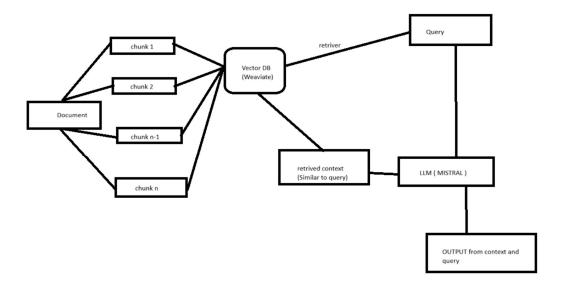
RAG PIPELINE DOCUMENTATION (PART - 1)

MODEL ARCHITECTURE:

- 1) DATA LOADER > Using Langchain PDFLOADER
- 2) EMBEDDING AND STORING IN DB: using weaviate db instance for cloud storage and efficient retrieval
- 3) RETRIVAL OF RELEVENT CHUNKS: passing query to database to retrieve relevant chunks.
- 4) Creating Prompt.
- 5) LLM: Mistral AI for Output generation using prompt which provide the context along with query

ARCHITECTURE:



Retrieval Process: Graph based Approximate Nearest Neighbour Search

Response generation:

The retrieved context along with the query in form of prompt is sent to LLM which generates the response according to the document relevance.

Sample queries and output

1)

```
output = rag_chain.invoke("what is the overall assignment work")
index = output.find('Answer_rag')
print(output[index:])

✓ 3.2s

Python

Answer_rag :

The overall assignment work involves developing a Retrieval-Augmented Generation (RAG) model for a Question Answering (QA) bot problem statement. The m

The task requirements include implementing an RAG-based model, using a vector database like Pinecone DB to store and retrieve document embeddings effic
```

2)

```
output = rag_chain.invoke("Do you know how RAG works ?")
index = output.find('Answer_rag')
print(output[index:])

✓ 3.0s

Answer_rag:

RAG stands for Retrieval-Augmented Generation. It is a model for question answering (QA) bot that uses a vector database like Pinecone DB and a generat

The RAG-based model can handle questions related to the provided document or dataset
```

3)

4)

```
output = rag_chain.invoke(" How much stipend will I be paid ?")
index = output.find('Answer_rag')
print(output[index:])

    1.5s

Answer_rag :
I don't know. The context does not provide information about the stipend amount.
```