

# A Robust Technique for Improved Color Image Segmentation by SVM Classifier followed by Integrated Techniques

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**Abstract** - In existing the segmentation is based on color or texture or on both color and texture but proposed method for color image segmentation is based on both color and texture with multi-class SVM (Support Vector Machine). For color feature extraction we used homogeneity model and for textural features we used PLD (Power Law Descriptor). With the help of SR-FCM (Soft Rough Fuzzy-C-Means) clustering. Membership functions based on the fuzzy set are facing the major problem of cluster overlapping. The rough set concepts can help us to get correct data from incomplete data, uncertainty of data. For defining the soft set theory there is no any requirement of parameterization tools. To get improved results of proposed algorithm the combination of aspects of fuzzy sets, rough sets as well as soft sets are used. The feature extraction for textural feature is done by using spatial domain which helps to reduce the run time complexity. Proposed method provides better performance compared to all state of art techniques which is developed and analyzed using MATLAB.



**Fig 1. Image Segmentation into Super pixels**

**Keywords:** Homogeneity, Soft Sets, Multi Class SVM, Texture, Power Law Descriptor. Segmentation, Classification, Clustering, Fuzzy Sets, Rough Sets.

## I. INTRODUCTION

### 1.1 Image Segmentation

The color image segmentation in preprocessing step of prime importance, connected application such as a robotic vision, used in numerous computer vision and image processing, content based image retrieval as well as medical imaging, face recognition. The four major groups of image segmentation algorithm can be categorized by clustering, thresholding and edge based as well as region based segmentation. Segmentation is the nothing but getting ROI (region of interest) by dividing image into blocks or we can say partitioning of an image. Image segmentation is an important task for specific application in digital image processing. There are different techniques used for image segmentation which can be done in both special and frequency domain.

The image segmentation is the process of partitioning a digital image into multiple segments is also known as sets of pixels or super pixels. The aim of segmentation is too simple as well as changes the representation of images is extra meaningful as well as easy to analyze. Therefore image segmentation is used to locate objects as well as boundaries (curves and lines etc.) in images. In exact term, the process of image segmentation to allocate a small pieces of paper to every pixel in an image to the extent that pixel with same label.

In image segmentation are set of segments that collect cover page of entire image, a set of an outline representing remove from the image. Each pixel of the region is similar with respect to the some computed property or characteristics, like that texture, intensity or color. Therefore take to a regions are different and significantly with respect to the same characteristics. When the applied to the stock of images, things in medical imagining, the resulting an outline representing of after image segmentation can be used to reconstruction with help of interpolation algorithm for example Marching cubes.

In this techniques clustering are explored in recent times for color image segmentation. Wang et al, in their work applied to pixel the homogeneity as well as texture features to SVM, hence the Fuzzy C Means (FCM) algorithm because using by the preliminary clustering Lingens et al, proposed to rough k means algorithm for used to the clustering of internet users, which applied for image segmentation applications. Sanker pal and Pradipta Maji proposed to RFCM, in which presented that, the crisp lower bound as well as fuzzy boundary of a class, to enables efficient selection of cluster prototypes.

It shows that, "Soft Rough Fuzzy C means as well as Multi Class SVM using Color image segmentation "is presented. To start the color as well as texture a thing said of the color image, the pixel levels are obtained through the Power Law Descriptor as well as homogeneity. In these characteristics are applied to the Soft Rough Fuzzy C means (SRFCM) are clustering algorithm. After the multi class SVM classifier is trained by using the samples secure from SRFCM clustering. The steps of image segmentation are completed with trained by the Multi Class SVM.

The association of the paper is as per the following. The primers of SRFCM bunching are explaining about in Section 2. The fundamental ideas of Two Class SVM and Multi-Class SVM are examined in area 3. The basics of Power Law Descriptor are explain about in Section 4. In segment 5 the proposed Color picture division utilizing SRFCM bunching and Multi class SVM is explain about, pursued by defense for utilizing this calculation. In Section 6 the presentation estimates utilized in assessing the division calculation are displayed. Area 7 demonstrates the pictorial and goal assessment after effects of the proposed calculation. The finishing up comments is given in area 8.

## II. LITERATURE REVIEW

[1] Augustin, Thomas. "The problem of meaningfulness: Weber's law, Guilford's power law, and the near-miss-to-Weber's law." *Mathematical Social Sciences* 57.1 (2009): 117-130.

The present paper gives reformulations of regular models of separation, as for example, Weber's law, Guilford's capacity law, and the close to-Weber's law. All models depend on the presumption that the model parameters may rely upon the hidden physical improvement scale by which the physical forces are estimated. By talking about the importance of the model parameters, we arrive at the resolution that the close to Weber's law can be connected to physical amounts which are estimated on a proportion, interim, or log-interim scale, though uses of Weber's law and Guilford's capacity law are essentially limited to physical proportion scales. At last, we talk about the importance of an exactly based sub model of the close to-Weber's law, alluded to as fixed-point model. The outcomes

demonstrate that, from a hypothetical perspective, the fixed-point model is better than Guilford's capacity law and the close miss-to-Weber's law.

[2] Cheng H.D, Jiang X.H, Sun Y, and Wang. J, "Colour Image Segmentation: Advances and prospects" *The Journal of Pattern Recognition Society* (2001)2259-2281.

Picture division is basic and basic to picture handling and example acknowledgment. This overview gives a synopsis of shading picture division procedures accessible at this point. Fundamentally, shading division methodologies depend on monochrome division methodologies working in various shading spaces. In this manner, we initially examine the significant division approaches for portioning monochrome pictures: histogram thresholding, trademark highlight grouping, edge discovery, district based strategies, fluffy procedures, neural systems, and so on.; at that point survey some real shading portrayal techniques and their favorable circumstances/detriments; at last outline the shading picture division methods utilizing diverse shading portrayals. The utilization of shading models for picture division is additionally talked about. Some tale methodologies, for example, fluffy strategy and material science based technique are explored too.

[3] Chen, Jie, et al. "A robust descriptor based on weber's law." *Computer Vision and Pattern Recognition, 2008. CVPR 2008. IEEE Conference on. IEEE, 2008.*

Motivated by Weber's Law, this paper proposes a basic, yet extremely ground-breaking and powerful neighborhood descriptor, Weber Local Descriptor (WLD). It depends on the way that human view of an example relies upon not just the difference in an upgrade, (for example, sound, lighting, et al.) yet additionally the first power of the improvement. In particular, WLD comprises of two segments: its differential excitation and direction. A differential excitation is an element of the proportion between two terms: One is the relative power contrasts of its neighbors against a present pixel; the other is the force of the present pixel. A direction is the angle direction of the present pixel. For a given picture, we utilize the differential excitation and the direction parts to develop a linked WLD histogram highlight. Trial results on Brodatz surfaces demonstrate that WLD stunningly beats the other old style descriptors (e.g., Gabor). Particularly, trial results on face recognition demonstrate a promising presentation. In spite of the fact that we train just a single classifier dependent on WLD highlights, the classifier gets a practically identical presentation to best in class strategies on MIT+CMU frontal face test set, AR face dataset and CMU profile test set.

[4] Christoudias.C.M, Georgescu. B.andMeer. P. "Synergism in low-level vision" in 16th I.E.E.E conference, Pattern

Recognition, vol.4, pp 150-155, IEEE, New York City, New York.

Managing picture division with edge data is a frequently utilized methodology in low level PC vision. To improve the exchange off between the affectability of homogeneous locale depiction and the over division of the picture, we have joined an as of late proposed edge extent/certainty map into a shading picture divided dependent on the mean move method. The new technique can recoup locales with feeble however sharp limits and along these lines can give a progressively exact contribution to abnormal state elucidation modules. The Edge Detection and Image Segmentation (EDISON) framework, accessible for download, actualizes the proposed strategy and gives a total tool kit to irregularity safeguarding separating, division and edge location.

### III. SOFT ROUGH FUZZY C-MEANS ALGORITHM (SRFCM)

SRFCM has its roots inside the k-manner set of rules proposed with the aid of J Mc Queen. Fuzzy C-Means (FCM) Algorithm became proposed via Bezdek. In FCM, items are not constrained to belong to a single cluster. Each object belongs to all clusters with certain diploma of belongingness. Rough ok-approach (RKM) become proposed by using Lingras and West [9] by means of borrowing a number of the concepts of difficult set theory [13]. Rough Fuzzy c-approach set of rules became proposed by way of Mitra et al., [11]. In this paper SRFCM is proposed with the aid of applying similarity standards of gentle sets to Rough Fuzzy Frame work.

The basic steps for SRFCM are as,

1. Assume the random clusters 'm' denoted by  $C_i$
2. Find membership between cluster center and data points.
3. Allocate all the data point to the lower approximation and upper approximation.
4. Difference between membership and next membership of data points are calculated.
5. Calculate similarity index of sample point.
6. Compute the updated cluster prototype for every cluster.
7. Iterate and run steps 2 to 6.

### IV. MULTI CLASS-SVM (SUPPORT VECTOR MACHINE)

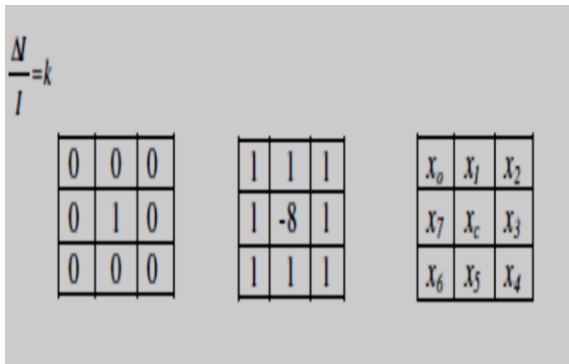
In artificial intelligence, support-vector machines (SVMs, additionally support-vector systems [1]) are managed learning models with related learning calculations that examine information utilized for grouping and relapse investigation. Given a lot of preparing models, each set apart as having a place with either of two classifications, a SVM preparing calculation fabricates a model that allots new guides to one class or the other, making it a non-probabilistic twofold straight classifier (in spite of the fact that techniques, for example, Platt scaling exist to utilize SVM in a probabilistic arrangement setting). A SVM model is a portrayal of the models as focuses in space, mapped with the goal that the instances of the different classifications are partitioned by a reasonable hole that is as wide as could reasonably be expected. New models are then mapped into that equivalent space and anticipated to have a place with a classification dependent on the side of the hole on which they fall.

In expansion to performing direct arrangement, SVMs can productively play out a non-straight characterization utilizing what is known as the bit stunt, verifiably mapping their contributions to high-dimensional element spaces. At the point when information are unlabeled, managed learning is preposterous, and an unaided learning approach is required, which endeavors to discover normal bunching of the information to gatherings, and after that guide new information to these shaped gatherings. The help vector clustering [2] calculation, made by Hava Siegelmann and Vladimir Vapnik, applies the insights of help vectors, created in the help vector machines calculation, to classify unlabeled information, and is one of the most broadly utilized grouping calculations in mechanical applications.

#### *Power Law Descriptor*

In insights, a power law is a useful connection between two amounts, where a relative change in one amount brings about a corresponding relative change in the other amount, free of the underlying size of those amounts: one amount shifts as an intensity of another. For example, thinking about the region of a square regarding the length of its side, if the length is multiplied, the region is increased by a factor of four.

Ernst Weber observed that the ratio of incremental threshold to the background intensity is a constant [1]. This relation known since as weber's law can be expressed as:



Where,  $\Delta I$  represents the increment threshold (just noticeable difference for discrimination);  $I$  represents the initial stimulus intensity and  $k$  signifies that the proportion on left side of the equation remains constant despite variations in the  $I$  term. The fraction  $\Delta I / I$  are known as the Weber fraction. Weber’s law more simply stated says that the size of a just noticeable difference is a constant proportion of the original stimulus value.

V. PROPOSED METHOD

5.1 Algorithm Steps

1. Texture, color as well as spatial features cues re remove from the image. Local binary pattern for texture features as well as Homogeneity model is used to the extract color features. Further, relating to information is embedded in the features make legally null the effect of outliers and noise.
2. Standard Cubic Feet per Minute (SCFPM) are based to clustering is applied on the features space because for selecting the training samples the first step which are applied to the Multi class SVM classifier and the other next stage of segmentation.
3. In the Multi Class SVM training once again all the Multi Class SVM classifier is trained using the samples obtained from the on step. For the image pixels in  $j$ th cluster of some pixels are most appropriate as training samples and left are used as test samples.
4. The classification of Multi SVM pixels are applied to the test set of SVM for classifying the new data. As combine the test set as well as training to obtain the result of final segmentation.

5.2 Colour Feature Calculation

All the pixels within the photo are marked as homogenous place bearing on an image. The photograph segmentation undertaking is now a class hassle and the system of segmentation is aimed toward assigning a label to every individual pixel or entire vicinity primarily based on

homogeneity. Color functions are extracted from the Lab color model, due to the fact color distinction can be measured effectively in LAB coloration space.

5.3 Texture Feature Extraction by Power Law Descriptor

Power law descriptor is robust for feature extraction as it irrespective of illumination changes.

Textural features are extracted as given below,

1. Convert image to gray scale.
2. Find the differential excitations for given gray scale.
3. Calculate orientations for given gray scale image.
4. Find 2D pixels histogram of given differential excitations.
5. Final resultant histogram at given point  $(i,j)$  forms clusters.

VI. RESULT ANALYSIS

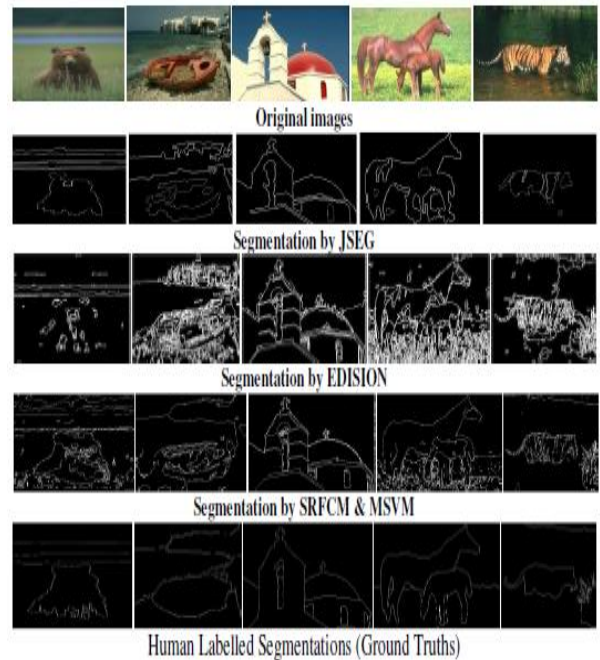


Fig.6.1 Results of Proposed Work and Ground Truth Images

Performance parameters are calculated are as below,

### 1. *Random Index(RI)*

Random index (RI) is the performance parameter which can be calculated between final segmented image and ground truth image.

### 2. *Variation of Information(VOI)*

Variation between computed image by proposed work and ground truth image is called variation of information.

**Table I : RI and VOM**

Image	RI			VOI		
	MSVM	JSEG	EDI	MSVM	JSEG	EDI
Bear	0.68	0.61	0.68	3.42	2.09	2.55
Boat	0.54	0.45	0.46	3.63	3.64	5.61
Church	0.72	0.45	0.67	2.70	3.03	3.06
Horse	0.60	0.45	0.46	3.30	3.34	5.33
Tiger	0.63	0.47	0.54	2.60	2.63	4.15

## VII. CONCLUSION

We successfully developed a robust technique for color image segmentation which is based on the integration of following techniques such as fuzzy sets, rough sets and soft sets. For final segmentation with better accuracy the results obtained from hybridization of the above techniques are fed to multi class SVM (Support Vector Machine) classifier which is machine learning tool for supervised classification. This implementation is applied to lot of images from the Berkeley Database which includes 500 images with ground truth images for original images. Proposed work is more efficient compared to state of art techniques which is proved by objective analysis. The results obtained from the soft fuzzy rough means clustering with SVM classifier, inter cluster distance is increased and intra class distance is minimized. Different performance metric shows the efficiency of the proposed work. Proposed algorithm is developed mostly for improving the quality as well as accuracy of clustering algorithm. This algorithm is also used for images with noise or noisy images.

## VIII. REFERENCES

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