

CNN and Keras based Road Safety Traffic Signs Recognition System

¹M.Mohsina Kousar, ²Dr. S. Shoeb Peer

¹Research Scholar, Department of CSE, Ashoka women's engineering College, Kurnool, India

²Professor, Department of CSE, Ashoka women's engineering College, Kurnool, India

ARTICLE INFO

Article History:

Accepted: 01 Aug 2023

Published: 08 Aug 2023

Publication Issue

Volume 10, Issue 4

July-August-2023

Page Number

368-374

ABSTRACT

Accident rates increased due to non-compliance with traffic signs and non-compliance with traffic rules. By using synthetic training data from images of train signs, we can overcome the problem of analysis of traffic data, which varies by country and region. This method is used to create files containing synthetic images to detect traffic signs from different perspectives. With this knowledge and a well-designed Convolutional Neural Network (CNN), we can create a data-driven train recognition and detection system that has accurate detection accuracy and still has action. This article presents the process of recognizing and classifying road signs and traffic signs to create their inventory to help great engineers replace and maintain them. It uses images captured by a camera from a moving car. The system is based on three main stages: color segmentation, recognition and classification. Many researchers in the world of artificial intelligence and technology and Tesla, Uber, Google, Mercedes-Benz, Toyota, Ford, Audi etc. big companies like self-driving cars and self-driving cars. Therefore, in order for this technology to be used correctly, vehicles must be able to interpret traffic signs and make decisions accordingly. This reduces accidents and also helps drivers concentrate on driving rather than looking at every traffic card. The purpose of this document is to provide an effective way of discovering and recognizing traffic signs in India.

Keywords : Road Accidents, Traffic Sign Recognition, Convolution Neural Network, Feature Extraction

I. INTRODUCTION

We present methods that overcome the limitations of existing methods such as neural networks and feature

extraction and improve traffic sign detection performance while reducing traffic accidents[1].

Traffic signification is the process of determining the class to which a traffic sign belongs. In fact, the

pattern of the symbol is always the same, the only change is the camera information or the view from the driver and the environment that is difficult to recognize, causing changes in authentication that can be different[2]. by changing the traffic signals in the training set to be completed. Our method does not require real-time data, as we use the dataset, which contains images similar to real-world data.

This research is done using machine learning techniques. Before we move on, we need to understand some of these concepts[3]. Machine learning is a method of training machines to do things on their own without human intervention. These estimates will be output. Neural network, deep learning etc. There are many branches of machine learning such as Among them, deep learning is considered the most popular branch of machine learning[4]. Traffic signs are an important part of the road. Must obey the traffic signs, obey the traffic rules and be safe. Road safety is a major concern for pedestrians and drivers. Road safety, traffic signs, weather, road congestion, etc. There are many factors that affect road safety, such as traffic signs. Sometimes drivers do not recognize these traffic signs, which is one of the causes of accidents[5]. As the car industry is booming, they are focusing on self-driving cars. It is expected to be safe and work better. Distribution of traffic signs is an important part of electric vehicles. It should be effective in reducing road accidents and increasing the reliability of self-driving cars. Otherwise, it may cause traffic damage. Traffic sign classification is still a difficult recognition problem for computer systems.

In this project, we will build a deep neural network model that can classify existing traffic signs in an image. With this model, we can read and understand traffic signs, which is a very important task for any driverless car[6]. Validating the method has become a major challenge in education and business. This can lead to conflict and increase risk. Section Confusion in identifying warning signs can be dangerous. However, due to differences in the environment, there are some

instances where road signs may differ, making the appear difficult to humans and machines. Article Problems found in records can be fixed by using different man-made documents without burdening workers with approval. The acknowledges the signal in several ways. The article uses the Hough transform to determine the convergence of the path signature, or similar, as in the paper where the circular Hough transform is used to determine the constraint marks[7,8]. The method is based on histogram features and support vector machine that uses information about the shape of the signature in the detection step.

II. EXISTING SYSTEM

Yadav et al. and others. Use support vector technology. For training and testing purposes, the data is split 90/10 by distribution. To achieve what is desired, there must be levels of color segmentation, classification and recognition. Anushree.A et al. and others. Use Raspberry Pi to detect and recognize traffic signs with little coding[9]. However, it needs a Raspberry Pi board that is very expensive to use. S. Harini et al. Another way to show traffic awareness is to use lots of pictures. Processing the image first, including separating the foreground and background, thinning, and so on. These symbols are then divided into hexagons, triangles or circles and are then sent to the comparison model. Objects with some similarities are determined by the pre-learning algorithm.

III. PROPOSED SYSTEM

In recent years there has been an interest in deep learning. The smartest of many deep learning models is the convolutional neural network (CNN), a class of neural networks central to computer vision because it shares the advantages of ImageNet Large Scale's known competitors. 2012 Visual Recognition Competition (ILSVRC). Medical research is no exception, and CNNs have achieved expert-level

performance in their domain. Demonstrated the potential of deep learning for diabetic retinopathy diagnosis, skin classification, and prostate cancer detection, respectively. Needless to say, radiologists are more interested in the capabilities of CNNs, and many studies have been published in areas such as lesion detection, classification, classification output, image processing, and natural language processing[10,11]. Knowing this best method will not only help researchers use CNNs for radiological and medical tasks, but will also help radiologists because deep learning will impact its future applications. This article focuses on the main aspects of CNNs and their application to various electronic projects and discusses their challenges and future directions. Other deep learning models such as neural networks for sequential models are beyond the scope of this article.

IV. METHODOLOGY

3.1. DETECTION USING CNN ENSEMBLE:

Traditional methods such as machine learning algorithms are used for traffic sign classification. Machine learning, a field that has emerged from the field of artificial intelligence, is important because it enables machines to gain human-like intelligence without being specially designed. The structure consists of stacked iterations of several convolutional and pooling layers followed by one or more layers[12]. The step in which the input data is converted to output through this process is called propagation. Although the integration and clustering methods described in this section apply to 2D-CNNs, similar tasks can be performed on three-dimensional (3D) CNNs. An overview of convolutional neural network (CNN) architecture and training process. CNNs are stacked by several building blocks: the convolutional layer, the pooling layer (like maximum pooling), and the full link (FC) layer[13]. The performance of the model with certain kernels and weights is calculated by the performance loss due to the subsequent propagation of the training data and

learning parameters (i). That is, the core and weight adjusted for the loss value through backward extension of the gradient descent optimization algorithm. ReLU, corrected linear unit

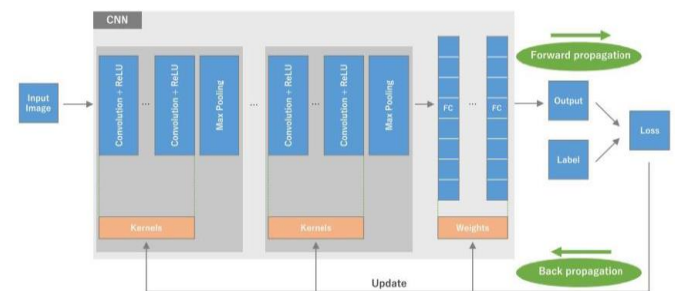


Fig.1. Convolutional neural network (CNN) architecture

Due to the complexity of fabrication, numerous deep learning algorithms have been proposed to learn from multiple hidden mechanisms in deep neural networks. Deep learning is the study of a combination of neural networks and other machine learning algorithms with many hidden layers. Deep learning enables multi-layered computational models to learn data representations with different levels of abstraction. The ability of deep learning to provide accurate analytics and predictions continues to evolve[14]. Thus, deep learning goes beyond machine learning algorithms. In this article, we propose a deep CNN model, the TS-CNN model, by adding four thick layers or all connected layers. The fully connected process learns deeply as it learns the properties of each feature combination from the previous process. Thus, our model achieves 95% accuracy. The technique used for detection and recognition is an image processing process with groups (CNN) for recognition.

3.2 color segmentation:

They used the technology of a small computer ARK-2121 and installed this machine in the car. The main techniques in character recognition are SVM and HOG. They achieved 91% detection accuracy and an average of 98% accuracy for the classification system. The high recognition of CNN makes it suitable for many studies in the field of vision. The method used to make CNNs is Tensor Flow. Members of the

newspaper achieved more than 99% accuracy on traffic signs using German data.

3.3 Convolutional Layers:

Convolutional layers are the principle of CNN architectures that perform feature extraction and often include a combination of linear and nonlinear operations, i.e. convolution operations and activation functions. Convolution is a special type of linear operation used for subtraction, in which a small number of numbers called kernels are used for input, a sequence of numbers called tensors[15]. The smart product between each element of the kernel and the tensor is calculated in each function of the tensor and summed to get the output value of the corresponding output tensor, called the special report. This process is repeated using multiple cores to create a set of unique maps that represent specific maps.

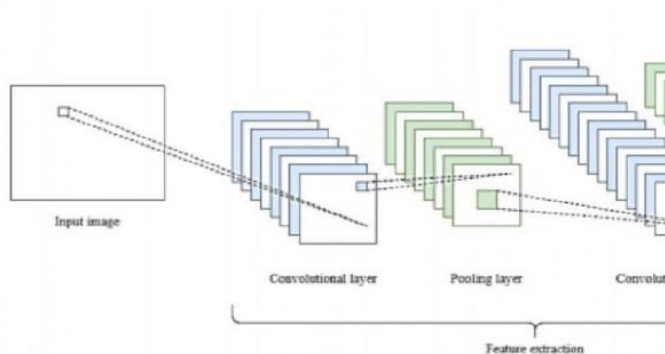


Fig. 2. CNN Architecture for Classification

A CNN architecture for classification includes convolution, max pooling, and all layers. Convolutional and maximum pooling layers are used for feature extraction. Convolutional layers are used for feature detection while maximum pooling layers are used for feature selection. Maximum pooling layers are used when the image does not need all the high resolution details or when the output of a small area extracted by the CNN after downsampling of the input data is needed[16]. The results of the convolution and pooling layers are fed to all layers for classification.

V. RESULTS



Fig.3. Upload Image

The image above represents the link of the website request before an image is uploaded to the server. In order to divide images into 43 different symbols, we can click the upload button and assign the signature image.

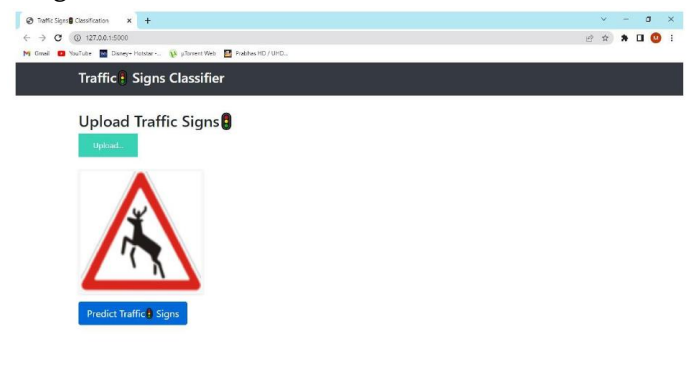


Fig.4. Predictive Traffic Sign

Now we can see that the image has been uploaded to the web page and we can now share the image by clicking on the "Forecast Traffic Sign" button.

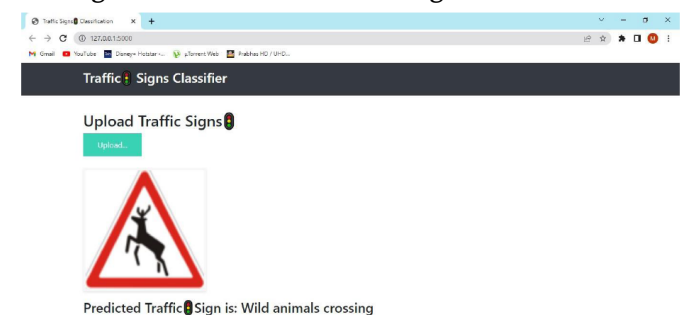


Fig.5. Predicted Traffic Sign is Wild animals crossing



Fig.6.

The sign indicates whether to go straight or turn right

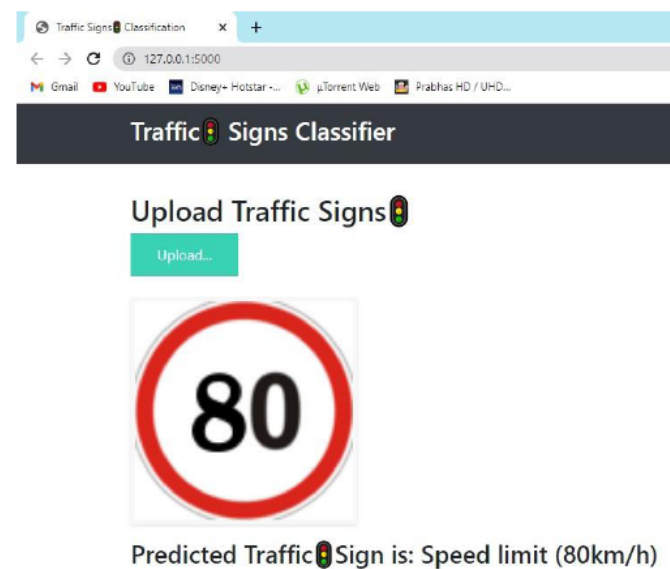
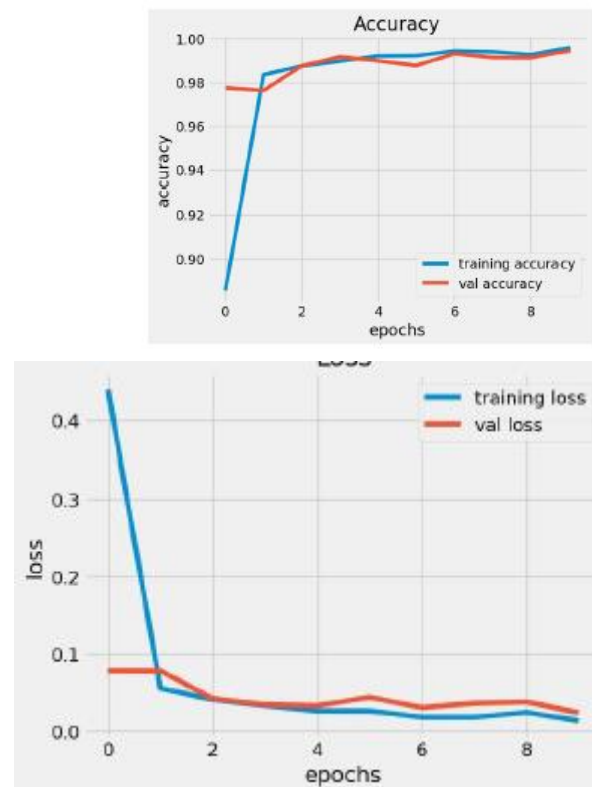


Fig.7. Traffic signs indicate the speed limit (80 km/h)



Fig.8. Predicted Traffic Sign is Slippery road

Accuracy and Loss Graph



VI. CONCLUSION

Classification of traffic signs is carried out using neural networks. Section learns the various models under CNN notification and uses the model with the highest accuracy on theGTSRB dataset. Section Create multiple categories for each vehicle race to increase standard accuracy. Knowing thesign will send a warning to the driver, helping him make the right decision[16,17]. Section This model is a major advance in driving as it reduces the driver's work without compromising safety.In addition, this system can be used easilywithout requiring much hardware, thus increasing its usefulness.Our plans will get us one step closer to realizing the best driver assistance (self-driving cars) or even not driving at all, and there is still much room for improvement. To detect marks, the text depends on the color and shape of the marks. The CNN algorithm is preferred because it provides high performance compared to otheralgorithms and can cause problems in case of interference between

symbols that affect its color. Another important issue to consider is night search. If you are using an infrared camera, there is no problem detecting the signature, but if you are using a non-infrared webcam, the signature will not be detected clearly, leading to a high risk of accident. This will reduce day and night events. Because we are using a low GPU (graphics processing unit), we cannot get 100% performance if the bus is too far from the system [18]. However, we provide a text-to-speech module that can be installed in the advanced driver service. Smaller materials were used in this project compared to other models, which reduces the cost and is not affected by the material. Traffic sign classification is very useful in driving assistance. Convolutional Neural Networks are a class of deep learning techniques used to analyze and analyze visual images. It is used to train image classification and recognition models because of its accuracy and precision.

VII. REFERENCES

- [1]. Shustanov, P. Yakimov, "CNN Design for Real-Time Traffic Sign Recognition," 3rd International Conference "Information Technology and Nanotechnology," ITNT- 2017, 25-27 April 2017, Samara, Russia.
- [2]. Wali, S. B., Hannan, M. A., Hussain, A., & Samad, S. A. (2015). An Automatic Traffic Sign Detection and Recognition System Based on Colour Segmentation, Shape Matching, and SVM. *Mathematical Problems in Engineering*, 2015, 1–11. doi:10.1155/2015/250461.
- [3]. R. Biswas, H. Fleyeh, M. Mostakim, "Detection and Classification of Speed Limit Traffic Signs," *IEEE World Congress on Computer Applications and Information System*, pp. 1-6, January 2014.
- G. Antipov, SA Berrani, JL Dugelay, "Minimalistic CNN-based ensemble model for gender prediction from face images," *Elsevier*, January 2016.
- [4]. Y. Xie, L. F Liu, C. H. Li, and Y. Y. Qu. "Unifying visual saliency with HOG feature learning for traffic sign detection." In *IEEE Intelligent Vehicles Symposium*, , 2009, pp. 24-29.
- [5]. R. Qian, B. Zhang, Y. Yue, Z. Wang, D. Coenen, "Robust Chinese traffic sign detection and recognition with deep convolutional neural network," *IEEE 11th International Conference on Natural Computation*, pp. 791-796, January 2016.
- [5]. J. Stallkamp, M. Schlipsing, J. Salmen, and C. Igel, "The German traffic sign recognition benchmark: a multi-class classification competition," in *Proc. IEEE IJCNN*, 2011, pp. 1453–1460.
- [6]. S. Houben, J. Stallkamp, J. Salmen, M. Schlipsing, and C. Igel, "Detection of traffic signs in real-world images: The German traffic sign detection benchmark," in *Proc. IEEE IJCNN*, 2013, pp. 1–8.
- [7]. Thakur Pankaj and D. Manoj E. Patil "Recognition Of Traffic Symbols Using K-Means And Shape Analysis" *International Journal of Engineering Research & Technology (IJERT)* Vol. 2 Issue 5, May - 2013 ISSN: 2278-0181.
- [8]. Zhou, L., & Deng, Z. (2014). "LIDAR and vision-based real-time traffic sign detection and recognition algorithm for intelligent vehicle". *17th International IEEE Conference on Intelligent Transportation Systems (ITSC)*. doi:10.1109/itsc.2014.6957752.
- [9]. Zakir, U., Edirishinghe, E. A., & Hussain, A. (2012). Road Sign Detection and Recognition from Video Stream Using HSV, Contourlet Transform and Local Energy Based Shape Histogram. *Lecture Notes in Computer Science*, 411– 419. doi:10.1007/978- 3-642-3156
- [10]. Yadav, Shubham & Patwa, Anuj & Rane, Saiprasad & Narvekar, Chhaya. (2019). Indian Traffic Sign Board Recognition and Driver Alert System Using Machine Learning. *International Journal of Applied Sciences and Smart Technologies*. 1. 1-10. 10.24071/ijasst.v1i1.1843.

- [11].Anushree.A., S., Kumar, H., Iram, I., &Divyam, K. (2019). Automatic Signboard Detection System by the Vehicles.
- [12].S. Harini, V. Abhiram, R. Hegde, B. D. D. Samarth, S. A. Shreyas and K. H. Gowranga, "A smart driver alert system for vehicle traffic using image detection and recognition technique," 2017 2nd IEEE International Conference on Recent Trends in Electronics, Information & Communication Technology (RTEICT), Bangalore, India, 2017, pp. 1540- 1543, doi: 10.1109/RTEICT.2017.8256856.
- [13].C. Wang, "Research and Application of Traffic Sign Detection and Recognition Based on Deep Learning," 2018 International Conference on Robots & Intelligent System (ICRIS), Changsha, China, 2018, pp. 150-152, doi: 10.1109/ICRIS.2018.00047.
- [14].Y. Yuan, Z. Xiong and Q. Wang, "VSSA-NET: Vertical Spatial Sequence Attention Network for Traffic Sign Detection," in IEEE Transactions on Image Processing, vol. 28, no. 7, pp. 3423-3434, July 2019, doi: 10.1109/TIP.2019.2896952
- [15].S. Huang, H. Lin and C. Chang, "An in-car camera system for traffic sign detection and recognition," 2017 Joint 17th World Congress of International Fuzzy Systems Association and 9th International Conference on Soft Computing and Intelligent Systems (IFSAS-CIS), Otsu, Japan, 2017, pp. 1-6, doi: 10.1109/IFSAS-CIS.2017.8023239.
- [16].Bi, Z., Yu, L., Gao, H. et al. Improved VGG model-based efficient traffic sign recognition for safe driving in 5G scenarios. Int. J. Mach. Learn. & Cyber. (2020).
- [17].G. Piccoli", E. De Michelib, P. Parodi", M. Campani, " Robust method for road sign detection and recognition", 1996 Elsevier

Cite this article as :

M. Mohsina Kousar, Dr. S. Shoeb Peer, "CNN and Keras based Road Safety Traffic Signs Recognition System", International Journal of Scientific Research in Science and Technology (IJSRST), Online ISSN : 2395-602X, Print ISSN : 2395-6011, Volume 10 Issue 4, pp. 368-374, July-August 2023. Available at doi : <https://doi.org/10.32628/IJSRST52310429>
Journal URL : <https://ijsrst.com/IJSRST52310429>