

# Efficient Approach for Traffic Sign Detection based on Saliency Detection via Graph-Based Manifold Ranking

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**Abstract** - Now days, traffic sign detection attracted large number of researchers interest due to its important in efficient intelligent transportation systems (ITSs). Traffic sign detection helps to minimize the road accidents and hence to minimize the loss. The efficient traffic sign detection is vital research problem since from last decade. In ITSs, there is more work conducted already on name plate detection and recognition, however there only few concrete research studies presented to solve the problem of traffic sign detection and recognition. For applications like road surveying, autonomous vehicles are mainly demanding the system of road side sign detection and recognition. The current methods for road side sign detection having issues of efficiency and accuracy due to different factors affecting on road sign detection such as shadow, non-uniform sizes of signs, illumination conditions, blurring, occlusion, and sign deterioration etc. In this work, we proposed novel method for road sign detection and recognition based on saliency regions detection. Saliency regions detection helps to locate the road sign efficiently and hence traffic sign detection properly. In this paper, author introduced the efficient segmentation method and graph based ranking approach for the accurate detection salient regions. Additionally we applied the RGB image smoothing algorithm to improve the detection accuracy. The performance results claims that proposed approach outperforming the previous method.

**Index Terms**— Saliency Detection, Traffic Sign Detection, Segmentation, Salient regions, Manifold ranking

## I.INTRODUCTION

The ITSs has been developed for the security of human for reducing the road accident in recent time. There are most of accident are happening on rural roads due to driver disturbance. And even many times drivers are not checking the traffic signs board & un-careful driving. The refined methods into the pc vision have been used by state of art for the detection of traffic sign it is analysing the past decade. On-road applications of vision have enclosed lane detection, driver distraction detection, and resident create logical thinking [1]. To someone better-known with recent advances in computer vision, the matter of traffic sign detection and recognition might sound simple to resolve. A traffic sign has been fairly easy objects with heavily affected appearances. The general traffic sign detection methods have been explained in this article. The standalone solution was not operate properly that's what there is the utilization of firm strategies, The true & false positive rates they are not able the forward, In order to collaborating the methods incorporates the synergistic effect, However, generally such as building blocks of the greater detection systems utilized by them [2]. The 3 main categories have been confirmed for traffic sign detection ways: shape primarily based, colour primarily based, &

strategies which are based on machine learning. once introducing the strategies, they present two traffic sign detection systems which use them. Computers are excellent at addressing numerical vector. In recent years, the functions of the algorithmic rule in image classification are additional and a lot of bigger, welcome by the bulk of image process enthusiasts [3]. Therefore, this paper carries out analysis behind the identification of many typical traffic signs. Traffic signs segmentation algorithm utilized to separate sign from the background is studied basis on the repetitive segmentation and most variance between clusters. Then with SIFT elements of sign extracted, the codebook is generated by these feature cluster and pictures are represented by histograms with the use of BOW model and traffic signs classifier is implemented on the basis of SVM at an equivalent time. The experimental results basis on the traffic sign images collected within the natural surroundings demonstrate the effectiveness and usefulness of the BOW model classification algorithm. The traffic sign images has been known normally in these methods: One is that the identification to the colour and shape elements of traffic signs combined with example matching, a picture is typically thought of as a separate object in these strategies that are simply suffering from the item angle, intensity of the light, occlusion of objects and

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alternative factors within the method of characteristic the focused area, and therefore the strategies are appropriate for a scenario of fewer samples [4].

Before collaborating of the image with the neural network & support vector machines they did the feature extraction of image. This type of methods has been extra hard for the little influence of the replacement of existing innovation which is creates the effect on it & it was conferrable for the classification of image into the additional samples. The algorithm of the image feature extraction has been SIFT options & Hu invariant moment [5]. Bag of Words algorithmic rule utilized in the sector of image process is developed in recent 10 years and it's been investigated wide [6]. Bag of Words model includes image sampling, feature description, code book and classifier architecture. Bag of Words model are often imaginary as a large document assortment D a complete of M documents in D, and every document are often described as the N vector.

For the recognition of variety of applications & accurate & efficient detection of traffic sign text the Intelligent Transportation Systems has been needed. With the use of graph based ranking, pre-processing algorithm & segmentation methods, for the traffic sign detection the novel approach has been proposed in this article. The related work study of the various methods has been shown in section II. In section III, the design and algorithms are presented for proposed work. In section IV, the current simulation results, performance metrics are discussed. Finally the conclusion and future work presented in section V.

## II. RELATED WORKS

This section shows the various methods of recent traffic signs. On the study of recognition & detection information of traffic road side signs texts there is very normal researches has been done still. There are several researches on the license plate recognition. On the road side traffic sign detection & recognition very rare study has been done by the authors.

In [1], after consideration of 2 various modules author confirmed traffic sign detection model. Extraction of ROI (region of interest) is the first module. This module transforming the images colour to gray pictures through learning analysis which is the specific colours for traffic signs which is extra described into the grey images It follows shape example matching, wherever a group of templates for every target class of signs has been designed. After that, a group of ROIs is generated. The second module is for recognition. It

validates if an ROI belongs to a target class of traffic signs by supervised learning. Native shape and colour elements have been extracted.

In [2], The traffic sign detection techniques has been described by author into this paper which is taking the primary base of prohibitory & needs signs & danger signs into the GTSDDB competition is the 3rd place. The coarse to fine window theme & oriented gradient histogram has been used by this strategy. Candidate ROIs are 1st roughly detected among a small-sized window, so additional verified among a large-sized window for higher accuracy. Experimental results show that the projected methodology achieves high recall and exactitude ratios, and is powerful to varied adverse conditions as well as unhealthy lighting condition, partial occlusion, poor quality and little projective deformation.

In [3], author proposed programmed street sign discovery and acknowledgment framework in view of support vector machines (SVMs). In programmed activity sign support and in a visual driver-help framework, street sign location and acknowledgment are two of the most imperative capacities. There approach could identify and perceive round, rectangular, triangular, and octagonal signs and, thus, covers all current Spanish activity sign shapes. Street signs give drivers critical data and help them to drive all the more securely and all the more effortlessly by controlling and cautioning them and accordingly managing their activities.

In [4], author proposed novel way to deal with discrete sign discovery from characterization, yet at the same time measure the execution on significant classifications of signs to take into account benchmarking specific arrangements. The considered standard calculations speak to the absolute most well known recognition methodologies, for example, the Viola-Jones finder in light of Haar highlights and a direct classifier depending on HOG descriptors. Further, an as of late proposed issue particular calculation abusing shape and shading in a model-based Hough like voting plan is assessed. At last, we introduce the best-performing calculations of the IJCNN competition.

In [5], For detecting speed signs, writer proposed radial symmetry detector. They assessed the detector itself in a system that is mounted inside a street vehicle. They also assessed its execution that is coordinated with classification over a series of

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sequence from streets around Canberra and demonstrate it while running on the web in our street vehicle. With high reliability in real time they show that it can detect signs w. For receiving to identification of street sign they have been analyzed the interior parameters of calculation. They exhibit the stability of the system under the variety of these parameters and show computational speed increases through their tuning. The detector is shown to work under a wide assortment of visual conditions.

In [6], two types of spectral segmentation algorithms are introduced by author: K-way division and hierarchical leveled division. The writer composed algorithms give superb divisions which save protest subtle elements by specifically fusing the full-go associations. Moreover, Throughout the opposite of scanty lattice the proclivity framework has been indicated, it was the Eigen deterioration which is processed effectively. The experimental comes about on the BSDS and MSRC picture databases show the prevalence of their division algorithms as far as significance and accuracy contrasted and existing mainstream strategies.

In [7], they presented novel technique for bland visual order: the issue of identifying object data of natural image while summing up crosswise over varieties inalienable to the object class. Strategy for pack key point is completely supported on the vector quantization of relative invariant descriptors which is known as patches of image. They proposed and thought about two option implementations using different classifier: Naïve Bayes and SVM.

In [8], /the approach of the novel graph based traffic sign detection has been conferred by author in this paper which is known as saliency measure stage, multi threshold segmentation stage & graph based ranking stage.

In [9], for the recognition of text in traffic signs & automatic detection author described the novel approach. Within the image for consisting the search regions scene structure has been utilized by researcher, where author found the traffic sign candidates. Maximally stable extremely regions (MSERs) and hue, saturation, and value color three holding are used to locate a large number of candidates.

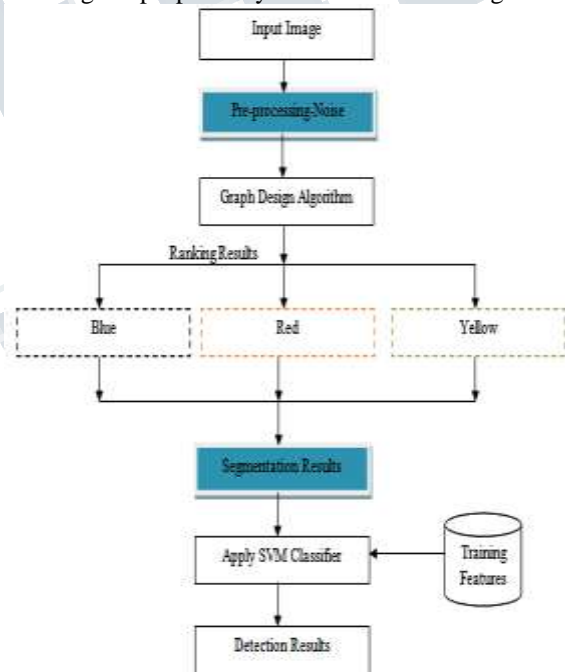
In [10], the appropriate extension of usually existing LISA-TS traffic sign data set has been described by researcher, even most equivalent its size, along with

the high definition quality for U.S. traffic signs detection.

In [11], the traffic sign detection has been proposed by author which is able to fast detecting multiclass traffic signs into the high resolution images when the higher definition node has been achieved.

**III. METHODOLOGY**

Into the previous many methods by many author nothing considered important work related to the noise removal & image smoothing. Into the ITSs, during transmission processes & video capturing noise has been described; and for performing the noise removal for image & video processing this was the essential task. In this research we implemented the approach of considering image pre-processing for RGB image and then applied graph based technique and segmentation methods for traffic sign detection. Below figure 1 is showing the proposed system architecture diagram:



**Figure 1: Architecture of proposed approach**

Based on above diagram, below are different algorithms involved in our process.

**Algorithm 1: Detection Algorithm**

Input: RGB image

Output: Detection Signs

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- Step 1: browse input image I
- Step 2: P = preprocess (I);
- Step 3: Apply Graph Ranking Algorithm on P
- Step 4: Perform Segmentation of Blue, Red and Yellow colours.
- Step 6: Load Train dataset
- Step 7: Apply SVM (Test, Train)
- Step 8: Display Classification Results on RGB image I.
- Step 9: Measure End Time
- Step 10: Stop

The algorithm for pre-processing is designed as:

**Algorithm 2: Preprocess (I)**

1. The input I needs to be preprocessed.
2. Extract R, G and B Channels
3. On each channel, 2D conversion, if the input image is 3 dimensional (3D) then it is first converted into 2D, as most of image processing methods are applied on 2D images only.
4. The next task of preprocessing phase is image smoothing and denoising.
5. This can be done by using two filters as mentioned below
6. Out\_1 = apply laplacian filter on R channel
7. Out\_2 = apply mean filter on R channel
8. Out\_R = Out\_1 – Out\_3
9. Out\_1 = apply laplacian filter on G channel
10. Out\_2 = apply mean filter on G channel
11. Out\_G = Out\_1 – Out\_3
12. Out\_1 = apply laplacian filter on B channel
13. Out\_2 = apply mean filter on B channel
14. Out\_B = Out\_1 – Out\_3
15. Out\_RGB = cate (Out\_R, Out\_G, Out\_B)
16. Return Out\_RGB

17. End

Rest the algorithms for graph-based saliency measure algorithm is referred from the [8] for our work. Also the multi-threshold segmentation algorithm is utilized from [8]. The quality enhancement of into the performance is the vital reason of the use of this algorithm. Into the following algorithm 3 steps are given.

**Input:**

$\Omega$ : Complete affinity matrix  
 W: affinity matrix is defined bt Eq. (1)  
 S: saliency map is defined bt Eq. (3)

**Output:**

$\Omega$ : complete affinity matrix which combines the information of saliency and contextual relationship of nodes.

Step 1: because all of the nodes belonging to traffic sign region have high saliency values, the relevance's of the i-node and the nodes having low saliency values should be ignored as follows:

**If**  $S(i) = 0 \quad i \in \{1, \dots, N\}$ ,

**Then**  $\Omega(j, i) = 0, \quad j = \{1, \dots, N\}$ ,

**Otherwise**  $\Omega(j, i) = \Omega(i, j)$

Step 2: due to the fact that the complex background is liable to affect the ranking results, the relevance's of the i-node and unconnected nodes should be ignores as follow:

**If**  $W(i, k) = 0, \quad i \in \{1, \dots, N\}, \quad k \in \{1, \dots, N\}$ ,

**Then**  $\Omega(i, k) = 0$ ,

**Otherwise**  $\Omega(i, k) = \Omega(i, j)$

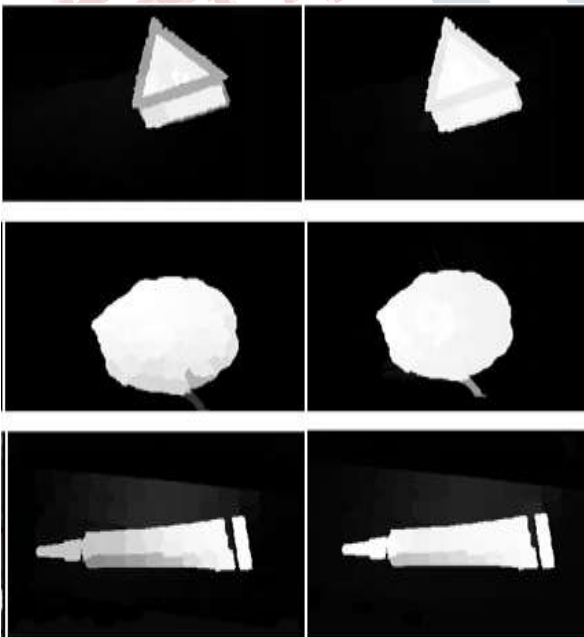
**IV. RESULTS**

The designed method will be simulated using the MATLAB tool by using the research datasets like Spanish Traffic Sign Set and GTSDDB Dataset. The comparative study among existing and proposed methods is done by using the four main parameters such as accuracy, precision, recall rates and processing time.





**Fig.2. a) Original Image b) Superpixel image**  
 Figure 2 is showing the deference between the original image and super pixel image after simulating the practical work



**Fig.3. a) Saliency Map Stage -1 Image b) saliency Map Stage-2 Image**  
 Figure 3 is showing deference between the Saliency Map Stage-1 and Stage-2 analysis.

**V. CONCLUSION AND FUTURE WORK**

We propose a bottom-up method to detect salient regions in traffic sign images through manifold ranking on a graph, which incorporates local grouping cues and boundary priors. We adopt a two-stage approach with the background and foreground queries for ranking to generate the saliency maps. We evaluate the proposed algorithm on large datasets and demonstrate promising results with comparisons to fourteen state-of-the-art methods. Furthermore, the proposed algorithm is computationally efficient. Our future work will focus on integration of multiple features with applications to other vision problems.

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