

M. Manimegala^{1*}, C. Aswin², M. Balamurugan³, P. K. Charumathi⁴, M. Hrithik⁵, M. S. Kamalesh⁶

¹Assistant Professor, Department of Computer Science and Engineering, Sri Shakthi Institute of Engineering and Technology, Coimbatore, India ^{2,3,4,5,6}Student, Department of Computer Science and Engineering, Sri Shakthi Institute of Engineering and Technology, Coimbatore, India

Abstract: The increase in road accidents caused by driver fatigue has necessitated the development of real-time driver drowsiness detection systems. This paper presents a hybrid solution for detecting driver drowsiness using facial landmark analysis and alternative vehicle behavior monitoring techniques. The system utilizes OpenCV and MediaPipe for real-time video processing to analyze Eye Aspect Ratio (EAR) and Mouth Aspect Ratio (MAR) for detecting prolonged eye closure and yawning, respectively. In scenarios where facial detection fails due to poor lighting or obstructions, the system monitors sudden vehicle speed changes and harsh braking as alternative indicators of drowsiness. The proposed solution triggers timely audio-visual alerts to prevent accidents and logs driver behavior for future analysis. This paper discusses the implementation, testing, and performance analysis of the system, highlighting its accuracy, reliability, and real-world applicability.

Keywords: Driver Drowsiness, Eye Aspect Ratio (EAR), Mouth Aspect Ratio (MAR), OpenCV, MediaPipe, Speed Monitoring, Real-time Alert System.

1. Introduction

Driver fatigue is one of the leading causes of road accidents worldwide, posing a serious risk to life and property. Traditional approaches to fatigue detection rely on subjective measures such as self-assessment, which are impractical for real-time applications. With advancements in computer vision and sensor Technology, automated drowsiness detection systems have emerged as effective tools for improving road safety. This paper proposes a Driver Drowsiness Detection System that combines facial landmark analysis with vehicle behavior monitoring. Facial detection is performed using OpenCV and MediaPipe to monitor eye closure duration and yawning through EAR and MAR calculations. An alternative detection mechanism involving sudden speed changes and abrupt braking ensures system reliability when facial data is unavailable. The integration of real-time alerts and behavior logging enhances driver safety, making the solution practical and scalable.

2. Literature Review

The literature on driver drowsiness detection highlights various approaches, with a predominant focus on facial feature recognition and physiological monitoring. Facial feature-based systems, such as those using Eye Aspect Ratio (EAR) and Mouth Aspect Ratio (MAR), have proven effective in detecting drowsiness by monitoring eye closure and yawning, as proposed by Ji et al. (2004). However, environmental factors like lighting and head movements can reduce the accuracy of such systems. Physiological monitoring, incorporating wearable devices such as smartwatches or EEG headsets, has also been explored to provide additional data for drowsiness detection, as seen in Abtahi et al. (2015). Despite the potential of these methods, wearable devices are often considered intrusive and expensive. These limitations have led to the development of hybrid systems that combine facial recognition with vehicle behavior monitor.

3. Methodology

A. Existing System

Existing drowsiness detection systems primarily rely on facial feature analysis or physiological monitoring. Some key drawbacks include poor accuracy under low-light conditions or occluded face positions, dependency on expensive wearable devices for physiological parameters, lack of alternative methods for detecting drowsiness if facial detection fails, high computational. costs for real-time deep learning systems, and limited integration with vehicle behavior analysis for fallback detection. These systems may also fail to provide timely warnings if the environmental conditions do not support effective facial detection, leading to an increased risk of accidents. The need for a more robust, multimodal solution is crucial for improving the safety and reliability of drowsiness detection.

B. Proposed System

The proposed system addresses the limitations of existing approaches by integrating facial landmark detection with vehicle behavior monitoring. Key features of the system include real-time video processing using OpenCV and MediaPipe frameworks, which allow the detection of key facial landmarks. Eye Aspect Ratio (EAR) and Mouth Aspect Ratio (MAR) are calculated to monitor prolonged eye closure and yawning, which are key indicators of drowsiness. Additionally, the system incorporates an alternative detection mechanism using sudden changes in vehicle speed and harsh braking when facial detection fails due to poor lighting or face occlusion. If drowsiness is detected, the system triggers an audio-visual alert to immediately warn the driver. The system also logs drowsiness events for future analysis and performance

^{*}Corresponding author: charupalani6@gmail.com

evaluation, ensuring that the system's effectiveness is constantly reviewed.

4. Software Description

The Driver Drowsiness Detection System is implemented using Python 3.9, leveraging key libraries like OpenCV for realtime video processing and facial landmark detection, and MediaPipe for accurate tracking of facial features to compute Eye Aspect Ratio (EAR) and Mouth Aspect Ratio (MAR) for detecting drowsiness indicators such as eye closure and yawning. The system also incorporates MySQL for logging drowsiness events, capturing data on the time, cause of detection, and alternative triggers such as vehicle speed or harsh braking. Flask is used to manage the alert system, providing real- time notifications through audio and visual cues when fatigue is detected. Additionally, the system monitors vehicle behavior by integrating sensor data to detect sudden speed changes or braking when facial detection is unreliable, ensuring fallback mechanisms for consistent operation. This combination of computer vision and vehicle behavior analysis makes the system robust and effective for preventing accidents caused by driver fatigue.

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6. System Specification

- A. Hardware Requirements
 - *Processor*: Intel Core i3 or higher
 - *RAM*: 4 GB
 - Hard Disk: 500 GB
 - Monitor: 17" Color Monitor
 - Webcam: 720p HD
- B. Software Requirements
 - Programming Language: Python 3.9
 - Libraries: OpenCV, MediaPipe, MySQL, Flask
 - Operating System: Windows 8.1 or higher



8. Result







Fig. 4. Eye detection



Fig. 5. Yawn detection



Fig. 6. Speedometer and brake

9. Future Enhancement

Future improvements to the system include integration with IOT devices to enable remote monitoring of driver behaviour in real time, particularly for fleet management application. The system could also benefit from advance deep learning models to improve detection accuracy and adaptability under challenging condition, such as varied head positions or extreme lighting. Wearable devices like smart watches could be integrated to provide additional psychological data, such as heart rate, which can further enhance drowsiness detection. A multi camera system could be implemented to improve detection from different angles, ensuring accurate facial recognition. Furthermore, the system could be integrated with vehicle control systems to automatically take preventive actions, such as reducing the vehicle speed or adjusting the seat position when drowsiness is detected.

10. Conclusion

The Driver Drowsiness Detection System develop in this paper effectively combines facial landmark analysis and vehicle behaviour monitoring to provide a reliable, real-time solution for detecting drowsy drivers. The Integration of EAR and MAR calculation with fallback mechanisms based on vehicle speed and braking ensures that the system functions reliably under various conditions. By triggering timely alerts, the system aims to reduce the risk of accident caused by driver fatigue. The system ability to log events for analysis further enhances its value for future improvements. With continued advancements, such as the integration of deep learning models and wearable devices, the system can evolve into an even more accurate and comprehensive tool for ensuring road safety.

References

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