

Vehicle Number Plate Detection Under Different Environmental Conditions

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Abstract

The main aim of the project is to detect the number plate of the vehicle and display the corresponding vehicle number. The motive of this project stems from the need for robust and reliable number plate detection systems that can perform effectively in real-world scenarios. Existing techniques frequently have trouble dealing with different lighting situations, like dimly lit areas, which can result in erroneous or missed detections. Furthermore, unfavourable angles of the number plates might make the detection issues worse. The problem statement of this project is to develop a number plate detection system that is capable of accurately locating number plates on vehicles, irrespective of the prevailing environmental conditions. To achieve this, we employ a multi-stage approach that combines computer vision techniques and image processing methodologies. Our proposed approach consists of three stages which includes: pre-processing, feature extraction and text extraction

Keywords: Machine learning, computer vision, image processing, feature extraction.

1. Introduction

The project aims to develop a robust vehicle number plate detection system capable of accurate operation under diverse environmental conditions. Advanced image processing techniques will be employed to tackle challenges such as varying lighting, different angles, and vehicle speed. The system's successful implementation holds the potential to improve traffic management, law enforcement, and intelligent transportation systems. By enhancing number plate detection capabilities, the project aims to contribute to safer roads and more efficient transportation systems, making it an asset for various domains reliant on accurate and reliable vehicle identification. The task area of this project focuses on the detection and localization of number plates on vehicles which involves the use of image processing techniques and involves handling challenging environmental factors, including varying lighting conditions, different vehicle speeds, and number plates at different angles. Employing advanced image processing methods such as edge detection, adaptive thresholding, and contour analysis will enable the system to accurately extract number plates, even under adverse conditions. To evaluate the above-mentioned task, we used different modules which includes cv2, NumPy, Py tesseract and matplotlib where cv2 provides functions to process and manipulate the images, NumPy provides support for multidimensional arrays and mathematical functions to work on arrays, Py tesseract is used to read text and Matplotlib is used for visualization of images.

2. Problem Statement

The current state of vehicle number plate detection systems reveals inherent challenges that hinder their

effectiveness in real-world scenarios. Existing technologies encounter difficulties when confronted with diverse environmental conditions, impacting their accuracy and reliability. The key problems that motivate the proposed project are:

- **Variability in Lighting Conditions:** Current systems struggle to consistently perform under varying lighting situations, leading to erroneous or missed detections, particularly in dimly lit or high-glare environments.
- **Challenges with Different Angles and Perspectives:** Detecting number plates becomes problematic when vehicles exhibit skewed or non-standard angles, making it difficult for existing systems to accurately locate and identify plates.
- **Speed-Related Issues:** The speed at which vehicles move can introduce motion blur, affecting the precision of number plate detection. Existing systems may face challenges in accurately identifying plates under varying speeds.
- **Font Style and Plate Variation:** Non-standardized font styles on number plates and variations in plate designs pose additional complexities. Existing systems may struggle to adapt to diverse font styles and plate formats.
- **Adaptability to Environmental Conditions:** The need for a robust system capable of operating seamlessly across different environmental conditions, including adverse weather, diverse lighting, and varying vehicle speeds, is evident.

3. Literature Survey

In 1990, Lotufo et al. has worked on automated number plate recognition (ANPR). In 1994, Lee et al. conducted yet another study on the identification of Korean licence plates in the past. In their research, the region of interest is first identified using neural networks for color classification. The classification process makes use of the pixels' Hue, Lightness, and Saturation (HLS) values. In 2018, Number plate detection was carried out by Surekha, however they did not record any character identification. In their research, they have used morphological processing, cunning edge detection, and neural networks. A 97% accuracy rate was recorded. They, however, did not consider many environmental conditions in their approach. Computer vision is a major area for vehicle number plate detection. Computer vision is a specialized branch of AI that is concerned with enabling computers to interpret and understand visual information from the world, much like human vision. The process typically involves a series of computer vision techniques and algorithms to achieve accurate and efficient number plate detection and the process includes image acquisition, image preprocessing, number plate localization, character segmentation, optical character recognition (OCR), and post-processing. These techniques enable machines to automate the detection and recognition of number plates, making vehicle number plate detection systems possible. By leveraging these techniques, systems can achieve accurate and reliable detection, enabling a wide range of applications in traffic management, law enforcement, security and automated systems. Enabling computers to see and comprehend visual data from the environment, akin to human vision, is the focus of computer vision, a specialized area of artificial intelligence. Image acquisition, image pre processing, number plate localization, character segmentation, optical character recognition (OCR), and post-processing are some of the computer vision techniques and algorithms that are typically used in the process to achieve accurate and efficient number plate detection. Vehicle number plate detection systems are made possible by these techniques, which allow machines to automatically detect and recognize number plates. These methods provide a broad range of applications

in traffic management, law enforcement, security, and automated systems by offering accurate and dependable detection for systems.

4. Proposed System

To develop an image processing system that can deal with cross angled, low illuminated, different font number plates or blurry plates. The goal is to create number plate detection systems that can reliably operate in various environmental conditions, contributing to traffic management, surveillance, law enforcement, and other applications. Continued advancements will drive the adoption of these systems, making our roads safer and transportation systems more efficient. our system can detect the plate and extract the text from the input image. To develop an image processing system that can deal with cross angled, low illuminated, different font number plates or blurry plates. The goal is to create number plate detection systems that can reliably operate in various environmental conditions, contributing to traffic management, surveillance, law enforcement, and other applications. Continued advancements will drive the adoption of these systems, making our roads safer and transportation systems more efficient. our system can detect the plate and exte 7 developed systems of three stages as shown in ract the text from the input image.



“Figure 1”

4.1: Image Preprocessing

The input is the image of the vehicle. The collected image is prepared for the entire plate identification procedure during the preprocessing stage. The preprocessing aims to reduce computational complexity while ensuring the image retains essential information for accurate identification. Techniques like gray scaling convert the colored image into a grayscale representation, reducing data processing costs. Adaptive thresholding segments the image into foreground and background regions, considering varying lighting conditions, enhancing consistency. Noise removal techniques further refine the image, eliminating unwanted artifacts that could interfere with the identification process. By effectively preprocessing the image through these techniques, subsequent stages benefit from reduced complexity, leading to faster and more accurate vehicle number plate detection and recognition. The preprocessing stage is an indispensable part of automatic license plate recognition systems, as it lays the foundation for reliable and efficient identification of number plates in various environmental conditions.

4.2: Feature Extraction

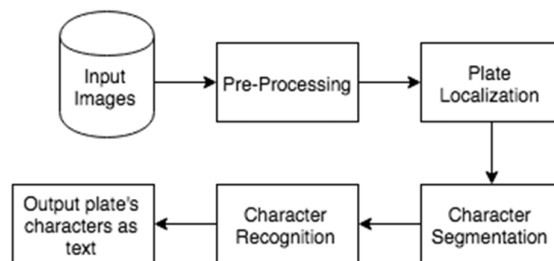
Feature extraction is a fundamental technique in computer vision and pattern recognition, aiming to capture meaningful information from raw data, such as images, to reveal underlying patterns or characteristics. In the context of pre-processed images, relevant features are extracted using methods like edge detection and morphological operations. Edge detection identifies the boundaries of objects in the image, representing prominent transitions in intensity, which can be crucial for recognizing object shapes. Morphological operations manipulate image structures to fill gaps, remove noise, and highlight

regions of interest. By extracting these relevant features, the processed data can be represented more effectively, aiding subsequent analysis and tasks, such as object recognition, classification, or detection. Feature extraction is a vital step in computer vision systems, as it enables the transformation of complex data into meaningful representations that facilitate advanced pattern recognition and understanding of visual information.

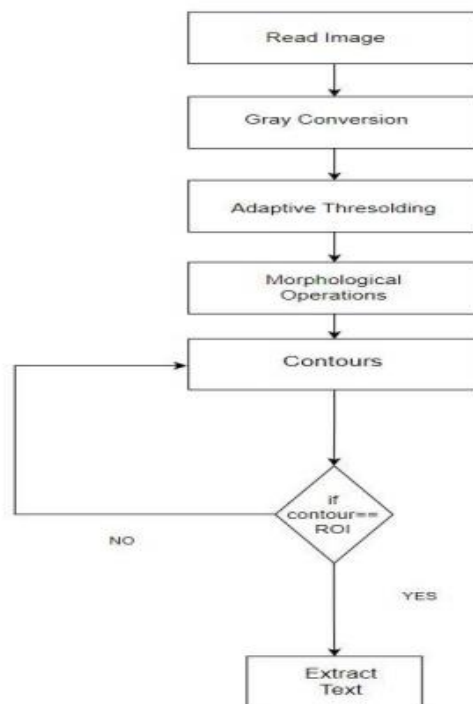
4.3: Text Extraction

Text extraction in vehicle number plate detection is facilitated by PyTesseract, a popular Python library that interfaces with the Tesseract OCR engine. Tesseract OCR, developed by Google, is widely used for accurate text extraction from images. In the context of vehicle number plates, PyTesseract can be applied to recognize the alphanumeric characters present in the localized number plate region. By extracting the text from the number plate, valuable information for vehicle identification and other applications can be obtained. The seamless integration of PyTesseract with Tesseract OCR makes it a valuable tool in the field of computer vision, enabling efficient and reliable text extraction from images.

5. Methodology



“Figure 2”



“Figure 3”

5.1: Read Image

The imread function from OpenCV is essential for reading image files, enabling analysis and processing of visual data. This step is vital to gain insights from the images and perform various tasks like object detection, recognition, and more, contributing to computer vision and AI applications.

5.2: Gray Conversion

Gray scaling, also known as gray conversion, is an essential image preprocessing technique widely employed in various computer vision applications, particularly in vehicle number plate detection. The process involves converting a colored image into a grayscale image, where each pixel is represented by a single intensity value instead of separate red, green, and blue (RGB) channels. The benefits of gray scaling are multifaceted. By reducing the image's complexity, it helps improve the accuracy of recognition algorithms, making them more efficient in identifying and processing visual data. Additionally, gray scaling proves advantageous in mitigating the impact of inconsistent or varying lighting conditions, which are common challenges in number plate detection tasks. Such lighting variations can distort colors, making color-based recognition less reliable. By simplifying the input data, gray scaling provides a consistent and standardized representation of images, facilitating subsequent analysis, and enabling more accurate and reliable results in vehicle number plate detection and other computer vision tasks. Fig 4 describes gray conversion .

5.3: Adaptive Thresholding

Adaptive thresholding is a vital image processing technique used in computer vision to binarize images, separating foreground objects from the background. This method is particularly beneficial when dealing with varying lighting conditions or uneven illumination. By converting the image into a binary format, adaptive thresholding allows the identification of regions of interest (ROIs) essential for further analysis, such as object recognition or tracking. The ROIs encompass the foreground objects, while the background objects, including noise or irrelevant elements, are excluded. This distinction aids in focusing on relevant information and improves the accuracy of subsequent computer vision tasks. Fig 5 visually demonstrates the process of adaptive thresholding, showcasing its significance in image segmentation and its valuable role in various computer vision applications.

5.4: Morphological Operations

Morphological operations are commonly used in vehicle number plate detection to manipulate the shape and structure of binary images. These operations help in removing noise, filling gaps, and enhancing the quality of the detected number plate regions. Two commonly used morphological operations are dilation and erosion. Dilation expands the boundaries of the regions in the binary image, while erosion shrinks them. These operations can be adjusted by changing the size and shape of the kernel as well as the number of iterations.

5.5: Contour Identification

The findContours function identifies the contours in the processed image. Contours represent the boundaries of objects in the image. Contours represent the boundaries of objects in an image, and by identifying and analysing these contours, we can determine the shape and location of the number plates. Once the relevant contours (regions of interest) are retained based on the area thresholding, the Hough

Transform is applied to 14 further refine the identification of the number plate regions. The Hough Transform is a technique used to detect straight lines or shapes in an image, such as the edges of the number plate. By applying the Hough Transform, the number plate's potential boundary lines can be identified, helping to confirm and precisely locate the number plate region. The combination of threshold area filtering and the Hough Transform assists in accurately and efficiently detecting and extracting the number plates from the processed image.

5.6: Text Extraction

It involves extracting the alphanumeric characters from the detected number plate region. One popular approach for text extraction is to utilize Optical Character Recognition (OCR) techniques.

6. Conclusion

This study focuses on Vehicle number plate detection and recognition. The main contributions of this study include considering difficult scenarios including change in lighting conditions, blurry, noisy pictures, skewed and non-standard and partially worn-out number plates. The preprocessing of this study includes image processing techniques including gray scaling, adaptive thresholding. Next, for segmentation of the plate contours are applied. Finally, by iterating through the contours we identify our region of interest to extract the vehicle number. Future technological and Artificial Intelligence developments are projected to increase the efficiency of vehicle number plate recognition in a variety of environmental settings. The following upcoming breakthroughs might strengthen the relationship between AI and number plate detection: enhanced imaging and sensors, AI based image processing, contextual understanding.

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