



# Enhanced Driver Vigilance System

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**Abstract:** The concerns rise in driver fatigue-related vehicle collisions has made drowsiness detection in drivers a significant area of study. Experts say that drivers who drive long distance without taking regular rests are at a high risk of experiencing fatigue. Research shows that exhausted drivers in need of rest account for around 25% of all major highway collisions. The purpose of our systems is to spot early indicators of driver exhaustion before they impact one's ability to drive. This system is a novel approach utilizing deep learning techniques, specifically 2D convolutional neural networks (CNNs), to identify signs of drowsiness in drivers face by analysing facial and eye features. The idea is aimed to use traditional models of multi-layer 2D CNN with multi-label classification and Haar-cascade algorithm. Multiple face signs like eye closures and yawning are considered through the input images to improve the detection accuracy under various driving conditions.

**Keywords:** Drowsiness detection, Facial recognition, eye detection, yawn detection, multi-label classification, Image-based analysis, Deep learning.

## I. INTRODUCTION

The growing number of automobiles on the road has led to an increase in traffic accidents, which are now the main cause of death in many countries. In the rapidly changing landscape of the automotive industry, the integration of technology-driven features, particularly advanced driver assistance systems (ADAS), has become increasingly common. Drowsiness among drivers serves as a serious risk on road and is a major factor in many traffic accidents. Addressing driver drowsiness is critical because it compromises road safety. This project focuses on the creation of a non-intrusive detection system for detecting driver tiredness from frontal face image inputs using computer vision and deep learning methods.

The novelty of this approach lies in its ability to analyse fatigue without intruding on the driver's personal space. By emphasizing the global concern of rising car accident fatalities, this research underscores the urgency of preventing driver fatigue to enhance overall road safety.

The current situation of escalating number of vehicles on the roads and the resulting surge in accidents, emphasizes the responsibility of drivers to ensure the safety of themselves and their passengers. This research not only underscores the global concern surrounding the escalating fatalities from car accidents but also highlights the need for proactive measures to prevent driver fatigue. By utilizing computer vision and deep learning in real-time driver monitoring, the study seeks to contribute to the ongoing efforts in road safety. The approach presented in this research serves as a testament to the significance of embracing technological solutions to address critical issues such as driver drowsiness, ultimately promoting a safer and more secure driving environment.

## II. LITERATURE SURVEY

[1] The paper introduces a novel approach to identify driver drowsiness using a deep 3D convolutional neural network (CNN) and a state probability vector. It involves facial detection, 3D CNN classification of facial image sequences, and concatenation of output probabilities for recognition. Despite improved performance over 2D CNN, challenges arise with large head rotations and occlusions.

[2] The paper introduces an advanced driving assistance system (ADAS) for detecting driver drowsiness, employing sequences of facial images to minimize false positives. Two solutions are proposed: a recurrent and convolutional neural network (R-CNN) and a fuzzy logic-based system. While both achieve comparable accuracy, the fuzzy logic-based system exhibits notably high specificity in reducing false alarms.



[3] The paper addresses driver fatigue and drowsiness with a Python-based "Driver Drowsiness Detector" prototype. It utilizes Eye Aspect Ratio (EAR) and Yawn Detection algorithms to analyse eye and mouth regions, enhanced by a Convolutional Neural Network (CNN) for improved accuracy. Leveraging Python's text processing capabilities and simplicity, along with the DLib library and OpenCV's Haar Cascade Classifiers, the prototype employs threading, networking, and numerical techniques to detect facial features, ensuring comprehensive drowsiness detection.

[4] The paper presents an intelligent framework for addressing driver drowsiness using computer vision technology. It employs the Viola-Jones face detection algorithm to train a stacked deep convolutional neural network (CNN) on dynamically retrieved key frames. Specifically, the CNN's SoftMax layer categorizes drivers based on their eye state, distinguishing between drowsy and alert states while driving. This approach surpasses conventional CNNs and mitigates issues like pose accuracy in regression. Additionally, an alarm system alerts drowsy drivers, contributing to road safety without overwhelming data acquisition.

[5] The paper presents an intelligent alerting system designed to mitigate accidents due to driver drowsiness. It integrates live video streaming of drivers with drowsiness detection techniques, utilizing parameters like Eye Aspect Ratio (EAR) and Euclidean distance of the eye. Experimental validation confirms its efficacy in reducing fatigue-related accidents and injuries. The study explores diverse detection methods including machine learning, deep learning, and vehicle data analysis. Employing an IoT processor like the Raspberry Pi3 CPU Module, the method processes recorded videos and images, enabling accurate assessment through face and eye behaviour analysis.

### III. SCOPE AND METHODOLOGY

#### Scope

The project's goal is to make driving safer by employing technology to identify indicators of tiredness in drivers. This project's scope includes the research, development, and implementation of face detection utilizing deep learning approaches across numerous datasets. The technology will be developed to capture and evaluate the user's facial expressions and eye closure patterns. The project's goal is to improve road safety, minimize accidents, and address the global concern of rising fatalities from car accidents.

#### Methodology

The process entails developing a deep learning model to categorize photos of eyes as open or closed, and yawning or not yawning. This is accomplished through pre-processing the photos, which includes face detection and cropping, shrinking to a predetermined size, and normalizing pixel values. Trained eye dataset and yawn dataset independently and pre-processing is achieved like cropping, resizing and normalising pixels. The both datasets are trained on Sequential CNN architecture. The process entails developing a deep learning model to categorize photos of eyes as open or closed, and yawning or not yawning. This is accomplished through pre-processing the photos, which includes face detection and cropping, shrinking to a predetermined size, and normalizing pixel values. The dataset is enhanced using a variety of techniques, including zooming and flipping. A Convolutional Neural Network (CNN) architecture is used, which consists of convolutional and max-pooling layers followed by dense layers with dropout regularization to prevent overfitting. The model is trained for 50 epochs with the Adam optimizer and categorical cross-entropy loss function, followed by performance validation on a test set. After training, the model is evaluated with previously unseen data. Unseen data is previously stored separately to ensure the model's correctness on unseen data.

The ultimate goal is to include this model into a driver drowsiness detection system that uses real-time video footage to inform drivers when tiredness is detected, potentially lowering accidents. The prediction label in real-time camera footage indicates if the driver is "Drowsy" or "Not Drowsy" followed by audio warning to the user. Model performance is evaluated using graphs exhibiting measures such as Accuracy, Precision, and Recall. Real-time camera footage is critical for identifying tiredness and giving voice warnings to the user, hence improving driving safety.

### IV. SYSTEM ARCHITECTURE

The facial detection system described in the architectural explanation is designed to increase driver safety. Beginning with data taken by a camera, the system uses a face detection module that uses a Haar cascade classifier to identify faces in the image being captured. After detecting a face, the system proceeds to an eye detection module, which targets the eye region for additional analysis. However, it remains uncertain whether the system undertakes any classification specifically on the eye region.



Finally, based on detected face features, an alarm module is probably activated to inform the driver or take further actions. By integrating image processing units for face identification and perhaps spotting signs of driver tiredness or distraction, this architectural concept hopes to promote increased road safety. However, clarification of the categorization procedure and the specific actions initiated by the alert module will enhance the system's functionality's accuracy and efficiency, assuring a more significant strategy for tackling driver safety issues.

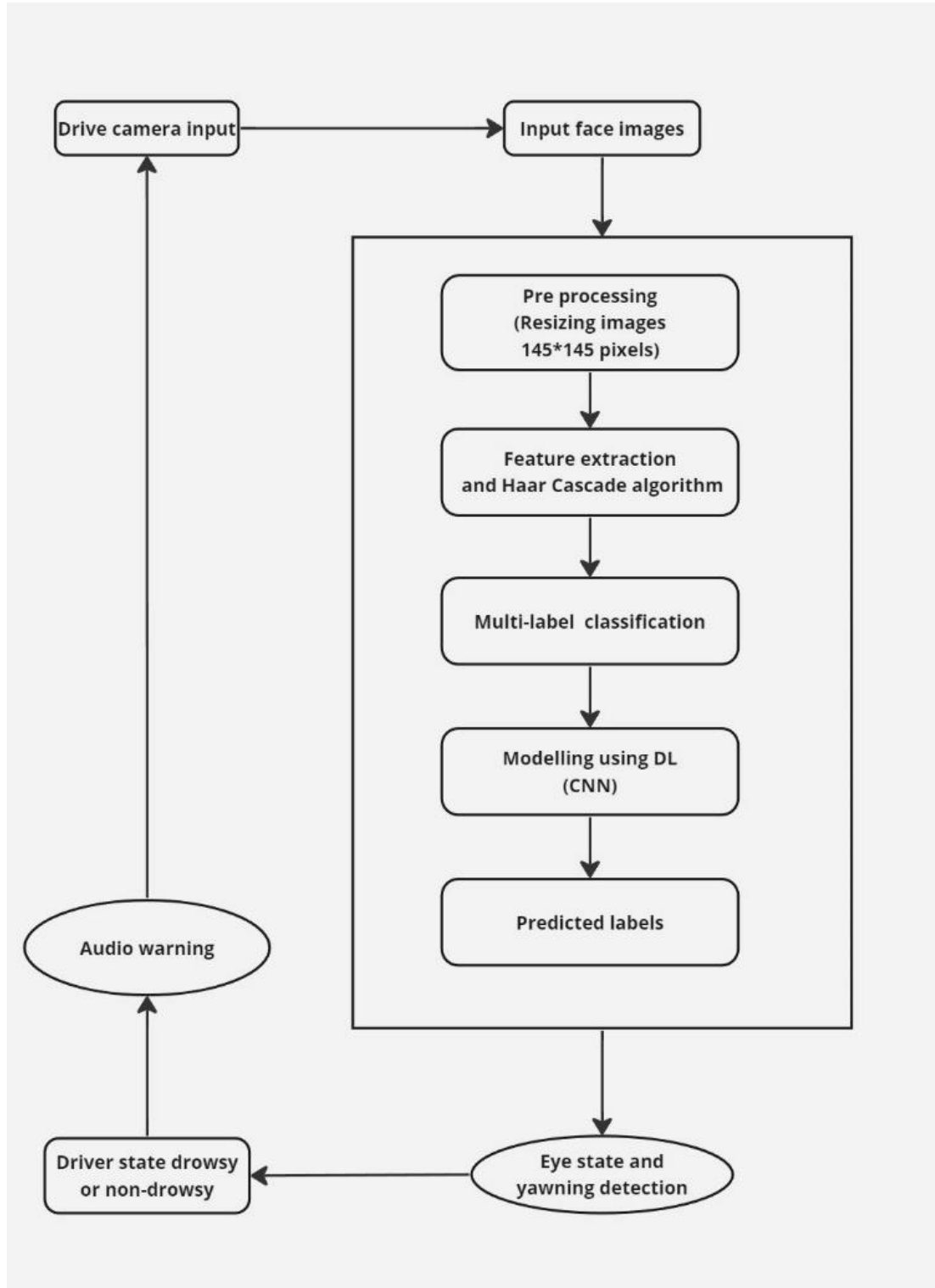


Fig. 1 System architecture

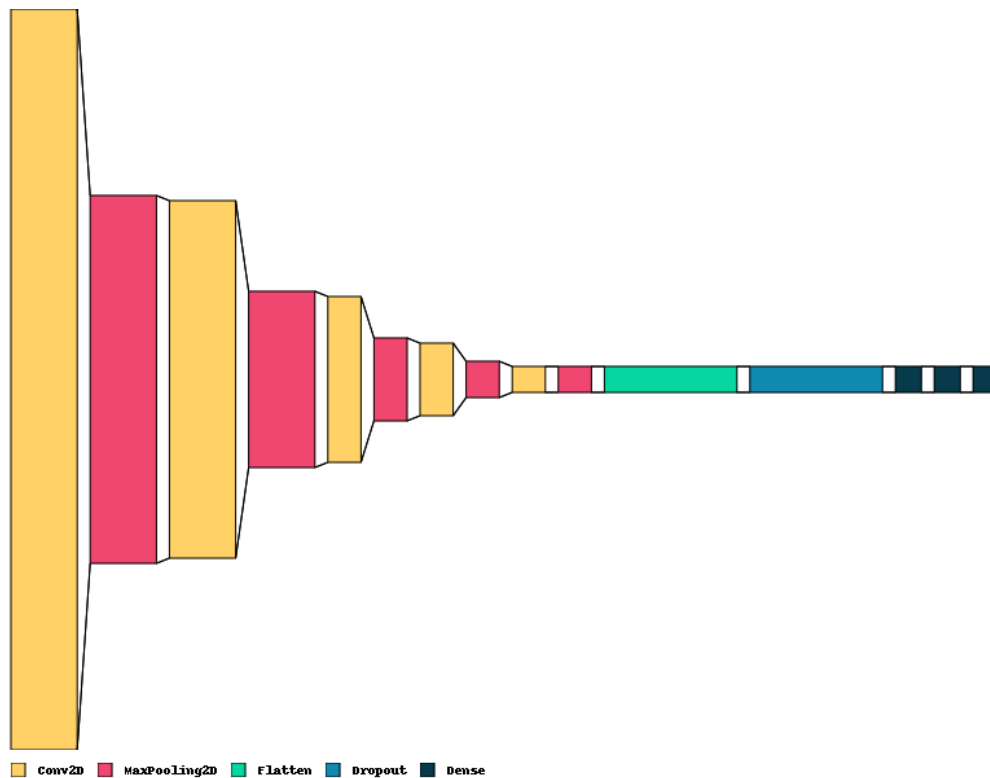


Fig. 2 CNN architecture

## V. CONCLUSION

The driver drowsiness detection system shows potential as a useful tool for monitoring drivers and detecting tiredness in real time. Its accuracy and effectiveness were assessed on two datasets and determined to be high. Integration into a comprehensive system increases its ability to improve driving safety. This technique, together with existing driver monitoring systems, has the potential to drastically reduce sleepy driving accidents. Future study should concentrate on technical factors that improve accuracy, generalization, real-time functioning, privacy protection, and data efficiency. These developments will result in more effective and practical strategies for reducing the risks of drowsy driving and improving road safety.

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