



# INTELLIGENT TRAFFIC LIGHT CONTROL: ENHANCING EFFICIENCY WITH YOLO V4, AI TECHNOLOGY

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## ABSTRACT:

The increasing quantity of metropolitan cars is making traffic management, an ongoing concern. In addition to seriously disrupting our daily lives and elevating stress levels, traffic bottlenecks have an adverse effect on the environment by raising carbon emissions. Megacities are facing enormous obstacles and delays in their daily transportation-related activities as a result of population growth. To continuously evaluate traffic density and implement necessary measures. We need an intelligent traffic management system. The availability of separate lanes for different kinds of vehicles does not, however, shorten commute wait times at traffic signal locations. A technology that gathers live pictures from signals using intelligent machines is presented as an approach to the situation with regards to existing methodology. This method determines Volume of traffic by applying the YOLOv4 image processing methodology to earn efficiency.

**KEYWORDS:** artificial intelligence, intelligent monitoring, signal switching algorithm, traffic jams, the traffic light system, and the traffic management.

## I. INTRODUCTION

Urban roadways suffer capacity, congestion, and control issues as the number and volume of cars on the road rise, particularly in big cities. The current traffic management systems cause long lines at junctions because they demand a lot of labor and effort to avoid and prevent accidents. Upgraded

infrastructure and advanced technology are needed for better traffic control. These days, intelligent transportation technologies are essential for enhancing traffic control. Vehicle movement sequences may be managed at both single and many junctions using WSN (Wireless Sensor Network). Conventional traffic control systems need the traffic police to employ signs, beacons, and whistles to manage traffic congestion in addition to requiring a large amount of labor to operate the system manually. At traffic signals, sensors and timers are essential elements in managing vehicle flow. Electronic Sensor: Another clever solution is to install a proximity or loop sensor in the lane. This sensor gathers data on the movement of vehicles along the path. 4,444 traffic lights are controlled by sensor data.

Conventional timer-controlled traffic signals: These signals are tracked by timers. The light turns green or red according to the timer value, which is set to a certain number. Following a thorough analysis of the literature, we discovered many methods for identifying and reacting to vehicle density. Thus, we made the decision in order to construct an adaptable traffic control system that can recognize items inside the pictures and modify the time of vehicle lights accordingly. Conventional technology has a lot of shortcomings. A manual control system takes a lot of

time and work to set up. We are unable to manually regulate traffic in every region of cities and towns due to a lack of staff.

More efficient traffic control systems are therefore required. Static traffic management does not respond to vehicle traffic; instead, it employs timed signals for each designated period. Because obtaining high-quality data frequently necessitates costly and sophisticated technology, and because financial restrictions may restrict the number of facilities available, correctness and coverage of electronic sensors are sometimes not constant.

Furthermore, most sensors only cover a small area, thus several sensors are usually needed for a facility's network to offer thorough coverage. By counting the number of automobiles at traffic lights and modifying the duration of the green light, traffic density is calculated in real time using live pictures from CCTV (video surveillance) cameras positioned at traffic junctions. The following categories of vehicles are used to calculate how long a green traffic signal will last: automobiles, bicycles, buses, trucks, persons, and bicycles.

## II. LITERATURE SURVEY

Radio waves are used by Radio Frequency Identification (RFID) tags to provide information about an object to a combined antenna and reader. A.K. Yerapada, P. Manikonda, and colleagues devised a technique utilizing an RFID tag and an RFID reader to determine a vehicle's speed. The average time at a certain junction was determined using the average speed of the specified (N) leading cars. Nevertheless, this calls for the installation of an RFID tag on every car as well as ongoing communication between the tag and the reader. Without utilizing any hardware, A. Kanungo, A. Sharma, et al. employed a background division approach @ thirty frames per sec and computed the vehicle density of the picture matrix. The result is then divided by the constant C, which is equal to the sum of the camera height, rows, and columns times thirty. The vehicle density at four-way junctions determines the signal timings. It's green time! Maximum: 60 seconds, Minimum: 10 seconds nice idea. Despite the importance of both, the launch and quantity of vehicles were not considered in this study. Since the motion of the vehicle doesn't change, the computation procedure needed to be completed swiftly.

The SATL (Self-Adaptive Control Traffic Signal System) was presented by Jiayang Li, Yuan Rui Zhang, and others as an enhancement to the traffic signal adjusts itself that takes automobile top speed

into account. A transmission module and a data collecting module that transmits data to the signal receiver are installed in the vehicle. The algorithm that determines how much time to allot and extend in the event of an accident or injury considers a signal from a 100-meter-range ZigBee transmitter with  $V_{max} = 40$  km/h and  $V_{min} = 20$  km/h. 4,444 further crashes or mishaps. Nevertheless, turning causes automobiles to slow down, which is particularly crucial at junctions. For this reason, A system of infrared sensors that are positioned on both sides of the road was created by Mr. Vignesh.S, Mr.K.S. Naresh, and others. When a car moves, the recognizes it and transmits the information to the Raspberry Pi that is attached to the signal. When a lot of cars are picked up by the IR sensor, this Raspberry Pi reacts by allocating a longer green time.

Vehicle position sensors are included with P. Rizwan, K. Suresh, et al., and the sensor data is used to determine when activity light timing changes. A portable program with images and time is created using input from video cameras and sensors to find a different path to a destination while taking four bearing convergences into account. Khushi suggested a method in which morphological picture era is calculated using MATLAB's length work, and activity thickness is determined by sending images gathered at convergences to the program.

The Arduino receives the duration of the greenish and reddish lightnings connected to a particular crossing point ( $T_{dur} = T_{max} - k * T_{max}$ ) (north, south, east, west). In line with Muhammad Fakhrie's investigation, his subsequent advances in ANN have resolved a great deal of controversy and have played a pivotal role in the establishment of a great deal of businesses. To establish the number of cars on the normalized image, a constructed neural arrangement with a sliding component and a specified bounding box was developed. Additionally, a fuzzy-based activity flag control device is suggested, which modifies the flag time and divides the activity zone into moo, slightly moo, medium, moderately tall, and tall categories. A. Chattaraj, S. Bansal et al. integrated both computations to increase location precision by varying the age learning rate and carrying out the calculation in accordance with the requirements for accuracy.

## III. METHODOLOGY

This approach is used in carrying out this analysis. It applies his "YOLO" philosophy to an effective

structure for activity planning in the car process of creation.

YOLO may be an incredibly accurate and fast state-of-the-art protest location calculator. YOLO may be an intelligent convolutional neural network (CNN) capable of identifying a wide range of things. Using a neural network connected to the full image, each district's bounding box and chance are predicted by this approach, which splits the picture into districts. These bounding boxes' weights are determined by projected changes. You will create enormous datasets with all the information found in one image. Based on the YOLO calculation, a single CNN predicts unique bundles' bounding boxes and lesson probabilities. YOLO advances the execution of acknowledgment by preparing based on a collection of images. It is possible to optimize the spine CNN used in YOLO to advance with preparation and execution.

Darknet is an open-source neural network architecture based on CUDA and C. It supports both CPU and GPU computation and is easy to set up. YOLO with dark web achieves 72.9 percent top-1 exactness and ninety-one percent top-5 preciseness on ImageNet. Basically, 1x1 channels are used for yield channel reduction and 3x3 channels are used for including extracting on the darknet. We also use global mean pooling to provide projections.

#### IV. OBJECT DETECTION

Python was used to design the model, and Chrome Collab—which has several tools and frameworks to execute the previously described algorithms—was used to complete the model's development. The YOLOV4 method with 53 convolutional layers was constructed utilizing CV technology. Blobs from the image were applied to resize and scale the picture before passing it to cv2.dnn.readNetFromDarknet, which loads the configuration file and reads its weights to load the CNN layer of the strategy. Each item has several boxes formed in each grid, which are deleted while accounting for the threshold. Finally, to determine which box is optimal for the item, use non-maximal suppresses. These values are used to rectangular boxes (x, y, w, h) cv2 and to indicate things (cars, trucks, motorbikes, persons, and buses). Rectangles are used to indicate things. Maus Various color classes are employed to produce item names surrounding the car. Depending on the kind of vehicle, the average time it takes for a car to cross a lane is hours.

#### V. SIMULATION ENVIRONMENT

A collection of cross-platform Python modules called Pygame program are used to make simulations

and video games. It is made up of sound and graphics libraries for computers that are designed to operate with Python, including extensions for the superb SDL library and programs for AI, arithmetic, and Pygame. Indeed, it has. Users will be able to utilize Python to construct fully functional games and multimedia apps thanks to this. Pygame is incredibly small and compatible with nearly any platform and OS. We build cars at random using Pygame's features, then we track their motions by changing their coordinates on a regular basis. makes use of threads, a method for carrying out several activities at once. We maintain current traffic signals. To finish the work and track time in seconds, use the time function. Lord takes pictures of her in vehicles as Blitz watches.

#### VI. APPROACH ADOPTED IN YOLO

YOLOv4 has the benefit over earlier versions and technologies in that it does away with the disadvantages of GPU consumption and laborious picture processing. The level of detection is higher than in previous YOLO generations. Moreover, the mean accuracy (AP) and images per second (FPS) rose by 10% and 12% (i.e.) accordingly in comparison to YOLOv3.

The beautifully designed activity location display in this concept may be an activity management extension of the YOLO computation. Items determined by this computation aid in demoting the present flag clock. It takes a while to identify various photographs using the R-CNN computation. YOLO is a computationally intensive image recognition system that was developed in 2011 following extensive research on image preparation for quick and accurate multi-frame picture recognition. Reenactments were used to demonstrate the accuracy of the estimate by conveying different vehicle concentrations at random intervals at crossing sites in every heading. The flag swapping computation raises the quantity of vehicles that might change in each period compared to normal vehicles passing through hard-coded ways, facilitating persistent and productive vehicle growth. To carry out Calculation uses this time to reduce activity clog efficiently since it takes 15 seconds each time it requires to snap a photo and identify a car in the image.

#### VII. RESULTS AND DISCUSSION

The current part provides a quick overview of the findings and analysis that were done to create a successful traffic light control software that uses the YOLO engine. To help you better understand the intended work, you will find related images, matching illustrations, and tables in this area.

The number of cars of each kind that the motor vehicle identification module detects while the A green signal is the GST (Green Signal Time). "number of cars of

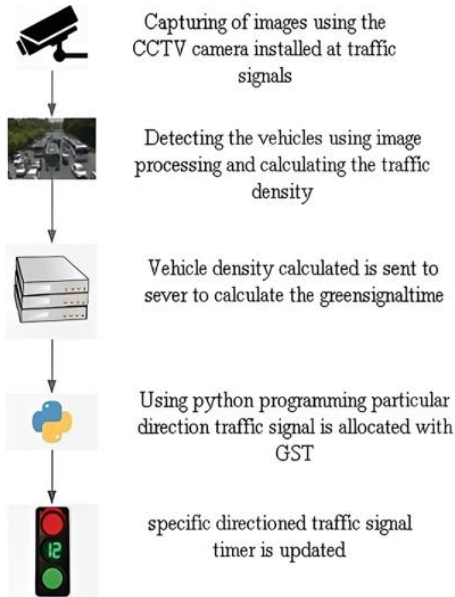


Fig. 1. Proposed System Model

traffic, particularly at congested crossings, while decreasing pollution and traffic congestion.



Fig. 2. Image captured from city traffic

"Class." The "Average Time of Class" refers to the typical amount of time it takes for cars in this class to arrive at the intersection. The number of lanes at the intersection is indicated by the variable "No of Lanes". With respect to the conditions, the mean duration needed For every kind of car to cross a street intersection can be adjusted. This can be done by district, city, or region, relying on the intersection's features.

The formula for calculating the Green Signal Time (GST) depending on velocity is shown below.

$$GST = \left[ \frac{\sum Noofvehicles_{vehicleclass} * averagetime_{vehicleclass}}{Nooflanes} \right]$$

Any transportation authority is able to evaluate data. The densest to the least dense states of the signals are periodically changed. This is consistent with modern technology, which enables users to switch routes in a predefined sequence when traffic lights turn green. With the exception of his yellow signal, which was taken into consideration, the signals stay in the same order. To improve traffic management, the average time it takes for each kind of vehicle to cross a junction can be adjusted according to the area, town, city, or even the design of the intersection.

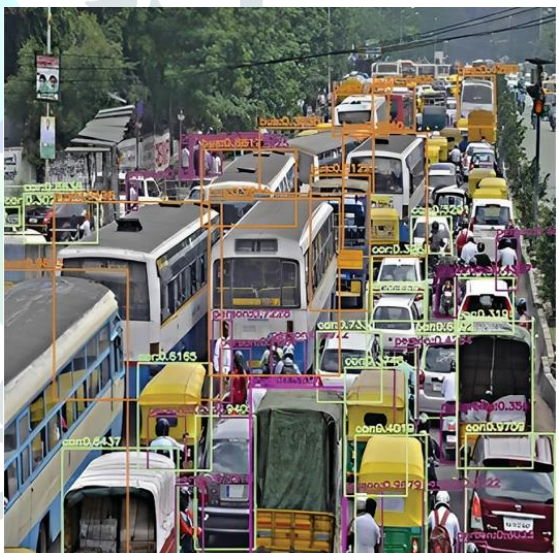


Fig.3. Image captured from the near bus station

**VIII. ANALYSIS AND DISCUSSION**

To develop an efficient traffic light control system utilizing the YOLO algorithm, this part evaluates the findings that were acquired and evaluated. The purpose of this study is to enhance everyday vehicle

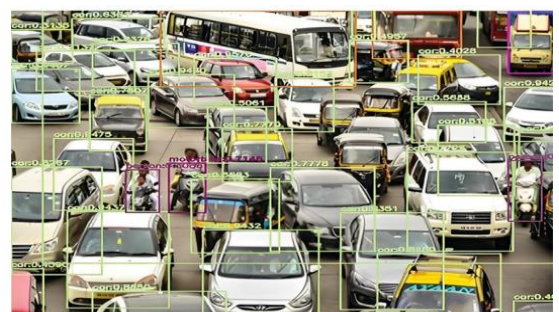


Fig.4. Image Captured from signal traffic

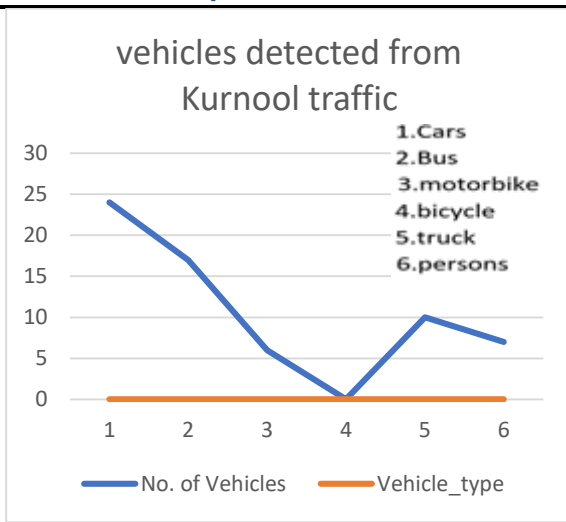


Fig.5. Visual depiction of traffic in Kurnool

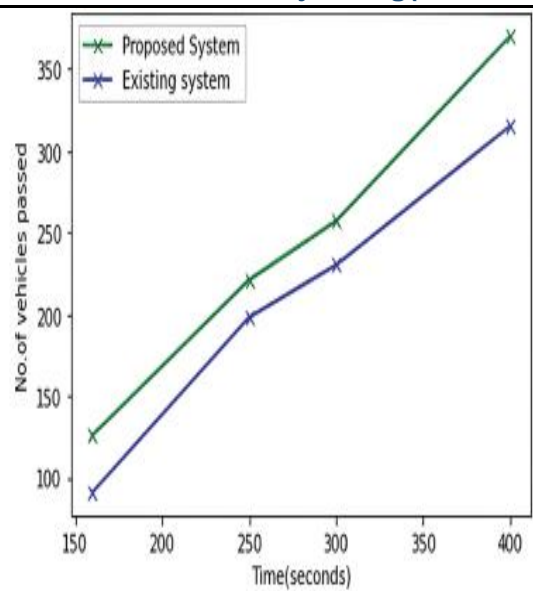


Fig.8. Current system versus system of intelligent traffic control

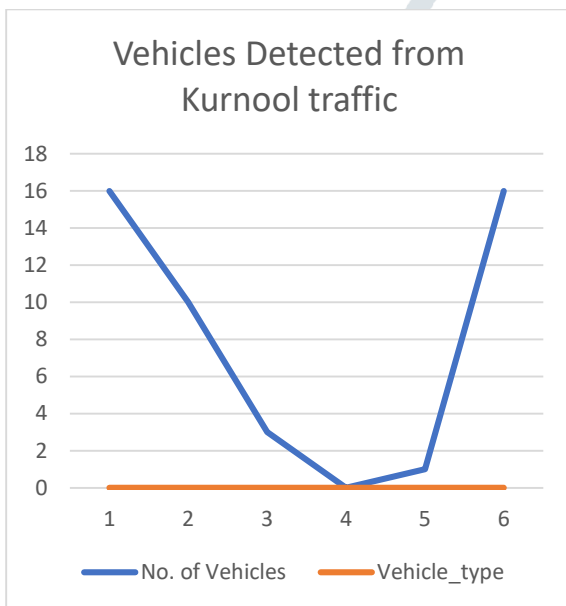


Fig.6. Graphical representation of Kurnool traffic near Nandyal Checkpost.

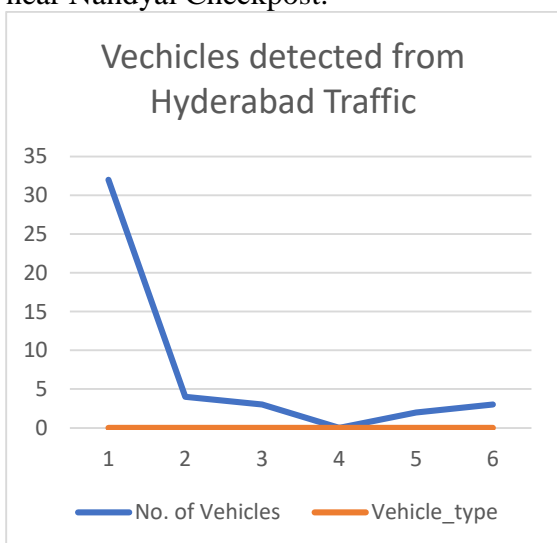


Fig.7.Graphical representation of Hyderabad traffic

**Table 1:** Allotment of Green Signal Duration

Rank of The figure	Area	Duration allocated in secs
1	Kurnool	46
2	Kurnool	23
3	Hyderabad	22

**Table 2:** Quantity of Vehicles Crossed in Per Lane of the Current Timer

Time Enacted	Lane 1	Lane 2	Lane 3	Lane 4	Unit time
160	18	33	15	23	0.55625
250	97	70	13	20	0.8
300	87	89	29	33	0.793333
400	122	109	40	40	0.7775

**Table 3:** The volume of cars that went through the right intelligent timer lanes

Time Enacted	Lane 1	Lane 2	Lane 3	Lane 4	Unit time
160	20	35	17	25	0.60625
250	100	73	15	24	0.848
300	90	93	30	38	0.83667
400	130	110	44	48	0.83

**Table 4:** The total grant of green light time (seconds) for various intervals time frames for the corresponding tracks

Time Enacted	Lane 1	Lane 2	Lane 3	Lane 4
160	47	32	23	23
250	72	74	35	23
300	118	72	34	33
400	125	113	38	43

Depending on the amount of automobiles in the image, the rate of detection accuracy ranges from 80 to 85 percent. the YOLOv4 algorithm has increased its ability to identify several vehicles in an image. In comparison to the conventional hard-coded approach, Using simulation, we also saw an important rise in the quantity of automobiles passing in a specific period of time.. As a result, there were more cars overall using all the lanes. The produced model may be applied in regions with a high vehicle density, although it is restricted to four-way junctions. The primary focus of this initiative is transportation, particularly in big, populated cities.

## IX. CONCLUSION

Depending on the vehicle density, the suggested YOLOv4 algorithm may identify a high number of cars in a picture with a rate of 85%. In comparison to conventional hard-coded methods. By using simulation, we were also able to gain an important rise in the number of automobiles crossing per unit of time. Every lane saw an increase in traffic. The created model may be used to four-way junctions and high traffic density locations.

Public transportation in cities is the main area of research. To guarantee that directions with heavy traffic take more time than directions with light traffic, the suggested technique also automatically modifies the green light period based on the volume of traffic at the traffic signal.

By doing this, inconvenient delays are removed, there is less waiting and traffic, as well as less fuel used and pollution produced. A certain number of cars crossing junctions is significantly reduced by the system, according to simulation data, as compared to current approaches. By adding actual CCTV data for modeling instruction and extra calibration, there may be ways to improve this device's performance.

## XI. REFERENCES

[1] Corovic, A., Ilic, V., Duric, S., Marijan, M., Pavkovic, B. "The Real-Time Detection of Traffic Participants Using YOLO Algorithm." 2018, 26th Telecommunications Forum (TELFOR). doi:10.1109/telfor.2018.8611986.  
 [2] A. Zaid, Y. Suhweil and M. A. Yaman, "Smart controlling for traffic light time," 2017 IEEE Jordan Conference on Applied Electrical Engineering and

Computing Technologies (AEECT), Aqaba, 2017, pp. 1–5. doi:10.1109/AEECT.2017.8257768.  
 [3] D.Y. Huang, Chao-Ho Chen, Wu-chin hu "Reliable moving vehicle detection based on the filtering of swinging tree leaves and raindrops." doi:10.1109/student.2019.6089337  
 [4] Tai Huu - Phuong Tran, Jae Wook Jeon. "Accurate Real-Time Traffic Light Detection Using YOLOv4." 2020. DOI: 10.1109/ICCE-Asia49877.2020.927706  
 [5] Hussain, J., Prathap, B.R., Sharma, A. (2022). "An Improved and Efficient YOLOv4 Method for Object Detection in Video Streaming." In: Shukla, S., Gao, XZ., Kureethara, J.V., Mishra, D. (eds), Data Science and Security. Lecture Notes in Networks and Systems, vol. 462. Springer, Singapore. https://doi.org/10.1007/978-981-19-2211-4\_27  
 [6] Muhammad Saleem, Sagheer Abbas, Taher M. Ghazal, Muhammad Adnan Khan, Nizar Sahawneh, Munir Ahmad. "Smart cities: Fusion-based intelligent traffic congestion control system for vehicular networks using machine learning techniques." Egyptian Informatics Journal, 2022.https://doi.org/10.1016/j.eij.2022003. [7] Alharbi, A., Halikias, G., Sen, A.A.A. et al. "A framework for dynamic smart traffic light management system." Int. J. Inf. Technol. vol. 13, 2021, 1769–1776. https://doi.org/10.1007/s41870-021007552.  
 [8] Ouallane, Asma Ait, et al. "Overview of Road Traffic Management Solutions based on IoT and AI." Procedia Computer Science, vol. 198, 2022, 518-523. https://doi.org/10.1016/j.procs.2021.12.279  
 [9] B. Ali Almansoori, S. Saif Almansoori, H. Almansoori, R. Ahmed Almansoori, I. Ahmed and K. Shahid, "AI-Based Adaptive Signaling for Traffic Control Around Roundabouts," Advances in Science and Engineering Technology International Conferences (ASET), 2022, pp. 1-5, Doi: 10.1109/ASET53988.2022.9735009.  
 [10] Dave, P., Chandarana, A., Goel, P., & Ganatra, A. "An amalgamation of YOLOv4 and XGBoost for Nextgen smart traffic management system." PeerJ. Computer Science, vol. 7, 2021, e586. https://doi.org/10.7717/peerj-cs.586.  
 [11] Adarsh P., Rathi, P., Kumar, M. (2020). "YOLO v3-Tiny: Object detection and recognition using one stage improved model." 2020 6th International Conference on Advanced Computing and Communication Systems (ICACCS). doi:10.1109/icaccs48705.2020.9074315 [12] Chong HF, Ng DWK. (2016). "Development of IoT device for traffic management system." 2016 IEEE Student

Conference on Research and Development (SCORED).doi:10.1109/scored.2016.7810059

[13] Manikonda P, Yerrapragada AK, Annasamudram SS. (2011). “Intelligent traffic management system.” 2011 IEEE Conference on Sustainable Utilization and Development in Engineering and Technology(STUDENT). doi:10.1109/student.2011.6089337

[14] Pranav Shinde, Srinand Yadav, Shivani Rudrake, Pravin Kumbhar. “Smart Traffic Control System using YOLO.” 2019 IEEE 8th Data-Driven Control and Learning Systems Conference(DDCLS).doi:10.1109/ddcls.2019.8908873

