Abstract—This paper presents FITORBIS, an innovative alarm system designed to promote morning productivity by requiring users to complete physical exercises (e.g., sit-ups) to deactivate the alarm. The system leverages skeleton-based human action recognition (HAR) using MediaPipe for real-time pose estimation and a Raspberry Pi for hardware integration. By combining lightweight machine learning frameworks with optimized algorithms, FITORBIS achieves real-time performance on low-cost hardware. Experimental results demonstrate 95% accuracy in exercise repetition counting, validated through a dataset of 500 exercise videos. The systems design addresses challenges in real-time HAR, including computational efficiency and pose ambiguity, while offering a novel application in fitness motivation.

Index Terms—Human Action Recognition, MediaPipe, Real-Time Systems, Fitness Applications, Pose Estimation

I. Introduction

Traditional alarm clocks often fail to encourage physical activity, leading to sedentary morning routines. FITORBIS addresses this gap by integrating fitness into wake-up rituals through AI-driven action recognition. Users preconfigure exercises (e.g., 10 sit-ups), and the alarm deactivates only upon successful completion.

A. Contributions

- Real-Time HAR Pipeline: Combines MediaPipe Pose
 for 2D joint extraction and angle-based validation for exercise counting.
- 2) **Hardware-Software Integration**: Raspberry Pi processes camera input and runs pose estimation using CPU-only optimization.
- Open-Source Implementation: Python-based code with OpenCV and MediaPipe ensures accessibility and customization.

II. RELATED WORK

A. Pose Estimation

Kim et al. [1] proposed a two-stage method using MediaPipe Pose (MPP) for 2D landmark detection and uDEAS optimization for 3D joint angles. While FITORBIS focuses on 2D pose estimation, their work validates the robustness of MPP in real-time applications.

B. Action Recognition

Rodrguez-Moreno et al. [2] introduced CSP-based filtering for video-to-image transformation, improving action recognition accuracy. Kang et al. [4] optimized skeleton-based HAR using joint-mapping strategies, inspiring FITORBISs anglebased exercise validation.

III. METHODOLOGY

A. System Overview

FITORBIS operates in three stages:

- 1) **Pose Estimation**: MediaPipe extracts 33 body landmarks (Fig. 1).
- 2) **Feature Extraction**: Joint angles (e.g., elbow/knee) are computed using geometric relationships.

Action Classification: Threshold-based validation confirms exercise completion.

Fig. 1. MediaPipes 33 skeletal landmarks [1].

B. Algorithm Design

blueclass poseDetector():

bluedef __init__(self):

1) Pose Detection: The poseDetector class (below) uses MediaPipe to extract joint coordinates:

bluereturn math.degrees(arctan2(y3 - y2, x3 - x2

2) Exercise Validation: A sit-up is validated if:

- Hip-knee angle exceeds 90 during contraction.
- Shoulder-hip distance decreases by 30% from the initial pose.

C. Hardware Integration

- Raspberry Pi 4: Runs pose estimation at 10 FPS using OpenCVs GPU-accelerated pipelines.
- Camera Module: Captures 720p video at 30 FPS.

IV. RESULTS

A. Performance Metrics

Value
10
95%
;1s

PERFORMANCE METRICS OF FITORBIS.

B. Comparison with Existing Work

Method	FPS	Hardware	Accuracy
FITORBIS (Ours)	10	Raspberry Pi	95%
Kim et al. [1]	30	Intel NUC	97%
Kang et al. [4]	15	GPU	96%
TABLE II			

COMPARISON WITH EXISTING METHODS.

V. DISCUSSION

A. Strengths

- **Cost Efficiency**: Achieves real-time performance on sub-\$100 hardware.
- Adaptability: Works under varying lighting/background conditions.

B. Limitations

- 2D Pose Ambiguity: Lateral movements may reduce angle accuracy.
- Hardware Constraints: Limited to single-user scenarios.

VI. CONCLUSION

FITORBIS demonstrates the viability of integrating fitness with AI-driven action recognition on low-cost hardware. Future work will explore 3D pose estimation using uDEAS [1] for complex exercises.

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