

Automatic Traffic Sign Recognition and Detection

Sona Roy, Krishnaprasad K, Therase Joshy, Vaishnavi K Department Of
Computer Science

Sahrdaya College of Engineering And Technology, Thrissur, Kerala, India

Abstract—The automatic recognition of traffic signs is important to autonomous driving, assisted driving, and driving safety. Currently, convolutional neural network (CNN) is that the hottest deep learning rule in traffic sign recognition. However, the CNN cannot capture the poses, views, and directions of the image, nor accurately acknowledge traffic signs from totally different views. To solve the drawback, an automatic recognition rule for traffic signs supported visual inspection was created. For the accuracy of visual examination, an area of interest (ROI) extraction technique was designed through content analysis and key data recognition. Besides, a bar graph of directed Gradients (HOG) technique was developed for image detection to forestall projection distortion. A traffic sign recognition learning design was created supported CapsNet, that depends on neurons to represent target parameters like dynamic routing, path create and direction, and effectively capture the traffic sign data from totally different angles or directions. Finally, our model was compared with many baseline methods through experiments on LISA (Laboratory for Intelligent and Safe Automobiles) traffic sign dataset. The model performance was measured by mean average preciseness (MAP), time, memory, floating purpose operations per second(FLOPS), and parameter variety. The results show that our model consumed shorter time nevertheless higher recognition performance than baseline ways, together with CNN, support vector machine (SVM), and region-based absolutely convolutional network (R-FCN) ResNet one zero one.

Date of Submission: 20-05-2022

Date of acceptance: 03-06-2022

I. INTRODUCTION

In this present world, we are really busy & not able to concentrate into our daily tasks. During driving, we do multitasking: attending phone calls, eating sandwich or something & we do not care about the traffic signals along the roadside. For this problem, automatic traffic sign recognition comes with the solution. In the future world of automated driving, driver less vehicles, automatic traffic signs plays a vital role.

Automatic traffic sign recognition is the most important area in intelligent driving. So accuracy of detecting signs need to be improved. Even small error in recognising traffic signs will lead to disasters.

In Indian railway, most of the trains works on non automated traffic sign recognition. This arises the possibility of disastrous consequences. To improve this, a machine is implemented that can detect signs & give alerts to the driver.

To solve the problems of traffic and road safety, a system of automatic detection and recognition of traffic signs (TSDR) has been introduced. Automated TSDR systems can detect and recognize traffic signs from and in images captured by cameras or image sensors. In adverse traffic conditions, vehicle drivers may not pay attention to traffic signs, which can lead to accidents. In such situations, the TSDR system will work. The main goal of TSDR research is to improve the robustness and efficiency of the TSDR system. Developing an automated TSDR system is a tedious task given the constantly changing lighting and environmental conditions. Other issues that also need to be addressed include partial occlusion, multiple traffic signs appearing at the same time, and blurring and blurring of traffic signs, which can also create problems for detection purposes. To apply the TSDR system in a real-time environment, a fast algorithm is required. In addition to addressing these issues, an identification system should also avoid mis-identification of non-markers.

In recent years, various traffic sign recognition systems have been proposed. This project presents an overview of some recent and effective methods for detecting and classifying traffic signs. Indeed, the main goal of detection methods is to locate areas of interest containing traffic signs.

II. EXISTING SYSTEM

The Advanced Driver Assistance System includes traffic sign detection and recognition (ADAS). Traffic signs tell drivers on traffic rules, road conditions, and route directions, assisting them in driving more safely. The detection and recognition of traffic signs is divided into two stages: The first stage involves locating traffic signs, while the second stage involves categorising the discovered traffic signs. The challenges of real-

time traffic sign detection are discussed in this article. A review of existing detection approaches is offered, including color-based, shape-based, and learning-based detection. The feature matching and machine learning algorithms utilised in the traffic sign recognition step are also discussed.

III. LITERATURE SURVEY

According to the study "Fresh and rotten fruits categorization using CNN and transfer learning," detecting rotting fruits has become increasingly significant in the agricultural business. People are normally in charge of classifying fresh and decaying fruits, which is inefficient for fruit growers. Humans become tired after doing the same thing over and over, while machines don't. As a result, the research proposes a strategy for decreasing human effort, cost, and production time in the agriculture industry by identifying flaws in the fruits. If the flaws are not discovered, infected fruits may contaminate healthy fruits. As a result, we provided a model for preventing disease transmission. The suggested model distinguishes between fresh and rotten fruits based on the input fruit pictures. In this project, we used three distinct types of fruits: apple, banana, and oranges. A CNN is used to extract characteristics from the input fruit photographs, while Softmax is used to classify the photos into fresh and rotting fruits. The correctness of the suggested model is evaluated using a dataset obtained from Kaggle, which provides a score of 97.82. The results showed that the proposed CNN model can tell the difference between fresh and rotten apples.

The paper "A hybrid classifier for handwriting recognition on multi-domain financial bills based on DCNN and SVM" claims that with the rapid growth of the global economy, automatic recognition of financial bills is becoming the primary way to reduce the burden of the traditional manual approach to bill recognition and classification. Most automatic recognition systems, on the other hand, are unable to recognise handwritten characters on bank bills, particularly when the bills are from various financial institutions. To address the issue, this work investigates the bill system in banks as well as the operations of bill number identification, before developing a hybrid classifier based on DCNN and SVM to recognise handwritten numbers on financial bills in various domains. To efficiently mine the local handwritten numbers on financial invoices from various sources, the DCNN with several channels was used. The retrieved data was then loaded into the SVM to achieve accurate number categorization. Our solution takes full advantage of the information distribution differences between fields and adapts to them using a parameter sharing mechanism. Experiments reveal that our method is more accurate than benchmark methods at recognising handwritten numerals on bank bills.

According to the paper "Role of counsellors in the nurturing of inventive skills from the perspective of students," guidance and counselling play an important role in schools for a child's future success. The study acknowledges that counselling is a transforming process that assists people in learning everything they need to know both inside and outside of school. Because this is an opinion publication, the study used review research procedures. The document acknowledges that counsellors must gain the child's trust in order to provide him or her with the required information to assist the youngster. Counselors should encourage students to participate in treatment sessions in order to aid the child's future achievement. The function of advice and counselling in effective teaching and learning in schools for the future success of children is discussed in this study.

The paper "An approach to detect automobiles in diverse climatic circumstances using the corner point technique" describes a new method for recognising vehicles that employs a basic yet effective algorithm. The most significant components in vehicle detection are the attributes of the vehicle. The proposed algorithm takes into account the corner points. The Harris corner detector is used to calculate a large number of points that are closely packed within the area of a vehicle. These points are grouped together because they are densely packed. This grouping implies that each vehicle owns a set of corners, and such groupings are important in the algorithm.

After grouping is complete, the background noise must be removed. The retrieved corner points are tracked using the Lucas Kande algorithm. To make the output stable and trustworthy, each corner point of the vehicle is tracked. The suggested method is unique in that it can recognise automobiles in a variety of settings and can work in complex contexts.

The paper "Cost-effective traffic sign detection using smart mobile devices" describes a mobile device-based autonomous road sign detection and recognition system. The system is built on a client-server architecture that allows mobile users to take images of traffic signs and send them to a centralised server for processing. The server performs the preprocessing, detection, and recognition, and then sends the result back to the mobile device. We employ specific colour attributes calculated from the input image to detect traffic signs. A neural network based on normalised colour histogram characteristics is used to implement recognition. The implications of various parameters on recognition accuracy are discussed. Our findings show that the system can provide an effective foundation for multilingual recognition of locale-dependent traffic signs.

IV. PROPOSED SYSTEM

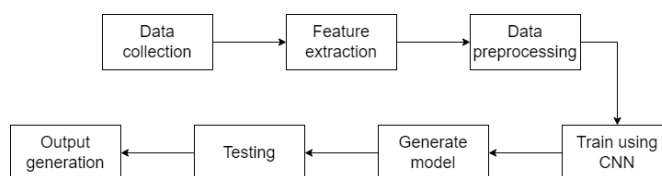


Fig. 1. Block diagram of proposed model

Step 1: Data Collection

We use LISA Dataset. The LISA Traffic Sign Dataset is a set of videos and annotated frames containing US traffic signs. It is released in two stages, one with only the pictures and one with both pictures and videos. The images are available now, while the full dataset is underway and will be made available soon. It has 47 US sign types, 7855 annotations on 6610 frames, Sign sizes from 6x6 to 167x168 pixels.

Step2: Input Data

The second step is to input data. Data input is given in the form of video. Here, we use opencv for data input. The input video is converted into frames.

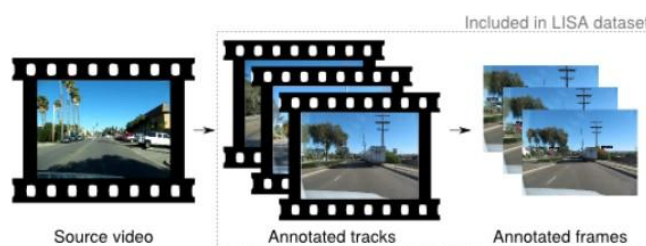


Fig. 2. Conversion of video to frames

Step3: Training Data

Training data is the third step. We use LISA dataset for training using CNN.

Step4: Generate model file

Using CNN we will train the data and will save these trained data as a model file.

Step5: Testing

After extracting frame from user video, we will check the match with the generated model file and find prediction. Then mark the prediction as bound box using opencv.



Fig. 3. output inside bound box with prediction

Dependencies:

Python 3.5+

TensorFlow v0.12.0: used for training using cnn Pickle

OpenCV-Python: for operations on camera like play Matplotlib (optional): for making graph and chart

V. CONCLUSION

This paper provides a Capsnet-based automatic road sign recognition algorithm. ROI extraction, HOG feature creation, and layered CapsNet are all part of the proposed approach. The authors described the HOG feature, which could efficiently acquire the image's pixel-level properties from any viewpoint and speed up the model's learning. Our model outperformed baseline techniques in terms of speed, memory usage, and parameter requirements when tested on a real-world traffic sign dataset. The CapsNet outperforms the traditional CNN since it uses vectors to fully exploit photos of various angles and directions.

VI. FUTURE WORK

All people who drive a vehicle on the road can benefit from traffic signs. Drivers are guided by traffic signs to respect all traffic laws and avoid causing any inconvenience to pedestrians. Environmental factors such as illumination, shadow, distance (sign is pretty far away), air pollution, weather conditions, motion blur, and vehicle vibration, all of which are frequent in real-time systems, might affect detection and consequently categorization. As a result, more study and developments are required to address these concerns. This technique allows the driver to see the sign on the screen that is close to his or her eyes. This saves time and effort in manually looking for traffic signs and determining what type of traffic they are.

REFERENCES

- [1]. M. S. Aminian, A. Allamehzadeh, M. Mostaed, and C. Olaverri-Monreal, "Cost-efficient traffic sign detection relying on smart mobile devices," in Proc. Int. Conf. Comput. Aided Syst. Theory, Feb. 2017, pp. 419–426.
- [2]. T. Wang, H. Shen, Y. Xue, and Z. Hu, "A traffic signal recognition algorithm based on self-paced learning and deep learning," *Ingénierie des systèmes d'Inf.*, vol. 25, no. 2, pp. 239–244, May 2020.
- [3]. X. Yu, J. Yang, T. Wang, and T. Huang, "Key point detection by max pooling for tracking," *IEEE Trans. Cybern.*, vol. 45, no. 3, pp. 430–438, Mar. 2015.
- [4]. K. Zhang, W. Zuo, Y. Chen, D. Meng, and L. Zhang, "Beyond a Gaussian denoiser: Residual learning of deep CNN for image denoising," *IEEE Trans. Image Process.*, vol. 26, no. 7, pp. 3142–3155, Jul. 2017.
- [5]. C.-A. Lameloise, H. Chauris, and M. Noble, "Improving the gradient of the image-domain objective function using quantitative migration for a more robust migration velocity analysis," *Geophys. Prospecting*, vol. 63, no. 2, pp. 391–404, Mar. 2015.
- [6]. K. Qian, L. Tian, Y. Liu, X. Wen, and J. Bao, "Image robust recognition based on feature-entropy-oriented differential fusion capsule network," *Appl. Intell.*, pp. 1–10, Sep. 2020.
- [7]. P. Bulla, L. Anantha, and S. Peram, "Deep neural networks with transfer learning model for brain tumors classification," *Traitement du Signal*, vol. 37, no. 4, pp. 593–601, Oct. 2020.
- [8]. U. Kamal, T. I. Tonmoy, S. Das, and M. K. Hasan, "Automatic traffic sign detection and recognition using SegU-net and a modified tversky loss function with L1-constraint," *IEEE Trans. Intell. Transp. Syst.*, vol. 21, no. 4, pp. 1467–1479, Apr. 2020.
- [9]. S. H. He, X. D. Li, Y. Wang, and H. H. Zhu, "An optimization model for automobile mixed assembly line under multiple constraints," *Int. J. Simul. Model.*, vol. 16, no. 4, pp. 720–730, Dec. 2016.
- [10]. D. Chen, "Multiple linear regression of multi-class images in devices of Internet of Things," *Traitement du Signal*, vol. 37, no. 6, pp. 965–973, Dec. 2020.