# RAG-powered Domain Specific/Time Sensitive Q&A Application

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#### INTRODUCTION

Given their exceptional text generating capability, LLM-backed Q&A systems are gaining significant popularity. However, they are limited by their training data which may not be domain-specific or up-to-date.

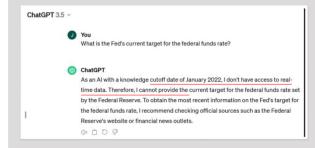


Fig 1. Result from ChatGPT

#### **SOLUTION**

#### **Retrieval-Augmented Generation**

- ✓ Support input of user-defined knowledge bases
- ✓ Retrieve relevant contexts
- ✓ Generate desirable answers

## **ARCHITECTURE** Source Document Tokenizer Text Chunks Retriever Top-k Chunks Question Generator Answer Fig 2. Architecture of this application Tokenizer: SpaCy Retriever: Sentence-Transformers + Cosine Similarity Generator: I lama 2 **INTERFACE** RAG: Please provide document source. Federal Reserve Press Release March 20.pdf RAG: What do you want to know about this document? What is the Fed's current target for the federal funds rate? RAG: According to the passage, the Fed's current target for the federal funds rate is between 5-1/4 and Fig 3. User interface of this application

#### **EVALUATION**

☐ Generated answers compared to ground truth

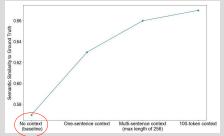


Fig 4. Performance of this application

Note: the evaluation was based on question-context-answer triplets for Apple's Annual Earnings Report 2022.

#### CONCLUSIONS

- This RAG-framed application can improve the Q&A performance
- Length of context matters a lot
- Future work will be focused on automating the process of analyzing source documents and deciding the right length of context.

#### **REFERENCES**

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