

Detection of Abnormal Driving Behaviors Patterns with Advanced Notifications

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Abstract

Road traffic injuries are the main source of death among individuals matured somewhere in the range of 15 and 29 years. Without supported activity, road traffic crashes are anticipated to end up the seventh driving reason for death by 2030. The greater part of the drivers are pompous and ignorant of their rash driving habits. Also, some strange driving practices are unapparent and are barely noticeable by drivers. To destroy the unsafe driving habits and to guarantee safe driving, the driver's conduct ought to be observed. In this paper, we will investigate some examination works, which screen the drivers' conduct. Driving conduct investigation is additionally a well-known course of Smartphone-based vehicular applications. Be that as it may, existing chips away at driving practices recognition utilizing dataset camera recordings can just give a coarse-grained result utilizing limits, i.e. recognizing anomalous driving practices from ordinary ones. Since limits might be influenced via auto compose and sensors' affectability, they can't precisely recognize the distinctions in different driving personal conduct standards.

Keywords

Sensing, Smartphone, Notifications, IMU, Data, Behavior, SenSpeed, VANET.

I. Introduction

(WHO), traffic accidents wound up one in every one of the best 10 driving reasons for death among the planet. In particular, traffic accidents asserted about 3500 lives on an ordinary in 2014. Studies demonstrate that the majority of traffic accidents zone unit caused by human components, e.g. drivers' anomalous driving practices. Along these lines, it's a need to search out drivers' irregular driving practices to caution the drivers or report Transportation Bureau to record them. in spite of the fact that there have been works [1] on strange driving practices discovery, the most spotlight is on police examination driver's standing bolstered pre-conveyed foundation, similar to liquor gadget, infrared gadget, and cameras, that bring about high establishment worth. Since dataset camera recordings have gotten expanding popularities over the ongoing years and blending in our day by day lives, a lot of and a lot of informational collection camera video-based transport applications region unit created in the savvy office. Driving conduct examination is to boot a well-known heading of Smartphone-based transport applications. In any case, existing works [2] on driving practices location utilizing informational index camera recordings can only give a coarse-grained result utilizing limits, i.e. trademark anomalous driving practices from old ones. Since edges are also stricken via car kind and sensors' affectability, they'll not precisely recognize the varieties in changed driving standards of conduct. Accordingly, those arrangements can't offer fine-grained recognizable proof, i.e. recognizing certain kinds of driving practices. Moving toward this path, we'd wish to consider a fine grained irregular driving practices perception approach, that utilizations Smartphone sensors to not solely acknowledge strange driving practices but

rather to boot affirm certain kinds of the driving practices while not requiring any further equipment. The fine-grained irregular driving practices perception is in an exceedingly position to flavor up drivers' familiarity with their driving habits as a large portion of the driver's territory unit arrogant and not mindful of their foolhardy driving habits to boot, some unusual driving conduct region unit unrealistic and straight forward to be unperceived by drivers. On the off chance that we will affirm drivers' irregular driving practices consequently, the drivers know about their unfortunate driving habits, so they will adjust them, serving to prevent potential car accidents in accordance with [3]. The issue of forceful driving seems, by all accounts, to be expanding in reality. Forceful Driver conduct indicates heedless driving rules which include thought of a standard method for disposition happen to finish in a brief period as well as aim. Generally, drivers are incognizant that they give possibly forceful action ordinary. As a foreseen result that is intended is difficult to turn up area with arranged require the foundation of undertone be met every now and again forceful driving charge as rash hasty. Conceivably forceful driver conduct is directly a main source of traffic in a city zone. Driving conduct let in the meaning of harsh or forceful driving could result from animosity, self-centeredness, or rivalry. As consequently of this individuals are experience the road as a more perilous territory. The most lethal factor is human mistake. This incorporates ignorance of traffic principles and roadway condition; absence of driving aptitudes; misguided thinking; and in particular, forceful driving. The primary goal of this examination is to recognize forceful driving practices and underline their impact on traffic security.

II. Related Work

"Relating Random Vector and Random Finite Set Estimation in Navigation, Mapping and Tracking", Keith Y. K. Leung, Member, IEEE, Felipe Inostroza, Student Member, IEEE, and Martin Adams, Senior Member, IEEE has presumed that The detailing of route, mapping, and following issues utilizing the RFS system can be identified with the RV plan, just through the application of perfect identification conditions, which expect no messiness, a solitary and right stage of information affiliation and a known guide cardinality. Conversely, the RFS approach thinks about every single conceivable change of information affiliation while representing location insights, and obscure guide cardinality. This is genuine for Bayesian separating, as well as ML cluster estimation. An attainable answer for RFS-based cluster estimation is anyway still an open issue and in this manner the discoveries were approved utilizing Bayesian separating approaches. The RFS based PHD-SLAM calculation was appeared to perform likewise to the RV based FastSLAM and MH-FastSLAM calculations under close-toideal discovery conditions. At the point when conditions moved toward becoming non-perfect with low probabilities of location and high mess, PHDSLAM kept on creating gauges with lower mistakes, while the assessments from the RV approaches wandered. This is because of their verifiable presumption of perfect recognition conditions, and the conceivable predisposition

caused by the usage of a solitary information affiliation theory. It ought to likewise be noticed that the PHD channel is the least complex estimation of a RFS Bayes channel. Further developed RFS channels, for example, the CPHD and the MB channels, are required to yield enhanced outcomes. Specifically, the LMB channel has been executed for taking care of the SLAM issue. Through the understanding acquired in this, the less difficult RV definition can be fruitful when discovery conditions are near perfect, where information affiliation is probably not going to fall flat. Be that as it may, in reasonable instances of highlight misdetections and high mess, the more mind boggling RFS structure gives a more powerful methodology. Wearable Mobile-Based Emotional Response-Monitoring System for Drivers Boon Giin Lee, Member, IEEE, Teak Wei Chong, Boon Leng Lee, HeeJoon Park, Yoon Nyun Kim, and Beomjoon Kim has proposed a negative passionate reactions are a developing issue among drivers, especially in nations with substantial traffic, and may prompt genuine accidents on the road. Estimating pressure and weakness instigated enthusiastic reactions by methods for a remote, wearable framework would be valuable for conceivably turning away roadway tragedies. The focal point of this investigation was to create and check an enthusiastic reaction observing worldview for drivers, got from electromyography signs of the upper trapezius muscle, photoplethysmography signs of the ear cartilage, and additionally inertial movement detecting of the head development. The significant sensors were associated with a microcontroller unit outfitted with a Bluetooth-empowered low-vitality module, which permits the transmission of those sensor readings to a cell phone progressively. A cell phone application was then used to separate the information from the sensors and to decide the driver's flow feeling status, by means of a prepared help vector machine (SVM). The enthusiastic reaction worldview, tried in ten subjects, comprised of 10 min pattern, 5 min prestimulus, and 5 min post boost estimations. Enthusiastic reactions were ordered into three classes: loose, pushed, and exhausted. The examination coordinated an aggregate of 36 highlights to prepare the SVM display, and the last improvement results uncovered a high exactness rate (99.52%). The proposed wearable framework could be connected to an insightful driver's security ready framework, to utilize those passionate reactions to counteract accidents influencing themselves or potentially other honest casualties. "Continuous Driver Drowsiness Detection System Based on PERCLOS and Gray scale Image Processing", Jun-Juh Yan, Hang-Hong Kuo, Ying-Fan Lin, Teh-Lu Liao proposed This examination builds up an ongoing tiredness location framework based on dark scale picture handling and PERCLOS to decide whether the driver is exhausted. The proposed framework involves three sections: first, it ascertains the inexact position of the driver's face in grayscale pictures, and afterward utilizes a little format to break down the eye positions; second, it utilizes the information from the past advance and PERCLOS to set up an exhaustion show; lastly, based on the driver's close to home weakness display, the framework ceaselessly screens the driver's state. Once the driver displays weariness, the framework cautions the driver to quit driving and take a rest. They inferred that it is a realtime, dark scale reenactment framework to distinguish driver languor by picture handling. In testing and results, and based on the exhaustion show, the framework can help screen the drivers' physical state, and remind drivers on the off chance that they are worn out, which they themselves may not took note. The greatest distinction between financially accessible items and the framework proposed in this examination is the utilization of dark scale pictures, which implies that recognition of skin shading isn't

required. Despite the fact that the proposed framework includes extra computation steps, it requires less memory and could be connected in various natural conditions. For instance, it could be utilized even

III. Detection Using Smartphone Sensors

To eliminate the necessity of pre-deployed infrastructures and further hardware, recent studies have confidence exploitation smartphones to note abnormal driving behaviors. Especially, [4] uses accelerometers, magnetometers and GPS sensors to figure out whether or not or huge motorcycle maneuvers or accidents occur. [5] Uses accelerometers, gyroscopes, and magnetometers to estimate a driver's driving vogue as Safe or Unsafe. [6] Use accelerometers to note drunk driving and sudden driving maneuver, severally. The works area unit similar in this they perform a coarse-grained driving behavior detection that uses some thresholds to hunt out abnormal driving behaviors. All identical, thresholds may be jam-packed with automobile type and sensors' sensitivity so as that they'll not accurately distinguish the variations in varied driving activity patterns. Therefore, none of the existing works can perceive fine-grained identification. Our work uses Smartphone sensing and machine learning techniques to understand a finegrained abnormal driving behaviors detection and identification. Though' machine learning technique already is utilized to some activity recognition work [5], our work is first to identify driving activities exploitation machine learning technique. In [7], since activities area unit are instant, the pattern of activities is simple. Therefore choices of activities' pattern would be legendary merely. However, in real driving environments, since the period of some driving behavior is long, not instant, like Weaving, the system ought to be compelled to verify the beginning and ending of the driving behavior first. Extracting and selecting effective choices of each type of abnormal driving behavior would be lots of sophisticated.

A. Driving Behavior Characterization

In this, we've got a bent to first describe the data assortment technique for driving behavior samples from real driving environments. Then we've got a bent to research patterns of each type of driving behavior from Smartphone sensors' readings.

B. Collecting data from Smartphone Sensors

We develop degree App to collect readings from the 3-axis measuring device and thus the 3-axis orientation detector. We have got a bent to align the two coordinate systems inside the Smartphone and inside the vehicle by making the accelerometer's axis on the moving direction of the vehicle. Therefore, we've got a bent to may monitor the vehicle's acceleration and orientation by retrieving readings from the Dataset camera video measuring device and orientation detector [8].

IV. Use of Smartphones to Detect the Driving Style

The smartphone could be used to warn for accidently lane-changes as seen in the work of recognize driver aggressiveness. There are smartphone systems which work as black-boxes in case of accidents [5] and inform other traffic participants about the accident so that they could avoid possible traffic jams [4]. Information could be shared among mobile phone owners in order to optimize their speed and avoid stop-and-go situations. Such information is gathered from traffic signals and cameras and distributed to the drivers [5]. Only the accelerometer readings taken from the phone is used to predict and classify the driving behavior. They

differentiate between vehicle conditions (speed and shifting), driving patterns (acceleration and deceleration, lane change) and road conditions. The GPS readings could be very efficient and could provide a more reliable speed information compared to the speed computed from the accelerometer readings, and the radius of curvature of the vehicle moving path could be used for recognizing certain types of driving behavior [6]. Nonetheless its localization accuracy represents a problem because of the localization error at the magnitude of several meters which may occur due to GPS. Moreover, the GPS consumes much more energy than the other sensors in the phone and thus leads to higher battery drain [9]. A future area also discussed in the work is that phones equipped with a camera could be very efficient during the drive since they could capture or mark road signs and follow the drivers' sight line. However, their high energy consume, complicated algorithms and "intensive computations for the limited computation capability of mobile phones" lead the authors to a more simple but also effective style for driving detection based on the accelerometer and orientation sensor of the phone. The light sensor in the phone to obtain information about the environment in which the car is moving, because the brightness directly affects the visibility of the driver and this influences his anticipation. Another novel method in the work of Magana is the weather information involved in estimating the driving behavior. This information is obtained from the Internet connection of the smartphone. A smartphone application which uses the information from the embedded sensors and the vehicles state information acquired from the vehicles CAN bus (speed, fuel consumption, GPS, etc.). The gathered data is passed to a fuzzy-based module which analyzes the data and classifies it and then an suggestion is presented to the driver how to optimize the fuel energy consumption/driving behavior. Langari, Murphey and Holmén attempt to classify the driving style by using "the ratio of the standard deviation and the average acceleration extracted from the acceleration profile within a specified window" [7]. A fuzzy rule classification was made referring to a conclusion made by the typical average acceleration ranges in a city are different for the various driving styles. The different driving styles according to the measure how fast the driver accelerates and decelerates. The developed algorithm extracts jerk features from the current vehicle speed within a short time-window, and classifies the current driving style into three categories: calm, normal and aggressive, by comparing the extracted jerk feature with the statistics of the driver styles on the current roadway. Few papers use dynamic time warping to detect repeating patterns of driving behavior.

V. Proposed System

As soon as the driver enters the vehicle, it would be mandatory for him to login into the system.. This login process will be biometric, to ensure the authenticity. His profile will then get activated in the monitoring system. All the data related to the driving is gathered by the virtue of various sensors such as : accelerometers, gyroscope, wind sensor, road surface sensor, humidity sensor, GPS sensor and camera. The weather conditions are taken into consideration using wind sensor, road surface sensor, humidity sensor etc. The use of camera would provide an added advantage by letting us know about the vehicle's surroundings. Road signs can also be identified using the camera. Information regarding the current traffic condition at vehicle's location can be obtained by sending the GPS location to the Google Maps. The amalgamation of all the above mentioned sensors would give us an insight into the vehicle's speed, orientation, acceleration, turning, braking,

surroundings etc. All the parameters would be compared with the historical database, and decision will be made whether the driving is risky or not.

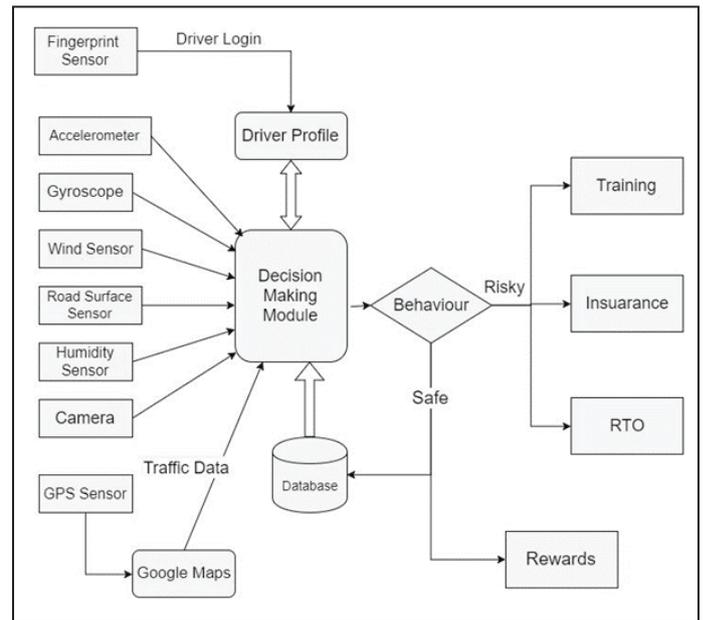


Fig. 1: Proposed Architecture Diagram

Algorithm

KNN Algorithm: In k-NN classification, the output is a class membership. An object is classified by a majority vote of its neighbors, with the object being assigned to the class most common among its k nearest neighbors (k is a positive integer, typically small). If k = 1, then the object is simply assigned to the class of that single nearest neighbor.

K-nearest neighbors KNN algorithm:

1. Determine parameter K = number of nearest neighbors
2. Calculate the distance between the query-instance and all the training samples
3. Sort the distance and determine nearest neighbors based on the K-th minimum distance
4. Gather the category y of the nearest neighbors
5. Use simple majority of the category of nearest neighbors as the prediction value of the query instance

KNN pseudo code:

Classify (X, Y, x) // X: training data, Y: class labels of X, x: unknown sample for i=1 to m do Compute distance d(Xi, x) end for Compute set I containing indices for the k smallest distances d(Xi, x). return majority label for {Yi where i ∈ I}

Haversine Algorithm: The haversine formula determines the great-circle distance between two points on a sphere given their longitudes and latitudes. Important in navigation, it is a special case of a more general formula in spherical trigonometry, the law of haversines that relates the sides and angles of spherical triangles.

- R is the radius of earth in meters.
- Lat₀ = latitude of origin point, Long₀ = longitude of origin point
- Lat₁ = latitude of target point, Long₁ = longitude of target point
- Difference in latitude = Lat₀ - Lat₁
Difference in longitude = Long₀ - Long₁
Φ = Difference in latitude in radians

- Λ = Difference in longitude in radians
 $O = \text{Lat}_O$ in radians.
 $T = \text{Lat}_T$ in radians.
- $A = \sin(\Phi/2) \cdot \sin(\Phi/2) + \cos(O) \cdot \cos(T) \cdot \sin(\Lambda/2) \cdot \sin(\Lambda/2)$
 $B = \min(1, \sqrt{A})$
 Distance = $2 \cdot R \cdot B$

VI. Conclusion

In this paper, we present a highly efficient mobile based rash driving detection system. The mobile phone, which is placed in the vehicle, collects and analyzes the data from its accelerometer sensors to detect any abnormal or dangerous driving maneuvers typically related to driving under alcohol influence and sends a message for help like alerting with voice message, flashing, vibration etc. We address the problem of performing abnormal driving behaviors detection (coarse-grained) and identification (fine-grained) to improve driving safety. In particular, we propose a system, to detect and identify specific types of abnormal driving behaviors by sensing the vehicle's acceleration and orientation using Smartphone sensors. Compared with existing abnormal driving detection systems, not only implements coarse-grained detections but also conducts fine-grained identifications.

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