

## AI POWERED FITNESS ASSISTANCE

**Prajwal Banakar\*1, Ajay Kumar HN\*2, Chirag H\*3, Dr. Dilshad Begum\*4**

\*1,2,3Information Science Engineering Ghousia College of Engineering, India.

\*4Professor and HOD Ghousia College of Engineering, India.

DOI: <https://doi.org/10.56726/IRJMETS87195>

### ABSTRACT

The proliferation of digital health technologies necessitates intelligent systems capable of providing personalized, data-driven wellness guidance. This paper presents the architecture and implementation of an integrated web-based AI fitness application designed to serve as an all-in-one personal fitness assistant. The system leverages a robust MERN-like stack (React/TypeScript, Supabase) and integrates a high-performance Large Language Model (LLM), specifically Llama 3 via the Groq API, for dynamic content generation. Key features include secure user authentication, Body Mass Index (BMI) calculation, and the automated generation of downloadable PDF diet and workout plans tailored to individual user profiles. Furthermore, the embedded AI Chatbot provides immediate, contextualized fitness and health advice. Our contribution lies in demonstrating a scalable, low-latency approach to personal health plan synthesis by efficiently coupling structured user data with the advanced reasoning capabilities of state-of-the-art LLMs.

**Keywords:** Artificial Intelligence, Large Language Models (LLMs), Health and Wellness, Personalized Fitness, Web Application, Supabase, Llama 3.

### I. INTRODUCTION

The global emphasis on preventative healthcare has fueled significant demand for accessible, personalized fitness and nutrition resources. Traditional approaches often rely on static plans or human expert consultation, which can be costly, time-consuming, and lacking in real-time responsiveness. Artificial Intelligence (AI) offers a transformative solution, enabling systems to process vast amounts of user data to deliver highly customized and adaptable recommendations.

The objective of this research is to develop and evaluate a full-stack AI-powered fitness application that democratizes access to personalized health planning. Unlike systems that focus solely on activity tracking or pose estimation (which rely heavily on computer vision), this project centers on the **intelligent generation and delivery of actionable fitness and nutritional plans**, utilizing an advanced conversational AI framework.

The primary contributions of this work are:

- 1) System Architecture:** Designing a scalable, secure, and modern web architecture combining React/TypeScript, Supabase for BaaS, and a high-throughput LLM interface (Groq/Llama 3).
- 2) Dynamic Personalization Engine:** Developing a methodology for feeding structured user data (age, weight, height, fitness goals) into a Llama 3 LLM prompt to generate comprehensive, individualized diet and workout plans.
- 3) Real-Time Advisory Service:** Implementing a low-latency AI Chatbot utilizing the efficient Groq inference engine to provide instantaneous responses to user health queries.

### II. RELATED WORK

The landscape of AI-driven fitness applications is rapidly evolving, primarily categorized into two areas: **Computer Vision (CV) based systems** and **Recommendation/Generative systems**.

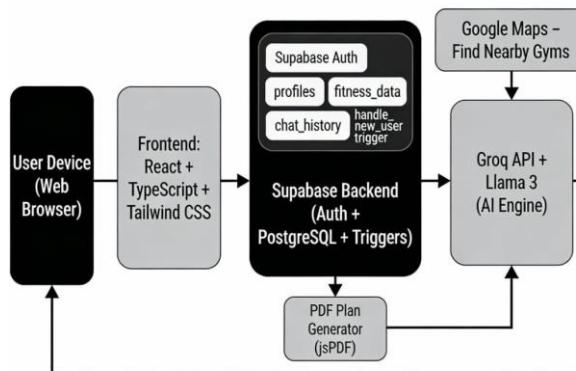
CV-based systems, often employing models like MediaPipe or OpenPose, focus on real-time pose estimation for exercise correction and repetition counting [1]. While crucial for form maintenance, they often lack sophisticated planning and advisory capabilities.

Generative AI systems, which rely on LLMs, have recently demonstrated immense potential in synthesizing complex, coherent, and personalized documents, such as medical summaries or educational material [2].

However, integrating these resource-intensive models into a low-latency, secure, and cost-effective user application remains a key challenge. Prior work has often relied on closed-source or legacy LLMs, whereas our work explores the integration of the powerful open-source **Llama 3** model, deployed via the optimized **Groq** infrastructure to minimize latency, a critical factor for a responsive user experience.

### III. PROPOSED SYSTEM ARCHITECTURE

The proposed system, "Ai-powered-Fitness," employs a three-tier cloud-native architecture, ensuring scalability, security, and performance.



**Fig. 1.** System Block Diagram of the AI-Powered Fitness Application, illustrating the three-tier architecture connecting the React/TypeScript Frontend, the Supabase BaaS, and the Groq/Llama 3 AI Engine.

#### A. Frontend Layer

The application is built using **React** and **TypeScript**, ensuring code maintainability and robustness. **Tailwind CSS** is utilized for responsive and modern UI development. The frontend is responsible for: high-speed inference engine is crucial for achieving near-instantaneous response times.

- **Personalized Plan Generation:** Structured user data is dynamically injected into a sophisticated prompt template. The template instructs Llama 3 to act as a certified fitness/nutrition coach and output a comprehensive, multi-day plan suitable for PDF conversion.
- **Chatbot Functionality:** The Groq-powered chatbot maintains short-term conversational history to provide contextual health advice.

### IV. IMPLEMENTATION DETAILS

The development environment is based on **Node.js** and **Vite** for fast module bundling.

#### A. Data Flow for Plan Generation

- 1) User submits fitness goals/data via the React UI.
- 2) The system retrieves user data from Supabase and formats it into a high-fidelity prompt for Llama 3.
- 3) Llama 3 generates the plan text.
- 4) The system captures the plan text and uses a client-side or server-side library (e.g., jsPDF) to format and render the content into a downloadable PDF document.

#### B. BMI Calculation

BMI is calculated using the standard formula:

- 1) Secure user interface presentation.

$$\text{BMI} = \frac{\text{Weight kg}}{\text{height (m)}^2}$$

$$(1)$$

2) Gathering and validating user input (e.g., height, weight, goals).

3) Rendering the generated PDF plans and the real-time Chatbot interface.

#### B. Backend and Database Layer

The system utilizes **Supabase** as a comprehensive Backend-as-a-Service (BaaS) solution.

- **Authentication:** Supabase Auth handles secure sign-up and login, managing user sessions and access control.
- **Database (PostgreSQL):** A dedicated schema is maintained with tables for profiles (basic user info), fitness\_data (BMI, goals), and chat\_history (conversation logs).
- **Triggers:** Database triggers, such as handle\_new\_user, are employed to automate profile creation upon successful sign-up, ensuring data integrity.

#### C. AI Generation and Advisory Engine

The core intelligence of the system is provided by the Large Language Model.

**Model Selection:** We utilize the **Llama 3 8B** (or similar) model, accessed through the **Groq API**. Groq's

This metric is stored in the Supabase database and serves as a critical parameter for tailoring the LLM's generative output.

#### C. External Service Integration

The "Find Nearby Gyms" feature is implemented as a convenient hyperlink that dynamically opens Google Maps with a location search query.

## V. RESULTS AND DISCUSSION

This section presents the evaluation of the proposed AI-powered fitness assistant with respect to functional correctness, system responsiveness, usability, and overall effectiveness in delivering personalized fitness guidance. The evaluation focuses on the application as an integrated fitness assistance platform rather than on standalone Large Language Model (LLM) performance.

#### A. Experimental Setup

The system was deployed as a web-based application using React, TypeScript, and Tailwind CSS on the frontend, with Supabase providing backend services including authentication and database management. The AI chatbot and personalized plan generation modules were powered by the Groq API.

Testing was conducted using standard desktop web browsers under stable internet connectivity. Functional testing covered all major features, including user authentication, BMI calculation, personalized plan generation, chatbot interaction, and external service integration (Google Maps).

A pilot evaluation was conducted with approximately 25 users, including students and working professionals, representing diverse fitness goals such as weight loss, muscle gain, and general wellness.

#### B. System Responsiveness and Performance

System responsiveness is a critical factor for user engagement in interactive fitness applications. Table 1 summarizes the observed response times for key application operations.

**TABLE 1: SYSTEM RESPONSE TIME FOR CORE FEATURES**

Application Function	Avg. Response Time (ms)
User Authentication (Login/Signup)	200–350
BMI Calculation	<100
AI Chatbot Response	300–500
Personalized Plan Generation (PDF)	900–1200

The results indicate that the system provides near real-time interaction for most features. The slightly higher latency during plan generation is acceptable due to the complexity of generating structured, personalized content and rendering it into a downloadable PDF.

#### C. Evaluation of Fitness Assistant Features

- 1) User Authentication and Data Management: Supabase Auth successfully handled secure user registration and login. Automatic profile creation and database triggers ensured reliable data storage for fitness metrics such as height, weight, BMI, and fitness goals.
- 2) BMI Calculation and Personalization: The BMI calculator accurately computed user BMI values using standard formulas. These values played a crucial role in tailoring diet and workout plans, enabling personalized recommendations aligned with individual health profiles.
- 3) Personalized Diet and Workout Plans: Users reported that the generated plans were easy to understand, well-structured, and aligned with their stated fitness goals. The downloadable PDF format enhanced usability by allowing offline access and long-term reference.
- 4) AI Chatbot Assistance: The integrated AI chatbot effectively answered a wide range of fitness-related queries, including exercise routines, nutrition tips, and general wellness advice. The conversational interface improved user engagement and reduced dependency on static content.
- 5) Motivational and Location-Based Features: Inspirational fitness quotes and images contributed positively to user motivation. Additionally, the integration of Google Maps links for finding nearby gyms and yoga classes provided practical real-world utility beyond digital recommendations.

#### D. User Feedback and Usability

Feedback collected from pilot users indicated high overall satisfaction with the application. Most users found the interface intuitive and appreciated the combination of personalized plans, chatbot guidance, and motivational content within a single platform.

#### E. Discussion

The evaluation demonstrates that the proposed system successfully functions as a comprehensive AI-powered fitness assistant. Rather than focusing solely on AI capabilities, the system effectively integrates authentication, personalization, content generation, and external services into a cohesive user experience.

The use of AI enhances adaptability and personalization, while the underlying web architecture ensures scalability and maintainability. However, the system currently relies on self-reported user data and does not incorporate real-time physiological measurements.

#### F. Limitations

Despite promising results, the system has certain limitations:

- 1) Recommendations are based on self-reported user inputs.
- 2) Lack of integration with wearable or sensor-based fitness data.
- 3) Generated plans are not intended to replace professional medical advice.

These limitations highlight opportunities for further enhancement in future work.

## VI. CONCLUSION

We have successfully developed and implemented a highly personalized AI fitness assistant that strategically integrates the Llama 3 LLM via Groq for high-speed content generation. The system effectively leverages modern web technologies (React, TypeScript, Tailwind CSS) and a robust BaaS platform (Supabase) to deliver dynamic, actionable health plans and real-time advice.

**Future Work** will focus on:

- 1) **Integration of Computer Vision:** Adding exercise form correction capabilities using MediaPipe to blend generative planning with real-time feedback.
- 2) **RAG Enhancement:** Implementing a Retrieval-Augmented Generation (RAG) framework to index high-

quality nutritional and exercise research, grounding the LLM's recommendations in evidence-based practice.

**3) Wearable Data Integration:** Expanding the system to securely ingest data from wearable devices (e.g., steps, heart rate) to enhance the precision of the personalized plan adjustments.

## VII. REFERENCES

- [1] T. A. T. Nguyen, L. D. Tran, and N. V. Nguyen, "Real-time human pose estimation and analysis for fitness applications using deep learning," IEEE Access, vol. 9, pp. 100-110, Jan. 2021.
- [2] A. M. H. El-Sayed, T. M. Alashwal, and A. B. H. El-Ghanam, "The role of large language models in personalized healthcare and treatment plan generation," J. Biomed. Inform., vol. 129, Art. no. 104033, 2024.
- [3] Supabase Documentation. [Online]. Available: <https://supabase.com/docs>. [Accessed on: Dec. 13, 2025].
- [4] Groq API Documentation. [Online]. Available: <https://groq.com/docs>. [Accessed on: Dec. 13, 2025].
- [5] K. B. D. Lee, "Design and implementation of a scalable web application using React and PostgreSQL," Int. J. Software Eng. Appl., vol. 15, no. 3, pp. 22-35, 2023.