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Driver drowsiness detection system

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ABSTRACT

Making insightful systems to prevent minor accidents can be particularly incredible in limiting the difficulty and loss of life. Street crashes and re-lated types of mishaps are a typical reason for injury and death among the human population. One of the elements which assume a significant part in mishaps is the human blunders including driving exhaustion depending on new brilliant methods; this application recognizes the indications of weariness and lethargy despite the individual at the hour of driving. The proposed framework depends on three separate calculations. In this model, the individual's face is monitored by a camera in the initial step by getting 14-16fps video grouping. At that point, the pictures are changed from RGB space into YCbCr and HSV spaces. The face territory is isolated from different parts and exceptionally precise HDP is accomplished. That the eyes are open or shut in a particular timespan is controlled by zeroing in on thresholding and conditions concerning the balance of human appearances. The proposed framework has been carried out in excess of thirty distinctive video successions with normal precision of 93.18 and discovery rate (DR) of 92.71 out of approximately 2500 picture outlines. High precision in division, low blunder rate and fast preparing of information recognizes this framework from comparable ones. This framework can limit the quantity of mishaps brought about by drivers' exhaustion.

Keywords— Drowsy Detection, Viola Jones, MATLAB

1. INTRODUCTION

Every year many individuals lose their lives because of car crashes around the planet. Street mishap is a worldwide misfortune with an over-rising pattern. The growing level of a traffic disaster has become a significant issue for society. Most of the accidents are caused by driver blunder (for example speeding) or poor working works including absence of safety belt use, interruptions, exhaustion, rash driving, and liquor or medication use while driving. Of all these issues, the main issue to be considered is the driver's focus as well as driver's heedlessness or interruption. In this unique situation, it is essential to utilize new advances to plan and fabricate frameworks that can screen drivers and to gauge their degree of consideration during the whole cycle of driving.

Therefore, there is a need to develop a system that will perceive and instruct a driver regarding his dreadful physiological conditions, which could basically lessen no. of fatigue related mishaps.

In this paper, a Viola Jones algorithm is presented to decrease the quantity of mishaps brought about by driver exhaustion and accordingly improve street security. It likewise measures the images rapidly.

2. LITERATURE REVIEW

In a proposal to increase exactness and accelerate drowsiness detection a couple of procedures have been proposed. This framework manages utilizing data obtained for the double form of the picture to find the edges of the face, which controls the area of where the eyes may exist.

T. Vesselenyi paper "Driver drowsiness detection using ANN image processing" detects accuracy of 100 positive results. This paper presents an examination regarding the possibility to build a driver drowsiness detection for vehicles based on 3 methods: EEG and EOG signal processing and driver image analysis. The authors have examined the likelihood to detect drowsy or alert state of the driver based on images taken during driving and by breaking down the condition of the driver's eyes. The modest number of neurons utilized in the secret layers to effectively arrange the images permits the usage of these organizations on minimal registering gadgets, utilizing a tiny part of their memory. The training of the network should be

possible explicitly for every driver, in this way improving the classification achievement rate.[1]

Eddie. E. Gallarza paper “Real Time Driver Drowsiness Detection Based on Driver’s Face Image Behavior Using a System of Human Computer Interaction implemented in a Smartphone” detects accuracy of 93.37. This paper presents a surveillance system to detect drowsiness. It is utilized a cellphone like small PC with a portable application utilizing android working framework to implement the Human Computer Interaction System. For the recognition of drowsiness, the most applicable visual pointers that mirror the driver’s condition are the behaviour of the eyes, the sidelong and frontal consent of the head and the yawn. The framework works enough under normal lighting conditions and no matter the utilization of driver accessories like glasses, hearing aids or a cap. The increment in the characteristics in cell phones made possible to develop an application of artificial intelligence.[2]

Younes Ed-Doughmi paper “Real-Time System for Driver Fatigue Detection Based on a Recurrent Neuronal Network” detects accuracy of 92. This paper presents an approach to analyze and anticipate driver drowsiness by applying a Recurrent Neural Network. He utilized a dataset to shape and affirm our model and implemented repetitive neural network design multi-layer model-based 3D Convolutional Networks to distinguish driver drowsiness. Main goal is to provide a neural network architecture for a reasonable and portable sleepiness recognition for drivers. To additionally improve these outcomes, a more customized data needs to be created that is more fitting to the subject of sleepiness in a climate near what the driver may experience in a genuine situation.[3]

Reza Ghoddosian paper “A Realistic Dataset and Baseline Temporal Model for Early Drowsiness Detection” presents a huge and public genuine dataset of 60 subjects, with video portions marked as concentrated, unwary and sleepy. The real-life dataset (RLDD) comprises of around 30 hours of video. The method presented in this paper is Hierarchical Multiscale Long Short-Term Memory (HM-LSTM) network. The proposed methodology has low computational and capacity requests. Our results showed that our method out-performs human judgment in two arranged estimations on the RLDD dataset.[5]

3. PROPOSED REVIEW

Viola Jones algorithm

The Viola Jones object recognition structure is the technique for reformist detection rates made in 2001 by Paul Viola and Michael Jones, the Viola-Jones calculation is an article acknowledgment structure that permits the location of image features progressively. It is quite powerful and its application has demonstrated to be particularly eminent in real time face detection. There are two phases in this algorithm:

- Detection
- Training

Training comes before detection, yet for the good of clarification, I’ll examine detection first.

3.1 Detection

Viola-Jones was intended for frontal appearances, so it can distinguish frontal the best instead of faces looking sideways, upwards or downwards. Prior to distinguishing a face, the picture is changed over into grayscale, since it is simpler to work with and there’s lesser information to measure. The Viola-Jones calculation initially distinguishes the face on the grayscale picture and afterward finds the area on the colored picture.

3.2 HAAR like features

HAAR-like highlights are named after Alfred HAAR, a Hungarian mathematician in the nineteenth century who built up the idea of HAAR wavelets (sort of ancestor of HAAR-like features). The highlights beneath show a crate with a light side and a clouded side, which is the way the machine figures out what the element is. Some of the time one side will be lighter than the other, as in an edge of an eyebrow. Sometimes the center segment might be shinier which can be interpreted as a nose.

In the detection phase, a window of the objective size is moved over the information picture, and for each section of the image the HAAR-like part has been determined. The difference is then compared to the learning threshold that isolates non-objects from objects. Since such a HAAR-like component is just a classifier (its location quality is marginally better compared to irregular speculators) an enormous number of HAAR-like highlights are essential to depict an item with adequate exactness.

Viola Jones Algorithm structures the premise of this hearty framework.[7] Viola Jones calculation helps in accomplishing high location rates. It additionally measures the pictures quickly. It shapes the premise of the greater part of the ongoing frameworks as it works just on the current single dim scale picture. There are three fundamental pieces of this calculation:

- (a) Integral Image which permits quick component assessment.
 - (b) Classifier function which is constructed utilizing modest number of significant highlights.
 - (c) The strategy for consolidating the classifiers in a course structure to increase the detector by focusing on region of interest.
- Viola Jones Algorithm is quick, productive and gives level of precision.

Drowsy Detection Algorithm: Laziness of an individual can be assessed by the widely inclusive time span for which his/her eyes are in shut state. In this system, fundamental thought is given to the faster ID and treatment of data. The amount of housings for which eyes are shut is checked. Accepting the amount of edges outperforms a particular worth, a caution message is created on the grandstand showing that the driver is feeling drowsy.

Eyes and mouth detection: After the face is identified utilizing Viola-Jones, the area containing the eyes and mouth must be isolated. To identify the coordinate from where the locale of eye is beginning, certain calculations are done. After the rectangular window is removed, we have considered that the eyes are situated a ways off of $(0.25 \times \text{height of window})$ from the top and $(0.15 \times \text{width of window})$ from the left. The size of window is $(0.25 \times \text{tallness of window})$ in height and $(0.68 \times \text{width of window})$ in width. To recognize the arrange from where the locale of mouth is beginning certain calculations are done. After the rectangular window is removed, we have thought about that the mouth are situated a ways off of $(0.67 \times \text{tallness of window})$ from the top and $(0.27 \times \text{width of window})$ from the left. The size of window is $(0.20 \times \text{height of window})$ in tallness and $(0.45 \times \text{width of window})$ in width.

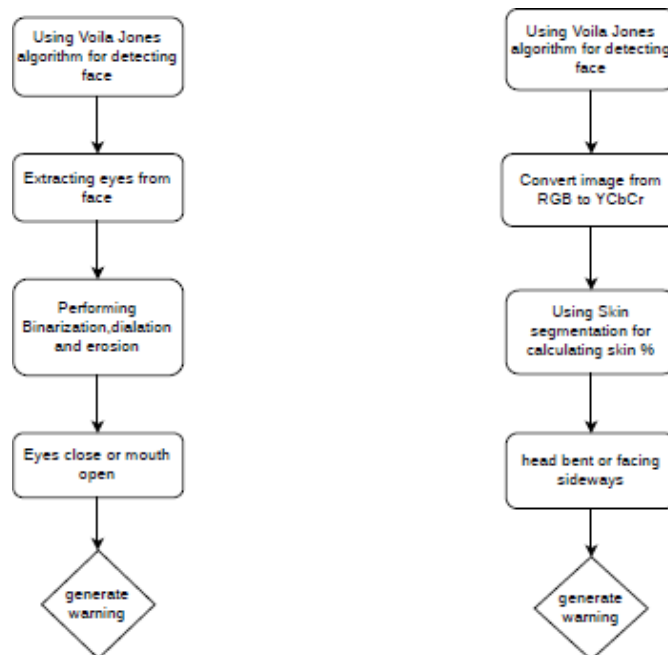


Fig. 1: Flowchart of proposed algorithm

Skin Segmentation: Starting with a shading picture, the main stage is to change it to a skin- probability picture. This includes changing each pixel from RGB portrayal to chroma portrayal and deciding the probability esteem dependent on the condition given in the past segment. The skin-probability picture will be a dim scale picture whose dark qualities address the probability of the pixel having a place with skin.

An image which is taken inside a vehicle joins the driver’s face. Usually a camera takes pictures inside the RGB model (Red, Green and Blue). In any case, the RGB model remembers brilliance for expansion to the tones. While investigating a human face, RGB model is extremely delicate in picture intensity. Hence, to eliminate the intensity from the pictures is second step. We utilize the YCbCr space since it is generally utilized in video pressure norms. Since the skin tone relies upon intensity, we non-linearly change the YCbCr shading space to make the skin bunch luma free. This likewise empowers hearty identification of dim and fair complexion tone tones. The primary benefit of changing the picture over to the YCbCr space is that impact of glow can be eliminated during our picture handling. In the RGB space, every part of the image (red, green and blue) has an alternate splendor. However, in the YCbCr area all data about the splendor is given by the Y part, since the Cb (blue) and Cr (red) segments are free from the radiance.

Decision Making: The first frame is used for learning. All the results are calculated taking first frame as ideal frame.



Fig. 2: Face Detection

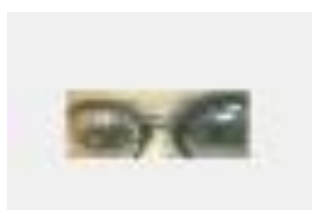


Fig. 3: Eyes Detection



Fig. 4: Mouth Detection

- (a) Eyes Closed: At the point when eyes are closed, the quantity of dark pixels in binary image diminishes impressively. In case eyes are found closed for at least 2 consecutive seconds (for instance $2 * 16 = 32$ frames, considering 16 edges each second), by then the warning will be generated.
- (b) Mouth Open: At the point when mouth is open, the subsequent dark pixels in binary image can be significantly bigger or more modest than the ideal edge. The distinction can be over 6 of the dark pixels in ideal edge. In the occasion that mouth is found open for at least 2 consecutive seconds (for instance $2 * 16 = 32$ frames, considering 16 edges each second), it suggests that the individual is yawning and accordingly the warning will be produced.
- (c) Head Lowering: If the head is lowered or turned the amount of skin pixels decrease when diverged from the ideal packaging. In case head is lowered or found turned in various manners for at least 2 progressive seconds (for instance $2 * 16 = 32$ frames, considering 16 edges each second), it infers that the individual is vulnerable for setback and appropriately the warning will be delivered.

4. RESULTS

Around 1000 positive pictures and 5000 negative pictures were taken as test datasets to prepare the Face, Eye and Mouth classifiers. Drowsiness of an individual can be assessed by the comprehensive time span for which his/her eyes are in closed state. In this system, fundamental thought is given to the faster acknowledgement and treatment of data. The amount of traces for which eyes are closed is checked. If number of frames exceed a specific value then an alert message is made on the show showing that the driver is feeling moderate. In this computation, first the video is uploaded for image and video processing. By then the Haar classifier is used to look moreover, perceive the faces in each individual packaging. In case no face is distinguished then another packaging is acquired. If a face is distinguished, a space of interest is set apart inside the face. This territory of interest contains the eyes and mouth. Describing a space of interest basically diminishes the computational necessities of the structure. There are likewise numerous alternate ways like Open CV to do this undertaking. Yet, we have selected MATLAB as it is a lot of easy to use also, handily comprehended.

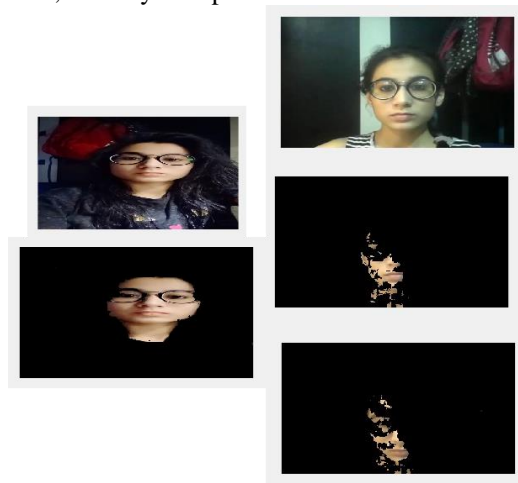


Fig. 5: Drowsiness Detected

5. ACCURACY

The estimation is minded thirty recordings of around 5-10 seconds. The estimation offers right reaction on around 25 recordings that makes it about 83.33 precise.

6. CONCLUSION

Viola Jones computation is used to distinguish various faces and space of captivated features like eyes, nose, mouth, etc. Exact area of features increases immediately when a camera of high judgments is used. The calculation introduced in this paper shapes the premise of many 'Real Time System' which gives exceptionally exact outcomes in less time. This framework was constructed remembering each conceivable behavior of the driver and furthermore the requirement for a speedy figured outcome which is precisely accomplished by the Viola Jones calculation which is the foundation of Drowsy Detector Algorithm. In the momentum framework when the numerous countenances are recognized the outcomes are incorrect. The ideal outcome would be, on the off chance that different countenances are recognized, just the driver's face ought to get distinguished. This could be accomplished utilizing the Depth Calculations wherein the space of the square shapes of multiple faces are determined and the biggest territory is chosen since in the majority of the cases the driver's face would be ahead and every one of the unessential appearances would be behind the driver. The square shape which is ahead would clearly have a bigger region than the remainder of the distinguished faces thus choosing the biggest region square shape would be appropriate.

Ceaseless observing of the Driver's weakness level ought to be done and dependent on that a choice ought to be made and a message ought to be shipped off the Driver's connected person. GSM technologies can be included. It is imperative to have a coordinated processor while carrying out this frame-work on the vehicle which thus would be associated with the GSM System. Likewise, it is important to have a presentation framework which would show the driver's face if the driver needs to do any change in accordance with the camera.

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