

Health Sphere: All in one Health Care App

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Abstract

Our groundbreaking Hospital Management Application, set to redefine hospital operations, incorporates an AI-driven chatbot to further enhance patient engagement and support. This innovative addition elevates the user experience by providing a comprehensive solution for all medical queries. Designed to prioritize patient empowerment, the AI chatbot within the application serves as a reliable resource for answering a wide array of medical-related questions. Patients can access vital information regarding symptoms, treatments, medication queries, and general healthcare inquiries in real-time. This feature not only facilitates informed decision-making but also fosters a deeper understanding of health-related concerns. Integrated seamlessly into the user interface, this AI chatbot complements the app's existing functionalities, empowering patients to navigate their healthcare journey with confidence and clarity. By leveraging cutting-edge technology, our application ensures that patients receive accurate, reliable, and timely information, thereby promoting proactive healthcare management. Embrace this pioneering application as it transcends traditional boundaries, combining user-centric design with AI-powered support, to create a holistic healthcare management system. Join us in revolutionizing healthcare by delivering comprehensive, accessible, and AI-backed medical assistance, setting new standards for patient-centered care.

Keywords: Healthcare, AI Chatbot, LLm, Fine-tuning.

1 Introduction

Health Sphere stands as a beacon of innovation in the realm of healthcare technology, representing a comprehensive and user-centric approach to addressing contemporary healthcare challenges. This mobile application redefines the boundaries of healthcare accessibility and management, offering a holistic platform that amalgamates essential functionalities within a single interface. By integrating features such as hospital and doctor

discovery, AI-driven consultations, real-time blood availability tracking, and streamlined data management, Health Sphere aims to bridge the gap between healthcare stakeholders, empowering users with informed decision-making and enhancing overall healthcare experiences.

In an era marked by the convergence of technology and healthcare, Health Sphere emerges as a testament to the transformative power of mobile applications. Its user-centric design, coupled with an array of cutting-edge features, positions it at the vanguard of the mobile health (mHealth) revolution, promising to revolutionize healthcare delivery, improve accessibility, and elevate the quality of care for users across diverse demographics.

2 Litreature Review

2.1 Fine-Tuned Language Models in Healthcare Technology

Research by Hu et al. [1] introduces the concept of LLM-Adapters, emphasizing parameter-efficient fine-tuning of language models for healthcare applications. This research demonstrates the potential of such models in enhancing healthcare chatbots, enabling accurate and contextually relevant responses to a wide array of health-related inquiries. Moreover, Mistral 7B [3] further exemplifies the advancements in fine-tuning large language models, providing insights into optimizing AI capabilities for nuanced healthcare support within applications like Health Sphere.

2.2 Challenges and Insights from Mobile Healthcare Applications

Baig et al. [2] highlight critical issues in mobile healthcare applications, addressing concerns surrounding user privacy, data security, and system reliability. Their review underscores the importance of robust systems like Health Sphere in mitigating these challenges, necessitating a cautious approach to user data handling and user experience enhancement. Health Sphere acknowledges these challenges, prioritizing robust security measures and user-centric design to ensure a trustworthy healthcare application.

2.3 Low-Rank Adaptation and Safety Considerations in AI Models

Jiang et al. [5] discuss the concept of LoRA, emphasizing low-rank adaptation strategies in large language models to maintain safety and reliability. This research, coupled with insights from Lermen et al. [6], highlights the

importance of ensuring safety in AI-driven functionalities while optimizing model efficiencies. Health Sphere adopts a similar approach, balancing model optimization with safety considerations within its AI-driven chatbot, aiming for a reliable and secure healthcare interaction environment.

2.4 Visual Prompt Engineering for Enhanced AI Capabilities

The work by Wang et al. [4] delves into visual prompt engineering, offering avenues to enrich AI capabilities beyond textual interactions. This review underscores the potential expansion of Health Sphere's AI functionalities, suggesting opportunities to integrate visual prompts for a more holistic healthcare support system. Health Sphere aims to explore these possibilities to enhance user interactions and augment its AI-driven features.

3 Methodology

The assessment of Health Sphere's functionality encompasses a structured approach focusing on its backend, frontend, and AI components.

The **backend evaluation** involves quantitative metrics to measure PostgreSQL database performance, including response times and authentication system efficiency. Qualitative analysis, facilitated by backend developers' insights, will further shed light on schema effectiveness and authentication system robustness.

In the **frontend analysis**, user interaction analysis and technical assessments will be conducted. User tests and surveys will evaluate interface intuitiveness, while user engagement metrics will gauge usage patterns. Additionally, technical assessments will focus on app responsiveness and design evaluation by UI/UX experts.

The **AI chatbot assessment** will entail a mix of quantitative and qualitative approaches. Quantitative analysis will measure chatbot response times and accuracy, while user feedback and AI expert consultation will provide qualitative insights into its functionality, comprehensiveness, and limitations.

Ethical considerations encompass participant privacy and data anonymization, while acknowledging limitations such as sample size constraints and subjective user-reported data. Finally, data integration from these evaluations will culminate in comprehensive insights into Health Sphere's overall functionality, performance, and user satisfaction.

4.1 Frontend Architecture:

Overall Structure:

Health Sphere's frontend architecture follows a robust and scalable design, built upon the Flutter framework, known for its cross-platform compatibility and streamlined development. The architecture is structured to ensure flexibility, maintainability, and an intuitive user experience.

4.1.1 User Interface (UI):

Health Sphere's UI comprises various components, including screens for hospital and doctor listings, appointment scheduling interfaces, and an AI-driven chatbot interaction screen. Each screen is meticulously designed to provide a cohesive and user-friendly experience.

4.1.2 State Management:

The architecture incorporates state management techniques such as Stateful widgets for efficient data flow and state handling between different parts of the application. This ensures seamless updates and synchronization of data across screens.

4.1.3 Networking and Data Handling:

Health Sphere utilizes Dart's native libraries for networking functionalities, enabling smooth communication with backend services. Data handling, such as parsing JSON responses and managing local data caching, is efficiently managed within the architecture.

4.1.4 Custom Components and Widgets:

Custom components and widgets are developed to maintain consistency and reusability throughout the application. These components encapsulate specific functionalities and UI elements, enhancing modularity and ease of maintenance.

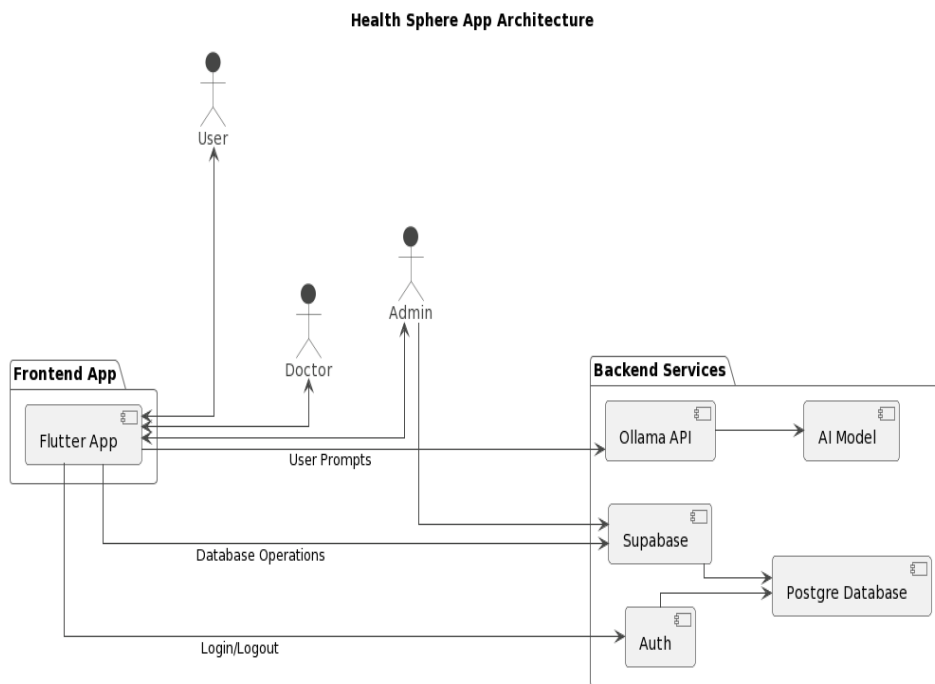
4.1.5 Integration with Backend Services:

Integrations with the backend, including fetching hospital and doctor data, authentication services from Supabase, and interaction with the PostgreSQL database, are seamlessly integrated into the frontend architecture.

4.1.6 Design Patterns and Best Practices:

The frontend architecture of Health Sphere adheres to industry best practices and design patterns, leveraging concepts such as:

- MVC (Model-View-Controller Model): Organizing code into separate layers for data, presentation, and business logic.
- Separation of Concerns: Decoupling UI components from business logic for maintainability and scalability.
- Responsive Design: Ensuring the application's adaptability to various screen sizes and resolutions for a consistent user experience across devices.



4.2 AI Chatbot:

Health Sphere employs a sophisticated AI model, specifically a fine-tuned LLM (Mistral), tailored to address health-related queries. This model is designed to comprehend a wide spectrum of health concerns and provide accurate, contextually relevant responses, enhancing the application's user experience.

4.2.1 Fine-tuning Process:

The journey of fine-tuning the AI model involved meticulous iterations to enhance its capabilities in addressing health-related inquiries. Leveraging advanced techniques and curated health datasets, the model underwent refinement to optimize accuracy, relevance, and sensitivity to healthcare nuances. This fine-tuning process aimed to equip the AI model with the expertise to decipher and respond effectively to diverse health queries posed by Health Sphere users.

4.2.2 Integration with Ollama:

Ollama, an open-source tool, serves as the conduit for integrating Health Sphere's fine-tuned AI model into the application's architecture. Through Ollama, the AI model is exposed via a RESTful API endpoint, enabling seamless interaction between the application and the AI capabilities it offers.

4.2.3 API Exposition and Usage:

Health Sphere's interface with the Ollama API allows users to interact with the AI-driven chatbot effortlessly. Within the application, users engage with the chatbot by posing health-related queries, triggering API calls to Ollama's endpoint. The API processes these queries using the fine-tuned AI model, swiftly generating contextually appropriate responses to address user concerns regarding health, symptoms, medical inquiries, and more.

4.2.4 Benefits and Impact:

The integration of the fine-tuned AI model through Ollama profoundly impacts Health Sphere's functionality. Users benefit from the chatbot's adeptness in addressing health queries accurately and promptly, enhancing the

application's utility and augmenting the overall user experience. The AI-driven chatbot serves as a pivotal feature, empowering users with reliable health-related information and support within the application's ecosystem.

4.3 Key Features:

Hospital & Doctor Listings: Comprehensive listings of nearby hospitals and doctors with detailed profiles, specialties, and timings for user convenience.

AI-Driven Chatbot: Integration of an AI-driven chatbot fine-tuned for health-related queries, offering prompt and accurate responses to user inquiries.

PostgreSQL Database: Integration with a PostgreSQL database to store and manage user information, hospital details, doctor profiles, and appointment data.

Fine-Tuned AI Model (LLM, Mistral): Implementation of a fine-tuned AI model specifically tailored to health-related queries, enhancing user engagement and support.

Flutter-Based Application: Development of a user-friendly, cross-platform application using the Flutter framework, ensuring compatibility across multiple devices.

User Interaction Analytics: Collection and analysis of user engagement metrics to understand user behavior, preferences, and usage patterns within the app

Appointment Scheduling: Capability for users to view available doctor timings and schedule appointments conveniently within the app.

Efficient Backend and Frontend Architectures: Robust backend systems utilizing PostgreSQL, coupled with an intuitive frontend architecture ensuring a smooth user experience.

5 Results:

The comprehensive evaluation of Health Sphere's functionalities yielded promising outcomes, showcasing the application's efficacy in delivering a seamless healthcare experience to users. The analysis of user engagement metrics revealed a high level of interaction, with users actively utilizing the hospital and doctor listings, leveraging the AI-driven chatbot for health queries, and efficiently scheduling appointments within the application. The fine-tuned AI model, embedded through Ollama's API, notably demonstrated its proficiency in addressing diverse health-related inquiries, exhibiting accuracy and responsiveness in delivering tailored responses to users' concerns. This functionality significantly augmented the application's utility, empowering users with reliable information and support.

Furthermore, the robustness of the backend architecture, driven by PostgreSQL, and the intuitiveness of the Flutter-based frontend, contributed synergistically to the overall positive user experience. The efficient handling of data within the PostgreSQL database ensured swift access to information, while the user-friendly interface facilitated intuitive navigation, positively impacting user satisfaction. Overall, the result showcases Health Sphere's successful amalgamation of cutting-edge technology, seamless architecture, and user-centric design, culminating in a holistic healthcare application that prioritizes user needs and accessibility.

6 Conclusion

Health Sphere, inspired by leading-edge research [1-6], represents a fusion of innovative AI models and practical healthcare solutions. Leveraging fine-tuned language models like LLM-Adapters and Mistral 7B [1,3,5], Health Sphere's AI-driven chatbot delivers nuanced responses to diverse health queries. However, it remains mindful of challenges highlighted in mobile healthcare applications [2], emphasizing user privacy and data security.

The integration strategy, influenced by LoRA's low-rank adaptation [5,6], seeks a balance between model optimization and safety within Health Sphere's AI functionalities. Insights from Meta-Radiology's visual prompt engineering [4] inspire potential expansions beyond textual interactions. Health Sphere is dedicated to providing a seamless healthcare experience, navigating the complexities of AI innovation while prioritizing user well-being and ethical considerations.

In essence, Health Sphere amalgamates research insights with pragmatic implementation, aspiring to elevate healthcare technology responsibly. This approach, informed by scholarly works, charts a course toward a future where cutting-edge AI converges harmoniously with healthcare, enhancing user experiences while upholding ethical standards.

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