Verbal RL via Self-Reflection

Noah Shinn

GPT-3.5-Turbo: The sky was ablaze with a stunning array of colors.

*Actual count = 10 words

GPT-3.5-Turbo: The sky was ablaze with a stunning array of colors.

User: Verbosely check if this sentence has exactly 11 words.

GPT-3.5-Turbo: 1. The 2. sky [...] 9. of 10. colors.

Yes, the sentence has exactly 10 words. I apologize for the previous mistake. Here is an alternative sentence with 11 words:

"I embarked on a journey to discover the hidden wonders of nature."

*Actual count = 12 words

[truncated]

GPT-3.5-Turbo: I embarked on a journey to discover the hidden wonders of nature.

User: Verbosely check if this sentence has exactly 11 words.

GPT-3.5-Turbo: 1. I 2. embarked [...] 11. of 12. nature.

Apologies for the previous oversight. This sentence indeed has 12 words, not 11. Here is an alternative sentence with exactly 11 words:

"She diligently practiced the piano, striving for perfection in every note."

*Actual count = 11 words

[truncated]

GPT-3.5-Turbo: She diligently practiced the piano, striving for perfection in every note.

User: Verbosely check if this sentence has exactly 11 words.

GPT-3.5-Turbo: 1. She 2. diligently [...] 10. every 11. note.

Yes, the sentence has exactly 11 words.

Task completed

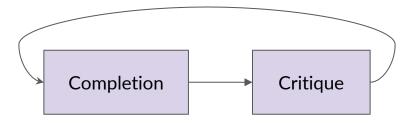
Reasoning in language space

Humans LLMs

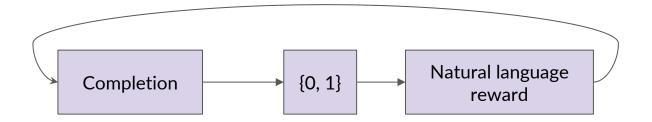
Q: What is 30,000 + 50,000 Q: What is 30,000 + 50,000

A: 80,000 A: 80,000

Critique methods



Natural language can amplify rewards



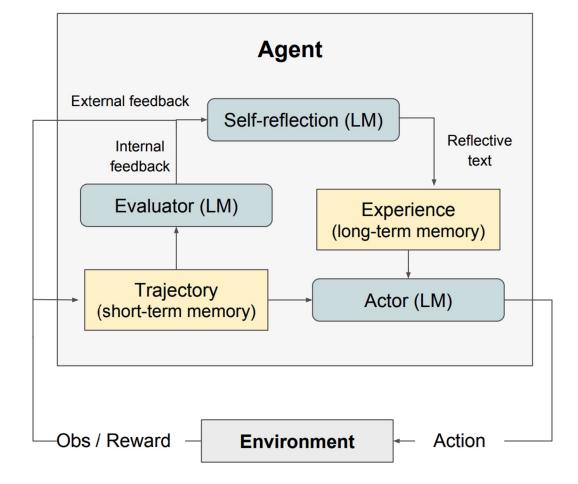
Verbal RL

$$\pi_{ heta}(a_i|s_i),\, heta=\{M_a\}$$
 $igg|$ $\pi_{ heta}(a_i|s_i),\, heta=\{M_a, egin{array}{c} mem \} \end{array}$

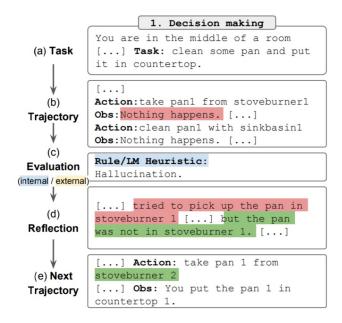
Algorithm 1 Reinforcement via self-reflection

```
Initialize Actor, Evaluator, Self-Reflection:
M_a, M_e, M_{sr}
Initialize policy \pi_{\theta}(a_i|s_i), \theta = \{M_a, mem\}
Generate initial trajectory using \pi_{\theta}
Evaluate \tau_0 using M_e
Generate initial self-reflection sr_0 using M_{sr}
Set mem \leftarrow [sr_0]
Set t = 0
while M_e not pass or t < \max trials do
    Generate \tau_t = [a_0, o_0, \dots a_i, o_i] using \pi_{\theta}
    Evaluate \tau_t using M_e
    Generate self-reflection sr_t using M_{sr}
    Append sr_t to mem
    Increment t
end while
return
```

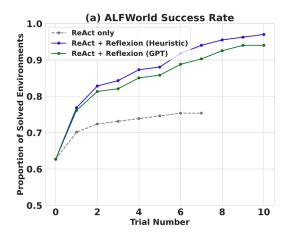
Verbal RL



Decision-making



- 1. (Task)
- (Action + observation trajectory)
- 3. (Evaluation)
- 4. (Self-reflection)
- 5. (Repeat)

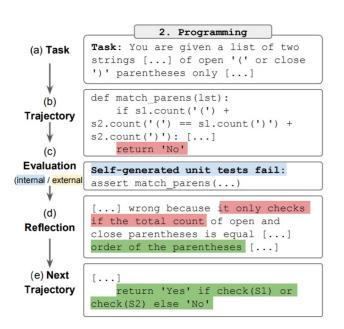


| Welcome!
| You are in the middle of the room.
| Looking around you, you see
| a diningtable, a stove,
| a microwave, and a cabinet.
| Your task is to:
| Put a pan on the diningtable.
| > goto the cabinet
| You arrive at the cabinet.
| The cabinet is closed.
| > open the cabinet
| The cabinet is empty.





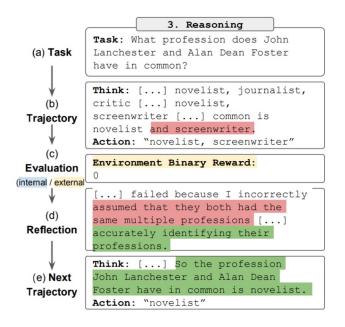
Programming

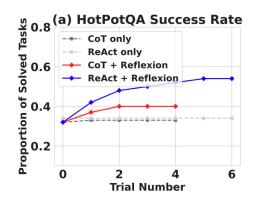


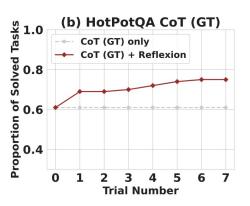
Benchmark + Language	Prev SOTA Pass@1	SOTA Pass@1	Reflexion Pass@1
HumanEval (PY)	65.8 (CodeT [4] + GPT-3.5)	80.1 (GPT-4)	91.0
HumanEval (RS)	_	60.0 (GPT-4)	68.0
MBPP (PY)	67.7 (CodeT [4] + Codex [5])	80.1 (GPT-4)	77.1
MBPP (RS)	_	70.9 (GPT-4)	75.4
Leetcode Hard (PY)	_	7.5 (GPT-4)	15.0

- 1. (Instruction)
- 2. (Function implementation)
- 3. (Unit test feedback) \leftarrow execution
- 4. (Self-reflection)
- 5. (Repeat)

Reasoning

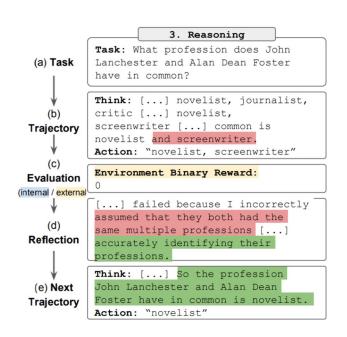


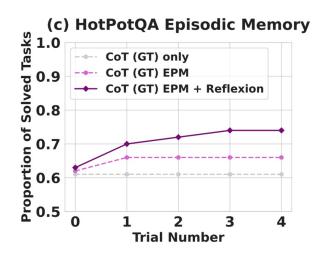




- 1. (Question)
- 2. (Knowledge retrieval)
- 3. (Answer)
- 4. (Exact match evaluation)
- 5. (Self-reflection)
- 6. (Repeat)

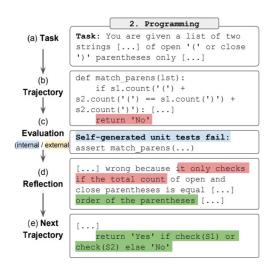
Is episodic memory enough for self-improvement?





- 1. (Question)
- 2. (Knowledge retrieval)
- 3. (Answer)
- 4. (Exact match evaluation)
- 5. (Self reflection)
- 6. (Repeat)

Is self-debugging enough for self-improvement?



Approach	Test Generation	Self-reflection	Pass@1 (Acc)
Base model	False	False	0.60
Test generation omission	False	True	0.52
Self-reflection omission	True	False	0.60
Reflexion	True	True	0.68

Test generation omission

- 1. (Instruction)
- 2. (Function implementation)
- 3. (Unit test feedback) ← execution
- 4. (Self-reflection)
- 5. (Repeat)

Self-reflection omission

- 1. (Instruction)
- 2. (Function implementation)
- 3. (Unit test feedback) \leftarrow execution
- 4. (Self-reflection)
- 5. (Repeat)

More directions

- Various exploration techniques exist
- Persisting memory can store new ideas
- Few-shot demonstrations are powerful
- Transfer learning in semantic space

References

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