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### **Comparison of Feature Selection Methods for Credit Risk Assessment**

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ABSTRACT- Fund is the greatest variable of the Banking Industry. In Banking Industry achievement and disappointment depends on the credit. Keeping money Industries are focused today with increment in volume, speed and assortment of new and existing information. Managing and analyzing the massive data is more difficult. The credit scoring databases are often large and characterized by redundant and irrelevant features. With this features, classification methods become more difficult. This difficulty can be solved by using feature selection methods. The main objective of the feature selection is to reduce the size of dimensions, costs and increase the classification accuracy. This research paper uses a filter feature selection model for finding the optimal feature subset to evaluate the credit risk. The filter model is implemented using WEKA tool. Comparison study is made to find the credit risk assessment.

*Key Terms* - *Classification, Data Mining, Credit Risk, Feature Selection, Filter*.

#### **1. INTRODUCTION**

Financial data analysis is used in many financial organizations for accurate analysis and assessment of the loan proposal. Credit risk is the greatest challenge for the Banking Industry. Granting credit approval depends on borrower's credit risk. Credit risk which encompasses the borrower's ability and willingness to pay and it is one of the main factors for defining a lenders credit policy [9]. The credit scoring databases are often large and characterized by redundant and irrelevant features. Managing and analyzing the massive data is more difficult. With this features, classification methods become more difficult. This difficulty can be solved by using feature selection methods. Datasets with high dimensional features have more complexity and spend larger computational time for classification [1]. Feature selection can be defined as a process that chooses a minimum subset of M features from the original set of N features, so that the feature space is optimally reduced according to a certain evaluation criterion [6]. Feature selection methods can be used to create



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an accurate predictive model and to identify and remove unneeded, irrelevant and redundant attributes from data. Feature selection methods are classified as three categories, such as Filter methods, Wrapper methods and embedded methods.

#### 2. LITERATURE REVIEW

Feature selection techniques improve the classification accuracy and develop a new method in combining the sample domain filtering, resampling and feature subset evaluation methods [1]. Many proposed feature selection methods have been compared and studied it is very difficult to find out the effectiveness of the feature selection methods, because data sets may include many challenges such as the huge number of irrelevant and redundant features, noisy data, and high dimensionality in term of features or samples [2]. Feature extraction carries out a transformation of the original features while variable selection selects a subset from the original features. Variable selection reduce the dimension of data without transforming data into a new set [3]. Janecek[4] showed the relationship between feature selection and data classification and the impact of applying Principle Component Analysis (PCA) on the classification process. Feature selections methods are used to improve the classification accuracy and reduce the subset of data analysis. It can be classified into two major groups such as, filter and wrapper [5].

#### **3. MODEL and RESULT**

This research work focus on reducing the feature set using the filter feature selection model. This research work uses feature selection techniques such as Relief, correlation based, entropy filter methods to evaluate the feature set. Managing and analyzing the massive data is more difficult. Financial industries comprise on huge amount of data. It is very difficult to handle the dimensionality data which uses huge number of features. In this study, we have analyzed the classification of credit risk data set using filter feature selection techniques. The bank dataset was taken from UC Irvine Machine Learning Repository [7]. It comprises of 21 attributes and 1000 instances. The feature selection methods used are Relief, Entropy and Correlation Feature selection (CFS). The three classification algorithms Decision Tree, Logistic Regression and Support Vector Machine are used to find the accuracy in the optimal feature subset. The WEKA software is used for analysis and evaluation. The metrics used for analysis are accuracy, recall and precision. Table 4.1 shows the result for three algorithms without using feature selection. Among the three algorithms the highest classification percentage is found in SVM.



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Classification	Classification	Precision	Recall
percentage	accuracy		
72.4	0.724	0.715	0.724
75.2	0.752	0.749	0.752
73	0.73	0.727	0.73
	percentage 72.4	percentage      accuracy        72.4      0.724        75.2      0.752	percentage      accuracy        72.4      0.724      0.715        75.2      0.752      0.749

#### Table 4.1. Results for Classification Accuracy without Feature Selection

The table 4.2 shows the classification accuracy for three algorithms using filter feature selection methods that is, entropy, relief and CFS.

sele		Feat ure	Entropy	Relief	CFS
	Classifier				
	%	1	70.7	70.6	71.6
C4	Accuracy	(	0.7	0.706	0.716
.5	Precision	(	0.689	0.701	0.688
	Recall	(	0.7	0.74	0.726
S V M	%	1	76.4	74.4	72.2
	Accuracy	(	0.764	0.744	0.722
	Precision	(	0.757	0.738	0.71
	Recall	(	0.764	0.744	0.722
L R	%	1	75.2	74.8	74.8
	Accuracy	(	0.752	0.748	0.748
	Precision	(	0.746	0.743	0.743
	Recall	(	0.752	0.748	0.748

The table 4.2 shows the classification accuracy for three algorithms using filter feature selection methods that is, entropy, relief and CFS.

Table 4.2 Results for Classification Accuracy with Filter Feature Selection



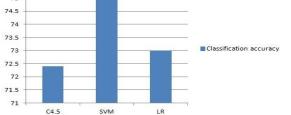
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The result shows that the highest accuracy is found in entropy filter based SVM.

Fig 4.1 shows the comparison of three algorithms. The graph is drawn with the classification algorithms on x-axis and percentage of classification accuracy on y-axis.



## Fig 4.1 Classification accuracy for three algorithms (without using feature selection)

Fig 4.2 shows the comparison of three algorithms with using filter feature selection techniques such as, entropy, CFS and relief. The graph is drawn with the feature selection techniques on x-axis and percentage of classification accuracy on y-axis.

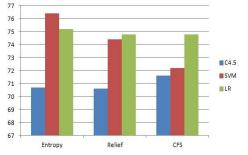


Fig 4.2 Comparison of filter feature selection techniques.

In Fig 4.2 the experimental results shows that the percentage of classification accuracy for bank

credit data is high when using feature selection methods as compared to fig 4.1 where the classification accuracy is low by using without feature selection methods.

3<sup>rd</sup> February 2017

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#### 4. CONCLUSION

The comparison study was made by using three classification algorithms such as C4.5, SVM and LR using with feature selection and without feature selection methods for credit risk assessment. The credit scoring databases are often large and characterized by redundant and irrelevant features. Thus the comparison study proves that classification accuracy is highest for support vector machine algorithm compared to C4.5 and LR. Thus the filter model helps banking Industry to make decision whether the loan can be approved or rejected for the new potential customer.

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