Healthcare Assistant Agent - Documentation

This document has setup instructions, followed by an implementation overview.

Data Setup

Full instructions can also be found at: https://github.com/justthea/CS410-Healthcare-Assistant-Chatbot/blob/main/healthcare-data/README.md

1. **Database Setup**
    a. Prerequisites
        i. PostgreSQL 14 Installation:
            1. macOS(Using Homebrew):
                a. brew install postgresql@14
                b. brew services start postgresql@14
            2. Linux (Ubuntu/Debian):
                a. `sudo apt update`
                b. `sudo apt install postgresql-14 postgresql-contrib-14`
                c. `sudo systemctl start postgresql`
            3. Windows:
                a. Download PostgreSQL 14 installer from https://www.postgresql.org/download/windows/
                b. Run the installer and follow the setup wizard
                c. Remember the password you set for the postgres user
        ii. PostgreSQL Vector Extension Setup
            1. macOS:
                a. brew install postgresql@14
                b. psql postgres -c 'CREATE EXTENSION vector;
            2. Linux:
                a. sudo apt-get install postgresql-14-vector
            3. Windows:
                a. Vector extension is included in PostgreSQL 14+ installation

1.2 Database Creation (if needed)

b. Connect to PostgreSQL:

macOS/Linux

Command: psql postgres

Windows

Command: "C:\Program Files\PostgreSQL\14\bin\psql.exe" -U postgres

Then create the database:

i. CREATE DATABASE medical_db;
ii. \q

1.3 Backend Setup

- Navigate to the backend directory:
  - cd CS410-Healthcare-Assistant-Chatbot/healthcare-data
- Create and activate virtual environment:

macOS/Linux
python -m venv venv
source venv/bin/activate

Windows
python -m venv venv
venv\Scripts\activate

- Install dependencies:
  pip install -r requirements.txt

- Configure environment:`
  - Copy .env.example to .env
  - Update values:
    DB_HOST=localhost
    DB_PORT=5432
    DB_NAME=medical_db
    DB_USER=postgres
    DB_PASSWORD=your_password
    FDA_API_KEY=your_api_key

- Run setup script:
  python src/scripts/setup.py

Backend CLI Usage Guide

Interactive Mode

Start the chat interface:
python src/interactive.py

Example Mode

Run example queries:
python src/main.py

UI / Web Application Setup

Pre-requisites (Common for Windows/Mac/Linux)

1. Node.js Installation:
    - Ensure you have Node.js installed
    - Download Node.js from https://nodejs.org/.
2. Verify Installation:
    - Run the following commands in your terminal or command prompt to verify installation:
        - node -v
        - npm -v

Clone the GitHub repository and navigate to the project directory ‘healthcare-ui’.

Steps for Windows/Mac/Linux

1. **Install Dependencies**:
    - Run the following command to install all the dependencies listed in package.json.
      Command: npm install
2. **Start the Development Server**:
    - Use the following command to start your React app
      Command: npm start
    - This will launch the app in your default browser at http://localhost:3000
3. **Open in Browser**:
    - If it doesn't open automatically, visit http://localhost:3000 in your browser.

Steps to Set Up the Healthcare LLM

1. **Obtain Gemini API Key**
    - Using a Google Account, create an API Key from https://aistudio.google.com/app/apikey
2. **Place API Key in Environment File**
    - Create a .env file in the healthcare-ui folder
    - Use VIM or another text editor to include this line replacing your_api_key_here with your API Key:

- REACT_APP_GEMINI_API_KEY=your_api_key_here

Common Commands for Troubleshooting

- **If npm install Fails**:
  - Delete node_modules and package-lock.json, then reinstall.
    Command for Mac/Linux: rm -rf node_modules package-lock.json
    Command for Windows: rd /s /q node_modules package-lock.json
    Then, run npm install
- **Clearing Cache**:
  - Clear npm cache if issues persist
    Command: npm cache clean --force
- **Port Conflict**:
  - If port 3000 is in use, start the app on a different port.
    Command for Mac/Linux: PORT=3001 npm start
    Command for Windows: set PORT=3001 && npm start

Implementation Overview

Our Healthcare Information Assistant project delivered an intelligent healthcare assistant capable of recommending over-the-counter medications based on user-reported symptoms. The chatbot leveraged large language models (LLMs) and public health data to provide personalized recommendations, demonstrating a functional and user-friendly integration of natural language processing with health informatics.

1. **Frontend Development**:
   - A fully functional healthcare-themed website was created with a professional design, featuring pages for Home, About, Services, Contact, and Appointments.
   - A chatbot interface was seamlessly integrated into the website, positioned as a widget in the bottom-right corner. Users could interact with the chatbot via a clean and responsive input box, simulating a realistic user experience.
2. **Data Integration**:
   - Extensive data was sourced from reliable public health databases, including the NIH, covering over-the-counter medication information, symptom descriptions, and their mappings.
   - The collected data was structured into organized datasets that categorized symptoms and medications and outlined usage guidelines.
   - These datasets were converted into a vectorized format to enable efficient retrieval during LLM processing, ensuring real-time performance in responding to user queries.

3. **LLM Implementation**:
    - The LLM was fine-tuned to process user symptoms provided in natural language, using advanced NLP techniques like entity recognition and contextual tokenization.
    - Integration with the vectorized data allowed the chatbot to recommend accurate and personalized over-the-counter medications.
    - Robust API-based communication was established between the frontend and the backend, enabling smooth interactions with the LLM.
4. **Evaluation and User Feedback**:
    - A systematic evaluation framework was developed to measure the chatbot's effectiveness. Metrics included precision, recall, and user satisfaction, providing a comprehensive assessment of system performance.
    - After comparisons of multiple models, we eventually decided to use Gemini for its consistent accuracy
    - User feedback was derived from real-world interactions and sample user responses. This data informed iterative improvements to enhance user experience and recommendation accuracy.
    - As seen in the video presentation, accuracy for each query is output as well to measure effectiveness.

Challenges and Solutions:

- Initial challenges in selecting an optimal vector database were overcome through team coordination and research into industry best practices.
- Sequential dependencies between frontend and backend tasks were managed efficiently to ensure timely integration.