



Pareto Prompt Optimization

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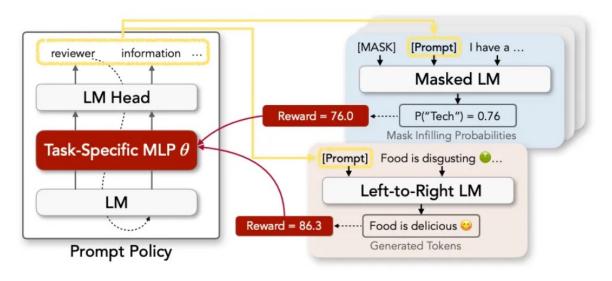


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What is Prompt Optimization

 A prompt is the input text to the LLM.

 Prompt optimization aims to automatically find effective prompts that produce desirable results from an LLM.



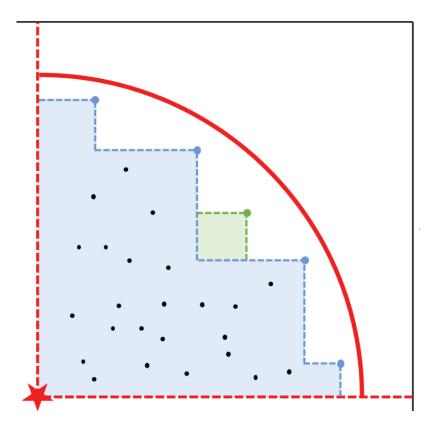
Workflow for RLPrompt

Source: Deng, Mingkai, et al. "Rlprompt: Optimizing discrete text prompts with reinforcement learning."



Multi objective Prompt Optimization

- Example: text summarization
 - Accuracy vs conciseness
- No single prompt excels in all objectives simultaneously
- We aim for a set of prompts from Pareto Front



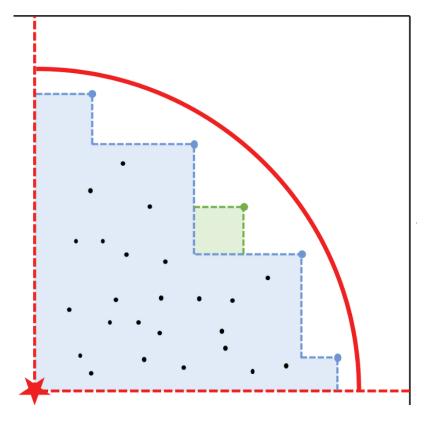
HyperVolume Metric illustration in Objective space



Multi objective Prompt Optimization

 RL or EA usually relies on single objective criterion.

Example: HyperVolume.

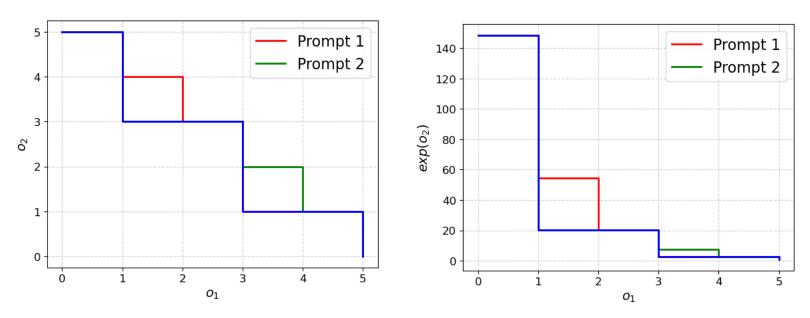


HyperVolume Metric illustration in Objective space



Decision-making criteria

- Requires accurate objective values
- Sensitive to non-linear transformations.



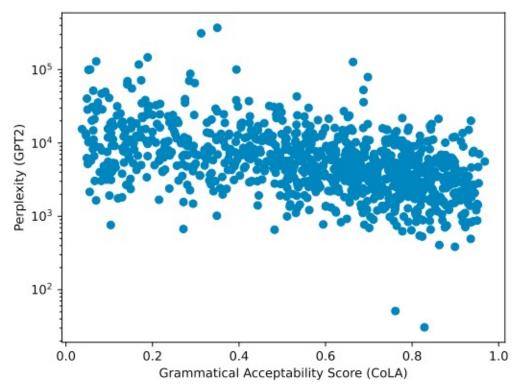
Effect of Non-Linear Transformation on Hypervolume Metric



Challenges of Multi-Objective Optimization in Text Generation

 Accurate objective values is often unavailable.

 Objectives may be measured by different metrics which have non-linear relationships.



Non-Linear relationship between different metrics.



ParetoPrompt

We propose ParetoPrompt, a **multi-objective** prompt optimization method driven by RL.

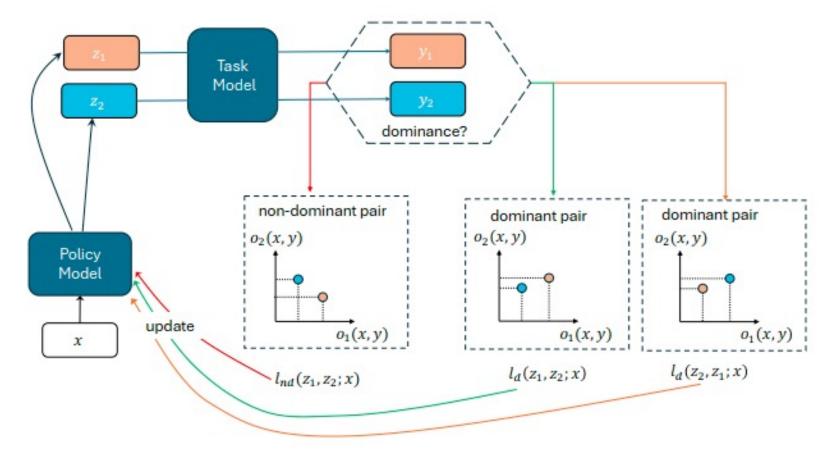
Our algorithm optimizes prompts based solely on the relative ranking of corresponding objectives, without requiring **precise objective values**.

Advantages:

- Avoids assumptions about how to combine objectives.
- Remains robust to changes in how each objective is scored (non-linear transformations, etc.).



ParetoPrompt Algorithm







Few-shot Text Classification with Misaligned Metrics

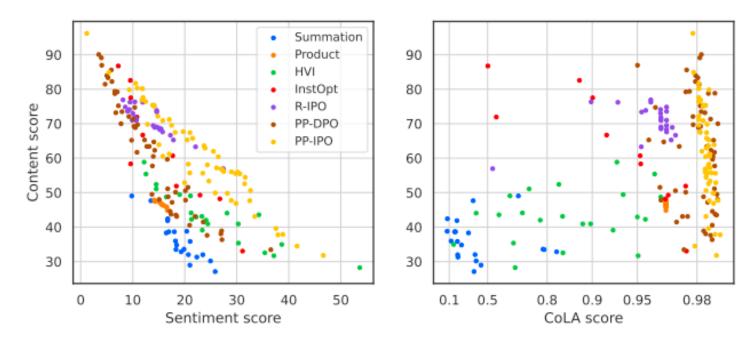
Dataset	Metric	Summation	Product	HVI	InstOpt	R-IPO	PP-DPO	PP-IPO
MR	C-HV	76.5(8.4)	63.3(5.2)	77.2(6.9)	70.6(9.1)	75.5(8.5)	80.9(4.3)	83.0(4.6)
	P-HV	54.4(11.1)	46.5(11.2)	55.3(12.0)	67.2(0.1)	61.0(5.1)	66.8(11.5)	68.3(7.1)
	Diff-HV	22.1	16.8	21.9	3.4	14.5	14.1	14.7
SST-5	C-HV	32.0(2.1)	27.4(1.3)	36.2(0.8)	31(3.7)	29.5(7.0)	34.1(2.5)	35.3(0.5)
	P-HV	23.6(3.1)	20.4(1.2)	24.4(2.5)	32.6(1.9)	29.9(1.3)	34.3(0.4)	34.0(0.5)
	Diff-HV	8.4	7.0	11.8	-1.6	<u>-0.4</u>	-0.2	1.3
Yahoo	C-HV	39.8(1.9)	27.5(1.2)	42.2(2.6)	40.7(3.9)	24.9(1.3)	47.1(1.7)	48.2(0.5)
	P-HV	21.2(4.8)	17.0(7.4)	28.3 5.4	31.5(4.2)	20.6(4.3)	31.3(5.2)	35.5(2.7)
	Diff-HV	18.6	10.5	13.9	9.2	4.3	15.8	12.7
Yelp-5	C-HV	35.5(2.8)	27.0(6.7)	40.4(2.6)	24.1(1.4)	40.0(3.7)	41.3(3.1)	40.6(2.1)
	P-HV	21.9(2.1)	19.7(4.8)	20.5(3.1)	35.1(2.5)	37.1(4.4)	39.7(2.5)	39.8(2.3)
	Diff-HV	13.6	7.3	19.9	-11.0	2.9	1.6	0.8



THREE-OBJECTIVE Text Style Transfer task

Three objectives:

- 1. Style
- 2. Content similarity
- 3. Prompt fluency



Comparative Performance Across Three-Objective Prompt Optimization Methods



ParetoPrompt

We prove that even without the specific objective value in the prompt optimization:

- ParetoPrompt achieves comparable performance with the algorithms relying on specific objective values.
- ParetoPrompt robust performance even when the training metric differs from the evaluation metrics used during testing.



Thank you

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