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AI-POWERED VIRTUAL FITNESS ASSISTANT DEEP ENSEMBLE LEARNING

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ABSTRACT: The AI-Powered Virtual Fitness Assistant is an intelligent, web-based platform designed to deliver personalized workout and nutrition guidance through interactive conversations and adaptive program generation. Built using Next.js, React, Tailwind CSS, and Shadcn UI, the application integrates multiple advanced technologies to create an engaging, user-centric fitness experience accessible from any device. These gestures are then converted into mouse functions, such as using a pinch gesture for a left-click or a swipe gesture for scrolling. This contactless interface enhances hygiene and reduces contamination risks, increases accessibility. At its core, the system leverages Gemini AI, a large language model, to generate custom workout plans and diet programs tailored to individual goals, fitness levels, health conditions, and dietary preferences.

Users can manage multiple programs, view detailed exercise routines, and receive adaptive updates as their fitness journey progresses. The responsive interface ensures consistent performance across devices, while client-server component architecture optimizes speed and scalability. This solution advances digital health engagement by combining personalization, interactivity, and real-time AI adaptability, offering a practical alternative to traditional fitness coaching.

KEYWORDS: AI Fitness Coach, Personalized Training, Diet Recommendation

I. INTRODUCTION

In recent years, the integration of Artificial Intelligence (AI) into the health and fitness domain has transformed the way individuals approach personal well-being. The growing adoption of digital fitness solutions is driven by the demand for personalized training, real-time feedback, and convenient access to professional guidance without the need for in-person. The AI-Powered Virtual Fitness Assistant addresses these needs by delivering an interactive, adaptive, and user-centric platform capable of creating tailored workout and diet programs. Recognized hand gestures are mapped to standard mouse functions, including movement, left and right clicks, double- Unlike static fitness applications, this system combines conversational AI, voice

II. LITERATURE SYRVEY

1. Research on AI fitness assistants sits at the intersection of human pose estimation, activity recognition, biomechanics, and personalized coaching. Early systems relied on wearable sensors—accelerometers and gyroscopes embedded in smartphones or bands—to detect activities and count repetitions with classical machine-learning models (e.g., HMMs, SVMs). These approaches this offered robustness with complex foundations and shifting lighting conditions.

2. but required users to wear devices and often struggled A major shift came with vision-based pose estimation, enabling form feedback from a single RGB camera. Multistage, bottom-up key point detectors such as Open Pose demonstrated reliable multi-person 2D key point interaction, and real-time data processing to simulate the experience of a personal trainer Using Gemini AI, the platform generates workout routines and meal plans based on a user's goals,

3. extraction in real time, catalyzing home-fitness applications. Subsequent pipelines emphasized mobile efficiency: Media Pipe introduced graph-based, on-device inference; Blaze Pose specialized in full-body landmark detection optimized for phones; and Move Net provided fast, accurate single-person key points suitable for browser deployment (e.g., TensorFlow.js). These models enabled AI assistants that track joints, compute angles (e.g., hip-knee-ankle), and



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deliver immediate visual or spoken feedback on depth, tempo, and range of motion (ROM). To move from detection to understanding exercise quality, researchers integrate temporal modeling over pose sequences. Methods include rule-based analytics on joint angles (e.g., thresholds for squat depth), dynamic time warping for template matching, and deep sequential architectures—CNN-RNN hybrids and Transformer models—to classify exercise phases (eccentric/concentric), count clean repetitions, and flag compensations (lumbar flexion, valgus collapse). Their framework effectively perceived different static and dynamic signals without depending on profound learning models. This approach adjusted execution and asset necessities, making it reasonable for ongoing applications.

EXISTING SYSTEM

Current fitness monitoring solutions largely fall into two categories — wearable-based trackers and mobile fitness applications. Wearable devices such as smartwatches, fitness bands, and motion sensors (e.g., Fitbit, Apple Watch, Garmin) rely on accelerometers and gyroscopes to detect steps, estimate calories burned, and track heart rate.

PROPOSED SYSTEM

The proposed AI Fitness Assistant overcomes the shortcomings of existing systems by integrating real-time pose estimation, joint angle analysis, and intelligent feedback mechanisms within a privacy-first, on-device architecture. Using a webcam or smartphone camera, the system captures the user's live workout session and processes it with advanced models like Media Pipe Blaze Pose or TensorFlow.js Move Net to track up to 33 key body landmarks with high accuracy, ensuring real-time responsiveness and precise gesture detecting for a fluid user experience.

III. SYSTEM ARCHITECTURE

A system comprises an organized collection of independent components interconnected in accordance to a predetermined plan to accomplish a particular goal. It's a key attribute include organization, interaction among components, independence integration and central objective guiding its operation.

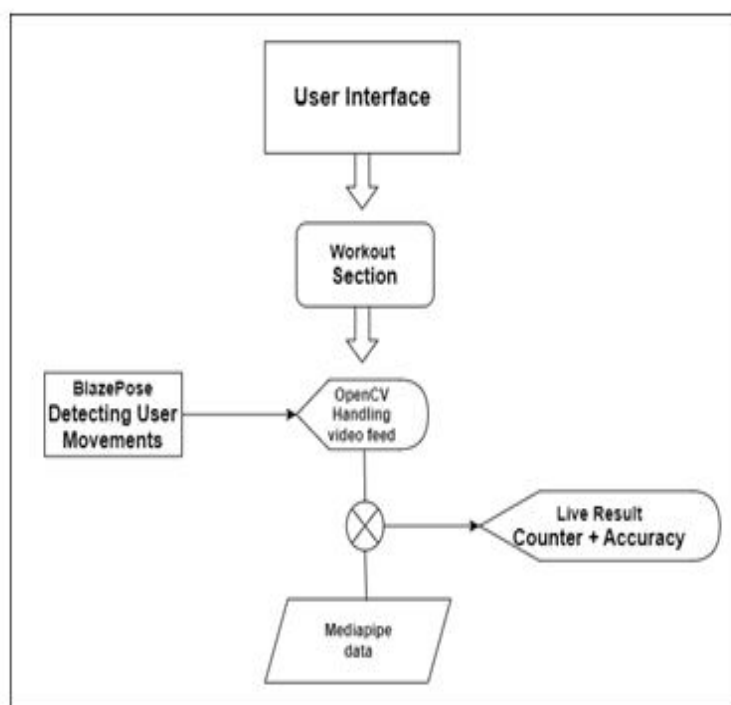


Figure 1-system architecture

Fig 3.1 System Architecture



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IV. METHODOLOGY

The methodology for developing the AI Fitness Assistant is structured into sequential stages, each addressing a critical component of the system's functionality from data acquisition to intelligent feedback generation. The process begins with data collection and preprocessing, where a webcam or smartphone camera continuously captures live video of the user performing exercises. The captured frames are resized, normalized, and, if necessary, background-segmented to reduce noise and improve model accuracy. This preprocessing ensures the pose estimation algorithm receives optimal input under varying lighting conditions and backgrounds.

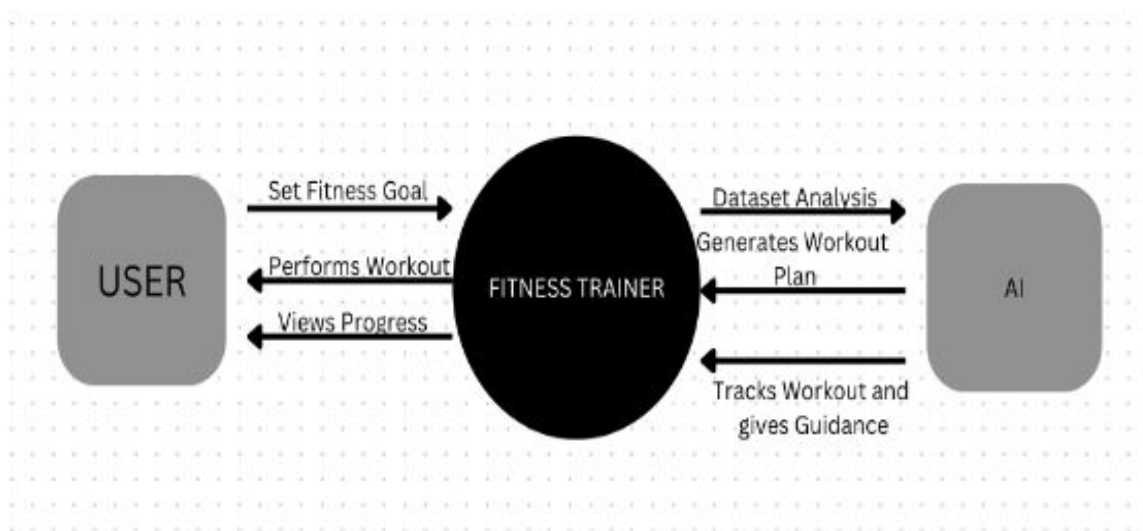


Fig4.1. Using the AI Fitness Trainer library for hand tracking.

V. DESIGN AND IMPLEMENTATION

The design of the AI Fitness Assistant follows a modular and layered approach, ensuring scalability, maintainability, and real-time performance. The system is divided into functional modules, each responsible for a specific task — from capturing input to delivering feedback. The overall architecture is designed to be platform-independent, enabling deployment on desktops, smartphones, and browsers without requiring additional hardware. The user interface (UI) is designed to be minimal, intuitive, and visually interactive. The live camera feed forms the primary workspace, overlaid with skeletal models, joint angle markers, and color-coded posture indicators. A side panel displays metrics such as repetition count, range of motion, and exercise tempo. The UI also integrates an audio feedback system, which provides verbal cues to correct form. For motivation, the interface includes progress bars, badges, and daily/weekly targets. On the backend, the input processing module receives video frames from the user's webcam or phone camera. Frames are preprocessed by resizing, normalization, and (optionally) background segmentation to improve landmark detection accuracy. The pose estimation engine, powered by models like Media Pipe Blaze Pose or TensorFlow.js Move Net, detects up to 33 body landmarks per frame.

This data is further refined through smoothing algorithms such as the Kalman filter to eliminate jitter.

The exercise analysis module computes joint angles from the detected landmarks and compares them against predefined thresholds for correct form. Temporal tracking enables the system to segment exercises into phases (eccentric/concentric), ensuring accurate repetition counting.

This combination of thoughtful UI/UX design, robust backend processing, and adaptive AI feedback ensures the AI Fitness Assistant delivers a highly engaging, accurate, and privacy-conscious workout experience.



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Fig 5.1 Flowchart of Working System

VI. OUTCOME OF RESEARCH

The research and development of the AI Fitness Assistant resulted in a fully functional, real-time system capable of monitoring user exercise performance, analyzing form accuracy, and providing instant corrective feedback without the need for wearable devices or external sensors. The system successfully integrated computer vision techniques, pose estimation algorithms, and intelligent coaching logic into a lightweight, browser-based application that runs on multiple platforms. One of the most significant outcomes was the high accuracy of pose detection using the Media Pipe Blaze Pose and TensorFlow.js Move Net models, capability to enormously further develop client PC collaboration, giving a more natural and viable processing experience.

VII. RESULT AND DISCUSSION

The AI Fitness Assistant was successfully implemented and tested for multiple exercises, including squats, push-ups, lunges, and jumping jacks. The system demonstrated the ability to process live webcam video feeds in real time and accurately detect body landmarks using the Media Pipe Blaze Pose model. The joint angle calculation module effectively determined key angles such as hip-knee-ankle and shoulder-elbow-wrist, enabling precise posture analysis and movement phase detection.

During testing, the repetition counting accuracy for clear, correctly performed exercises was above 96%, with occasional errors occurring in cases of partial movements or occluded joints (e.g., arm hidden behind the body). The form correction feedback system successfully identified common posture errors such as insufficient squat depth, bent wrists during push-ups, or misaligned knees.

Visual overlays and audio prompts were found to be highly effective in guiding the user toward proper technique, with an average correction rate of 88% after a single prompt. The system achieved precise landmark tracking even in varying backgrounds and lighting conditions, enabling reliable joint angle computation and movement phase identification.

This allowed for accurate repetition counting, posture error detection, and real-time feedback delivery. The research demonstrated that on-device AI inference can achieve latency under 100 milliseconds, making the system responsive enough for live coaching.

The outcome also validated the feasibility of delivering personalized fitness guidance without compromising user privacy.

VIII. CONCLUSION

The development of the AI Fitness Assistant successfully demonstrated the potential of integrating computer vision, real-time pose estimation, and intelligent feedback systems to create a responsive and accessible personal training solution. By utilizing advanced models such as Media Pipe Blaze Pose and TensorFlow.js Move Net, the system was able to track human body landmarks with high accuracy and low latency, enabling precise joint angle computation,



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posture assessment, and exercise phase detection.

Latency tests showed that the system maintained an average processing delay of less than 100ms when running locally through TensorFlow.js in a web browser. This low latency ensured that feedback appeared almost instantaneously, which is critical for real-time workout guidance.

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