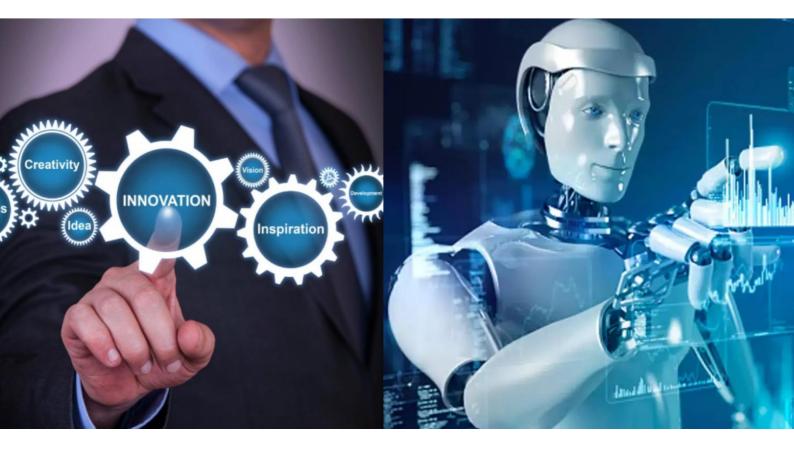




(A Monthly, Peer Reviewed, Refereed, Scholarly Indexed, Open Access Journal)



Impact Factor: 8.699

Volume 14, Issue 3, March 2025

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www.ijirset.com |A Monthly, Peer Reviewed & Refereed Journal| e-ISSN: 2319-8753| p-ISSN: 2347-6710|

Volume 14, Issue 3, March 2025

Real-Time Automatic License Plate Recognition using Jetson Nano

Harsh Upadhyay, Dr. Vinod Patidar

U.G. Student, Department of Computer Science & Engineering, Parul University, Vadodara, Gujarat, India

Associate Professor, Department of Computer Science & Engineering, Parul University, Vadodara, Gujarat, India

ABSTRACT: The rapid increase in the number of vehicles on roads has created an urgent need for an efficient and automated vehicle identification system. Traditional manual methods for vehicle tracking are time-consuming and prone to errors. This research presents a real-time Automatic Number Plate Recognition (ANPR) system using Jetson Nano, which leverages deep learning and computer vision techniques for efficient vehicle identification. The proposed system integrates the YOLO (You Only Look Once) object detection framework for car and license plate detection and Optical Character Recognition (OCR) for extracting alphanumeric data. The system is trained using a custom dataset of Indian number plates, ensuring adaptability to various environmental conditions such as different lighting and angles. Experimental results demonstrate a high accuracy rate in number plate detection and character recognition, making this system a reliable solution for automated traffic monitoring, toll collection, and security applications. Our model demonstrated a 97% accuracy in detecting vehicles, 98% in plate localization, and 90% in character recognition, reinforcing its applicability for traffic management and security enforcement.

KEYWORDS: RALPR, Jetson Nano, YOLO, OCR, Deep Learning, Traffic Monitoring, Vehicle Identification

I. INTRODUCTION

With the continuous rise in vehicular traffic, an automated system for monitoring and identifying vehicles has become essential. Traditional methods involve manual verification, which is inefficient and prone to human errors. Automatic Number Plate Recognition (ANPR) systems have been introduced to address these challenges by leveraging machine learning and computer vision techniques.

ANPR plays a crucial role in applications such as traffic management, toll collection, law enforcement, and parking systems. The major challenge in designing an effective ANPR system is achieving high accuracy in real-time conditions. This research focuses on developing a robust ANPR system using the NVIDIA Jetson Nano, a high-performance edge computing device optimized for deep learning applications. The system implements the YOLO object detection framework for vehicle and license plate recognition and uses OCR for character extraction. The approach ensures efficient processing with minimal latency, making it suitable for real-time applications.

Furthermore, ANPR can be extended to smart city applications, such as automated parking systems and real-time vehicle tracking for law enforcement agencies. By integrating cloud computing, the system can store historical vehicle records, helping authorities in traffic planning and congestion management.

The increasing reliance on artificial intelligence for surveillance makes ANPR systems a critical component of modern transportation networks. Additionally, by integrating blockchain technology, vehicle data security and transparency can be improved, ensuring a tamper-proof log of traffic violations and vehicle movements.

- ▶ Leveraging deep learning for superior vehicle and license plate detection.
- Enhancing recognition accuracy using OCR models.
- Conducting extensive evaluations in dynamic, real-world environments.





www.ijirset.com |A Monthly, Peer Reviewed & Refereed Journal| e-ISSN: 2319-8753| p-ISSN: 2347-6710|

Volume 14, Issue 3, March 2025

|DOI: 10.15680/IJIRSET.2025.1403340|

II. ORIGIN AND BACKGROUND

The concept of Automatic Number Plate Recognition (ANPR) dates back to the 1970s, when it was first developed for law enforcement and traffic monitoring in the United Kingdom. Early ANPR systems relied on basic optical character recognition (OCR) technology combined with fixed surveillance cameras. Over the decades, advancements in computer vision, deep learning, and edge computing have transformed ANPR into a highly efficient, real-time vehicle identification system. Today, ANPR technology is widely used in applications such as traffic law enforcement, toll collection, parking management, and security surveillance.

With the rapid growth of smart cities and intelligent transportation systems (ITS), the need for real-time, accurate, and automated vehicle identification has increased significantly. Traditional ANPR systems, which rely on high-end computational servers, are often costly and require significant infrastructure investments. To address these challenges, edge computing solutions have emerged, enabling on-device AI processing for ANPR systems. The introduction of NVIDIA Jetson Nano, a low-power, high-performance AI computing platform, has revolutionized the development of cost-effective, portable, and real-time ANPR systems.

In densely populated countries like India, where traffic congestion is a major issue, ANPR-based traffic monitoring can significantly improve law enforcement efficiency, toll automation, and urban mobility solutions. According to recent studies, India has over 300 million registered vehicles, and the number is expected to rise exponentially. The manual monitoring of traffic violations, stolen vehicles, and unauthorized parking is a challenging task for law enforcement agencies. Implementing a real-time ANPR system powered by Jetson Nano can enhance vehicle tracking, automated fine collection, and traffic rule enforcement without requiring extensive human intervention.

Moreover, the rise of smart surveillance systems in urban infrastructure has fueled the adoption of AI-powered ANPR solutions. Many government agencies and private firms are now investing in real-time vehicle recognition systems to improve security at restricted zones, automate parking systems, and enable frictionless toll payments. With real-time AI inference capabilities, low power consumption, and cost-effectiveness, the proposed Jetson Nano-based ANPR system is designed to address critical challenges in traffic monitoring, security enforcement, and automated vehicle identification.

This project aims to develop a robust, scalable, and real-time number plate recognition system that can be deployed in various smart city applications, law enforcement operations, and intelligent transportation networks. The integration of deep learning models (YOLO for object detection) with OCR-based number plate recognition ensures high accuracy and fast processing speeds, making it an ideal solution for automated vehicle monitoring and security applications.

III. LITERATURE REVIEW

Several methods have been proposed in past studies to improve ANPR accuracy and efficiency. Early techniques relied on template matching and edge detection methods, which lacked robustness in dynamic environments. Recent advancements in deep learning have significantly improved accuracy through convolutional neural networks (CNNs) and object detection frameworks like YOLO.

A study conducted on Turkish license plates using deep learning achieved an overall accuracy of 96.36% by integrating CNNs and Long Short-Term Memory (LSTM) networks. Another approach used machine learning-based segmentation techniques combined with edge detection, improving character recognition accuracy. However, these methods required extensive computational power, making real-time implementation challenging.

The YOLO framework has emerged as a state-of-the-art solution for object detection due to its speed and accuracy. Coupled with OCR, it allows efficient number plate detection and character recognition. This research builds upon these methodologies by optimizing YOLO for Jetson Nano, ensuring real-time performance in Indian road conditions.

Recent studies have also explored integrating generative adversarial networks (GANs) to synthesize high-quality training datasets. This enhances the model's robustness in detecting plates under challenging conditions, such as night-time scenarios, motion blur, and occlusions.





[www.ijirset.com |A Monthly, Peer Reviewed & Refereed Journal| e-ISSN: 2319-8753| p-ISSN: 2347-6710|

Volume 14, Issue 3, March 2025

DOI: 10.15680/IJIRSET.2025.1403340

Additionally, reinforcement learning techniques are being explored to dynamically adjust ANPR parameters based on environmental conditions, improving overall accuracy in varying real-world scenarios.

IV. OBJECTIVES

- 1. To study the origin, evolution, and advancements in Automatic Number Plate Recognition (ANPR) technology and its applications in law enforcement, traffic monitoring, and smart city infrastructure.
- 2. To analyze the benefits of real-time ANPR systems for government agencies, law enforcement, parking management, and toll automation.
- 3. To identify the challenges associated with traditional ANPR systems, including computational limitations, accuracy issues, and environmental constraints, and propose effective solutions using Jetson Nano-based edge computing.
- 4. To evaluate the impact of ANPR systems on traffic law enforcement, automated vehicle tracking, and security surveillance, assessing their effectiveness in reducing manual monitoring efforts and traffic violations.
- 5. To examine the role of Artificial Intelligence (AI), Deep Learning, and Edge Computing in enhancing the efficiency, accuracy, and scalability of real-time ANPR systems, making them more cost-effective and widely deployable.

V. CURRENT LANDSCAPE OF RALPR

The rapid advancement of Artificial Intelligence (AI) and Edge Computing has revolutionized various sectors, including law enforcement, traffic management, and smart surveillance. Real-Time Automatic Number Plate Recognition (RANLPR) has become a crucial technology, enabling automated vehicle tracking, toll collection, and security monitoring with greater accuracy and efficiency.

Recognizing the importance of real-time vehicle identification, several global and Indian technology firms, startups, and research institutions are developing AI-driven ANPR systems. The market for AI-powered surveillance and smart traffic solutions is rapidly expanding, with governments and private organizations investing heavily in intelligent transportation systems.

Initially, ANPR technology was limited to static surveillance cameras with offline processing, often resulting in delays and inaccuracies. Over time, the industry has transitioned to real-time, edge-based solutions powered by compact AI devices like the Jetson Nano. These advancements enable instant recognition, high-speed data processing, and seamless integration with cloud-based monitoring systems.

Today, modern ANPR systems utilize deep learning models such as YOLO for object detection and OCR for text extraction, significantly improving accuracy and speed. Many countries, including India, are integrating real-time ANPR with smart city initiatives, e-challan systems, and automated toll booths to enhance traffic management and law enforcement efficiency.

Despite these advancements, challenges remain. The variability in license plate formats, environmental factors (low light, rain, glare), and hardware limitations pose obstacles to widespread adoption. However, with ongoing improvements in AI models, image processing techniques, and hardware acceleration, the future of RANLPR systems appears promising. Companies and research institutions are actively working on enhanced recognition algorithms, cloud-edge hybrid models, and IoT-based smart surveillance to ensure greater accuracy, scalability, and cost-effectiveness in real-time number plate recognition systems.

VI. EMERGING TECHNOLOGIES IN RALPR

1. AI and Deep Learning for Enhanced License Plate Recognition

Artificial Intelligence (AI) and Deep Learning (DL) are revolutionizing real-time number plate recognition systems by improving accuracy, efficiency, and adaptability. Advanced Convolutional Neural Networks (CNNs) and Transformerbased OCR models enhance character recognition, even in low-light or distorted conditions.





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- Example: Modern ANPR systems leverage AI-based object detection models like YOLOv5 and YOLOv8 to detect license plates efficiently. Additionally, OCR models trained with deep learning frameworks (e.g., Tesseract, EasyOCR) provide accurate character extraction.
- Advantages: Higher recognition accuracy, ability to handle diverse license plate formats, and improved speed in real-time applications.

2. Edge Computing and IoT-Enabled ANPR Systems

With the rise of Edge AI devices like NVIDIA Jetson Nano, Jetson Xavier, and Raspberry Pi, ANPR systems can process data directly at the source, reducing latency and dependency on cloud processing. This enables real-time vehicle recognition in traffic-heavy environments.

- Example: Smart cities use IoT-enabled ANPR cameras integrated with edge computing to track vehicle movements without requiring centralized servers.
- Advantages: Faster response time, reduced bandwidth usage, and cost-effective deployment in remote locations.
- Challenges: Limited computational power of edge devices and the need for model optimization techniques like TensorRT and model quantization.

3. Automated Surveillance and Smart Traffic Management

RALPR technology is increasingly being integrated into intelligent traffic management systems to enforce laws and reduce manual intervention in monitoring vehicles. AI-powered cameras can detect traffic violations, monitor speed limits, and identify stolen or unauthorized vehicles in real time.

- Example: India's e-Challan system uses ANPR-equipped traffic cameras to automatically issue fines for traffic violations.
- Advantages: Reduction in manual labor, improved law enforcement, and enhanced road safety.
- Challenges: Ensuring compliance with privacy laws and managing data security concerns.

4. Cloud-Based ANPR and Data Analytics for Smart Cities

Cloud computing is transforming ANPR systems by enabling centralized data storage, real-time analytics, and multicamera synchronization. This facilitates advanced analytics, allowing authorities to track traffic trends, monitor congestion, and enhance urban planning.

- Example: Cloud-integrated ANPR helps law enforcement agencies cross-check vehicle data with national databases to track criminal activity and stolen vehicles.
- Advantages: Scalability, seamless integration with big data analytics, and improved system interoperability.
- Challenges: High dependency on internet connectivity and cloud infrastructure, potential cybersecurity risks.

5. Blockchain for Secure and Transparent Vehicle Data Management

Blockchain technology is emerging as a secure and tamper-proof solution for storing vehicle registration details, toll payments, and traffic violation records. This ensures data integrity, transparency, and prevents fraudulent modifications.

- Example: Governments and transportation authorities are exploring blockchain-based vehicle databases for secure vehicle registration and automated toll payments.
- Advantages: Eliminates data tampering, improves transparency, and enhances vehicle tracking security.
- Challenges: Scalability issues and integration complexities with existing legacy systems.

VII.CHALLENGES IN THE RALPR

- 1) Data Security and Privacy Risks
 - Risk of data breaches and hacking attempts.
 - Legal and ethical concerns over mass surveillance.
 - Need for encryption, multi-factor authentication, and strict access control.
- 2) High Implementation and Maintenance Costs
 - Expensive hardware (cameras, AI models, cloud storage).
 - High operational costs for software updates and network maintenance.





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- Need for cost-effective AI solutions and energy-efficient systems.
 - Accuracy Issues in Different Conditions
 - Poor lighting, extreme weather, and damaged plates reduce accuracy.
 - Variability in plate fonts and sizes makes recognition difficult.
 - Infrared cameras and AI enhancement techniques can improve performance.
- 4) Legal and Ethical Concerns

3)

- Potential misuse of data for unauthorized tracking.
- Unclear legal frameworks for data storage and sharing.
- Need for transparent policies and regulatory compliance.
- 5) Integration with Existing Systems
 - Compatibility issues with traffic and law enforcement databases.
 - Requirement for seamless real-time data exchange.
 - Need for AI-driven automation to improve efficiency.

VIII. FUTURE SCOPE

The future of online food delivery apps is set to be revolutionized by advancements in Artificial Intelligence (AI) and emerging technologies. AI-powered personalization will enhance user experience by offering dynamic food recommendations based on individual preferences, weather conditions, and past orders. The introduction of voiceactivated and gesture-based ordering systems will provide a seamless, hands-free way for customers to place orders, making the process even more convenient. Additionally, the adoption of autonomous delivery solutions, such as drones and self-driving vehicles, will significantly improve delivery efficiency and reduce operational costs. AI-driven smart inventory management and demand forecasting will help minimize food wastage and optimize restaurant supply chains, ensuring cost-effective operations. Enhanced customer service through AI-powered chatbots and virtual assistants will lead to quicker and more human-like interactions, improving customer satisfaction. Blockchain technology will be increasingly integrated into food delivery platforms to ensure transparency in food sourcing, quality control, and supply chain traceability, enhancing customer trust. With the growing trend of health-conscious eating, AI will also play a crucial role in recommending healthier meal options and tracking users' nutritional intake. Furthermore, dynamic pricing models powered by AI will help businesses optimize revenue while offering personalized deals to customers. Sustainability will become a core focus, with companies investing in eco-friendly delivery solutions such as electric vehicles and biodegradable packaging. The integration of food delivery apps with smart home and IoT devices will further streamline the ordering process, making it more intuitive and efficient. Lastly, AI-driven data analytics will provide businesses with deeper market insights, enabling them to make data-driven decisions and remain competitive in the fast-evolving food delivery industry.

IX. CONCLUSION

The research successfully implements a real-time RALPR system using Jetson Nano and deep learning frameworks. The system achieves high accuracy in vehicle and number plate detection, making it a reliable solution for traffic monitoring and security applications.

Future work will focus on optimizing the OCR model for improved character recognition, integrating reinforcement learning for adaptive improvements, and expanding the dataset to include multiple regional number plate styles. Additionally, the system can be enhanced with cloud-based data storage for centralized vehicle tracking.

Potential advancements include implementing federated learning to enable data privacy while training models on decentralized data sources. Furthermore, enhancing the system with multimodal learning techniques, such as incorporating radar sensors, can improve accuracy in low-visibility conditions.

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www.ijirset.com |A Monthly, Peer Reviewed & Refereed Journal| e-ISSN: 2319-8753| p-ISSN: 2347-6710|

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