

An Approach to Credit Card Fraud Detection

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ABSTRACT

Due to the rapid advancement in the electronic commerce technology in today's world the use of credit cards has gradually increased. Since credit card is the most favored mode of payment, the number of fraud cases which are associated with it is also rising every day. Fraud detection means that identifying the fraud as quickly as possible once it has been performed. Fraud detection methods are continually developed to defend the criminals in adapting to their strategies used. The transaction made are classified as normal, abnormal or suspicious depending on the initial belief. Once a transaction is found to be suspicious, belief is further strengthened or weakened according to the similarity with fraudulent or genuine transaction history using various algorithms. It is necessary that the credit card companies are able to identify fraudulent credit card transactions so that the customers are not charged for the items that they didn't purchase. Such problems can be tackled with various technologies of Machine Learning and its relevant areas.

The Credit Fraud Detection project plans to demonstrate the modelling of a dataset making use of machine learning with Credit Card Fraud Detection. The Credit Card Fraud Detection Problem consists of modelling past credit card transactions with the data of the ones that turned out to be fraud among that. This model is then used so as to recognize whether a new transaction is fraudulent or not in all. The objective here is to detect 100% of the fraudulent transactions in the dataset while minimizing the incorrect fraud classifications. Credit Card Fraud Detection is a classic sample of classification. In this process, it is focused on analyzing and pre processing data sets as well as the deployment of multiple anomaly detection algorithms such as Random Forest Classifier, Naive Bayes, etc. on the given Credit Card Transaction data.

Keywords-Machine Learning, Data Mining, Credit Card Fraud, Normal and Fraud Transaction.

Date of Submission: 06-07-2021

Date of acceptance: 20-07-2021

I. INTRODUCTION

The popularity of on-line shopping has grown day by day in today's world. According to a 2005 AC Nielsen survey, 1 in 10 of the world's population buys online. Nowadays, credit card is the most popular method of payment. As the number of credit card users increases worldwide, identity theft is on the rise and fraud is on the rise. Credit card-based purchases can be divided into two types: 1) physical card purchases and 2) visual card purchases. At the purchase of a physical card, the cardholder personally presents the card to make a payment. While buying a physical card, the attacker needs to steal the credit card and build a signature to buy it. For visual card purchases, only card details are required such as card number, expiration date, secure code, etc. Such purchases are usually made online or over the phone. To commit fraud in these types of purchases, a person simply needs to know the details of his or her card. The online shopping mode is made mostly by credit card. Credit card fraud has been increasing day by day. The amount of financial losses due to credit card fraud increases as the use of credit cards becomes more common. Security means using your credit card safely and avoiding fraudulent appearances. The purpose of security is to prevent the use of fraudulent credit cards. In cases of fraud there are issues such as lost cards, stolen cards, application fraud, fraudulent mail, postal fraud and unpaid fraud (NRI). To reduce this risk, credit card security is required. 'Fraud' in credit card transactions is the unauthorized and unwanted use of an account by a person other than the owner of that account. Necessary preventive measures can be taken to prevent this abuse and can be studied in such fraudulent behaviour to reduce and prevent similar incidents in the future. Someone else's credit card for personal reasons while the owner and the issuing authorities do not know that the card is in use.

II. LITERATURE SURVEY

1. A Survey on Different Data Mining & Machine Learning Methods for Credit Card Fraud Detection

In the above paper, different methods of fraud detection in credit cards were investigated. Firstly, significance of the subject was stated and existing deficiencies in traditional systems were mentioned. Fake transactions have also varying degrees of risks and ways should be found for finding transactions with highest risk in quicker and more accurate manner. For identification of these transactions, common data mining methods alone do not suffice. Innovative algorithms should be used for the effective results. Due to rapid growth in cashless transaction, the chances of number of fraudulent transactions can also be increasing. A fraudulent transaction can be identified by analyzing various behaviors of credit card customers from the previous transaction history datasets. If any deviation is noticed in spending behavior from available patterns, it is possibly of fraudulent transaction. Data mining and machine learning techniques are widely used in credit card fraud detection.

2. Credit Card Fraud Detection Using Random Forest Algorithm

In this paper it mainly focuses on credit card fraud detection in real world. Here the credit card fraud detection is based on fraudulent transactions. Generally credit card fraud activities can happen in both online and offline. But in today's world online fraud transaction activities are increasing day by day. So in order to find the online fraud transactions various methods have been used.

In the paper mentioned, the authors made use of Random Forest Algorithm for finding the fraudulent transactions and the accuracy of those transactions. This algorithm is based on supervised learning algorithm where it uses decision tree for classification of the dataset. After classification of dataset a confusion matrix was obtained. The performance of Random Forest Algorithm is evaluated based on the confusion matrix. The results obtained from processing the dataset gives accuracy of about 90%. Thus, using this Random Forest algorithm and decision tree algorithm they have extracted the accurate percentage of detection of fraud from the given dataset by studying its behavior. A confusion matrix is basically a summary of prediction results or a table which is used to describe the performance of the classifier on a set of test data where true values are known. It provides visualization of an algorithm's performance and allows easy identification of classes. Thus, resulting in the computing of most performance measures by giving insights not only the errors being made by the classification model but also tell the type of errors being made.

3. Credit Card Fraud Detection Using Predictive Modelling

In this paper author proposed that fraud detection is a critical problem affecting large financial companies that have increased due to the growth in credit card transactions. This paper presents detection of frauds in credit card transactions, using data mining techniques of Predictive modeling, logistic Regression, and Decision Tree. The dataset is highly unbalanced, the positive class (frauds) Account for 0.172% of all transactions.

Decision trees are used to choose between several courses of action. It provides effective structure to investigate the possible outcomes. Decision trees use tree structure to build classification or regression model. A decision tree is a flowchart like tree structure, where non leaf node denotes a test on attribute. In the results, the decision tree will have a decision node and leaf nodes. Predictive modeling is used to analyze the data and predict the outcome. Predictive modeling used to predict the unknown event which may occur in the future. In this process, we to create, test and validate the model. There are different methods in predictive modeling. They are machine learning, artificial intelligence and statistics. Once created a model, it can use many times, to determine the probability of outcomes. So predict model is reusable. Historical data is used to train an algorithm. The predictive modeling process is an iterative process and often involve training the model, using multiple models on the same dataset.

4. Credit Card Fraud Detection and Prevention using Machine Learning

In this survey, a review of a contextual investigation including the identification of Credit Card misrepresentation where information standardization is applied prior to cluster analysis and with results obtained from the use of Cluster Analysis and Artificial Neural Network on the discovery of extortion has indicated that neuronal data Sources may be limited by bundling properties. What's more, encouraging outcomes can be gotten by utilizing standardized information and information ought to be MLP prepared. This examination depended on solo learning. Noteworthiness of this paper was discovering an estimate and reducing the measure of costs. The result was 23% and the calculation they found was the minimum chance of Bayes. In this system, a collective replacement comparison measure is proposed that represents profits and losses due to fraud detection. Using the existing cost measure, a cost sensitive method that depends on the Bayes minimum risk is used.

5. CreditCardFraudDetectionusingMachineLearningAlgorithms

In the above mentioned survey, the authors focused on - Card transactions are always unfamiliar when compared to previous transactions made by the customer. This unfamiliarity is a very difficult problem in real-world when it is recalled concept drift problems. Concept drift can be said as a variable which changes over time and in unforeseen ways. These variables cause a high imbalance in data. The main aim of the research was to overcome the problem of Concept drift to implement on real-world scenario. They aimed to overcome three main challenges with card frauds related dataset i.e., strong class imbalance, the inclusion of labelled and unlabelled samples, and to increase the ability to process a large number of transactions.

III. SYSTEM DESIGN

System design is the process of defining the architecture, components, modules, interfaces, and data for a system to satisfy specified requirements. Systems design could be seen as the application of systems theory to product development.

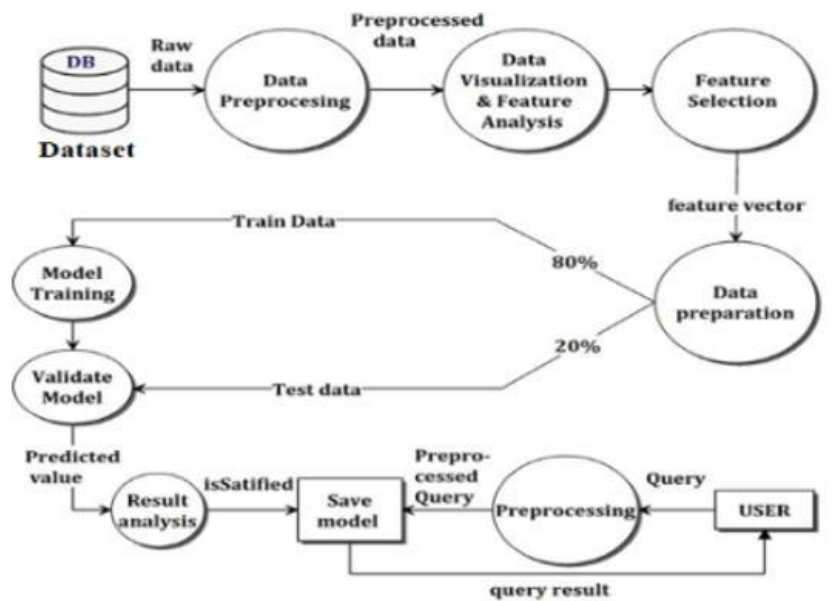


Figure 1. System Architecture

The architecture shows the flow of the process of the system. It consists of various levels; from the dataset being provided with its pre-processing, extraction, training and testing model to the evaluation of results of the transaction. It also shows the percentage of accuracy of the training model by the specific algorithm used in the implementation.

IV. IMPLEMENTATION

In computer science, an implementation is a realization of a technical specification or algorithm as a program, software component, or other computer system through computer programming and deployment. Many implementations may exist for a given specification. In this system, python is used to build the functionality of the system and graphical user interface (GUI). It makes use of different libraries of Python, like tkinter, numpy, panda, matplotlib and seaborn to build the system.

The different Modules of the Credit Card Fraud Detection Project:

Data Collection - Data is the most important part when we work on prediction systems. It plays a specific role in whole project i.e., the system depends on that data. So, collection of data and then select is the first and the critical step which should be performed properly. The data which is used in the process is a set of transactions collected from the credit card transactions records. This step is basically concerned with selecting the subset of all the available data that is playing its role in detection.

Data Preprocessing – The next step after the selected data is organize it by the process of by formatting, cleaningand finally,samplingfromit.The datapre-processingstepsincludes:

- **Formatting:** The data have been selected may not be in a format which is suitable to work with further.Like, the data may be in a relational database and it should be in a flat file or it may be in a proprietary fileformatbutit shouldbe inarelational database or anytextfile.
- **Cleaning:** Cleaning data means here that it is the removal or fixing of missing data. There may be the datainstances which are incomplete and do not carry the data that is required or needed to address the givenproblem.These instancesmayneedtobe removedfromthedatast.
- **Sampling:** The final step in data pre processing is Sampling which deals there may be far more selecteddata available than required to be used. The more the data, it can result in much longer running times foralgorithmsused and larger computationalaswellasmemory requirementswithinthesystem.
- **Feature Extraction and Algorithm** – Nowadays, it is becoming quite common to work with datasets ofhundreds or even thousands of features. If the number of features becomes similar or even bigger than thenumber of observations stored in a dataset, then this can further lead to a machine learning model sufferingfromoverfitting.So,toavoidsuchascenario,itisnecessarytoapplyeitherregularizationordimensionality reductiontechniqueswhichisfeatureextraction.

Data Visualisation - Data visualization is the step wherein all the data will be transformed into some form of plotsand analyzed further from that instead of any such tables. As a human being, we are more likely to take a lot ofinformation from diagrammatic representation thanits substitutes. If we want to convert the data from aboringtable into an interesting pictorial form like a scatter plotting, then it can be done by making use of very goodpackageswhich areavailablefrom availablepopular programming languageswhich areused commonly.

Feature Extraction and Algorithm – Nowadays, it is becoming quite common to work with datasets of hundredsor even thousands of features. If the number of features becomes similar or even bigger than the number ofobservations stored in a dataset, then this can further lead to a machine learning model suffering from overfitting.So, to avoid such a scenario, it is necessary to apply either regularization or dimensionality reduction techniqueswhich is feature extraction. In machine learning, the dimensionality of the dataset is same as thenumber ofvariableswhichareusedtorepresentit.Featureextractiontechniquescanalsoleadtoothertypesofadvantagesuchas accuracy improvements, overfitting riskreduction, speed upintraining, improved data visualization,increase in efficiency of the model. Feature Extraction is used to reduce the number of features in a dataset bycreating the new features from the existing ones. It discards the original features. The algorithm used are RandomForest, Decision Tree and Naïve Bayes. These algorithm are stable and works well. So, they contribute in reducingtheoverall performanceof thesystem.

V. RESULTS

Out[3]:

| | Time | V1 | V2 | V3 | V4 | V5 | V6 | V7 | V8 | V9 | ... | V21 | V22 | V23 | V2 |
|--------|----------|------------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----|-----------|-----------|-----------|----------|
| 0 | 0.0 | -1.359807 | -0.072781 | 2.536347 | 1.378155 | -0.338321 | 0.462388 | 0.239599 | 0.098698 | 0.363787 | ... | -0.018307 | 0.277838 | -0.110474 | 0.06692 |
| 1 | 0.0 | 1.191857 | 0.266151 | 0.166480 | 0.448154 | 0.060018 | -0.082361 | -0.078803 | 0.085102 | -0.255425 | ... | -0.225775 | -0.638672 | 0.101288 | -0.33984 |
| 2 | 1.0 | -1.358354 | -1.340163 | 1.773209 | 0.379780 | -0.503198 | 1.800499 | 0.791461 | 0.247676 | -1.514654 | ... | 0.247998 | 0.771679 | 0.909412 | -0.68928 |
| 3 | 1.0 | -0.966272 | -0.185226 | 1.792993 | -0.863291 | -0.010309 | 1.247203 | 0.237609 | 0.377436 | -1.387024 | ... | -0.108300 | 0.005274 | -0.190321 | -1.17557 |
| 4 | 2.0 | -1.158233 | 0.877737 | 1.548718 | 0.403034 | -0.407193 | 0.095921 | 0.592941 | -0.270533 | 0.817739 | ... | -0.009431 | 0.798278 | -0.137458 | 0.14126 |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 284802 | 172786.0 | -11.881118 | 10.071785 | -9.834783 | -2.066656 | -5.364473 | -2.606837 | -4.918215 | 7.305334 | 1.914428 | ... | 0.213454 | 0.111864 | 1.014480 | -0.50934 |
| 284803 | 172787.0 | -0.732789 | -0.055080 | 2.035030 | -0.738589 | 0.868229 | 1.058415 | 0.024330 | 0.294869 | 0.584800 | ... | 0.214205 | 0.924384 | 0.012463 | -1.01622 |
| 284804 | 172788.0 | 1.919565 | -0.301254 | -3.249640 | -0.557828 | 2.630515 | 3.031260 | -0.296827 | 0.708417 | 0.432454 | ... | 0.232045 | 0.578229 | -0.037501 | 0.64013 |
| 284805 | 172788.0 | -0.240440 | 0.530483 | 0.702510 | 0.689799 | -0.377961 | 0.623708 | -0.686180 | 0.679145 | 0.392087 | ... | 0.265245 | 0.800049 | -0.163298 | 0.12320 |
| 284806 | 172792.0 | -0.533413 | -0.189733 | 0.703337 | -0.506271 | -0.012546 | -0.649617 | 1.577006 | -0.414650 | 0.486180 | ... | 0.261057 | 0.643078 | 0.376777 | 0.00879 |

284807 rows x 31 columns

Figure 1. Displayed the dataset stored in the system



Figure2.HeatmapofCorrelation

```
0: 284315
1: 492
Name: Class, dtype: int64
```

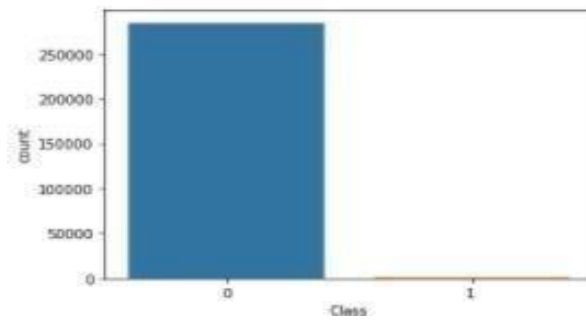


Figure3.Displaysthenumberoffraudandnormaltransactiondata

Out[32]: <AxesSubplot: xlabel='Algorithms', ylabel='Accuracy score'>

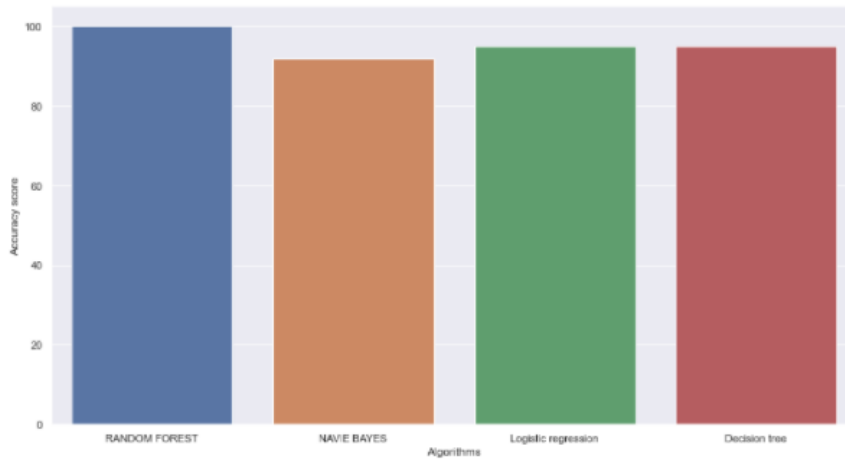


Figure4.Accuracyscoreobtainedbydifferentialgorithms

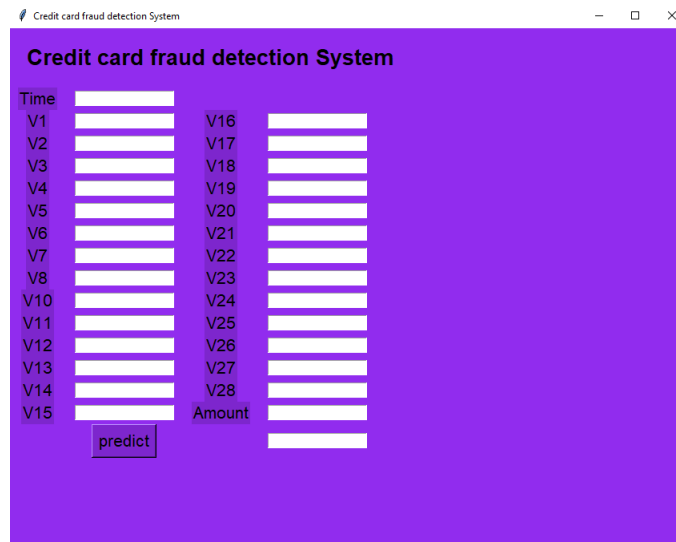


Figure5. GUIofCreditCardFraudDetection

VI. CONCLUSION

Credit Card Fraud Detection is undoubtedly an act of criminal dishonesty. This system has listed out the most common methods of fraud along with their detection methods; also reviewed recent findings in this field.

This system has also explained in detail that how machine learning techniques can be applied to get better results in fraud detection consisting of algorithm, code, explanation its implementation and the results. When the entire dataset was fed into the algorithm then the precision rose to 33%. This high percentage is as per the huge imbalance between the number of provided valid and genuine transactions. Since it is based on machine learning algorithms so the program will only increase its efficiency over time when more data is put into it.

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