

Securing Human Life By Fatigue Detection During Driving

Engr. Muhammad Jibran, Engr. M. Fahad Khan

Department of Computer Engineering Faculty of Telecommunication and Information Engineering University of Engineering & Technology Taxila

ABSTRACT

Traffic accidents are mainly caused by driver inattention. Physiologica and bahavioral elements need to be measured in order to detect the drivers inattention. Different approaches have been made, and among them Computer Vision has the potential of monitoring the person behind the wheel without interfering with his driving.

A computer vision system for driving monitoring uses face location and tracking as the first processing stage. On the next stage the different facial features are extracted and tracked for monitoring the driver's vigilance. In this thesis I have developed a system that can monitor the alertness of drivers in order to prevent people from falling asleep at the wheel. The other main aim of this algorithm is to have efficient performance on low quality webcam and without the use of infrared light which is harmful for the human eye. Motor vehicle accidents cause injury and death, and this system will help to decrease the amount of crashes due to fatigued drivers. The proposed algorithm will work in three main stages. In first stage the face of the driver is detected and tracked. In the second stage the facial features are extracted for further processing. In last stage the most crucial parameter is monitored which is eye's status. In this last stage it is determined that whether the eyes are closed or open. On the basis of this result the warning is issued to the driver to take a break.

1.Introduction

Driving is a complex task where the driver is responsible of watching the road, taking the correct decisions on time and finally responding to other drivers' actions and different road conditions. Vigilance is the state of wakefulness and ability to effectively respond to external stimuli. It is crucial for safe driving. Among all fatigue related accidents, crashes caused by fell-asleep-drivers are common and serious in terms of injury severity.

According to recent statistics driver fatigue or vigilance degradation is the main cause of 17.9% of fatalities and 26.4% of injuries on roads [1]. Vigilance levels degrade mainly because of sleep deprivation, long monotonous

driving on highways and other medical conditions and brain disorders such as narcolepsy. Majority of the road accidents are mainly due to the driver fatigue. Driving for a long period of time causes excessive fatigue and tiredness which in turn makes the driver sleepy or loose awareness. The study states that the cause of an accident falls into one of the following main categories: (1) human, (2) vehicular, and (3) environmental. The driver's error accounted for 93% of the crashes. The other two categories of causative factors were cited as 13% for the vehicle factor and 34% for environmental factors. It is important to note that in some cases; more than one factor was assigned as a causal factor.

The three main categories (human, vehicular, and environmental) are related among each other, and human error can be caused by improper vehicle or highway design characteristics. The recognized three major types of errors within the human error category: (1) recognition, (2) decision, and (3) performance [2]. Decision errors refer to those that occur as a result of a driver's improper course of action or failure to take action. A recognition error may occur if the driver does not properly perceive or comprehend a situation. To perform all these activities in time and accurately its necessary that driver must be vigilant.

In Pakistan, in the year 2006, police has recorded 10125 crashes, of which 4193 were fatal causing death or severe injuries to the driver. Agha Khan university conducted a study which shows that traffic accidents cause 56% deaths and only 4% serious injuries. These values are much higher compared to the statistics given by the government officials which shows death rate of 11.2 of 100,000. [3]

The focus of that system was to detect drowsiness, although the addition of the fixed gaze visual cue would make it possible to detect some kinds of cognitive distractions. The face and eye detection step is at the base of the image processing of the system. Human face detection and tracking is a very vast domain in computer vision and image processing field. A substantial number of algorithms have been developed in the last decade for this purpose. Face detection is an area of large interest as facial features contain large amount of useful information which can be used for several different purposes. Techniques of facial feature detection have also been employed in the proposed system.

. Face detection and tracking is the first step for other algorithms that work in face recognition or expression analysis in some areas of computer vision. The second step is the detection and tracking of the driver's eyes. Then in from the eyes it is determined that wither the eyes are open or closed. The aim of this paper is to develop a computer vision method able to detect and track the face of a driver in a robust fashion, also determine the status of the eyes, and with the highest precision possible. It is to serve as the bases of an automatic driver fatigue monitoring system.

2. Existing Systems and Approaches

Many researchers have worked in recent years on systems for driver inattention detection.. Sleep has a long history of research in the fields of psychology and medicine, where accurate measurements and indicators have been developed [4]. Electroencephalograms (EEG) [5] represent the electrical changes in the brain, measured with a series of electrodes placed in the scalp. The electrodes detect small voltages produced in the brain cortex. These potentials form waves at several frequencies, known as delta, theta, alpha, beta and gamma waves, which are linked to different cognitive and motor processes, including drowsiness and the different sleep stages, as shown in figure 1. Brain studies couple EEG with electrooculography (EOG), which detects eye movements, and electromyogram (EMG) that monitors muscular tone. These measurements provide the best data for detection of drowsiness, and as such have been used by several drowsiness detection systems, usually in conjunction with heart rate and breathing rate. The problem of these techniques is that they are intrusive to the subject. They require electrodes and other sensors to be placed on the head, face and chest as in figure 2, which may annoy the driver. They also need to be carefully placed: installing the electrodes to obtain an EEG requires external help and takes a few minutes, and medical equipment is always expensive. Recent research has introduced some contact-less readings, but no remarkable results have been achieved so far. Nonetheless, physiological measures such as EEG have been used in some projects [6], and are frequently used as the ground-truth for testing other, less invasive methods.

A driver's state of attention can also be characterized using indirect measurements and contact-less sensors. Lateral position of the vehicle inside the lane, steering wheel movements and time-to-line crossing are commonly used, and some commercial systems have been developed. These systems do not monitor the driver's condition, but its driving. Volvo Cars introduced its Driver Alert Control system [Volvo Car Corp. 08] in 2008, which is available on its high-end models. This system uses a camera, a

number of sensors and a central unit to monitor the movements of the car within the road lane, to assess whether the driver is drowsy (see figure 3). Mercedes-Benz has introduced a similar system (ATTENTION ASSIST) [DaimlerAG 09] in its newest E-Class vehicles

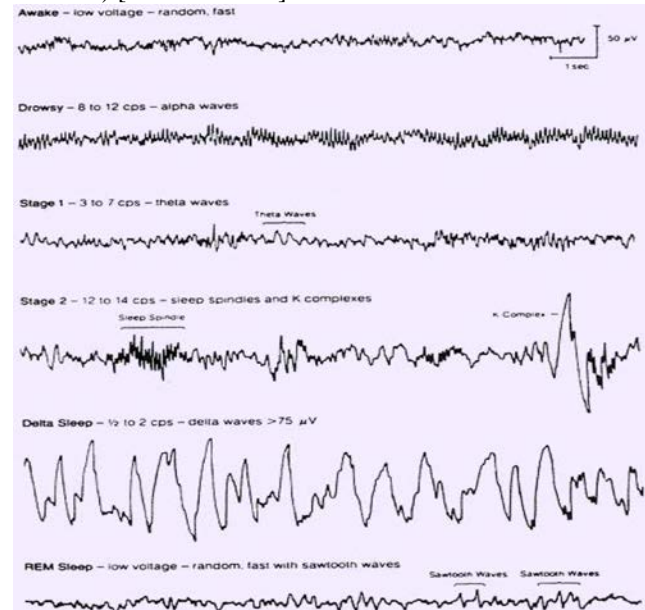


Figure 1: An example of brain wave activity for different sleep stages.[19]

(shown in figure 4). Daimler AG, owner of Mercedes-Benz, was in 2001 one of the first to develop a system based on vehicle speed, steering angle and vehicle position relative to road delimitation (recorded by a camera) to detect if the vehicle is about to leave the road [7]. Other manufactures have conducted research and presented prototypes. Toyota [5] used steering wheel movement sensors and pulse sensor to record the heart rate. Mitsubishi has reported the use of steering wheel sensors and measures of vehicle (such as lateral position of the car) to detect driver drowsiness in their advanced safety vehicle system [6].



Figure 2: A driver wearing a helmet with electrodes for EEG.

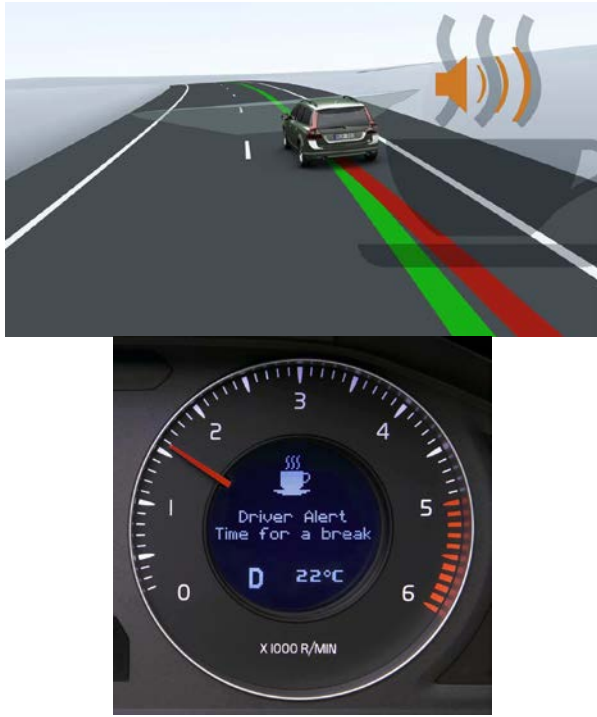


Figure 3: Volvo's Driver Alert Control. Images from Volvo Cars

These techniques are not invasive, and to date they obtain the most reliable results with the least number of false positives, a critical problem in this type of systems. However, they face several limitations such as geometric characteristics and state of the road, and driver experience. They also require a training period for each person, during which the driving style of the user is learned and modeled, and thus are not applicable to the occasional driver. Despite the high number of parameters involved, these systems are basic in that the behaviors they are able to detect are few: the measurements may not reflect user behaviors such as the so-called micro-sleeps: if a drowsy driver falls asleep for few seconds, the lateral position may not change in a straight road [8]. Response time of these systems may compromise their effectiveness.

The driver had to wear a wrist band in order to measure heartbeat in the project held by Toyota called ASV (Advanced Safety Vehicle). Special contact lens or a helmet is used to detect eye and gaze motion. These techniques are not in common practice although they are intrusive.

A "Driver Fatigue Monitor System" by Attention Technologies, Inc. It is based on the PERCLOS (PERcent of the Time Eyelids are CLOSED) measure of driver fatigue. They have discovered that the PERCLOS-based system does work but has some flaws.



Figure 4: Mercedes-Benz's Attention Assist.[20]

Drivers in fatigue exhibit changes in the way their eyes perform some actions, like moving or blinking. These actions are known as visual behaviors, and are readily observable in drowsy drivers, and also in distracted ones. More precisely, typical characteristics include longer blink duration, modified blinking frequency, slow eyelid



Figure 5: System by Attention Technologies. Images from Attention Technologies.

movement, a smaller degree of eye opening and gaze (narrowed field of view, with reduced response to objects in the peripheral areas of vision). Although not purely visual, other characteristics that are included in this group are yawning, nodding, sluggish facial expression (due to relaxed muscular tone) and dropping posture. Of all of them, the percent of eye closure (PERCLOS) has been found to be the most reliable indicator of drowsiness [11]. Computer vision has been the tool of choice for many researchers to be used to monitor visual behaviors, as is non-intrusive. Most systems use one or two cameras to track the head and eyes of the subject [12]. Commercial products are available for general applications not focused on driving problems. A few companies commercialize systems as accessories for installation in vehicles, but are not part of the car manufacturers' developments: reliability is not high enough for car companies to take on the responsibility of its production and possible liability in case of malfunctioning. By installing the system

themselves, the owners take the responsibility instead. Seeing Machines sells the FaceLAB software that uses two cameras to track the face in 3D. They have also presented the Driver State Sensor (DSS), which calculates PERCLOS. The Swedish company SmartEye AG [13] offers mono- and multi-camera systems that detect eye movements, gaze fixation and blink detection. Mono-camera systems have been a major focus on late years, because integration in industrial production is much easier and less costly. As it can be seen, some systems have indeed entered the market, but in the literature there are few details available regarding the methods and parameters of those systems.

Computer vision systems use natural light, infrared (IR) or both to illuminate the face of the driver. This is an important problem of system that must work 24/7 and day and night scenarios are very different. Usually daytime algorithms need to be adapted to work during nighttime. [14] presented a system using 3D techniques to estimate and track the line of sight of a person using multiple cameras. In [15] a system with active IR illumination and a camera is implemented. In addition to providing illumination, IR light reflects on the eye's cornea and produces the red-eye effect, similar to the one appearing in photography when flash light is used. This reflection can be detected and used for locating and tracking the rest of the face. They suggested to estimate the local gaze direction based on pupil location. In [16] a system based on natural light was presented. Systems relying on a single visual cue may fail when the required features can not be detected accurately or reliably. Also, people's visual behaviors under fatigue or distraction change from person to person, and a single indicator may not be representative of the overall cognitive state [15]. Uncertainty and the ambiguity are reduced by multiple visual cues as compared to that of relying on only one source. Recent research points in this direction. In [17], another multi-sensor system was presented. The authors tested two decision making methods, fuzzy inference system (FIS) and artificial neural networks (ANN) to fuse the data and obtain an estimation of the drowsiness state of the driver. The AWAKE European project proposed a multisensory system that integrated multiple visual cues with information from the vehicle and the environment. This system should be configured explicitly for every driver, requiring a learning stage. Other European initiative, SENSATION, carried on with this line of research. A non-intrusive system fusing driver's condition information with data from his/her driving, with minimal to no per-person customization would be the best candidate for mainstream adoption, and thus research has concentrated on this option lately.

3 Proposed System

The algorithm proposed is a computer vision algorithm that aids in the detection of the current driver state of vigilance. It detects the current state of the driver eyes in every frame (open or closed). Applying this algorithm on consecutive video frames may aid in the calculation of eye closure period. Eye closure periods for drowsy drivers are longer than normal blinking, a fact that can be exploited to monitor a driver state of vigilance. It is also a very critical parameter because the closure of eyes for a little longer time could result in severe crash. So we will warn the driver as soon as closed eye is detected. The flow chart of the algorithm is represented in Figure 7.

3.1 Algorithm Stages

The major stages of algorithm are as:

3.1.1 Image Capture

The image is captured from the video, where image is a numeric representation (normally binary) of a two-dimensional image. The video is acquired by using a low cost web camera. The camera provides 30 frames per second at VGA mode. The web camera is shown in figure 6.



Figure 6: The web camera used [18]

Then the recorded video is opened in MATLAB and the frames are grabbed, then the algorithm is run on every frame detecting the driver's vigilance.

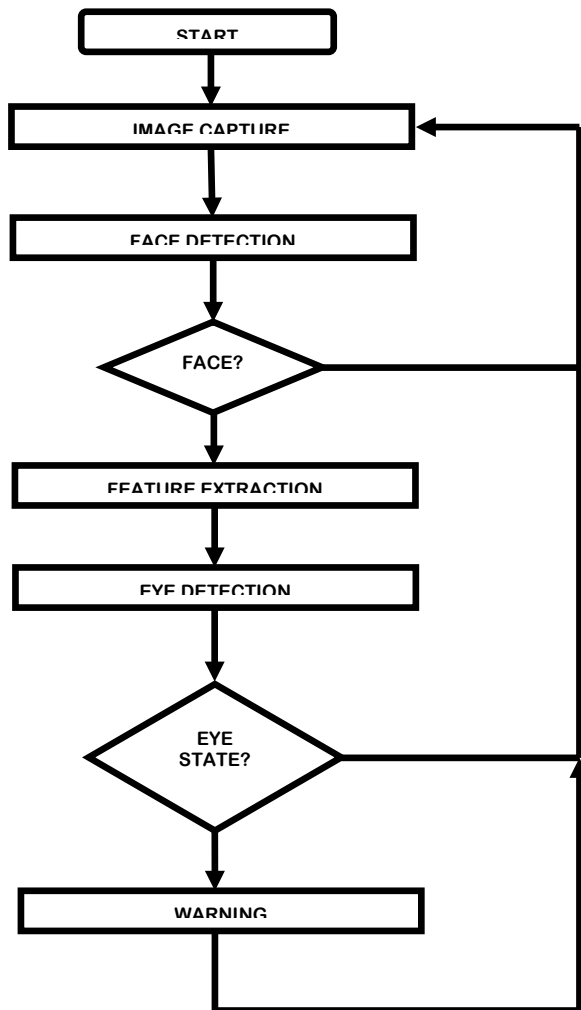


Figure 7: The proposed Algorithm.

3.1.2 Face Detection

In every given frame the face is detected using an specified algorithm. Human face size and location in digital images is detected using the technique of face detection. This technique ignores all the information such as building, cars, trees etc from the image except the facial features, leaving us with the detected face. Face detection is considered specialized form of object class detection. Object class detection technique involves the detection of the object belonging to a specific class. Size and location of the object is detected in order to differentiate the object from its background. Frontal human face detection was the focus of earlier human face detection algorithms where as newer algorithms focus more on multi-view human face detection and other complex scenarios including the detection of the faces which are in plane rotation with the observer or the scenarios in which the face is vertically rotated or left-right rotated i.e. out of plane rotation. New techniques focus on the variation or difference between

images or video frames based upon pose, face detection and lightening. The technique of pattern classification is often employed for face detection purpose in many algorithms. In this technique the content is transformed into features and these features are then trained by a classifier which shows if the detected image content is a face or not.

If the result of face detection comes positive then the algorithm proceeds to the next, otherwise the flow of algorithm goes back to the image capture stage. The different face detection algorithms are explained in detail in later chapters.

3.1.3 Facial Feature Detection

Facial Feature Detection (FFD) is to find the exact location of facial features, including mouth and eyes corners, lip contour, jaw contour, and the shape of the entire face. Face and facial feature detection are difficult problems, due to the large variations a face can have in a scene caused by factors such as intra-subject variations in pose, scale, expression, color, illumination, background clutter, presence of accessories, occlusions, hair, hats, eyeglasses, beard etc. The easiest case considers a single frontal face and divides it into region of interest like for mouth, eye and nose etc. The feature detection is used to determine the region of eyes so that their status can be determined easily and quickly. It is performed by segmenting the face that has been detected. This segmentation is based on the experimental results obtained from the different research papers.

3.1.4 Eye Detection

Eye detection is the essence of eye tracking and iris recognition – without it, there's no way to identify the eye itself. It sounds simple, but it's really quite complicated. In this stage the eyes are detected in the specified region by the feature detection. In the beginning it looks for the Eigen eye. This process is time taking and it is done just once. After the detection of Eigen eye it is just matched in the other frames for the same candidate. The eye detection process is explained in detail in later chapters.

3.1.5 Eyes State

In this stage, it is determined that if the eyes are closed or open. The detection of eyes status is very important. It is done by an algorithm which will be explained in the later chapters. If the eyes are detected to be closed then the warning is sounded. If the eyes are open then the algorithm goes to the first step of the image capture. The same pattern repeats to check the status of the eyes.

4 Experimental Results

4.1 Image Capturing

As mentioned earlier, the video is recorded by web camera for testing the system. The recorded video is used in

MATLAB by using “mmreader” command. The algorithm is performed on the each frame of the video.

4.2 Face Detection

The face detection is done by the Viola and Jones algorithm. For the accurate setup we used the cascade which is part of the OpenCV library. This cascade contains 24 stages and has a total of 2913 weak classifiers. Its starting window size is 24×24 pixels. The starting scale was set to 1.0, the scale step size was set to 1.1 and the position step size Δ was set to 1.0. In total 32 different scales were checked, yielding a total of more than 1.8 million possible detection windows. The OpenCV has the trained cascade so its easy to use. On the other hand the processing speed in OpenCV is very fast as compared in MATLAB because OpenCV is basic programming language as compared to MATLAB which is high level programming language. The results of face detector are as:

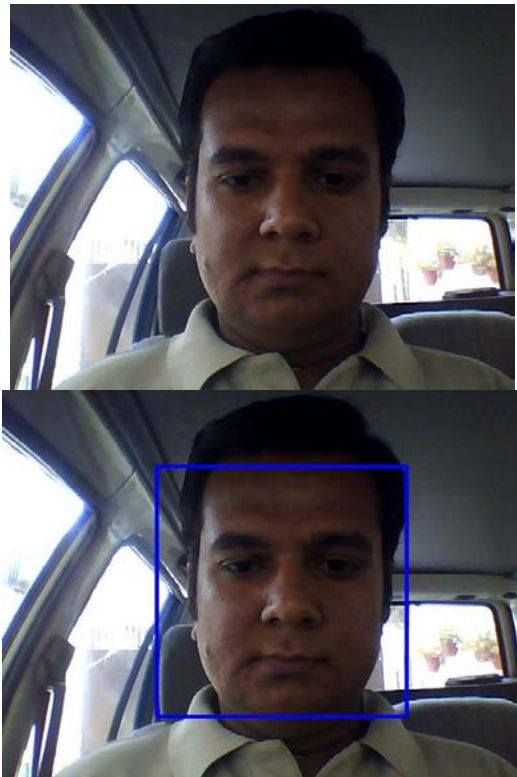


Figure 8: Input and output of face detector.

4.3 Facial Feature Extraction

For the facial feature extraction firstly the search areas are defined according to the geometry of the face. Then in these search areas specific content is found by its own algorithm.

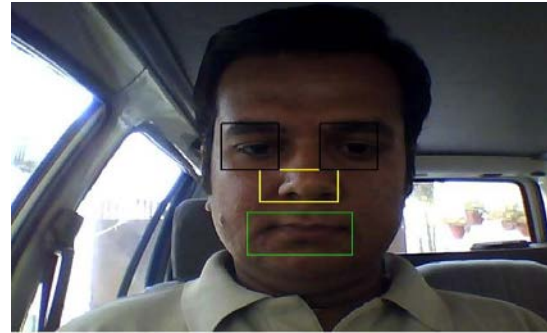


Figure 9: The highlighted search areas for mouth, nose and eyes.

4.3.1 Nose holes

Finding nose holes in an area given from face's geometry depends on the angle between camera and face. If there isn't a direct line of sight between nose holes and camera, it is obviously not possible to track them. Nose holes colors have a significant saturation, depending on its color black. The threshold must be defined and over geometry or clustering two centers of saturation can be found.

4.3.2 Mouth

Detecting the middle of the mouth isn't as simple as it is thought. There are a lot of possibilities, going over gradient horizontal and/or vertical decent, hue or saturation. At the moment it is implemented utilizing the distinct hue of lips. Light reflects on lips and this point is fetched by a defined hue value. In contrast to the methods, this method is not light independent, thus intensity and direction of the light can influence results. A better method should be included in the future.

4.3.3 Eyes and pupils

A lot of ways can be developed to find pupils in the given area surrounding the eyes. It can also be done using hue or saturation, which leads – controlled conditions given - to good results, but it highly depends on the current light situation. Different pupils were used for testing and the best results were gained by pupils directly from the tester, which was not really surprising. Obtaining them is not that simple though. An algorithm from Anirudh S.K. called “Integrodifferential operator” [7] was used, which requires too much calculation time to be used in real-time environments, but is fast enough for getting the first pupils, so it takes time only first time for finding the Eigen pupil. Once the Eigen pupil is there the rest is very simple. But

sometimes due the lack of lighting conditions it takes a little longer time to find the pupil. Still this algorithm is very accurate and the it perform incredibly well under very low lighting condition. This feature of operating under low lighting conditions is very useful in our system. Then for tracking the Eigen pupil found by the Integro differential operator, there are few steps:

- The face is detected from the image.
- In the detected face the geometric region of the eye is defined.
- The Eigen pupil is resized according to the size of the newly found eye region.
- Then Eigen pupil is run as mask on the whole eye region and the men difference is found at every point.
- The point which gives the minimum mean difference is defined as the pupil of the eye.

This method of finding the pupil is very quick and easy. It takes very less time as compared to the Integrodifferential operator.

4.4 Detection of Eyes Status

To detect the eyes are whether open or close was a quite challenging task. Many different approaches were implemented for the detection of the status of the eyes. Some of them are listed below:

4.4.1 Status Determination using edge detection

In this novel and simple approach, the eye region from the detected face is subjected to the edge detector. The advantage of using MATLAB is that it has built in edge detector. so I applied “Canny Edge detector” and “Sobel Edge detector”. The results of both the edge detectors are presented below. The Canny Edge detector ha lot of noise and it makes some extra edge. On the other hand the Sobel Edge detector gives better results in case of the noisy image. Both the edge detectors provide the binary output.

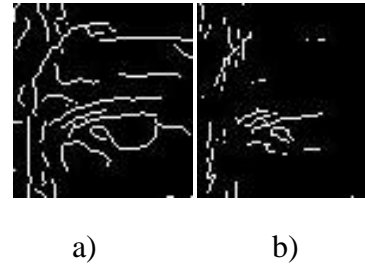


Figure 10: Eye edge diagram using a) Canny Edge Detector and b) Sobel Edge detector.

The eye status detection is performed by calculating the sum of the binary image. As the sum of the image will be higher in case of open eyes, because there are more edges than the closed eye. The problem arises when the lighting conditions and the background interfere with the image. It results in the change in number of edges.

4.4.2 Eye status detection using correlation

The in correlation approach the eye region is correlated with the previous eye region. The result will be different in the case of change of status of eye. It was implemented with the built-in function of 2D correlation in MATLAB. But the positioning of eye in each frame and the external factors affect the correlation results. The experimental results show that this system is also not very good for the implementation.

4.4.3 Proposed Method for Eye Status Detection

Initially the average intensity of every x coordinate is calculated. These values are separately calculated for each eye. The plot of the average values show that there are two prominent changes of intensity. One is the upper edge and other is the eyebrow. The face position of the eye is detected using these two values. First two intensity changes found in the above steps gives us information whether the eye is closed or not. The distance between x coordinates of the intensity is larger in case if eyes are open, compared to smaller distance in the case of opened eyes.

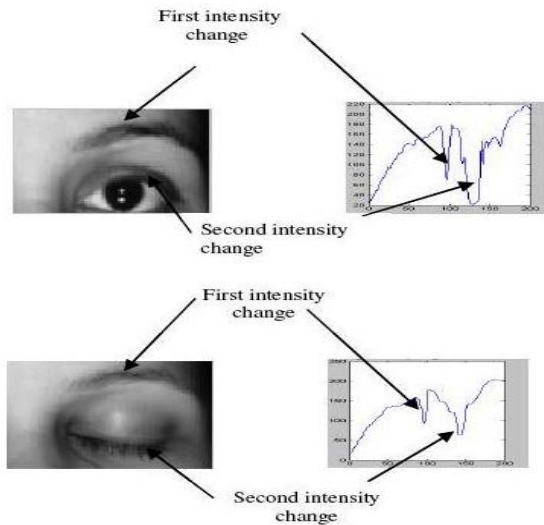


Figure 11: Intensity variation on the face

4.5 Drowsiness detection

The eye status defines the drowsiness state of the driver. If the eyes are closed then the driver is fatigued and need to take some rest. Some of the systems don't warn the driver on the first closed eye is detected, but for a vehicle moving at a high speed the closed eye even for a moment can be very crucial. Because if the vehicle is moving on 100 Miles/Hour, which means its 44.7 Meters/Second. So if we waste a second or so that could prove to be lethal.

4.6 Results

The result of the tracking is as:



Figure 12: Tracking for eyes open.



Figure 13: Tracking for eyes closed.

For the Figure 12 and 13 we can see that incase of eyes closed the pointer does not exactly points to the eyes. This is because the use of matching algorithm for the eyes tracking. It will match it to the eye brows, because they also high concentration of dark pixels. For drowsiness detection the results of the eyes status algorithm is very accurate.

5 Future Recommendations

System can be improved in many dimensions some of which are discussed including The other parameters of the driver like yawning etc. should be included to get the better vigilance status of the driver. The algorithm should be made at night with low lighting, as the light from the inside light unit of the car. The warning system should be modified to either stop the car slowly or make some vibrations to wake up the driver.

References

- [1] Y. Elzohairy, "Fatal and Injury Fatigue-Related Crashes on Ontario's Roads: A 5-year Review", Working Together to Understand Driver Fatigue: Report on Symposium Proceedings, February 2008.
- [2] Dingus, T.A., Jahns, S.K., Horowitz, A.D., and Knipling, R. (1998). Human factors design issues for crash avoidance systems. In W. Barfield, and T.A. Dingus (Eds.), Human factors in intelligent transportation systems. (pp. 55-93). Mahwah, New Jersey: Lawrence Erlbaum Associates.
- [3] "Pakistan Car Crash Accidents" available online at: <http://www.caraccidents.com/country-car-accidents/pakistan-car-accidents-crash.html>
- [4] A. Rechtschaffen. "Current perspectives on the function of sleep". Perspectives in biology and medicine, vol. 41, no. 3, pages 359-390, 1998.
- [5] S.B. Klein & B.M. Thorne. Biological Psychology. Worth Pub, 2007.
- [6] Albert Kircher, Marcus Uddman & Jesper Sandin. Vehicle control and drowsiness.

- Tech. Rep. VTI-922A, Swedish National Road and Transport Research Institute, 2002.
- [7] DaimlerAG. The electronic drawbar, June 2001. <http://www.daimler.com>.
- [8] H. Ueno, M. Kaneda & M. Tsukino. Development of drowsiness detection system. In Proceedings of Vehicle Navigation and Information Systems Conference, pages 15–20, 1994.
- [9] A. Kircher, M. Uddman, and J. Sandin, “Vehicle control and drowsiness,” Swedish National Road and Transport Research Institute, Tech. Rep. VTI-922A, 2002.
- [10] Anon, “Perclos and eyetracking: Challenge and opportunity,” Applied Science Laboratories, Bedford, MA., Tech.Rep., 1999.
- [11] D. Dinges. F. PERCLOS: A valid psychophysiological measure of alertness as assessed by psychomotor vigilance. Tech. Rep. MCRT-98-006, Federal Highway Administration. Office of motor carriers, 1998.
- [12] Y. Matsumoto & A. Zelinsky. An algorithm for real-time stereo vision implementation of head pose and gaze direction measurements. In Procs. IEEE 4th Int. Conf. Face and Gesture Recognition, pages 499–505, mar 2000.
- [13] SmartEyeAG. AntiSleep, 2009. www.smarteye.se.
- [14] W. Shih & Liu. A calibration-free gaze tracking technique. In Proc. 15th Conf Patterns Recognition, volume 4, pages 201–204, Barcelona, Spain, 2000.
- [15] Qiang Ji & Xiaojie Yang. Real-time Eye, Gaze and Face Pose Tracking for Monitoring Driver Vigilance. Real-Time Imaging, vol. 8, pages 357–377, Oct 2002.
- [16] T. D’Orazio, M. Leo, C. Guaragnella & A. Distanto. A visual approach for driver inattention detection. Pattern Recognition, vol. 40, no. 8, pages 2341–2355, 2007.
- [17] P. Boyraz, M. Acar & D. Kerr. Multi-sensor driver drowsiness monitoring. Proceedings of the Institution of Mechanical Engineers, Part D: Journal of Automobile Engineering, vol. 222, no. 11, pages 2041–2062, 2008.
- [18] <http://www.a4tech.com>
- [19] www.sleephomepages.org
- [20] emeredesbenz.com
- September, 2007. His areas of interest are Digital Image Processing, and Computer vision.



Muhammad Jibran Is providing technical assistance as R4 NSS Core Engineer in Core Network Operations regarding Operation and Maintenance of different Core Entities in Ericsson UAE. He obtained BSc degree in Software Engineering from University Of Engineering and Technology Taxila, Pakistan in October 2006. His areas of interest are Digital Image processing, video Summarization and Data communication.



Muhammad Fahad Khan Is working as a Instructor in Software Engineering Department at University of Engineering & Technology Taxila, Pakistan since September, 2007. He obtained Masters degree in Computer Engineering from the University of Engineering & Technology Taxila, Pakistan in January, 2010. He graduated in Software Engineering from University of Engineering & Technology Taxila, Pakistan, in