Retrieval-Augmented Generation (RAG) Chatbot



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Introduction

- In the ever-evolving digital landscape, Multi-Modal Al is revolutionizing the way we interact with information.
- By integrating various data types text, images, audio, and more.
 Multi-Modal Al systems enable more sophisticated and context-aware responses.
- This presentation delves into the innovative application of Multi-Modal AI in enhancing Retrieval-Augmented Generation (RAG) systems, ultimately driving more intelligent and dynamic user experiences.



PROJECT GOALS

1. Multimodal Chatbot purpose mainly as an assistant

2. Chatbot must fetch the relevant text and images from the knowledge base for the user query and present it, with minimal latency and hallucination



TECH STACK

Language: Python

Technology: RAG

Embedding model: text-embedding-004

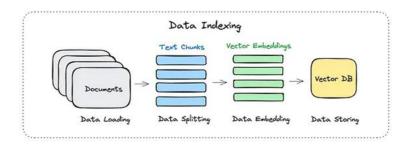
image model: gemini-2.0-flash

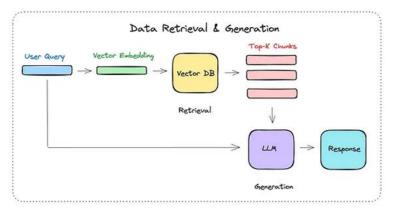
Chat model: Llama3-70b-8192

image processing: OpenCV

Vector DB: Pinecone Db

Note: No frameworks like langchian is used.





RAG PIPELINE

Vector Embedding : They are numerical representations of data points.

Vector Database: It is a specialized data management system designed to store, manage, and index high-dimensional vector embeddings.

Chunking: A process by which small individual pieces of information are bound together to create a meaningful whole in memory.

LLM: A large language model (LLM) is a type of artificial intelligence (AI) system capable of understanding, generating, and processing human language by training on vast amounts of text data.

Prompt: The input or instruction given to the model to elicit a specific response or output.

Project Overview

- This Retrieval-Augmented Generation (RAG) Chatbot uses Django framework to build a full-stack web application.
- Implemented a MultiModal RAG system, that queries from different document types like pdf, docs,
 csv, image and video, used different processing techniques for different document types.
- Knowledge Base:
 - O Students Grading Dataset [CSV File]
 - O Data Incident Management [PDF File]
 - O Luxury Cars Image Dataset [Image Files]
 - O Linear Regression in 3 Minutes [Video File]
 - O Introduction to Machine Learning [DOC File]
- The user can also upload their own data into the system and can be queried from that also.

User Query Processing

- This is handled by rag_bot.py [rag\Utils\rag_bot.py]
- The RAGbot class has two methods retrieve_context() and normal_response()
- The retrieve_context() will take the user query as its input, the user query is embedding and searched in pinecone for all the related context and the function returns top 4 the matches.
- The normal_response() will take the user query and matches from the db and extracts all the context from the matches and feeds it to the Ilama3-70b-8192 used via groq API to generate response for the user query. Used prompt engineering for optimal response.

Csv Query Processing

- If a csv content is present in the related context for the user query it will be processed differently.
- The summary of the csv along with the user query will be given to a LLM which will produce a pandas code for the user query, which will be executed and the pandas output is stored.
- The pandas code, pandas output along, csv summary along with the user query will be given to another LLM which will give a final response to the user query.

Insert file Processing

Each file type in handled differently to extract the context. Every file processing module is written as a class for modularity.

PDF File handling:

- This is handled by process_pdf.py [rag\Utils\process_pdf.py]
- Main function extract_content() and process_pdf()
- The extract_content() read the pdf and extracts all the text, tables and images.
- The process_pdf() chunks the content in the file each image and tables is a separate chunk and the text is splited using _split_text() function with chunk_size=1000, chunk_overlap=300
- Finally all these chunks are embedded upserted along with some metadata into vector DB (pinecone)

Docx File handling:

- This is handled by process_doc.py [rag\Utils\process_doc.py]
- Main function extract_content() and process_documents()
- The extract_content() read the Doc or txt file and extracts all the text, tables.
- The process_documents() chunks the content in the file each image and tables is a separate chunk and the text is splited using _split_text() function with chunk_size=1000, chunk_overlap=200
- Finally all these chunks are embedded upserted along with some metadata into vector DB (pinecone)

Csv File handling:

- This is handled by process_csv.py [rag\Utils\process_csv.py]
- Main function generate_csv_summary() and process_csv()
- The **generate_csv_summary()** read the csv file and extracts all the important information.
- The process_csv() create embedding for the csv summary and it is upserted into the vector
 DB (pinecone)

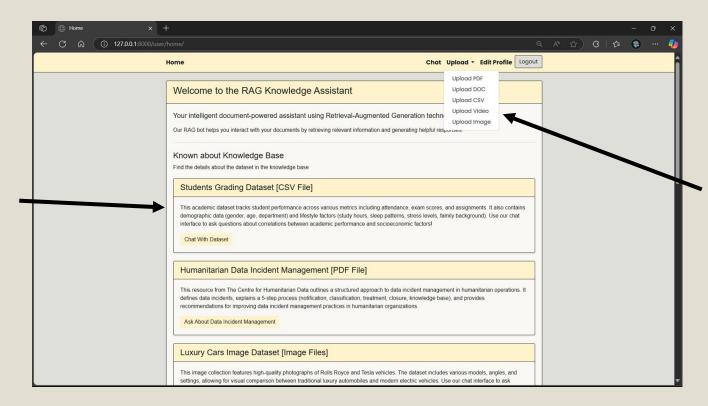
Image File handling:

- This is handled by process_image.py [rag\Utils\process_ image.py]
- Main function generate_image_summary() and process_image()
- The optimize_image() is used to reduce the size of the image for storing, the generate_image_summary() is used to create a summary for the image using the gemini-2.0-flash which is the optimal model for image summarization.
- The image summary is embedded and upserted in the vector DB along with the metadata which contains the base64 encoding of the image file. This base64 encoding will be later used for displaying the image when required.

Video File handling:

- This is handled by process_video.py [rag\Utils\process_video.py]
- The main functions process_video() and process_frames_with_transcript() handle video processing.
- The extract_frames() is used to extract key frames from the video, while transcribe_audio() is used to extract audio transcribe and these both are given to process_frames_with_transcript() which creates context with gemini-2.0-flash using both audio and frames.
- Finally all these are embedded upserted along with some metadata into vector DB (pinecone).

Outputs



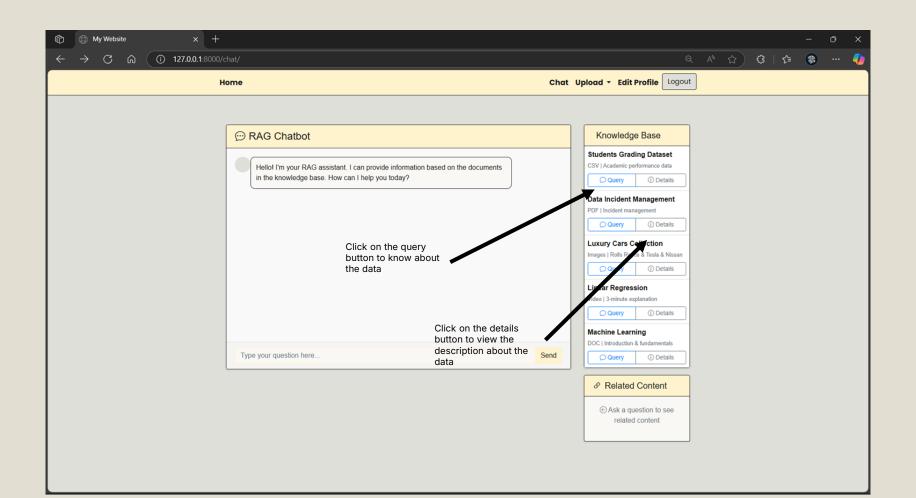
Description of the

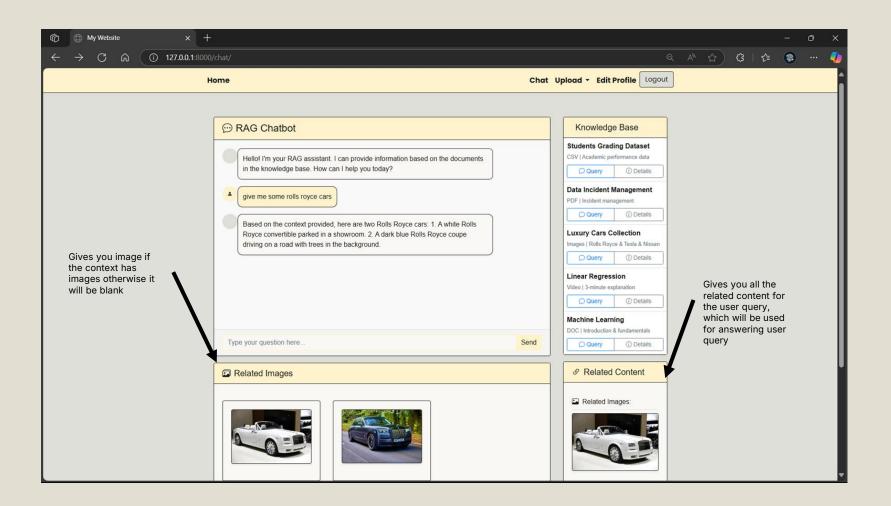
knowledge base is

data in the

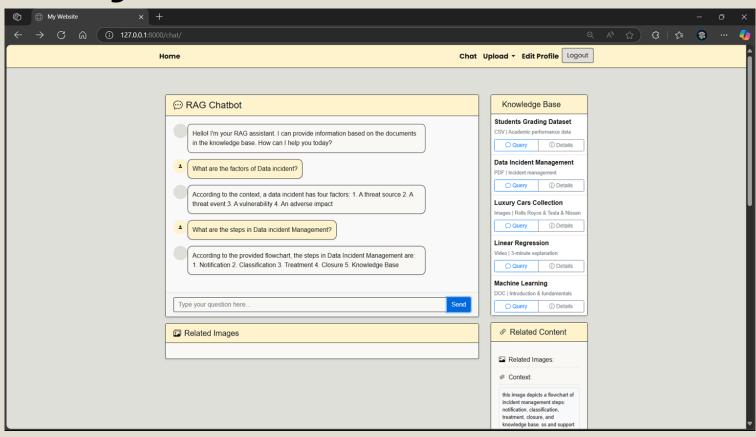
given

Different types of documents can be upload here, which will be processed and stored in pinecone (Vector DB)

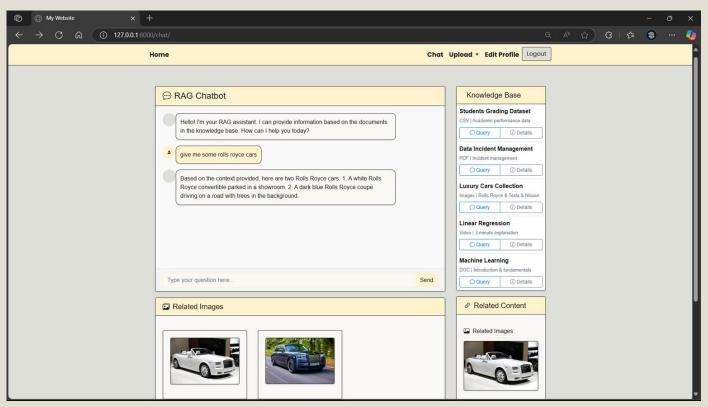




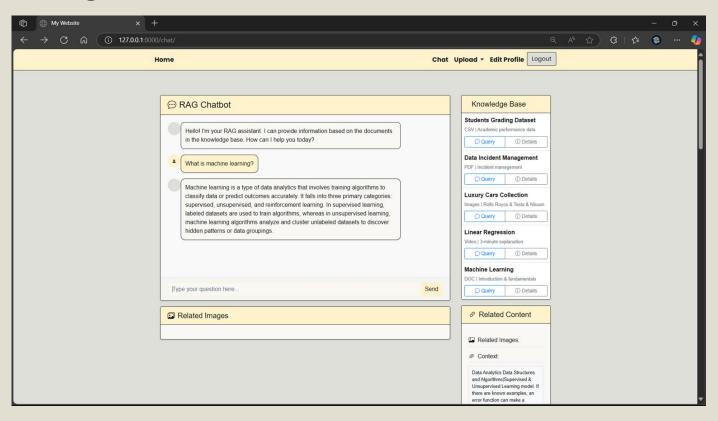
Query from Pdf Data



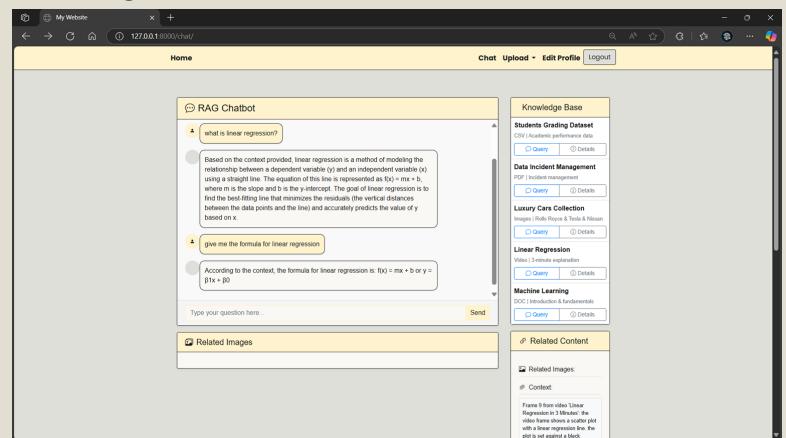
Query from Image Data



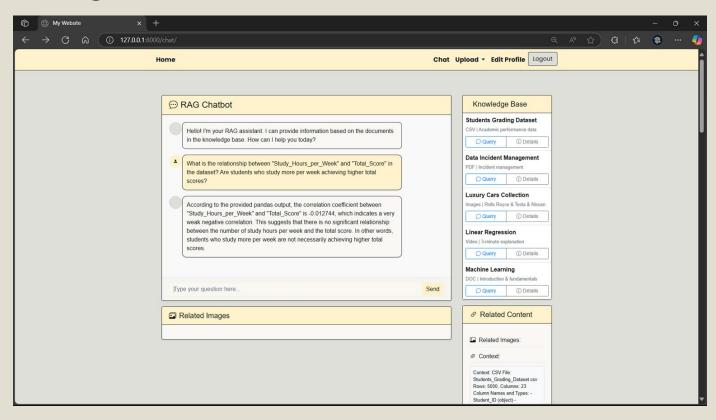
Query from Docx Data



Query from Video Data



Query from CSV Data



Thank you