



Software Fault Detection using the Data Pre-Processing and Support Vector Machine

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Abstract-With increases use of high speed internet huge organization like banking, hospital and industrial work are done using the software rather than manual work. So this organization are required high quality software. If failure happen in the system is effect the financial cost of the organization. Software fault detection model is used for to identify the fault in software which cause the failure. In this paper fault detection model is proposed which used data pre-processing and support vector machine classifier. In data pre-processing feature ranking is perform using information gain and feature cluster is form using fuzzy c-means clustering.

Keywords: Data mining, data pre-processing, fuzzy c-means, software fault detection, SVM.

I INTRODUCTION

The growing use of the software in daily life is need to provide the high quality software at low cost. Software fault detection is important activity for providing the high quality software. Because of the software fault detection model decrease the effort cost for maintenance of the software. Because of the software fault detection model used limited testing resources at the testing phase of the software development life cycle.

Software fault detection is done by classifying the faulty or non-faulty module. Software metrics are used for identify the fault or non-fault module of the software. Metrics are compute form the source code of the software. This metrics value is called feature of the software. Data mining techniques like Naïve Bayes, Neural Network, Support Vector Machine and genetic programming are used to identify the faulty module. Data pre-processing is increased the result of the supervised learner. Public NASA datasets are used for software defect detection which is collect from the PROMISE repository.

The objective of the paper is to design the software fault detection model which includes data pre-processing and support vector machine. In this paper includes the methodology, proposed framework with results and comparison with another techniques.

II RELATED WORK

The quality of dataset is improved by data pre-processing, which incorporates feature selection and sampling which reduce instances. In feature selection is the method of distinctive and removing irrelevant and duplicate features from a dataset in order that which increase the performance of classification model. [1]-[2]

Wangshu Liu et al. used data pre-processing approach which consists the two stage used for identify the fault in software. Compare result using the Naïve Bayes, C4.5, IB1. [3]

Rohit Mahajan et al. proposed the framework for software fault prediction using the Bayesian Regularization (BR) and compare with Levenberg-Marquardt (LM) and Back propagation (BPA). Bayesian regularization give better performance. [4]

Gupta, Deepika, Vivek K. Goyal, and Harish Mittal Proposed Estimating of Software Quality with Clustering Techniques. This paper focus on clustering with very large dataset and very many attribute of different types. Effective result can be produced by using fuzzy c-mean clustering. [5]

Arashdeep Kaur et al. Proposed model for software fault prediction. In this paper, investigate the fuzzy c mean and k-mean performance. Fuzzy c mean is better than k- mean for requirement and combination metric model. Also investigate the metrics used in early life cycle can be used to predict fault module or not. [6]

Wangshu Liu et al. use the clustering based feature selection method for software fault prediction. FF-relevance and FF- Correlation Measure use. Heuristics approach use for cluster formation. [7]

Issam H. Laradji et al. use the greedy forward feature selection and Average Probability Ensemble learning model is to classify data. This model contain seven algorithm such as W-SVM, Random Forest etc. [8]

III METHODOLOGY

3.1. Data Pre-Processing

In data pre-processing step feature ranking is performed using information gain. The feature ranking ranks the feature based on the information that provides. Correlation between the feature and class measure to rank the feature. For measuring the correlation information gain (IG) is used. Information gain measures the amount of information provided by the feature f , whether the instance is fault or non-fault. The formula used for measuring the information gain is: [3]

$$IG(f) = H(A) - H(A|B)$$

Where $H(A)$ computes the entropy of the discrete random variable A (i.e. class). Consider $p(a)$ denote prior probability of a value a of A then $H(A)$ is computed by formula:

$$H(A) = - \sum_a p(a) \log_2 p(a)$$

$H(A|B)$ computes the conditional entropy which quantifies the uncertainty of A given the observed variable B . Consider $p(a|b)$ denote posterior probability of a for value b . $H(A|B)$ is computed by formula:

$$H(A|B) = - \sum_b P(b) \sum_a p(a|b) \log_2 p(a|b)$$

3.2. Fuzzy C-Means Clustering

In the second step of the framework cluster will be generated using the fuzzy c-means clustering method. Fuzzy c-means clustering moves the data centre iteratively to the right location.

Fuzzy c-means (FCM) generates the cluster in which allows data to belong to more than one cluster. The minimum value of the objective function is used to generate the cluster. Formula for objective function is given below:

$$J_m = \sum_{i=1}^N \sum_{j=1}^C u_{ij}^m \|x_i - c_j\|^2, 1 \leq m < \infty$$

Where, m = real number greater than 1.

u_{ij} = membership degree of x_i in the cluster j .

x_i = i th at d -dimensional measured data.

c_j = d -dimension center of the cluster.

Iterative optimization of the objective function is carried out in FCM. During the iteration process cluster centers c_j and membership u_{ij} are updated by:

$$u_{ij} = \frac{1}{\sum_{k=1}^C \left(\frac{\|x_i - c_j\|}{\|x_i - c_k\|} \right)^{\frac{2}{m-1}}}$$

$$c_j = \frac{\sum_{i=1}^N u_{ij}^m x_i}{\sum_{i=1}^N u_{ij}^m}$$

This iteration process will stop when $\max_{ij} \{ |u_{ij}^{(K+1)} - u_{ij}^{(K)}| \} < \epsilon$.

Where, k are steps of iteration and ϵ is a termination value between 0 and 1.

3.3. Support Vector Machine

Support vector machine (SVM) is used as the classifier for the software defect detection. Support vector machine is a supervised learning algorithm. Linear function is used in SVM. So it works as binary classification. It classifies the software module into two classes: Fault and Non-Fault.

IV PROPOSED SOLUTION

Figure 1 shows the proposed framework. This framework has two steps: the first step is data pre-processing in which information gain and fuzzy c-means clustering are used. In the second step, classification will be performed using support vector machine. Data are taken from the PROMISE data repository.

In the proposed framework, features are ranked using information gain. After that, clustering of the features is done. For that, fuzzy c-means clustering is used. After that, clusters are given as input to the classification model. For the implementation of the proposed framework, MATLAB and WEKA 3.6 are used.

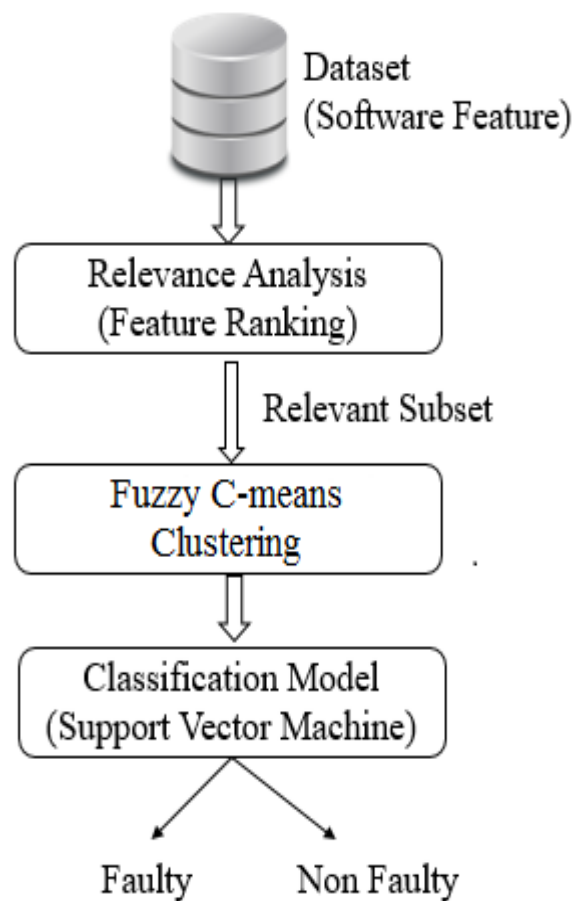


Figure 1. Proposed Framework

V EXPERIMENTAL RESULT AND ANALYSIS

The implementation of data mining algorithms requires the more powerful software tools. The number of available tools for data mining is grow therefor the choice of the suitable tools becomes difficult. We have used WEKA 3.6 and MATLAB for implementing proposed work.

5.1. Dataset

Features are extracted from the source code of the software. For software detection model collect the dataset from the promise data repository. Take CM1 dataset from the repository. Which contain the 21 features of the software and 498 instances. All the feature value is numeric. The basic feature avialble in dataset are:

- Line of code(Loc)
- Haslated complexity metrics
- Mccabe Complexity metrics
- Unique oprands
- Unique opreater
- Total operands
- Total operator

5.2. Analysis of the Result

First step of the proposed framework is to count the information gain and raking the feature. Figure 2 shows the information gain results for the CM1 dataset.

Second step of the proposed framework is to generate cluster of the feature using the Fuzzy C-means algorithms. In this step three cluster will generate which will show in figure 3.

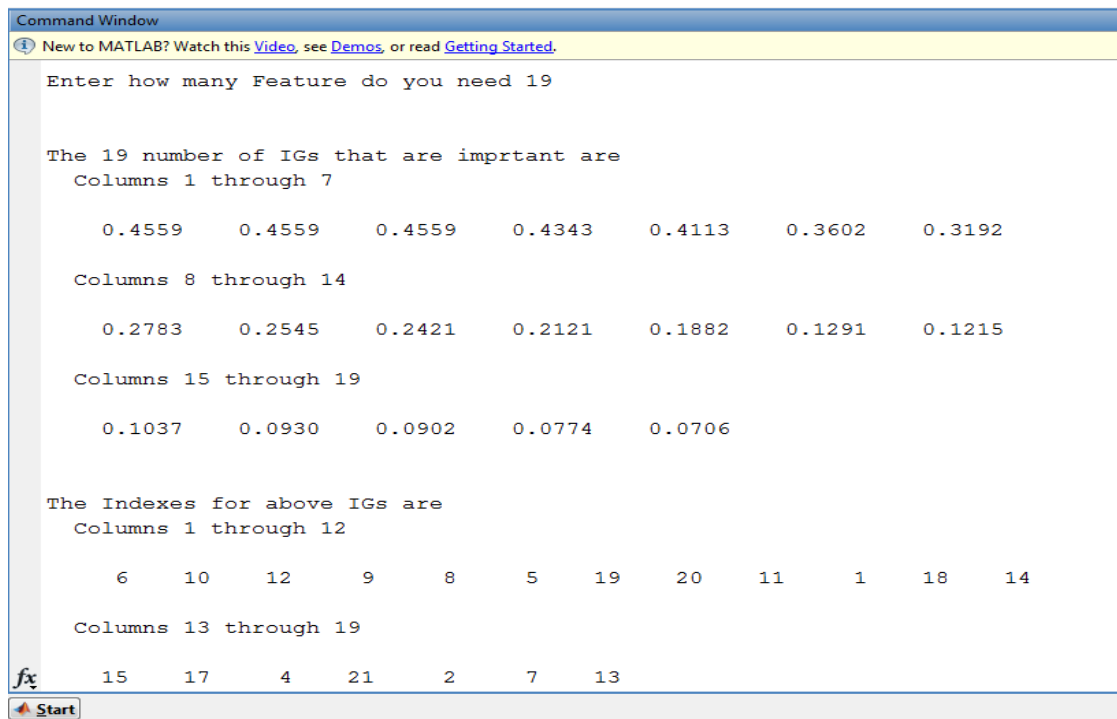


Figure 2 information gain for CM1 dataset

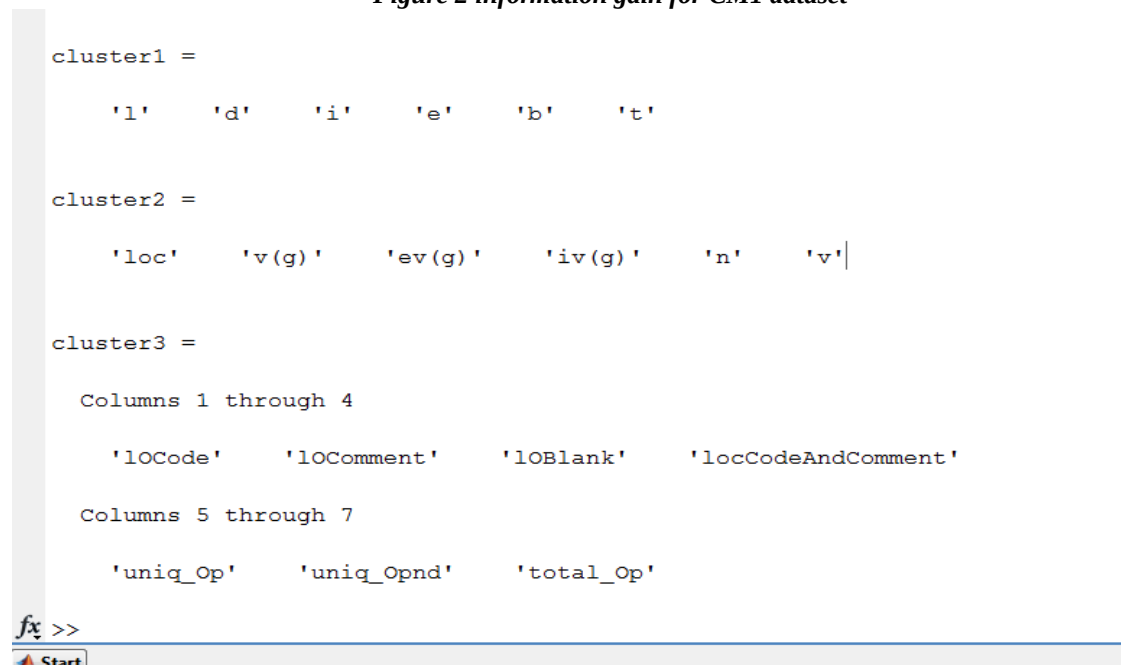


Figure 3 Cluster of features

In the third step classification model is used for the classifying the fault and non-fault modules. For that different classifier is applied directly to the CM1 dataset and compare that result. Here we consider the three classifier which are Neural Networks, Naïve Bayes and Support Vector Machine. Results of this Classifier are show in table 1.

Algorithm	Accuracy
Neural Network	68.7
Naïve Bayes	77.7778
Support Vector Machine(SVM)	81.82

Table 1. Accuracy of algorithms

Form this results identify that the support vector machine gives the better accuracy for software fault detection dataset. So we used the SVM as classifier in proposed framework. The results for the proposed framework is shown in table 2 and figure 4.

Performance Measure	Cluster1	Cluster2	Cluster3	Data
Accuracy	86.8687	84.8485	83.8384	81.8182
Precision	0.9425	0.987	0.9205	0.9737
Recall	0.9111	0.8444	0.9	0.8222

Table 2. Results of proposed framework

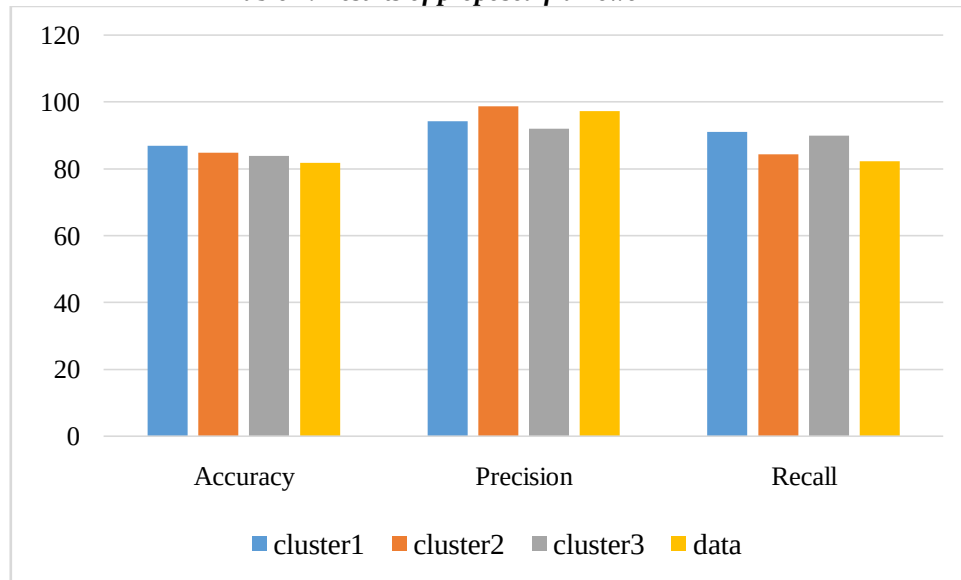


Figure 4. Results of proposed Framework

Figure 5 shows the comparison graph of the proposed framework and existing solution. For that accuracy of the proposed framework is better.

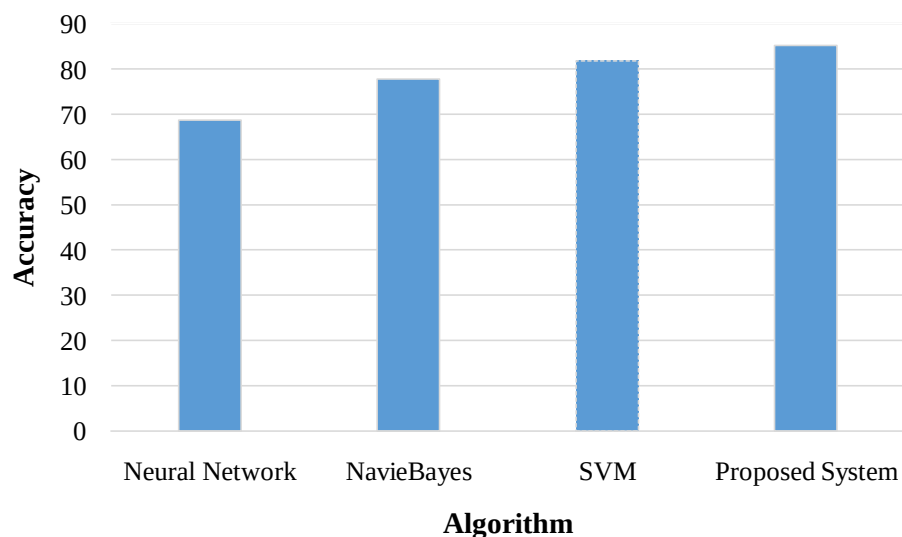


Figure 5. Comparison of Results

VI CONCLUSION

In this research paper we introduce the software fault detection model for identify the faulty modules in software. Also comparing the results of the proposed framework and exiting classifier and from that we can say that proposed framework gives the better results.

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