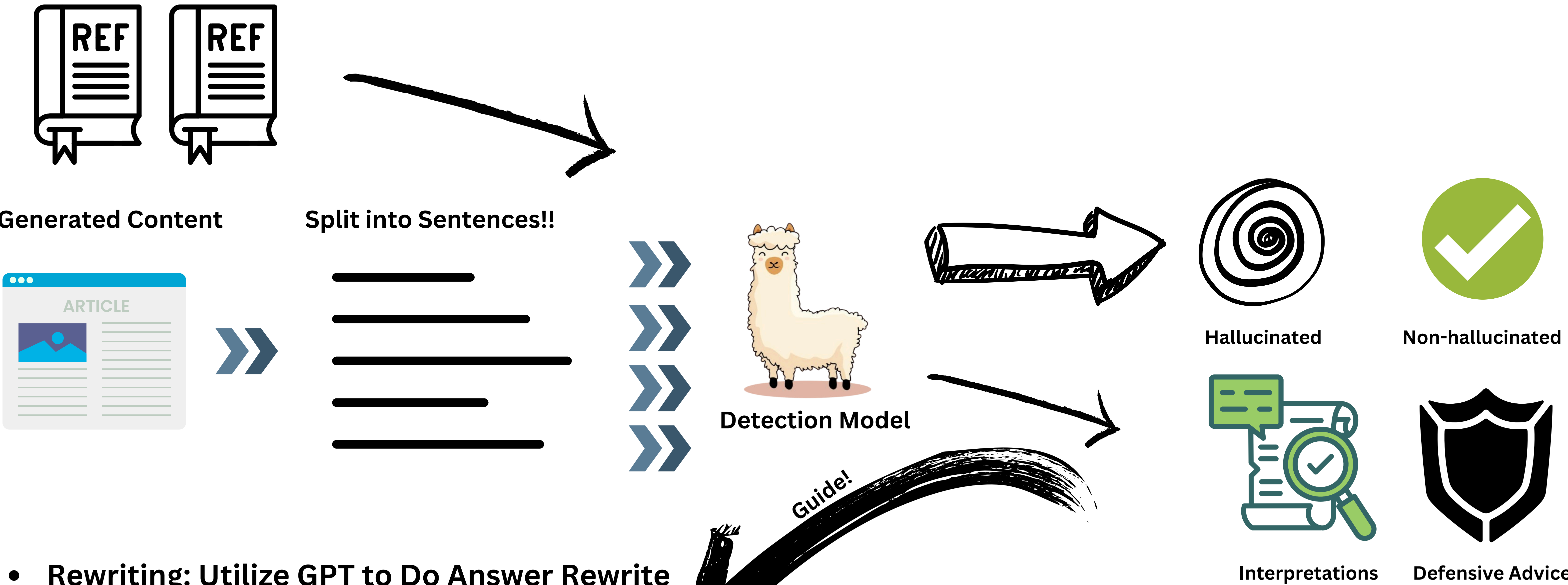


RAG-HAT: A Hallucination-Aware Tuning Pipeline for LLM in Retrieval-Augmented Generation

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• Detection: Sentence-Level Interpretable Hallucination Detector



• Rewriting: Utilize GPT to Do Answer Rewrite



• Mitigation: Overly Cautious Penalization

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• Mitigation: Conduct DPO Training



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.

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(↓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%)

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.

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

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