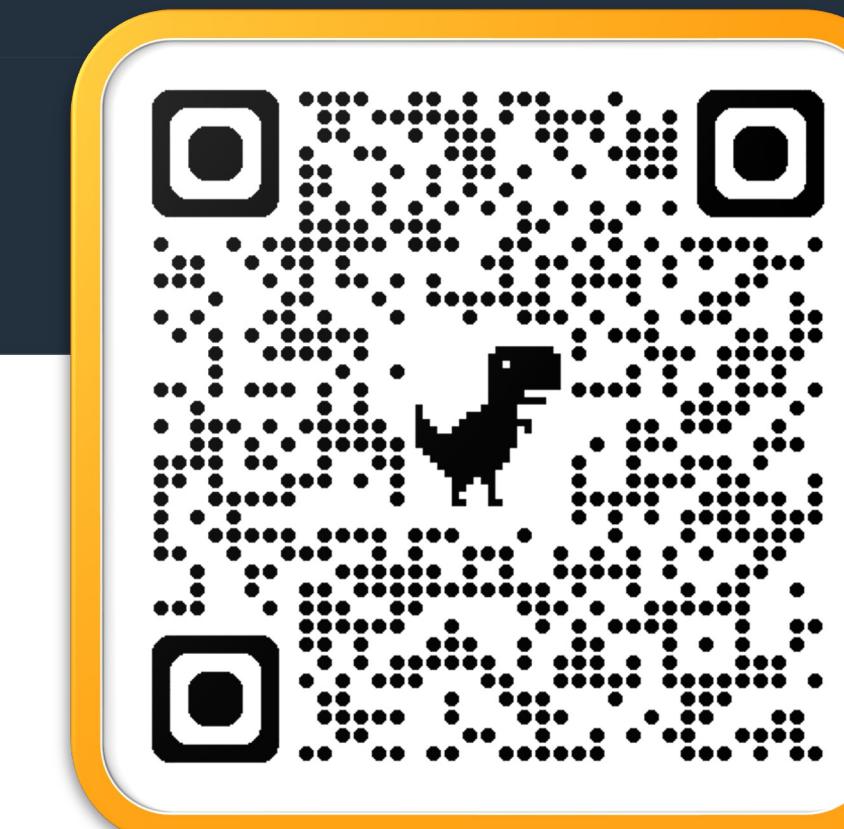


# CriSPO: Multi-Aspect Critique-Suggestion-guided Automatic Prompt Optimization for Text Generation

amazon | science

Han He\*, Qianchu Liu\*, Lei Xu\* [Equal Contribution], Chaitanya Shivade, Yi Zhang, Sundararajan Srinivasan, Katrin Kirchhoff  
hankcs, liufqian, leixx @ amazon.com Amazon AWS AI Labs



← code available!

## Background / Motivation

Writing good prompts for text generation is challenging

**Example of naïve prompt:**

*Please write a detailed clinical note summary for the input conversation.*

**Example of a good prompt:**

*Write a structured patient summary from the conversation. Your 300-350 word summary should reconstruct the clinical rationale by comprehensively addressing in each section:*

- <Chief Complaint>: Presenting symptom in 2-3 precisely worded sentences
- <HPI>: ...

Manually crafting good prompts is tedious, requires expertise, and doesn't scale well.

Current automated methods rely on numerical metrics to guide a search, but these approaches are less effective and lack robustness.

## Our Novelty

LLM generated Critiques and Suggestions

Use LLM to compare reference with generation, and provide multi-dimensional critiques and actionable suggestions.

**Example:**

Dimensions: length, professionalism

Critiques: (1) The generated text is shorter than reference.

(2) The generated text is written in layman words.

Suggestions:

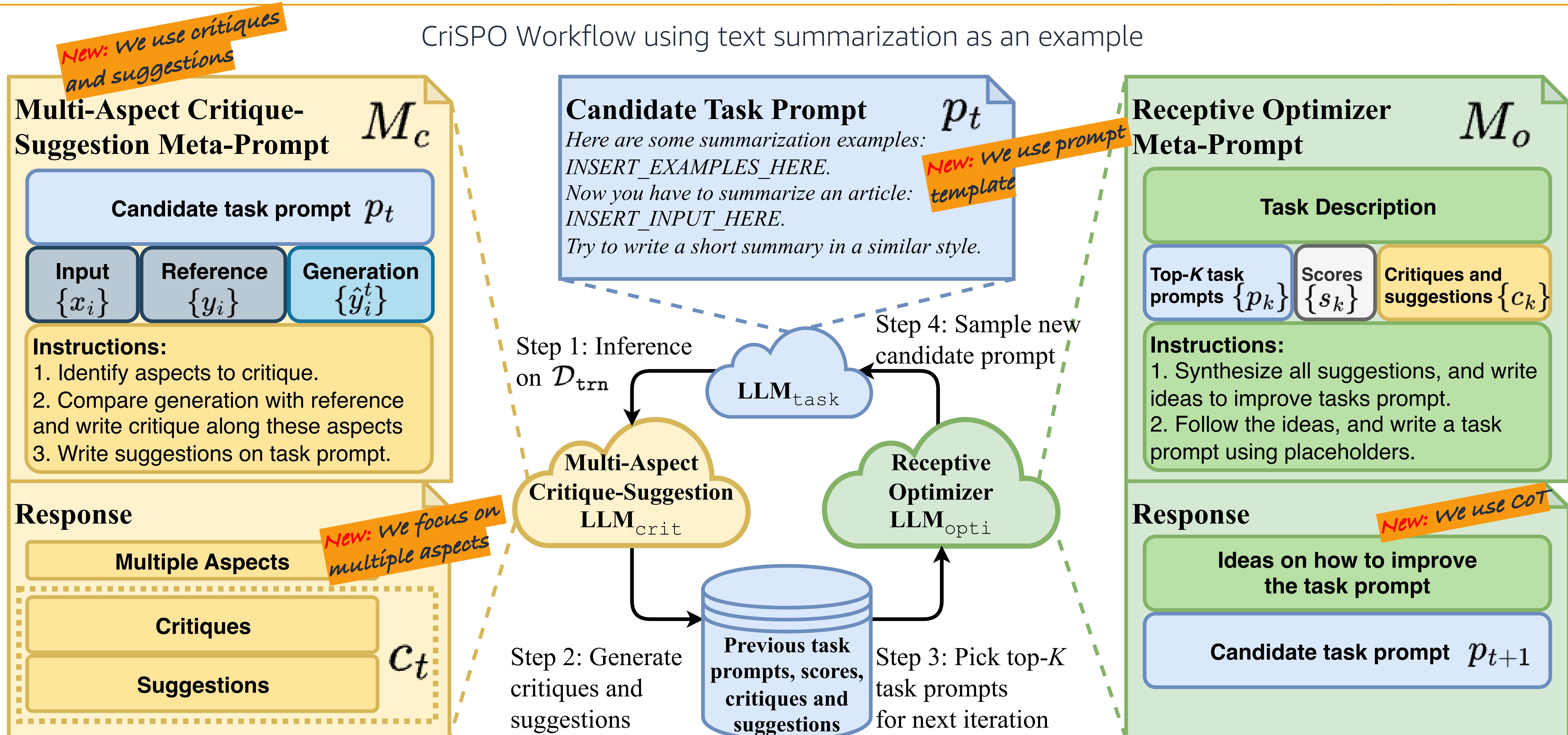
(1) Add "300-350 words" in prompt.

(2) Add "use medical language" in prompt.

Dimensions discovered by the LLM

Scope	Detail	Number_of_sentences
Focus	Length	Syntax
Concision	Word_Count	Conciseness
Style	structure	Recall
	Specificity	Detail_level
	Level_of_detail	Stylistic_elements
	Grammatical_structure	Number_of_words
	Number_of_characters	Specific_phrases

## CriSPO Workflow using text summarization as an example



## Experiments

We conduct thorough experiments on 4 summarization datasets, 5 QA/RAG datasets, and 2 other natural language generation (NLG) tasks to demonstrate the effectiveness. We achieve new SoTA on medical note generation dataset (ACI-Bench).

### Extension – multi-metric

Text generation usually involves optimizing multiple metrics. We first optimize for a primary metric, then use CriSPO to optimize a suffix to improve secondary metrics.

Best prompt found by CriSPO and OPRO (baseline) on SAMSUM dataset

**OPRO [Best Prompt]:** Generate a **one to two sentence** summary within the `<summary>` tags that concisely describes the key details of the conversation and **any conclusions** reached.

`INPUT_DOC`

**CriSPO [Best Prompt]:** The text below contains a discussion expressing several **key facts and events**. Your concise **1-sentence** summary should relate only the **2 most** important pieces of information stated, **without assumptions or extra context**.

`INPUT_DOC Write the summary within <summary> tags.`

**OPRO [Example Output]:** Ralph asked Andrew if he heard a Polish joke, then told a joke about sinking a Polish battleship by putting it in water. Andrew responded that the joke was terrible and so unfunny that it made his mouth dry, requiring a sip of water.

**CriSPO [Example Output]:** Ralph tells Andrew a Polish battleship joke that Andrew finds unfunny.

**[Reference]:** Ralph told Andrew a joke.

CriSPO shows high prompt diversity

Dataset	Length↑	Vocab↑	ROUGE-L↓	Cosine↓
CNN				
OPRO	41±6	36±5	57.5	0.93
CriSPO	<b>149±24</b>	<b>96±12</b>	<b>50.3</b>	<b>0.90</b>
MeetingBank				
OPRO	31±5	28±4	44.9	0.84
CriSPO	<b>216±41</b>	<b>135±19</b>	<b>39.7</b>	<b>0.80</b>
SAMSUM				
OPRO	34±6	30±5	57.0	0.94
CriSPO	<b>172±22</b>	<b>112±12</b>	<b>46.0</b>	<b>0.88</b>
D2Note				
OPRO	58±11	46±8	62.7	0.95
CriSPO	<b>247±40</b>	<b>117±13</b>	<b>54.3</b>	<b>0.93</b>

We achieve an average of **10%** ROUGE-1 and ROUGE-L improvement. Check our paper for details!