

# Survey Paper on Traffic Rules Violation Detection

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## Abstract

The detection of traffic rule violators is a highly desired but tough task because of several challenges including occlusion, illumination, etc., which make it difficult to enforce safety measures on Indian roadways. The sophisticated technology known as the traffic rules violation detection system uses different techniques to detect and evaluate traffic offenses. The system records pictures and videos of cars breaking traffic laws using cameras positioned at intersections. Artificial intelligence algorithms are then used to evaluate these photos and videos in order to identify and categorize the specific infraction, such as lane violations, speeding, or red light jumping. In order for law enforcement to take action against the violators, the system can also automatically create alerts and notifications. Through the enforcement of traffic laws and regulations, the system contributes to a decrease in accidents and an improvement in overall road safety.

*Index Terms*—traffic rules violation detection, Artificial intelligence etc

## I. INTRODUCTION

Two-wheelers are a very popular form of transportation, but they carry a high danger because they have less protection. Bicyclists are strongly encouraged to wear helmets in order to lower the risk. Because governments recognize the importance of helmets, they have made it illegal to ride a bike without one and ordered that anyone who do so face legal action. But the technology used for video surveillance today are passive and heavily rely on human intervention. Because human intervention tends to reduce efficiency over time due to human weariness, these systems are typically inefficient. Furthermore, human intervention for helmet violations is ineffective since only one rider from a large number of two-wheeler riders may be apprehended at a time. Riders on two-wheelers and four-wheelers disregard crosswalks at signals. When crossing violations are detected manually, it is ineffective and would cause traffic jams since it would disrupt the flow of traffic. Since traffic movement would not be disrupted and several violations could be detected at once, using CCTV cameras to detect crossing violations would be very advantageous.

India currently uses a manual approach for identifying infractions and prosecuting offenders. The traffic is continuously recorded by the CCTVs at the signals. The traffic police crew keeps an eye on these real-time videos in the control room, and when they spot a violation, they record the license plate and capture a screenshot as evidence. The offender is then notified of the infraction and charged with the fine after the infraction is recorded in the database. The staff finds it exhausting to constantly watch the screens and make sure no infraction goes unnoticed in the midst of the nation's heavy traffic because this process is manual.

## II.LITERATUREREVIEW

Prem Kumar Bhaskar, et al, "Image Processing Based Vehicle Detection and Tracking Method" [1], In the field of traffic surveillance systems, where effective traffic management and safety are the primary concerns, vehicle detection and tracking are important and effective. The problem of identifying car and traffic information from video frames is covered in this work. There is still opportunity for development in this field despite the numerous studies that have been conducted and the several approaches that have been used. It is suggested that a novel algorithm for vehicle data tracking and recognition be developed utilizing blob detection techniques and the Gaussian mixture model in order to make improvements. We start by learning the backdrop in order to distinguish the foreground from the background in frames. In this case, the object is identified by the foreground detector, and rectangular zones surrounding each detected object are defined by a binary computation. Some morphological procedures have been used to eliminate noise and accurately detect moving objects. After then, the detected items and their territories are tracked in order to complete the final counting. Using the Gaussian Mixture Model and Blob Detection techniques, we achieved an averaged detection and tracking accuracy of over 91%, which is encouraging.

Kunal Dahiya, Dinesh Singh, and C. Krishna Mohan, "Automatic Detection of Bike-riders without Helmet using Surveillance Videos in Real-time" [2], The architecture for automatically identifying bikers without helmets using real-time surveillance footage is presented in this paper. The suggested method uses object segmentation and backdrop subtraction to first identify bike riders from surveillance footage. Next, it uses a binary classifier and visual cues to assess whether or not the biker is wearing a helmet. Additionally, we offer a violation reporting consolidation method that enhances the suggested method's dependability. We have presented a performance comparison of three different feature representations for classification in order to assess our methodology. According to the experimental results, the detection accuracy on real-world surveillance data is 93.80%. Additionally, it has been demonstrated that the suggested method works in real-time with a processing time of 11.58 ms per frame and is computationally less expensive.

Abdullah Asım, et al, "A Vehicle Detection Approach using Deep Learning Methodologies" [3], In order to maximize the success rate of the trained detector by delivering effective results for vehicle detection, the trained vehicle detector will be tested on test data. The study's goal is to successfully train our vehicle detector using R-CNN and Faster R-CNN deep learning techniques on sample vehicle data sets. There are six major phases to the working procedure. In that order, they are loading the data set, convolutional neural network design, training options configuration, Faster R-CNN object detector training, and trained detector evaluation. Furthermore, the study's scope included references to Faster R-CNN and R-CNN deep learning techniques, as well as comparisons between experimental analysis and vehicle detection findings.

J. Chiverton, "Helmet presence classification with motorcycle detection and tracking" [4], Although helmets are necessary for motorcyclist riders' safety, enforcing helmet use is a labour-intensive and time-consuming task. Therefore, a method for automatically classifying and monitoring motorcyclist riders wearing and not wearing helmets is explained and put to the test. The method makes use of support vector machines that have been trained on histograms created from motorbiker riders' head region image data utilizing both still photos and individual image frames from video footage. The learned classifier is integrated into a tracking system that uses background removal to automatically segment motorcyclist riders from video data. The trained classifier is used to classify the riders' heads after they have been segregated. Every motorcyclist creates a series of areas in neighbouring time periods known as tracks. The mean of the individual classifier findings is then used to classify these tracks collectively. Experiments demonstrate that the classifier can correctly identify whether or not riders are wearing helmets in still images. Additionally, tests on the tracking system show that the classification approach is valid and useful.

Gomathi, et al, “Automatic Detection of Motorcycle without helmet using IOT” [5], India’s large population, growing commuter population, poor traffic signal management, and rider behavior have all contributed to the country’s growing traffic violation monitoring and control problems. It has proven ineffective to handle such high traffic volumes and track violations by relying only on physical traffic police for monitoring. India’s large population, growing commuter population, poor traffic signal management, and rider behavior have all contributed to the country’s growing traffic violation monitoring and control problems. It has proven ineffective to handle such high traffic volumes and track violations by relying only on physical traffic police for monitoring.

Aaron Christian P. Uy, et al, “Automated Traffic Violation Apprehension System Using Genetic Algorithm and Artificial Neural Network” [6], The issue of packed and congested roadways is a concern in developing nations due to ineffective traffic law enforcement. Because they can simply escape and are not caught, drivers disregard the law. An intelligent traffic system that can automatically identify and stop traffic infractions specifically, drivers that swerve or obstruct the pedestrian lane is proposed in this study. Three processes violation detection, plate localization, and plate recognition are included into the system’s design. While an artificial neural network was used for the plate recognition process, genetic algorithms were used for violation detection and plate positioning. The position of the identified vehicle in relation to the camera is crucial for the plate number recognition. Therefore, the physical characteristics of the vehicle that are recorded by the violation detection procedure will be the primary information about the offender; the recognized plate number will only be supplemental information. With an average accuracy of 90.67% and a program duration of 1.34 seconds, the system as a whole was able to identify the plate number of the vehicles that were identified as traffic offenders and detect the aforementioned offenses based on the findings of 48 photos that were tested.

Samir Ibadov, et al, “Algorithm for detecting violations of traffic rules based on computer vision approaches” [7], In order to increase public safety at uncontrolled pedestrian crossings, we suggest a novel technique for automatically detecting traffic law breaches. There are multiple steps in the algorithm. They are the detection of autos, pedestrians, and zebras. We employ the quicker R-CNN deep learning algorithm for automobile detection. When it comes to identifying traffic law infractions, the algorithm performs admirably.

Amey Narkhede, et al, “Automatic Traffic Rule Violation Detection and Number Plate Recognition” [8], The global population is becoming more urbanized. As a result, there are now much more cars on municipal roads, which has led to a sharp rise in traffic violations. People are more critical these days. In addition to causing more accidents that could endanger lives, this causes serious harm to the environment. The people. The detection of traffic violations and their unthinkable consequences are necessary to address the concerning situation and stop similar incidents in the future. System implementation is essential.

Pooya Sagharichi Ha, Mojtaba Shakeri, “License Plate Automatic Recognition based on Edge Detection” [9], In this research, we offer an image processing program for identifying license plates: the Automatic License Plate Recognition System (ALPRS). There are four steps in the ALPRS primary process: The FMH filter eliminates the image’s noise. Background subtraction is done using a straightforward approach. The location of the license plate is localized using Canny edge detection. Lastly, the template matching technique is used to extract the letters and numbers. There are two benefits to the suggested algorithms: First, the approach is very resilient to noise. Secondly, it is capable of handling license plates with varying colors. A live video stream is used to test the algorithm’s performance. Our algorithm’s results indicate that, out of 70 car photos, the missing rate is nearly 16%.

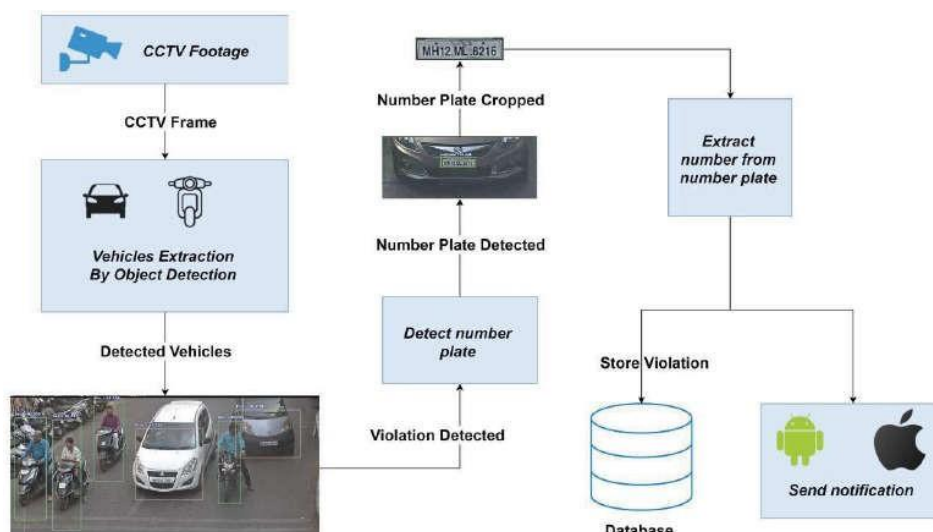
Kaiming He, Georgia Gkioxari, Piotr Dollár, Ross Girshick, “Mask R-CNN” [10], We offer a generic, adaptable, and conceptually straightforward framework for object instance

segmentation. Our method concurrently creates a high-quality segmentation mask for each instance and effectively recognizes items in an image. By incorporating a branch for object mask prediction alongside the current branch for bounding box identification, the technique, known as Mask R-CNN, expands on Faster R-CNN. Mask R-CNN runs at 5 frames per second, is easy to train, and adds only a minor overhead to Faster R-CNN. Furthermore, it is simple to adapt Mask R-CNN to other applications, such as estimating human poses using the same framework. In the three tracks of the COCO suite of challenges—person keypoint detection, bounding-box object detection, and instance segmentation—we provide the best results. On all tasks, including the COCO 2016 competition winners, Mask R-CNN surpasses all current, single-model entrants without any extras. We anticipate that our straightforward and efficient method will provide a strong foundation and facilitate further instance-level recognition research.

Joseph Redmon, Santosh Divvala, Ross Girshick, Ali Farhadi, “You Only Look Once: Unified, Real-Time Object Detection” [11], We introduce a novel method for object detection called YOLO. Classifiers are repurposed to carry out detection in earlier object detection operations. Rather, we formulate object detection as a regression issue to bounding boxes that are geographically separated and the class probabilities that go along with them. Bounding boxes and class probabilities are directly predicted from complete images in a single evaluation by a single neural network. The detection pipeline may be directly adjusted end-to-end on detection performance because it is a single network. The speed of our unified architecture is really high. At 45 frames per second, our fundamental YOLO model processes images in real time. Fast YOLO, a scaled-down variant of the network, achieves double the mAP of other real-time detectors while processing an incredible 155 frames per second. While YOLO is less likely to forecast false positives on background, it makes more localization errors than state-of-the-art detection algorithms. Lastly, YOLO picks up extremely broad object representations. When it comes to generalizing from natural images to other domains, such as artwork, it performs better than other detection techniques like DPM and R-CNN.

Dr. Praveen Blessington Thummalakunta, Manav More, Rhotuja Kakade, Omkar Nagare, Rutuja Sawant, “Traffic Offender Detection System Using Deep Learning Approach” [12], India's large population, growing commuter population, poor traffic signal management, and rider behavior have all contributed to the country's growing traffic violation monitoring and control problems. It has proven ineffective to handle such high traffic volumes and track offenses by relying just on physical traffic police for surveillance. India's large population, growing commuter population, poor traffic signal management, and rider behavior have all contributed to the country's growing traffic violation monitoring and control problems. It has proven ineffective to handle such high traffic volumes and track offenses by relying just on physical traffic police for surveillance.

### III. PROPOSED SYSTEM



### Figure1-SystemArchitecture

In the proposed system, we use Object Detection to recognize vehicles after initially feeding the machine an image frame from CCTV footage. YOLO, or "you look only once," is used in object detection to find cars in an image frame. Following vehicle detection, each vehicle is cropped out using the coordinates that the object detection algorithms have provided from the bounding boxes. Every car is now inspected for various infractions. This suggested system includes the following violations: Crosswalk Violation (Violating Zebra Crossing) and Helmet Violation (Two-Wheeler Rider Not Wearing a Helmet). Vehicles with two wheels will be inspected for helmet and crosswalk violations, and vehicles with four wheels will be examined for crosswalk violations. A CNN (Convolutional Neural Network) based classifier, which performs well on visual data, is used to detect helmet violations. By employing MaskRCNN, which uses instance segmentation to compare the coordinates of the vehicle's bottom tire with those of the detected crosswalk, crosswalk violations can be identified. Following the detection of a vehicle violation or violations, object detection is used to find the relevant car's license plate. Once more, YOLO is utilized to identify a vehicle's license plate. The license number is extracted from the license plate using optical character recognition, or OCR. Violations are added to the database and associated violations are communicated to vehicle users. Statistical analysis of past infractions of traffic laws can be obtained from the database.

### IV. CONCLUSION

This research examines various approaches for a Traffic Rules Violation Detection. The suggested method uses ideas like CNN, MaskR-CNN, OCR, and YOLO to automatically detect infractions of traffic laws. It easily and precisely accomplishes the intended result, but because it uses ideas like object detection and image segmentation, it demands a lot of processing power. The benefit of the suggested system is that, in comparison to the human-intervened system, it can detect a greater number of infractions. Furthermore, when put into practice, the suggested methodology's end-to-end autonomous system would provide it an advantage in identifying infractions. As a result, stringent laws against breaking traffic laws can be put into place, improving road safety and raising awareness among drivers.



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