An Approach For Prediction Of Loan Approval Using Machine Learning Algorithm

Abstract—In our banking system, banks have many products to sell but main source of income of any banks is on its credit line. So they can earn from interest of those loans which they credits. A bank’s profit or a loss depends to a large extent on loans i.e. whether the customers are paying back the loan or defaulting. By predicting the loan defaulters, the bank can reduce its Non-Performing Assets. This makes the study of this phenomenon very important. Previous research in this era has shown that there are so many methods to study the problem of controlling loan default. But as the right predictions are very important for the maximization of profits, it is essential to study the nature of the different methods and their comparison. A very important approach in predictive analytics is used to study the problem of predicting loan defaulters: The Logistic regression model. The data is collected from the Kaggle for studying and prediction. Logistic Regression models have been performed and the different measures of performances are computed. The models are compared on the basis of the performance measures such as sensitivity and specificity. The final results have shown that the model produce different results.

I. INTRODUCTION

Small loan is an important aspect of our everyday life: it allows aspiring entrepreneurs to get started on ideas that could be grown into business; it allows curious students to afford higher education that is otherwise unavailable without a stable income; more importantly, it allows ordinary people who have no friends or relatives for support to obtaining short-term financial assistance and get back on their feet to fight for the American Dream. Nevertheless, with loan it comes with the possibility of default as well. Default is a financial term describing the failure of meeting the legal obligation of a loan - paying back the principal and interest. It’s a common problem in the financial industries and one of the major risks of offering loans. Of course, default does not happen the majority of the time and the lending banks usually able to make up the loss from a defaulting loan from other fully paid loans and their accompanied interests. Furthermore, banks issuing loans with higher interest rate to individuals with high probability of default - the financial institutions are trading off an increased chance of default with an increased profit from the high interest. All things considered, default is a fact of life and most financial institutions have a well-established practice to minimize its impact and absorbing the loss. But what about a situation where instead of a single bank is issuing the loan, the loan is comprised of funds from several investors? Lending Club is one of the many peer-to-peer lending company that gives rise to this peculiar situation. In plain words, peer-to-peer lending company acts as a broker between borrowers and investors. The company creates a platform where borrowers can create small unsecured personal loans, and investors can seek out these loans and decide which loans to invest from. Borrowers obtain the loan they want, investors get to profit from the loan interest, and the company gets a cut from both parties (origination fee from borrowers and service fee from investors). This also means that when a loan goes default, it’s no longer a single bank that is absorbing the loss - single or multiple individual investors will be absorbing it instead. The overall profit might be positive if all the loans were originated from a single lender as other fully paid loans could cover the loss, but this is no longer the case as there will be winners and losers among this new form of lending practices if the investors did not diversify. An obvious solution to this problem is to predict whether a particular loan will go default based on initial information provided by the borrowers and their credit report. There’s no doubt LendingClub already has an existing model in place to approve loans posted on their website. This paper will explore the process and result on formulating a new machine learning model that could predict a loan default; but more importantly, the model will focus on
minimizing the overall loss in investment of bad loans in order to lessen the burden passed onto individual investors. As a side note, the paper will also explore privacy-preserving mechanism on sensitive information provided from the borrower’s credit report. The end goal is to evaluate a simplified version of RAPPOR (Randomized Aggregately PrivacyPreserving Ordinal Response) and determine whether data that have beenhashed by this algorithm could still be used to predict loan default as stated previously.

II. LITERATURE SURVEY

Amira Kamil Ibrahim Hassan, Ajith Abraham (2008) uses a prediction model which is constructed using three different training algorithms to train a supervised twolayer feed-forward network. The results show that the training algorithm improves the design of loan default prediction model.

Angelini (2008) used a neural network with standard topology and a feed-forward neural network with ad hoc connections. Neural network can be used for prediction model. This paper shows that the above two models give optimum results with less error.

Ngai (2009) uses the classification model for predicting the future behaviour of costumers in CRM. In CRM domain, the mostly used model is neural network. He recognized eighty seven articles associated to data mining applications and techniques between 2000 and 2006.

Dr. A. Chitra and S. Uma (2010) introduced a ensemble learning method for prediction of time series based on Radial Basis Function networks (RBF), K - Nearest Neighbor (KNN) and Self Organizing Map (SOM). They proposed a model namely PAPEM which perform better than individual model.

Akkoç (2012) used a model namely hybrid Adaptive Neuro-Fuzzy Inference model, grouping of statistics and Neuro-Fuzzy network. A 10-fold cross validation is used for better results and a comparison with other models.

Sarwesh Site, Dr. Sadhna K. Mishra (2013) proposed a method in which two or more classifiers are combined together to produce an ensemble model for the better prediction. They used the bagging and boosting techniques and then used random forest technique.

Maher Alaraj, Maysam Abbod, and Ziad Hunaiti (2014) proposed a new ensemble method for classification of costumer loan. This ensemble method is based on neural network. They state that the proposed method give better results and accuracy as compared to single classifier and any other model.

AlarajM., AbbodM (2015) introduced a model that are based on homogenous and heterogeneous classifiers. Ensemble model based on three classifiers that are logistic artificial neural network, logistic regression and support vector machine.

III. PROBLEM STATEMENT

To design and implement the system using machine learning and data mining to predict the probability of the user to get loan or not from bank to improve the accuracy and to minimize the frauds. Banks, Housing Finance Companies and some NBFC deal in various types of loans like housing loan, personal loan, business loan etc in all over the part of countries. These companies have existence in Rural, Semi Urban and Urban areas. After applying loan by customer these companies validates the eligibility of customers to get the loan or not. This paper provides a solution to automate this process by employing machine learning algorithm. So the customer will fill an online loan application form. This form consist details like Sex, Marital Status, Qualification, Details of Dependents, Annual Income, Amount of Loan, Credit History of Applicant and others. To automate this process by using machine learning algorithm, First the algorithm will identify those segments of the customers who are eligible to get loan amounts so bank can focus on these customers.

IV. PROPOSED SYSTEM

Decision tree algorithm in machine learning methods which efficiently performs both classification and regression tasks[2]. It creates decision trees. Decision trees are widely used in the banking industry due to their high accuracy and ability to formulate a statistical model in plain language. In Decision tree each node represents a feature (attribute), each link (branch) represents a decision (rule) and each leaf represents an outcome (categorical or continues value). Using different data analytics tools loan prediction and there severity can be forecasted. In this process it is required to train the data using different algorithms and then compare user data with trained data to predict the nature of loan. Several R functions and packages were used to prepare the data and to build the classification model. The work proves that the R package is an efficient visualizing tool that applies data mining techniques. Using R Package, customer’s data analysis can be done and depends on that bank can sanction or reject the loan. In real time customers data sets may have many missing and imputed data which needs to be replaced with valid data generated by making use of the available completed data. The dataset has many attributes that define the credibility of the customers seeking for several types of loan. The values for these attributes can have outliers that do not fit into the regular range of data.

Fig. System Architecture

V. ALGORITHM USED

DT is a supervised learning algorithm used to solve classification and regression problems too. Here, DT uses tree representation to solve the prediction problem, i.e., external node and leaf node in a tree represents attribute and class labels respectively. The pseudo code for DT model is depicted in the following section:

Step 1: Best attribute is chosen as the tree’s root.
Step 2: Training set is divided into subsets, such that, each subset comprises similar value for an attribute.
Step 3: Step 1 and Step 2 are repeated for all subsets until all the leaf nodes are traversed in a tree.
The analytical process started from data cleaning and processing. Missing value imputation with mice package, then exploratory analysis and finally model building and evaluation. The best accuracy on public test set is 0.811. This brings some of the following insights about approval. Applicants with Credit history not passing fails to get approved. Probably because they have a probability of not paying back. Most of the Time, Applicants with high income sanctioning low amount is more likely get approved which make sense, more likely to pay back their loans. Some basic characteristic gender and marital status seems not to be taken into consideration by the company.

VI. CONCLUSION

VII. REFERENCES


[10] Trevor Hastie, Robert Tibshirani, and Jerome Friedman,”The Elements of Statistical Lear


