Chatbot with Google Gemini Al

The system uses the following **libraries and tools**:

- **1.LangChain** framework for integrating LLMs with utility tools such as text chunkers, vector stores, and retrievers.
- **2.Chroma** vector database for efficient storage and retrieval of embeddings (dense representations of text).

3.Google Generative Al:

Embeddings: Converts text into semantic embeddings for similarity search.

Chat Model: Generates conversational responses with safety and respect guidelines.

- **4.RecursiveCharacterTextSplitter** Splits large documents into smaller, coherent chunks for processing.
- **5.PyPDFLoader** -Specialized loader for extracting text from PDF documents.

Features

- **1. PDF Processing** Extracts text from PDF files and processes it for AI consumption.
- 2. Text Chunking -Divides the text into manageable, semantically meaningful units.
- 3. Embeddings Converts text chunks into dense vectors to capture semantic meaning.
- **4. Similarity Search** Retrieves relevant chunks based on user queries using a vector database.
- **5. Conversational AI** Generates human-like responses, ensuring they are context-aware and safe.

- 1. **PDF Loading** The PDF is read using PyPDFLoader, Text is extracted and cleaned for further processing.
- 2. **Text Chunking** The text is divided into smaller chunks using RecursiveCharacterTextSplitter, ensuring meaningful divisions.
- 3. **Embedding Creation** Each chunk is transformed into a dense vector representation using Google Generative AI embeddings.
- 4. **Vector Store** The embeddings are stored in Chroma, a vector database, enabling efficient similarity search.

- 5. **Query Handling** User queries are compared with the stored embeddings using MultiQueryRetriever to retrieve the most relevant text chunks.
- 6. **Response Generation** Retrieved chunks and the query are passed to the Google Generative AI chat model, which generates a detailed response.
- 7. **Safety Features** The system includes safety settings to filter harmful or inappropriate content.

Speech-to-Text Application

This application converts audio into text using WhisperX, a fast and efficient extension of OpenAl's Whisper. It supports high-quality transcription with optional GPU acceleration.

Libraries Used

- 1. WhisperX Transcribes audio into text efficiently, Supports GPU for faster processing.
- 2. Google Colab Tools Allows access to Google Drive for storing and loading audio files.

- 1.Install WhisperX Ensures the library is ready for transcription tasks.
- 2. Mount Google Drive Provides access to the audio file stored in Drive.
- **3.Load the Model** Loads the small model for fast transcription.
- **4.Transcribe Audio** Converts the audio file into text and timestamps.
- 5.Display Results Extracts the text from the result object for easy readability.

Frontend

ChatInterface.tsx - A React-based component that provides a chat interface for sending queries to the backend and displaying responses.

Key Features

- 1. **User Input**: Allows users to type a question or select a predefined sample question.
- 2. Chat UI: Displays messages (from both user and bot) with a scrolling interface.
- 3. Backend Integration: Sends user queries to the backend for processing.

Code Explanation

• State Management:

- messages: Array of chat messages.
- o input: Current user input.
- o isLoading: Indicates whether a backend request is in progress.

Main Components:

- **Chat Header**: Displays the title and a close button.
- Message Area: Shows the conversation between user and bot.
- **Input Form**: Allows users to type and send messages.
- o Sample Questions: Quick options for user queries.

Backend Communication:

- Sends user queries to the backend (http://localhost:5000/chat) via a POST request.
- Handles errors with toast notifications.

Key Functions

handleSubmit(e):

- Sends the user's guery to the backend.
- Updates the chat with the bot's response.

scrollToBottom():

Automatically scrolls to the latest message.

Dependencies

- React for building the UI.
- lucide-react for icons.
- Custom UI components like Button, Input, ScrollArea.

- User types a question and clicks "Send."
- 2. The app sends the query to the backend.
- 3. The response is displayed in the chat area.
- 4. Sample questions provide quick options for users.

Backend

app.py - Provides an API endpoint to handle transcription and translation requests.

Key Features

- 1. **Speech-to-Text Transcription** Uses Whisper to transcribe audio files into English text.
- 2. Translation Uses NLLB-200 to translate English text into Telugu.
- 3. Chunking Handles long text by splitting it into smaller chunks for processing.

Code Explanation

- Endpoints:
 - /chat (POST) Accepts a question or transcription request, Returns the transcription or translated response.
- Components:
 - Whisper Model Transcribes audio files into English text.
 - NLLB-200 Model Translates English text to Telugu.

Key Functions

- **split_text_into_chunks(text, max_tokens, tokenizer)** Splits long text into manageable chunks for translation.
- translate_large_text(text, translation_function, tokenizer) Translates text by processing it in chunks.
- english_to_telugu(text) Translates English text into Telugu using NLLB-200.
- **transcribe_and_translate(audio_path)** Handles the full transcription and translation pipeline:
 - 1. Transcribes audio into English.
 - 2. Translates English text into Telugu.

Dependencies

- whisper: For speech-to-text transcription.
- transformers: For handling NLLB-200 translation.
- google.colab: For accessing audio files in Google Drive (specific to Colab).

- 1. An audio file is transcribed into English text using Whisper.
- 2. The user selects a translation option (currently only English-to-Telugu).
- 3. The text is translated and returned to the frontend.

<u>Documentation for English-to-Telugu Translation Code</u> Overview

This code enables the translation of English text into Telugu using the Meta NLLB-200 model, a multilingual translation model designed for low-resource languages. It supports the handling of long text by splitting it into smaller, manageable chunks and translating each chunk sequentially.

Key Features

- 1. Translation Converts English text into Telugu using the NLLB-200 model.
- 2. Chunking Handles long text by splitting it into smaller pieces for processing.
- 3. Multilingual Support Leverages NLLB-200, which is capable of translating between 200+ languages.
- 4. Pre-trained Model Utilizes a pre-trained transformer model for high-quality translation.

Workflow

- 1. Input English Text Accepts text in English as input.
- 2. Tokenization and Chunking Splits long text into smaller chunks to comply with token limits.
- 3. Translation Translates each chunk into Telugu.
- 4. Output Telugu Text Combines translated chunks into the final Telugu output.

How to Use

- 1. Install the required libraries, including Hugging Face's transformers.
- 2. Load the pre-trained NLLB-200 model and tokenizer.
- 3. Call the main translation function with the English text.
- 4. Receive the Telugu translation as output.

Dependencies

- Transformers: For loading and running the NLLB-200 translation model.
- PyTorch: For model computations