IoT Based Traffic Sign Detection and Recognition

¹Dr Vijaya Geetha,²Bhargavi S, ³Kavya raj, ⁴Kanisha R, ⁵Mahalakshmi S Department of Electronics and Communication, Dr T Thimmaiah Institute of Technology

Abstract: In today's world, almost everything we do has been simplified by automated tasks. In an attempt to focus on the road while driving, drivers often miss out on signs on the side of the road, which could be dangerous for them and for the people around them. This problem can be avoided if therewas an efficient way to notify the driver without having them to shift their focus. Traffic Sign Detection and Recognition (TSDR) plays an important role here by detecting and recognizing a sign, thus notifying the driver of any upcoming signs. This not only ensures road safety, but also allows the driver to be at little more ease while driving on tricky or newroads. This paper presents a summary of traffic sign detection and recognition. It describes the characteristics and requirements and also difficulties between the road sign identification and recognition of the road signs. It shows the color detection from road sign technique used for verification and classification of the road signs. The paper introduces a traffic sign detection and recognition system that accurately estimates the situation and exact boundary of traffic signs signs using Arduino UNO and color sensor TCS230. During this project, we'll identify the colors of the and recognize the signs accordingly. With this model, we are ready to read and understand traffic signs which are a really important task for all autonomous vehicles. Another commonly faced problem is not being able to understand the meaning of the sign. With the help of this Advanced Driver Assistance Systems (ADAS) application, drivers will no longer face the problem of understanding what the sign says. In this paper, we propose a method for Traffic Sign Detection and Recognition using image processing for the detection of a sign and an ensemble of Convolutional Neural Networks (CNN) for the recognition of the sign. CNNs have a high recognition rate, thus making it desirable to use for implementing various computer vision tasks. Tensor Flow is used for the implementation of the CNN. We have achieved higher than 95% recognition accuracies for circular signs on the Belgium and German data sets.

Keywords: Traffic sign detection, traffic sign recognition, color sensor TCS230, Arduino UNO.

I. INTRODUCTION

Traffic sign detection may be a major crisis in intelligent vehicles; traffic sign recognition provides critical information like directions and alerts in autonomous driving or driver assistance systems. In self-driving cars, the passenger can fully depend upon the car automation for traveling. But to realize level 5 autonomous, it's necessary for vehicles to know and follow all traffic rules. In the world of AI and advancement in technologies, many researchers and large companies like Tesla, Uber, Google, Mercedes-Benz, Toyota, Ford, Audi, etc are performing on autonomous vehicles and self-driving cars. So, for achieving accuracy during this technology, the vehicles should be ready to interpret traffic signs and make decisions accordingly. There are several differing types of traffic signs like speed limits, no entry, traffic signals, turn left or right, children crossing, no passing of heavy vehicles, etc. Traffic signs classification is that the process of identifying which class a traffic sign belongs to.

The traffic signs focus on reduction of the traffic load on existing road network through various travel demand management measures. Traffic signs should remove the encroachments, congestion and improve the traffic signal, road condition and geometrics features at intersections. The traffic signs should be as a guidance or speaker on a road network. Traffic sign reduce the traffic congestion along the road and also provide facilities for the road users. Road signs notify road users of regulations and provide warning

and guidance needed for safe, uniform and efficient operation. The purpose of road signs is to promote road safety and efficiency by providing for the orderly movement of all road users on all roads in both urban and non-urban areas. The use of traffic sign recognition system is very challenging. Since rain, fog, snow etc, affect the whole system. Another thing is the light variation i.e., shadows, sun, clouds etc. The geometrical shape of the object and the perspective is also a big concern. So, we must be able to come up with a system which can work under light variation and geometrical transformation of the objects in a scene. If we narrow down the scope, we are particularly interested in detection of these signs under low light condition. In short, our objective is to make the system very accurate and efficient. In the detection stage, color information is exploited to detect regions of interest (ROI) that may correspond to traffic signs. The shape of these regions is tested in the classification stage, allowing rejecting many of the initial candidates and grouping traffic signs into classes.

The pictogram contained on each ROI (if exist) is extracted, analyzed and compared with the pictogram database. The best match between the ROI and database pictogram, if high enough, is considered the sign that is more likely to appear in that ROI. Each recognized sign is part of the output result of the recognition stage.

II. LITERATURE SURVEY

The method described by Shustanov, P. Yakimov [1] used for Road Sign Detection and Recognition is image processing technique which consist of a group of (CNN) for the recognition called as ensemble. The recognition rate for the CNN is very high, which makes it more desirable for various computer-based vision tasks. The method used for the execution of CNN is TensorFlow. The members of this paper achieved more than 99 percent of accuracy for circular signs on using German data sets.

Walietal [2] describes how they have used to implement a novel method for sign recognition.

They used advanced ARK-2121 technology which is small computer which they installed this tech on the car. The major techniques in the recognition step of the sign were SVM and HOG. They achieved an accuracy of 91% in detection and about 98% average on the classification process.

R. Qian etal [3] describes the analysis and design process of "German Traffic Sign Recognition Benchmark" dataset. The outputs of this project showed that algorithms of machine learning showed very well in recognition of traffic signs. The participants got a very good percentage of 98.98 recognition rate which is as high as human perfection on these datasets.

Liu et al. in [4] is used SIFT matching based SVM for detecting and recognizing broken signs. In this method, images captured from a on board camera and then readjusted SIFT matching technique to a standard camera axis and compared to a reference image by SVM.

Gao et al. [5] used LCH spacing model.fit.chromatic-adaptation transform, which is called Bradford transform. FOSTS model is then applied for shape analysis. Recognition process is performed by comparing the 49-dimensional vectors representing a current image with template vectors stored in the database that has been classified into several color/shape subgroups with an accuracy of 95%.

III. METHODOLOGY

Traffic sign detection may be a major crisis in intelligent vehicles, traffic sign recognition provides critical information like directions and alerts in autonomous driving or driver assistance systems. you all have heard about the self-driving cars during which the passenger can fully depend upon the car for traveling. But to realize level 5 autonomous, it's necessary for vehicles to know and follow all traffic rules. In the world of AI and advancement in technologies, many researchers and large companies like Tesla, Uber, Google, Mercedes-Benz, Toyota,

Ford, Audi, etc are performing on autonomous vehicles and self-driving cars. So, for achieving accuracy during this technology, the vehicles should be ready to interpret traffic signs and make decisions accordingly. There are several differing types of traffic signs like speed limits, no entry, traffic signals, turn left or right, children crossing, no passing of heavy vehicles, etc. Traffic signs classification is that the process of identifying which class a traffic sign belongs to. The framework we proposed is categorized into three stages: Detection and feature extraction and recognition. The detection stage is just used to find a road sign. At the point when a vehicle is travelling at aspecific speed, the camera catches the road sign in nature, and our calculation verifies whether a sign is available in that outline or not available in that perimeter. Distinguishing the traffic sign depends on shape and color. In the feature extraction stage, the proposed calculation characterizes the distinguished road sign. This is accomplished with the assistance of "Convolutional Neural Network" algorithm which classifies the image into sub classes.

In this project, we make a CNN block where predictions are directly preformed across multiple feature levels. For this project, we are using the general public dataset available at Kaggle i.e GTSRB. Our approach to putting together this traffic sign classification model is discussed in four steps:

Step 1: Explore the data set Our 'train' folders contain 43 folders each representing a special class. The range of the folder is from 0,1,2, up to 42. With the assistance of theOS module, we iterate over all the classes and append images and their respective labels within the data and labels list.

Step 2: Build a CNN model to classify the pictures into their respective categories, we'll build a CNN model (Convolutional Neural Network). CNN is best for imageclassification purposes.

Step 3: Train and validate the model After building architecture, we then teach the model using

Step 4: Test our model with test data set Our data set contains a test folder and during a test.csv file,

we've the small print associated with the image path and their respective class labels. Now we are getting to build a graphical interface for our traffic signs classifier with Tkinter. Tkinter is nothing but a GUI toolkit within the standard python library. Here we upload the pictures and classify the image.

This project we'll develop using python. We are getting to develop a model which can detect the traffic sign. We build deep neural network model which will identify which traffic sign is present therein image. We also used PIL library to open image content into array. Our data set contain train folder which carries folder each represents different classes and in test folder we've the small print associated with the image path and their respective class labels.

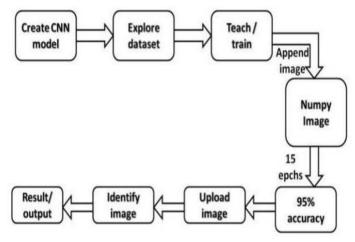


Fig 1: System architecture

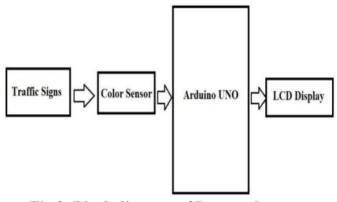


Fig 2: Block diagram of Proposed system

• The figure 2 describes the block diagram of IoT based traffic sign detection and recognition.

188N [Unline]: 2583-26

• First, provide the input as traffic sign images that as to be detected and recognized.

• The color sensor which is used to sense the colors, will identify the type of input traffic image based on the RGB colors.

• The Arduino UNO will take the analog input from the color sensor and converts it into digital values.

• Finally the desired output is displayed on the LCD.

IV. HARDWARE AND SOFTWARE REQUIREMENTS

Hardware Requirements

A. Color Sensor (TCS230)

Colour sensors are developed based on diffuse technology that can detect a wide range of colours. The combination of colour sensitive filters and sensors array perform colour sensing, which is further used to analyse the colour present in an image or in a specified object. The colour measurement process involves a light source to illuminate the surface, the target surface, and a receiver that measures the reflected wavelengths. A white light emitter is used to illuminate the surface. The sensor then activates three filters with three wavelength sensitivities to measure the wavelengths of RGB colours respectively. Based on these three colours, the colour of the material is determined. Modern colour sensing has seen the involvement of fibre optics in the colour detection process. In this technology, the light transmission to the object and back depends on the optical glass fibres, which operate on the principle of total internal reflection. This phenomenon causes fibre to act as a waveguide and enables the complete reflection of light. In a fibre optic colour sensor, the white light spot is projected via the fibreopticonto the target surface; the part of light that is reflected back from the target is directed onto the detector via the same optical fibre. The reflected light is then separated into long, medium and short-wave light components. and transformed into L*a*b* colour coordinates. Colour

sensors are majorly used to grade coloured products, distinguish coded markings, detect the presence of adhesives or data codes on a package. The technology has a wide range of applications in various industries such as textile, automation, automotive, food, printing, pharmaceutical, and many more. These sensors can also be programmed to identify any one colour or multiple colours for sorting applications, based on the level of sophistication required in the colour measurement process.

B. Arduino UNO

Arduino UNO is a microcontroller board based on the ATmega328P. It has 14 digital input/output pins (of which 6 can be used as PWM outputs), 6 analog inputs, a 16 MHz ceramic resonator, a USB connection, a power jack, an ICSPheader and a reset button. It contains everything needed to support the microcontroller; simply connect it to a computer with a USB cable or power it with a AC-to-DC adapter or battery to get started.

C. LCD (Liquid crystal display)

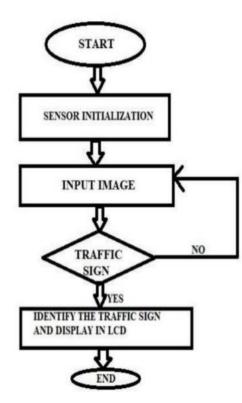
A liquid-crystal display (LCD) is a flat-panel display or other electronically modulated optical device that uses the light-modulating properties of liquid crystals combined withpolarizers. Liquid crystals do not emit light directly[1] but instead use a backlight or reflector to produce images in color or monochrome. LCDs are available to display arbitrary images (as in a general-purpose computer display) or fixed images with low information content, which can be displayed or hidden. For instance: preset words, digits, and seven-segment displays, as in a digital clock, are all good examples of devices with these displays. They use the same basic technology, except that arbitrary images are made from a matrix of small pixels, while other displays have larger elements. LCDs can either be normally on (positive) or off (negative), depending on the polarizer arrangement. For example, a character positive LCD with a backlight will have black lettering on a background that is the color of the backlight, and a character negative LCD will have a black background with the letters being of the same color as the backlight. Optical filters are

added to white on blue LCDs to give them their characteristic appearance.

Software Requirements

• Jupyter Notebook

V. WORKING OF PROPOSED MODEL



Step 1: Start

Step 2: Sensor Initialization

Step 3: Input Image

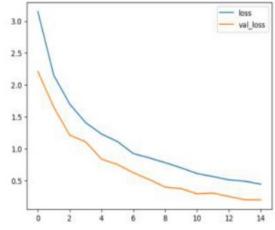
Step 4: Is The Given Image A Traffic Sign If No Go To Input Image If Yes Identify The Traffic Sign And Display In Lcd Step 5: End

VI. RESULTS

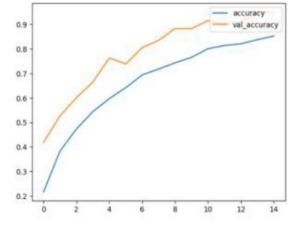
In this project we've successfully classified the traffic signs using color sensor to sense the colors of road signs and identify the signs respectively & CNN classifier with 95% accuracy and also visualized how our accuracy and loss changes with time, which is pretty good from an easy CNN model.

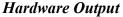
				CL	assid		Name
20 -	-			0	0	Speed limit (5km/h)
	~			1	1	Speed limit (1	
**	11	1		2		Speed limit (3	
60-11				3		Speed limit (4	
80 - 08	-	2		4			
00				4	4	Speed limit (5	0km/h)
0 20	40	60 80					
				"sequential_s"			
Out[5]:	Classid	Name	1. approx	Clubel	43 4	National Stragon Charana, J.S.S., J.S.S., 1	Par an 1
53		Give Way		aling (Mancaling) of (Linno2013	- 14	(More, 225, 224,) (More, 222, 223, 1	0 0
			*****	nolingit (Hadro	11-620	(Barre, 111, 111, (Barre, 120, 100, 1	
54		No stopping		poolinged_2 (Made	ond long	(neme, 54, 58, 64	r) #
55		No entry	- C CARTAN	nt_# (Convers) monlinged_# (Mase	miling	(Mane, 12, 12, 13) (Mane, 21, 25, 13)	
56	56	Unknown7	6 10010	ongguding"s (umo		(Mare, 34, 24, 25) (Mare, 52, 52, 25	() 200104 (6) 6
57	57	Unknown8	#3.40	ten (Flatten)		Dione, Juana;	
			dation	· (Derrya)		(Naume, SAN)	\$259364
			eed limit (15kn	n/h) Uni	cnow	nl Spee	d limit (5
ratch out for ca		9	(15)		-		No entr
0	6	take from Len	tatch out far co	rispeed to		60km/h)	So No entr
CO beed limit (Skr	6	take from Len	tatch out far co	rispeed to	-	60km/h)	50
Epoch 18/15		Take from Left	tatch out far co	inspeed to	6 C	sokm/h)	No entr
Epoch 18/15 185/105 [Take from Left	tatch out for ce	inspeed to	6 C	sokm/h)	No entr
Epoch 18/15 185/105 [y: 0.8825 Epoch 11/15	vitipeed li	Take from Len mil (70km/h)	atch out for ca Dont Go Right (step - loss: 8.6972	nispeed to Ni - accuracy: 0	600 bor	eokm/h)	No entr Unknowr 696 - val_acc
Epoch 10/15 185/105 [90 0.8825 Epoch 11/15 185/105 [vitipeed li	Take from Len mil (70km/h)	tatch out for ce	nispeed to Ni - accuracy: 0	600 bor	eokm/h)	No entr Unknowr 696 - val_acc
Epoch 18/15 185/185 [9: 0.8825 Epoch 11/15 185/185 [0.9149	vitipeed li	Take from Len mil (70km/h)	atch out for ca Dont Go Right (step - loss: 8.6972	nispeed to Ni - accuracy: 0	600 bor	eokm/h)	No entr Unknowr 696 - val_acc
Epoch 18/15 185/185 [9: 0.825 Epoch 11/15 185/185 [9: 0.825 Epoch 11/15 Epoch 12/15		mit (70km/h)	atch out for ca Dont Go Right (step - loss: 8.6972	- accuracy: 8.8	6 0 .7653 -	60km/h)	No entr Dinknown 695 - val_accur
Epoch 18/15 185/185 [9: 0.825 Epoch 11/15 185/185 [9: 0.825 Epoch 11/15 Epoch 12/15		mit (70km/h)	Dont Go Right /step - loss: 0.6890 -	- accuracy: 8.8	6 0 .7653 -	60km/h)	No entr Dinknown 695 - val_accur
Epoch 18/15 185/185 285/185 285/185 285/185 285/185 285/185 2000 2010 2010 2010 2010 2010 2010 201		mit (70km/h)	Dont Go Right /step - loss: 0.6890 -	- accuracy: 8.8	6 0 .7653 -	60km/h)	No entr Dinknown 695 - val_accur
Epoch 13/15 185/105 [195/105 [195/105 [195/105 [9.949 Epoch 13/15 185/105 [Bopch 13/15 195/105 [V78peed II	mit (70km/h)	Dont Go Right /step - loss: 0.6890 -	- accuracy: 8.8 accuracy: 8.8	mit (0 6 0 1,7653 - 1 135 - 1	60 km/m)	No entr Unknown 696 - val_accur 4 - val_accur
Epoch 14/15 185/185 [190 11/15 185/185 [8.9149 8.9149 [8.9149 Epoch 12/15 185/185 [185/185 [185/185 [185/185 [9.9899	V78peed II	mit (70km/h)	Dont Go Right /step - loss: 0.6872 tep - loss: 0.6890 - tep - loss: 0.5687 -	- accuracy: 8.8 accuracy: 8.8	mit (0 6 0 1,7653 - 1 135 - 1	60 km/m)	No entr Unknown 696 - val_accur 4 - val_accur
Epoch 14/15 Epoch 14/15 Epoch 11/15 Epoch 11/15 Epoch 11/15 Epoch 11/15 Epoch 12/15 Epoch 13/15 Epoch		- 5225 95/5 - 11815 115	Domt Go Rught (step - loss: 0.697 tep - loss: 0.697 tep - loss: 0.5697	- accuracy: 0.8 accuracy: 0.8	60 0 hor 1,7653 - 1 135 - 1 .8281 -	60km/dv) 1 val_loss: 0.3 val_loss: 0.34 val_loss: 0.362 val_loss: 0.362	No.entr Unknown 696 - val_accur 4 - val_accur 477 - val_accur
Epoch 14/15 185/185 Epoch 14/15 185/185 Epoch 14/15 185/185 Epoch 12/15 185/185 Epoch 12/15 185/185 Epoch 13/15 185/185 Epoch 13/15 185/185 Epoch 14/15 185/185 Epoch 14/15 185/185 Epoch 14/15 Epoch		- 5225 95/5 - 11815 115	Dont Go Right /step - loss: 0.6872 tep - loss: 0.6890 - tep - loss: 0.5687 -	- accuracy: 0.8 accuracy: 0.8	60 0 hor 1,7653 - 1 135 - 1 .8281 -	60km/dv) 1 val_loss: 0.3 val_loss: 0.34 val_loss: 0.362 val_loss: 0.362	No.entr Unknown 696 - val_accur 4 - val_accur 477 - val_accur
Epoch 14/15 185/185 [29/14/15 185/185 [8.9149 Epoch 11/15 185/185 [8.9149 Epoch 12/15 185/185 [9.9889 Epoch 13/15 185/185 [185/185 [8.9492 Epoch 14/15 185/185 [- 5225 95/5 - 11815 115	Domt Go Rught (step - loss: 0.697 tep - loss: 0.697 tep - loss: 0.5697	- accuracy: 0.8 accuracy: 0.8	60 0 hor 1,7653 - 1 135 - 1 .8281 -	60km/dv) 1 val_loss: 0.3 val_loss: 0.34 val_loss: 0.362 val_loss: 0.362	No.entr Unknown 696 - val_accur 4 - val_accur 477 - val_accur
Epoch 14/15 185/105 [197] v 0.3825 Epoch 11/15 185/105 [Epoch 12/15 185/105 [185/105 [185/105 [97] v 0.3989 Epoch 12/15 185/105 [95] v 0.3989 Epoch 14/15 185/105 [9, 542 Epoch 15/15		rrife (70km/h) r	Domt Go Rught (step - loss: 0.697 tep - loss: 0.697 tep - loss: 0.5697	- accuracy: 0.8 accuracy: 0.8 - accuracy: 0.8 - accuracy: 0.8	6 0 6 0 6 0 6 0 6 0 6 0 6 0 6 0	COLONY (*) (* val_loss: 0.3) (* val_loss: 0.3) (* val_loss: 0.30) (* val_loss: 0.30) (* val_loss: 0.20) (* val_loss: 0.20) (* val_loss: 0.20)	No entr No entr Conknown (mknown))

0+1+1+1+



Out[21]: <matplotlib.legend.Legend at 0x2b25d143220>











VII. CONCLUSION

In this paper, we proposed an efficient traffic sign detection and recognition method to the present end; we generalized the traffic sign templates with precise boundaries and high accuracy. To achieve practical detection speed, we have used color sensor to recognize the road signs and we explored the bestperforming convolution neural network for both detection & recognition considering the characteristics of traffic signs. By using the images of traffic signs, our method effectively utilizes strong information of target shapes to the drivers. The hardware used in this project is very less when compared to other models, which reduces cost and also free from hardware impairments. Color features extracted using vision models can perform accurate recognition for traffic signs located at a reasonable distance for still images under various viewing conditions. This approach shows a good performance for a wide variety of traffic signs of different colors, forms, and informative content. For detection of a sign, this paper depends on color and shape of the sign. CNN algorithm is preferred more compared to

Swanirman Sunirmit Publications of Research Volume 4, Issue 1-March 2024 | [2024-25]

other algorithms since it provides high efficiency and there is a problem if there is a reflection on the sign which impacts its color. Similarly, if the sign is not proper or cut off, the shape of the sign is impaired, thus resulting in improper detection of the sign which leads to fault detection. Another important issue to consider is detection in the night. If the camera used is infrared then there is no problem in detecting signs but in case of no infrared webcam which may not detect signs accurately which causes high chance of accidents. Adding text to speech module in our model makes the driver effortless and makes him to concentrate completely on driving rather than checking for traffic signs. This reduces the occurrence of accidents during night and as well as day time. The hardware used in this project is very less when compared to other models, which reduces cost and also free fromhardware impairment

REFERENCES

[1] Jisha Elizabeth Shaji1, Hari S "Automatic Traffic Sign Detection and Recognition using CNN" 2019

[2] Alexander Shustanova "CNN Design for Real-Time Traffic Sign Recognition" 2017..
[3] C. Liu, F. Chang, and C. Liu, "Occlusion-robust traffic sign detection via cascaded colour cubic feature," IET Intell. Transp. Syst., vol. 10, no. 5, pp. 354–

360, 2015 Explore resulting or any accomplite for Statisticable Ecosystem or Society

[4] A. Mogelmose, D. Liu, and M. M. Trivedi, "Detection of U.S. traffic signs," IEEETrans. Intell. Transp. Syst., vol. 16, no. 6, pp. 3116–3125, Dec. 2015.

[5] O. Dabeer et al., "An end-to-end system for crowdsourced 3D maps for AL vehicles: The mapping component," in Proc. IEEE/RSJ Int. Conf. Intell. Robots Syst., Sep. 2017, pp. 634–641..'

[6] Gudigar, C. Shreesha, U. Raghavendra, and U. R. Acharya, "Multiple thresholding and subspace based approach for detection and recognition of traffic signs," Multimedia Tools Appl., vol. 76, no. 5, pp. 6937–6991, 2017.

[7] Shrivastava, A. Gupta, and R. Girshick, **"Training region based on object detectors with online hard example mining,"** in Proc. Comput. Vis. Pattern Recognize., Jun. 2016, pp. 761–769.

[8] E. Perez and B. Javidi, "Composite filter bank for road sign recognition," presented at 13th Annual Meeting IEEE Lasers and Electro-Optics Society, Rio Grande, Puerto Rico, 2000.

[9] Ebner, "A parallel algorithm for color constancy," J. Parallel Distrib. Comput., vol. 64, pp. 79-88, 2004.

[10] Shima Damadam, Zourbakhsh, Reza, Azadeh, "An Intelligent IoT Based Traffic Light Management System: Deep Reinforcement Learning" 5(4), 1293-1311, 2022.

[11] Archana, Nikitha, Himabindu, Manvita, "IoT based Signal Vioted Vehicle Detector" 4(3), 814-821, 2022.

[12] S. Loyola Samraj, N.V. Bhalke, A. Aarthi, R. Srinath and E. Prabhu, "Robust

Smart Home Monitoring System Based on 802.11 Mesh Network", Lecture Notes in Networks and Systems, vol. 98, pp. 38-47,2020.

Sunirn

ISSN [Online]: 2583-2654

www.swanirmanconsultancy.in