

Micro Credit Defaulter Project

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INTRODUCTION

• Business Problem Framing

This problem helps you understand the field of 'Micro Credit' and about the financial background of the people availing them.

Conceptual Background of the Domain Problem
 The problem has been divided by EDA steps, then PCA analysis, confusion matrix and logistic regression, ROC and AUC curve, decision tree, Gradient boosting, knn classifier, Random Forest classifier, hyperparameter optimization.

Review of Literature

This is a comprehensive summary of the research done on the topic of "Micro Credit'. It also shows about the financial background of the people availing them. People who can return the loan in time and about the people who will be defaulter list.

Motivation for the Problem Undertaken
 To find out about the group of people who can return the loan in time and about the people who will be defaulter list.

Analytical Problem Framing

 Mathematical/ Analytical Modeling of the Problem PCA Standardization

$$z = \frac{value - mean}{standard\ deviation}$$
 Covariance Matrix computation

$$\left[\begin{array}{ccc} Cov(x,x) & Cov(x,y) & Cov(x,z) \\ Cov(y,x) & Cov(y,y) & Cov(y,z) \\ Cov(z,x) & Cov(z,y) & Cov(z,z) \end{array} \right]$$

Covariance Matrix for 3-Dimensional Data Linear Regresion

$$Y = a + bX$$

ROC curve is a graph with: The x-axis showing 1 - specificity (= false positive fraction = FP/(FP+TN)) The y-axis showing sensitivity (= true positive fraction = TP/(TP+FN))

- Data Sources and their formats
 Data Provided by Flip Robo Technologies
- Data Preprocessing Done
- Remove duplicate or irrelevant observations
- Fix structural errors
- Filter unwanted outliers
- Handle missing data
- Changed strings into integers
- Hyperparameter optimization using grid search cv
- Validate

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Data Inputs- Logic- Output Relationships

Micro credit defaulter model

EDA steps, then PCA analysis, confusion matrix and logistic regression, roc and auc curve, knn and hyperparameter optimization using grid search cv. This process helps us in isolating low-income families, high-income families and their money usage pattern. This also helps

microfinancing organizations to pinpoint the families that can and cannot repay loan amount within specified time.

 State the set of assumptions (if any) related to the problem under consideration

Dropped down the columns and created an updated datafile.

Hardware and Software Requirements and Tools Used

Hardware - Laptop

Software -

import pandas as pd, import os, import csv, import sklearn, import numpy as np, import matplotlib.pyplot as plt, import seaborn as sns, %matplotlib inline, from sklearn.preprocessing import MinMaxScaler, PCA, knn, ROC and AUC curve, Hyperparameter optimization using grid search cv.

Model/s Development and Evaluation

 Identification of possible problem-solving approaches (methods)

EDA Method, PCA analysis, Covariance, Confusion Matrix, Linear Regression and prediction, PCA, knn, ROC and AUC curve, Hyperparameter optimization using grid search cv.

Testing of Identified Approaches (Algorithms)

EDA Method, PCA analysis, Covariance, Confusion Matrix, knn, ROC and AUC curve, Hyperparameter optimization using grid search cv.

Run and Evaluate selected models

Linear Regression, prediction, knn, ROC and AUC curve, Hyperparameter optimization using grid search cv.

 Key Metrics for success in solving problem under consideration

Classification Matrix and Confusion Matrix.

Visualizations

Contour, Heat map, pair plot, box plot, histogram, gridsearchcv.

• Interpretation of the Results

This process helps us in isolating low-income families, high-income families and their money usage pattern. Families taken or not taken loans. This also helps microfinancing organizations to pinpoint the families that can and cannot repay loan amount within specified time.

CONCLUSION

Key Findings and Conclusions of the Study

This process helps us in isolating low-income families, high-income families and their money usage pattern. Families taken or not taken loans. This also helps microfinancing organizations to pinpoint the families that can and cannot repay loan amount within specified time. This project also helps us to learn how big the world of microfinancing is and how it helps in eradicating poverty.

 Learning Outcomes of the Study in respect of Data Science

EDA and PCA algorithm helped in cleaning excess data and helped it in putting it in visualised matter. Hyperparameter optimization using grid search cv helped to create a model.

Limitations of this work and Scope for Future Work

Excessive unfiltered data.

Can help microfinancing organization identify needy families and can assist them.