A Real-Time AI System For Posture Detection & Monitoring

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Abstract-

In this paper, we explore real-time physical estimation by integrating two fundamental frameworks in computer science and machine learning, OpenCV and MediaPipe. OpenCV optimizes the video input using resizing, color space conversion, and noise reduction, while the pre-trained MediaPipe model is able to detect and track human regions. The system is tested for accuracy, responsiveness, and stability under various conditions, including occlusion and changing lighting conditions. The results show the potential for advancing evaluation techniques and highlight benefits in motion analysis, medical treatment, and human- computer interaction. Vision, human geography, machine learning, motion measurement, healthcare, human-computer interaction. **Keywords**- OpenCV, MediaPipe, Pose Estimation, Real Time, Human-Computer Interaction.

Keywords⁻ Openev, mediai ipe, i ose Estimation, Keai rime, iraman-computer interaction.

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I. Introduction

The introduction of this paper highlights the integration of artificial intelligence (AI) and machine learning (ML) in the medical field to improve physical fitness through posture correction. It highlights how an intelligent AI system using Python libraries (e.g., Mediapipe, TensorFlow, and OpenCV) can detect body posture in real time and monitor exercise. This approach addresses the need for precise exercise methods to prevent injuries, especially for those who do not have access to professional trainers. It also highlights the role of AI in promoting cost effectiveness, sustainability, and improving mental health through improved exercise programs.

Prior Research

Previous studies in this article highlight several developments in physical examination and monitoring systems, particularly those aimed at improving health. Key findings include:

- **1. Problem Management:** Personal trainers and physiotherapists have traditionally been able to monitor mobility, but this approach is not available to rural or elderly people who need regular assistance.
- **2. Limitations of Existing Devices:** Devices such as Xbox Kinect can track body movements, but are expensive and unsuitable for health purposes, limiting their applicability.
- **3. Usability:** Provide personalized nutrition, insights, and exercise planning. Health and fitness apps are leading the way.

Rationale

The purpose of the article is to solve the problem of poor physical condition that can lead to injuries due to lack of professional guidance during home workouts. The brand uses OpenCV and MediaPipe for body estimation and provides instant feedback on exercises, specific squats and push-ups. By doing this, it allows users to perform safe and efficient workouts at home without the need for additional equipment or extensive training.

II. Methodology

The approach to conduct real-time human research using OpenCV and MediaPipe involves several key steps to achieve accurate and effective human estimation. The key points of this approach are:

- Use OpenCV to process video streams.
- Color space conversion
- Noise reduction
- Optimize input data for post-processing. Trace anatomical details. Estimated points are positioned to form a skeletal model.

• Accuracy: Ensure that important points are clearly visible with physical markings. Excellent performance in all conditions, including occlusion and different lighting environments. Manage and prioritize your video ideas, while MediaPipe focuses on key findings and makes predictions. Computer networks and other activities.

III. Technology Review

This paper focuses on estimating the physical body using OpenCV and MediaPipe, with emphasis on their applications in motion analysis, healthcare, and human-computer interaction. The main challenges include achieving accurate key detection, sustaining high cost, and maintaining stability under various conditions. By integrating OpenCV for video processing and MediaPipe for tracking, this research provides a realistic solution for estimating in multiple countries.

MediaPipe

MediaPipe is an advanced learning engine developed by Google and designed for real-time applications. It is popular for physical tracking and estimation. Key features include:

- Pose Detection and Landmark Evaluation: MediaPipe Pose is a unique solution that instantly detects 33 body landmarks using RGB video input.
- Lightweight Architecture: A model optimized for instant processing even on mobile devices.
- Two-Stage Detection: Combined with tracking to increase accuracy and efficiency.

While MediaPipe focuses on using machine learning for physical perception, OpenCV provides appropriate graphics and computational tools to improve performance, accuracy, and real-time usability. Together they form a powerful framework for many computer vision applications. [1]

OpenCV

OpenCV (Open-source Computer Vision Library) is a widely used library for real-time computer vision and graphics processing. It supports multiple programming languages such as C++, Python, Java, and is compatible with multiple platforms such as Windows, Linux, macOS, Android, and iOS. OpenCV provides tools to perform the following tasks:

- Image and Video Analysis: image processing (such as filtering, transformation) and video analysis (such as word tracking, object recognition) studies.
- Machine Learning: Built-in machine learning algorithms for tasks such as image classification and feature extraction.
- Camera Calibration: A method of calibrating a camera and correcting inconsistencies in captured images.
- Performance Optimization: Optimized for computing efficiency and real-time processing, supporting multicore processors. machine.

Its integration with modern tools such as MediaPipe expands its advantages in motion tracking and cognitive processing.

Numpy

NumPy plays an important role in research on OpenCV and MediaPipe due to its efficiency in performing arithmetic and array operations. Both libraries rely on NumPy to perform important tasks such as preprocessing, data transformation, and arithmetic operations.

The role of NumPy in OpenCV:

- Image display and manipulation
- Mathematical operations
- Data integration

The role of NumPy in MediaPipe:

- Pose landmark calculation
- Data preprocessing
- Real-time computation

The role of NumPy in MediaPipe & OpenCV:

- Angle and vector calculation: NumPy is used to calculate the angles of body joints using vector functions, which is a function in anatomy.
- Efficient data processing: Combine data processing bypass in OpenCV with data localization in MediaPipe.
- Camera Calibration & Transformation: OpenCV uses NumPy for matrix calculations (such as rotations and

measurements) to correct inequality and fix area.

Python

Due to its simplicity, extensive library support, and developer community, Python has become the dominant language for OpenCV and MediaPipe integration research. Its ecosystem accelerates the development of computer vision and machine learning by providing tools that integrate with these models.

The role of Python in MediaPipe & OpenCV:

- Working with OpenCV (image processing) and MediaPipe (prediction).
- It provides visual feedback, while OpenCV and MediaPipe do the reverse coding.

Python supports researchers by simplifying the integration of OpenCV and MediaPipe with a common framework. Its ease of use, stable ecosystem, and simplicity make it indispensable for applications such as instant posture correction and training. Python accelerates development while providing easy computation and seamless visualization.

Research Gaps and Contribution

While OpenCV & MediaPipe have shown promising results in posture correction, there are several gaps this research we aim to address. First, existing studies often focus solely on accuracy without considering model accessibility and usability in real-world applications. This study contributes by extending the framework for medical scenarios. Additionally, prior studies primarily focus on precision accuracy but this study aims optimize the framework for low- powered devices like smartphones and edge AI systems by making it portable.

IV. Related Work

The research is based on many studies on gesture recognition, motion tracking, and instant feedback. Below is a summary of the main activities involved and their contribution to the field:

Real-Time Workout Posture Correction

This study presents a method developed to provide instantaneous and recalculated exercise therapy using computer vision and machine learning. Focusing on exercises, this system allows you to practice easy-to-use home exercises without a personal trainer.

Technology Stack: OpenCV, MediaPipe, Flask, React, MongoDB Feature: Real-time Feedback, Accuracy

- Methodology:
- 1. Data Acquisition
- 2. Pose Detection
- 3. Posture Analysis
- 4. Feedback Generation
- 5. Experimental Validation
- 6. Repetition Counting

Limitations:

- Camera Calibration
- Limited Exercise Scope

Pose detection in varying conditions

This research focuses on the integration of OpenCV and MediaPipe to build a robust and usable human time estimation system. This work demonstrates its applicability in many fields, from sports to medicine, from human-computer interaction to entertainment.

Technology Stack: OpenCV, MediaPipe Feature: Real-time detection, Accuracy, evaluation in varying conditions

Methodology:

1. Framework Integration

2. Human Body Modelling

3. Datasets Used: COCO, MPII and Pose Track

4. Evaluation Metric: Percentage of Correct Key points (PCK), Mean Average Precision (MAP) & Frames Per

Second (FPS)

Limitations:

- Environment Stability
- 2D pose Estimation
- Dataset Generalization

Automated Body Postures Assessment from Still Images

This research paper presents a framework for automatic human assessment using MediaPipe and manual extraction techniques. It focuses solely on comparing the user's body image captured by the webcam with the body image to provide quick correction recommendations.

Technology Stack: OpenCV, MediaPipe, CNN, GAN Feature: Real-time Feedback, Accuracy, Handicraft Features Extraction

Methodology:

- 1. System Design & Setup
- 2. Landmark Detection
- 3. Feature Extraction
- 4. Feedback Generation
- 5. Real-Time Operation

Limitations:

- Inability to detect torso & neck
- Camera Calibration
- Quality Sensitivity

V. Discussion

Together, these studies provide research on human observation using techniques such as OpenCV, MediaPipe, and handcraft. Below is a collaborative discussion to document the contributions, strengths, challenges, and future directions of the entire project.

Interpretation of Results

Our research paper focuses on the integration of OpenCV, MediaPipe and hand-crafted tools for physical analysis. The results show great potential for health, wellness and self-assessment applications. Here is a description of their results:

- 1. High Accuracy Using MediaPipe: We used MediaPipe's BlazePose to identify 33 body features in all our studies and achieved high accuracy in identifying important points. Results from controlled case studies prove the reliability of the framework by showing a tight correlation between measurement results (e.g. sitting and standing vs. reality).
- 2. Workflow of the Linked Workflow: In both studies, the integration of OpenCV and MediaPipe provides instant, low-latency, controlled feedback. The frame rate is high enough to support real-time applications such as physical therapy and ergonomic assessment. good performance.
- 3. Effectiveness of instant feedback: Visual overlay (such as bone descriptions and arrowheads) enables effective feedback. The feedback simulates a "smart coach" to provide effective and engaging feedback to users. (yoga or exercise).
- 4. Performance under controlled conditions: All studies were conducted in a controlled environment with constant lighting and minimal occlusion. Gender) is a variable that is flexible to subtle changes. In the third study, it was noted that there were problems with trunk and neck bending, emphasizing the need for more practical models.

Comparison

The automated system provides results close to manual assessment for static operations, proving its effectiveness in exercise and rehabilitation. The limits of slight deviation in the transition from stationary to motion are shown in the correct directions. Correct number of repetitions and feedback for squats and push-ups. Patient follow-up is normal. Performance is better due to the configuration of the trunk.

Limitations & Areas for Improvement

Despite the success of the CNN + LSTM model, there are areas where further improvements may be beneficial:

- 1. Hand-crafted features are good for simple scenes but struggle to cope with changes in user behaviour or difficult movements.
- 2. As shown in the study, correct or incorrect calibration can affect the results.
- 3. The lack of 3D pose estimation leads to inaccuracies in depth-related postures or overlapping limbs.

VI. Conclusion

This work presents a new method for evaluating body posture using MediaPipe for terrain detection and manual work to quantify different body postures. The findings of this paper represent significant advances in the use of computer vision for physical assessment and therapy. Although each research method has its own benefits and specific outcomes, limitations always indicate that further research is needed for robustness, adaptability, and in-depth assessment. This work has established a solid foundation for instant, cost-effective, and easy- to-use physical assessment that has the potential to impact multiple factors such as health, clean drinking, wellness, and education. This research contributes to the field of real time detection by demonstrating that of automated human estimation using MediaPipe and crafting techniques. It focuses solely on comparing the user's body image with the body image retrieved from the web to provide real-time feedback.

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