

Vehicle Number Plate Detection and Recognition Techniques: A Review

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ABSTRACT

Vehicle number plate detection and recognition is an integral part of the Intelligent Transport System (ITS) as every vehicle has a number plate as part of its identity. The quantity of vehicles on road is growing in the modern age, so numerous crimes are also increasing day by day. Almost every day the news of missing vehicles and accidents are perceived. Vehicles tracking is often required to investigate all these illegal activities. So, vehicle number plate identification, as well as recognition, is an active field of study. However, vehicle number plate identification has always been a challenging task for some reasons, for example, brightness changes, vehicle shadows, and non-uniform license plate character type, various styles, and environment color effects. In this review work, various state-of-the-art vehicle number plate detection, as well as recognition strategies, have been outlined on how researchers have experimented with these techniques, which methods have been developed or used, what datasets have been focused on, what kinds of characters have been recognized and how much progress have been achieved. Hopefully, for future research, this review would be very useful.

1. Introduction

Vehicle Number Plate Recognition (NPR) or License Plate Recognition (LPR) or Registration Plate Recognition (RPR) is an enhanced computer vision technology that connects vehicles without direct human connection through their number plates [1-3]. Day by day, the number of vehicles on the road is continuing to grow. For this reason, the news spread almost every day about the vehicle being filched from the parking garage or any other place in the city or having an accident and fleeing. To recognize these vehicles [4, 5], authorities should therefore install a number plate detection and recognition device on CCTV at every street corner in every region. This system enhances the police's ability to track illegal activities involving the use of vehicles. NPR systems are effectively used by provincial establishments and manufacturing groups in all facets of safety, inspection, traffic management applications [6, 7].

The number plates vary from country to country. There are some rules and regulations for vehicle number plates. Number plate consists of (1) 2 letters (these refer to the region in the country where the vehicle was first registered) (2) 2 numbers (when it was issued) (3) 3 letters chosen at random. Some basic information about vehicle number plates like dimension, styles, and characters

of number plates fitted after 1st September 2001 is shown in Figure 1.

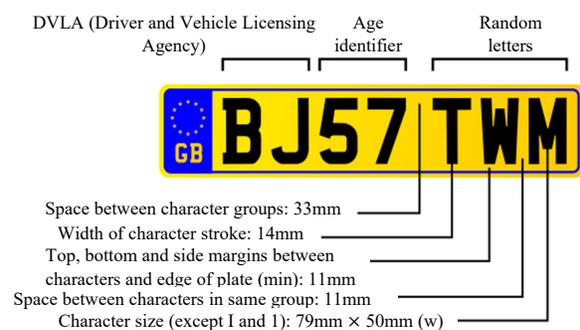


Figure 1: Vehicle number plate fonts and spacing [8]

Some variations are often seen on the vehicle number plates. The difference between American and European number plates is that American vehicle number plates have more things than identification numbers, sometimes little pictures, different color text but in European vehicle plates are used just for identification. Front number plates must show black characters on a white background and the rear number plate must have black letters on a yellow reflective background [8]. The number plates dimension of

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the car and motorcycle in the UK (United Kingdom) is shown in the form of Table 1.

Table 1: Dimension of the vehicle's number plate in UK standard.

Properties	Dimension	
	Car	Motorcycle
Character Height	79 mm	64 mm
Character Width	50 mm	44 mm
Character stroke	14 mm	10 mm
Space between characters	11 mm	10 mm
Space between groups	33 mm	30 mm
Space between vertical lines	19 mm	13 mm

For nearly half a century, vehicle number plate detection, as well as recognition, has been a topic of interest. This technique in the field has opened new challenges. In terms of consistency, color, number plate shape, and type of vehicle, the major challenges of vehicle number plate detection as well as recognition are focused on the various categories of features and are related to changing illumination level, the geometry of visualization, and background [9,10]. In Figure 2, typical samples of vehicle number plates [11] are shown.



Figure 2: Samples of vehicle number plate [11]

Number plate recognition procedure is divided into three key functions: Identification of Plate Area, Segmentation of Plate Character, and Recognition of Character [12-16]. In terms of traffic management, traffic optimization, traffic law enforcement, vehicle access control, automated collection of tolls, traffic speed control, automatic parking, monitoring of stolen cars, and tracking of possible acts of terrorism, each of these aspects plays a crucial role [6, 7, 14, 17, 18].

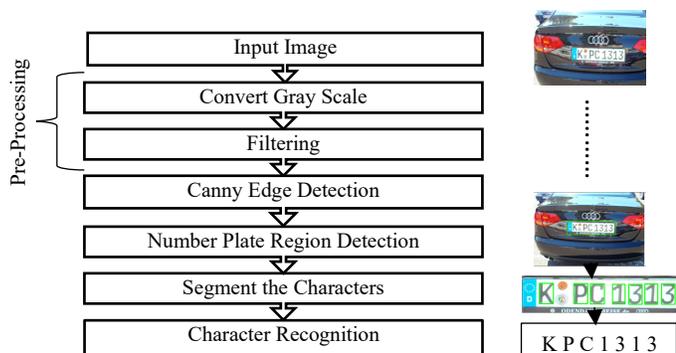


Figure 3: Flow diagram of common number plate detection and recognition method

Figure 3 shows common vehicle number plate detection and recognition method based on the edge detection method. At first, the vehicle registration plate detection as well as the recognition system capture the image using the camera and then apply some

image processing techniques for pre-processing the image such as input image to grayscale image conversion, filtering technique to eliminate noise. Next, to extract the license plate area, apply the canny edge detection technique. After that, apply the appropriate detection method to detect the vehicle registration plate effectively, and apply the segmentation technique to segment the characters of the registration plate. Finally, the appropriate character recognition method is used to recognize each of the characters separately.

Due to the lighting conditions, the noisy image captured, fast-moving vehicles, are always a difficult task in vehicle number plate identification as well as recognition. Several researchers have been working on vehicle number plate recognition and are still working in this field. They have adopted several image processing techniques and presented some of their development strategies for vehicle number plate detection. As much research has been done so far in this paper on vehicle number plate detection as well as recognition and their success behind their proposed method and exactly what caused their proposed method to fail is discussed here. And this paper explores how to resolve their limitations or what more can be achieved in this area in the future.

Vehicle number plate detection studies, as well as recognition techniques, have been categorized into three sections in this review paper: (1) Related Works on Vehicle Number Plate Detection Techniques (2) Related Works on Vehicle Number Plate Recognition Techniques (3) Related Works on Vehicle Number Plate Detection as well as Recognition Techniques.

The residual of the paper is arranged in a structured way. The number plate detection strategies are demonstrated in section 2. Techniques for number plate recognition are discussed in section 3. In section 4, techniques of vehicle number plate detection, as well as recognition, are illustrated. Finally, section 5 states the conclusions.

2. Related Works on Vehicle Number Plate Detection Techniques

Number plate detection (NPD) is a technology that uses certain image features to understand vehicle registration plates to assess location data for vehicles [14,19]. To determine a location going to the next frame, NPD identifies a region of the vehicle number plate with similar structures. The consecutive frame fixes the area of detection in the prior frames with the observed area of the vehicle [20]. During the identification of the registration plate of the vehicle, various difficulties of the surrounding environment were observed. In addition to these, several vehicle number plate considerations are concise in Table 2.

Table 2: Some factors of vehicle number plates [14,21]

Variants of the number plates	Variants of the environment
Plate size	Brightness
Plate background	Similarity in background
Plate location	
Quantity	
Font	
Angle	
Screw	

Different researchers have talked about their proposed techniques for identifying vehicle number plates at different times and still a lot of work is being done following their proposed method. Many image processing techniques are existing to detect vehicle number plates such as segmentation, edge detection, color code-based techniques, feature-based techniques, and machine learning techniques. This section discusses different strategies associated with the identification of vehicle number plates. Centered on various methods, the following section is split into several sub-sections.

2.1. Edge Detection

In image processing, it is possible to recognize the edges of the image through different edge detection techniques, such as Sobel, Prewitt, Laplacian, and Canny edge band detectors. The Sobel edge detector effectively transforms a compact, detachable, and numeral valued filter to the image in a horizontal and vertical direction. Prewitt is used in frames to detect vertical and horizontal edges. Hence Sobel and Prewitt are kind of similar. Canny edge detector probably the most effective method for complex edge detection. Below, discussed the previous literature of vehicle number plate detection techniques based on edge detection.

In [22], an algorithm has been suggested for vehicle number plate detection in practical situations by Wazalwar. To define the region of interest (ROI), they used the Euler number of a binary image and for edge detection, they used the Mexican hat operator. They have claimed that a license plate had been successfully identified through their suggested technique and their success rate was about 94-99% and the average accuracy was about 96.17%. Yet there is a situation during their prosperity where they have suffered. The edge detection system fails to properly recognize the edge if the license plate is black.

In [23], a license plate detection system founded on an enhanced Prewitt arithmetic operator has been suggested by Chen under various backgrounds and lighting conditions. The projection method was also carried out horizontally and vertically to change the top and bottom edge areas along the edge to get the vehicle number position. They have achieved 96.75% precision in their proposed technique, and they have stated that their proposed system meets efficiency in real-time.

An innovative technique for vehicle number plate detection using the special technique of edge detection [24] has been introduced by Tejas. They have used the Sobel edge detection technique to obtain accurate boundaries of the number plate in the image. The system scanned the connected component and then fill them with holes. Thereafter, the system searches the rectangular region that is filled with holes which is probably the size of the license plate and then extracts it. Their proposed system is based on the Internet of Things (IoT). Therefore, online databases have been developed and regularly updated. They have also estimated that the accuracy of their acquisition is around 96.842%. In Figure 4, their suggested technique is shown.

2.2. Morphological Operation

Morphological Operations in image processing attempts to remove these imperfections by considering the image's shape and structure. To reduce noise or to brighten the frame, morphological

operations are essentially applied to grayscale images. Morphological operations are referred to as a blend of erosion, dilation, and basic set-theoretical functions, such as a binary image supplement [25]. The corresponding study on morphological operation-based vehicle number plate detection techniques has been discussed below.

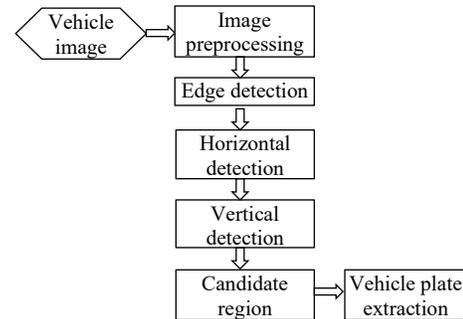


Figure 4: Block diagram of the proposed method [24]

In [26], an existing system used for license plate location on a Raspberry Pi has been improved by the Yepez. Their improved morphological algorithm that reduces computational complexity is based on morphological operations. The strength of this strategy is that the emerging LPR algorithm can operate with the computer as well as low processing power on portable devices. They have also claimed that their enhanced algorithm can detect license plates effectively and have achieved a high precision is about 98.45%. In Figure 5, the flowchart of their proposed method is shown.

2.3. Convolutional Neural Networks (CNNs)

A convolutional neural network (CNN) based framework for the detection of vehicle number plate was proposed by the authors in [27]. They have enhanced the existing blurred and obscure image method. They believed that their suggested method effectively detects the number plate of the vehicle under various lighting conditions. The accuracy obtained by their proposed method is around 100%.

2.4. Machine Learning (ML) based Approaches

Machine Learning (ML) likewise means that by providing a collection of training data, the machine is trained to do something in image processing. Machine learning has models/architectures, functions of loss, and many methods that can be used to decide which will provide better processing of images. For image enhancement, this approach is commonly applied. The corresponding work of machine learning-based vehicle number plate detection techniques is given below.

In [5], a new technique to detect a vehicle authorization plate has been developed in the Miyata study. The license plate detection technique detects only the edge vertical parts and the candidate license plates that use the contours acquired by dilation and erosion processing and area fill processing. The SVM (Support Vector Machine) has applied to decide whether a license plate is a candidate region or not, and eventually recognizes the location of the license plate. They have claimed that the suggested method efficiently detects license plates and achieved the rate of detection is 90%.

In [28], an innovative method has been presented for detecting and locating a vehicle's license plate in color images by Yaseen. AdaBoost, a multi-boosting model based on HOG features, is part of the development process. They have claimed that the accuracy achieved by their proposed method is around 89.66%. Figure 6 displays the flow chart of their suggested system.

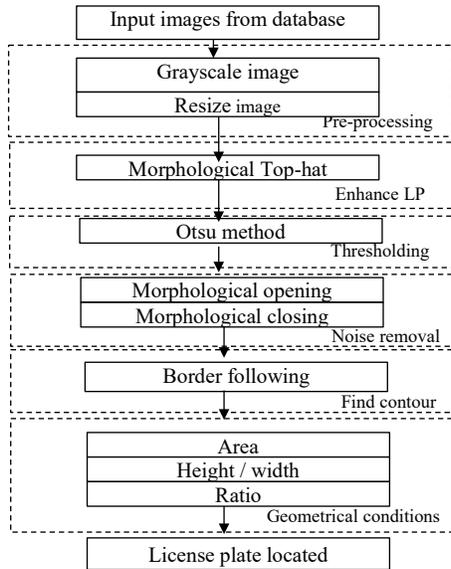


Figure 5: Flowchart of the proposed method [26].

In [29], a vehicle registration plate detection method in the natural image by AdaBoost using the Modified Census Transform (MCT) feature has been suggested by Ban. There are too many noises in the natural image, so detecting the number plate in natural images is too difficult. MCT features, which are robust to illumination change, and AdaBoost for the feature selection to overcome this restriction. They have also reported that the high detection rate achieved by the proposed technique is about 98.7%.

In [30], a systematic style for vehicle registration plate detection applying boosting and part-based models was proposed by Molina-Moreno, which is an algorithm for boosting. They used two datasets and stated that better performance on these datasets was obtained 97.52% accuracy by their proposed method. With several algorithms, they have also compared their proposed method.

In [31], a novel vehicle number plate detection system has been suggested to improve identification in low lights and over corrosive environments by Babbar. For the extraction of license plates, they used CCA (Connected Component Analysis) and Ratio Analysis (RA). Some OCR strategies have also been used, for example, LR+RF, SVC+KNN, Extra Trees, SVC (Linear, Poly, Rbf, Linear.svc). They stated that the car localization achieved by the developed system is 92.7% and the segmented characters' accuracy is about 97.1%.

In [32], a KNN (K-nearest Neighbor) machine learning system for automatic vehicle license plate detection was developed by Akshay Iepcha. The KNN classifier has been used according to the aspects of the license to retrieve the registration plate from the image. They have also stated that a license plate is correctly identified through their suggested method and achieved an accuracy higher than 90%.

Table 3 provides an overview of the strategies for detecting the vehicle number plate. This table has been sorted based on the year and accuracy.

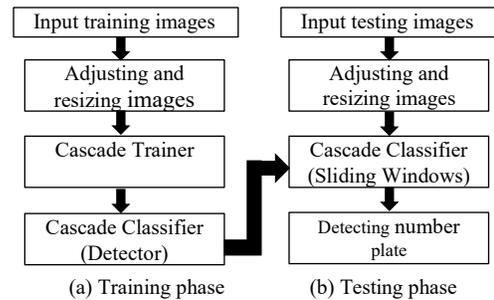


Figure 6: Flowchart of the proposed ANPD system [28].

3. Related Works on Vehicle Number Plate Recognition Techniques

Vehicle Automatic Number Plate Recognition (ANPR) is a technology applied for the observation as well as recognition of vehicle number plate characters from static and moving vehicle images [14, 28, 32, 33]. Due to its effect on the rapid development of traffic monitoring and surveillance [15, 22, 34, 35], vehicle number plate recognition has become a key research field in recent years. For the identification of number plates, several methods are used, such as machine learning, neural networks, BAM (Bidirectional Associative Memories) [35]. Various researchers have given their useful ideas on their proposed vehicle number plate recognition method at various times. In this review paper, various vehicle number plate recognition techniques have been explored. Vehicle number plate recognition techniques have been categorized into some subsections based on distinct approaches in the following section.

3.1. Neural Network (NN) based Approaches

Image recognition algorithms in neural networks (NN) can recognize anything, from text to images, audio files, and videos. Neural networks are an interlinked set of neurons or perceptron's called nodes. Each node uses a single input data, generally a single pixel of the image, and uses a simple calculation called an activation function which produces results and each neuron has a numerical score that determines its outcome

1. Artificial Neural Networks (ANNs)

In [15], a high-performance-based system for vehicle number plate recognition has been introduced by Türkyılmaz. They have applied edge-based image processing techniques for registration plate detection and have also used a three-layer feedforward artificial neural network for vehicle number plate character recognition using a learning algorithm for back-propagation. The feedforward ANN model for three layers is shown in Figure 7. The input layer receives information from the external environment and transmits it to the nodes (processing units) of the hidden layer without any modification. Network outputs are calculated by processing information in hidden layers and output layers. The most well-known back-propagation learning algorithms are used efficiently at the training stage of this ANN. The authors have verified that the number plate has been successfully identified and recognized by their developed system and their performance rate

Table 3: Summary of the vehicle number plate detection techniques

First Author & Year	Detection Methods	Datasets	Accuracy	Advantages	Limitations	Future Opportunities
Eswar, 2020 [27]	Convolution Neural Network (CNN)	Private dataset	100%	Able to detect number plate at different lighting conditions and processing time will be less.		
Molina-Moreno, 2019 [30]	Scale-adaptive deformable part-based boosting algorithm	Caltech, LPR and MMR Database	97.52%		The proposed system, due to the presence of noise, lack of lighting, and blurring of remote license plates, is not high enough in many realistic scenarios.	To improve the process of segmentation, morphological structures can be assumed.
Yaseen, 2019 [28]	AdaBoost based HOG features	North Iraq Vehicle Images (NI-VI) dataset,	89.66%	The dataset must cover all real-life vehicle conditions likely to start, such as weather conditions, size, color, and license plates.		This dataset can be further used for Automatic Number Plate Recognition systems.
Yepez, 2018 [26]	Morphological operations	MediaLab LPR database	98.45%	Able to work on both a computer and low power portable device.		Any other image processing technique can be applied for further improvement.
Babbar, 2018 [31]	CCA, Ratio analysis, LR+RF, SVC+KNN, Extra Trees, SVC (Linear, Poly, Rbf, Linear.svc)	Vehicles at IIIT institution	Detection rate 92.7%, accuracy 97.1%	The system is successfully detecting number plates from skewed angles.		Perhaps this system can be improved by locating the reversed vehicle number plate in the event of an accident and warning the nearest hospital and policing station of the accident, thus saving lives.
Akshay Iepcha, 2018 [32]	KNN Classifier	Their own dataset with videos	Higher than 95%	Improved performance is 11%.		
Tejas, 2017 [24]	Sobel edge detector	Their own dataset	96.842%	Proposed system makes easier to update database.		Genetic algorithm can be applied for better performance and web application can be integrated.
Miyata, 2016 [5]	Support Vector Machine (SVM)	Their own dataset with 100 images.	90%		The detection rate is significantly influenced by the luminosity of the body of the vehicle license plate.	The detection rate can be increased by improving brightness or other features.
Chen, 2012 [23]	Prewitt operator for edge detection		96.75%	Suggested system performs in real-time.		Different edge detection method can be used.

Ban, 2012 [29]	AdaBoost based on Modified Census Transform (MCT) features	Their own dataset with 3373 LP images	98.7%		Proposed method is failed to detect the numbers, which have different width/height ratio when the training stage.	The shortcomings can be improved in the future.
Wazalwar, 2011 [22]	Mexican hat operator for edge detection	Medialab LPR Database	96.17%		Black license plate cannot detect the edge properly.	Motion analysis can be applied to overcome failure.

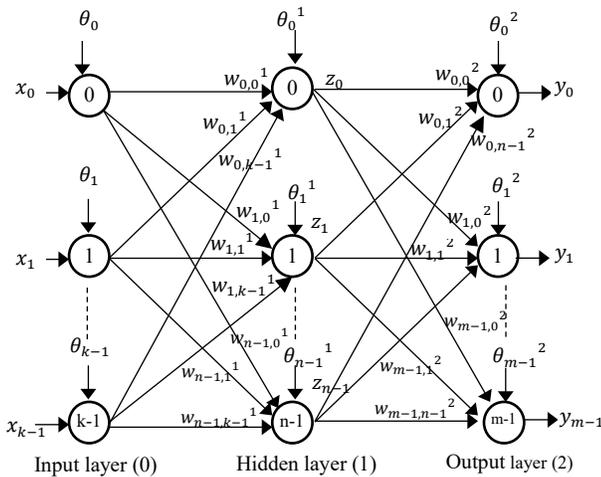


Figure 7: Three-layer feedforward ANN model [15]

II. Convolutional Neural Networks (CNNs)

CNN is familiar to describe the characters that appear in the segmented License plates (LPs). CNN consists of, as seen in Figure 8, a set of layers of conv (Convolution), pooling, and fully connected (FC) layers [36].

In [11], a system for automatic number plate recognition applying convolutional neural networks (CNN) centered on self-synthesized features was proposed by Mondal. The self-synthesized feature of CNN can recognize the states of the vehicle from the number plate. They have confirmed that their system is robust and effective with accurate identification of the license plate of the vehicle from the images above 90%.



Figure 9: Vehicle number plate recognition results using dataset [37].

In [37], a set of vehicle number plate recognition techniques has suggested by Yang. They have first introduced a contour

reconstruction method with edge-detection to accurately detect the number plates and then used a zero-one-alternation technique to effectively remove the misleading top and bottom borders around plates to allow more precise character segmentation on plates. Subsequently, for character recognition, a convolutional neural network (CNN) was applied. Furthermore, the SIFT (Scale Invariant Feature Transform) feature has been used in CNN for successful training. SIFT is a feature detection algorithm and it helps to locate the local features in an image. Finally, a two-phase verification approach has been implemented, the first phase is a statistical filter in the LPD phase to effectively remove the wrong plates and the second phase is shortening the system pipeline, which increases the LPDR system's performance. They have confirmed that the intended method essentially recognizes the vehicle number plate in real time and achieved a precision rate is about 84.3%. Figure 9 displays the recognition results of the proposed system.

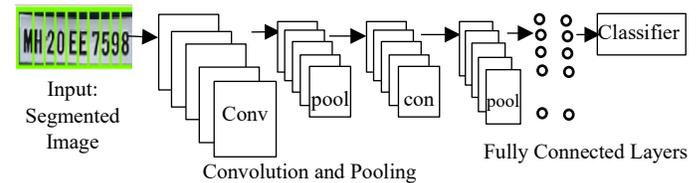


Figure 8: The design of CNN for character recognition [36].

In [38], an interference occurrence on CNN classifiers in the License Plate Recognition Systems (LPR) was introduced in the study of Qian, which adds pre-arranged alarms to definite parts of license plate images, pretending to have indeed formed spots. They have used the genetic algorithm technique to enhance the difficult issues. During vehicle number plate identification, spots that are not usually accessible to humans will be at great risk at any point. They have argued that they were able to identify the number plate character effectively despite getting several spots and their performance rate is 93%.

In [39], a framework for the identification of vehicle license plates on urban roads focused on vehicle tracking and data integration was implemented by Zhu. An object detection framework is trained, centered on a plate detector, to detect each vehicle's license plate from the video series. The convolutional neural networks (CNN) have been applied for vehicle registration plate recognition from the video sequences. Besides, the continuous frames have combined recognition effects to achieve the result. The proposed LPR system layout focusing on vehicle tracking and outcome incorporation is shown in Figure 10. They claimed that under the real urban road climate, their license plate detection accuracy and recall were 82.5% and 89% respectively.

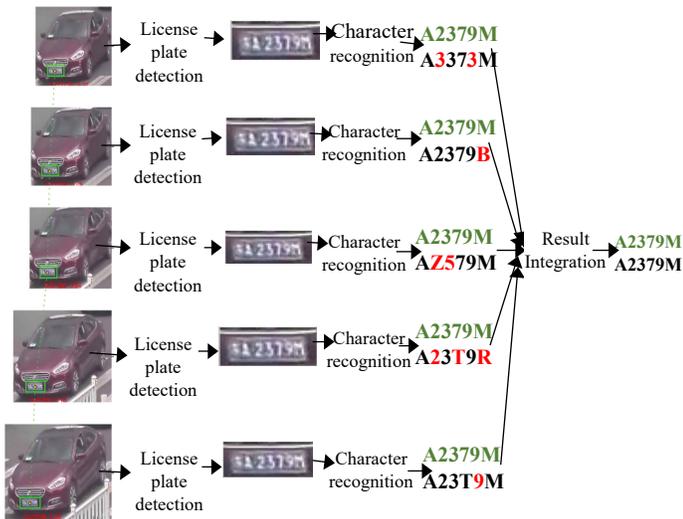


Figure 10: Structure of the LPR system centered on vehicle tracking and result integration [39].

In [40], the authors presented a system of image de-noise supportive of defining the license plate of the vehicle. They also combined a new de-noising and rectification approach conducted by CNN that focuses on jointly solving both problems. They argued that their proposed approach effectively recovers the image issues of the low-quality license plate and identifies the character successfully. They claimed that their proposed method achieves 93.08% accuracy for detecting the license plate.

III. Generative Adversarial Networks (GANs)

In [41], a new method of making text images of high-resolution license plates has been introduced by Gupta where the style and textual content of the images are parametrically represented. To acquire the conditional generation of photo-realistic images, the proposed system combines text to image recovery techniques with Progressive Growing of Generative Adversarial Networks (PGGANs). They have used the American license plate dataset for the proposed system and achieved higher accuracy to recognize the registration plate characters.

In [42], a method of registration plate recognition for speeding vehicles using a motion camera was proposed by Wang, implying whether something is feasible to create synthetic training data using GAN to enhance identification precision. They used a Deep Convolutional Neural Network (DCNN) accompanied by a long-term short-term memory (LSTM), Bidirectional Recurrent Neural Network (BRNN), which performs the learning function and sequence labeling. They reported that the recognition accuracy achieved by the proposed system was 89.4% for moving cars on demanding test datasets.

In [43], a method for registration plate recognition in the natural environment has been suggested by Zhang. Their suggested method contains a customized model of Cycle GAN for license plate image generation. They have employed a 2D attention plate recognizer with an Xception-based CNN encoder which can reliably and efficiently differentiate license plates with different designs. Four datasets were also used by them to assess the efficiency of their proposed framework and achieved an accuracy higher than 80%.

IV. Recurrent Neural Networks (RNNs)

In [44], a combined ConvNet-RNN model was developed by Cheang to identify legitimate captured registration plate images. To develop feature extraction, a Convolutional Neural Network (ConvNet) is included. For computation, a Recurrent Neural Network (RNN) has been applied. They tackled this problem by promoting the whole image as a contribution to ConvNet, sliding windows could not access the whole image context. They have confirmed that the combined model achieved over 76% accuracy in recognizing the license plate characters in their dataset, with a per-character accuracy of 95.1%.

3.2. Computer Vision (CV)

In [45], an algorithm based on computer vision technology for automatic vehicle license plate recognition has been introduced by Akila. The suggested system addresses various lighting conditions by capturing the image file collected at different times. They used Optical Character Recognition (OCR) to retrieve the numbers from the number plate. They have tested their proposed system for different data with different characteristics of number plates, such as black, inverted color, bold or stylish pattern. Recursive subdivisions have been used to extract character image features. They have stated that the proposed system was successfully identical, extracted, and segmented by the license plate, and achieved a higher, good, and acceptable rate.

3.3. YOLO (You Only Look Once)

YOLO is an actual algorithm for object detection, among the most effective and significant object detection algorithms, which integrates several pioneering ideas in computer vision from the scientific community [46]. All of the previous algorithms for object detection use regions to locate the object within the image. YOLO greatly differs from region-based algorithms. The bounding boxes and the class probabilities for these boxes are predicted by a single convolutional network in YOLO. YOLO is faster (it can deal with 45 frames per second) than other algorithms for object detection. But the YOLO algorithm is limited by the fact that it manages with small objects within the image.

In [47], a robust and efficient YOLO object detector-based ALPR system has been implemented by Laroca. They have used an inverted License Plates (LPs) system for the segmentation and identification of characters applying basic techniques for data improvement. Both Fast-YOLO and YOLOv2 models were evaluated at this point to be able to handle simpler (i.e., SSIG) and more realistic (i.e., UFPR-ALPR) data. For simpler situations, Fast-YOLO should be able to correctly detect vehicles and their LPs in a much shorter time. The resulting ALPR process has also obtained crucial results in two datasets. They reported that their system achieved a recognition rate of 93.53%.

In [48], an efficient and effective YOLO object detector-based layout-independent Automatic License Plate Recognition (ALPR) framework has been suggested by Laroca that includes a coherent technique for detection and layout classification of license plate (LP). In their proposed ALPR system, they performed experiments with the Fast-YOLOv2 and Fast-YOLOv3 models. In the validation set, Fast-YOLOv2 obtained slightly better results than its successor. This is since YOLOv3 and FastYOLOv3 have

relatively high performance on small objects but comparatively worse performance on medium and larger size objects. Eight public datasets were used by them and many data augmentation techniques were used on the datasets. They have confirmed that an overall identification rate of 96.8% on the datasets was reached by the proposed method.

In [49], an inventive vehicle license plate location system using the latest YOLO-L model and pre-identification plate was developed by Min. The proposed model modifies two parts to discover the area of the license plate precisely. The k-means++ clustering algorithm was first used to choose the appropriate size and amount of the candidate boxes for a plate. Next, the YOLOv2 network model and depth were amended. To separate license plates from related items, they also used a plate pre-identification algorithm. They have claimed that precision of 98.86% and a recall of 98.86% were achieved by the proposed method.

In [50], a global vehicle license plate recognition scheme has been suggested by Henry. The intended method was founded on the YOLOv3 networks. The suggested system consists of three key steps: 1) identification of License Plate (LP), 2) recognition of unified characters, and 3) detection of global LP layouts. They used their Korean dataset to test their success and used the dataset of the other four countries. They have confirmed that the proposed ALPR method consumes an average of around 42ms per image to extract the number of LPs and achieves an accuracy of up to 90%.

3.4. Deep Learning (DL) based Approaches

In [51], a system for vehicle license plate recognition in complex environments using a deep learning approach was suggested by Weihong. First, more sophisticated algorithms for key issues such as skewing of the license plate, image noise, and blurring of license plate were implemented. Then the deep learning algorithms were listed as algorithms for direct detection and indirect detection, and the detection and recognition of license plates and algorithms were analyzed. Besides, contrasts were made between the variations in data sets, workstations (special computers that provide higher performance, graphics, memory space, and multitasking capabilities), precision, and time complexity of various license plate recognition systems. Finally, the existing public datasets of license plates were compared and illustrated as per the set of images, the resolution, and the sophistication of the area. They reported that their model achieved a segmentation rate of 82.6% and recognition precision of 87.3%.

In [52], an innovative deep learning-based vehicle registration plate recognition approach for general road surveillance cameras is presented by Elihos. In the character detection and recognition process, the suggested free segmentation license plate recognition technique employs deep learning object detection techniques. They used their private dataset containing 2000 images captured on a highway, which were tested. They also stated that the overall accuracy of the proposed achievement is 73.3%.

In [53], an SSD (Single Shot Detector) based natural environment registration plate recognition system has been suggested by Yao. The proposed LPR-SSD network is composed of two networks centered on SSDs. The proposed method is subdivided into two sections. The first part consists of locating and classifying the plate detection, and the second part is to locate and

identify character recognition. They reported that the LPR-SSD achieved a greater acceleration in testing and the accuracy of identification and classification of license plate location exceeded 98.3% and the accuracy rate of character recognition exceeded 99.1%.

3.5. Image Processing Techniques

I. Edge-based Approach

In [54], a system for automatic license plate recognition founded on integrated edge-based Connected Component Analysis (CCA) techniques was proposed by Arafat where license plate identification, segmentation, and recognition of different shapes have focused. They ensured that better character segmentation was accomplished by the proposed approach and that 96.5%, 95.6%, and 94.4% were correct for identification, segmentation, and recognition respectively.

II. Gradient Segmentation

In [55], a system for vehicle license plate tracking through gradient-based segmentation was developed by Kumar. Gradient-based segmentation adjusts the lighting level of the image to ascertain the position of the license plate. The proposed approach filters the region of interest using the Hue, Saturation, and Value (HSV). They also ensured that the proposed system accurately tracks the vehicle's license plate to recognize the registration plate characters and achieved the precision is about 94%.

III. Optical Character Recognition (OCR)

In [56], an automated number plate recognition system manipulating image processing techniques was introduced by Kashyap. To recognize the characters on the license plate, Optical Character Recognition (OCR) converted the lettering on the number plate image to text. They have achieved accuracy is about 82.6%.

In [57], an effective process for automatic license plate recognition was intended by Pechiammal. The proposed method consists of three portions: segmentation of characters, identification of optical characters, and matching of models. They have demonstrated that the suggested method effectively extracts character from the plate and 85% is the extraction rate.

In [58], an innovative vehicle number plate recognition method using OCR and template matching strategies for the Pakistani language has been suggested by Rehman. Several real-time images from different formats of number plates used in Pakistan were evaluated by the proposed ANPR system. They stated that for law enforcement agencies and private organizations to enhance home security, the ANPR model has both time and money-saving profit. They reported that 93 % accuracy of their proposed ANPR approach was achieved. This system can be further expanded to identify the number plate of the crashed vehicle in an accident and warn the nearest hospital and police station about the accident, thereby protecting the number plate of the accident.

3.6. Feature Extraction Technique

In [59], an innovative method was intended for the framework of vehicle registration plate recognition based on compressive sensing techniques using reduction of dimensionality and

extraction of features by Joki'c. To extract the features, they used the Support Vector Machine (SVM). They announced that the proposed method has achieved an average accuracy is about 98.81%.

3.7. K-means Clustering-based Approach

In [36], an efficient deep learning-based approach to recognition plate for vehicles, including appropriate optimal K-means clustering segmentation and Convolutional Neural Network (CNN), was implemented in the research of Pustokhina. Optimal K-means clustering is used for segmenting the license plate and a Convolutional Neural network is used for recognizing the license plate characters. The Bernsen Algorithms (IBA) and the Connected Component Analysis (CCA) models were used to classify and locate the license plates. They have reported that the maximum accuracy obtained by the proposed Optimal K-Means with Convolutional Neural Network (OKM-CNN) system on the datasets is about 98.1%.

3.8. Genetic Algorithm (GA) based Approach

In [60], introduced the latest approach to image-processing algorithms and the optimized genetic algorithm (GA) of the Neutrosophic Set (NS) by Yousif. Certain techniques including

edge detection and morphological localization were initially introduced. Besides, they also used a new method using a new approach to optimize the (NS) operations for extracting the most salient features (GA). Furthermore, the clustering algorithm k-means was introduced for the segmentation of (LP) characters. Finally, the Connected Components Labeling Analysis (CCLA) algorithm has been used to identify the associated pixel domains and the labeling accuracy obtained by the efficiency of the suggested new method was 96.67% for Arabic-Egyptian (LP) and 94.27% for English (LP) and that the computations in both databases had an estimated completion time of approximately 0.996 seconds. Language is the most important factor to recognize characters. Each researcher uses different methods for the different languages for which the recognition rate varies. But English is the common language, and a very good number of techniques for English language recognition compared to other languages. Although the Arabic-Egyptian language is harder to recognize than English, the reason for the higher recognition rate is the image resolution.

Table 4 provides an overview of the techniques used to recognize vehicle number plates. This table has been sorted based on the year and accuracy.

Table 4: Summary of the proposed vehicle license plate recognition methods.

First Author & Year	Recognition Methods	Recognition Character	Datasets	Accuracy	Advantages	Limitations	Future Opportunities
Pustokhina, 2020 [36]	OKM-CNN, Improved Bernsen Algorithm (IBA), CCA	English	Stanford Cars, FZU Cars and HumAIn.	98.1%	Performs in real time.		Multilingual LPs can be increased to recognize the efficiency of the OKM-CNN model.
Yousif, 2020 [60]	Neutrosophic set (NS) based Genetic Algorithm (GA), K-means Clustering, CCLA, edge detection	Arabic – Egyptian, English	Private dataset, Media Lab benchmark LP and AOLP benchmark LP dataset	96.67% for Egyptian and 94.27% for English	Easily recognizes Arabic or Egyptian characters as well as English characters.		Optimization techniques such as particle swarm, ant colony, chicken swarm, and fuzzy techniques can be added.
Arafat, 2020 [54]	Connected component analysis, integrated edge based technique	English	Malaysian LPs	94.4%	For real-time applications, this technique is useful.		In the future, it is possible to recognize font similarity issues in LP characters using the DL architecture.
Rehman, 2020 [58]	OCR, Template Matching	Pakistani	Private Dataset contains 900 images	93%		The identification rate of their proposed scheme is lower for unclear plates, blurring and non-standard vehicle number plates.	The accuracy can be improved, and this system can be further expanded to identify the number plate of the crashed vehicle

Henry, 2020 [50]	YOLOv3 Networks	Korean and English	KarPlate, AOLP, Caltech Cars, Medialab LPR, University of Zagreb	Higher than 90%	The proposed scheme is applicable to the license plate for vehicles in several countries.		
Weihong, 2020 [51]	Deep learning approaches	English and Chinese	Caltech Car, English LP, Chinese LP, UFPR-ALPR	87.3%			An algorithm with an image deblurring and plate correction can be implemented, or the license plate detection rate can be increased.
Zhang, 2020 [43]	CycleGAN model, Xception-based CNN encoder	Chinese	CCPD, AOLP, PKUData, CLPD	More than 80%		Images with extreme blur or occlusion are unable to recognize.	A transformer-like decoder may be explored to accelerate training speed.
Yao, 2019 [53]	SSD based approach	Chinese	Their own dataset contains 16 types of license plates	99.1%	Efficiency of the proposed system is real timing.		
Jokić, 2019 [59]	Compressive Sensing Technique, SVM	English	Character Image set in CV toolbox for matlab	98.81%	The proposed system has great performance in classification.		
Laroca, 2019 [48]	Fast-YOLOv2 and Fast-YOLOv3 models.	Chinese and English	Caltech Cars, EnglishLP, UCSD-Stills, ChineseLP, AOLP, OpenALPR-EU, SSIG-SegPlate, UFPR-ALPR	96.8%	Proposed system achieved an impressive trade-off between accuracy and speed.		Further optimization the system can be used a new CNN architecture.
Kumar, 2019 [55]	Gradient based Segmentation, Edge detection techniques	English	Their own dataset contains 78 images.	94.87%	This system to be helpful for the security of the vehicles.	This system could not extract the license plate with a yellow base.	This technique can be applied for any type of character segmentation and recognition.
Lee, 2019 [40]	Image De-noising, Rectification, CNN	English	AOLP-RP and VTLPs dataset	93.08%		Some cases LPR makes a mistake in detection and classification.	Adjacent context can be added in the future.
Zhu, 2019 [39]	Convolutional Neural Networks (CNN)	Chinese	Their own dataset, contains 19020 images	82.5%	This method is feasible and accurate in real time.		Performance can be further improved.

Elihos, 2019 [52]	Deep learning techniques	English	Private dataset	73.3%		In weak character signals, the proposed system cannot detect properly.	It is possible to apply adequate methods of deep learning-based object classification.
Akila, 2019 [45]	Optical Character Recognition (OCR), Recursive subdivisions	English	Private dataset	Achieved higher rates.	This type of technology involves identifying vehicles that are unknown.	There is no classification between the customer and the visitor in this system.	In the future, to identify the visitor, a separate scanner will be installed.
Laroca, 2018 [47]	Fast-YOLO, YOLOv2, and CNN	English	SSIG and UFPR-ALPR.	SSIG: 93.53%, UFPR: 78.33%		For certain real-world ALPR applications, this outcome is still not acceptable.	In the future, character segmentation and recognitions techniques can be improved.
Gupta, 2018 [41]	Progressive Growing of Generative Adversarial Networks (PGGANs)	English	American license plate dataset	More than 90%		Quality of synthesized images suffers when there are too few samples of a given style in the training data.	This system can be used in more complex scene text synthesis.
Kashyap, 2018 [56]	Image processing techniques, OCR	English		82.6%			Multi-level genetic algorithms can be added for further improvement.
Türkyılmaz, 2017 [15]	Edge-based method and three-layer feedforward ANN	English	Their own database contains 357 images	97%	The developed system performs in real-time.		Advance image processing techniques can be applied.
Cheang, 2017 [44]	CNN and RNN (Recurrent Neural Network),	English	Their own Malaysian VLP dataset contains 2713 images	95.1%	This system performs in real time.		Substituting for long-term short-term memory (LSTM) for the CNN module would improve performance.
Mondal, 2017 [11]	CNN based self-synthesized feature learning algorithm	English	Their own dataset contains 800 images	90%	This system runs on automation.		System performance can be done in real-time.
Wang, 2017 [42]	GAN (Generative Adversarial Networks), DCNN, BRNN (bidirectional recurrent neural network), LSTM (long short-term memory)	Chinese	Dataset1 contains 203774 images and dataset2 contains 45139 images	89.4%	The significance of GAN is magnified when real annotated data is limited.		Accuracy can be improved in the future.

Pechiamal, 2017 [57]	Image Processing Techniques	English		85%	Low processing time.		An influential ANPR framework can use used to manage multi-style plates.
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4. Related Works on Vehicle Number Plate Detection and Recognition Techniques

In [61], a new system for detecting and recognizing the Indian vehicle number plate has been suggested by Varma that can compete with noisy, low-light, cross-angled, non-standard font number plates. This work uses many image processing techniques in the pre-processing stage, such as morphological transformation, Gaussian smoothing, and Gaussian thresholding. They have used the K-nearest neighbor (KNN) approach for recognizing the character. They have stated that their proposed system achieved 98.02% accuracy for vehicle number plate detection and 96.22% accuracy for character recognition.

In [62], Automatic Number Plate Detection (ANPD) and Automatic Number Plate Recognition (ANPR) systems were intended for the detection and recognition of vehicle number plates in the research of Yaseen. A new realistic vehicle image dataset for three cities, called North Iraq-Vehicle Images (NI-VI), has been presented (Duhok, Erbil, and Sulaimani). Three types of images, such as rotated, scaled, and translated, are included in the collection of data.

In [63], the latest approach to identify and recognize the license plate centered on a hybrid feature extraction model and BPNN,

which is adaptable in poor lighting and complex contexts, was introduced by Xie. They reported that the accuracy achieved by the proposed technique is 97.7% and the processing time is 46.1ms.

In [64], a full unregulated scenario ALPR method has been proposed and implemented a new Convolutional Neural Networks (CNN) to detect as well as recognize the number plate of the vehicle in an input image. To identify the character, they used OCR technology. They have reported that an average accuracy of more than 80% was reached by the proposed method.

A systematic technique was developed in [65] for the identification, segmentation, and recognition of characters within the license plate. To extract the characters from the number plate, they utilized Hough Transform and horizontal projection. They ensured that more than 90% higher accuracy was reached by the proposed system.

In [66], a Bangla license plate recognition system based on Convolutional Neural Networks was suggested by Shaifur Rahman, which could be used for various purposes, such as roadside assistance, vehicle license status identification. Six CNN layers and a fully connected layer were used by the authors for training. They have reported that 89% testing precision was achieved by the proposed Bangla license plate recognition system (BLPRS).

Table 5: Summary of the proposed vehicle number plate detection and recognition techniques.

First Author & Year	Detection & Recognition Methods	Recognition Character	Datasets	Accuracy	Advantages	Limitations	Future Opportunities
Alam, 2021 [68]	CNN and Deep Learning	Bengali	VLPR vehicle dataset	98.2%	This system is used for smart cities.		The system can be used for LP in other languages.
Varma, 2020 [61]	Morphological transformation, Gaussian smoothing, Gaussian thresholding, and KNN	Indian	Private dataset	Detection: 98.02% and Recognition: 96.22%		When font size of LP is smaller, the suggested method gave poor prediction.	In the future Convolutional Neural Network can be integrated that incorporates both detection and recognition into a single structure.
Onim, 2020 [69]	YOLOv4, CNN, Tesseract (OCR engine)	Bengali	Private dataset	90.50%		When it is under shade or under direct sunlight, their proposed system fails to detect VLP.	To reduce the effects of blurry VLP and by deploying preprocessing, to overcome the deterrents of OCR.
Yaseen, 2019 [62]	ANPD and ANPR technologies.	Arabic	North Iraq (NI-VI) dataset	-----	Provides a realistic dataset.	The proposed data set is connected to only north Iraq vehicle license plates.	In future, the research for the entire country of Iraq can be strengthened.

Xie, 2018 [63]	Feature extraction model and BPNN	Chinese	Private dataset	97.7%	This system to be helpful for real time applications.		With RFID devices and Bluetooth devices, this work can be enhanced to better precision of recognition.
Hossen, 2018 [67]	Horizontal & Vertical projections, Back-propagation feed-forward neural networks	Bangla	Private dataset	90.5%	Proposed method is very effective for different viewpoints, illumination conditions, and small distances.		In the future, accuracy can be improved.
Shaifur Rahman, 2018 [66]	Convolutional Neural Networks (CNN)	Bangla	Their own dataset	89%		For smaller memory and computational power, the proposed system faced some limitations.	With a higher number of function maps and more layers, the proposed framework can be augmented.
Silva, 2018 [64]	CNN, OCR	English	AOLP Road Patrol, SSIG, OpenALP R, CD-HARD	Higher than 80%.			This research can be extended to detect motorcycle LPs.
Prabhakar, 2014 [65]	Hough Transform, Horizontal Projection	English	Private dataset	94%	This system effectively reduces the computation time.		In the future, the system can be developed at a low cost in real-time.

In [67], a novel method for detecting the Bangla license plate was proposed by Hossen Firstly, the location of vehicles is determined. Next, compare the RGB intensity of the plate with the vehicle’s license and material properties to localize the license plate area. Thereafter, they have separated horizontal projection-based registration using the required threshold value. After that, using vertical projection of the same threshold value, the characters and the digits are also separated. Finally, using the back-propagation feed-forward neural networks, the characters and digits have been established. Authors have reported that 93.89%, 98.22%, and 92.77% respectively are the success rate of the license plate identification, segmentation, and recognition process. In Figure 11, the proposed method is shown.

In [68], a method using Convolutional Neural Network (CNN) and Deep Learning strategies to identify and recognize vehicle number plates in the Bengali language has been suggested by Alam. A super-resolution technique has been used with the CNN in the recognition portion to reconstruct the pixel quality of the input image. Each number plate character is segmented using a bounding box technique. 700 vehicles were appointed to test the experiment outcomes. They reported that in the validation set, CNN gained 98.2% accuracy and obtained 98.1% accuracy in the

evaluation set and the error rate was 1.8%. Their proposed system can be connected to a cloud-based system where all registered vehicle numbers will be stored.

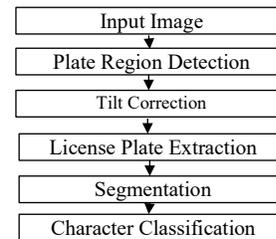


Figure 11: Overview of the proposed method by Hossen [67]

In [69], a prototype of YOLOv4 object detection has been implemented in which the Convolutional Neural Network (CNN) is trained and configured to detect the vehicle’s Bengali license plate and to recognize characters from the detected license plates using Tesseract (OCR engine). They reported that the model of license plate detection is trained at 90.50 % to mean average accuracy (mAP) and recall of 0.86 during training.

An overview of the vehicle number plate detection as well as recognition techniques is shown in Table 5. This table has been sorted based on the year and accuracy.

The existing methods for the identification and recognition of vehicle license plates have been classified based on accuracy that is shown in Figure 12.

5. Conclusion

This study paper presents a concise description of the vehicle number plate detection as well as recognition techniques used for effective traffic monitoring and observation of the reliability of the methods. In the construction of a smart transport network, vehicle number plate detection, as well as a recognition system, plays an important role. Although identification of vehicle number plates has always been a difficult proposition for certain reasons including changes in lighting, glare, non-uniform type of license plate, different styles, and color effects in the environment. Recognitions may also use some image processing techniques in conjunction with neural networks to identify the number plate characters, moving distance images, numbering schemes, angled or side-view images. In this study, the methods of vehicle number

plate detection and recognition have been classified based on accuracy. In the future, the preference is to use high-resolution cameras with an improved number of frames for better performance and effective license plate recognition. The classification section can be further improved with the complexity, speed, and chronological order. This study includes a comprehensive evaluation of the progress and future patterns in the identification and recognition of recent vehicle number plates which could be of value to researchers interested in such development.

Conflict of Interest

There is no conflict of interest reported between the authors.

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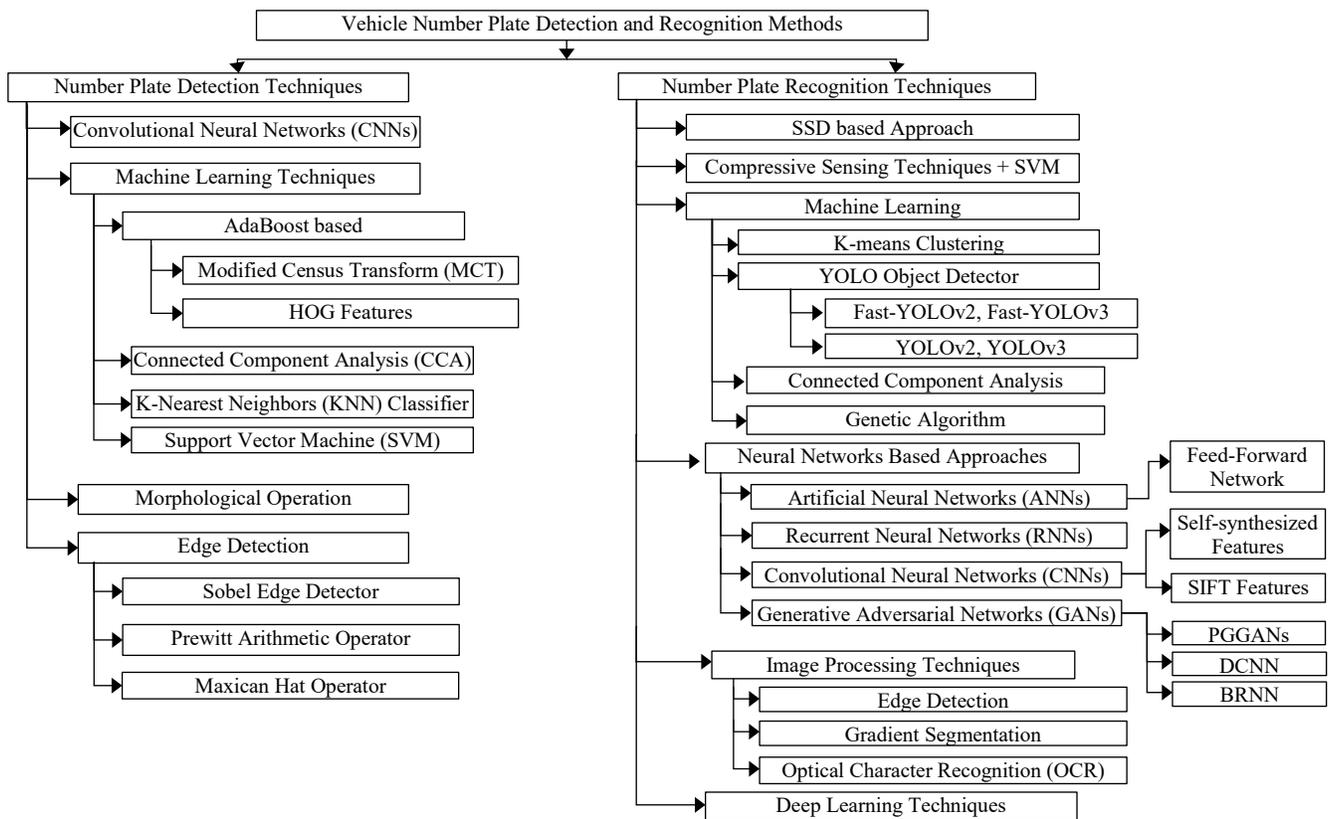


Figure 12: Existing frameworks for previous research.

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