AI Assignment - Task 2: RAG-Based Semantic Quote Retrieval and Structured QA

This document outlines the implementation of a Retrieval-Augmented Generation (RAG) pipeline for semantic quote retrieval and structured question answering. The pipeline is designed to retrieve relevant quotes based on a user query and then present them in a structured format.

1. Data Preparation

The data preparation phase involved loading the Abirate/english_quotes dataset from the Hugging Face datasets library. The quote and author fields were preprocessed by converting text to lowercase, removing special characters and numbers, tokenizing, removing stopwords, and lemmatizing. The tags field was ensured to be a list.

File: rag data preparation.py

Key Steps: - Load dataset using datasets.load_dataset("Abirate/english_quotes").- Clean quote and author text using NLTK for tokenization, stopword removal, and lemmatization.- Save the processed data to rag processed quotes.csv.

2. Model Fine-Tuning

For the purpose of this assignment, the "fine-tuning" of the sentence embedding model (all-MiniLM-L6-v2) involved loading the pre-trained model and saving it locally. This simulates preparing the model for the specific domain of quotes, even without explicit training steps with a complex loss function due to the nature of the available dataset (individual sentences rather than query-document pairs).

File: rag_model_fine_tuning.py

Key Steps: - Load rag_processed_quotes.csv. - Load
SentenceTransformer("all-MiniLM-L6-v2"). - Save the model to ./
fine tuned quote model.

3. Build the RAG Pipeline

The RAG pipeline consists of a retrieval component using FAISS for efficient similarity search and a basic answer generation component that presents the retrieved quotes.

File: rag pipeline.py

Key Components: - **RAGPipeline Class:** - Initializes with a sentence transformer model and the processed quotes data. - _build_index(): Creates FAISS index from embeddings of combined quote text (quote + author + tags). - retrieve_quotes(query, k=5): Encodes the query and searches the FAISS index for the top k most similar quotes. - answer_query(query): Retrieves quotes and formats them into a human-readable response.

4. RAG Evaluation

The evaluation of the RAG pipeline was performed by testing its retrieval capabilities on a few sample queries. The relevance of the top-k retrieved quotes was manually inspected to assess the effectiveness of the semantic search.

File: rag_evaluation.py

Key Steps: - Instantiate RAGPipeline . - Define a set of test_queries . - For each query, retrieve top-k quotes and print their details (quote, author, similarity score).

5. Streamlit Application

A simple Streamlit web application was developed to provide an interactive interface for the RAG pipeline. Users can input a query and view the retrieved quotes and their details.

File: streamlit_app.py

Key Features: - **User Input:** A text input field for entering queries. - **Quote Display:** Displays the retrieved quotes, author, tags, and similarity scores. - **Caching:** Uses @st.cache_resource to load the RAG pipeline once for efficiency.

How to Run the Application

To run the Streamlit application, follow these steps:

1. Clone the repository (if applicable) or ensure all files are in the same directory.

- 2. **Install the required libraries:** bash pip install pandas openpyxl nltk datasets sentence-transformers faiss-cpu streamlit
- 3. Download NLTK data (if not already downloaded by
 rag_data_preparation.py): python import nltk
 nltk.download("punkt") nltk.download("stopwords")
 nltk.download("wordnet")
- 4. Run the data preparation script: bash python3.11 rag data preparation.py This will create rag processed quotes.csv.
- 5. Run the model fine-tuning script: bash python3.11 rag_model_fine_tuning.py This will save the fine_tuned_quote_model directory.
- 6. Launch the Streamlit application: bash streamlit run streamlit_app.py --server.port 8501 --server.enableCORS false -- server.enableXsrfProtection false The application will be accessible via a local URL (e.g., http://localhost:8501) and potentially an external URL if exposed.

Deliverables

- rag data preparation.py: Script for data loading and preprocessing.
- rag_model_fine_tuning.py: Script for loading and saving the sentence transformer model.
- rag_pipeline.py: Implements the RAG pipeline with FAISS indexing and retrieval.
- rag evaluation.py: Script for evaluating the retrieval component.
- streamlit app.py: Streamlit web application for interactive demo.
- rag processed quotes.csv: Processed dataset.
- fine_tuned_quote_model/: Directory containing the saved sentence transformer model.
- README RAG.md: This documentation file.
- README RAG.pdf: PDF version of this documentation file.