

## Revolutionizing Toll Collection with Automatic Number Plate Detection Systems

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**Abstract**—One kind of security system is the number plate recognition system. The NPR system uses image processing principles. Additionally, this makes use of an OCR (Optical Character Recognition) system to decipher images of license plates. Tollway authorities utilize number plate recognition systems for a variety of reasons, one of which is to automatically recognize a vehicle's license plate, give the driver with a pay-slip, and then open the toll road to that vehicle. To authorize the car to park in their designated location, parking authorities also use this technique. The technology works by taking a picture of the license plate, processing it, and then reading each character on the plate to ensure flawless identification. The most important part is optical character recognition (OCR), which converts the images of license plates into text that can be deciphered later. This study presents an ANPR algorithm and network flow in its entirety, along with examples of its effective use. The Automatic Number Plate Recognition (ANPR) system was found to have an accuracy rate of 75- 85% for Indian number plates, thanks to its foundation in template matching.

**Keywords** - Automatic Number Plate Recognition,Django, OCR, Toll Collection ,OpenCV, Vehicle Detection, Web Application

### I. INTRODUCTION

Road maintenance and infrastructure development are financed in large part by toll collection, making it an essential part of today's transportation infrastructure. Congestion, human mistake, and operational inefficiency are common problems with traditional toll collecting technologies like RFID- based devices and manual cash payments. Not only can these systems cause travel delays, particularly during rush hour, but they also need substantial infrastructure and maintenance expenditures. The demand for more effective and scalable toll collecting technologies is rising in tandem with the expansion of transportation networks and the corresponding increases in traffic volumes. The use of Automatic Number Plate Detection (ANPD) systems is one potential answer.

When vehicles go through toll booths, ANPD's state-of-the-art image processing and Optical Character Recognition (OCR) systems automatically identify and scan the license plates. With this data, the driver may be charged without stopping or interacting with toll booth workers. Toll collection can be made more efficient, operating expenses can be reduced, and user comfort can be increased by combining ANPD with automated toll systems. A high-resolution camera installed at toll booths may capture photos of passing automobiles and is the central component of the system. In order to correctly identify the license plate number, these photos are analyzed using algorithms developed by ANPD. Payment for the toll is automatically debited from the linked account when the detected number is compared with a pre-registered database. It is possible to send out notifications for human intervention or legal action in the event that the vehicle is unregistered or has an inadequate balance.

Automated Number Plate Recognition (ANPR) systems use Optical Character Recognition (OCR) and high-resolution cameras to read license plates as cars go through toll lanes. Toll costs are determined by comparing the extracted license plate information with a central database; payments are then automatically debited from a bank account or digital wallet that is connected. Improving efficiency and lowering traffic congestion, this method removes the need for actual toll booths, cash payments, and RFID tags. One solution to these problems is the Automated Number Plate Detection (ANPD) Toll System, which uses ANPR technology to collect tolls in an efficient and automated manner.

In addition to collecting tolls, automated number plate recognition (ANPR) systems also improve safety by letting authorities keep tabs on illegal or stolen cars and traffic infractions. Although it has many benefits, it still has certain problems that need fixing in order to work at its best, including different types of license plates, bad lighting, and database administration. A smart, automated, and congestion-free transportation network may be achieved by introducing ANPD-based toll systems on highways, urban roads, and parking facilities. This will lead to quicker and more secure transactions.

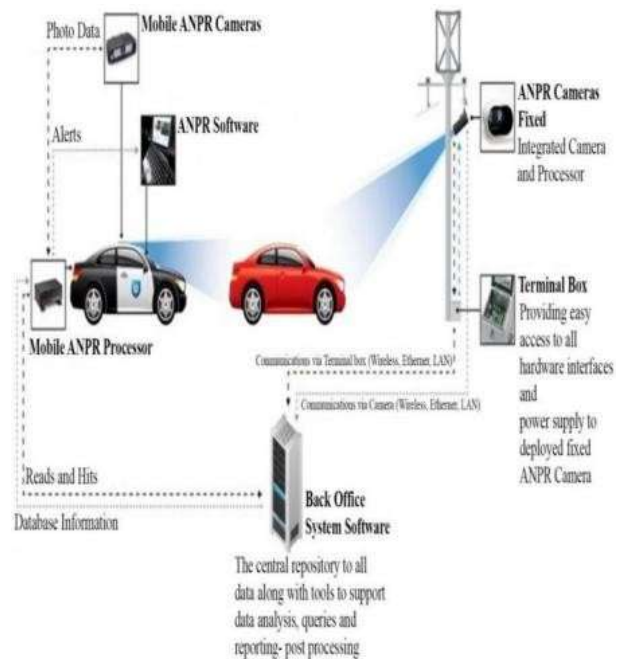
## II. LITERATURE REVIEW

Xiaojun Zhai and Faycal Bensaali [1] proposed an FPGA-based standard definition ANPR system, which demonstrated the feasibility of implementing license plate recognition in hardware for real-time processing. Their system, while effective, was limited by the resolution and scalability to high-definition (HD) environments. Kocer and Cevik [2] introduced an artificial neural network (ANN)-based approach to vehicle license plate recognition, emphasizing the model's learning ability to adapt to diverse datasets. This work highlighted the potential of neural networks in handling various font styles and noisy images, a common challenge in toll booth scenarios. Roy and Ghoshal [3] worked on segmentation improvements to enable ANPR systems across multiple countries with different plate formats. This flexibility is particularly relevant in international highways and border toll systems, where plate designs vary significantly. Ozturk and Ozen [4] developed a system using probabilistic neural networks (PNNs), showcasing the application of statistical models in recognizing number plates with higher confidence levels. Their work provided a robust framework for high-accuracy recognition in variable conditions. Prabuwo and Idris [5] explored OCR in car park control systems, a concept extendable to toll systems. Their research showed how optical character recognition could be integrated into access control, forming the basis of modern automated toll collection systems. Lakshmi et al. [6] introduced a novel method tailored to Indian license plates, accounting for unique regional characteristics. Their system addressed font inconsistencies and background variations, both critical for reliable toll processing in diverse traffic conditions. Jiao et al. [7] proposed a configurable method for recognizing multiple license plate styles, enabling systems to adapt dynamically to regional plate variations. This adaptability is key for building scalable and generalizable toll systems. Zhang and Wang [8] applied a Backpropagation (BP) neural network for vehicle plate recognition, which demonstrated effective training capabilities in handling complex datasets with high accuracy, ideal for real-time toll systems. Wen [9] presented an algorithm suited for intelligent transportation systems, where speed and accuracy are crucial. The algorithm's effectiveness in high-traffic environments makes it highly relevant to toll collection systems that need to process numerous vehicles in real-time. Patel et al. [10] provided a comprehensive survey of ANPR techniques, outlining challenges such as plate damage, lighting variation, and occlusion, offering insights for overcoming practical implementation issues in toll systems. Rafique et al. [11] introduced a toll plaza monitoring system that incorporated space-invariant vehicle recognition, enhancing accuracy in scenarios where camera angles and vehicle positions vary, as is common in toll lanes. Finally, Rahati et al. [12] proposed a method combining the contourlet transform with Support Vector Machines (SVMs) for vehicle recognition. Their approach captured fine edge details, which are essential in distinguishing characters on plates under challenging imaging conditions.

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## III. METHODOLOGY

### A) SYSTEM ARCHITECTURE



The planned toll collecting system employs a Tesseract OCR-based Automatic Number Plate Recognition (ANPR) technology to circumvent the shortcomings of conventional RFID-based systems. By taking pictures of license plates, reading the text, and processing payments without using RFID tags, this technology simplifies the toll collecting procedure. This system offers a more efficient and cost-effective alternative to current toll collecting systems by using computer vision, image processing, and machine learning. First, the system uses the toll gate's high-resolution cameras to identify when a vehicle is approaching. Regardless of the weather or illumination, these cameras can clearly catch the vehicle's license plate.

To make the license plate more legible, the acquired picture is preprocessed using tools including edge detection, noise reduction, and contrast enhancement. As a result, the Optical Character Recognition (OCR) step is more likely to provide accurate results. The license plate's alphanumeric text is extracted from the picture using the Tesseract OCR engine after preprocessing. In order to confirm the car's registration information and if the toll payment account is current, the retrieved text is then compared with a centralized vehicle database. Paying the toll is as easy as linking your bank account, digital wallet, or FASTag system if your car is registered.

After the toll is paid, the boom barrier is opened automatically so the vehicle may travel through without stopping. To prevent illegal cars from avoiding toll costs, an alarm is sent up for human verification in the event that payment fails. Toll plaza traffic flows smoothly since this procedure is totally automated, which drastically cuts down on human interaction. The elimination of RFID tags—and the expenses related with their issue, maintenance, and fraud concerns such tag cloning—is a major benefit of this technology. The ANPR-based toll system allows for entirely contactless and smooth transactions, which improves efficiency and reduces congestion, in contrast to RFID systems, which need that cars briefly slow down or stop in order for the tag to be read. Also, even in low light, with varied number plate forms, or with partly covered plates, the high accuracy is guaranteed by using AI-driven upgrades and deep learning-based optical character recognition models. Problems with the proposed system include changes in license plate forms, low-quality photos caused by external conditions, and the need for processing speed in real-time, notwithstanding its benefits. To overcome these obstacles, one may use deep learning character recognition models in conjunction with Tesseract OCR, improve picture quality using adaptive contrast enhancement methods, and speed up OCR processing with GPU acceleration and multithreading.

## B) IMPLEMENTATION

**Pre-Processing Techniques:** This first stage gets the picture ready for the next one, which will probably include improving the image and reducing oise.

**Number Plate Extraction:** In this step, we employ characteristics such as edge detection, picture information, color features, texture features, character features, and combination features to seek for and isolate the license plate within the picture.

**Character Segmentation:** This stage involves employing techniques like linked component analysis, projection methods, boundary information, extracted features, and integrated algorithms to segment individual characters on the number plate once it has been retrieved.

**Character Recognition:** Optic character recognition (OCR), template matching, artificial neural networks (ANNs), and combination methods are used in this last step to identify the segmented characters.

**Output:** The next stage is to export the detected vehicle number plate data, with the option to do post-processing if necessary.

## IV. EXPERIMENTAL RESULTS AND ANALYSIS

The system was tested using 60 vehicle images captured under different conditions. Table I presents the accuracy of OCR in various scenarios.

Test Scenario	Condition	Accuracy(%)
1	Daylight, front-facing	94
2	Low light, angled	83
3	Night with flash	89
4	Rainy weather	78
5	Shadowed region	85

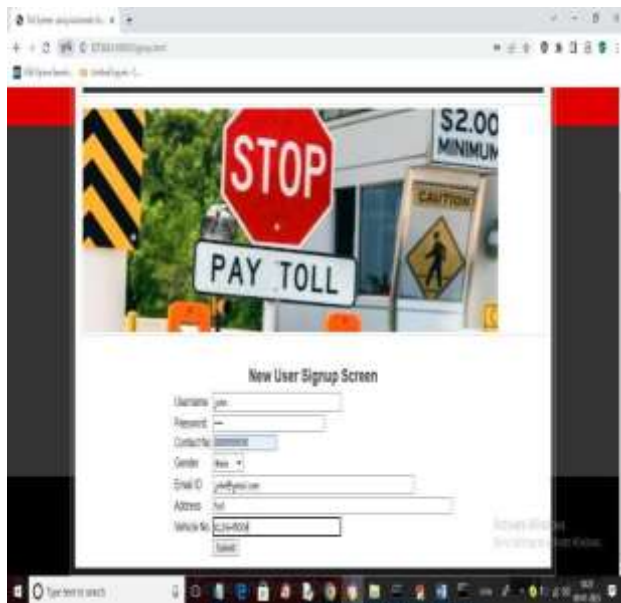


Number plate detection



Home page





Login page



Image upload



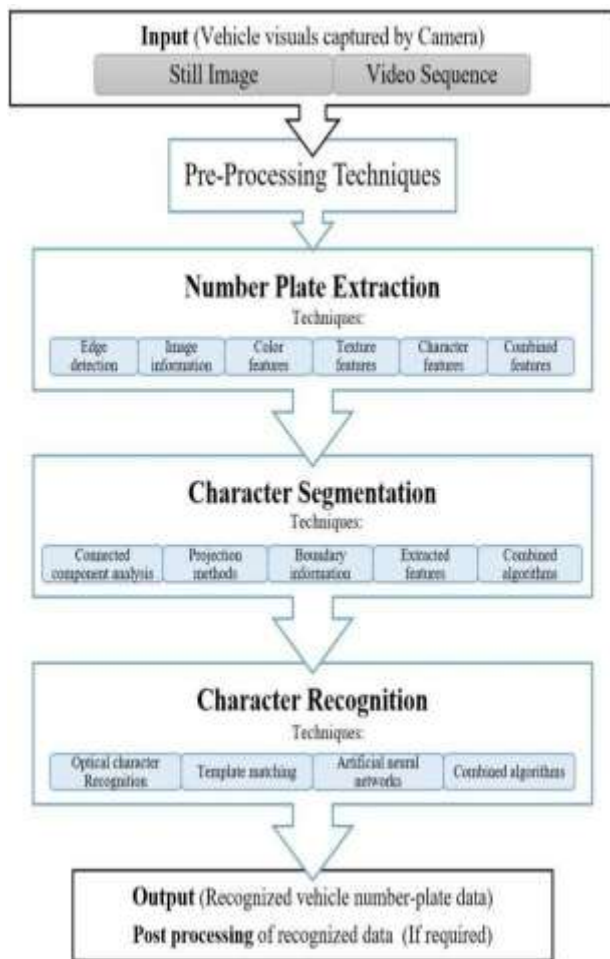
Prediction



Output

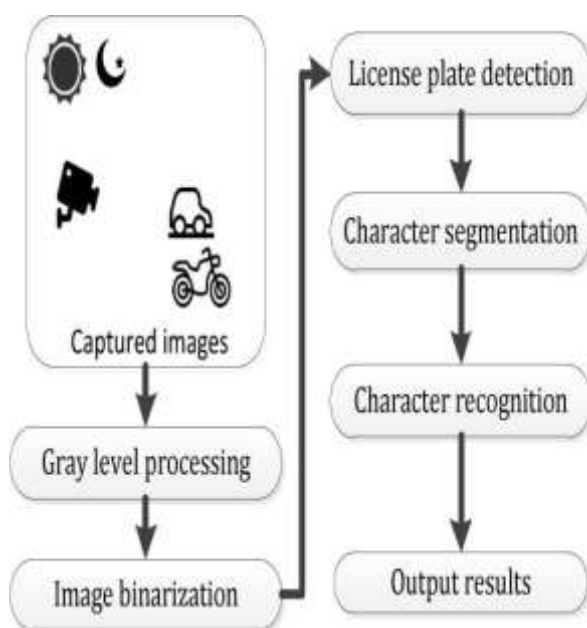
### Logical design:

### Url listing:



Websites	Data collected
<a href="https://wikipedia.org">https://wikipedia.org</a>	Searching of any information that can be used in documentation.
<a href="https://dev.sqlserver.com/doc">https://dev.sqlserver.com/doc</a>	SQL- Server it performing in mainly depending on the one of the database using.
<a href="https://www.answers.com">https://www.answers.com</a>	Answers.com, online dictionary, encyclopedia and much more.
<a href="https://google.co.in">https://google.co.in</a>	Any information searching and downloading.
<a href="https://training-classes.com">https://training-classes.com</a>	Designing part information as Gathered

### Physical design:



## V. CONCLUSION

Finally, compared to conventional rfid and human-operated toll collecting methods, the suggested anpr based system is more reliable, expandable, and efficient. The technology successfully automates toll processing without human involvement or rfid infrastructure by combining high-resolution imagery with powerful image preprocessing and tesseract ocr for optical character recognition. Aside from improving toll collecting efficiency, this technology has wider applications in traffic monitoring, parking automation, and law enforcement.

## VI. FUTURE SCOPE

**Multi-Lingual Support:** To make it more accessible throughout all of India's states, we can include support for regional languages that are used for number plates that aren't printed in English. **Dashboard for Real-Time Monitoring:** One way to enhance system supervision is to establish a centralized dashboard that tracks traffic, identifies anomalies (such as stolen or banned cars), and provides analytics for tolls.

Connectivity to E-Wallets and UPI: Making it easy to pay using digital wallets or UPI may make the system more accessible and convenient for users.

Smart cities and intelligent transportation systems stand to benefit greatly from the development of ANPR technology. Automated, transparent, and efficient road infrastructure may be built upon ANPR-based toll systems with ongoing development and validation in the real world.

## VII. REFERENCES

- [1]. XiaojunZhai, FaycalBensaali, "Standard Definition ANPR System on FPGA and an Approach to Extend it to HD" in 2013 IEEE GCC Conference and exhibition, November 17-20, Doha, Qatar. pp.214
- [2]. H. ErdincKocer and K. KursatCevik, "Artificial neural networks based vehicle license plate recognition," Procedia Computer Science, vol. 3, pp. 1033-1037, 2011 [3]. A Roy and D.P Ghoshal, "Number Plate Recognition for use in different countries using an improved segmentation," in 2nd National Conference on Emerging Trends and Applications in Computer Science(NCETACS), 2011, pp. 1-5
- [4]. FikriyeÖztürk and FigensÖzen, "A New License Plate Recognition System Based on Probabilistic NeuralNetworks," Procedia Technology, vol. 1, pp. 124- 128,2012
- [5]. Anton SatriaPrabuwono and Ariff Idris, "A Study of Car Park Control System Using Optical Character Recognition," in International Conference on Computer and Electrical Engineering, 2008, pp. 866-870
- [6]. Ch. Jaya Lakshmi, Dr. A. Jhansi Rani, Dr. K. Sri Ramakrishna, and M. Kanti Kiran, "A Novel Approach for Indian License Recognition System," International Journal of Advanced Engineering Sciences and Technologies, vol. 6, no. 1, pp. 10-14,2011
- [7]. Jianbin Jiao,Oixiang Ye , and Qingming Huang,"A Configurable Method for multi style license plate recognition."PatternRecognition,vol.42,no.3,pp.358- 369,2009.
- [8]. Zhigang Zhang and Cong Wang, "The Research of Vehicle Plate Recognition Technical Based on BP Neural Network," AASRI Procedia, vol. 1, pp. 74- 81, 2012
- [9]. Ying Wen, "An Algorithm for License Plate recognition Applied to Intelligent Transportation System", IEEE Transactions of Intelligent Transportation Systems. pp. 1-16, 2011
- [10]. Chirag Patel, Dipti Shah, Atul Patel," ANPR: A Survey", International Journal, 2013 [11] SaimaRafique, Mahboob Iqbal and Hafiz Adnan Habib, "Space Invariant Vehicle Recognition for Toll Plaza Monitoring and Auditing System", Multitopic Conference, 2009. INMIC 2009,IEEE 13th International, pp. 1-6
- [11]. SaeidRahati, ReihanehMorvejian, Ehsan M. Kazemi and Farhad M. Kazem "Vehicle Recognition Using Contourlet Transform and SVM," Proceedings of the Fifth International Conference on Information Technology,200