COMPARATIVE STUDY OF VARIOUS SUPERVISEDCLASSIFICATION METHODSFORANALYSING **DEFORESTATION FACTORS**

S.P.Rajagopalan¹, C.Lalitha²

¹Professor, G.K.M College of Engineering & Technology, Chennai, TamilNadu ²*Research Scholar, Vels University, Chennai, TamilNadu*

Abstract

In this paper, various supervised classification techniques are compared and the results are demonstrated. Here the classification techniques like Decision tree, Bayesian method, Neural Networks and Rule Based method are discussed with regards to the data sets given. The population, built up and agriculture are the factors that play a vital role in the development of country which directly affects economic condition. In this paper, factors such as road, population, built up development, agriculture and industry are considered as drivers of deforestation in the study area which is located in the Erode District of TamilNadu, India.

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Keywords: Supervised Classification, Decision tree, Bayesian method, Neural Network.

1. INTRODUCTION

There are two types of classification, supervised and unsupervised. Supervised methods classify the data which is known and observed by the user specifically. Unsupervised methods are classified unknowably. The results are obtained with the given data sets by using the WEKA. The aim of this paper is to compare the various supervised methods by using the factors such as demographic, built up, road and agriculture. The primary methods used in Data mining are Data selection, data reduction and filtration. Data mining examines and discovers various algorithms under several computational efficiency. It integrates machine learning, pattern recognition, statistics, databases, and visualization techniques into one so that the information can be extracted from the large databases.

The tasks of data mining are association rules mining, classification, prediction and cluster analysis. Generally speaking, association rule mining and classification rule mining are the most effective and efficient techniques in data mining. Classification rule mining is used for the prediction future objects whose class label is not known. Recently it has been determined that primary factor for the degradation of ecosystem is deforestation of forests.

Classification results are basis for interpretation, analysis and modelling for various environmental and socioeconomic applications. Data mining techniques can be applied for generating the class association rules for analysing the deforestation. In this paper, we applied various supervised classification techniques with our data sets.

2. WEKA

The Waikato Environment for Knowledge Analysis (WEKA) is a tool for machine learning algorithms which can be used for classification and clustering. In this paper, we use decision tree methods like J48 and Random forest. Bayes algorithm such as Naive Bayes and Neural Network methods like Multilayer perceptron are implemented in WEKA. We divided our data set into 10 cross validation folds and all the methods are tested and compared according to the data.

3. CLASSIFICATION METHODS

3.1 Decision Tree

Decision Tree (DT) worksaccording to the processing and deciding upon attributable data. Here attributes in DT are considered as nodes and each leaf node as a class. J48 and Random Forest were used in our experiments. It follows a recursivemethod for a given set of data.It searches the attributes as Depth-first strategy. It divides the class into several nodes and tests each node that gives the best result.It classifies the datasets invariably. This method is not suitable for finding anddid not show good results to the given datasets. Accuracy value of J48 methods is compared and shown in Table 1 and Figure 1.

J48 CLASSIFICATION METHOD											
	1990		2000		2010			14	18 CLASS	IFICATION	
Forests	CCI	MAE	CCI	MAE	CCI	MAE	150				Populatio
Population	83.95	9E-04	96.29	2E-04	74.07	0.002	100				
crop	79.01	0.002	97.28	2E-04	80.24	0.002	0				crop
Built up	78.02	0.002	41.48	0.006	45.18	0.005	Ŭ	CCI MAE	CCI MAE	CCI MAE	🛚 Built up
Industry	94.32	0.012	84.44	0.01	78.76	0.011		1000	2000	2010	Industry
Road	69.87	0.002	92.83	4E-04	78.51	0.001		1990	2000	2010	

Random forest is also called regression trees that induces the data from bootstrap samples of the training data. It uses random feature selection by induction process. It is comparatively gives better result than CART and C4.5.

It shows a better performance, after modelling the result. The disadvantages of DT are focus on continues attributes, computational efficiently with growing tree size. According to comparison provided for different classification methods in emotion recognition, Random Forest is the best classifier method on that group with 5 attributes and the results are compared and shown in Table 2 and Figure 2.

Base Function (RBF)were used in this work. MLP is a feed forward network that makes a model to map input data to

output data. Hidden layer in MLP can include various layers between input and output. It classifies 3 factors gradually.

The accuracy result of MLP is shown in Table 3 and Figure



3.2 Artificial Neural Networks

3.2.1 MLP

Artificial Neural Network (ANN) is the common classification methods in data mining. Neural Network based classifiers, Multi Layer Perceptron (MLP) and Radial

AYER PE	RCEPTRO	N CLAS	SIFICATIO	25000									
19	1990 2000			2010		20000							Populati
MAE	C.COEFF	MAE	C.COEFF	MAE	C.COEFF	10000			-				crop
22347	0.228	2864.9	0.3141	11191	0.2381	5000							Built up
2417	0.292	15707	0.2553	21556	0.1219	Ū	Ч	쁖	ЧЧ	쁥	Ч	Ë	a buit up
68.017	0.1542	113.16	0.379	145.75	0.3932		Σ	B	Σ	B	Σ	Ö	Industry
2.21	0.1388	3.187	0.1746	8.3707	0.0617			Ŭ		ŭ		Ŭ	Road
68.017	0.1542	355.67	0.1538	349.82	0.1307		19	90	20	00	20	010	
	AYER PE 19 MAE 22347 2417 68.017 2.21 68.017	AYER PERCEPTRO 1990 MAE C.COEFF 22347 0.228 2417 0.292 68.017 0.1542 2.21 0.1388 68.017 0.1542	AYER PERCEPTRON CLASS 1990 20 MAE C.COEFF MAE 22347 0.228 2864.9 2417 0.292 15707 68.017 0.1542 113.16 2.21 0.1388 3.187 68.017 0.1542 355.67	AYER PERCEPTRON CLASSIFICATIO 1990 2000 MAE C.COEFF 22347 0.228 2864.9 2417 0.292 15707 0.2553 68.017 0.1542 113.16 0.379 2.21 0.1388 3.187 0.1746 68.017 0.1542 355.67 0.1538	AYER PERCEPTRON CLASSIFICATION METHING 200 200 MAE C.COEFF MAE C.COEFF MAE 22347 0.228 2864.9 0.3141 11191 2417 0.292 15707 0.2553 21556 68.017 0.1542 113.16 0.379 145.75 2.21 0.1388 3.187 0.1746 8.3707 68.017 0.1542 355.67 0.1538 349.82	AYER PERCEPTRON CLASSIFICATION METHOD 190 200 2010 MAE C.COEFF MAE C.COEFF 22347 0.228 2864.9 0.3141 11191 0.2381 2417 0.292 15707 0.2553 21556 0.1219 68.017 0.1542 113.16 0.379 145.75 0.3932 2.21 0.1388 3.187 0.1746 8.3707 0.0617 68.017 0.1542 355.67 0.1538 349.82 0.1307	AYER PERCEPTRON CLASSIFICATION METHOD 25000 2000 15000 15000 15000 15000 15000 15000 15000 15000 15000 15000 15000 15000 15000 15000 15000 10000 5000 10000 5000 0 10000 <td>AYER PERCEPTRON CLASSIFICATION METHOD 25000 20000 20000 10000</td> <td>AYER PERCEPTRON CLASSIFICATION METHOD 25000 2000 2010 20000 15000 15000 15000 15000 15000 10000 15000 10000 15000 100</td> <td>AYER PERCEPTRON CLASSIFICATION METHOD 25000 1990 2000 2010 2000 MAE C.COEFF MAE C.COEFF 22347 0.228 2864.9 0.3141 11191 0.2381 2417 0.292 15707 0.2553 21556 0.1219 68.017 0.1542 113.16 0.379 145.75 0.3932 2.21 0.1388 3.187 0.1746 8.3707 0.0617 68.017 0.1542 355.67 0.1538 349.82 0.1307 1990 2000</td> <td>AYER PERCEPTRON CLASSIFICATION METHOD 25000 25000 20000 15000 MAE C.COEFF MAE C.COEFF MAE C.COEFF MAE C.COEFF 22347 0.228 2864.9 0.3141 11191 0.2381 10000 5000 0 2417 0.292 15707 0.2553 21556 0.1219 5000 0 10000 0 10000 0 10000 0 10000 0 10000 0 10000 0 10000 0 10000 0 10000 0 10000 0 10000 0 10000 0 10000 0 0 10000 0 0 10000 0 0 10000 0</td> <td>AYER PERCEPTRON CLASSIFICATION METHOD 25000 1990 2000 2010 20000 MAE C.COEFF MAE C.COEFF MAE C.COEFF 22347 0.228 2864.9 0.3141 11191 0.2381 10000 2417 0.292 15707 0.2553 21556 0.1219 5000 0 68.017 0.1542 113.16 0.379 145.75 0.3932 145.75 0.3932 1990 2000 2000 68.017 0.1542 355.67 0.1538 349.82 0.1307 1990 2000 2000</td> <td>AYER PERCEPTRON CLASSIFICATION METHOD 25000 25000 20000 10000</td>	AYER PERCEPTRON CLASSIFICATION METHOD 25000 20000 20000 10000	AYER PERCEPTRON CLASSIFICATION METHOD 25000 2000 2010 20000 15000 15000 15000 15000 15000 10000 15000 10000 15000 100	AYER PERCEPTRON CLASSIFICATION METHOD 25000 1990 2000 2010 2000 MAE C.COEFF MAE C.COEFF 22347 0.228 2864.9 0.3141 11191 0.2381 2417 0.292 15707 0.2553 21556 0.1219 68.017 0.1542 113.16 0.379 145.75 0.3932 2.21 0.1388 3.187 0.1746 8.3707 0.0617 68.017 0.1542 355.67 0.1538 349.82 0.1307 1990 2000	AYER PERCEPTRON CLASSIFICATION METHOD 25000 25000 20000 15000 MAE C.COEFF MAE C.COEFF MAE C.COEFF MAE C.COEFF 22347 0.228 2864.9 0.3141 11191 0.2381 10000 5000 0 2417 0.292 15707 0.2553 21556 0.1219 5000 0 10000 0 10000 0 10000 0 10000 0 10000 0 10000 0 10000 0 10000 0 10000 0 10000 0 10000 0 10000 0 10000 0 0 10000 0 0 10000 0 0 10000 0	AYER PERCEPTRON CLASSIFICATION METHOD 25000 1990 2000 2010 20000 MAE C.COEFF MAE C.COEFF MAE C.COEFF 22347 0.228 2864.9 0.3141 11191 0.2381 10000 2417 0.292 15707 0.2553 21556 0.1219 5000 0 68.017 0.1542 113.16 0.379 145.75 0.3932 145.75 0.3932 1990 2000 2000 68.017 0.1542 355.67 0.1538 349.82 0.1307 1990 2000 2000	AYER PERCEPTRON CLASSIFICATION METHOD 25000 25000 20000 10000

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3.2.2RBF

RBF is another type on ANN. The input is linear and the output is nonlinear. Here it hides several non sequential values because of random input. At first it shows a good

result for the year 1990 and 2010, but not for 2000.So that this algorithm is not suitable for the given datasets. The RBF networks are divided in two feed-forward layer is shown in Table 4 and Figure 4.

NOP CLASSIFICATION METHOD														
	19	90	2000		2010									Population
Forest	CCI	MAE	CCI	MAE	CCI	MAE	100					-1 m		E crop
Population	83.95	9E-04	96.29	2E-04	74.07	0.002	50							
crop	79.01	0.002	97.28	2E-04	80.24	0.002	0							Built up
Built up	78.02	0.002	41.48	0.006	45.18	0.005		8	AE	8	AE	8	μ	Industry
industry	94.32	0.012	84.44	0.01	78.76	0.011			≥ ≥		≥		2	Road
Road	69.87	0.002	92.83	4E-04	78.51	0.001		19	990	20	00 20		10	

3.3 Bayesian Methods

Bayesian methods are one of the solution for the classification methods in data mining. In our work, Naive Bayesis implemented for classification. It follows an independent feature model with strong independence assumptions. This method is applicable for the statistical data.

Classification is done by appropriately to the attributes Cropand built up. It classifies and shows the better performance to 10% for all the years. This method shows good results for two attributes and is displayed in Table 5 and Figure 5.



3.4 Rule Based Classification

Decision trees can be translated into a set of rules by creating a separate rule for each path from the root to a leaf in the tree. However, rules can also be directly induced from training data using a variety of rule-based algorithms. Classification accuracy of rule learning algorithms can be improved by combining features using the background knowledge of the user automatic feature construction algorithms.

3.4.1 1R Algorithm

One Rule is a simple accurate, classification algorithm that generates one rule for each predictor in the data and then selects the rule with the smallest total error as its "one rule". It is the simplest rule-based classification learning algorithm for discrete attributes. It shows a gradual increase in the years 1990 and 2000 for all the attributes but suddenly no changes for the year 2010 because OneR produces rules only slightly less accurate .The result of this method is shown in Table 6 and Figure 6.



3.4.2 Prism

Prism is a greedy algorithm that finds a minimum spanning tree for a connected weighted undirected graph. This means it finds a subset of the edges that forms a tree that includes every vertex, where the total weight of all the edges in the tree is minimized. It can also be used to find the minimum spanning forest and reflects the same result for all the three decades , is in Table 7 and Figure 7.

PR	120 -													
	1990		2000		2010		100 ·							Population
Forest	CCI	MAE	CCI	MAE	CCI	MAE	80 · 60 ·							. crop
Population	99.5	0	99.5	0	99.5	0	40 ·			-18		-		
crop	100	0	100	0	100	0	20							Built up
Built up	100	0	100	0	100	0		ō	щ	ō	W	ō	삝	Industry
Industry	100	0	100	0	100	0		0	Ξ		Ξ	0	ŝ	Road
Road	99.25	0	99.75	0	97.53	0		19	990	20	000	20	10	

4 COMPARATIVE RESULTS

The Machine learning techniques such as Naïve Bayes, Bayes network, J48, Random Forest, Multi Layer Perceptron (MLP) and Radial Base Function (RBF) ,one R, PRISM were used for simulation. Here we split our original dataset of 405 samples into 66% for training purpose and remaining 34% for testing purpose. Weka incorporates k-fold crossvalidation, in which the original sample is randomly partitioned into k subsamples. The cross-validation process is then repeated into several subsamples .Here ,we have used 10-fold cross validation .Kappa gives a numerical rating of -1 to 1 scale, where 1 is perfect agreement, 0 implies expected by chance, and negative values indicate agreement lesser than chance, Comparatively PRISM gives best result and shows the degradation of forest. The correctly classified Instances of various years are shown in Table8,9,10and the accuracy rates are displayed in Figure 8,9,10.





5. CONCLUSIONS AND FUTURE WORK

In this paper, four classifiers including Neural Network, Naïve Bayesian, Rule based, Decision tree were tested to determine the deforestation from the dataset of demographic factors. All the result were classified as 1,-1 or 0. There are many different mining and classification algorithms, and parameter settings in each algorithm. Experimental results in this paper are based on the default settings. Extensive experiments with different settings are applicable in WEKA. J48 is very simple classifier to make a decision tree, but it gives the invariable result in the experiment. Naïve Bayesian classifier also showed good result only for two attributes(Crop and Built up), but RBF classified another two attributes(Industry and Road) properly. Rule Based classifiers such as OneR and PRISM also showed good result than compared with J48 or Naïves Bayesian classifier. From this experiment, we can find that a simple Random Forest classifier can provide best classification result for deforestation(except one attribute).It is planned to incorporate other techniques like different ways of feature selection, classification using ontology.

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