

AUTOMATIC LICENSE PLATE DETECTION USING YOLO

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ABSTRACT

Identity of license plate inclusive of characters and numerals is proposed using CNN. This is executed by way of amassing the numerals from (zero-nine) and characters from (A-Z) from which a dataset is constructed to train convolutional neural networks (CNN). The challenge is quite hard because of non uniform illumination situations the variety in plate codecs. License plate recognition is the effective manner for figuring out routinely, there also exists different algorithms but they not as effective in turning in pleasant effects. Number Plate recognition algorithms in pictures and videos typically includes 3 stages 1)license plate extraction 2)character segmentation 3)character by character recognition. With the skilled model, character recognition device is designed and implemented. In this thesis we make use of object detection algorithm YOLOV3 with the help of network structure Darknet.

Keywords: Yolo, OpenCV, Image Recognition, CNN.

I. INTRODUCTION

In recent years, the wide variety of privately-owned automobiles has improved significantly and this has, in turn, exacerbated the traffic control burden. The resultant congestion can reason excessive problems, like traffic injuries or public location vulnerability to crime or terrorist attacks. Physical management of this ubiquity of cars is quite hard, and this problem has advocated the improvement of computerized systems to manage traffic jams. In particular, automated license plate detection can accurately monitor the vehicles and greatly alleviate visitors control burdens; therefore, this method has attracted full-size interest of researchers.

Numerous techniques of figuring out automobile license plates were accomplished. In standard, those algorithms are advanced from 3 steps, namely searching the wide variety plate place, segmenting the characters from the registration code, and recognizing each individual character. The vehicle registration plate recognition gadget is an application that replaces the feature of human imaginative and prescient in recognizing motorized car license plates.

II. MOTIVATION

India's population is expanding daily, which has a significant impact on the growth of both the private and public transportation options. This rise in the number of cars is also contributing to an increase in traffic and the variety of crimes that go along with it. There have been several theft, hit-and-run, theft, kidnapping, smuggling, on-avenue fatalities, etc. incidents. Unsolved because the cars involved couldn't be precisely recognised. The project's motivation is to automate the laborious process of locating a vehicle's licence plate that is still used in India. Therefore, this version will let authorities keep track of those who are disobeying traffic laws.

III. OBJECTIVE

To recognize characters and digits on a license plate in a photograph by creating a dataset of characters in English and training them with the yolo approach. Create a system that uses YOLOV3 to detect characters and numbers in a licence plate image. to collect a large amount of data from government databases in order to train the model. The collection comprises of unique subsets of light requirements collected from all lighting conditions.

IV. LITERATURE SURVEY

An end-to-end DL-ALPR device for Brazilian licence plates based on modern Convolutional Neural Network architectures was proposed by Sérgio Montazzolli and Claudio Rosito Jung[1]. The system was once able to

accurately see and interpret all seven characters of a licence plate in 63.18% of the test set, and 97.39% while thinking about at least five proper characters. They used a publicly available dataset of Brazilian plates (partial match). They are able to segment 99% of the characters and accurately comprehend 93% of them by paying attention to each character individually and segmenting them.

Vasyl Koval, Volodymyr Turchenko, V. Kochan, A. Sachenko, George Markowsky[2] proposed the Smart Vehicle Screening System, which can be established into a tollbooth for computerized focus of car license plate data the usage of a image of a vehicle. An computerized device may want to then be carried out to manipulate the fee of fees, parking areas, highways, bridges or tunnels, etc. There are regarded an strategy to pick out car thru recognizing of it license plate the use of picture fusion, neural networks and threshold strategies as nicely as some experimental consequences to understand the license plate successfully.

Chomtip Pornpanomchai , Nuchakarn Anawatmongkon[3] proposed a study that aims to evaluate the effectiveness of a straightforward algorithm used to extract the location of licence plates from a video frame. This study has the ability to recognise a licence plate on a car's front or back. Users can take screenshots from the video file using the project's application, which will subsequently produce the screenshot. The application will also determine where the licence plate is located in that screenshot. The image that our programme takes will be saved as a.jpg file. This project's accuracy for a video screenshot is approximately 82%, and for a digital camera image, it is approximately 92%.

Shantha Lakshmi S, Basker N, Sathiyabhama B, Vinothkumar R B, Revathi T[4] proposed a deep learning-based system which is suggested for detecting both vehicles and number plates. A video dataset that has been divided up into numerous frames is used as the input for the suggested model. The dark flow toolkit detects the vehicles and their licence plate using pre-trained weights and labels from the dataset. This programme allows you to extract the vehicle's region with accurate annotation. The proposed model's ultimate goal is to determine the vehicle's speed based on the immediate surroundings.

V. HARDWARE AND SOFTWARE REQUIREMENTS

Table 1. Hardware Requirements

SR. No.	Hardware	Description
1.	Processor	Intel Core i7 @ 2.70 GHz
2.	Memory	8.00 GB
3.	Hard Disk Space	256 GB
4	Device	HP Pavilion
5.	Others	Other required standard computer peripherals, such as keyboard and mouse.

Table 2. Software Requirements

Sr. No.	Software	Description
1.	Operating System	Microsoft Windows 10
2.	Database server	MySQL
3.	Web Server	Xampp
4.	Programming IDE	Jupyter Notebook , GoogleCollab
5.	Browser	Google Chrome , MicrosoftEdge

VI. ARCHITECTURE

Yolo is an entirely convolutional network, and the final result is produced by using a 1 x 1 kernel on a function map. The detection in YOLO v3 is accomplished by applying 1 x 1 detection kernels on function maps of three amazing sizes at 3 specific locations within the community. The recognition kernel's shape is 1 x 1 x (B x (5 + C)). Here, B is for the number of bounding packing containers that a cellular at the characteristic map can anticipate, "5" stands for the four bounding container attributes and one item confidence, and C stands for the quantity of

instructions. The kernel size in YOLO v3 trained on COCO is 1×1 , with $B = 3$ and $C = 80$. so the kernel size is $1 \times 1 \times 255$.

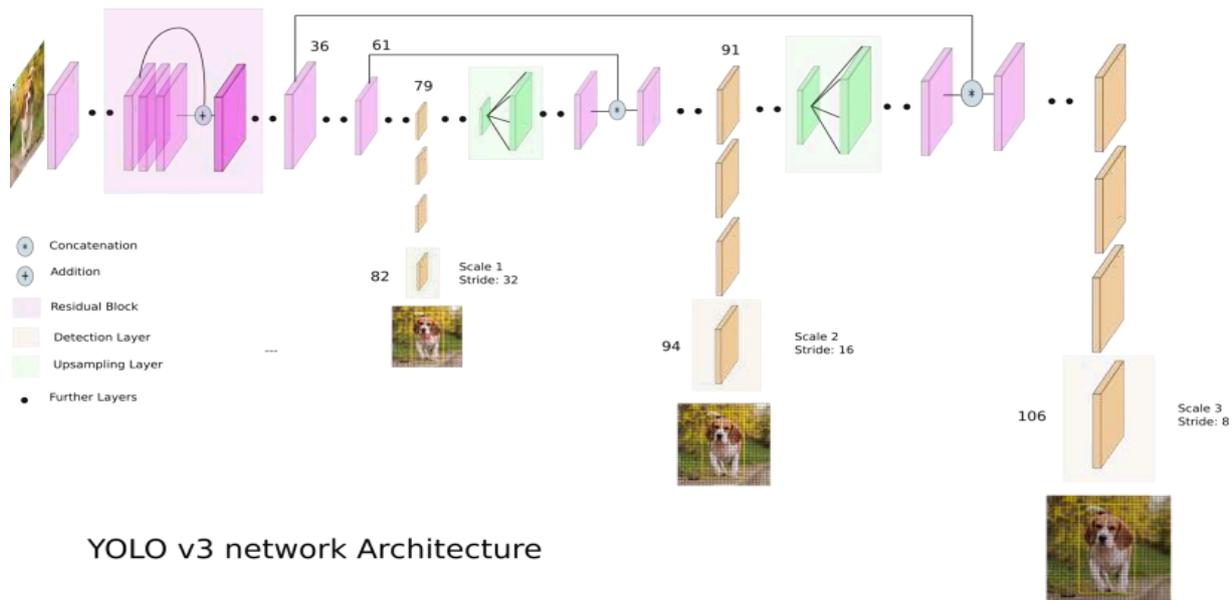
The 82nd layer does the initial detection. The network downshifts the image for the first 81 layers until the 81st layer has a goose-step of 32. The generated feature map may have dimensions of 13×13 if the image is 416×416 . The 1×1 recognition kernel is used to make a single verification, giving us a recognition function map of $13 \times 13 \times 255$. After that, layer 79.99's characteristic map is sent through three convolutional layers before being 2x up-sampled to a size of 26×26 . The characteristic map from layer 61 is then intensified with this singleverification, giving us a recognition function map of $13 \times 13 \times 255$.

After that, layer 79.99's characteristic map is sent through three convolutional layers before being 2x up-sampled to a size of 26×26 .The characteristic map from layer 61 is then intensified with this feature map.

The features from the earlier layer are then combined by subjecting the combined characteristic maps to a few more 1×1 convolutional layers(61). The 94th layer then performs the second detection, producing a detection function map with the dimensions 26 by 26 by 255.

Similar to before, some 1×1 convolutional layers follow to combinethe data from the prior layer. We create the final of the three at layer 106, producing a distinctive map with dimensions of 52 by 52 by 25.

Another comparable approach is used, in which a feature map from layer 91 is depth concatenated with a characteristic map from layer 91 after being exposed to a few convolutional layers.



YOLO v3 network Architecture

Figure 1: YOLOv3 Architecture

VII. METHODOLOGY

• Data modelling

We employed a security camera that could observe licence plates of moving cars from a distance as well as licence plate photographs from the government database as part of our endeavour. The licence plates are easily visible from this. The camera is within reach, and video of regular information is provided. New films from other days in the government database that are not part of the training dataset are obtained to finish testing.

Then it is labelled using the Labeling tool. Only the licence plates are displayed here from the collection of photographs, and the registration number and the car's whole image have also been cropped.

• Annotating Images

Since we underestimated how long it would take to name the photographs, we decided to use an open supply for our task. All of the files were then saved in the txt format once the images had been annotated. The total number of lessons for numbers is (0.9), and there are (zero, 26) lessons for characters.

• Data Preparation

Yolov3's data is in txt format and comprises the instructions as well as a few values. The values are entered into www.irjmets.com

the device using a certain format where the absolute x, y, width, and peak values are listed in italics.

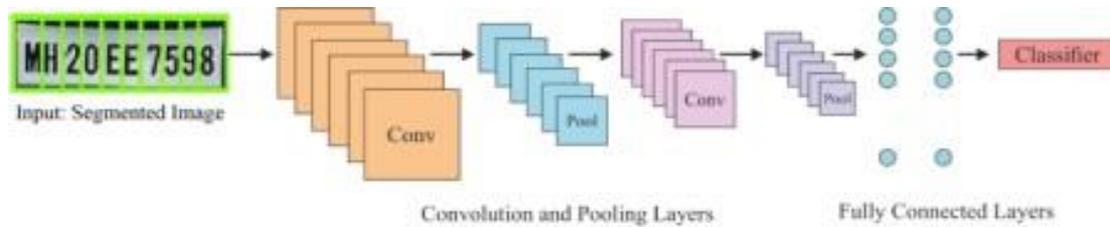


Figure 2: Model process flow.

VIII. CONCLUSION

We outline a procedure in this work for determining and locating the license plate number on Indian license plates. Since there isn't a publicly accessible dataset on the Indian license plate, we created a collection of license plates for different classes of vehicles. We extracted the license plate for recognition using the CNN-based approach (YOLO). The first CNN network's recovered image serves as the input for the second CNN network, which is in charge of segmenting and identifying the license plate number. We believe that a more diverse training dataset will produce higher performance on the test dataset since deep neural network models are data dependent. Additionally, a broad dataset will enable us to classify future vehicle kinds without having to modify our algorithm.

IX. REFERENCES

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