RAG-HAT: A Hallucination-Aware Tuning Pipeline for LLM in Retrieval-Augmented Generation



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• Detection: Sentence-Level Interpretable Hallucination Detector



Hallucinated Non-hallucinated **Detection Model** Guide **Rewriting: Utilize GPT to Do Answer Rewrite Defensive Advice** Interpretations **Detected Hallucinated Content** --- { "review_stars": 5.0, ... According to their "review_date": "2021-12-17 02:46:28", structured data, they offer a "review_text": "Beautiful views, great variety of beers and a Honey beer, and family friendly. Highly GPT-4 Avocado Ale that is **popular** recommend the Honey among customers ... Avocado Ale ... Some might argue that the claim of the Honey

• Mitigation: Overrly Cautious Penalization



ARTICLE

Teaching yourself can be done by learning something new with each task you complete, as mentioned in passage 1. However, taking notes is not always necessary. To learn a language quickly, studying or working abroad can be helpful, as mentioned in passage 2. Wikipedia ...



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Mitigation: Conduct DPO Training





JSON data does not explicitly state its popularity ...

Avocado Ale being "popular among customers"

could be considered a hallucination since the

Figure 1: An Example of Defensive Advice: The LLM made a minor extension partially based on the provided references. Defensive advice highlights that the statement is not well supported.

Preference

Dataset

Model Hallucinates Less

Evaluations

DATASET	METHOD	Detector	GPT-4 Turbo	Human	Average
RAGTruth Test Set	Qwen	36.9(-)	51.3(-)	34.4(-)	40.9(-)
	Qwen(Regenerate)	-	44.2(↓13.8%)	-	44.2(↓13.8%)
	RAG-HAT	$22.7(\downarrow 38.5\%)$	41.3(↓ 19.5%)	25.7(↓ 25.3%)	29.9(↓ 26.9 %)
WebGLM 1000	Qwen	21.3(-)	46.7(-)	-	34(-)
	Qwen(Regenerate)	-	38.8(↓17.0%)	-	38.8(↓17.0%)
	RAG-HAT	12.0(↓ 43.7%)	37.9(↓ 19.0%)	-	24.9(\26.8%)

DATASET	METHOD	GPT-4 Turbo	Human
RAGTruth Dataset	Qwen RAG-HAT	41.1 57.3	33.2 40.8
WebGLM 1000	Qwen RAG-HAT	39.5 58.5	-

Table 5: Answer Quality Win Rates: 1,000-Example WebGLM Set and RAGTruth Test Set

Table 3: Hallucination Rate: 1,000-Example WebGLM Set and RAGTruth Test Set (Total 450 Examples): Our detection model cannot fairly benchmark the hallucination rate of the regeneration approach since it serves as the trigger for regeneration.