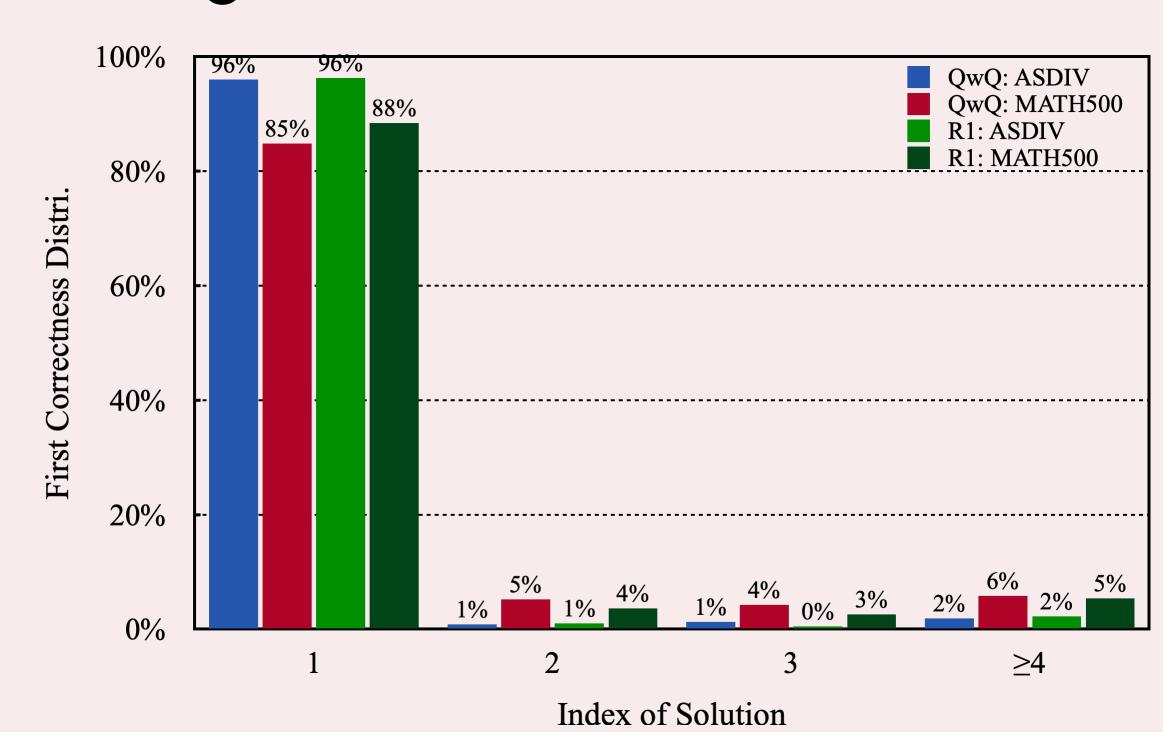
QwQ-32B-Preview (901 tokens)



QwQ-32B-Preview generates 13 solutions!

### **FINDINGS**

Finding 1: Redundant solutions contribute less to the accuracy



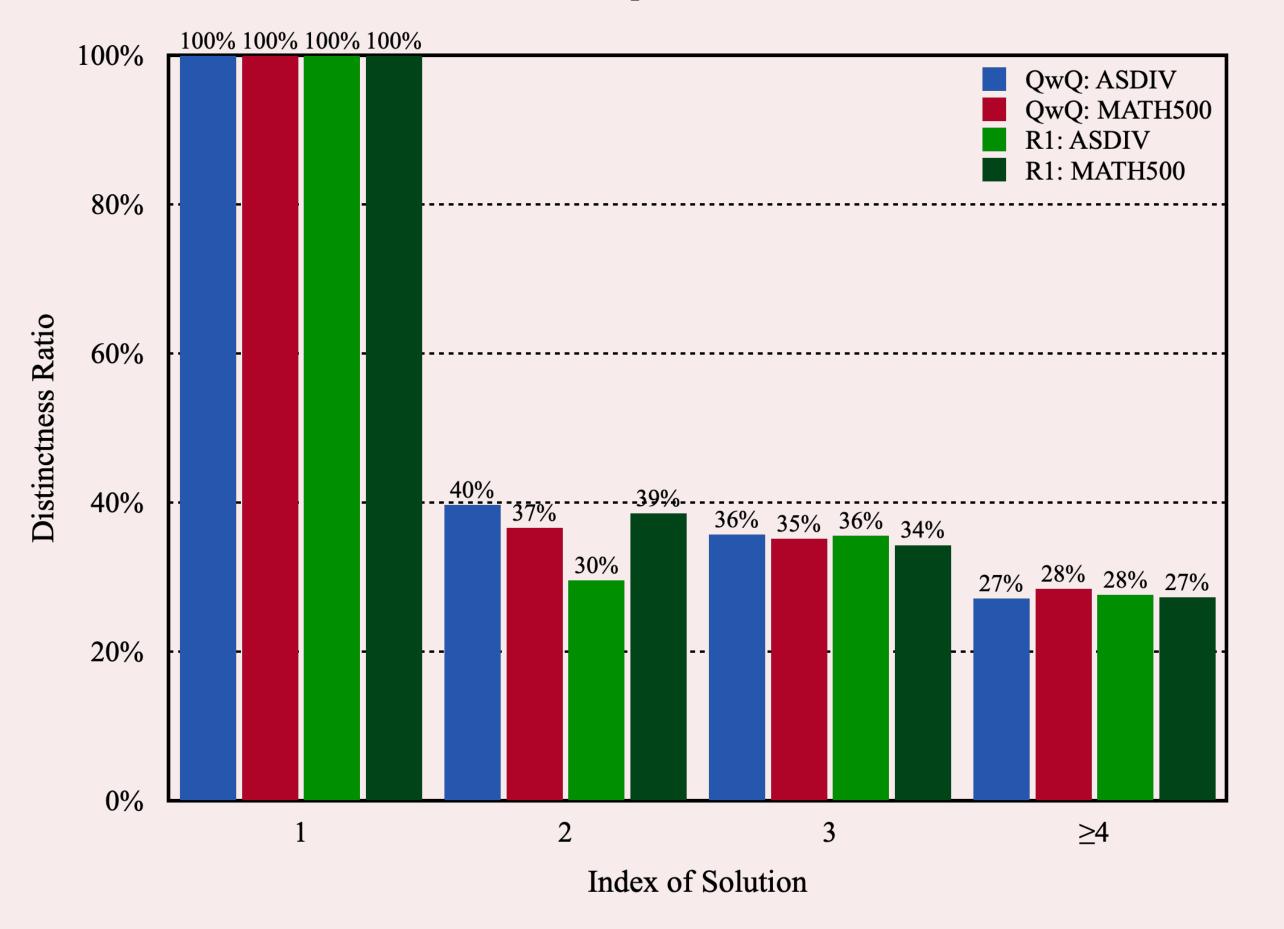
- ► More than 85% of the time the first answer is already correct.
- ► Subsequent solutions mainly verify previous solutions.

(Figure: The distribution of First Correct Solution)

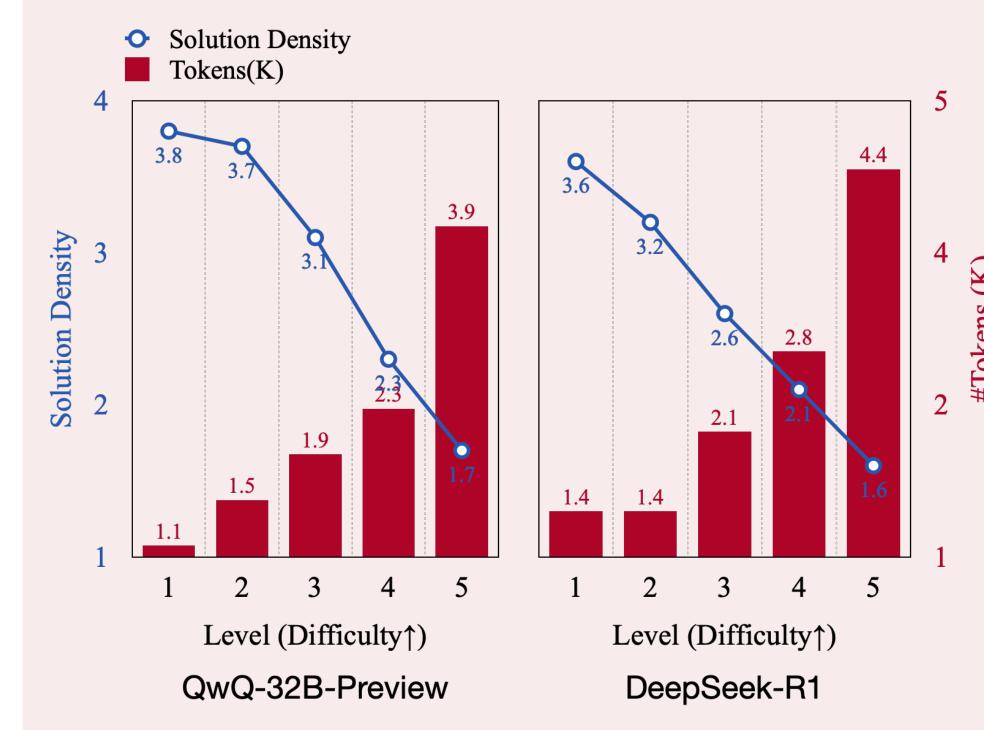
Finding 2: Redundant solutions lack diversity

- ► The second solution has only a 50% chance of introducing a new reasoning strategy.
- ► As the number of solutions increases, the possibility decreases.

(Figure: The probability that the solution of each position introduces a new reasoning strategy)



Finding 3: Overthinking is more severe in simple questions



- ► Solution Density: the number of solutions per 1000 tokens
- ► Overthinking is more prominent in simple questions

(Figure: The Solution Density in different difficult level of problems (MATH500))

#### **METRICS**

► Outcome Efficiency: The ratio of the tokens in first correct solution to the total tokens

$$\xi_O = \frac{1}{N} \sum_{i=1}^{N} \sigma_i \frac{\hat{T}_i}{T_i}$$

- $\sigma_i$ : The correctness of the *i*-th response.
- $\widehat{T}_i$ : The tokens in the first correct solutions
- $T_i$ : The tokens in the full response
- ► Process Efficiency: The ratio of the tokens in different reasoning strategy to the total tokens

$$\xi_P = \frac{1}{N} \sum_{i=1}^N \frac{D_i}{T_i}$$

- $D_i$ : The tokens in the different reasoning strategies
- $T_i$ : The tokens in the full response

## METHODS

# Length Preference Optimization

- ► Step 1: Sampling on training set
- ► Step 2: Streamline sampling results
- ► Step 3: Construct Preference Pair
- ► Positive Example: Keep the *first correct solution* and another round of *verification (reflection)*
- ► Negative Example: The longest response in sampling results

#### **EXPERIMENT RESULTS**

Methods	Accuracy	Response		Efficiency	
		#Solution	#Token	Outcome	Process
	AS	SDIV			
QwQ-32B-Preview	96.5	3.5	713.7	53.7%	69.0%
+SimPO <sub>FCS+Reflection</sub>	96.6	1.9	381.5	82.5%	87.9%
	GS	M8K			
QwQ-32B-Preview	94.7	2.9	756.1	11.8%	75.2%
+SimPO <sub>FCS+Reflection</sub>	95.9	1.8	416.5	86.0%	91.0%
	MA	TH500			
QwQ-32B-Preview	91.2	4.3	2398.5	51.4%	70.3%
+SFT <sub>Shortest Response</sub>	92.6	4.4	2359.0	59.7%	72.8%
+DPO <sub>Shortest Response</sub>	93.2	3.4	1928.8	64.3%	77.8%
+RPO <sub>Shortest Response</sub>	90.2	3.5	2015.2	64.7%	76.6%
+SimPO <sub>Shortest Response</sub>	91.0	3.5	1871.5	64.7%	78.1%
+SimPO <sub>First-Correct Solution</sub>	90.4	1.3	1015.6	85.5%	96.3%
+SimPO <sub>FCS+Reflection</sub> (Ours)	91.4	2.4	1330.3	79.1%	88.9%
+SimPO <sub>Greedily Diverse Solutions</sub>	91.2	1.7	1285.8	80.1%	90.2%

- ► Our proposed method maintains comparable math reasoning performance.
- ► Also greatly reduces generated tokens and improves efficiency.