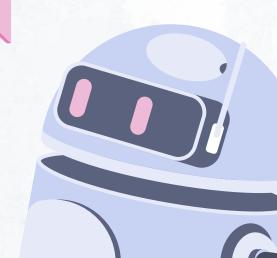
The Impact of **Embedding Methods on** Faithfulness of Retrieval-Augmented Generation (RAG)





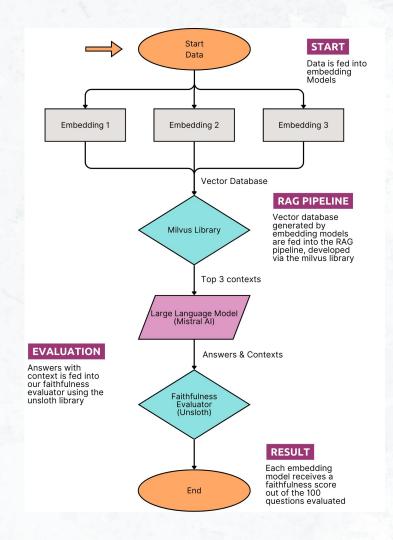
Chang Jin, Reagan Keeney, and Amandeep Kaur Singh

Problem Statement

- Different embedding models are designed with specific goals in mind
 - Embeddings vary on effectiveness
 - No one-size-fits-all solution; performance is highly context-dependent.
- Evaluating RAG systems is a challenging task
- Faithfulness
 - Faithfulness reflects how well retrieved contexts are utilized in LLM responses. Measures "hallucination".
 - Ensuring accurate incorporation of context is critical for reliable outputs.

Pipeline + Dataset

- RAG Mini BIOASQ Dataset
- Embedding Models:
 - Baseline:bge-small-en-v1.5
 - MedEmbed-small-v0.1 (fine-tuned BERT)
 - o All-MiniLM-L6-v2
- Milvus Vector DB
- Mistral Al LLM
- Evaluation



Faithfulness Overview

- Metric for evaluating output
 - Non-traditional- unique to RAG
- Number of Truthful Claims/Total Number of Claims



Results & Discussion



Model	Faithfulness
bge-small-en-v1.5(baseline)	97%
MedEmbed-small-v0.1	93%
all-MiniLM-L6-v2	92%

Thank You!

