

Automated Traffic Monitoring System

Darshan N Y¹, Dr. Sharath M N²

¹PG Student, Department of Computer Science and Engineering, Rajeev Institute of Technology, Hassan, Visvesvaraya Technological University, Belagavi, Karnataka, India.

²Department of Computer Science and Engineering (AI&ML) Rajeev Institute of Technology, Hassan, Visvesvaraya Technological University, Belagavi, Karnataka, India.

To Cite this Article: Darshan N Y¹, Dr. Sharath M N², "Automated Traffic Monitoring System", Indian Journal of Computer Science and Technology, Volume 04, Issue 03 (September-December 2025), PP: 153-159.



Copyright: ©2025 This is an open access journal, and articles are distributed under the terms of the [Creative Commons Attribution License](#); Which Permits unrestricted use, distribution, and reproduction in any medium, provided the original author and source are credited.

Abstract: An automated traffic monitoring system uses sensors and advanced algorithms to continuously collect, analyze, and manage real-time traffic data. This system can be used to optimize traffic flow, reduce congestion, improve safety, and provide drivers with accurate travel information. These systems use various sensors, such as loop detectors, cameras, and radar, to gather data on traffic volume, speed, density, and other relevant parameters. The collected data is then analyzed using algorithms and software to identify patterns, predict traffic flow, and detect potential problems. Based on the analyzed data, the system can implement various traffic management strategies, such as adjusting traffic signals, providing real-time traffic updates to drivers, and implementing incident management procedures. Automated traffic monitoring systems can be integrated with other transportation systems, such as public transportation, parking management, and ride-sharing services, to create a more comprehensive and efficient transportation ecosystem. By optimizing traffic signals and providing real-time traffic information, these systems can reduce congestion and improve traffic flow. Automated traffic monitoring systems can help to detect and respond to accidents, weather events, and other incidents that could pose a threat to road safety.

I. INTRODUCTION

The rapid growth of population and urban development in India is occurring simultaneously, leading many smaller towns to evolve into cities with populations exceeding a million. This surge has significantly strained existing infrastructure, including transportation, roadways, lighting, and housing. As a consequence, The number of vehicles on the roads has risen sharply in these cities, causing frequent congestion and prolonged traffic jams. These conditions have increased travel times and slowed down overall mobility. If this trend continues unchecked, it could lead to severely clogged roads, environmental degradation, health issues, and ultimately make these urban areas unsuitable for healthy human habitation. Traffic issues in India have long been overlooked, and our approach to addressing them has remained largely conservative. However, the situation now demands the integration of advanced technologies and innovative solutions to effectively manage urban traffic challenges. Several critical factors contribute to the growing congestion in Indian cities. Automated Traffic Monitoring Systems (ATMS) offer a promising solution by using modern technology to gather and analyze traffic data. These systems employ Sensors, cameras, and other devices to continuously track traffic conditions—such as vehicle speed, volume, and incidents. This real-time data helps traffic authorities make smarter decisions, such as modifying signal timings or diverting routes to ease congestion and enhance road safety. Urban traffic problems that once seemed limited to highly industrialized nations are now becoming increasingly common in Indian cities as well. The situation has escalated to a level where congestion and persistent traffic jams are seriously disrupting mobility. Commuters face longer travel times and slower average speeds, leading not only to frustration but also to a negative impact on public health and the environment. Moreover, in terms of road safety, the current conditions in many cities remain far from ideal. The adaptive traffic management system will offer real-time traffic flow simulation, capable of modeling traffic patterns, vehicle movements, queues, and turning maneuvers across the entire primary road network within the designated study area, including all ATCS-controlled intersections. The suggested solution involves using radar sensors at intersections to detect vehicles and measure queue lengths. These sensors transmit the data to local controllers at each junction. The local controllers are then connected to software on a central server in the Control room, enabling real-time coordination between all the intersections.

II RESEARCH MOTIVATION

- Traffic congestion, road safety, and inefficient traffic management are persistent challenges in urban areas worldwide. As cities grow and vehicle numbers increase, traditional traffic monitoring systems—often manual, limited in scope, and prone to human error—struggle to keep up with real-time demands. The need for smarter, scalable, and automated solutions is more urgent than ever.
- Automated Traffic Monitoring Systems (ATMS), powered by modern technologies such as computer vision, machine learning, and the Internet of Things (IoT), offer a promising alternative. These systems can continuously analyze traffic flow, detect violations, and provide real-time data to traffic authorities for informed decision-making. This not only improves traffic efficiency but also contributes to reducing accidents, enhancing public safety, and minimizing environmental impact caused

by idle vehicles.

- Moreover, accurate traffic data is crucial for long-term urban planning and infrastructure development. By automating data collection and analysis, an ATMS provides consistent, high-resolution insights that manual methods cannot match.
- This research is motivated by the potential of an Automated Traffic Monitoring System to revolutionize traffic management, enhance safety, and support the development of smart cities. It aims to address the technical challenges involved—such as object detection accuracy, real-time processing, and system scalability—while contributing to sustainable urban mobility solutions.

1. Need for Real-Time Traffic Data

Accurate and real-time traffic data is essential for dynamic traffic signal control, congestion management, and emergency response. Manual data collection methods are often outdated and inconsistent. An automated system ensures continuous, scalable, and high-resolution monitoring, enabling more responsive and informed decision-making by traffic authorities.

2. Improving Road Safety

Thousands of lives are lost annually due to preventable road accidents, many of which result from speeding, signal violations, and poor visibility. Automated systems can enhance surveillance and enforcement capabilities, ensuring traffic rules are followed more consistently and violators are identified promptly.

3. Urban Planning and Policy Making

Long-term infrastructure development and urban mobility planning require reliable traffic data. An ATMS provides critical insights into peak usage times, high-risk zones, and traffic behavior patterns, which can guide policy decisions and infrastructure investments.

4. Reducing Human Dependency and Error

Human monitoring is prone to fatigue, bias, and limited coverage. Automation reduces the dependency on personnel for routine monitoring and allows human operators to focus on higher-level analysis and strategic tasks.

5. Scalability and Integration with Smart Cities

With the global shift towards smart cities, integrating intelligent traffic systems with broader urban management platforms (e.g., public transport, emergency services, and environmental sensors) becomes vital. An automated traffic monitoring system fits well into the smart city ecosystem, offering seamless integration and interoperability.

6. Technological Advancements

Recent advancements in deep learning, high-definition cameras, edge-AI hardware, and wireless communication protocols have made the deployment of such systems technically feasible and economically viable. This research seeks to explore and harness these technologies to create a system that is not only functional but also efficient, cost-effective, and robust under real-world conditions.

III OBJECTIVES

1. Real-Time Traffic Monitoring

- Develop a system capable of continuously monitoring traffic flow using sensors or camera feeds.
- Capture and analyze live traffic conditions such as vehicle count, lane usage, and congestion levels.

2. Vehicle Detection and Classification

- Implement computer vision or sensor-based algorithms to detect moving and stationary vehicles.
- Classify vehicles based on type (e.g., car, truck, bus, motorcycle) for traffic analysis and regulatory enforcement.

3. Traffic Violation Detection

- **Automatically detect common traffic rule violations, such as:**
 - Red-light running
 - Speeding
 - Wrong-way driving
 - Illegal lane changes
 - No-entry violations
- Generate automatic alerts or evidence (e.g., snapshots, video) for enforcement agencies.

4. Data Collection and Analytics

- Collect and store historical traffic data for pattern analysis.
- Provide visual dashboards and statistical reports for traffic management authorities.
- Enable data export for integration with other urban systems or policy-making platforms.

5. Integration with Smart City Infrastructure

- Ensure compatibility with existing urban traffic systems (e.g., traffic lights, public transport monitoring, emergency response).
- Enable data sharing through APIs or cloud platforms to support smart city applications.

6. Automated Alerting System

- **Provide real-time alerts for:**
 - Accidents
 - Traffic jams
 - Road blockages
 - Emergency vehicle detection
- Notify concerned authorities through SMS, email, or mobile app.

7. Enhance Road Safety and Law Enforcement

- Support law enforcement with accurate and timestamped incident records.
- Deter violations by increasing the perceived risk of detection.

8. Scalability and Deployment Flexibility

- Design the system to be scalable across different urban and rural locations.
- Support easy deployment on:
 - Roadside units
 - Traffic signals
 - Drones or mobile surveillance vehicles
 - Cloud or edge computing platforms

9. Optimize Traffic Flow

- Analyze traffic data to optimize signal timings and suggest alternate routes.
- Enable adaptive traffic control based on real-time congestion levels.

10. Cost Efficiency and Automation

- Minimize operational costs by reducing reliance on manual monitoring.
- Automate tasks such as data logging, report generation, and rule enforcement notifications.

IV LITERATURE SURVEY

1. Traffic Monitoring Technologies & Sensing Modalities

Early work in urban traffic management emphasized the use of wireless sensor networks (WSNs) for real-time traffic data collection (vehicle count, speed, density) at intersections. For example, one survey highlights WSN usage combined with RFID, VANETs, Bluetooth, camera and infrared sensing in urban schemes. Non-intrusive sensing approaches have also been studied: e.g., the system “DeepWiTraffic” uses just two WiFi transceivers and deep learning on WiFi Channel State Information to detect and classify vehicles into five types—with high accuracy—demonstrating that even low-cost sensing can be viable in rural road contexts. More recently, computer vision and object-detection techniques dominate: a review paper examines vision-based traffic control and monitoring, covering density estimation, traffic sign detection, accident detection and emergency vehicle detection.

2. Automated Data Processing, Computer Vision & Machine Learning

Deep learning applied to traffic monitoring is surveyed in “Deep Learning Serves Traffic Safety Analysis: A Forward-looking Review” where the authors outline a processing pipeline: video enhancement, semantic/incident segmentation, object detection/classification, trajectory extraction, speed estimation and anomaly detection. Another paper “Artificial Intelligence Enabled Traffic Monitoring System” describes a system that uses deep convolutional neural networks on video surveillance data for queue detection, tracking stranded vehicles, vehicle counts, and works under challenging conditions (rain, low light).

3. Violation Detection, Traffic State Estimation & Applications

Vehicle classification surveys (turn0search7) frame one major application of traffic monitoring: classifying vehicles for road usage, tolling, violation detection. There are systems specialised for automated data collection: e.g., “Development of a tracking-based system for automated traffic data collection for roundabouts” describes a system dedicated to roundabout scenes where camera calibration, background subtraction, tracking and data mining are used to extract trajectories, vehicle speeds and volumes. They achieved >90% accuracy in comparison with manual measurement.

4. Deployment Challenges, Scalability & Smart City Integration

A broad survey “Intelligent Transportation Systems Using External Infrastructure: A Literature Survey” reviews ITS solutions using external infrastructure (roadsides, sensors) covering hardware setups, software analytics, and summarizes open challenges: plug-and-play deployment, secure real-time distribution, sensor reliability, and

scalability. The vision-based review highlights that datasets, performance metrics vary widely, making meaningful comparisons difficult; this emphasizes the deployment challenge of reproducible performance in new contexts.

5. Gaps & Open Research Directions

Based on the surveys cited, the key gaps and future directions include:

- **Standardised datasets** and benchmarking frameworks for urban traffic monitoring tasks (vehicle classification, queue estimation, violation detection) remain limited.
- **Robustness** under challenging conditions (night, bad weather, occlusion, camera movement) is still a serious challenge. The AI/vision survey points this out. [arXiv](#)
- **Edge processing / real-time processing:** While data collection is heavy, many works mention latency and hardware cost as constraints.

6. Suggested Structure for Your Literature Survey Section

- Introduction to Traffic Monitoring Systems: Motivation, scope, why automation is needed.
- Sensing Modalities & Data Acquisition: WSN, RFID, camera, WiFi CSI, LiDAR – strengths & weaknesses.
- Data Processing Methods: Computer vision, machine learning, deep learning – detection, classification, tracking, trajectory analysis.

V PROBLEM STATEMENT

Rapid urbanization, increasing vehicle density, and outdated traffic management infrastructure have led to severe traffic congestion, high accident rates, and ineffective law enforcement in cities worldwide. Traditional traffic monitoring techniques—such as manual surveillance, road sensors, or static CCTV systems—are often labor-intensive, limited in coverage, prone to human error, and incapable of providing real-time, actionable insights. In addition, the lack of integrated, automated systems makes it difficult for traffic authorities to make data-driven decisions, enforce traffic laws efficiently, and respond swiftly to incidents such as accidents, illegal parking, and signal violations. These limitations directly impact road safety, emergency response times, commuter satisfaction, and urban environmental quality.

Proposed Model

1. Input Layer: Camera Feed or Sensors

- High-resolution surveillance cameras or edge-based IoT sensors are installed at intersections or road segments.
- Drones can be optionally used for aerial monitoring in high-traffic areas.

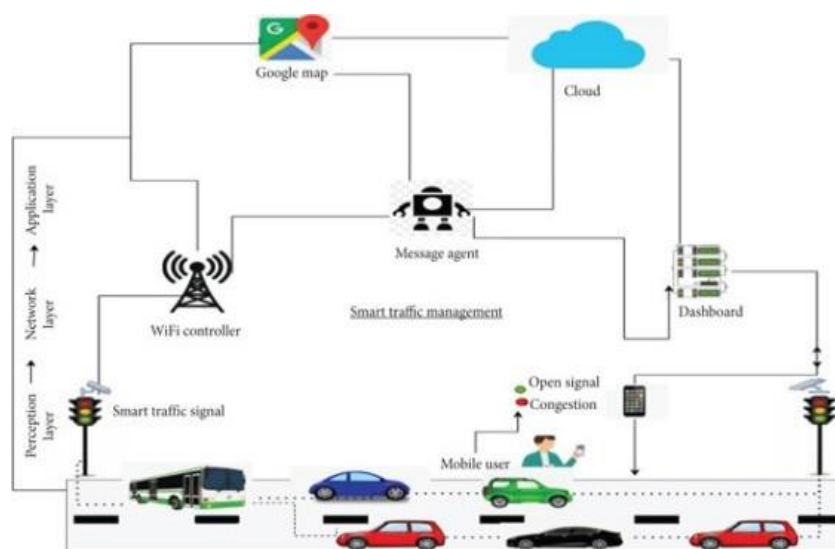


Fig a: Communication with mobile users

2. Image/Video Preprocessing

- Frames are extracted from the live video feed at fixed intervals.
- Image enhancement techniques (e.g., noise reduction, contrast correction) are applied to improve detection accuracy under low-light or poor weather.

3. Vehicle Detection and Classification

- Object detection models (e.g., YOLOv5, YOLOv8, Faster R-CNN) are used to detect vehicles in real time.
- Vehicles are classified into categories: car, truck, bus, motorcycle, etc.
- DeepSORT or ByteTrack can be used to assign persistent IDs and track vehicles across frames.

4. Violation Detection Module

- **Red-light Violation:** Use traffic signal status + vehicle motion analysis.
- **Speed Violation:** Use time-stamped tracking between two virtual lines to compute speed.
- **Wrong-Way Detection:** Use motion vectors against designated lane direction.
- **Lane Discipline Violations:** Use lane detection with deviation analysis.

5. Traffic Flow Analysis

- Count vehicles per time unit (e.g., per minute) for traffic density analysis.
- Detect congestion or bottlenecks based on vehicle clustering and motion slowdowns.

6. Data Storage and Logging

- Store all detection events, counts, and violations in a structured database.
- Evidence (e.g., frame snapshots of violations) is stored and linked to event records.

7. Notification/Alert System

- Real-time alerts (SMS, email, or push notifications) sent to traffic authorities for violations or accidents.
- Dashboard view for control rooms showing live data and flagged events.

8. Web-Based Dashboard (Frontend)

- User-friendly UI for traffic operators or law enforcement.
- Displays:
 - Live camera feeds
 - Vehicle counts and types
 - Real-time violations with evidence
 - Downloadable reports and charts

VI METHODOLOGY

1. Requirement Analysis & System Design

- **Define Scope:** Determine which features to implement (e.g., vehicle detection, counting, red-light violation detection, speed monitoring).
- **Identify Deployment Environment:** Urban intersections, highways, or toll booths.
- **Select Hardware & Software:** Cameras, edge devices (e.g., Raspberry Pi, Jetson Nano), server/cloud support, software libraries (OpenCV, YOLO, TensorFlow, etc.).

2. Data Collection

- **Video Acquisition:**
 - Capture real-time traffic videos using surveillance cameras or publicly available datasets.
 - Include various lighting, weather, and traffic conditions (day/night, rain, congestion, etc.)
- **Annotation (for Training, if required):**
 - Label vehicle bounding boxes and classes (car, truck, bike, bus) using tools like LabelImg or Roboflow.
 - Annotate violation events (red-light crossing, overspeed, wrong direction) if supervised training is needed.

3. Preprocessing

- **Frame Extraction:** Extract frames from video at a fixed interval (e.g., 10 fps).
- **Image Enhancement:** Apply filters (e.g., histogram equalization, denoising) for better visibility under poor lighting.
- **Region of Interest (ROI):** Define specific monitoring zones (e.g., stop lines, traffic lanes) to reduce processing load.

4. Vehicle Detection and Classification

- **Object Detection Model:**
 - Use pretrained models like **YOLOv5/YOLOv8**, **Faster R-CNN**, or **SSD** for real-time detection of vehicles.
 - Fine-tune the model on local dataset if accuracy is low.
- **Classification:**
 - Classify detected vehicles into categories: car, truck, bike, bus, etc.
 - Assign unique IDs to vehicles using tracking algorithms.

5. Vehicle Tracking

- **Tracking Algorithm:**
 - Implement tracking using **DeepSORT**, **Kalman Filter**, or **ByteTrack** to maintain vehicle identity across frames.
 - Calculate speed using frame-by-frame displacement over time.
- **Path Analysis:**

- Track the trajectory to detect lane changes, wrong-way driving, or U- turns.

6. Violation Detection Module

• Red-Light Violation:

- Monitor traffic signal status (manually or through smart integration).
- Check if any vehicle crosses the stop line during the red signal.

• Speed Violation:

- Set two virtual lines on the road with known distance.
- Measure time taken to cross both lines→ calculate speed = distance/time.

VILRESULTS

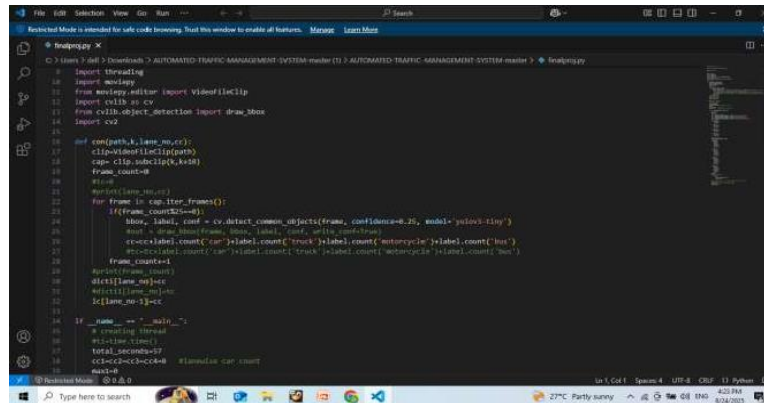


Fig a : Multi threaded finalproj.png

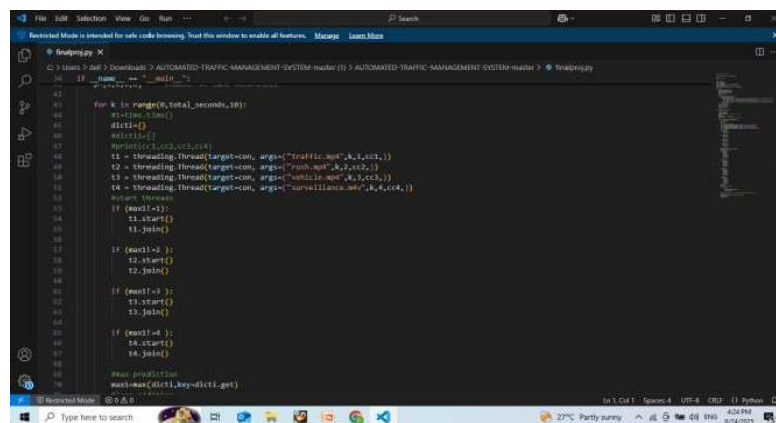


Fig b : Traffic video threading code.png

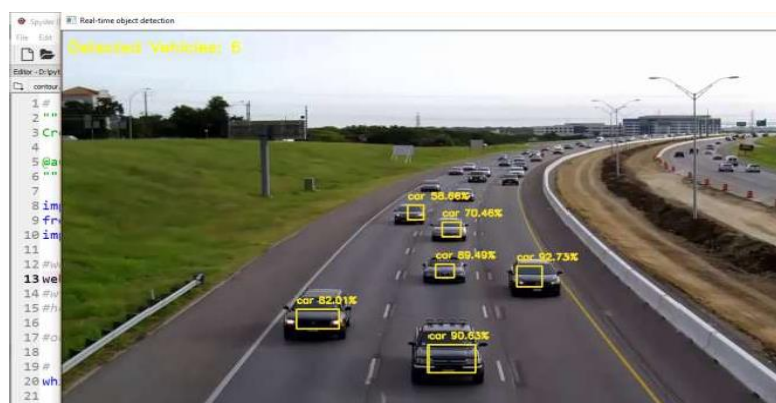


Fig c : Vehicle detection result highway.png

• Wrong-Way Driving:

- Compare direction of motion with allowed traffic flow in ROI.

• No Helmet / Number Plate Detectin (optional):

- Use secondary Models for Fine grained detection.

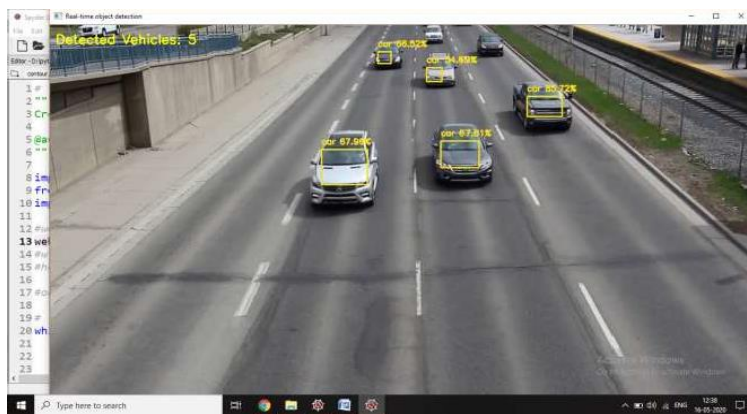


Fig d : Real time vehicle detection urban road.png

This image shows a real-time vehicle detection system running on a video feed from a highway. It looks like it's built using computer vision and deep learning models (possibly with frameworks like OpenCV + TensorFlow/PyTorch+ YOLO/SSD/Faster R-CNN).

Here's a breakdown of what's happening:

Detection Boxes (Yellow Rectangles)

- Each yellow rectangle surrounds a detected vehicle.
- The label shows "car" along with a confidence score (e.g., car 92.73%).
- This score indicates the model's confidence that the detected object is indeed a car.

How It Works

- A deep learning-based object detection model (such as YOLO, SSD, or Faster R-CNN) analyzes every video frame.
- **The system :**
 1. Identifies vehicles
 2. Categorizes them as cars
 3. Tallies the total count
 4. Shows the results live on the display.

VIII.CONCLUSION

The Automated Traffic Management System (ATMS) is designed to improve traffic flow efficiency, reduce congestion, and enhance road safety through the integration of modern technologies such as sensors, IoT devices, and intelligent algorithms. By automating traffic signal control, monitoring real-time traffic conditions, and providing data-driven decision-making, the system minimizes human intervention while increasing reliability and accuracy. This method enhances efficiency while also travel time for commuters but contributes to lowering fuel usage and environmental pollution. The system's adaptability to changing traffic patterns makes it a sustainable and scalable solution for urban mobility challenges. The Automated Traffic Management System effectively addresses the challenges of traffic congestion, road safety, and inefficient signal control through automation and real-time monitoring. By utilizing sensors, IoT devices, and intelligent algorithms, the system optimizes traffic flow, reduces travel time, minimizes fuel consumption, and lowers environmental pollution. Its scalability and adaptability make it a sustainable solution for modern urban transportation needs, contributing to a safer, faster, and more efficient traffic management process.

References

1. Li, L., & Jayakumar S., Lokesh Kumar K., Purva Darshini S. K., & Sanjeev D & Kumar et al. (2021). Traffic Monitoring System Using IoT and DL. In Advances in Parallel Computing Technologies and Applications. IOS Press.
2. Zhao, X., Dawson, D., Sarasua, W. A., & Birchfield, S. T. & Zhao et al (2020) (2020). Automated Traffic Surveillance System with Aerial Camera Arrays Imagery: Macroscopic Data Collection with Vehicle Tracking.
3. Chen, J., Xing, H.L., Yang, H., & Xu, L.X. (2019).Network Traffic Analysis Using LSTM Neural Networks. Open Journal of Applied Sciences, 9(12), 1–10.
4. Singh, T., Rajput, V., Satakshi, Prasad, U., & Kumar, M. (2022). Real-time traffic light violations using distributed streaming.The Journal of Supercomputing. <https://doi.org/10.1007/s11227-022-04897-9>
5. Li, T., Bian, Z., Lei, H., Zuo, F., Yang, Y.-T., Zhu,Q., Li, Z., & Ozbay, K. (2024). Multi-level Traffic- Responsive Tilt Camera Surveillance through Predictive Correlated Online Learning. arXiv. <https://arxiv.org/abs/2408.02208>
6. Wang, Y., Wang, Q., Suo, D., & Wang, T. (2020). Intelligent traffic monitoring and traffic diagnosis analysis based on neural network algorithm. Neural Computing and Applications.
7. Mandal, V., Mussah, A. R., Jin, P., & Adu- Gyamfi, Y. (2020). Artificial Intelligence Enabled Traffic Monitoring System. Sustainability, 12(21), 9177.
8. Dong, Z., Lu, Y., Tong, G., Shu, Y., Wang, S., & Shi, W. (2020). WatchDog: Real time Vehicle Tracking on Geo-distributed Edge Nodes.