

RLHF using natural language feedback.

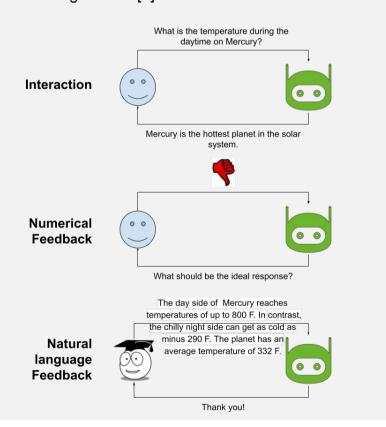
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1 Introduction

Reinforcement learning from human feedback(RLHF) has proven beneficial in aligning language models[1], however, majority of these approaches rely on numerical feedback[2]. RLHF using natural language feedback remains largely unexplored[2]. We propose two techniques to factor in natural language feedback alongside numerical feedback.

2 Problem Overview

After deployment, the BOT can gather human feedback from the user in the following manner[4]:



3 Methods

- 1. Joint loss function
- 1.1 Policy gradient objective[3]:

$$J_{ ext{PG}}(\pi_{ heta}) = \mathbb{E}_{ au \sim
ho_{\pi_{ heta}}} \left[\sum_{t=0}^{T-1} \gamma^t \cdot \hat{Q}_t \log \pi_{ heta}(a_t|s_t)
ight],$$

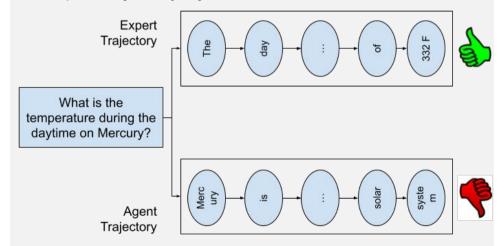
1.2 Interactive Imitation learning objective[2]:

$$J_{\mathrm{IL}}(\pi_{\theta}) = E_{s \sim d^{\pi_E}, a \sim \pi_E(.|s)}[\log(\frac{\pi_E(a|s)}{\pi_{\theta}(a|s)})]$$

1.3 Joint objective:

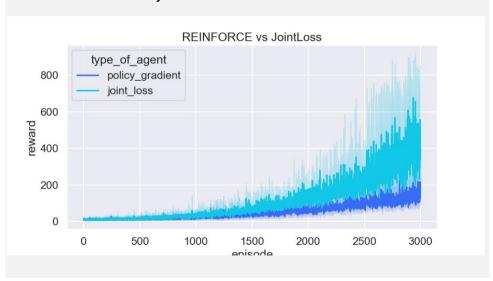
$$J(\pi_{\theta}) = \alpha J_{PG}(\pi_{\theta}) + (1 - \alpha) J_{IL}(\pi_{\theta})$$

2. Expert trajectory injection



4 Experiment setup and Results

- The proposed approaches are compared against REINFORCE algorithm(policy_gradient).
- Metric of comparison is cumulative reward after n episodes.
- The environment used for comparison is Mujoco's InvertedPendulum-v4. We will move to real world environment proposed in fits[4] once we get positive results in Mujoco.



References

[1] Long Ouyang Training language models to follow instructions with human feedback. 2022.
[2] Patrick Fernandes. Bridging the Gap: A Survey on Integrating (Human) Feedback for Natural Language Generation. Transactions of the Association for Computational Linguistics
[3] Richard S Sutton and Andrew G Barto. Reinforcement learning: An introduction. MIT press, 2018.

[4] Jing Xu Learning new skills after deployment: Improving open-domain internet-driven dialogue with human feedback, 2022.