

# A VISION-BASED SYSTEM DESIGN AND IMPLEMENTATION FOR ACCIDENT DETECTION AND ANALYSIS VIA TRAFFIC SURVEILLANCE VIDEO

Anuradha U<sup>1</sup>, Lathesh D R<sup>2</sup>, Ajay Kumar G L<sup>3</sup>, Akash H G<sup>4</sup>, Mahendra N V<sup>5</sup>

<sup>1</sup>Asst Professor, Dept. Of CSE, Sambhram Institute of Technology, Bengaluru-560097

<sup>2,3,4,5</sup>Eight Semester, Dept. Of CSE, Sambhram Institute of Technology, Bengaluru-560097, India

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## ABSTRACT

This study aims to test the entire framework on an AI demo board and investigate the issue of automatically and effectively recognizing and analyzing traffic incidents captured by surveillance cameras. To begin, the motion interaction field (MIF) method, which is capable of detecting collisions in video, is used to locate damaged automobiles based on the interactions of various moving objects. Second, the location of the destroyed vehicles is determined using the YOLO v3 model. Using a hierarchical clustering method, the vehicle trajectories prior to the collision are recovered, and the related trajectories are reconstructed. Finally, a perspective transformation is used to project the trajectory into a vertical view to help traffic cops make better decisions. The unbiased finite impulse response (UFIR) method, which does not require statistical information about the external noise, is used to estimate the vehicle's velocity. The estimated velocity and impact angle from the vertical view can then be used to investigate the traffic accident. Finally, a Huawei AI demo board called HiKey970, which was used to code all of the aforementioned algorithms, is used in an experiment to show how useful and effective the proposed method is in practice. A few mishap observation recordings are sent onto the demo board. Mishaps are recognized and the proper vehicle directions are gathered.

## 1. INTRODUCTION

Over the past few decades, the significance of utilizing technologies for traffic monitoring has increased. Crash detection is heavily dependent on human oversight at the traffic management center (TMC). Despite the fact that manual observation is frequently trustworthy, it has a few drawbacks. On the one hand, it is difficult for people to promptly recognize all traffic accidents in the city, which means that the injured may not receive adequate treatment in many instances. On the other hand, because it is difficult to obtain the trajectory and speed from surveillance footage, manual investigation of the cause of a traffic accident occasionally results in errors. Therefore, technologies that automatically recognize and analyze traffic incidents are required. Vision-based collision detection systems have advanced in three ways over the past two decades: modeling of vehicle interactions, vehicle behavior analysis, and patterns of traffic flow [1]. To imitate regular traffic patterns, the first strategy makes use of traffic restrictions derived from large data sets. When a vehicle's trajectory departs from normal patterns [5]-[7], an accident occurs. However, the lack of collision trajectory data in the real world makes it difficult to identify collisions. The second method looks at vehicle motion metrics like speed, acceleration, and the distance between two vehicles

to look for accidents [8]-[10]. This suggests that all vehicles should be watched constantly. Consequently, processing capacity typically limits the method's accuracy in a crowded traffic environment. In the third approach, vehicle interactions are depicted using the intelligence driver model [12] and the social force model [11]. Since it only detects crashes based on changes in vehicle speed, this method requires a large number of training samples, but its accuracy is limited.

## 2. LITERATURE SURVEY

Video analytics, inexactly characterized as independent comprehension of occasions happening in a scene checked by a few camcorders, has developed quickly during the most recent twenty years. In spite of this work, genuine observation frameworks are as yet unequipped for autonomously breaking down complex occasions in camera view. This is a critical issue since video takes care of from a great many observation cameras all through the world are not reviewed progressively, making them incapable for mishap, wrongdoing, or psychological oppression counteraction and relief, which are all serious issues in the present society. These feeds are right now exclusively caught to assist with post-occasion video legal sciences.

Using the visual intervention influence of pavement marking for rutting mitigation— Part II: Visual intervention timing based on the finite element simulation as made sense of and demonstrated in a buddy study, visual mediation impacts driving way of behaving, which might bring about the rearrangement of wheel tracks, stress decrease from the grouping of hub pushes, and rutting relief (Part I). A three-stage mediation strategy with enough visual intercession time might assist with lessening rutting. This study proposes an underlying improvement rate approach and fosters a rutting forecast procedure in view of a limited component model. The rutting profundity information is segmentally fit to give the rutting disfigurement rate bend, which is used to expect the mediation timings of three sorts of ordinary asphalt frameworks. SUPERPAVE asphalt is uncovered to be the latest to foster intercession, though AC asphalt is the most seasoned. The investigation likewise discovered that the higher the protection from rutting deformity, the more extended the rutting misshapening takes to arrive at the subsequent stage (fixed condition), it is delayed to imply that mediation. For a similar asphalt development, the mediation of the longitudinal incline section is sooner than that of the level slant segment. Moreover, an intercession cycle might expand the assistance life of black-top asphalt by 16-31%. Synergies of electric urban transport systems and distributed energy resources in smart cities Transportation frameworks and designs utilize the most energy inside urban communities. These frameworks have been the subject of much exploration (offices and transport). Notwithstanding, cooperative

energies between them are frequently disregarded, bringing about an inability to profit by the expected advantages of their joined coordination and the board. This work gives a direct programming model to tracking down the ideal activity and arranging of distributed energy resources (DER) in a private zone while considering electric private and public transportation frameworks, for example, electric cars and metro, into thought. As a result, the essential commitment of this study is a gander at the cooperative energies of such an organized plan. A portion of the metro's regenerative slowing down energy is expected to be saved in the batteries of electric vehicles (EVs) and utilized later for different trains or the actual EV. A few contextual analyses in light of information from a private area in Madrid and a metro line have been introduced. The got information recommend huge expense reserve funds in the entire framework, prominently a critical decrease in power uses for the metro framework.

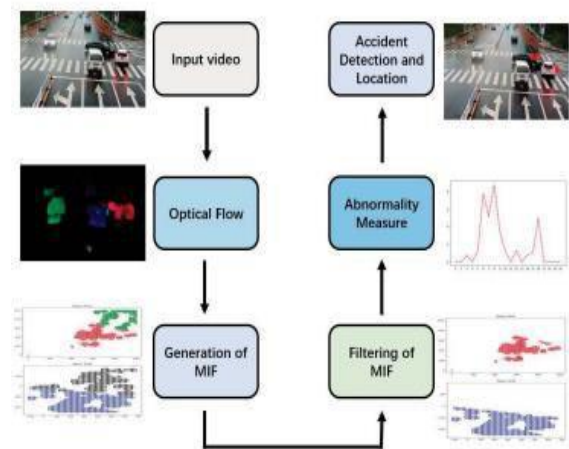
### 3.PROBLEM STATEMENT

The World Health Organization describes the road traffic system as the most complex and the most dangerous system with which people have to deal every day.

- In Existing approaches use relatively small datasets using data from only a few years or only a few roads. Indeed, it can be hard to collect all the necessary information to perform road accident prediction on a larger scale, and dealing with big datasets is more difficult.
- It is extremely tedious and time-consuming to keep watching all the day and identify congestion from the current surveillance system using in traffic monitoring hall. Furthermore, it is impossible to watch all the cameras relies on human eyes considering numerous cameras covering a large-scale region using in the freeway. However, prompt detection of the traffic congestion in large-scale region is important. Prompt detection can prevent extended congestion with devastating evolution from the initial controllable traffic congestion, which is one of the important applications in intelligent transport system (ITS). Deep learning algorithms have the potential implementation meanings to be intensely used in many fields of the transportation system, from traffic flow prediction to traffic congestion recognition. Classification of traffic condition is one of the most important parts of an ITS, which can be widely utilized in traffic control strategies, traffic flow analysis and so on. Thus, it is necessary to propose an intelligent transport system.

### 4.SYSTEM DESIGN

A few deep learning-based frameworks for distinguishing independent car crashes have been depicted. To distinguish impacts in motion pictures, these frameworks need delayed preparing with huge measures of information and utilize complex brain organizations. Notwithstanding, in light of the fact that to an absence of preparing information and high handling costs, these structures are challenging to execute practically speaking. Besides, with an ascent in the quantity of



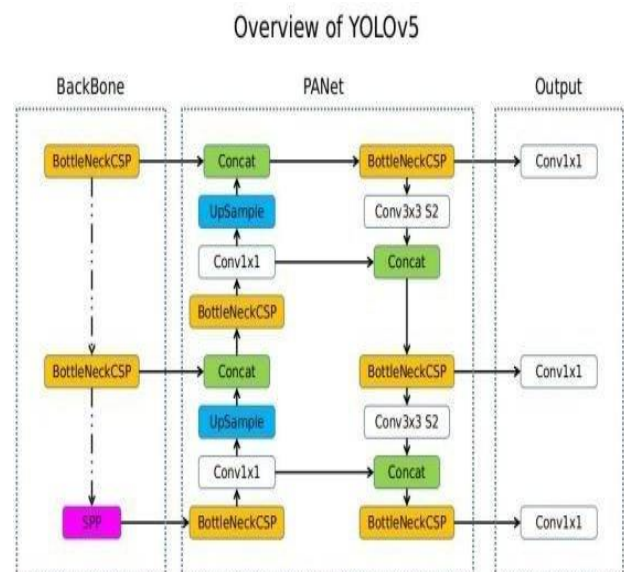
rush hour gridlock observation films, utilizing a concentrated framework to identify and examine mishaps all through the entire city is troublesome. A dispersed engineering comprised of implanted gadgets introduced in each city block is important. As a result, a lightweight structure that can run on inserted gadgets is required.

### 5.IMPLEMENTATION

#### ALGORITHMS:

##### YOLOV5:

YOLO (You Only Look Once) is a thing distinguishing proof methodology that partitions pictures into networks. Every lattice cell is answerable for recognizing things inside itself. Consequences be damned is one of the most notable item distinguishing proof strategies because of its speed and exactness. For elite execution object discovery, YOLO (You Only Look Once) models are used. YOLO partitions a picture into matrices, every one of which recognizes objects inside it. They may be utilized for constant article ID in view of the information streams. As a Convolutional Neural Network Scheme, the YOLOv5 Architecture (CNN). The key components are the Backbone, Neck, and Head. In the Backbone, CSPNet is utilised to extract features from pictures used as input images. The Pyramid feature is created with the Neck.

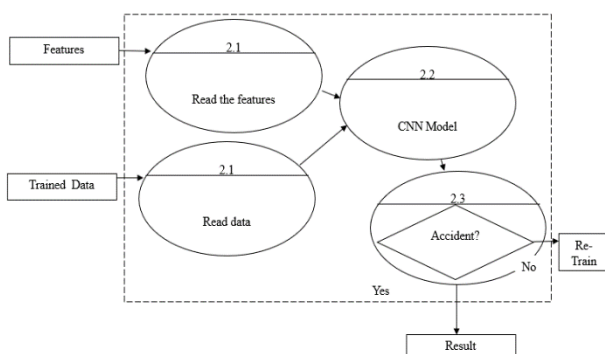
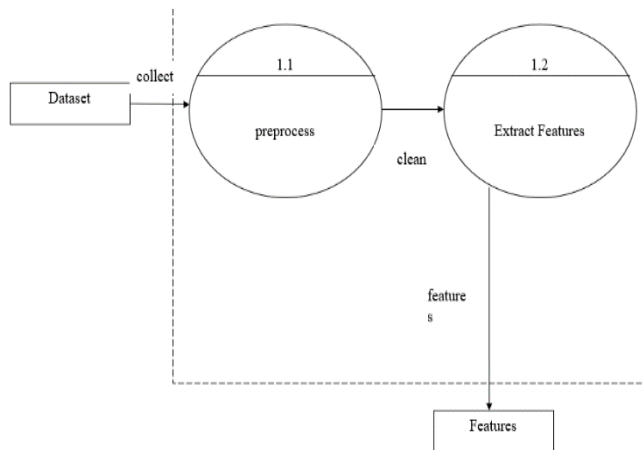


## 6. TRAINING

After preprocessing and EDA, we had the final dataset that had been thoroughly cleaned and analyzed. The 80-20 train-test validation method was used, which specifies that 80% of the information is used for planning and 20% is used for testing. The sklearn library is used to divide the data into training and testing portions. Out of 303 samples, 242 examples or instances are chosen and used to create the model. The remaining 61 samples are used as testing data to judge how well the constructed model performs. To implement the model, we proceeded forward. We carried out two distinct sorts of experiments during the implementation. Both the considered deep neural network and the artificial neural network were implemented individually. As a result, we were able to determine how accurately each of these models performed. A Deep neural network has multiple hidden layers. Whereas the Artificial Neural network has one or two hidden layers in it. The activation of neurons is present at the output layer.

$$f(x) = \frac{1}{1 + e^{-x}}$$

In the output layer, the sigmoid activation function is applied. The dataset's redundant features are removed using feature selection. Feature extraction and feature selection are different. Finding relevant components from the existing data is called feature extraction. By removing unnecessary features through feature selection, the neural network is fed with pertinent information.



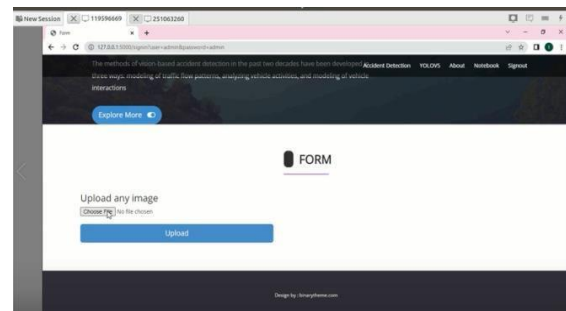
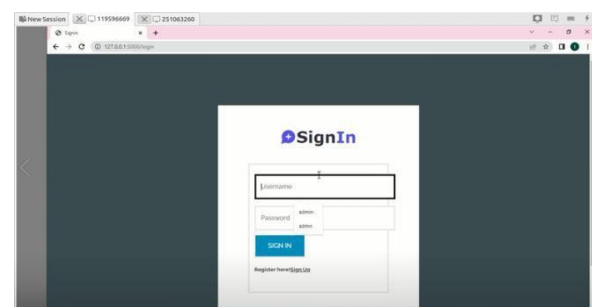
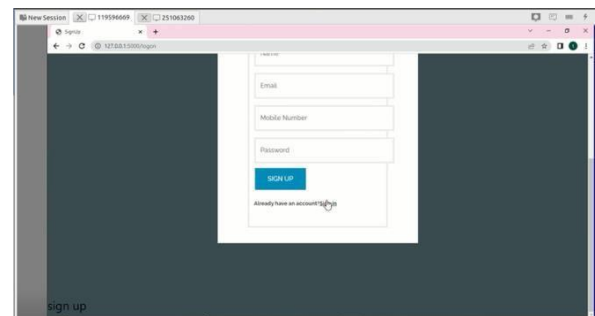
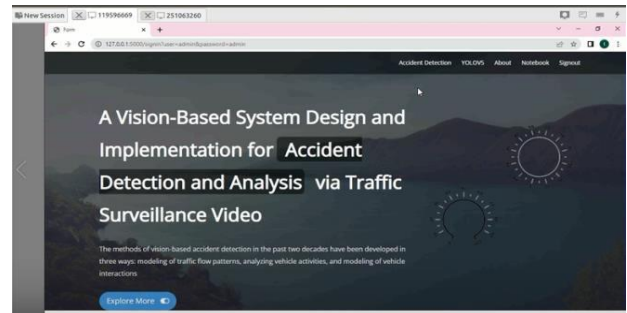
## 7. RESULTS

Fig 1 shows the home screen.

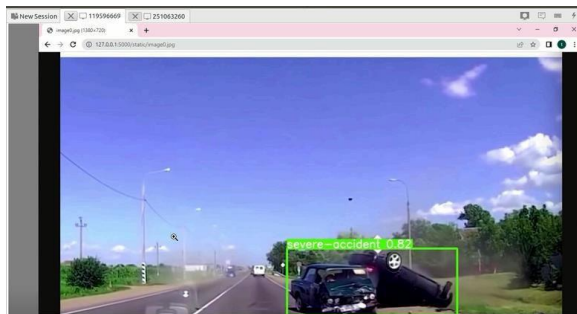
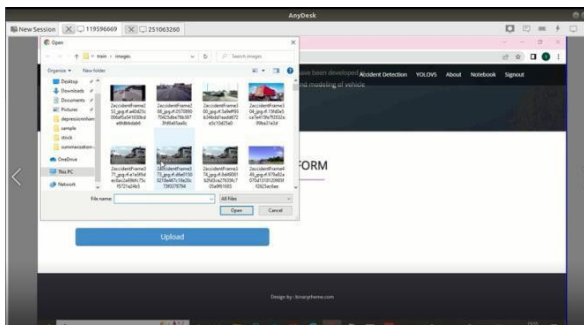
Fig 2 and Fig 3 depict the registration page.

Fig 4 and Fig 5, you input an image.

Fig 6 predicts the accident based on the input image. This suggests a process where you navigate from the home screen to the registration page, then input an image, and finally, the system predicts whether an accident is likely based on that image.







## 8.CONCLUSION AND FUTURE SCOPE

A framework for consequently distinguishing and examining car crashes utilizing observation video was proposed in this exploration. To start, the MIF model strategy was used to distinguish and find crashes in films. Second, a Consequences be damned v3 model was used to distinguish broken cars. Third, the directions were recovered utilizing the various leveled bunching strategy preceding the crash. To work with traffic police's independent direction, the directions were projected to an upward picture utilizing viewpoint change. The vehicle speed was determined after the directions were separated utilizing UFIR sifting. The determined speed and the gathered vertical effect point were then used to dissect a mishap. At long last, an equipment practice test was finished utilizing HiKey970, a Huawei computer-based intelligence demo board, to code all of the previously mentioned calculations. The demo board got a mishap observation video as information. The occasion was accurately Volume 12, Issue 01, Jan 2023 ISSN 2456 – 5083 Page 483 distinguished, and the significant vehicle directions were recorded. By 28.85%-45.72%, HiKey970 outscored the Intel Center i7-9750H central processor @ 2.60-GHz machine

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