

SRPO: Enhancing Multimodal LLM Reasoning via Reflection-Aware Reinforcement Learning

Zhongwei Wan,

The Ohio State University,

https://arxiv.org/abs/2506.01713

https://srpo.pages.dev/



Background and Motivation

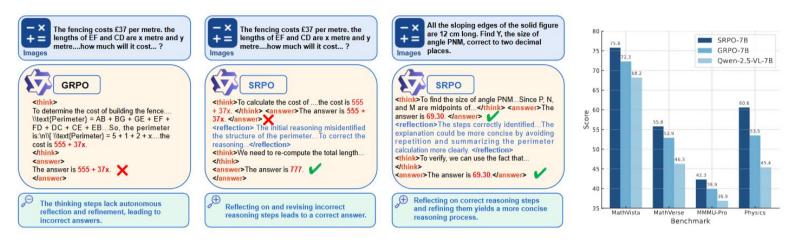
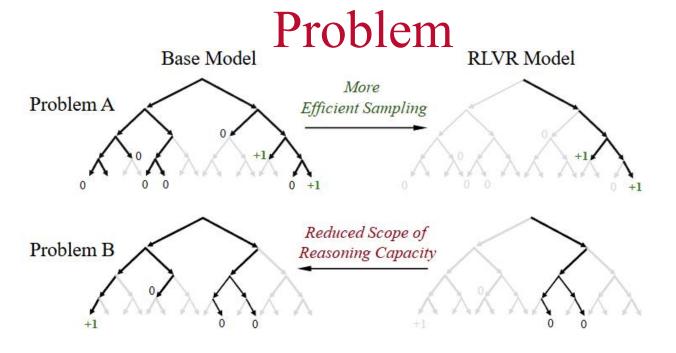


Figure 1: Left: Illustrative examples of reflection improving reasoning. Right: Quantitative comparison on benchmark datasets.

- MLLMs struggle on complex visual-text reasoning.
- Outputs often verbose or incorrect.
- Missing piece: explicit self-reflection.
- We propose SRPO to inject reflection in SFT + RL.

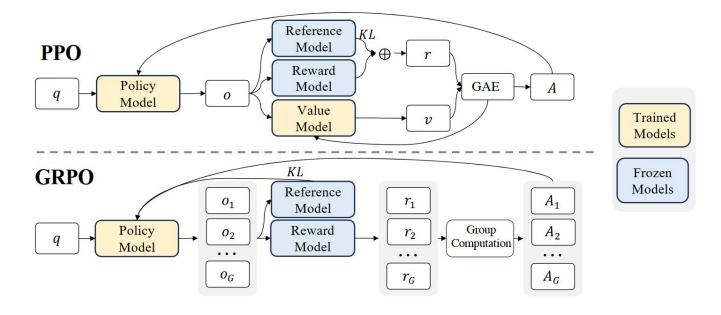




- Local token dependency \rightarrow drift, verbosity.
- Pretraining constrains behaviors; vanilla RL can't teach reflection well.
- Need a unified way to learn reflect—revise.



State of the Arts & Limitations



- Large-scale RL (R1, GRPO variants) boosts CoT but seldom rewards reflection.
- Multimodal RL (Vision-R1, MM-Eureka, VL-Rethinker) lacks reflection utility checks.
- Vulnerable to length redundancy and reward gaming.



Overview of SRPO

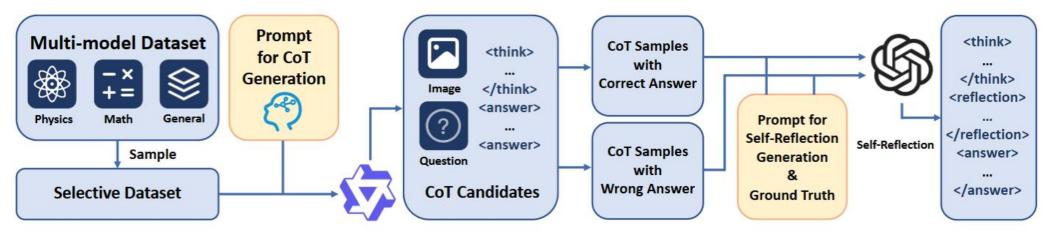


Figure 2: Pipeline of Self-Reflection SFT data construction, including CoT and self-reflection generation.

- Two stages: reflection-SFT \rightarrow reflection-aware RL.
- Pattern: first solution \rightarrow reflection \rightarrow refined solution.
- Reward = task + reflection.



Key Technique (1/3): Reflection-Oriented SFT

SFT Training Data

This data first generated by the multimodel LLM you want to enhance, then used an advanced Multimodel LLM to give a highquality reflection

Question:

In $\triangle PQR$, let ZQ=3a-11, ZP=a+5, PY=2c-1, YR=4c-11, $\angle PRZ=4b-17$, $\angle ZRQ=3b-4$, $\angle QYR=7b+6$, and $\angle PXR=2a+10$. PX is an altitude of $\triangle PQR$. Find a.

A: 10

B: 20

C: 30 D: 40

PZX

```
To solve for 'a', we need to use the fact that the sum of the angles around point Z is 360 degrees since Z is a vertex of triangle PQR. However, the problem provides us with the measures of angles at Z and at other points, which suggests that we might need to use the properties of the triangle and the given information about the altitude PX.
```

The key here is recognizing that the sum of the angles in triangle PQR is 180 degrees. But since we have angles at point Z and PX is an altitude, we can use the fact that the sum of the angles around point Z is 360 degrees minus the angle at Z itself, which is part of the triangle's interior angles.

Let's start by summing up the given angles around point Z:

 $m\angle PZR + m\angle ZQR + m\angle QYR = 360^{\circ} - m\angle PRZ$ Substituting the given values:

(4b - 17) + (3b - 4) + (7b + 6) = 360° - (a + 5)

Combining like terms

14b - 15 = 360° - a - 5

Simplifying further:

14b - 15 = 355 - a

Now, let's isolate 'a':

a = 370 - 14b

Since we don't have the value of 'b' directly, we need to look at the other given angles and the properties of the triangle. However, the problem seems to be designed such that we can find 'a' directly from the options provided.

Given the complexity of the equation and the need to find 'a', we can test the options:

В

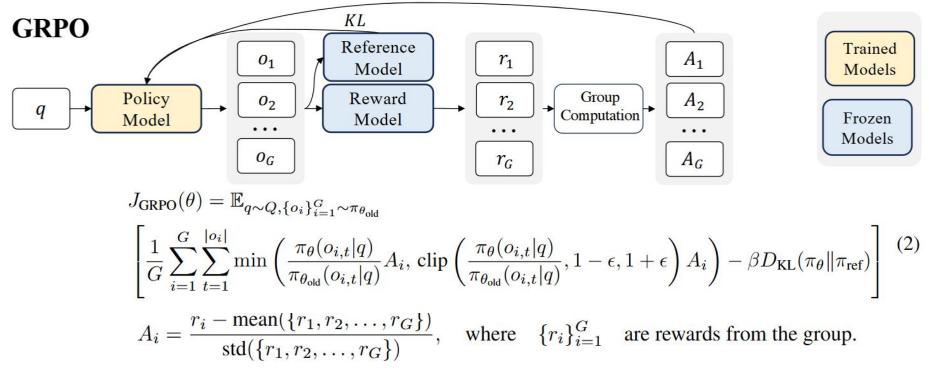
⚠ The original chain-of-thought reveals several issues. First, the reasoning starts with an incorrect assumption: it incorrectly treats the sum of angles around point Z as 360° minus the angle at Z from the triangle, confusing the concept of full rotational angle sum with interior angle sums in triangles. This leads to an inconsistent approach because it mistakenly attempts to sum angles that are not necessarily co-terminal or relevant to the triangle's interior angle sum (180°). In addition, the chain-of-thought relies on a direct substitution and solution for variable 'b' when in fact, the altitude information (PX perpendicular to QR) could offer necessary relationships to set up right triangle equations, a possibility which is completely overlooked. The reasoning would benefit from clarifying assumptions such as the correct use of angle relationships in triangles (e.g., complementary angles due to the altitude and the fact that PX forms right angles with QR) and using correct geometric constructions to relate angles and sides. Moreover, testing the answer choices without fully establishing equations to relate "a", "b", and "c" seems more like guesswork and might lead to trivialization of the problem's structure. A more robust approach might be to: 1) check proper angle relationships induced by the altitude; 2) clearly identify all triangles and apply the triangle sum theorem directly; 3) ensure that the angles being summed are those that form a closed figure. Finally, carefully verifying each step and avoiding mixing rotations with interior angles would lead to a correct and more methodical solution

D

- Build ~10k high-quality reflective samples.
- Two types: refine correct CoT; revise incorrect CoT.
- Distill reflective skills from stronger MLLMs.



Key Technique (2/3): GRPO Recap & Group Advantage



- Grouped sampling; intra-group advantage.
- Clipped ratio + KL to reference.
- No critic network needed.



Key Technique (3/3): Reflection-Aware Rewards

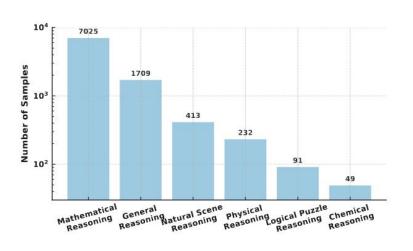
• Structure: think \rightarrow reflect \rightarrow rethink.

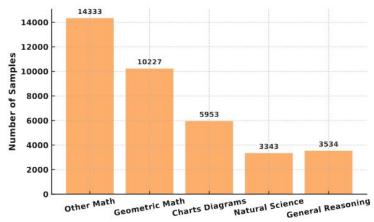
$$egin{aligned} R_{ ext{total}} &= R_{ ext{task}} + R_{ ext{reflection}}. \ R_{ ext{task}} &= R_{ ext{format}}(0/0.5) + R_{ ext{accuracy}}(0/0.5). \ R_{ ext{reflection}} &= I_{ ext{eff}} + I_{ ext{ref}} + lpha f_{ ext{len}}(L_{ ext{response}}). \ f_{ ext{len}}(L) &= \left(\exp(-|L-T_{ ext{target}}|/(T_{ ext{max}}-T_{ ext{target}}))\right)^2. \ I_{ ext{eff}} &\in \{-0.25, 0, 0.25, 0.5\} \ ext{(penalize harmful reflection)}. \end{aligned}$$

$$I_{\rm eff} = \begin{cases} 0.25, & \text{if reflection keeps a corrected answer,} \\ 0.5, & \text{if reflection corrects the wrong answer,} \\ 0, & \text{if reflection fails to correct the wrong answer,} \\ -0.25 & \text{if reflection misconducts the right into wrong answer.} \end{cases}$$



Experimental Settings





(a) Self-reflection SFT data statistic

- (b) RL training data statistic
- SFT: ~10k reflective samples from LLaVA-CoT, Mulberry, MathV360K.
- RL: ~30k diverse multimodal reasoning tasks.
- Models: Qwen-2.5-VL-7B/32B; OpenRLHF; 3 epochs; α =0.1; T=1.0; Adam 1e-6.



Results (1/3): Main Benchmarks

- SRPO-7B tops open-source peers across many tasks.
- SRPO-32B competitive with strong closed models.
- Strong on MathVista/MathVerse/MMMU-Pro/EMMA.

MathVista MathVerse MathVision OlympiadBench WeMath MMMU-Pro	MMMU	TTO 43 5 1
Claude3.7-Sonnet 66.8 52.0 41.3 48.9 72.6 51.5 GPT-40 63.8 50.2 30.4 35.0 68.8 51.9 GPT-01 73.9 57.0 60.3 68.0 98.7 62.4 Gemini2-flash 70.4 59.3 41.3 51.0 71.4 51.7 Seed1.5-VL-T 85.6 - 68.7 65.0 - 67.6 Deen-Source General MLLMs (7B-16B)		EMMA
GPT-40 63.8 50.2 30.4 35.0 68.8 51.9 GPT-01 73.9 57.0 60.3 68.0 98.7 62.4 Gemini2-flash 70.4 59.3 41.3 51.0 71.4 51.7 Seed1.5-VL-T 85.6 - 68.7 65.0 - 67.6 Open-Source General MLLMs (7B-16B)		
GPT-01 73.9 57.0 60.3 68.0 98.7 62.4 Gemini2-flash 70.4 59.3 41.3 51.0 71.4 51.7 Seed1.5-VL-T 85.6 - 68.7 65.0 - 67.6 **Open-Source General MLLMs (7B-16B)** InternVL2-8B 58.3 22.8 17.4 †10.1 †47.2 29.0 InternVL2.5-8B 64.4 39.5 19.7 12.3 53.5 34.3 QwenVL2-7B 58.2 19.7 16.3 †9.7 †51.6 30.5 Llava-OV-7B 63.2 26.2 †18.5 †8.5 †49.9 24.1 Kimi-VL-16B 68.7 44.9 21.4 QwenVL2.5-7B 68.2 46.3 25.1 20.2 62.1 36.9 **Open-Source Reasoning MLLMs (7B)** **MM-Eureka-8B¹** 67.1 40.4 22.2 8.6 †55.7 27.8	68.3	35.1
Gemini2-flash Seed1.5-VL-T 70.4 85.6 59.3 41.3 51.0 71.4 51.7 67.6 Open-Source General MLLMs (7B-16B) InternVL2-8B 58.3 22.8 17.4 †10.1 †47.2 29.0 InternVL2.5-8B 64.4 39.5 19.7 12.3 53.5 34.3 QwenVL2-7B 58.2 19.7 16.3 †9.7 †51.6 30.5 Llava-OV-7B 63.2 26.2 †18.5 †8.5 †49.9 24.1 Kimi-VL-16B 68.7 44.9 21.4 — — — — QwenVL2.5-7B 68.2 46.3 25.1 20.2 62.1 36.9 Open-Source Reasoning MLLMs (7B) MM-Eureka-8B¹ 67.1 40.4 22.2 8.6 †55.7 27.8	69.1	32.7
Seed1.5-VL-T 85.6 - 68.7 65.0 - 67.6 Open-Source General MLLMs (7B-16B) InternVL2-8B 58.3 22.8 17.4 † 10.1 † 47.2 29.0 InternVL2.5-8B 64.4 39.5 19.7 12.3 53.5 34.3 QwenVL2-7B 58.2 19.7 16.3 † 9.7 † 51.6 30.5 Llava-OV-7B 63.2 26.2 † 18.5 † 8.5 † 49.9 24.1 Kimi-VL-16B 68.7 44.9 21.4 - - - - QwenVL2.5-7B 68.2 46.3 25.1 20.2 62.1 36.9 Open-Source Reasoning MLLMs (7B) MM-Eureka-8B ¹ 67.1 40.4 22.2 8.6 †55.7 27.8	78.2	45.7
Den-Source General MLLMs (7B-16B)	70.7	33.6
InternVL2-8B 58.3 22.8 17.4 †10.1 †47.2 29.0 InternVL2.5-8B 64.4 39.5 19.7 12.3 53.5 34.3 QwenVL2-7B 58.2 19.7 16.3 †9.7 †51.6 30.5 Llava-OV-7B 63.2 26.2 †18.5 †8.5 †49.9 24.1 Kimi-VL-16B 68.7 44.9 21.4 QwenVL2.5-7B 68.2 46.3 25.1 20.2 62.1 36.9 Open-Source Reasoning MLLMs (7B) MM-Eureka-8B ¹ 67.1 40.4 22.2 8.6 †55.7 27.8	77.9	2
InternVL2.5-8B		
QwenVL2-7B 58.2 19.7 16.3 †9.7 †51.6 30.5 Llava-OV-7B 63.2 26.2 †18.5 †8.5 †49.9 24.1 Kimi-VL-16B 68.7 44.9 21.4 - - - - QwenVL2.5-7B 68.2 46.3 25.1 20.2 62.1 36.9 Open-Source Reasoning MLLMs (7B) MM-Eureka-8B¹ 67.1 40.4 22.2 8.6 †55.7 27.8	51.2	19.8
Llava-OV-7B 63.2 26.2 † 18.5 † 8.5 † 49.9 24.1 Kimi-VL-16B 68.7 44.9 21.4 - - - QwenVL2.5-7B 68.2 46.3 25.1 20.2 62.1 36.9 Open-Source Reasoning MLLMs (7B) MM-Eureka-8B¹ 67.1 40.4 22.2 8.6 †55.7 27.8	56.0	†20.6
Kimi-VL-16B 68.7 44.9 21.4 - - - - QwenVL2.5-7B 68.2 46.3 25.1 20.2 62.1 36.9 Open-Source Reasoning MLLMs (7B) MM-Eureka-8B¹ 67.1 40.4 22.2 8.6 †55.7 27.8	54.1	20.2
Kimi-VL-16B 68.7 44.9 21.4 - - - - QwenVL2.5-7B 68.2 46.3 25.1 20.2 62.1 36.9 Open-Source Reasoning MLLMs (7B) MM-Eureka-8B¹ 67.1 40.4 22.2 8.6 †55.7 27.8	48.8	18.3
Open-Source Reasoning MLLMs (7B) MM-Eureka-8B¹ 67.1 40.4 22.2 8.6 †55.7 27.8	55.7	220
MM-Eureka-8B ¹ 67.1 40.4 22.2 8.6 †55.7 27.8	54.3	21.5
R1-VL-7B 63.5 40.0 24.7 †10.8 †53.8 7.8	49.2	†21.5
	44.5	8.3
R1-Onevision-7B 64.1 46.4 23.5 17.3 61.8 21.6	====	20.8
OpenVLThinker-7B 70.2 47.9 25.3 20.1 64.3 37.3	52.5	26.6
VL-Rethinker-7B 74.9 54.2 32.3 †20.5 †70.2 41.7	56.7	29.7
Vision-R1-7B 73.5 52.4 †27.2 †19.4 †62.9 †37.7	†54.7	†22.4
MM-Eureka-7B ² 73.0 50.3 26.9 20.1 66.1 †37.6	†55.2	†23.5
* (Ours - SRPO-7B) 75.8 55.8 32.9 22.8 71.6 42.3	57.1	29.6
Open-Source General and Reasoning MLLMs (32B)		
InternVL2.5-VL-38B 71.9 49.4 31.8 32.0 67.5 46.0	57.6	
Qwen-2.5-VL-32B 74.7 48.5 38.4 30.0 69.1 49.5	59.4	31.1
InternVL2.5-38B-MPO 73.8 46.5 32.3 25.6 66.2 -	<u> 22</u> 8	_
MM-Eureka-32B 74.8 56.5 34.4 35.9 73.4 †50.4	†62.3	†34.5
* (Ours - SRPO-32B) 78.5 58.9 39.6 38.5 76.4 51.3	66.1	38.2



Results (2/3): Cross-Disciplinary & RL Variants

Table 2: Performance comparison across different disciplines in MMK12.

Model	Math	Phys	Chem	Bio
Closed Models	ORDANI NE	Deline.	LUCACIO DE C	5000.0000
Claude3.7	57.4	53.4	55.4	55.0
GPT-40	55.8	41.2	47.0	55.4
01	81.6	68.8	71.4	74.0
Gemini2	76.8	53.6	64.6	66.0
Open General MLLMs		CHARLES TO	40.000.000.000	Secretarion
IntVL2.5-8B	46.8	35.0	50.0	50.8
Qwen-2.5-7B	58.4	45.4	56.4	54.0
IntVL2.5-38B	61.6	49.8	60.4	60.0
Qwen-2.5-32B	71.6	59.4	69.6	66.6
Qwen-2.5-72B	75.6	64.8	69.6	72.0
Open Reasoning MLL	Ms	Altonomical	273327777	7 - 200-53
IntVL2.5-8B-MPO	26.6	25.0	42.4	44.0
IntVL2.5-38B-MPO	41.4	42.8	55.8	53.2
R1-OneVision	44.8	33.8	39.8	40.8
MM-Eureka-7B	71.2	56.2	65.2	65.2
OpenVLThinker	63.0	53.8	60.6	65.0
MM-Eureka-32B	74.6	62.0	75.4	76.8
SRPO-7B	75.3	60.6	70.3	69.5
SRPO-32B	77.5	64.2	77.5	79.2

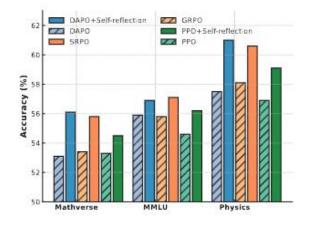


Figure 5: Performance of various RL methods with and without self-reflection.

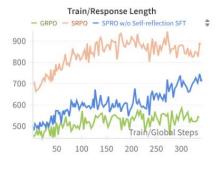


Results (3/3): Ablations & Dynamics

Table 3.	Ablation study	of SRPO-7B on R	L training da	ta size and	self-reflection	components
Table J.	ADIAHOH SILIUV	OI DIXI O-1D OILK	L uanime da	na size and	SCH-ICHCCHOIL	components.

Model Components	RL Data Size	MathVista	MathVerse	MathVision	MMMU-Pro	Physics	Avg.
Qwen-2.5-VL-7B		68.2	46.3	25.1	36.9	45.4	44.4
+ GRPO	37K	72.3	52.9	30.3	39.9	53.5	49.8
* (Ours - SRPO-7B)	37K	75.8	55.8	32.9	42.3	60.6	53.5
SRPO-7B	15K	74.5	54.9	32.2	41.4	60.1	52.6
SRPO-7B	5K	73.7	53.6	31.2	40.3	57.7	51.3
w/o Self-Reflection SFT	37K	74.2	53.3	30.3	39.7	58.6	51.2
w/o Self-Reflection RL	37K	70.3	48.2	27.2	38.7	48.5	46.6
 no Length Reward (f_{len}(·)) 	37K	75.3	56.2	32.4	41.7	60.1	53.1
 no Effectiveness Reward (I_{eff}) 	37K	73.9	54.7	31.6	40.9	58.8	52.0







- More RL data \rightarrow steady gains; even 5k > GRPO.
- Removing SFT or reflection-RL hurts; I_eff is crucial.
- SRPO converges faster; length well-controlled.



Samples Analysis

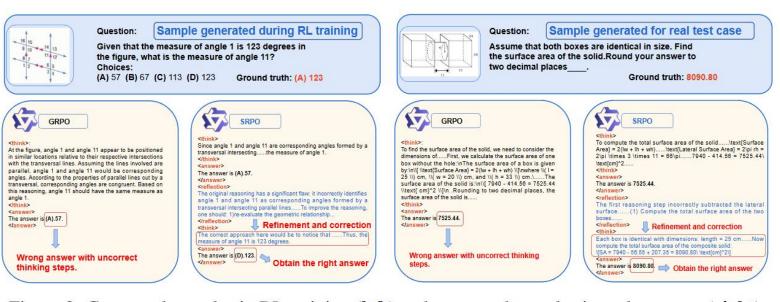
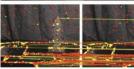


Figure 3: Generated samples in RL training (left) and generated samples in real test case (right).



Question: Based on the upper image, what task is the robotic arm most likely performing?





Options: (A) Sorting playing cards based on a predefined algorithm.

(B) Performing a magic trick with the cards.

(C) Playing a card game against a human opponent.

(D) Conducting an experiment on object recognition. Ground truth: (D)

The image shows a robotic arm with a camera mounted on it, which suggests that it is involved in a task that requires visual input and processing. The presence of playing cards on the table and the arm's position above them indicate that the robot might be interacting with the cards in some way. Given the context of the setup, the most logical task for the robotic arm would be to sort the playing cards based on a predefined algorithm, as this is a common application for robotic arms equipped with cameras for visual recognition.

</think>

<answer> The answer is A. (Incorrect)

</answer>

The original reasoning is overall correct in observing the camera and the cards, but it overlooks deeper experimental context. The red and yellow point clouds and structured laser-like patterns shown in the lower part of the image are indicative of 3D perception or object localization, which are typical in object recognition experiments. Additionally, there's no evidence of sorting mechanisms or card repositioning. The reflection should thus account for the experimental nature of the setup, making option D more consistent with the evidence provided.

</reflection>

<think> (Generate the right answer after effective self-reflection)

Let's rethink, the image shows a robotic arm with a camera, indicating a task that involves visual data collection or processing. The presence of playing cards on the table suggests an interaction with objects, and the alignment of the dots and lines might represent data points collected for object recognition or spatial analysis. Given these observations, the most likely task for the robotic arm is related to object recognition, where the camera would be used to identify and analyze the cards' features.

</think> <answer>

The answer is D</answer>

13

Conclusion & Future Work

- SRPO unifies reflection in SFT + RL.
- Rewards: structure, brevity, effectiveness.
- Next: scale (MoE/larger), harder multimodal tasks, better reflection data generation, dynamic self-reflection.

