HEALTHCARE QA ASSISTANT

AI-POWERED CLINICAL Q&A WITH RESEARCH-BACKED EVIDENCE

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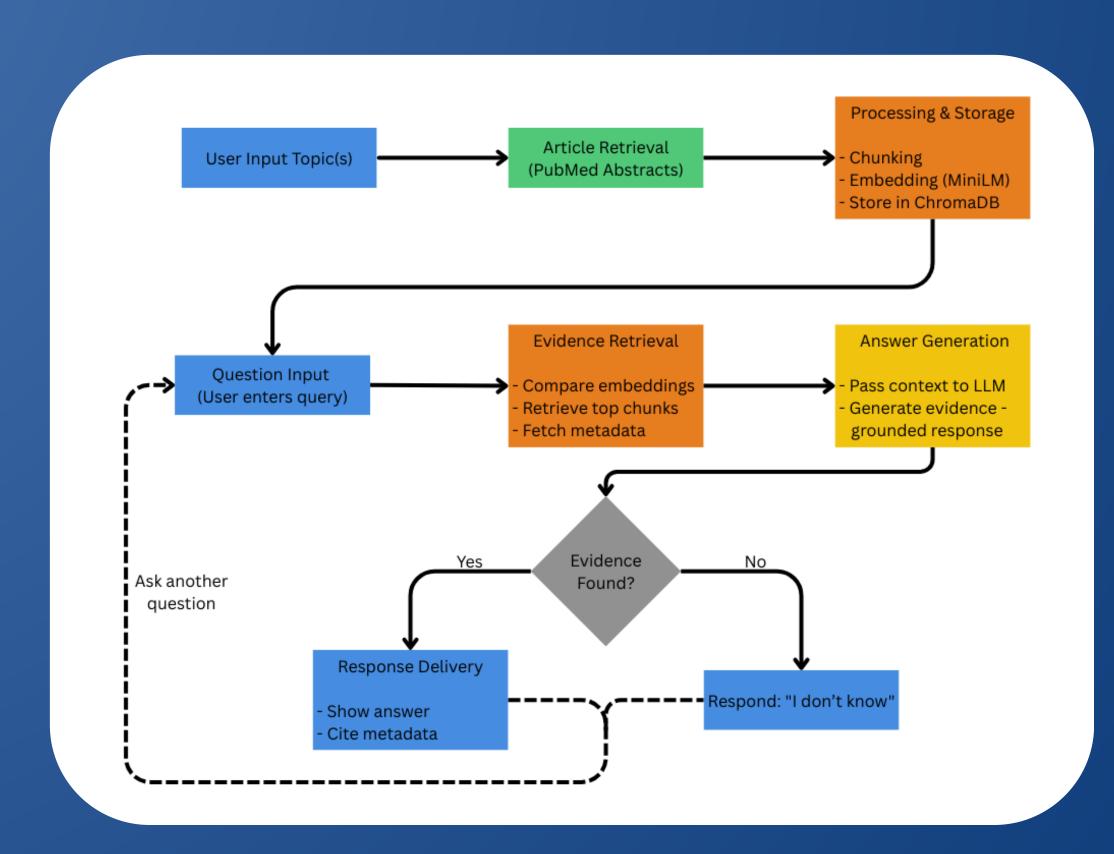
PROBLEM STATEMENT

- Healthcare professionals and patients often struggle to quickly access clear, evidence-based answers to medical questions, as reliable knowledge is scattered across vast volumes of research literature.
- This leads to delayed decision-making, uncertainty in treatment choices, and over-reliance on generic search engines that may surface incomplete or non-credible information.
- There is a growing need for a trusted, intelligent system that can retrieve, process, and summarize medical research into concise, transparent insights.
- An Al-powered assistant, equipped with PubMed article retrieval, semantic search, and evidence-grounded QA capabilities, can help clinicians and patients make informed decisions with confidence and efficiency.

PROJECT OBJECTIVES

- Build an AI-powered conversational assistant that helps healthcare professionals and researchers find evidence-based answers from PubMed.
- Enable efficient literature exploration by retrieving, chunking, and embedding medical articles for semantic search.
- Provide accurate, context-aware answers through a RAG pipeline powered by LLaMA3 (via GROQ).
- Ensure transparency in responses by displaying metadata such as authors, journals, and publication dates alongside generated insights.
- Deliver a user-friendly interface via Streamlit for intuitive querying and seamless interaction.

SYSTEM ARCHITECTURE



TECHNOLOGY STACK

This project integrates Al-driven retrieval pipelines and lightweight web technologies to deliver evidence-based, real-time medical insights from PubMed:

Frontend

• **Streamlit:** Builds a simple, responsive chat interface where users can enter medical queries and view evidence-backed answers.

Language Model

 Groq + LLaMA 3.3: Powers real-time, natural language question answering with concise, evidence-grounded medical responses.

Article Retrieval & Processing

- PubMed: Retrieves biomedical article abstracts and metadata directly from PubMed based on user-provided topics.
- **Text Splitter:** Splits long abstracts into manageable chunks for efficient embedding and retrieval.

• Semantic Embeddings & Storage

- Sentence Transformers (MiniLM): Generates dense vector embeddings for semantic similarity search.
- ChromaDB: Stores embeddings and enables fast, local vector-based retrieval of relevant medical evidence.

QA Pipeline

• RAG (Retrieval-Augmented Generation): Combines semantic retrieval from ChromaDB with LLaMA3 to produce reliable, citation-backed answers.

FUNCTIONAL WORKFLOW

This assistant system retrieves and synthesizes biomedical knowledge from PubMed to provide concise, evidence-based medical insights — all within a unified conversational interface.

1. Article Retrieval

- Users provide medical topics in the **Streamlit** sidebar (comma-separated).
- The app queries PubMed to fetch article abstracts along with metadata (authors, journal, publication date).

2. Processing & Storage

- Retrieved articles are split into smaller, manageable chunks.
- Each chunk is embedded using **Sentence Transformers (MiniLM)**.
- Embeddings and metadata are stored in **ChromaDB** for efficient semantic search.

3. Question Answering

a. Evidence Retrieval (ChromaDB Search)

- User questions in natural language are compared against stored embeddings.
- **Relevant chunks** (abstracts + metadata) are retrieved based on similarity.

b. Answer Synthesis (RAG Pipeline)

- The **Groq-hosted LLaMA 3.3** model synthesizes concise, evidence-grounded answers.
- If no strong evidence is found, the system responds with "I don't know".

4. Response Generation & Delivery

- Responses are transparent, citing relevant authors and journals.
- LLaMA 3.3 (Groq) ensures coherent, medically accurate text generation.
- Streamlit UI presents the answer in real time with source attribution.

KEY FEATURES

End-to-End Medical Q&A

• From query understanding to evidence-backed answers, the system automates the full retrieval-augmented QA flow for healthcare queries.

Evidence-Based Insights

 Generates concise medical responses grounded in PubMed abstracts, with transparent attribution to authors, journals, and publication dates.

Al-Powered Retrieval & Reasoning

 Uses Sentence Transformers for semantic query embedding and ChromaDB for retrieving the most relevant biomedical literature.

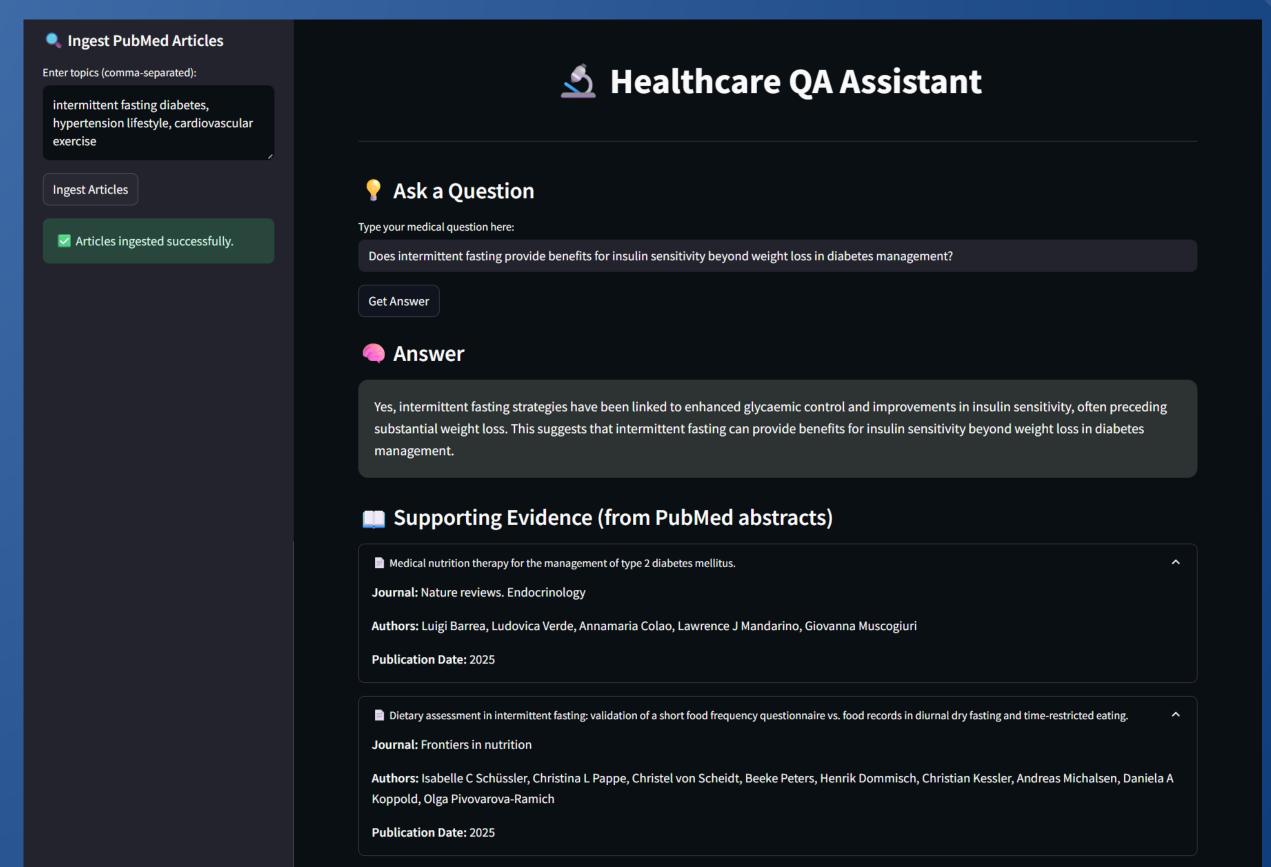
Streamlit-Powered Web UI

• An intuitive chat interface lets users ask questions in natural language and receive clear, real-time medical insights.

Modular & Scalable Architecture

 Decouples ingestion, retrieval, and answer generation, making the system extendable to additional medical domains or larger datasets.

USER INTERACTION PREVIEW



PROJECT SUMMARY

- Developed an AI-powered healthcare assistant that retrieves and synthesizes biomedical knowledge from PubMed to provide concise, evidence-based medical insights.
- Implemented **PubMed** API integration to fetch article abstracts and metadata on user-specified medical topics.
- Processed articles by splitting abstracts into smaller chunks, embedding them with **Sentence Transformers (MiniLM)**, and storing them in **ChromaDB** for efficient semantic search.
- Integrated a Retrieval-Augmented Generation (RAG) pipeline using Groq's LLaMA 3.3 to generate transparent, source-linked answers from retrieved biomedical literature.
- Designed an interactive **Streamlit** app with a sidebar for topic selection and a chat interface for natural language Q&A, with real-time display of metadata.
- Live App: https://vaibhav-project-healthcare-app-assistant.streamlit.app/
- GitHub Repository: https://github.com/vaibhavgarg2004/Healthcare-QA-Assistant

