

Using CNN for Detection of Driver Drowsiness¹

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ABSTRACT

This study presents a Convolutional Neural Network (CNN) based Driver Drowsiness Detection system. The device determines a driver's level of attention by analysing face features captured in real-time by in-car cameras. The CNN model is appropriate for use in the real world because it has been trained on various datasets and exhibits good accuracy in recognising indicators of tiredness. This device helps to keep drivers safe on the road by sending out timely alerts to stop tired drivers from causing accidents. Building intelligent technologies to reduce driver-related dangers is essential because road safety is still a significant concern. One of the leading causes of traffic accidents is driver fatigue, which highlights the necessity for reliable and quick detection systems. This study presents a novel method for driver drowsiness detection with convolutional neural networks (CNN).

The suggested method uses CNN architecture to analyze facial features taken from in-car cameras' real-time video streams. The driver's degree of awareness is deduced by analysing their facial expressions and landmarks. The CNN model is trained on various datasets, including awake and sleepy facial expressions, guaranteeing its flexibility to various driving scenarios and personal traits. Extensive tests use different datasets and scenarios to assess the system's efficiency. The outcomes show how well the CNN can detect indicators of driver fatigue, with high recall and precision rates. Moreover, due to its real-time processing capabilities, the model can be deployed in realistic on-road settings.

By promptly alerting drivers when indicators of Drowsiness are identified, the suggested Driver sleepiness Detection system offers a possible approach to improve road safety. By incorporating artificial intelligence into car safety systems; this research helps to lower the likelihood of accidents brought on by tired drivers.

INTRODUCTION

Intelligent system integration is critical as technological developments change the car industry and improve road safety. Creating Driver Drowsiness Detection systems, which attempt to reduce the risks of driving when tired, is a crucial component of this quest. Across the globe, driver fatigue continues to be a significant cause of traffic accidents, emphasising the need for prompt and efficient detection systems. The difficulty of driver drowsiness detection is the main topic of this research, which focuses on utilising the power of convolutional neural networks (CNNs). Conventional approaches frequently depend on elementary thresholding strategies or rudimentary image processing, which might need to be more robust for real-world situations.

In contrast, CNNs present a promising way to reliably determine a driver's attention level based on facial expressions because of their capacity to learn hierarchical characteristics from data autonomously.

The widespread use of in-car cameras makes it possible to continually and instantly observe drivers. This research will use CNNs to analyze facial features collected from these video streams to develop a system that can discriminate between alert and drowsy states. The practical application of the suggested approach depends on its capacity to adjust to various driving situations and individual characteristics. The CNN-based Driver Drowsiness Detection system's approach is covered in detail in the following parts. We review the model architecture, the evaluation metrics that were utilised, and the dataset used for training and validation. The results of this study hold

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great promise for improving road safety by promptly alerting sleepy drivers, in addition to being a valuable contribution to computer vision and artificial intelligence.

RELATED WORK

1) Bappaditya Mandal- Creating a reliable method for visual analysis of eye states to identify bus driver fatigue is the main emphasis of this paper's research. The suggested method uses sophisticated computer vision algorithms to precisely analyze the driver's eye movements and facial expressions to identify sleepiness indicators. A large-scale dataset assesses the system's efficacy, showing encouraging outcomes. The study's findings have the potential to significantly improve passenger safety and reduce the dangers posed by tired bus drivers.

2) Zuojin Li: The research presented in this paper develops an autonomous method for detecting driver weariness, addressing the critical requirement for transportation safety. The suggested method uses machine learning techniques to analyze driving operation data, including pedal usage and steering patterns, to detect indicators of driver fatigue. The system's effectiveness is assessed through comprehensive testing, demonstrating its ability to improve road safety by timely notifications and reducing the dangers of intoxicated driving.

3) Mr. Phil Hanley- This study examines the everyday problems of stress and exhaustion among bus drivers. The goal of the research is to find viable interventions and solutions by thoroughly investigating the elements contributing to driver fatigue and stress. The integration of operational data, physiological measurements, and survey results offers a comprehensive comprehension of the difficulties encountered by bus drivers. The study's findings provide important information that may be used to raise transportation safety, improve working conditions, and reduce the hazards related to driver stress and weariness.

4) Thobias Sando: The possible causes of driver weariness among Florida's transit bus operators are examined in this work. The study determines the leading causes of weariness through questionnaires, interviews, and data analysis. The findings aim to improve knowledge and provide insights for implementing focused interventions to address driver fatigue issues, which will ultimately enhance transit operations safety.

5) Dayang Nailul Munna- This study examines the various aspects that affect bus driver weariness and how it relates to the likelihood of accidents in this paper. Using surveys, driver interviews, and incident data analysis, the research attempts to identify the leading causes of fatigue and how they affect road safety. The results provide essential information for formulating plans to reduce fatigue-related hazards and improve the general safety of bus transit.

6) The research qualitatively examines the elements influencing urban bus drivers' weariness, according to author Herbert Biggsa. This study uses theme analysis and in-depth interviews to identify the primary factors contributing to driver weariness in urban transportation environments. The results offer significant perspectives for formulating focused interventions and tactics to enhance the welfare and security of city bus operators.

7) Bindu Bhatt- This report examines the health risks associated with driving buses for a living. The research finds and looks at the common health concerns connected to bus going through questionnaires, health evaluations, and work environment analysis. The results offer significant perspectives for formulating focused interventions and regulations to enhance bus drivers' occupational health and welfare, hence augmenting workplace safety within the transportation industry.

FLOWCHART

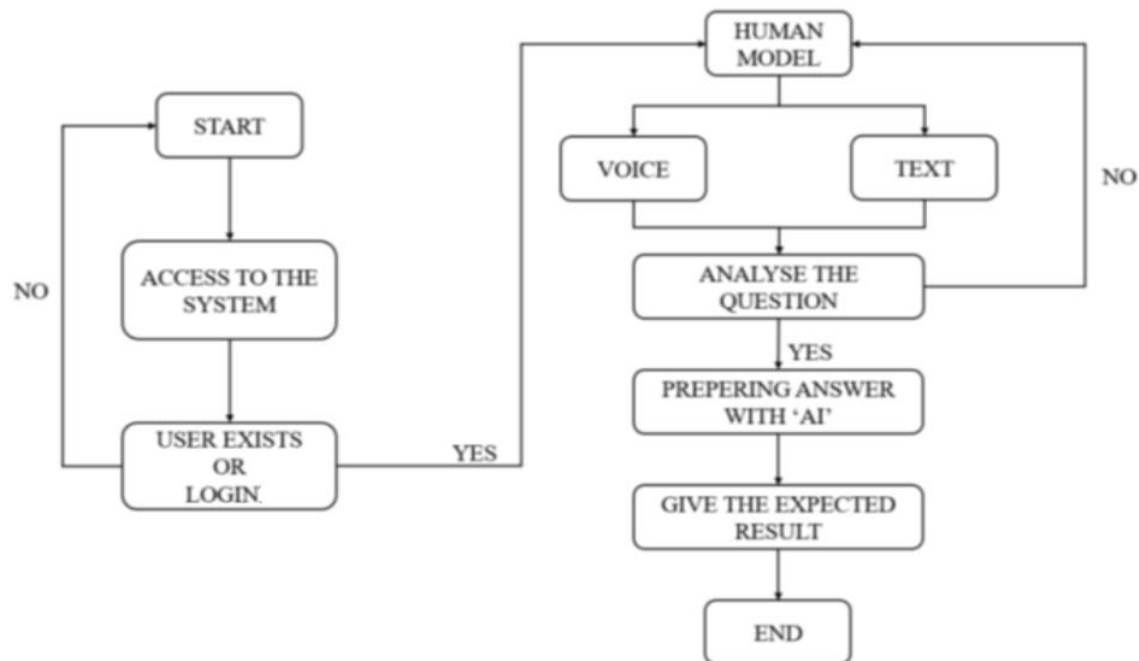


Fig 1.1 Flowchart of Driver drowsiness

Driver drowsiness detection systems typically utilize various sensors and algorithms to monitor a driver's behaviour and physiological signals, aiming to identify signs of fatigue or drowsiness. The working of these systems often involves the following components:

- 1) *Image/Video Capture*: In-vehicle cameras capture real-time images or video footage of the driver's face.
- 2) *Facial Feature Extraction*: Computer vision techniques are applied to extract facial features from the captured images or video frames. This may include eye movement, blinking patterns, head pose, and facial expressions.
- 3) *Data Pre-processing*: Extracted facial features are pre-processed to enhance the quality of data and remove any noise that might affect the accuracy of detection.
- 4) *Feature Analysis*: Extracted features are analyzed to identify patterns associated with drowsiness. For example, frequent blinking, slow eye movements, or changes in facial expressions may indicate fatigue.
- 5) *Machine Learning/Deep Learning Models*: Many systems employ machine learning algorithms or deep learning models, such as Convolutional Neural Networks (CNNs), to learn and recognize patterns indicative of drowsiness. These models are trained on labeled datasets containing examples of both alert and drowsy states.
- 6) *Classification*: The trained model classifies the input data into categories, typically distinguishing between an alert and a drowsy state. Some systems use predefined thresholds, while others leverage the continuous output of the model.
- 7) *Alert Generation*: When the system detects signs of drowsiness beyond a certain threshold, it triggers an alert. This alert can be in the form of a sound, visual warning, or haptic feedback to alert the driver.
- 8) *Real-Time Monitoring*: The entire process is designed to work in real-time, continuously monitoring the driver's condition during the journey.

It's important to note that the effectiveness of driver drowsiness detection systems depends on the quality of sensors, robustness of algorithms, and the ability to adapt to different driving conditions and individual variations in behaviour. Additionally, user acceptance and ethical considerations, such as privacy concerns, are crucial aspects that need to be addressed in the development and implementation of these systems.

CONCLUSION

In conclusion, the application of Convolutional Neural Networks (CNNs) in driver drowsiness detection demonstrates significant promise for enhancing road safety. Through real-time analysis of facial features, CNNs can accurately identify signs of driver fatigue, offering a proactive mechanism to prevent potential accidents. However, ongoing research and development are essential to address challenges such as diverse datasets, real-world adaptability, and user acceptance, ensuring the continued improvement and effectiveness of CNN-based driver detection systems on the road.

REFERENCES

1. J. May and C. Baldwin, "Driver fatigue: The importance of identifying causal factors of fatigue when considering detection and countermeasure technologies," *Transp. Res. F, Traffic Psychol. Behav.*, vol. 12, no. 3, pp. 218–224, 2009.
2. S. Lal and A. Craig, "A critical review of the psychophysiology of driver fatigue," *Biol. Psychol.*, vol. 55, no. 3, pp. 173–194, 2001.
3. E. Hitchcock and G. Matthews, "Multidimensional assessment of fatigue: A review and recommendations," in *Proc. Int. Conf. Fatigue Manage. Transp. Oper.*, Seattle, WA, USA, Sep. 2005.
4. Williamson, A. Feyer, and R. Friswell, "The impact of work practices on fatigue in long distance truck drivers," *Accident Anal. Prevent.*, vol. 28, no. 6, pp. 709–719, 1996.
5. W. Dement and M. Carskadon, "Current perspectives on daytime sleepiness: The issues," *Sleep*, vol. 5, no. S2, pp. S56–S66, 1982.
6. L. Hartley, T. Horberry, N. Mabbott, and G. Krueger, "Review of fatigue detection and prediction technologies," *Nat. Road Transp. Commiss., Melbourne, Vic., Australia, Tech. Rep.*, 2000.
7. Sahayadhas, K. Sundaraj, and M. Murugappan, "Detecting driver drowsiness based on sensors: A review," *Sensors*, vol. 12, pp. 16 937–16 953, 2012.
8. S. Kee, S. Tamrin, and Y. Goh, "Driving fatigue and performance among occupational drivers in simulated prolonged driving," *Global J. HealthSci.*, vol. 2, no. 1, pp. 167–177, 2010.
9. M.-H. Sigari, M.-R. Pourshahabi, M. Soryani, and M. Fathy, "A review on driver face monitoring systems for fatigue and distraction detection", *Int. J. Adv. Sci. Technol.*, vol. 64, pp. 73–100, 2014.