

## Causes of Driver Distraction in-vehicle : A Survey

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**Abstract** -Drivers with fatigue have perpetually been a serious drawback in several accidents. The driving force assumes that it's necessary to spot the item ahead of the vehicle and to know the causes of accidents and issues in vehicle management. Currently that traffic conditions are increasing, the likelihood of accidents in an exceedingly giant space is accrued as a result of the driving force is unable to sight the sort of fatigue they're experiencing whereas driving. This paper shows the driving force distraction detection survey. Driver distraction could be a hazard state that may simply result in traffic accidents. This study focuses on detective work the driver's distraction standing supported eye behavior (glass detection and blinking), arm position (right arm up, down, right forward), head orientation, and facial features. Every module shall give relevant data for the assessment of driver basic cognitive process.

**Key Words:** Driver Distraction, Fatigue Condition, Driver Inattention

### 1. INTRODUCTION

Distracting driving is one among the most causes of auto crashes. As per the statistics discharged by the National Road Safety Administration (NHTSA), 3092 folks were killed and 416 000 out of action in vehicle accidents involving distracted drivers in 2010[1]. Driver diversion is described as a type of inattention that "delays the identification of the information needed to perform the driving task safely because some incident, operation, item or individual inside or outside the vehicle compels or induces the driver to move his attention away from the driving task" [2]. In specific, with the widespread use of electronic route navigation systems, navigation map-viewing behavior is turning into a vital supply of driving distraction and vehicle accidents. A significant safety issue has drawn the eye of the many researchers to distracted driving performance related to map-viewing behaviors and navigation systems. The corresponding human factors analysis of such systems is

Reducing the danger of driving distractions and avoiding venturesome things, eye movement trailing, video camera recognition, and lane retention square measure fashionable techniques employed in current driver distraction studies. These ways square measure effective in classifying distraction and non-distraction periods. However, there square measure limitations within the detection of navigation-related distractions in time period.

From basic analysis & surveys, it's thought-about that, within the case of continuing driving, the driving force is tired when 2-3 hours and thus the steering performance is deteriorated. The driver's temporary state is higher at the hours of darkness, when lunch, in the afternoon, compared to other times in the day. Alcohol, drugs, too, are the reasons for the loss of concentration of the driver. Throughout the world, many countries have presented their different statistics on accidents that have occurred due to driver somnolence and distraction. Usually, around 20%-30% of accidents happen due to driver somnolence and distraction. Fig. 1 The reason for the driver's distraction while driving is shown.

believed to contribute to the event of safe, usable and acceptable help systems for vehicle customers.

**Fig -1:** Example of driver distraction[3].





**Fig -2:** Example of Driver Fatigue[3].

It is understood, according to valuable studies, that the lack of attention of the driver is the main cause of many accidents. Reducing accidents and improving public safety has become an important objective of the research system. Considering the safety instructions, the driver's response analysis becomes the current ground in the investigation. Recent studies have shown that the driver's eyes have recently been used as driver workload metrics. Because of the complexity and variations of the light intensity involved. The number of road accidents worldwide has steadily increased in the last few years. It is therefore important to study the causes of driver distractions.

## 2. RELATED WORKS

This section summarizes some of the relevant and important literature work on distracted driver detection.

This paper states that road accidents are caused by risky drivers who drive vehicles very rudely on the roads. Risk drivers are the people that drive vehicles at very high speeds on all roads. To avoid accidents caused by these people, there's a mechanism during this system for measuring the speed of various vehicles. The mechanism is predicated solely on the speed of the vehicle and its statistic . By making use of this technique , the speed is measured and therefore the speed range is decided . The analysis is then made supported the knowledge retrieved. Later on, finding the risky drivers, they're punished. The advantage of this is often to scale back the amount of accidents occurring on the roads. Therefore, risky drivers are recognized during this way[6].

The driver's basic cognitive process may be a major think about road crashes. The driver's basic cognitive process is split into 2 classes, distraction, and fatigue. The common definition of driver basic cognitive process is "decreased

attention to activities that ar important to safe driving within the absence of competitive activities." this technique is principally accustomed cut back road accidents and therefore the driver's basic cognitive process system is additionally being put in by prime motor vehicle firms like Toyota, Nissan, and Volvo, etc. This mechanism measures numerous factors, like biological measures for drivers, physical measures for drivers and performance measures. this is often helpful for analysis and is additionally accustomed cut back the distraction of the driving force whereas driving. Driver basic cognitive process in the main will increase the danger of driving. The advantage of this technique is principally to cut back accidents caused by the distraction of the driving force. The Disadvantage of the system is that we tend to conjointly got to think about natural disasters like climatic conditions, road type, and traffic density. As a result, a brand new system are projected to beat the higher than problem[7].

UCSD's Intelligent and Safe cars Laboratory has created a major contribution during this space however has proscribed solely three styles of distractions. Radio adjustment, mirror adjustment, and operative gear. It's Martin et al.[8] given a vision-based analysis framework that acknowledges in-vehicle activities victimisation two Kinect cameras that provided the motive force with front and back views to supply "hands on the wheel" data Ohn-bar et al.[9] projected a merger of classifiers wherever the image is to be segmental into three regions: the wheel, the gear and also the panel to infer actual activity. They conjointly given a regionally primarily based classification approach for the detection of the presence of hands in bound pre-defined regions within the image[10]. for every region, a model was learned one by one and joined employing a second-stage classifier. Authors have extended their analysis to incorporate eye cues to antecedently existing head and hand cues[11]. However, solely three styles of distractions were still thought of.

Road detection is of high importance to the warning of road departures and to support the detection of pedestrians. In this road detection system, image control is used; the road detection algorithm supports algorithms such as the illuminated-invariant features of space and the probability-based classifier. The incoming images are analyzed by light-invariant space, and the design of the road is estimated. The results will be shown in a map format. The result is presented in the form of a small hole from the middle to the bottom of the images. The road detection evaluation is validated by an illuminated-invariant algorithm. It performs in three rows, in the first row it contains the original images, in the second row it contains the illuminated-invariant image, in the third row it contains the path detected, and finally, in the bottom row, it shows the comparison with the hand-segmented results shown

in (Yellow). Properly classified pixels (green) have been falsely detected by road pixels (red)[12].

Accidents are not only due to Overspeed, driver distractions, or drunk driving. Abnormal conduct of the driver will also lead to accidents and pose a risk to both the public and the driver. To avoid an accident caused by the abnormal behavior of the driver, this paper introduces a mechanism for the detection of abnormal driving. This mechanism is based on the abnormal behavior of the driver by analyzing normal driving behavior. It also uses a driver model, driving style for detection and analysis. It has proposed a system called an abnormality index, which is used in the analysis of normalized driving behaviors to evaluate the abnormality. This is mainly done with the help of ordinary vehicle data such as brake pedals, gas, etc. The advantage of using this system is that the number of accidents occurs mainly based on abnormal driving. As a result, abnormal driving is recognized. However, this system also has limitations on longitudinal driving due to the limitation of vehicle test data and will be enhanced in the future[13].

In today's world, in-vehicle information (IVIS) is used by everyone, such as cell phones, satellite radios, navigation and driver distraction, which are important aspects of safety concern. According to the Survey, between 13 and 50 percent of accidents occurred due to the distraction of the driver resulting in 5,000 fatalities and \$40 billion in damages each year. As a result, a mechanism was introduced to overcome the problem of driver distraction. In this system, a strategy called Support Vector Machines (SVM) is a data mining method used to detect distractions by driving eye movements and performance. The data was trained and tested in both the SVM and the logistic regression model. This system tracks the movement of the eye and measures the number of distractions. The main advantage is that it is used to reduce the occurrence of accidents and ensure safe driving. The disadvantage is difficult eye-tracking, which may lose accuracy when vehicles travel on rough roads or under lighting conditions. These drawbacks will, therefore, be increased in the future[14].

In spite of driver behavior and driving behavior, other studies used physiological sensors to detect driver distraction and other abnormal conditions. According to the study in[15], driver monitoring systems for sleepiness and distraction detection can be classified into visual and non-visual methods. Visual-based methods to monitor driver head pose, eye movement and blinking, yawning, and facial expression. Non-visual based systems, on the other hand, detect driver status with physiological sensors such as EEG, ECG, and EOG, along with vehicle CAN bus signals. However, the effects of hand, arm, and body on the recognition of driver status have not been discussed. The stress detection system for drivers has also been studied in[16]. A specific type of continuous recurrent neural network called a neural cellular network was used for a binary classification task.

When drivers perform secondary tasks while driving, they are considered to be distracted, and many studies use eye-off-road duration to detect whether the driver is distracted by secondary tasks. The most common features for driver distraction detection are head pose and eye gaze information. In addition to driving behavior, vehicle information such as vehicle speed, steering, and acceleration are important features for assessing the level of driver distraction. An integration method combining hand, head, and eye for driver activity recognition was proposed in[ 17]. Rezaei and Klette have introduced an intelligent driver assistance system to prevent rear-end bass crashes[18]. The head pose was estimated based on the proposed face appearance model and 3D head model mapping. In[ 19], the driver's sleepiness alert system was proposed based on the driver's head and eye dynamics. The driver head pose was estimated based on an Euler angle comparison between a single head region image and a 3D head model with known rotations.

### 3. DATASET DESCRIPTION

The dataset created by Aboulenaga et al.[4] is employed during this paper. the info set consists of 10 categories. Safe driving, texting on mobile phones with the correct or the paw, talking on mobile phones with the correct or the paw, adjusting the radio, uptake or drinking, hair and make-up, reaching behind and speech the rider. Sample pictures of every category within the dataset ar shown in fig. 3. knowledge were collected from 31 participants from seven completely different|completely different} countries exploitation four different cars and enclosed many variations in driving conditions and driving conditions. for instance, drivers ar exposed to completely different lighting conditions, like daylight and shadows.



Fig -3: Ten Classes of Driver Postures from the Dataset

Among all the states in Asian country, Madras is that the highest accident rate. In 2013, a survey showed that fifteen,563 fatalities and fourteen,504 accidents occurred in Madras, the best accident rate in Asian country. there's an inventory of additional accidents within the state from 2002 to 2012. in line with the

report, 2 things may cause associate accident: one is that the technology and also the different is driving and driving and different health issues for the driving force. In India, seventieth of the prevalence of death inadvertently, in war. the amount of vehicles exaggerated from eighty two lakhs in 2007 to one.6 crores in 2012 and also the road infrastructure has not modified considerably as a results of this accident. A survey reported that roughly eight accidents occur each hour and fifteen % of accidents occur in each country. 9.663 accidents were recorded in metropolis by the National Crime Bureau[5].

The survey reported the number of accidents that have occurred in Tamil Nadu from 2000 to 2016 id given below [5].

**Table -1:** Survey report

| YEAR | Total No. of accidents | Total no. of peoples killed | Total no. of peoples involved |
|------|------------------------|-----------------------------|-------------------------------|
| 2000 | 48923                  | 9300                        | 62706                         |
| 2001 | 51978                  | 9571                        | 63853                         |
| 2002 | 53503                  | 9939                        | 65069                         |
| 2003 | 51025                  | 9275                        | 64517                         |
| 2004 | 52508                  | 9507                        | 66790                         |
| 2005 | 53878                  | 9706                        | 71727                         |
| 2006 | 55145                  | 11009                       | 75350                         |
| 2007 | 59140                  | 12036                       | 83135                         |
| 2008 | 60409                  | 12784                       | 83035                         |
| 2009 | 60794                  | 13746                       | 84250                         |
| 2010 | 64996                  | 15409                       | 90854                         |
| 2011 | 65873                  | 15422                       | 89667                         |
| 2012 | 67757                  | 16175                       | 94523                         |
| 2013 | 66238                  | 15563                       | 91244                         |
| 2014 | 67250                  | 15190                       | 92915                         |
| 2015 | 67250                  | 15642                       | 95343                         |
| 2016 | 71431                  | 17218                       | 99381                         |

#### 4. CONCLUSION AND FUTURE WORK

A literature survey on the security of driving accident detection for intelligent vehicles is conferred during this paper. The survey is especially supported a study of however accidents area unit detected and prevented. This additionally ensures the security of each the motive force and therefore the public. The survey mentioned varied solutions like collisions, congestion, road accidents and additionally provided temporary info on the driver's assistant system. Driver distraction may be a significant issue that has diode to an outsized variety of road crashes worldwide. Detection of the distracted driver is, therefore, a necessary element of the system in self-driving cars. Also, within the future, develop a system that may discover visual and psychological feature distractions furthermore as manual distractions.

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