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AN APPROACH FOR THE CLUSTERING AND CLASSIFICATION OF IMAGES USING ABC-GA AND NAIVE BAYES CLASSIFIER

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ABSTRACT

This paper presents a hybrid clustering algorithm and Naive Bayes classifier for trees, shade, building and road. It starts with the single step preprocessing procedure to make the image suitable for clustering. The pre-processed image is clustered using the ABC-GA algorithm that is developed by hybridizing the ABC and genetic algorithm to obtain the effective clustering in satellite image and classified using Naive Bayes classifier. Classification accuracy of hybrid algorithm is found better than ABC.

Keywords: Clustering, classification, features extraction, Naïve Bayes classifier, ABC-GA algorithm.

I. INTRODUCTION

The main objective of image segmentation is to partition the image into parts of strong correlation with objects or areas of the real world image

At present , in the Literature, an extensive variety of satellite image categorization methods are: Cluster, Statistic, Bayesian Net, Artificial Neural Networks (ANN), etc [4]. By visual understanding of satellite imagery and aerial photography, now, operators can manually remove cartographic features, such as buildings, roads, and trees.

frequently consigned to as ‘‘soft’’ classifiers, as opposed to the non-fuzzy ‘‘hard’’ classifiers, with their applicability being of still larger significance under the survival of varied pixels.

II. REVIEW OF LITERATURE

In the image processing domain, Satellite image categorization is an extremely taxing task. In order to progress precision, researchers have applied different kinds of classification techniques. According to the database of skilled knowledge for a more focused satellite image classification, a hybrid biologically inspired method was adapted that was presented by Lavika Goel [10].

As a competent land cover classifier for satellite image, a hybrid FPAB/BBO based algorithm has been presented by Navdeep Kaur Johal *et al.* [11].

A FPAB/BFO based algorithm for the categorization of satellite image has been presented by Parminder Singh *et al.* [12]. To choose the optimal and minimum set of fuzzy rules, the use of a genetic algorithm (GA) has been presented by O. Gordo *et al*. [13] and to categorize remotely sensed images using a fuzzy classifier.

A generalized multiple-kernel fuzzy C-means clustering (MKFCM) methodology for satellite image segmentation has been presented by M. Ganesh and V. Palanisamy [14]. Aissam Bekkari *et al*. [15] have progressed a methodology using a combination kernel that effortlessly combined multi-spectral features, Haralick consistency characteristics and Hybrid Median Filter, with dissimilar window sizes. The result demonstrated that the common use of spectral and texture data as one considerably enhanced the precision of satellite image categorization.

This paper, proposes a significant satellite image classification technique using ABC-GA clustering algorithm and Naïve Bayes classifier. The proposed approach consists of three steps

i)Pre-processing
ii) Clustering ABC-GA algorithm
iii) classification using Naïve Bayes Classifier.

Initially pre-processing is performed to make the image suitable for segmentation. In segmentation, the pre-processed image is segmented using hybrid ABC-

GA algorithm that is developed by hybridizing the ABC algorithm [16] and genetic algorithm to obtain the effective segmentation in satellite images. Then, feature is extracted and the classification of satellite image into four different labels (tree, shade, road and building) is done using feed-forward neural network classifier.

III. Hybrid Clustering Algorithm and Naïve Bayes Cassifier for Satellite Image Classification

3.1 Segmentation using Genetic-ABC Algorithm

In this work different colors of trees, roads , shades and buildings regions are taken in the database. Extract H,S,L,T,U,V layers. For each layer calculate features like mean and maximum value of histogram. Train these features using Naïve Bayes classifier.

During testing phase , a test image is given. For this apply the ABC-GA clustering algorithm.

Step1: Generate the new solutions using Eq.(3)

Step 2: *Fitness Computation*

The fitness of each chromosome in the solution is evaluated with objective function presented by Eq(1). The objective function of the clustering process taken for this paper is intra cluster distance that is computed by the following equation

$$J_1(Y)=\frac{1}{\sum_{i=1}^{N}\min\left\{\left\|X_i-X_j\right\|\right\}};\ j=1,2,...,k \quad \text{----} \quad (1)$$

Where, $\left\|X_i-X_j\right\|$ is a chosen distance measure between a data point X_i and the cluster centre X_j, 'N' and 'k' are the number of input data and the number of cluster centres, respectively. The finding of objective function of clustering process is illustrated in Fig(4).

Step 3*: Employed bee phase*

After initialization, in the employed bees' phase, each employed bee is sent to the food source in its memory and finds a neighboring food source. Here, 20% chromosomes are randomly generated for every new chromosome generation process. Subsequently, we have determined best fitness from the employed bee phase.

Step 4*: Onlooker bee phase*

In the onlooker bees' phase, the onlookers receive the information of the food sources shared by employed bees. Then each chromosome has chosen a solution to exploit depending on a probability related to the nectar amount of the solution (fitness values of the solution). That is to say, there may be more than one onlooker bees choosing a same solution if the source has a higher fitness. The probability P_i is calculated according to Eq(2) as followed:

$$P_i=\left[\frac{0.25}{Max\ fitness}\right]*fit+1 \quad (2)$$

After solutions have been chosen, each onlooker bee finds a new solution in its neighbourhood following Eq(3),

$$x_{i,j}=x_j^{\min}+rand(0,1)(x_j^{\max}-x_j^{\min}) \quad (3)$$

Step 5*: Mutation*

After onlooker bee phase, some selected chromosomes are considered for mutation. Mutation consists of changing the value of a random bit, which is randomly chosen in the chosen vector.

Step 6*: Scout bee phase*

In scout bees' phase, if the value of trials counter of a solution is greater than a parameter, known as 'limit', the solution is abandoned and the bee becomes a scout bee. A new food solution has produced randomly in the search space using Eq(3), as in the case of initialization phase. The employed, onlooker, mutation and scout bees' phases will recycle until the termination condition is met. The best food source which presents the best solution.

3.3 Classification using Naïve Bayes Classifier

Classification step is to identify road, building, tree and shadow regions from original satellite image.

3.4 Evaluation Metrics

The evaluation of proposed technique in different satellite images are carried out using the following metrics as suggested by below equation.

$$\text{Accuracy} = \frac{\text{number of true positives} + \text{number of true negatives}}{\text{number of true positives} + \text{false negatives} + \text{true negatives} + \text{false positives}}$$

Experimental Results

The proposed technique is designed for identify the four different labels (tree, shade, road and building) of satellite images. The obtained experimental results from the proposed technique are given in figure 7 (a)-(e).These figures shows the extracted regions (tree, shade, road and building) from the Input satellite images using Naïve Bayes Classifier with kernel – normal.

Fig.7(a) Input image

Fig.7(b) Road

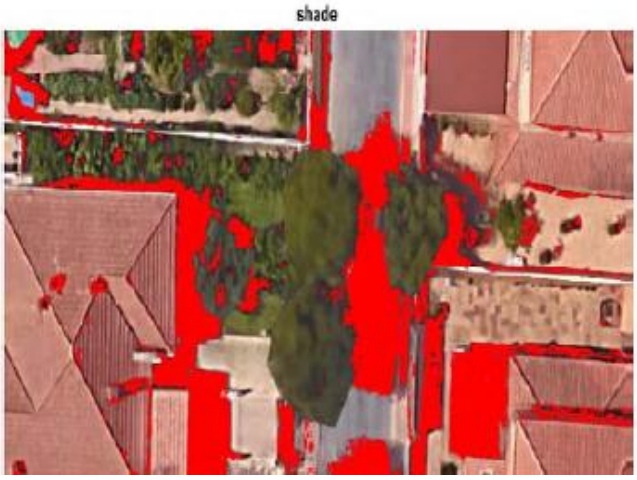

Fig.**7(C)** Shade

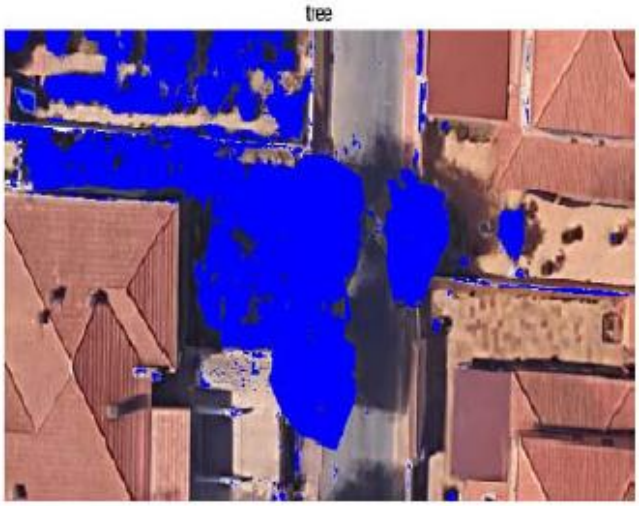

Fig.7(d) Tree

Fig.7(e) Building

Comparative Analysis

In this paper, the proposed algorithm results are compared with against ABC. The performance of this is analyzed with parameter accuracy for five images. The accuracy value is computed by dividing the total number of similar pixels identified as land use to the number of pixels in the tree, shade, building and road region. The detailed results obtained for proposed technique is shown in Table(1). As shown in Table(1) the proposed technique is achieved the maximum accuracy value compared to ABC algorithm for all images Hence, it is clear from result analysis that the proposed algorithm shows better results as compared with existing techniques.

Image index	Techniques	Tree	Shades	Roads	Building
I_1	ABC-GA	**0.81**	**0.6834**	**0.779**	**0.79975**
	ABC	0.77	0.6502	0.765	0.74619
I2	ABC-GA	**0.79**	**0.557**	**0.767**	**0.70625**
	ABC	0.78	0.650	0.7	0.68436
I3	ABC-GA	**0.79**	**0.719**	**0.884**	**0.74581**
	ABC	0.78	0.714	0.803	0.73980
I4	ABC-GA	**0.85**	**0.815**	**0.939**	**0.90891**
	ABC	0.80	0.765	0.866	0.86733
I5	ABC-GA	**0.82**	**0.857**	**0.911**	**0.95346**
	ABC	0.80	0.82	0.861	0.92116

Fig(10): Comparison of Accuracy

IV. CONCLUSION

In this paper, optimization algorithms for segmentation with the intention of improving the segmentation in satellite images using Naïve Bayes classifier is proposed. genetic-ABC algorithm, and iii) classification using feed-forward neural network classifier. In segmentation, the image is clustered using hybrid ABC algorithm that is developed by hybridizing the ABC algorithm and genetic algorithm to obtain the effective segmentation in satellite images. Then, feature is extracted and the classification of satellite image into four different labels (tree, shade, road and building) is done using feed-forward neural network classifier. Finally, classification accuracy of the proposed algorithm in satellite image classification is calculated and the performance is compared with ABC.

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