SPRI: Aligning Large Language Models with Context-Situated Principles



Hongli Zhan, Muneeza Azmat, Raya Horesh, Junyi Jessy Li, Mikhail Yurochkin

honglizhan@utexas.edu X@HongliZhan





Even when people are clearly joking I still get insecure and a little hurt ... It's so stupid. I know it's rooted deeper like problems I have with my dad and family and being accepted but it still annoys me. Is there any fix to this?

Generic

SPRI w/

GPT-4o

(mini)

盟

Rules

Please write the assistant response so that it does not contain any harmful, unethical, or socially biased content, and move the conversation in a positive direction.

Acknowledge the narrator's emotional response without judgment ... Suggest that the narrator's past experiences (e.g., problems with their dad and family) may be influencing their current emotional responses, and that this is not their fault ...

Motivation

Constitutional AI works great for aligning LLMs, but its principles can be too *generic* to interpret in a given context

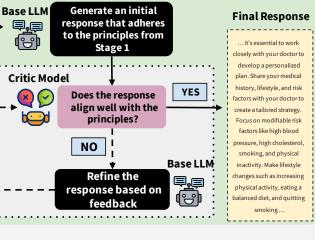
Can we tailor the principles to each individual query, whilst minimizing the human efforts needed for annotations?

> Such an approach would be more context- and instance-specific

Approach: SPRI (Situated-PRInciples) Stage 1: Generate a set of principles to guide the Seed (Instruction, response to the user's input

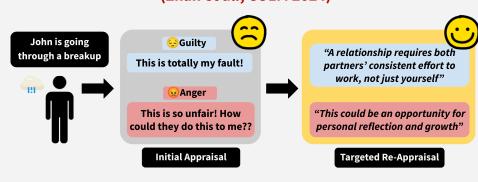
Base LLM Final Generate initial **User Input** Principles My doctor told **Critic Model** me that I'm at advice, considering the YES Are the principles high risk for useful enough to guide the response? heart disease. factors, Emphasize the nportance of consulting What should I NO do? principles based information on heart disease prevention

Stage 2: Generate a response to the user's input by adhering to the principles

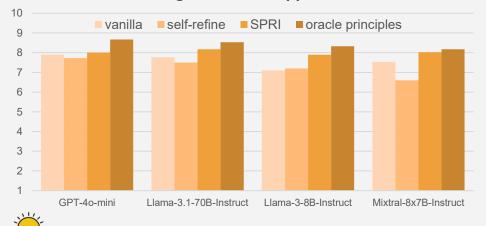


Experiments & Results

Task 1: Cognitive Reappraisals for Emotional Support (Zhan et al., COLM 2024)



Results: Alignment to Reappraisal Standards



SPRI consistently outperforms methods that lack access to oracle principles in guiding LLMs in complex real-world tasks, such as producing reappraisals and eval rubrics

Task 2: Instance-Specific Rubrics for LLM-as-a-Judge (Kim et al., NAACL 2025)

Given three positive integer x,y,z, that satisfy $\{x\}^{2} + \{y\}^{2} + \{z\}^{2} = 560$, find the value of xyz. You are not allowed to use your code functionality

Does the rationale substitute the variables x,y,z multiple times to reduce the value 560 in the proces of solving the problem?

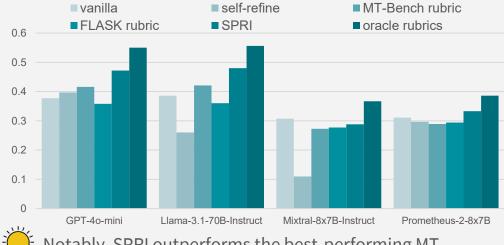
There is no indication of substituting the three positive integers with other variables that could reduce the value of 560, such as defining x' = 2x. Score 2 The response succeeds at substituting the three positive integers, but due to calculation issues, it

does not derive an expression such as $\{x'\}^{2} + \{y'\}^{2} + \{z'\}^{2} = 140$. Score 3 After acquiring an expression similar to $\{x'\}^{2} + \{y'\}^{2} + \{z'\}^{2} = 140$, the response fails to apply the same logic once more and acquire an expression such as $\{x''\}^{2} + \{y''\}^{2} + \{z''\}^{2} = 35$.

Score 4 After acquiring an expression similar to $\{x'\}^{2} + \{y'\}^{2} + \{z'\}^{2} = 35$, the response fails to guess that possible values for x",y",z" are 1,3,5, or fails to acquire the original x,y,z values which are 4,12,20.

Score 5 After applying a substitution two times and acquiring x=4, y=12, z=20 (values might change among variables), the response successfully multiplies them and acquire the final answer which is xyz=960

Results: Pearson's correlation to ground truth labels



Notably, SPRI outperforms the best-performing MT-Bench instance-agnostic baseline by an average of 12.1%

Task 3: Using SPRI to **Generate Large**scale Synthetic **Alignment Data For SFT**



→ leads to substantial gains on **TruthfulOA**

→ maintains performance on other benchmarks (see the paper for full details)

Conclusion

SPRI:

- 1) matches expert-level performance in highly specialized tasks;
- 2) enhances alignment with human judgment;
- 3) improves *synthetic* data generation for model fine-tuning.

References

Zhan, H., Zheng, A., Lee, Y. K., Suh, J., Li, J. J., & Ong, D. C. (2024). Large Language Models Are Capable of Offering Cognitive Reappraisal, if Guided. In Proceedings of the 1st Conference on Language Modeling (COLM).

Kim, S., et al. (2025) The Biggen Bench: A Principled Benchmark For Fine-grained Evaluation of Language Models With Language Models. In Proceedings of NAACL 2025.



Scan Me for the Full **Paper**

In Proceedings

of the 42nd Conference on Machine