RAG-based Al Application that answers questions about yourself!

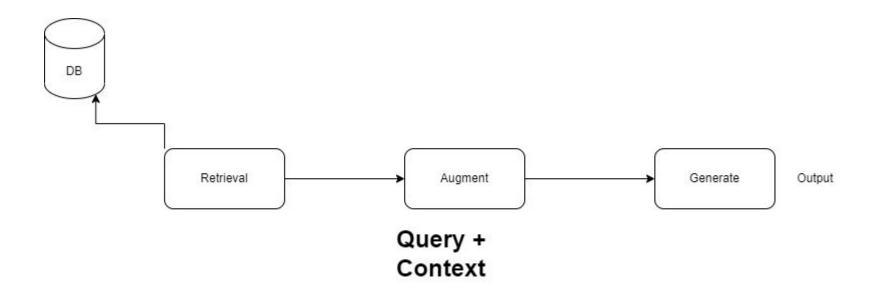
Problem Objective

Develop a Retrieval-Augmented Generation (RAG) based AI system capable of answering questions about yourself. The system should handle inquiries in English and manage follow-up questions effectively.

Solution Pipeline

- Data ingestion
- RAG using Langchain/ LLamaIndex

High level system design



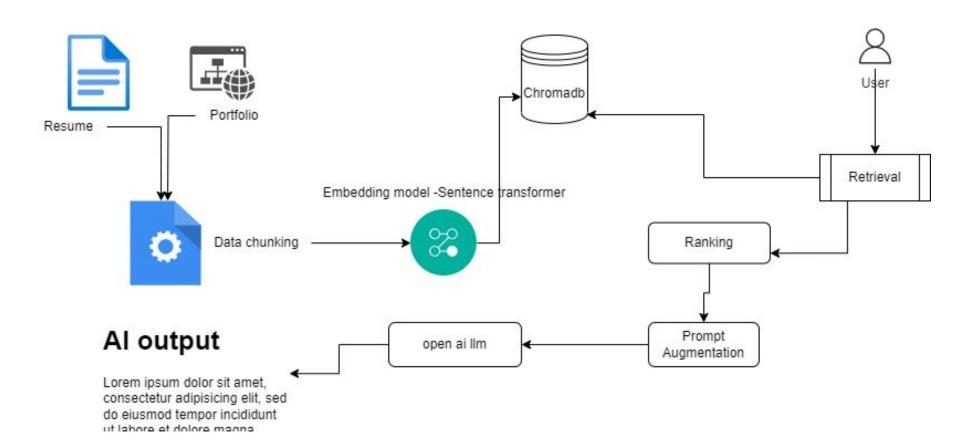
High level system design

- 1. **Retrieval Engine**: Searches vast databases for relevant data using advanced algorithms, ensuring retrieval of the most pertinent information.
- 2. **Augmentation Engine**: Integrates retrieved data with input queries, enriching context for generating more informed responses.
- 3. **Generation Engine**: Formulates coherent and contextually relevant responses based on augmented input, utilizing sophisticated language models to enhance understanding.

High level architecture

Data from resume, portfolio site, text

| Data from resume, portfolio site, text | Data from resume, portfolio site, text | Data from the data chunks | Data from the data from the data chunks | Data from the data from the data chunks | Data from the data from the data chunks | Data from the data from the data chunks | Data from the data from the data chunks | Data from the data from the data chunks | Data from the data chunks | Data from the data from the data chunks | Data



How the system works?

- We collect user data like resume, their portfolio site and use langchain based document loader, web based loader to read the contents of the user data.
- Then we use langchain text splitter to split the text documents to multiple chunks with overlaps
- The chunks of data are converted to embeddings using fastembed/sentence transformer and uploaded to local chroma db.
- We initialize a retriever using the vector db and keyword based search(BM25) and combine as hybrid retriever.

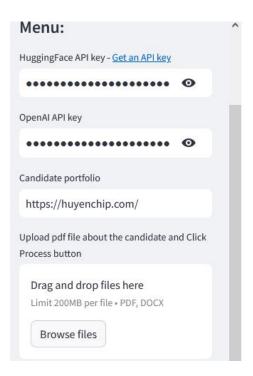
How the system works?

- We then use fastrank reranker to rerank the results from retriever.
- We then create a retrieval chain with memory to remember the pervious contexts so that the user can follow up with the questions
- When the user enters a query, the query is converted to embeddings and using retriever - we search based on vector similarity and keyword similarity and return the list of context which are relevant to the user query
- This context and query with memory history is augmented to prompt and passed to LLM for result generation.

Future experiments

- 1) Try different embeddings model
- 2) Play around with the reranking model, num of documents to be retrieved
- 3) Use LLM to eval if the retrieval for the query is correct or not
- 4) After the output generation, store the results as dataframe and use another LLM to eval the results. Exploring phoenix by arize ai.

Screenshots



Get information about the candidate from the resume

Ask questions..

describe about her

Chip Huyen is a writer and computer scientist who helps companies deploy machine learning into production. She has worked on GPU-native data processing and open data standards at Voltron Data, and has also built machine learning tools at companies like NVIDIA, Snorkel AI, and Netflix. Chip Huyen has founded Claypot AI, which was acquired. She has a background in teaching machine learning systems design at Stanford University and has authored a book called "Designing Machine Learning Systems." Additionally, Chip Huyen is working on a new book titled "AI Engineering," expected to be released in late 2024.

