

BANKRUPTCY PREDICTION SYSTEM FOR CREDIT CARD USING MACHINE LEARNING TECHNIQUES : A SURVEY

Surabhi Girare¹, Vaibhav Patel², Anurag Shrivastava³

¹ M.Tech Scholar, CSE, NIIST, Bhopal

² Assistant Professor, CSE, NIIST, Bhopal

³ Associate Professor, CSE, NIIST, Bhopal

ABSTRACT- Prediction of Bankruptcy is the most important task. The techniques to assist fraud investigators, for efficient Credit Card Fraud Detection (CCFD), rely on machine learning algorithms. Proposing a predictive model for Credit Card Fraud determination is however mainly exigent due to the highly distributed and imbalanced data and the availability of only few transactions labeled as fraud in overall transactions. To seek out whether the transaction is fraud on E-commerce websites, is role of prediction models. To find out such transaction can be treated as a sort of machine learning (ML) problem. Many researches use machine learning techniques to improve performance of prediction and classification accuracy. This paper surveys fine classification methods that are helpful in building model for CCFD. Further in this paper, a predictive model for CCFD based on machine learning is proposed. New naïve bayes classifier is used for better accuracy. The dataset from UCSD-FICO Data Mining competition is used for building and testing the model. The results obtained shows that the predictive model has potential in determining fraud and minimizing the risk in e-commerce transactions. The paper directs about the future research in the field.

Keywords - Bankruptcy, Credit Card Fraud Detection, E-commerce, Machine Learning, Financial Organization, Classification

1. INTRODUCTION

The Credit Card is used as a mode of payment provided to the customer of bank or such financial organization. It allows buying goods or services to its holder. It is generally made up of Plastic with some secret numbers and Cardholder's Promise to pay for these goods and services availed [1, 2]. In Figure 1 'Clearing and Settlement under Credit Card System' is depicted. Bankruptcy is a situation in which a firm is incapable to resolve its monetary obligations leading to legal threat. The financial assets of companies are sold out to clear the

debt which results in huge financial losses to the investors. Bankruptcy results in decreased liquidity of capital and minimized financial improvement.[24]

Financial fraud is a growing concern with far reaching consequences in the government, corporate organizations, finance industry, In Today's world high dependency on internet technology has enjoyed increased credit card transactions but credit card fraud had also accelerated as online and offline transaction.

In this paper, different machine learning techniques are employed to predict bankruptcy. The support system can be utilized by stock holders and investors to predict the performance of a company based on the nature of risk associated.

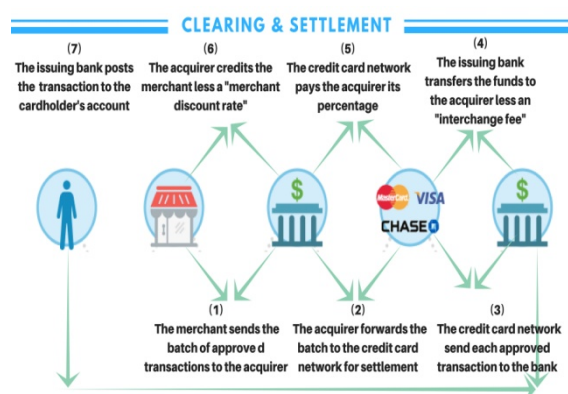


Figure 1: Clearing and Settlement under Credit Card System

2. LITERATURE REVIEW

Machine Learning (ML) through Ensembles is an important technique that combines outputs from multiple individual classifiers for improving classification accuracy [19, 20].

Theoretical and experimental results suggest that combining classifiers can give effective improvement in accuracy if classifiers within an ensemble are not correlated with each other [21, 22].

3. PROPOSED WORK

The framework proposed in this work is depicted in Figure 2. The proposed framework for prediction works for each transaction and separates the transaction with high or low risk using the method proposed. The proposed predictive model can be further used to generate alerts for transaction with high risks. Investigators check these alerts and provide a feedback for each alert, i.e. true positive (fraud) or false positive (genuine). The proposed model uses suitable pre-processing, attributes selection techniques along with proposed Bagging EM. K-fold cross validation is used as ‘test split’ method.

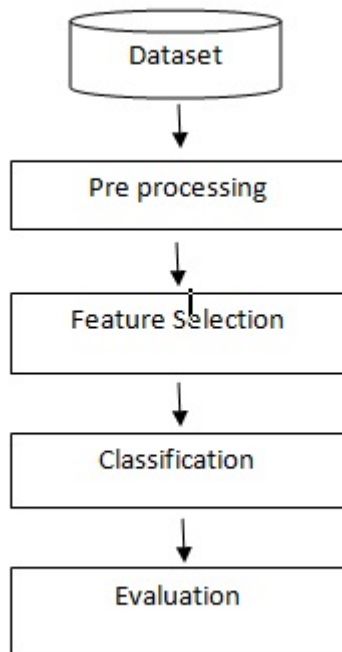


Figure 2: Proposed Model for Credit Card Fraud Detection

4. EXPERIMENTAL SETUP, METHODOLOGY AND PERFORMANCE ANALYSIS

4.1 Experimental Setup

Weka 3.6.11 is used as DM tool for simulation purpose. Weka is installed over Windows 10 Operating System. For this research a state of art research dataset from USCD-FICO competition [7, 8] is used. The competition is hosted

by: FICO a leading provider of technologies and University of San Diego. Dataset description is presented in Table 1.

| Dataset | No. of Features | Total Instances | No. of Instances (Yes) | No. of Instances (No) |
|-----------|-----------------|-----------------|------------------------|-----------------------|
| USCD-FICO | 20 | 10,0000 | 97346 (97.35%) | 2654 (2.65%) |

Table 1: Details of Dataset

4.2 Methodology

The experiment methodology involves following steps:

1. Preprocessing of Dataset
2. Applying Feature selection
3. Applying new naïve bayes classifier
4. Evaluate result

4.3 Performance Analysis

The performance analysis is done on the basis of following metrics:

Prediction Rate: Prediction rate refers to the percentage of correct predictions among all test data.

$$\text{Prediction Rate} = \frac{TP}{TP + TN} * 100$$

False Alarm Rate (FAR): The percentage of normal data which is wrongly recognized as of different class is FAR, and is defined as follows:

$$\text{False Alarm Rate} = \frac{FP}{FP + TN} * 100$$

5. CONCLUSION & FUTURE WORK

Credit Card Frauds are common as attackers gather information through transactions and registered accounts. This opens new confronts in the field of fraud detection and prevention, but prevention is of course better than detection. The simple techniques like database comparison and pattern matching are not enough for detecting such frauds because fraudulent transactions are rare within huge number of genuine transactions. So, Predictive models are of prime importance for banks to detect CCFs. The proposed ensemble based CCFD predictive model is compared with base learner and state-of-art model. The proposed work is compared on basis of two functional metrics: Prediction Rate and FPR proved to be better. The efforts shown that, Ensemble ML methods are more suitable for detecting frauds with credit cards. In future, more efforts methods will be worked out to improve the Fraud Catching Rate. At the same time proposed predictive model would be integrated with live stream to find the online fraudulent transaction instantly. In future we intend to build up a cloud based ML

application for detecting frauds in financial transactions done with cards.

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