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Drowsiness Sensing System of Driver Based on Behavioral Characteristics to Prevent Road Accidents Using Real-Time Optimized Computer Vision

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Abstract: A computer vision-based system called the Drowsiness Sensing Device using OpenCV was designed to identify driver drowsiness. The technology uses video frames from a camera positioned inside a car to identify different sleepiness indicators, including the length of eye closure and head position. The Eye Aspect Ratio (EAR), which aids in trying to assess drowsiness, is determined using the OpenCV library, which is also used to extract feature points and detect eye blinks. The system also has an alarm mechanism that sounds when a certain level of drowsiness is attained, alerting the driver to take the appropriate action. The proposed approach can be possibly employed to reduce the number of accidents occurred due to driver drowsiness. The suggested system is a real-time drowsiness sensing system that makes use of OpenCV to gauge a person's level of drowsiness. The technology employs a camera to take pictures of the driver's face, assessing the features like the mouth and eyes to determine how sleepy they are. The system can identify drowsiness by noticing changes in the eyes, such as drooping eyelids, and mouth movements, such as yawning. When the amount of drowsiness surpasses a predetermined threshold, the system informs the driver by assessing the photos using machine learning techniques. By prompting the driver to take a break, the proposed technology may help prevent accidents brought on by drowsy driving.

1. INTRODUCTION

A drowsiness detection system is a technological device that keeps track of a person's behavior and physiology for signs of drowsiness using sensors and algorithms. When someone is about to do something vital, like drive, operate heavy machinery or labor in a dangerous setting, it is intended to warn them before they fall asleep. Systems for detecting drowsiness often include physiological and behavioral tests to evaluate a person's level of drowsiness. Among these techniques are tracking eye movements, keeping an eye on respiration and heart rate, measuring brain waves, and examining body posture and facial expressions. The device then analyses these metrics using algorithms to ascertain if the user is exhibiting signs of sleepiness. If the system notices that a user is starting to nod off, it may sound an alarm or vibrate the vehicle to notify the user to take action to avoid an accident. Some systems can also change the surroundings or make recommendations to keep the user alert, such as changing the temperature or suggesting a break. In fields like transportation, healthcare, and manufacturing where employees must remain aware for extended periods, drowsiness detection technologies are becoming more and more common. They may also assist people in keeping track of their sleep habits and spotting potential sleep problems. They can help prevent accidents and increase workplace safety.

Sensors and algorithms are frequently used in drowsiness-sensing devices, which track a person's level of wakefulness and look for indicators of drowsiness. Here are a few typical methods:

1. Eye monitoring: This technique includes following the driver's gaze with a camera. The device can recognize drowsiness if the driver's eyes are closed for a long time if they are not fixed on the road, or if their eye motions do so.

2. Face scanning is a technique used to identify drowsiness in drivers. Typical drowsiness indicators include drooping eyelids, a slack mouth, and a blank stare.
3. Examination of the driver's steering and acceleration patterns is used in this technique to look for indicators of tiredness. For instance, it could be a sign of tiredness if the motorist frequently corrects themselves to stay in their lane or drives erratically.
4. EEG (Electroencephalogram) analysis: This technique uses sensors applied to the driver's scalp to record the brain's electrical activity. The device may identify drowsiness-related changes in brain activity, such as an increase in alpha waves.

The system can inform the driver in several ways once it has identified indicators of intoxication, including ringing an alarm, flashing lights, or shaking the seat. Depending on the method, different sleepiness detection algorithms are employed. For instance, drowsiness-related patterns in eye movements may be identified by eye-tracking algorithms using machine learning models, and similar patterns may be identified by facial analysis algorithms using computer vision techniques. Signal processing methods may be used by EEG analysis algorithms to examine the electrical signals coming from the brain.

2. EXISTING SYSTEM

Neural Networks: It is a set of algorithms that aims to identify hidden connections in a piece of data utilizing a procedure that imitates how the human brain works. In this context, neural networks are systems of neurons that can be either organic or synthetic in origin. Local Binary Pattern extraction: It has a useful texture descriptor for images that thresholds the neighboring pixels depending on the worth of the current pixel. LBP descriptors effectively capture an image's local spatial patterns and grayscale contrast. Support Vector Machine: It is a supervised machine learning algorithm used for both classification and regression. Although we also refer to these issues as regression issues, categorization is where they fit in best. Finding a hyperplane in an N-dimensional space that clearly classifies the data points is the goal of the SVM method. The number of features determines the hyperplane's size. The hyperplane resembles a line when there are just two input features.

Few Drawbacks faced in existing systems:

1. Low Accuracy.
2. Fewer Features due to the local binary pattern.

3. PROPOSED SYSTEM

1. Input Video: An external webcam or default webcam is utilized to record video.
2. Cascade classifier: They are trained using several positive (with faces or objects) and arbitrary negative images (without faces or objects). OpenCV contains several pre-trained cascading classifiers used in image processing to detect frontal views of faces and the upper body.
3. Feature Extraction: It is a method of converting unprocessed converting raw data into numerical features that may be handled while keeping the original data set's content intact.
4. Classification of fatigue person: A fatigued person gets classified into various categories based on that, the system verifies if the person is under drowsiness.

Advantages:

1. Has high accuracy.
2. Extraction of features is done properly.

Disadvantages:

1. Pose variations are quite sensitive to the facial recognition technology.
2. Changing camera angles or head movements might alter the texture of a person's face and produce an inaccurate result.
3. Occlusion refers to the face with a moustache, beard, and other accessories.

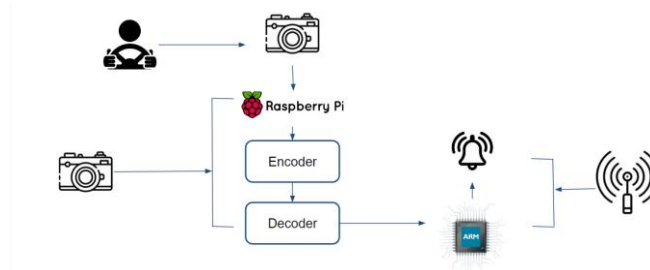


FIGURE 1. Architecture Diagram

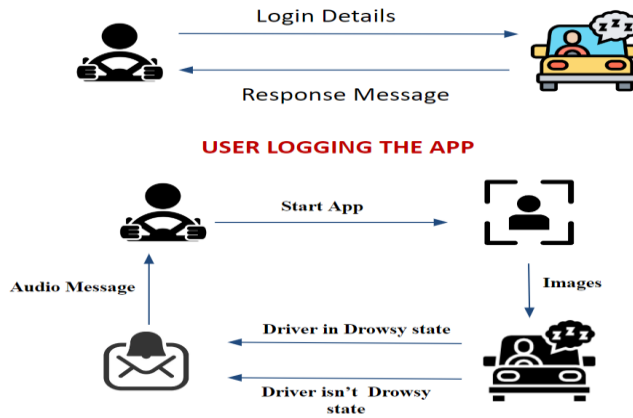


FIGURE 2. Dataflow Diagram

4. ALGORITHMS USED

Haar Cascade algorithm: An algorithm that can find items in pictures regardless of their size or position. This algorithm can operate in real-time and is not overly complicated. A haar-cascade detector can be trained to recognize a variety of items, including automobiles, bikes, structures, fruits, etc. It is also a machine learning-based strategy, and the classifier is trained using various positive and unfavorable images.

Positive images – These photos contain those that we want our classifier to be able to identify.
 Negative Images – Images containing everything else, omitting the object we're trying to find.

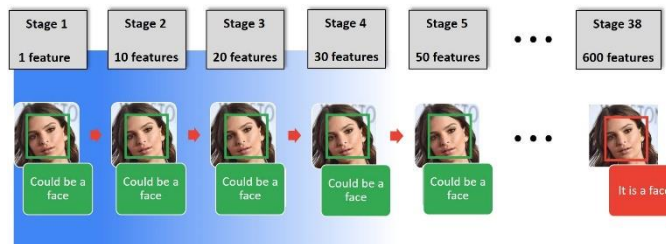


FIGURE 3. working of Haar Cascade algorithm

5. METHODOLOGY & IMPLEMENTATION

A key part of a driver sleepiness detection system is the **head posture estimate module**. It is employed to make educated guesses regarding the location and tilt of the driver's head in live video footage shot by a camera. The module frequently makes use of computer vision techniques and algorithms for figuring out the camera's head's 3D translation and rotation. Driver sleepiness detection systems can employ a variety of well-known head pose estimate libraries. OpenCV is among the most popular libraries. To predict the 3D rotation and translation of the head from a 2D image, OpenCV offers a pre-trained head pose estimation model.

1. These procedures the ability to incorporate the OpenCV head posture estimate module into a driver sleepiness detection system:
Set up OpenCV on your computer.
2. Configure the video capture module to capture the camera's video stream.
3. Using the `cv2.dnn.readNetFromCaffe()` function, load the previously practiced head position estimation model.
4. From each frame of the video stream, estimate the head pose using the model.
5. To identify changes in driving behavior that can point to tiredness, and compute parameters like head orientation or head movement.

The accuracy and robustness of the head pose estimate module's posture estimation could be further improved by using multiple cameras or depth sensors. The accuracy and resilience of other methods, such as machine learning-based head posture estimation models, can be improved under difficult lighting or occlusion settings. An integral component of a driver fatigue detection system is the **bio-sensors module**. It is employed to gauge bodily signals such as heart rate, skin conductance, and brainwaves coming from the driver. These indicators can be utilized to deduce the cognitive and emotional states of the driver, which may point to sleepiness or weariness. Driver sleepiness detection systems can make use of several well-liked bio-sensors. Electroencephalography (EEG), electrocardiography (ECG), and skin conductance sensors are a few of the most often utilized sensors. The following is how to use the bio-sensors module in a system:

1. Depending on the relevant physiological signals, choose the best biosensors.
2. Set up the sensors to record the driver's bodily physiological signals.
3. To extract pertinent elements from the signals, such as alpha power or heart rate variability, and process the signals.
4. Classify the driver's mental and emotional states, such as awareness or tiredness, using machine learning techniques.
5. To increase the system's overall accuracy and robustness, integrate the bio-sensors module's results with those from other modules, such as the face identification and eye tracking modules.

By applying cutting-edge signal processing methods, like wavelet transforms or independent component analysis, to extract more illuminating aspects from the physiological inputs, the biosensors module can be further enhanced. The accuracy and robustness of the bio-sensors module can also be increased using additional strategies like sensor fusion or deep learning-based models. A sleepiness detection system for drivers **data recording and analysis module** is a crucial part of the system. Data from numerous sensors and modules are recorded using it, pre-process the data, and analyze it to find relevant information and sleepy events. The following methods can be used to implement a system's data recording and analysis module:

1. Depending on the volume of data and processing needs, choose the best database technology and storage format.
2. To save the data from various sensors and modules, set up a data storage and retrieval system.
3. The data should be pre-processed to eliminate noise, remove unimportant information, and extract key features.
4. Analyse the data using statistical methods or machine learning algorithms to look for snooze-worthy occurrences or other significant information.
5. Visualize the outcomes and give the driver or other stakeholders feedback using a user interface or another method.

Using strategies like data compression, data fusion, or distributed computing can help the data recording and analysis module handle enormous amounts of data and increase processing effectiveness. The accuracy and robustness of the data analysis can also be increased by using additional methods like anomaly detection or time-series analysis. **Percentage of Eye Closure**, or **PERCLOS**, is a key sign of tired drivers in a vehicle drowsiness detection system. It is based on a study of how much time the driver spends, on average, with their eyes closed. The following steps can be taken to incorporate the PERCLOS module into a system:

1. To identify eye blinking and eye closing events, use the eye tracking module.
2. The PERCLOS metric can be calculated by dividing the total amount of time the driver's eyelids were closed by the whole length of observation time.
3. Decide on a PERCLOS threshold value after looking at data from previous research or the real world.
4. To identify sleepiness episodes, compare the estimated PERCLOS value to the threshold value.

5. When the PERCLOS value is greater than the threshold value, activate an alarm or other alert mechanism.

The accuracy and resilience of the PERCLOS measure can be increased by further optimizing the PERCLOS module by utilizing cutting-edge signal processing techniques, such as machine learning-based classification algorithms or statistical models. Additionally, by combining the PERCLOS measure with additional variables like heart rate variability or head posture estimation, the driver sleepiness detection system's accuracy can be improved. Several driver drowsiness detection systems use the **Internal Zone of Mouth Opening (IZMO)** feature to track the driver's level of weariness. The distance between the front and rear teeth is measured by IZMO to determine when the mouth is open wider than a predefined threshold. Usually, this criterion is established using the typical mouth-opening distance of a focused driver. When the IZMO reading is higher than the acceptable level, the possibility of the driver feeling sleepy or worn out exists. A complete driver sleepiness detection system often combines the IZMO feature with additional elements like eye tracking, head posture, and steering wheel movements. These technologies can provide a more precise assessment of the driver's state of attentiveness and, if necessary, warn the driver by simultaneously monitoring various signals. These devices' main objective is to stop accidents brought on by driver weariness or drowsiness. A camera or sensor is used by the Internal Zone of Mouth Opening (IZMO) module in a driver drowsiness detection system to detect the driver's mouth opening. Following that, the module gauges mouth opening based on predetermined thresholds. Typically, the IZMO module operates as follows:

1. To use a camera or sensor, the system records photos or video of the driver's face.
2. The algorithm recognizes the mouth region in the photos or videos that were taken.
3. The device measures how far apart the top and lower lips are and compares that measurement to benchmarks to determine the extent of mouth opening.
4. The algorithm then evaluates how much the driver opens their mouth over time to determine if they are getting sleepier.

The device may notify the driver by issuing an auditory or visual warning, jolting the chair or the handle, or both if their mouth opens beyond a predetermined limit. It's crucial to remember that the IZMO module is just one of a variety of characteristics that may be incorporated into driver drowsiness sensing systems and that depending on the environment, the success of each implementation will vary. To provide accurate and trustworthy results, these systems must be designed and implemented with a number of factors and tactics in mind.

RESULT & CONCLUSION

The outcome of a Driver Drowsiness Sensing System project depends on its particular aims and objectives, along with the methods and tools used to carry them out. The major goal of a Driver Drowsiness Sensing System project is to create a device that can detect when a motorist is starting to nod off and notify them of it. The project should produce a working system that can complete this task. Once the system is developed, it should be tested to see how effectively it functions in actual scenarios. To assess the system's precision and efficacy, it may be tested on a range of drivers and under various driving circumstances. To describe how the system was created, how it functions, and how its effectiveness was assessed, thorough documentation of the project is required. This data can be utilised as a basis for additional study and development in addition to show stakeholders the project's success. To detect drowsy driving and lower the risk of accidents, a successful Driver Drowsiness Sensing System initiative has the potential to significantly advance driver safety. The project might catalyze additional study and advancement in this field. To increase traffic safety, driver drowsiness-sensing devices are becoming more and more crucial. These systems often monitor the driver's behavior and warn them if they seem to be getting carried away or drowsy using a range of sensors, cameras, and algorithms. The Internal Zone of Mouth Opening (IZMO) module, which measures the extent of mouth opening to produce a reading of the driver's alertness level, is one of the fundamental components involved in many drivers' drowsiness-detecting systems. Eye tracking, head position monitoring, and steering wheel angle detection are additional functions frequently found in driver sleepiness detection systems. IZMO is typically paired with these features to provide a more detailed assessment of the driver's state. By warning drivers ahead of potential risks, the device has the potential to save countless lives. It is crucial to keep in mind that these systems are not perfect and could have drawbacks or problems in specific circumstances. To guarantee that these systems are as efficient and dependable as possible, it is crucial to continuously assess their performance and make necessary improvements.

Future Enhancement: With a Driver Drowsiness Sensing System project, there are several potential future improvements that might be taken into account, including:

1. Combination with additional safety systems: Integration of the sleepiness monitoring system with other safety features like automated emergency braking or lane departure warning systems is one improvement that may be made. This would enable a more thorough approach to driving safety.
2. Continuous monitoring in real-time: The system might be improved to continually monitor the driver rather than waiting for the system to notice indicators of intoxication. This would make it possible for the system to recognize tiredness before it starts to impair driving performance.
3. Personalization: Based on unique driving characteristics, an improved system might be created to tailor the sleepiness detection algorithm. For instance, the system may take into account a driver's age, gender, and driving history to better anticipate when they are likely to feel sleepy.
4. Multimodal detection: At the moment, the majority of driver sleepiness detecting systems depend on a single sensor modality, including head movement detection or eye tracking. Several sensing modalities, such as biometric data or facial expression detection, might be used in an improved system to increase accuracy and dependability.
5. Data analysis and artificial intelligence: An enhanced system that combines data analysis and artificial intelligence techniques may be developed in order to gradually raise the accuracy of sleepiness detection. The system would be better equipped to distinguish between normal driving behavior and signs of intoxication as a result.

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