

# IME672A Data Mining & Knowledge Discovery

## Project Report LOSS GIVEN DEFAULT

Group Number: 3

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## Introduction

Loss-given default (LGD) is the amount of money a bank or other financial institution loses when a borrower defaults on a loan. It is a very important parameter in risk models used by banks to calculate their economic capital and expected losses.

In this project, we aim to predict the loss of an asset(lgd\_time) to a bank. In the given dataset we have multiple parameters which will be used in various models to project this loss.

The data set has been kindly provided by a European bank and has been slightly modified and anonymized. It includes 2,545 observations on loans and LGDs. Key variables are:

- LTV: Loan-to-value ratio, in %
- Recovery\_rate: Recovery rate, in %
- lgd\_time: Loss rate given default (LGD), in %
- y\_logistic: Logistic transformation of the LGD
- Inrr: Natural logarithm of the recovery rate
- Y\_probit: Probit transformation of the LGD
- $\blacksquare$  purpose1: Indicator variable for the purpose of the loan; 1 = renting purpose, 0 = other
- event: Indicator variable for a default or cure event; 1 = event, 0 = no event

#### Our Approach in Brief

After visualizing and preprocessing the data, we split the data into training and testing data using stratified sampling, splitting it into a 7:3 ratio. We used the training dataset for training the following models:-

- 1. Linear Regression
- 2. Transformed Linear Regression
- 3. Probit Transformed Regression
- 4. Decision Tree Regression
- 5. Support Vector Regression
- 6. Random Forest Regression
- 7. Tobit Regression

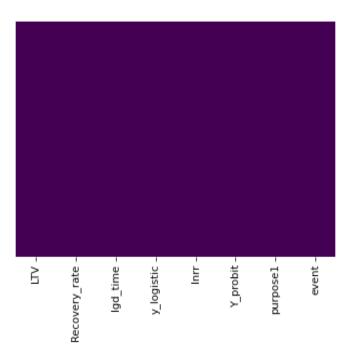
We tested these models on the testing dataset and compared their performance on 2 factors:

- b. Mean-Square Error Value
- c. R^2 Value

#### **Salient Features Of Dataset**

- The given dataset has 2545 rows(data points) and 8 columns ( attributes).
- LTV is a continuous-valued numeric attribute.
- Igd\_time is a numeric attribute with values between 0 and 1.
- Recovery\_rate is 1 lgd\_time.
- y\_logistic, Y\_probit and lnrr are transformations of the lgd\_time variable.
- purpose1 is a binary variable that takes value 1 if the loan is for renting purposes and for other purposes 0.
- event is also a binary variable which is 1 when the borrower defaults the loan and else 0

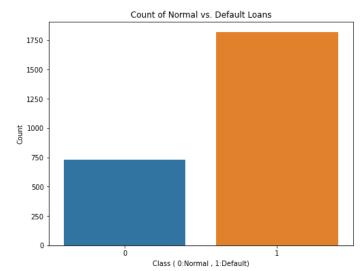
## Finding the NULL values



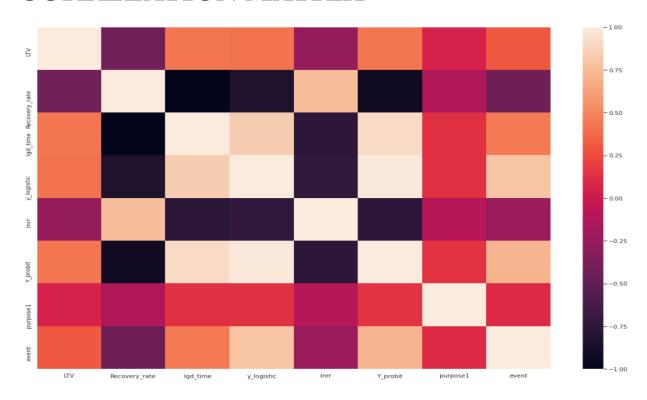
• There are no NULL values and missing data present in our dataset. So we didn't have to do any data preprocessing.

#### **DATA VISUALISATION**

- Default loans Distribution
- In our dataset 1817 loans are defaulted while the other ones are non default loans or the loans which were cleared on time
- It is clear that for non-default loans that the recovery\_rate will be 1 hence lgd\_time will be 0.
- We then visualized the distribution of various variables through box plots, violin plots.



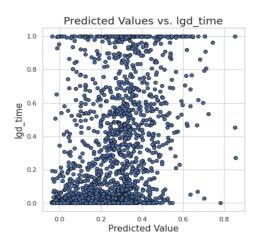
## **CORRELATION MATRIX**



## **Models**

• Linear Regression

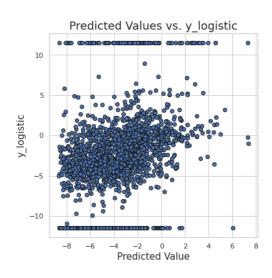
$$LGD_i = \beta' x_i + \epsilon_i$$



Mean Square Error: 0.108 R<sup>2</sup> Score: 0.193

#### • Logistic Transformed Linear Regression

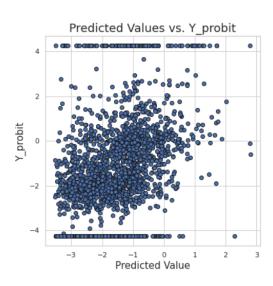
$$\ln \ \frac{LGD_i}{1 - LGD_i} = \beta' x_i + \epsilon_i$$



 $Mean\ Square\ Error: 36.88 \qquad R^2\ Score: 0.182$ 

#### • Probit Transformed Linear Regression

$$\Phi^{-1}(LGD_i) = \beta' x_i + \epsilon_i$$



Mean Square Error: 5.30 R<sup>2</sup> Score: 0.197

• Random Forest Regression - MSE: 46.47

- Decision Tree Regression MSE: 52.22
- Support Vector Regression MSE: 30.39
- Tobit Regression MSE: 52.22

#### **Results and Interpretation**

- The Simple Linear Regression using statsmodels OLS gives a mse of 0.10 and a low R<sup>2</sup> value of 0.19. The correlation between the dependent variable and the independent features is weak.
- Logistic and Probit Transformed Regressions give a pretty solid relation with a similar R<sup>2</sup> value but they seem to fit the residuals well.
- Other regressions like Decision Tree, Support Vector, and Random Forest give a similar mean squared error and also do not seem to capture the relation between the features.
- The best model which fits the data is the Probit Transformed Regression as it captures the residuals properly and also provides a better relation between the variables.