Retrieval-Augmented Generation (RAG) Chat Application

Objective

The objective of this assignment is to build a RAG-based chatbot capable of accurate and contextually relevant responses, using Python and LangChain tools.

Code Documentation

Module-Level Documentation

Module Name: rag chatbot

Description: Implements a Retrieval-Augmented Generation (RAG) pipeline using LangChain, FAISS, and Hugging Face.

Dependencies:

- transformers
- langchain
- faiss-cpu
- sentence-transformers
- huggingface-hub

Class and Function Documentation

Hugging Face Login

```
from huggingface_hub import login

def login_to_hugging_face(api_token):

"""

Logs into the Hugging Face platform using the provided API token.

Args:

api_token (str): API token for authentication.

Returns:

None

"""

login(token=api_token)

print("Logged in to Hugging Face successfully!")
```

Document Loader and Splitter

```
from langchain.document_loaders import TextLoader from langchain.text_splitter import RecursiveCharacterTextSplitter
```

def load_and_split_documents(file_path, chunk_size=500, chunk_overlap=100):

```
Loads a text file and splits its content into chunks.
Args:
    file path (str): Path to the document file.
    chunk size (int): Size of each chunk in characters.
    chunk_overlap (int): Overlap between consecutive chunks.
Returns:
    list: A list of split document chunks.
  loader = TextLoader(file_path)
  documents = loader.load()
  text_splitter = RecursiveCharacterTextSplitter(chunk_size = chunk_size, chunk_overlap =
chunk_overlap)
  return text_splitter.split_documents(documents)
Embedding and Vector Store
from langchain.vectorstores import FAISS
from langchain.embeddings import HuggingFaceEmbeddings
def create_vector_store(documents, model_name="sentence-transformers/all-MiniLM-L6-v2"):
  Creates a FAISS vector store from a list of documents using Hugging Face embeddings.
  Args:
    documents (list): List of document chunks.
    model name (str): Name of the Hugging Face embedding model.
  Returns:
    FAISS: A FAISS vector store instance.
  embedding model = HuggingFaceEmbeddings(model name=model name)
  vector_store = FAISS.from_documents(documents, embedding_model)
  return vector_store
```

Deliverables

- 1. Python Code:
 - o File Name: rag_assignment.ipynb
 - o Includes:
 - Document Loading
 - RAG Pipeline Implementation
 - Hugging Face Authentication
 - Commented and documented code for clarity.

2. Sample Questions and Responses:

- o File Name: responses.txt
- o Contents: Sample user questions and corresponding chatbot responses.
- 3. GitHub Repository:
 - o Repository Link: GitHub Repository
 - Contents:
 - Code (rag_assignment.ipynb)
 - responses.txt
 - Readme
- 4. Optional: Streamlit Deployment:
 - o Hosted Link: Streamlit Application

Installation and Usage

Prerequisites

- Python 3.8 or higher.
- Libraries:
 - o langchain
 - faiss
 - pandas
 - streamlit

Steps to Run

- 1. Clone the GitHub repository:
- 2. git clone <repository_link>
 - cd <repository_name>
- 3. Install dependencies:
 - pip install -r requirements.txt
- 4. Launch the chatbot (if deployed with Streamlit):
 - streamlit run app.py
- 5. Interact with the chatbot via the web interface

References

- LangChain Documentation: https://langchain.com/
- FAISS Documentation: https://faiss.io/
- Python Documentation: https://docs.python.org/3/