END-TO-END RETRIEVAL-AUGMENTED GENERATION WITH AMAZON BEDROCK

A Seamless Data Extraction and Response System

Github-Link

Project Overview

Objective:

Develop a comprehensive RAG system using Amazon Bedrock, focused on efficient information retrieval and natural language processing.

INTRODUCTION TO AMAZON BEDROCK

- What is Amazon Bedrock?
- A managed service that provides access to state-of-the-art foundation models for generative AI.
- Offers pre-trained models from various AI providers, including those optimized for different use cases like text generation, summarization, and more.
- Benefits:
- Ease of Use: Simplifies the integration of advanced AI models into applications.
- Scalability: Automatically scales resources to meet demand.
- Security & Compliance: Provides robust security measures and compliance certifications.

Project Workflow

- Data Source: Start with a PDF file.
- Data Extraction: Extract and preprocess text from the PDF.
- Text Chunking: Split the extracted text into smaller chunks.
- Embedding Generation: Use Amazon Bedrock models to convert text chunks into vector embeddings.
- Vector Database Storage:
- Database Used: FAISS (Facebook AI Similarity Search)
- Purpose: Efficiently store and search for vector embeddings based on similarity.

Query Handling

- User Query Input: User submits a query.
- Similarity Check: Embedding model retrieves similar chunks from FAISS.
- Ranked Results: Retrieves and ranks similar results.

Language Model Interaction

- LLM Query: The query is sent to an LLM via Amazon Bedrock.
- Comparison & Matching: LLM compares the query with the retrieved embeddings.
- Result Selection: LLM selects the best matching response.

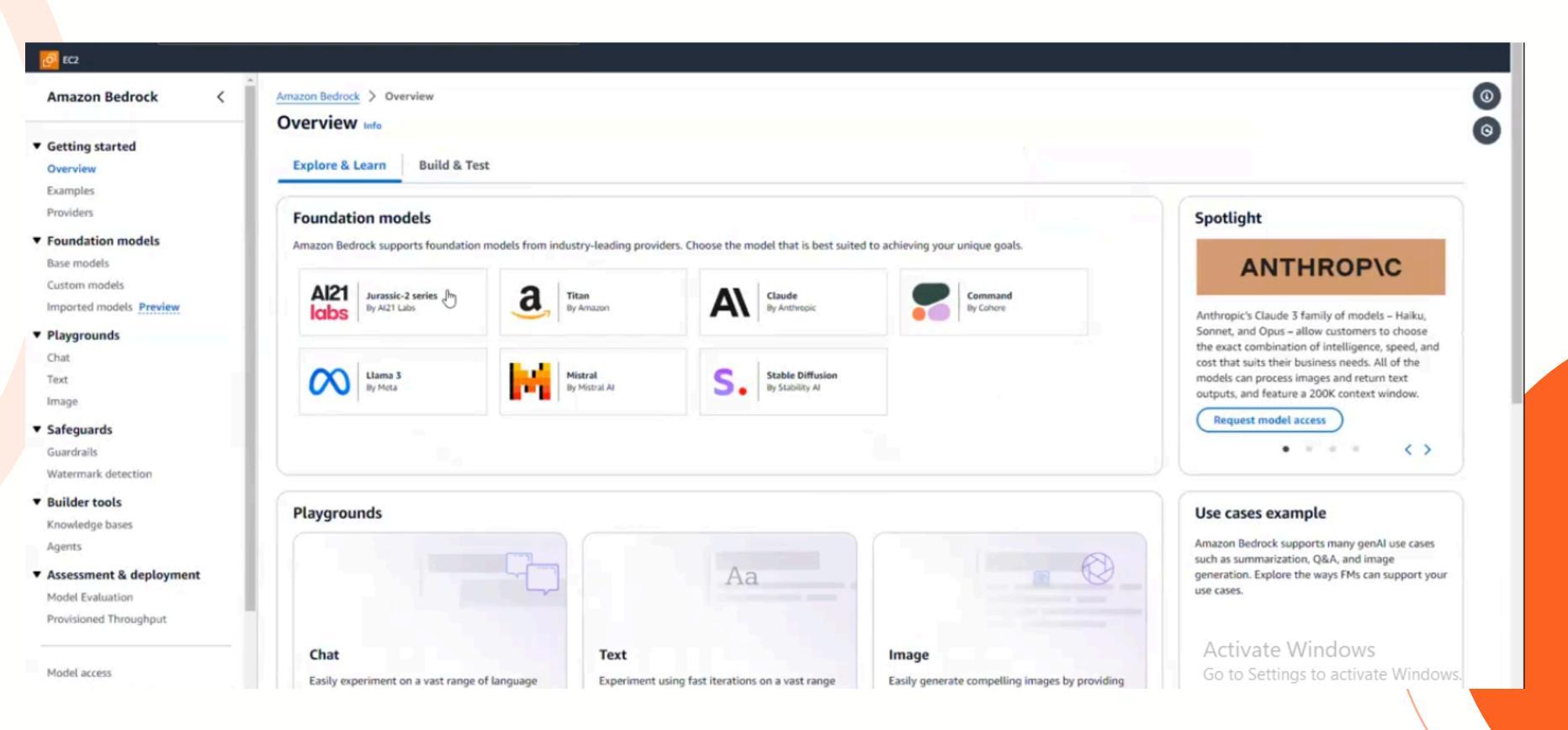
TECHNOLOGIES & TOOLS

- Amazon Bedrock: Central to utilizing pre-trained models.
- Vector Database: FAISS for storing and retrieving vector embeddings.
- Previous & Current Models:
- Previous: Sentence Transformer, Mistral-7B-Instruct (Hugging Face)
- Current: Mistral model from Amazon Bedrock
- Deployment Environment: EC2 instance, Streamlit app, and additional AWS services.

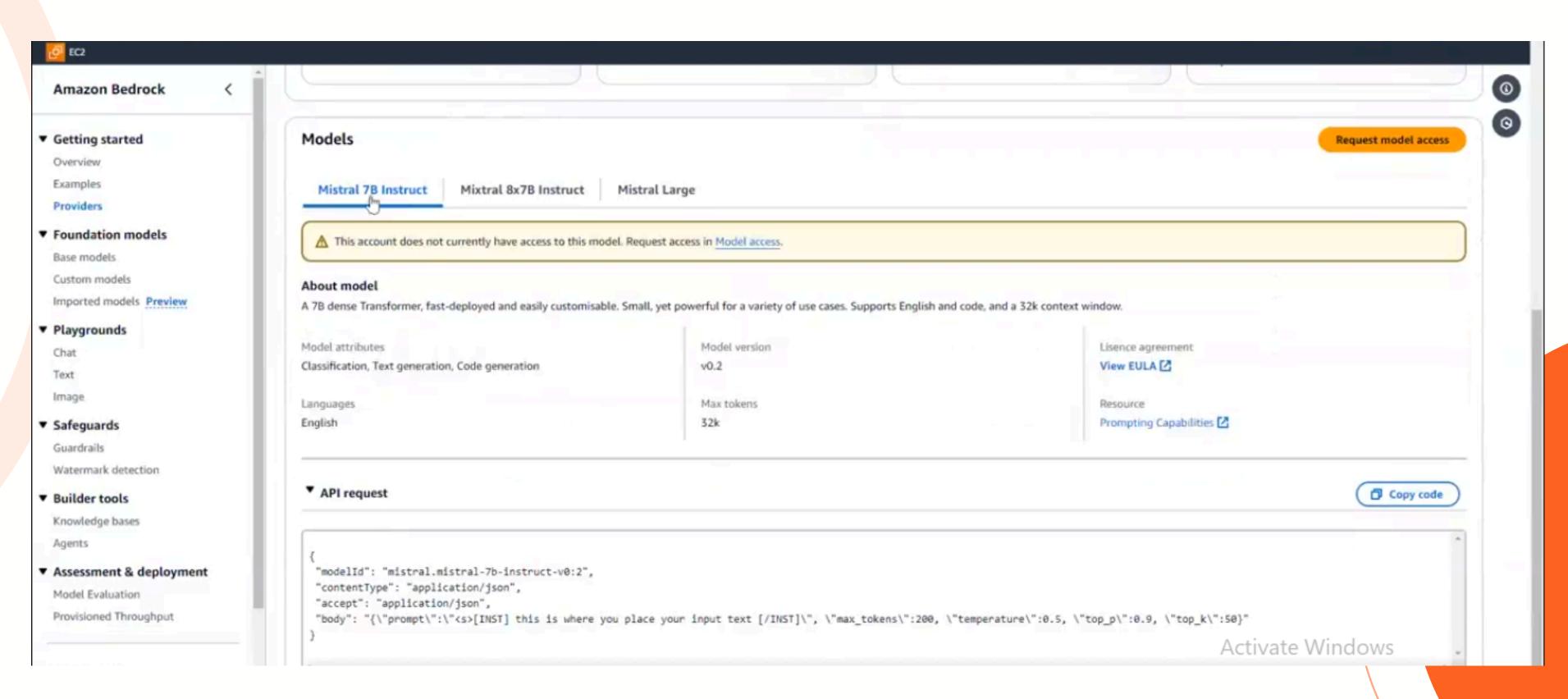
IMPLEMENTATION DETAILS

- Environment Setup: Conda environment and package installation.
- Deployment Process: EC2 instance setup, repository cloning, and app deployment.

WORKING WITH AMAZON BEDROCK



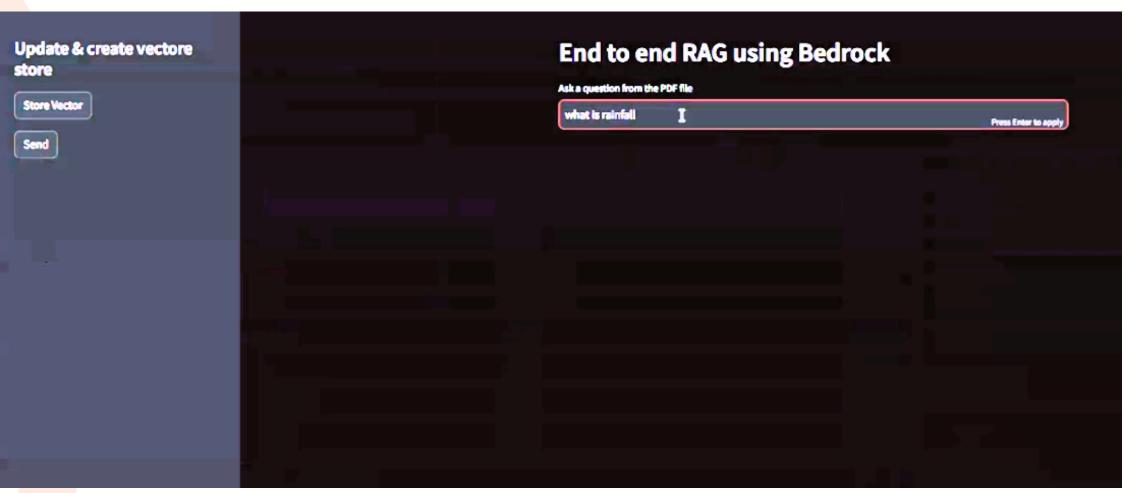
WORKING WITH AMAZON BEDROCK



SOURCE CODE

```
bedrock = boto3.client(
    service_name = "bedrock-runtime",
    region_name = region_name,
    aws_access_key_id = aws_access_key_id,
    aws_secret_access_key = aws_secret_access_key,
#Get embeddings model from bedrock
bedrock_embedding = BedrockEmbeddings(model_id="amazon.titan-embed-text-v1", client= bedrock)
def get documents():
    loader = PyPDFDirectoryLoader("Data")
    documents = loader.load()
    text_spliter = RecursiveCharacterTextSplitter(
                                        chunk_size=1000,
                                        chunk_overlap=500)
    docs = text_spliter.split_documents(documents)
    return docs
def get_vector_store(docs):
   vectorstore_faiss = FAISS.from_documents(
        docs,
        bedrock_embedding
   vectorstore_faiss.save_local("faiss_local")
def get_llm():
    11m = Bedrock(model_id = "mistral.mistral-7b-instruct-v0:2", client = bedrock)
```

RESULTS



i asked my model ..what is rainfall?

this is what my model replied....!



