



Archives available at [journals.mriindia.com](http://journals.mriindia.com)

## International Journal of Recent Advances in Engineering and Technology

ISSN: 2347-2812

Volume 14 Issue 01, 2025

### Driver Drowsiness Detection using ML and IOT

Madhuri Borawake<sup>1</sup>, Abhishek Patil<sup>2</sup>, Shreyash Yadav<sup>3</sup>, Vishwajeet Nagwade<sup>4</sup>, Hemant Somwanshi<sup>5</sup>

<sup>1</sup> Professor PDEA's College of Engineering, Pune.

<sup>2,3,4,5</sup> Student PDEA's College of Engineering, Pune.

pdeacoemmch@gmail.com<sup>1</sup>, abhishekpatil172002@gmail.com<sup>2</sup>, shreyashsy9990@gmail.com<sup>3</sup>, Vishwajeetnagwade6111@gmail.com<sup>4</sup>, kakapatil009@gmail.com<sup>5</sup>

Department of Computer Engineering,

Pune District Education Association's College of Engineering,

Manjari Bk., Hadapsar, Pune, Maharashtra, India – 412307

Email: coem@pdeapune.org

#### Peer Review Information

*Submission: 16 Jan 2025*

*Revision: 17 Feb 2025*

*Acceptance: 11 March 2025*

#### Keywords

*Drowsy Driving*

*Traffic Accidents*

*Facial Features*

*Mobile Device Alerts*

#### Abstract

According to new statistics across the globe, drowsy driving accidents become one of the very significant problems related to safety. This proposed Driver Sleep Detection System (DDS) takes a proactive approach with the help of ML and IoT technologies to track the behavior of the driver through features in facial expressions, head movements, and eye activity monitored by camera through the vehicle. Computer vision methods, such as Haar Cascades or CNNs, determine eye closure and yawning as drowsiness symptoms. The ML model enhances precision even more by training on diverse facial data while constantly adapting to varying driving environments and individual drivers. The key performance metrics are accuracy and response time, for the assessment of the system's performance. IoT integration allows for real-time alerts over an onboard buzzer or vibration system and can even forward data to a monitoring system or mobile device for remote viewing. Data is also stored in the cloud for long-term analysis.

JEL Classification Number: O33, R41

#### INTRODUCTION

Car accidents are ranked the leading cause of deaths with around 1.3 million lives lost yearly, most being due to driver distraction or drowsiness. Drowsiness significantly reduces a driver's focus, activity, and alertness which in some cases leads to slow or poor decision-making, whereas in other cases, the decision would just seem non-existent. This decreased level of alertness increases the risk of human error which goes on to

lead to injuries or even fatalities. Moreover, most truckers spend most of their time crossing many miles on the road, day and night, many not having enough sleep or being distracted by time spent talking on the phone, among other activities that heighten the risks of accidents. It is a monitoring system where it is proposed to monitor the driver and provide warnings in case he is distracted or drowsy. For this purpose, techniques in image processing will be used with facial images

captured by the camera showing signs of drowsiness or distraction. Open-source libraries like OpenCV and Dlib are chosen for the implementation of the functionality of image processing, whereas Python would serve as the programming language for the project. An infrared camera captures the driver's facial landmarks and eye movements continuously, with a focus on the eyes. The system supervises eye features to detect drowsiness, and in case the eyes are closed for some time, it renders the driver drowsy. In such a scenario, an audible alarm is sounded, informing the driver. This innovative approach reduces accidents caused by the lack of driver attention by giving real-time warnings for safer driving.

### **Financial Issues**

#### **1] Development Cost:**

Hardware Expenses: IoT devices, sensors (e.g., cameras, infrared sensors, EEG headbands), microcontrollers, and cloud infrastructure.

Software Development: Costs for training the AI model, integration, and app/firmware development.

Data Collection & Processing: Cost of gathering and labeling datasets (e.g., facial recognition or eye-tracking data).

#### **2] Operational Costs:**

Cloud Services & Data Storage: Maintaining real-time data processing and cloud-based AI models.

Device Maintenance & Updates: Upgrading firmware, replacing faulty sensors, and optimizing algorithms.

#### **3] Regulatory & Compliance Costs:**

Safety Certifications: Automotive and IoT security standards compliance, such as ISO 26262 and GDPR for data privacy.

Liability & Insurance: Risk of lawsuits in case of false negatives that may lead to accidents.

#### **4] Market & Business Costs:**

Prototyping & Testing: Developing working prototypes and testing in real-world conditions.

Production & Scaling: IoT hardware mass production and ML deployment on edge devices.

Marketing & Distribution: cost associated with branding, marketing, and installing in vehicles or industries.

### **LITERATURE SURVEY**

Verma et al. (2023) evaluate Driver Drowsiness Detection. This study makes use of image processing tools including OpenCV, Dlib, and the EAR method to detect drowsiness in real-time through live webcam feeds. The system sends alerts through emails or SMSes, thus this is a very cost-effective and efficient monitoring method. The system's major strength is its open-source tool base, which reduces the cost factor and increases access. However, it works poorly in poor lighting conditions, and its correctness depends

on a clear view from the camera, which is quite difficult to maintain in real-life driving.

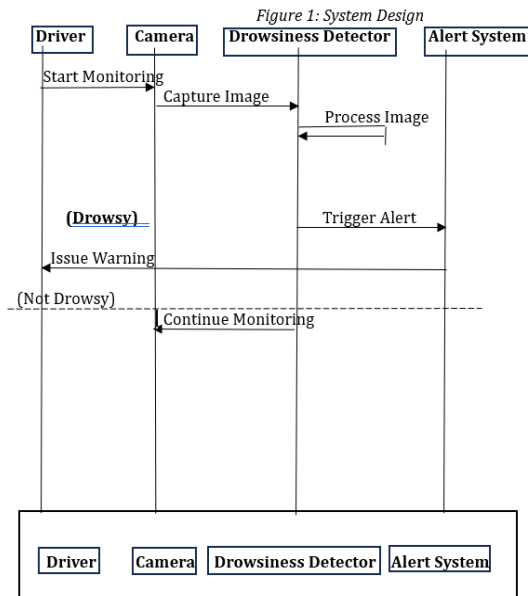
Hussain et al. (2022) evaluate Driver's Drowsiness Detection. This project uses OpenCV for capturing images and a CNN model for the classification of eye status (open or closed). It uses thresholding to identify drowsiness and trigger the alarm. It is non-intrusive, low-cost, and operates in real time, hence it is very practical for safety issues of the driver. The drawback is that it only monitors eye-tracking, which may not be a very accurate indicator of drowsiness at all times. Additionally, in poor lighting conditions, its reliability is decreased with a chance of wrong detection. Titare et al. (2021), evaluate Driver Drowsiness Detection and Alert System. This study applied image processing techniques like OpenCV, Dlib, and EAR to determine drowsiness in real time from live webcam feeds. The system will prompt alerts via email or SMS; hence, it is relatively cost-effective and an efficient monitoring option. The major advantage of this approach is based on the open-source tools, which contribute to the affordability and accessibility of the tool. Its effectiveness is limited in low-light conditions, and accuracy depends on an unobstructed camera view, which is not often possible in real-world driving scenarios.

Jain et al. (2021), it proposes Real-Time Driver Drowsiness Detection using Computer Vision. This research unites facial characteristic tracking with EAR and Mouth Aspect Ratio (MAR) to detect drowsiness through computer vision. The system is highly accurate for bright illumination conditions and it is simple to deploy at low cost. However, its performance is hindered by obstacles such as hats or facial occlusions, and it struggles for low light conditions, which makes it less reliable for real-world driving environments where lighting conditions vary.

Baharu et al. (2013), Driver Drowsiness Detection by Using Webcam This system relies on MATLAB and the Viola-Jones algorithm to track eye and mouth movements for drowsiness detection. It can be a low-cost solution fit for real-time in-vehicle monitoring but is otherwise impaired in effectiveness in shadowed or poorly lit environments. In addition, it has difficulty in determining if a driver is drowsy if he is wearing eyeglasses, making it less adaptable to diverse user conditions.

Ibrahim et al. (2023) The study "DDD TinyML" develops a lightweight driver drowsiness detection system using TinyML for IoT devices. It evaluates models like SqueezeNet, AlexNet, CNN, MobileNet-V2, and V3, optimized via quantization. MobileNet-V2 achieved 99.64% accuracy with DRQ, while CNN had the smallest size (0.05 MB). The models were tailored for microcontroller

deployment, ensuring real-time performance with resource efficiency.



## THE MODEL

As shown in Figure1: System Design a Drowsiness Detection System, whereby a camera captures images of the driver. Images from the camera are processed in a drowsiness detector, deciding whether or not the driver has drowsiness. Should drowsiness be detected, an alerting system prompts warning; else monitoring continues.

### 1. Behavioral and Physiological Detection Model:

**Eye Aspect Ratio (EAR):** A mathematical model for measuring the ratio of opening and closing eyes. Constantly low EAR is indicative of drowsiness.

**Mouth Aspect Ratio (MAR):** Sudden frequency of yawning to further prove drowsiness.

**Head Pose Estimation:** Rapid tilting of the head or nodding proves the occurrence of micro-sleeps.

### 2. IoT-Based Real-Time Communication:

**Data Transmission:** The system transfers driver condition data to cloud servers for storage and analytics.

**Remote Monitoring:** SMS, emails, or mobile application alerts are transmitted to fleet managers or family members.

**Vehicle Control:** The system can control speed or send out emergency alerts to the vehicle at times.

### 3. System Decision Logic and Automation:

**First Warning:** When minor drowsiness is detected, an audio and vibration alert is activated.

**Escalation:** If drowsiness continues, the system activates a buzzer and starts to slow down the vehicle speed.

**Intoxication Detection:** If BAC levels are beyond the threshold, the system refuses to ignite the engine or alert the authorities.

### 4. Model Selection and Training of Machine Learning:

For enhanced accuracy, the system uses a CNN that is trained on a diverse dataset of facial expressions and alcohol impairment patterns. The model is fine-tuned to reduce false positives and improve detection efficiency under different lighting and environmental conditions.

### 5. System Optimization and Deployment:

The system is tested in several rounds under simulated and real-world conditions. Threshold tuning is performed to enhance performance while minimizing false alarms. The final model is deployed in a test vehicle, where real-time performance is evaluated and refined for enhanced safety.

## SUMMARY AND CONCLUSIONS

Machine learning (ML) and IoT-based driver drowsiness detection system can increase road safety by detecting early signs of drowsiness to provide timely warnings. These intelligent algorithms and data-enabled real-time systems are specifically advantageous in critical scenarios such as long-haul trucking, public transport, and ride-hailing services. The benefits of these systems include real-time feedback, scalability, and precision, while problems include environmental factors, privacy concerns, and hardware limitations that are not fading away. Advancements in sensors, ML models, and privacy solutions can address these issues, thereby opening the door to wider applications. As a technological development, this will become an interesting step toward handling driver, passenger, and all road users' safety.

## References

Harshit Verma A., "Driver Drowsiness Detection," Proceedings of the Journal of Data Acquisition and Processing, April 2023.

A.Milan, Bhanu Sankar Ravi "Driver Drowsiness Detection" Proceedings of the Hindustan Institute of Technology and Science Chennai, 2022.

Mahek Jain, Bhavya Bhagerathi "Real-Time Driver Drowsiness Detection using Computer Vision" Proceedings of the International Journal of Engineering and Advanced Technology (IJEAT), 2021.

Swapnil Titare, A., "Driver Drowsiness Detection and Alert System" Proceedings of the

International Journal of Scientific Research in Computer Science, Engineering and Information Technology,2021.

Badiuzaman Bin Baharu “Driver Drowsiness Detection by Using Webcam” Proceedings of the Universiti Teknologi PETRONAS Bandar Seri Iskandar 31750 Tronoh Perak Darul Ridzuan,2013.

Sushma Sharad Tandle, Akshita S, Amtul Malik, Joncia Fernandes, Shilpa NS “Driver Drowsiness and Alcohol Detection System Using Arduino” proceedings of International Journal of Advances in Engineering and Management, 2023.

Anil Kumar Biswal, Debabrata Singh, Binod Kumar Pattanayak, Debabrata Samanta ,and Ming-Hour Yang “IoT-Based Smart Alert System for Drowsy Driver Detection” proceedings of Hindawi Wireless Communications and Mobile ComputingVolume,2021.

Mrs.R. Sathya, T. Surya Reddy, S. Akshith Reddy, K. Raghavendra “An IoT Based Driver Drowsiness Detection System and Deterrent System for Safety and Driving” proceedings of International Journal of Future Generation Communication and Networking,2020.

Tarek M. Hassan, Walaa Yahia Ibrahim, Hesham A. Ali, Hanaa Salem Marie “Deep Learning-Based Driver Drowsiness Detection Using Facial Expression Analysis” proceedings of Delta University Scientific Journal Vol.07,2024.

Sarah Saadoon Jasim, Alia Karim Abdul Hassan “Modern drowsiness detection techniques: a review” proceedings of International Journal of Electrical and Computer Engineering, 2021.