



AI Fitness Coach: Personalized Fitness Planning using Large Language Models

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ABSTRACT

Artificial Intelligence has rapidly transformed multiple domains, including healthcare, wellness, and fitness industries, by enabling intelligent decision-making and automation. Traditional fitness systems generally provide static and generalized workout and diet plans, which fail to consider individual user differences such as metabolism, body composition, lifestyle, and personal fitness goals. This lack of personalization often results in reduced user engagement and ineffective outcomes over time.

In this research, an AI-based Fitness Coach system is proposed that leverages the capabilities of Large Language Models (LLMs) to generate personalized workout routines and dietary recommendations. The system interacts with users through a conversational interface, collects essential user data such as age, weight, height, and fitness objectives, and processes this data to generate customized plans. Technologies such as LangChain are used to manage intelligent interactions, Ollama is utilized for local model execution to ensure privacy and reduced latency, and Gradio provides a user-friendly interface for seamless interaction. Furthermore, a feedback mechanism is incorporated to continuously monitor user performance and dynamically update recommendations. The proposed system enhances personalization, improves user motivation, and provides a scalable and efficient solution for modern fitness management systems.

1. Introduction

In the modern era, maintaining a healthy and balanced lifestyle has become increasingly challenging due to rapid urbanization, sedentary work environments, increased screen time, and unhealthy dietary habits. A large portion of the population is now facing lifestyle-related health issues such as obesity, stress, and lack of physical fitness. Although numerous fitness applications and digital platforms are available, most of these systems provide generalized workout and diet plans that do not consider individual differences such as body composition, metabolism, age, and personal fitness goals. As a result, users often fail to achieve their desired outcomes and gradually lose motivation, leading to discontinuation of their fitness routines.

The emergence of Artificial Intelligence (AI) has introduced new opportunities to address these challenges by enabling intelligent, adaptive, and personalized systems. In particular, Large Language Models (LLMs) have demonstrated remarkable capabilities in natural language understanding and generation, allowing systems to interact with users in a conversational manner and provide context-aware recommendations. These models can analyze user inputs, understand preferences, and generate meaningful outputs, making them highly suitable for applications in personalized fitness and healthcare.

The proposed AI Fitness Coach system leverages the power of LLMs to provide customized workout routines and diet plans tailored to individual users. The system collects essential user information such as age, height, weight,

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gender, and fitness goals, and processes this data to determine the user's current fitness level and target objectives. By integrating LangChain, the system enables seamless interaction between the user and the AI model, while Ollama ensures local execution of the model to maintain data privacy and reduce latency. Additionally, the system incorporates a feedback mechanism that continuously monitors user progress and dynamically updates recommendations, ensuring long-term effectiveness and user engagement.

This research aims to overcome the limitations of traditional fitness applications by introducing an intelligent, interactive, and adaptive solution that enhances user motivation and provides a personalized fitness experience. The proposed system not only improves the accuracy and relevance of recommendations but also promotes consistency and long-term adherence to fitness routines. Therefore, the AI Fitness Coach represents a significant step forward in the integration of Artificial Intelligence with healthcare and wellness systems, offering a scalable and efficient solution for modern fitness management [1][2][3].

2. Literature Review

The application of Artificial Intelligence in the domain of fitness and healthcare has evolved significantly over the past decade, with researchers exploring various techniques to improve user experience and effectiveness of fitness systems. Early fitness applications primarily focused on basic activity tracking functionalities such as step counting, calorie estimation, heart rate monitoring, and workout duration analysis. While these systems provided useful data, they lacked the capability to interpret this data in a meaningful way and deliver personalized recommendations tailored to individual users. Consequently, their effectiveness in improving long-term fitness outcomes was limited.

With the advancement of machine learning and deep learning technologies, more sophisticated systems were developed that could analyze user data and generate recommendations based on patterns and trends. These systems introduced features such as personalized workout suggestions and dietary recommendations; however, they were still limited in terms of adaptability and interaction. Most of these systems relied on predefined rules or static models, which restricted their ability to handle dynamic user requirements and evolving fitness goals.

The introduction of Large Language Models has significantly enhanced the capabilities of AI-based systems by enabling natural language interaction and context-aware response generation. LLMs can understand complex user queries, process contextual information, and generate human-like responses, making them highly effective for conversational AI applications. Recent studies have shown that LLM-based systems can improve user engagement and provide more relevant and personalized recommendations compared to traditional approaches.

Despite these advancements, several challenges still exist in current implementations. Many existing systems depend on cloud-based infrastructures, which can introduce latency and raise concerns regarding data privacy and security. Additionally, most systems do not incorporate real-time feedback mechanisms, limiting their ability to adapt recommendations based on user progress. The lack of continuous learning and adaptability reduces the overall effectiveness of these systems in long-term fitness management.

The proposed AI Fitness Coach addresses these limitations by integrating local model execution using Ollama, which ensures faster processing and enhanced data privacy. Furthermore, the system incorporates a dynamic feedback mechanism that continuously updates recommendations based on user performance, making it more adaptive and effective. By combining the strengths of Large Language Models, conversational AI, and real-time feedback, the proposed system provides a more comprehensive and intelligent solution for personalized fitness management compared to existing approaches [2][3][4].

3. Proposed System

The proposed AI Fitness Coach is designed as an intelligent, adaptive, and user-centric system that provides personalized fitness and nutrition recommendations using the capabilities of Large Language Models (LLMs). The primary objective of this system is to overcome the limitations of traditional fitness applications by delivering customized solutions that are tailored to individual user requirements. Unlike conventional systems that provide

fixed and generalized plans, the proposed system dynamically generates workout routines and diet plans based on user-specific data and continuously updates them according to user progress.

The system begins by collecting essential user information through an interactive interface. This information includes parameters such as age, height, weight, gender, current fitness level, and fitness goals (such as weight loss, muscle gain, or general fitness improvement). This data is then analyzed to determine the user's current physical condition and target objectives. The use of structured input ensures that the recommendations generated by the system are accurate and relevant.

To enable intelligent interaction, the system integrates LangChain, which acts as a framework for managing communication between the user and the Large Language Model. This allows the system to understand user queries, maintain conversational context, and generate meaningful responses. Ollama is used for executing the LLM locally on the system, which ensures data privacy, reduces dependency on external servers, and minimizes response latency. This approach makes the system more secure and efficient compared to cloud-based solutions. Once the user data is processed, the system generates personalized workout routines and diet plans that align with the user's goals and preferences. The recommendations are presented in a structured and easy-to-understand format. A key feature of the proposed system is its feedback mechanism, which continuously monitors user performance and collects feedback after each interaction. Based on this feedback, the system updates its recommendations dynamically, ensuring that the fitness plan evolves over time according to the user's progress. Additionally, the system can provide motivational messages, reminders, and suggestions to encourage users to maintain consistency in their fitness routines. This improves user engagement and helps in achieving long-term fitness goals. The combination of personalization, adaptability, real-time interaction, and privacy-focused design makes the proposed AI Fitness Coach a powerful and effective solution for modern fitness management systems [1][2][3].

4. System Architecture

Fig. 1 illustrates the overall system architecture of the proposed AI Fitness Coach, which is designed using a layered and modular approach to ensure efficient data processing, scalability, and real-time interaction. The architecture consists of four main components: the input layer, processing layer, output layer, and a feedback mechanism.

The input layer is responsible for collecting user-specific information such as age, height, weight, gender, and fitness goals through a user-friendly interface developed using Gradio. This layer acts as the entry point of the system and ensures that the data provided by the user is structured and ready for further processing.

The processing layer forms the core of the system, where all computations and decision-making processes take place. In this layer, LangChain is used to manage the interaction between the user and the Large Language Model, enabling smooth conversational communication. The Large Language Model is executed locally using Ollama, which ensures data privacy, reduces latency, and eliminates dependency on cloud-based services. This layer analyzes user inputs, identifies patterns, and generates personalized workout and diet recommendations.

The output layer presents the generated recommendations in a clear and structured format, including customized workout plans and dietary guidelines. This ensures that users can easily understand and follow the recommendations provided by the system.

Additionally, a feedback loop is integrated into the architecture to continuously monitor user progress and update recommendations accordingly. This feedback mechanism allows the system to adapt dynamically based on user performance, making it more effective and personalized over time.

Overall, the proposed architecture ensures a seamless flow of data between different components, enabling efficient, secure, and intelligent fitness management. The modular design also allows for future enhancements such as wearable device integration, real-time health monitoring, and advanced analytics capabilities [2][3][4].



Fig. 1: System Architecture of AI Fitness Coach

5. Methodology

The methodology of the proposed AI Fitness Coach follows a systematic and structured approach to ensure accurate, personalized, and adaptive fitness recommendations. The process begins with the **data collection phase**, where user-specific information such as age, height, weight, gender, current fitness level, and fitness goals is collected through a user-friendly interface developed using Gradio. This step is crucial as the quality and accuracy of input data directly influence the effectiveness of the generated recommendations.

After data collection, the system performs **data preprocessing and analysis**, where the input data is organized and key parameters such as Body Mass Index (BMI), fitness category, and goal type are identified. This preprocessing step helps in understanding the user's current physical condition and aligning the recommendations accordingly. The system ensures that the data is structured in a way that can be effectively utilized by the AI model.

In the next stage, the system integrates **Large Language Models (LLMs)** using LangChain, which enables intelligent and context-aware interaction with users. LangChain manages the flow of communication, maintains conversational context, and ensures that the responses generated by the model are relevant and meaningful. Ollama is used to execute the LLM locally, which enhances data privacy, reduces latency, and ensures faster processing compared to cloud-based solutions.

Once the model processes the user data, the system performs **plan generation**, where personalized workout routines and diet plans are created based on the user's goals and fitness level. These plans are structured in a clear and easy-to-follow format, including exercise types, duration, frequency, and dietary suggestions. The system ensures that the recommendations are practical and achievable for the user.

A key component of the methodology is the **feedback mechanism**, which continuously monitors user performance and collects feedback after each interaction. This feedback is used to update and refine the recommendations dynamically. The system adapts to changes in user progress, ensuring that the plans remain relevant and effective over time. This iterative process improves the overall accuracy and efficiency of the system. Finally, the entire system is implemented using Gradio, which provides a real-time and interactive interface for seamless communication between the user and the AI model. This structured methodology ensures reliability, scalability, and improved user satisfaction, making the system highly effective for personalized fitness management [1][2][3].

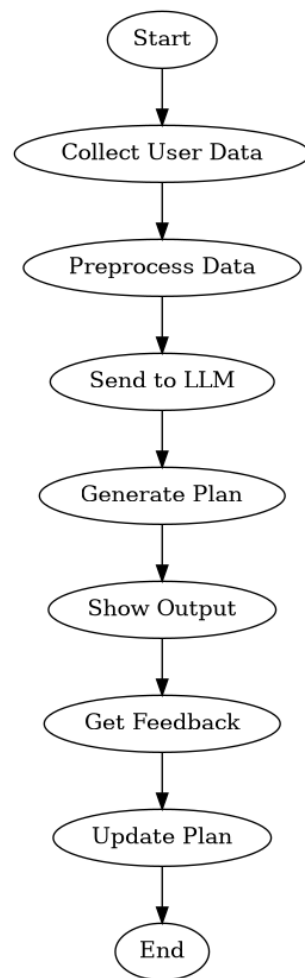


Figure 2: System Workflow

6. Results and Discussion

The implementation of the proposed AI Fitness Coach demonstrates significant improvements over traditional fitness systems in terms of personalization, adaptability, and user engagement. The system successfully generates customized workout routines and diet plans based on individual user profiles, which leads to better alignment with user goals and improved effectiveness of fitness programs. Unlike conventional applications that provide static recommendations, the proposed system continuously updates its suggestions based on user feedback and performance, making it more dynamic and responsive.

The interactive and conversational nature of the system plays a crucial role in enhancing user engagement. By allowing users to communicate with the system in natural language, the AI Fitness Coach creates a more intuitive and user-friendly experience. This increases user motivation and encourages consistent participation in fitness activities. As a result, users are more likely to adhere to their fitness routines and achieve their desired outcomes. Another important aspect of the system is its **privacy-focused design**, achieved through local execution of the Large Language Model using Ollama. This eliminates the need for constant internet connectivity and reduces dependency on external servers, making the system more secure and reliable. Additionally, the reduced latency ensures faster response times, improving the overall user experience.

The feedback mechanism integrated into the system further enhances its effectiveness by enabling continuous

improvement of recommendations. As users provide feedback and update their progress, the system adapts its suggestions accordingly, ensuring that the fitness plans remain relevant and personalized. This dynamic adaptability is a key advantage over traditional systems, which often fail to adjust to changing user needs. Comparative analysis with existing fitness applications indicates that the proposed system performs better in terms of personalization, adaptability, and user satisfaction. The ability to combine conversational AI, real-time feedback, and local model execution makes the system a comprehensive solution for modern fitness management. Overall, the results demonstrate that the AI Fitness Coach has strong potential for real-world implementation and can significantly contribute to improving health and fitness outcomes [2][3][4].

7. Advantages

The proposed AI Fitness Coach offers several significant advantages over traditional fitness systems by incorporating advanced Artificial Intelligence techniques and personalized recommendation strategies. One of the primary advantages of the system is its ability to provide **highly personalized workout and diet plans** based on individual user profiles. Unlike conventional applications that offer generic recommendations, this system analyzes user-specific parameters such as age, weight, height, and fitness goals to generate tailored fitness plans, resulting in improved effectiveness and user satisfaction.

Another key advantage of the system is its **real-time interaction capability** through conversational AI. By leveraging Large Language Models, the system enables users to communicate naturally and receive instant responses, making the interaction more engaging and user-friendly. This conversational approach enhances user motivation and encourages consistent participation in fitness activities.

The system also ensures **data privacy and security** by executing the Large Language Model locally using Ollama. This eliminates the need to send sensitive user data to external cloud servers, thereby reducing privacy risks and ensuring secure data handling. Additionally, local execution significantly reduces latency, leading to faster response times and improved system performance.

A major strength of the proposed system is its **adaptive feedback mechanism**, which continuously monitors user progress and updates recommendations accordingly. This dynamic adaptability ensures that the fitness plans remain relevant and effective over time, unlike static systems that fail to adjust to changing user conditions.

Furthermore, the system is designed with a **modular and scalable architecture**, allowing easy integration of additional features such as wearable device connectivity, mobile applications, real-time health monitoring, and advanced analytics. This flexibility makes the system future-ready and suitable for large-scale deployment. Overall, the combination of personalization, adaptability, privacy, and scalability makes the proposed AI Fitness Coach a powerful and efficient solution for modern fitness management [2][3][4].

8. Conclusion

In conclusion, this research presents an intelligent AI-based Fitness Coach that leverages the capabilities of Large Language Models to deliver personalized and adaptive fitness recommendations. The proposed system effectively addresses the limitations of traditional fitness applications, which often rely on static and generalized plans that fail to meet individual user requirements. By integrating conversational AI, real-time interaction, and a dynamic feedback mechanism, the system provides a more engaging, flexible, and user-centric fitness experience.

The use of technologies such as LangChain for managing interactions, Ollama for local model execution, and Gradio for user interface development ensures efficient system performance, enhanced data privacy, and seamless communication. The system not only generates customized workout routines and diet plans but also continuously updates them based on user feedback and progress, ensuring long-term effectiveness and improved health outcomes.

The results demonstrate that the proposed system significantly improves user engagement, motivation, and adherence to fitness routines compared to conventional approaches. The ability to combine personalization with real-time adaptability makes the system highly effective for modern fitness management.

Moreover, the modular architecture of the system allows for future enhancements and integration with emerging technologies such as wearable devices, mobile platforms, and advanced health analytics. With further development and real-world implementation, the AI Fitness Coach has the potential to evolve into a comprehensive digital health solution that can contribute to improving overall well-being and promoting healthier lifestyles. Thus, this research highlights the growing importance of Artificial Intelligence in transforming the fitness and healthcare domain and provides a strong foundation for future innovations in personalized fitness systems [1][2][4].

9. Future Scope

The proposed AI Fitness Coach system demonstrates significant potential; however, there are several opportunities for further enhancement and expansion. One of the major future improvements includes the **integration of wearable devices** such as smartwatches and fitness bands. By incorporating real-time data such as heart rate, step count, sleep patterns, and caloric expenditure, the system can provide more accurate and dynamic recommendations, thereby improving overall effectiveness.

Another important direction for future development is the creation of a **mobile application** that can make the system more accessible and user-friendly. A dedicated mobile app would allow users to interact with the AI Fitness Coach anytime and anywhere, increasing convenience and engagement. Additionally, push notifications and reminders can be integrated to help users stay consistent with their fitness routines.

The system can also be enhanced by incorporating **voice-based interaction** using speech recognition and text-to-speech technologies. This would enable users to communicate with the system more naturally, making it more interactive and accessible, especially for users who prefer hands-free operation. Another promising area is the integration of the system with **healthcare platforms and professional trainers**, which can provide expert validation and improve the reliability of recommendations. The system can also be extended to support multiple languages, making it accessible to a wider audience globally. Overall, the future scope of the AI Fitness Coach lies in transforming it into a **comprehensive digital health ecosystem** that combines Artificial Intelligence, IoT, and data analytics to deliver intelligent, real-time, and personalized fitness solutions [2][3][4].

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