

AI-Powered Health Assistant

A Project Report

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by

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ABSTRACT

This project presents the development of an AI-powered health assistant designed to provide personalized health advice, particularly focusing on dietary changes and wellness guidance. The primary problem addressed is the lack of accessible, accurate, and region-specific health recommendations, especially for individuals seeking to make informed dietary choices based on their personal needs. The project aims to develop a system that can deliver real-time advice and guidance through natural language processing (NLP), leveraging AI to understand queries related to health and nutrition.

The methodology involves integrating speech recognition, text-to-speech conversion, and AI-based language models, such as Hugging Face embeddings and language model APIs. The system processes user queries in multiple languages, offering tailored responses that address specific health concerns like diet, exercise, and preventive care. The assistant is designed to operate in both English and Hindi, catering to users in India, with the ability to recommend dietary changes based on regional foods and common health practices.

Key results include the successful implementation of a bilingual interface, with real-time processing of voice inputs and conversion to actionable health advice. The system's ability to analyze user queries, understand health-related contexts, and provide personalized suggestions is validated through testing with a diverse set of users.

In conclusion, the AI-powered health assistant significantly enhances access to personalized health advice. It bridges the gap between technological advancements in AI and the practical needs of users seeking to improve their diet and overall well-being. Future improvements may involve expanding the database of regional dietary recommendations and further optimizing the system for specific health conditions.

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CHAPTER 1

Introduction

1.1 Problem Statement:

Access to reliable medical information is often delayed due to long wait times, limited availability of healthcare professionals, and misleading online resources. Many individuals struggle to determine whether their symptoms require urgent medical attention or simple home care. Additionally, language barriers and accessibility issues make it difficult for some users to obtain accurate health advice.

This project aims to provide an **AI-powered medical assistant** that offers **instant, voice-enabled, and multilingual responses** to medical queries. It enhances accessibility, detects emergency situations, and ensures users receive clear, context-aware information anytime. While not a replacement for professional healthcare, it serves as a **bridge between concern and consultation**, reducing uncertainty and guiding users toward informed decisions.

1.2 Motivation:

The increasing reliance on unreliable online sources for medical advice poses a risk to public health. This project was chosen to bridge the gap between information and action by offering a fast, AI-driven, and accessible medical assistant. Its potential applications include preliminary symptom assessment, emergency keyword detection, and multilingual health guidance. The impact extends to reducing anxiety, improving healthcare accessibility, and minimizing unnecessary ER visits by directing users toward appropriate care. Beyond being a simple chatbot, the system is designed to act as an intelligent health companion that encourages self-care while recognizing when professional medical attention is necessary. The increasing reliance on the internet for medical advice often leads to misinformation, misdiagnosis, or unnecessary panic. This project was chosen to develop an AI-powered medical chatbot that provides accurate and context-aware responses to health-related queries. By integrating **Ollama LLM**, **Hugging Face embeddings**, and **Chroma for retrieval**, the chatbot enhances information accuracy and accessibility. Features like **voice-to-text**, **text-to-speech**, and **emergency keyword detection** make it more interactive and practical, especially for users who need quick assistance. The potential applications range from **preliminary medical guidance** and **health education** to **assisting individuals with disabilities** who prefer voice-based interactions.

1.3 Objective:

Clearly state the objectives of the project.

This project aims to develop an AI-powered medical assistant capable of providing quick and reliable medical insights using natural language processing, detecting emergency scenarios through keyword recognition, and offering multilingual and voice-enabled interaction for greater accessibility. The system is designed to minimize misinformation while ensuring structured and trustworthy responses. Beyond just answering medical queries, it is intended to deliver empathetic and human-like interactions that reduce fear and uncertainty. In the future, data-driven personalization could be introduced, allowing the system to refine its responses based on user history and preferences.

- Utilize **retrieval-augmented generation (RAG)** to enhance accuracy.
- Offer **text and voice-based input/output** for accessibility.
- Detect **emergency medical keywords** and prompt users to seek immediate help when necessary.
- Support **multiple languages** for broader usability.
- Provide **exportable chat history** for reference and documentation.
- Implement a **user-friendly interface with customizable settings** such as **dark mode, voice speed control, and favourite's**.

1.4 Scope of the Project:

The project focuses on developing an AI-based medical Q&A system with instant responses, voice-to-text and text-to-speech support, multilingual capabilities, emergency keyword detection, and basic symptom guidance. However, it does not serve as a replacement for professional medical consultation, nor does it integrate with real-time hospital or doctor networks. The system will not provide personalized medical treatment plans and remains dependent on AI training data, which may limit response depth. Designed as more than just a chatbot, it functions as an interactive health companion that could expand beyond medical advice in the future, with potential integration into wearable health technology or mental health support systems.

CHAPTER 2

Literature Survey

2.1 Review relevant literature or previous work in this domain.

The application of AI in healthcare has evolved significantly, with numerous studies highlighting the effectiveness of natural language processing (NLP) and retrieval-based AI models in providing medical guidance. Existing AI chatbots like IBM Watson Health, Ada Health, and WebMD's Symptom Checker have demonstrated the potential of automated medical assistance. Research in retrieval-augmented generation (RAG) further supports the integration of LLMs with knowledge bases to enhance response accuracy. However, misinformation, limited contextual understanding, and lack of personalization remain major challenges in existing systems.

2.2 Mention any existing models, techniques, or methodologies related to the problem.

Methodology/Model	Description	Limitations
Rule-Based Medical Chatbots	Early AI assistants followed predefined decision trees and scripts for symptom assessment and diagnosis.	Lacks adaptability, cannot handle complex or dynamic queries.
ML-Based Conversational Agents	Uses deep learning models (e.g., BERT, GPT) to generate responses based on trained medical data.	May generate inaccurate or hallucinated medical advice if not fine-tuned with verified healthcare data.
Hybrid Models (Retrieval-Augmented Generation - RAG)	Integrates LLMs with vector databases (e.g., ChromaDB, Pinecone) to retrieve contextually accurate medical information.	Still requires frequent updates and filtering to avoid misinformation.

Voice-Based AI Assistants	Platforms like Google Assistant, Alexa, and Siri provide general health-related responses using voice interaction.	Lack domain-specific accuracy and may not provide reliable medical advice.
Emergency Response AI	Some AI-driven systems detect critical health conditions using real-time symptom analysis and keyword detection.	Prone to false positives or delayed alerts, requiring manual verification.

2.3 Highlight the gaps or limitations in existing solutions and how your project will address them.

Challenges	Solutions
Lack of context-aware responses – Many existing chatbots generate generic or inconsistent answers that don't consider the user's specific situation.	Uses a retrieval-augmented model with Hugging Face embeddings to provide relevant and accurate responses tailored to the user's query.
Limited emergency detection – Most AI models fail to prioritize urgent medical situations, which could delay critical help.	Integrates keyword-based emergency alerts that instantly notify users when their query indicates a potential emergency.
Poor multilingual support – Many models are limited to English, making them less accessible to non-English speakers.	Offers bilingual support (English & Hindi) , ensuring broader accessibility, especially in the Indian regional context.
Lack of personalized interaction – Traditional chatbots do not retain user history, leading to impersonal and repetitive responses.	Implements session-based history tracking , allowing personalized responses by remembering past user interactions.
No voice interaction support – Many chatbots only support text input, making them less accessible to people with disabilities or those who prefer voice communication.	Integrates speech-to-text (voice input) and text-to-speech (audio responses) , enhancing usability for all users.

CHAPTER 3

Proposed Methodology

3.1 System Design

Provide the diagram of your Proposed Solution and explain the diagram in detail.

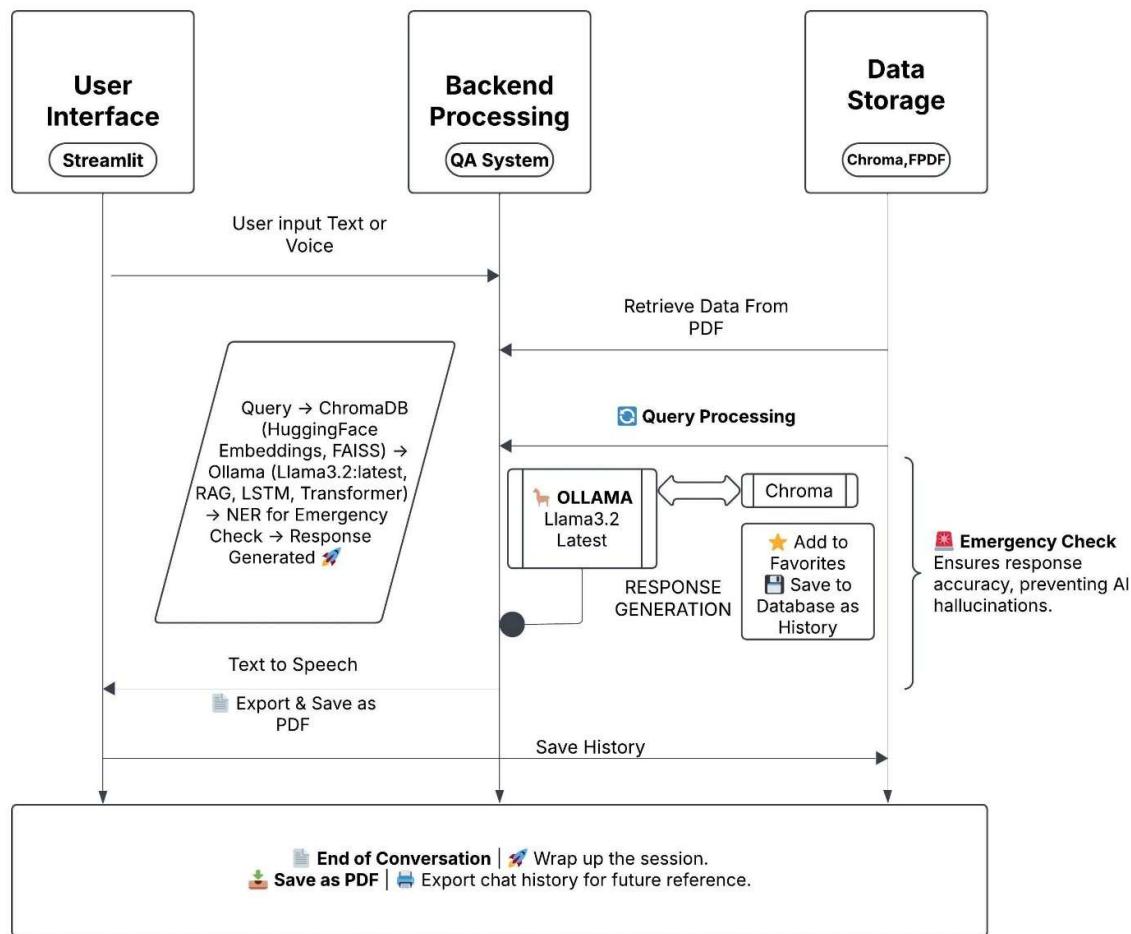


Figure 1

3.2 Requirement Specification

Mention the tools and technologies required to implement the solution.

3.2.1 Hardware Requirements:

-  **Processor:** Intel i5/i7 or AMD Ryzen 5/7 (or higher)
-  **RAM:** Minimum 8GB (16GB recommended for smooth performance)
-  **Storage:** At least 20GB free space (SSD preferred)
-  **Internet:** Stable broadband connection for API calls and model updates
-  **Microphone & Speaker:** Required for voice-based interaction
-  **GPU (Optional):** NVIDIA RTX 2060 or higher for faster model inference (if using local LLM processing)

3.2.2 Software Requirements:



Figure 2

- ❖  Programming Language: Python 3.11+
- ❖  Frontend: Streamlit (for web-based UI)
- ❖  LLM Model: Ollama (Llama 3.2 latest)
- ❖  Retrieval Mechanism: ChromaDB (Vector database)
- ❖  Embedding Model: Hugging Face (all-MiniLM-L6-v2)
- ❖  Speech Processing: Google Speech Recognition, gTTS (Text-to-Speech)
- ❖  Libraries:
 - LangChain (For RAG-based query processing)
 - FPDF (For exporting chat history)
 - NumPy, Pandas (For data handling)
- ❖  Development Tools:
 - Jupyter Notebook / VS Code / PyCharm
 - Postman (For API testing)

CHAPTER 4

Implementation and Result

4.1 Snap Shots of Result:

Kindly provide 2-3 Snapshots which showcase the results and output of your project and after keeping each snap explain the snapshot that what it is representing.

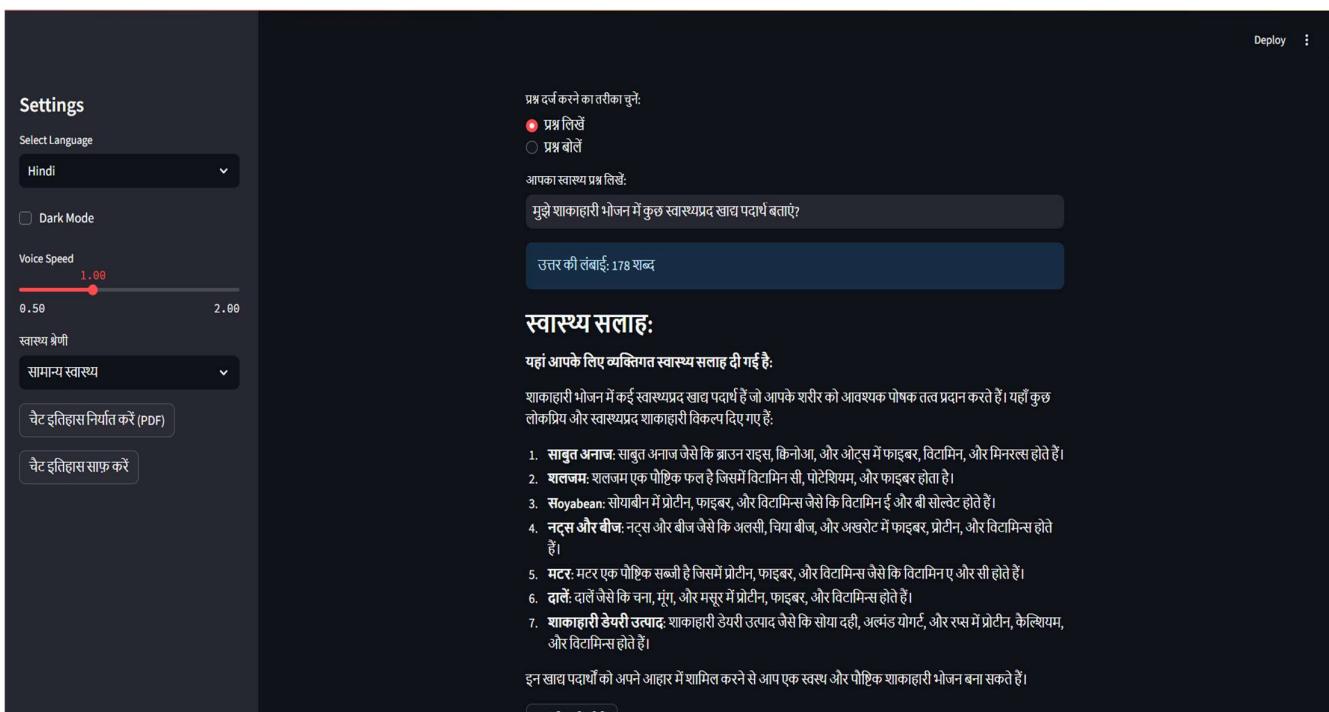


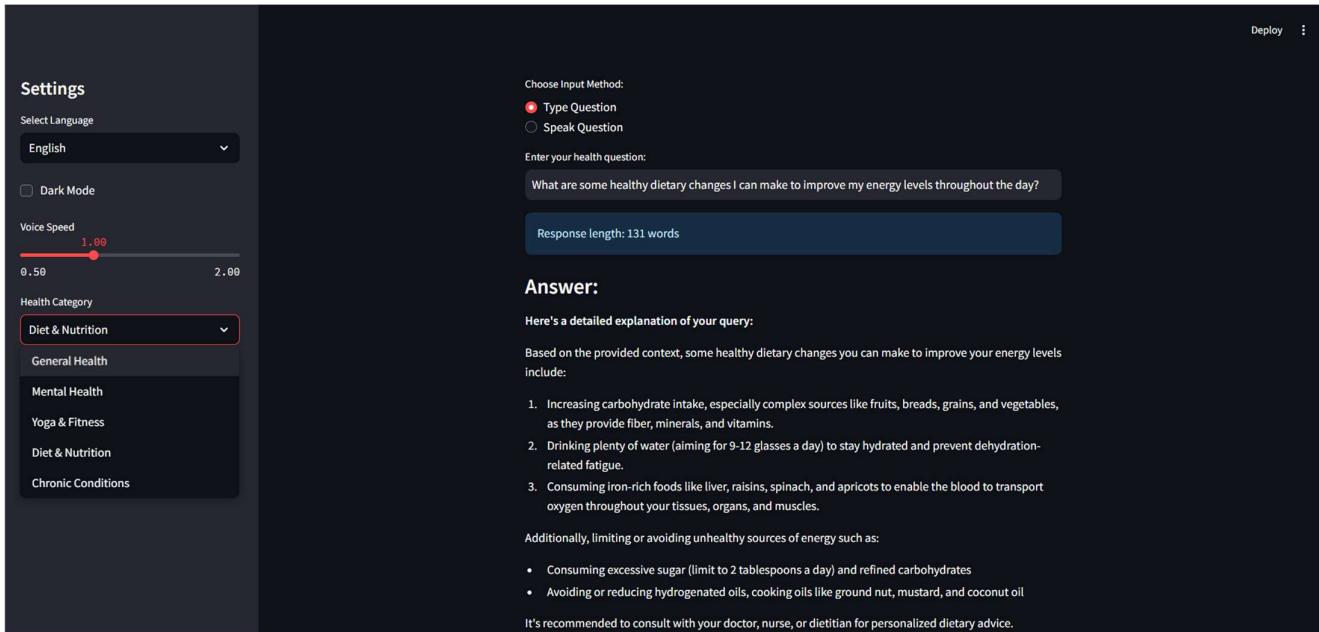
Figure 3

Document Ingestion in the CMD prompt:

```
(ai) C:\Users\Public\Project>python ingest.py
C:\Users\rudra\AppData\Local\ Packages\PythonSoftwareFoundation.Python.3.11_qbz5n2kfra8p0\LocalCache\local-packages\Python311\site-packages\tokenization_utils_base.py:1601: FutureWarning: 'clean_up_tokenization_spaces' was not set. It will be set to 'True' by default. This behavior will be deprected in transformers v4.45, and will be then set to 'False' by default. For more details check this issue: https://github.com/huggingface/transformers/issues/31884
  warnings.warn(
Creating new vectorstore
Loading documents from source_documents
Loading new documents: 100%|██████████| 2/2 [00:47<00:00, 23.58s/it]
Loaded 4603 new documents from source_documents
Split into 41945 chunks of text (max. 500 tokens each)
Creating embeddings. May take some minutes...
```

Figure 4

English Output :



The screenshot shows the AI-Powered Health Assistant interface. On the left, there is a sidebar titled "Settings" with options for "Select Language" (set to English), "Dark Mode" (unchecked), "Voice Speed" (set to 1.00), and a "Health Category" dropdown menu containing "Diet & Nutrition", "General Health", "Mental Health", "Yoga & Fitness", "Diet & Nutrition", and "Chronic Conditions". The main area has a "Choose Input Method:" section with "Type Question" selected. Below it is a text input field containing the question: "What are some healthy dietary changes I can make to improve my energy levels throughout the day?". A status bar indicates "Response length: 131 words". Under the "Answer:" heading, there is a detailed explanation: "Based on the provided context, some healthy dietary changes you can make to improve your energy levels include: 1. Increasing carbohydrate intake, especially complex sources like fruits, breads, grains, and vegetables, as they provide fiber, minerals, and vitamins. 2. Drinking plenty of water (aiming for 9-12 glasses a day) to stay hydrated and prevent dehydration-related fatigue. 3. Consuming iron-rich foods like liver, raisins, spinach, and apricots to enable the blood to transport oxygen throughout your tissues, organs, and muscles. Additionally, limiting or avoiding unhealthy sources of energy such as: • Consuming excessive sugar (limit to 2 tablespoons a day) and refined carbohydrates • Avoiding or reducing hydrogenated oils, cooking oils like ground nut, mustard, and coconut oil". At the bottom, a note says "It's recommended to consult with your doctor, nurse, or dietitian for personalized dietary advice."

4.2 GitHub Link for Code:

[GitHub Link](#)

The AI-Powered Health Assistant is an intelligent system designed to provide personalized health advice using AI, NLP, and speech recognition. It offers bilingual support, emergency detection, and symptom analysis, enhancing healthcare accessibility. The project integrates retrieval-augmented generation (RAG) and cloud-based deployment for real-time, reliable medical guidance.

https://github.com/rudrakadel/AI-Powered_Health_Assistant_TECHSAKSHAM/tree/main

CHAPTER 5

Discussion and Conclusion

5.1 Future Work:

While the AI-powered health assistant has shown promising results, there are several areas for future enhancement:

1. **Expanding Language Support:** Currently, the system supports English and Hindi. Future versions could incorporate additional regional languages to broaden accessibility and reach a larger audience in India and other multilingual regions.
2. **Integration with Wearables:** Future improvements could involve integrating the assistant with health-tracking devices (e.g., smartwatches) to gather real-time data about users' activity levels, heart rate, and sleep patterns, which could enhance personalized dietary recommendations.
3. **Advanced Machine Learning Models:** The model could be enhanced by incorporating more sophisticated NLP techniques or deep learning algorithms, such as transformers, to better understand nuanced health-related queries and provide even more accurate responses.
4. **Dynamic Diet Plan Generation:** Introducing an interactive feature that allows the system to generate complete meal plans based on dietary preferences, medical conditions, and personal goals could be a valuable addition.
5. **User Feedback Loop:** Implementing a feedback mechanism where users can rate responses or suggest improvements will help in continuously refining the system's accuracy and relevance.

5.2 Conclusion:

The AI-powered health assistant has demonstrated significant potential in providing personalized health and dietary advice through an accessible interface. By combining natural language processing, speech recognition, and AI-driven recommendations, the project addresses a critical need for accessible health guidance, especially in India, where regional dietary habits and languages vary widely. This system not only assists users with

making informed decisions about their health but also promotes healthier lifestyles through easy access to information.

The project's impact lies in bridging the gap between technology and health by creating an intuitive and responsive assistant that helps users manage their well-being. Moving forward, expanding the capabilities and integrating new features could further elevate its contribution to the growing field of AI-based healthcare solutions.

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