

VIRTUAL INTERNSHIP EXPERIENCE

ID/X PARTNERS

Hosted By:

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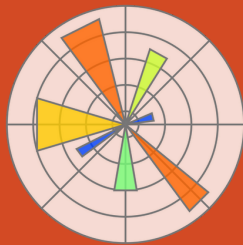
2 0
2 3



Business Understanding

- Credit risk is known as the risk of borrower's failure to repay a loan
- Assessing borrower's risk to repay the loan is a crucial thing in credit risk assessment
- We can use machine learning to automate the process

Tools:



Analytical Approach

- Descriptive analysis
- Graph analysis
- Predictive modelling (classification)



Data Requirements & Collection

- I required a dataset of customer loan from financial company
- The dataset is collected by ID/X Partners from a company

Data Understanding

- This dataset has 74 columns / features
- Consists of 52 numerical & 22 non-numerical features
- Many features have missing values
- There are 17 null features

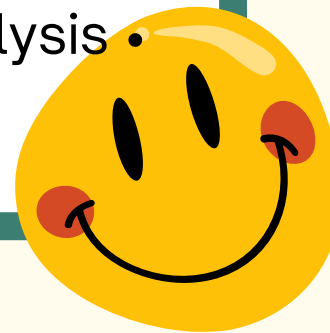




Data Preparation

Missing Value: Removing and Imputing Feature
Engineering: Categorical Encoding, Log
Transform, Standardization

Feature Selection using Correlation Analysis • Removing outliers using IQR Method



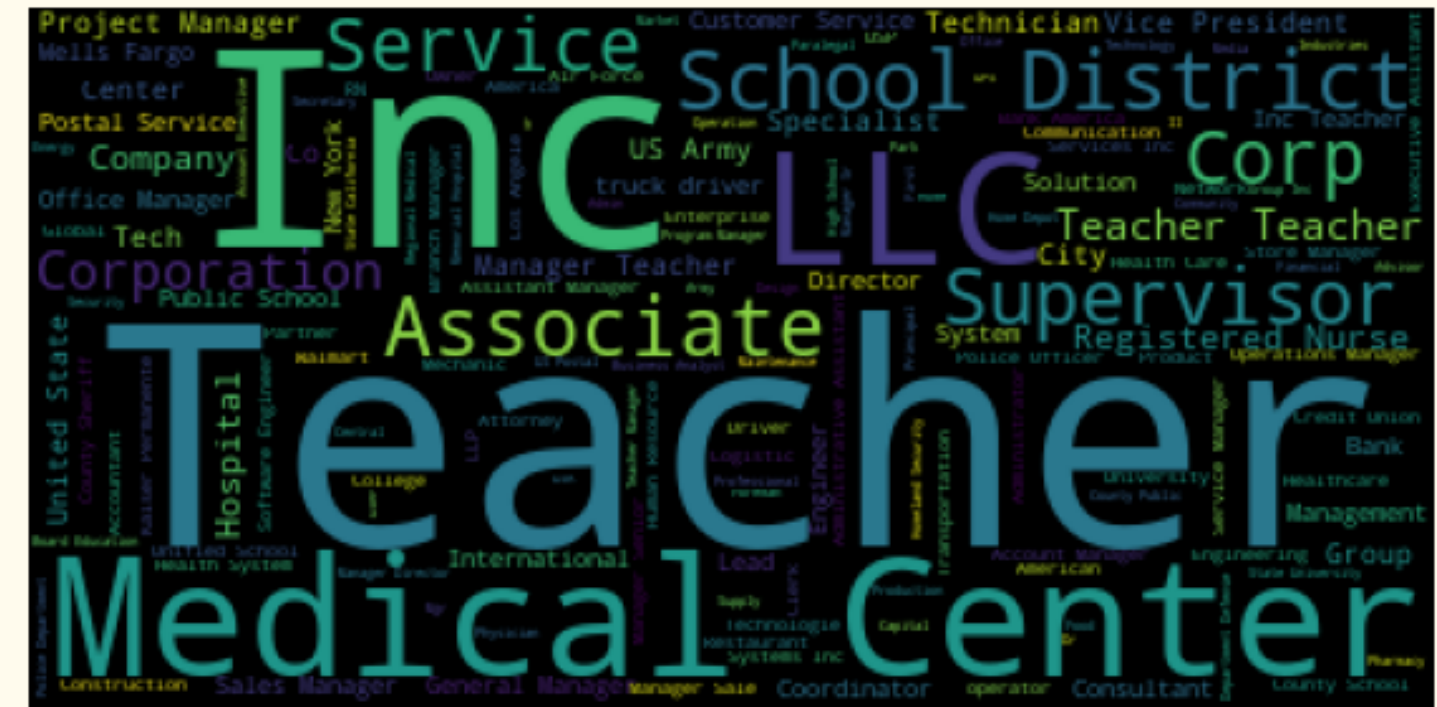
What are the employee titles of our borrowers?



Exploratory Data Analysis

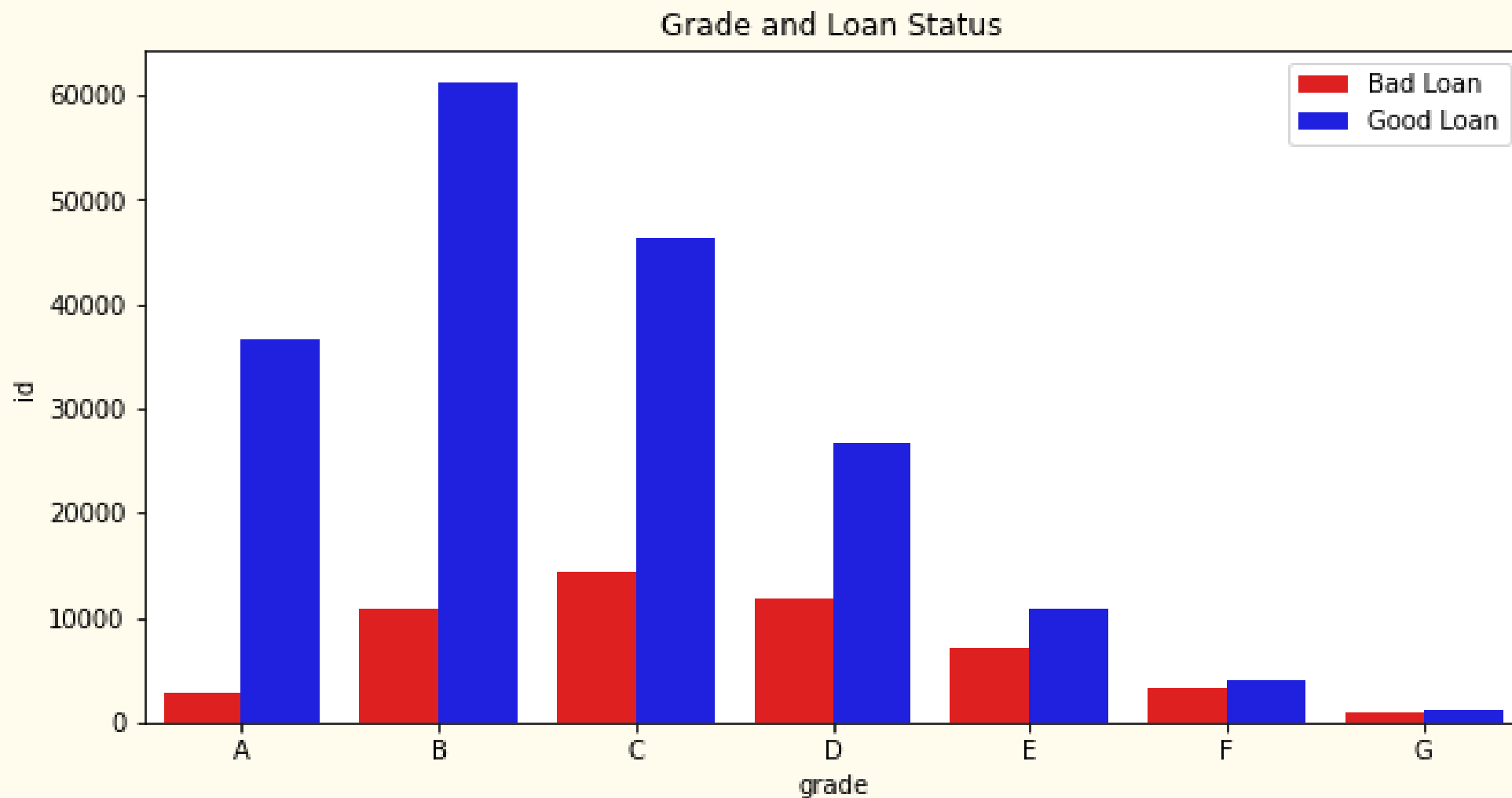
Good Loan (1) : Fully Paid, Does not meet the credit policy. Status:Fully Paid

Bad Loan (0) : Charged Off, Does not meet the credit policy. Status:Charged Off, Default, Late (31-120 days)

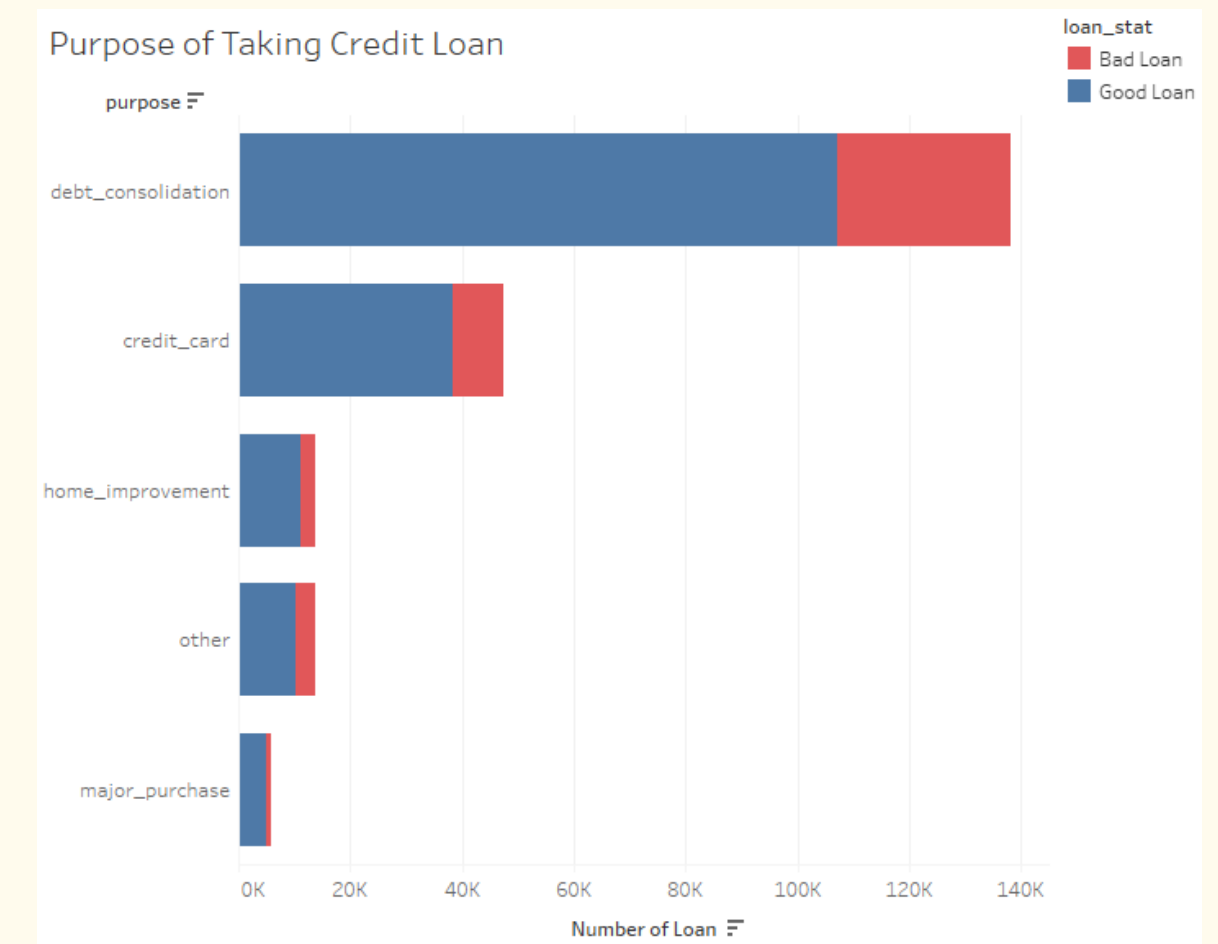


Visualization

How about classifying grade towards our borrower and the loan status?



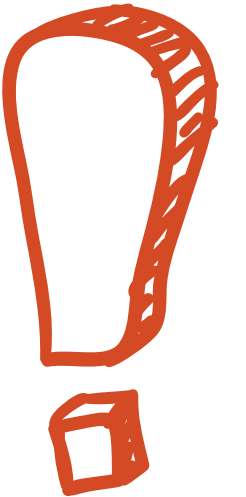
Why do our borrower take credit loan?



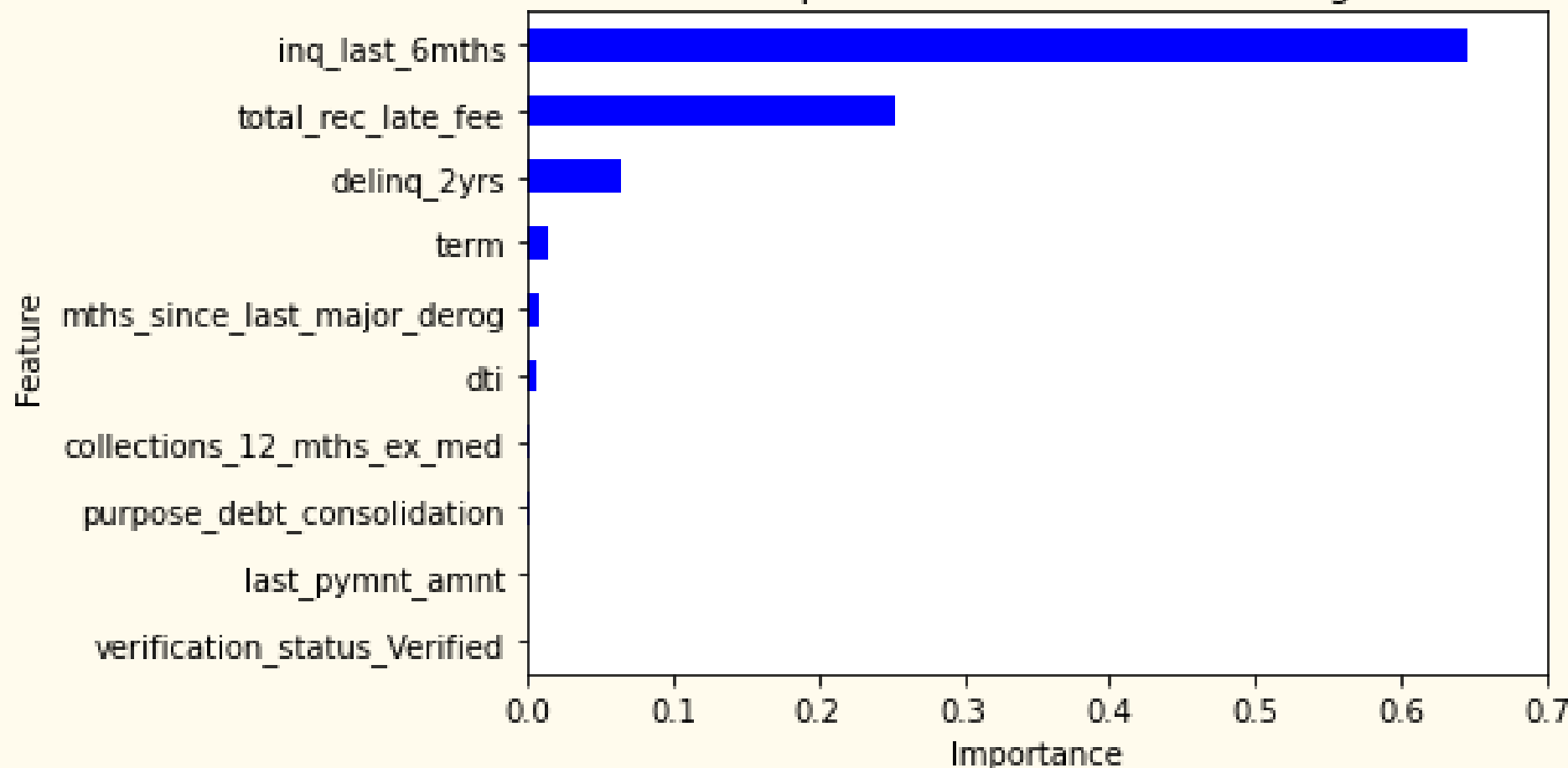
Modelling & Evaluation

70% Training & 30% Testing

- I used SMOTE for handling imbalanced class
- All steps are handled by Pipeline

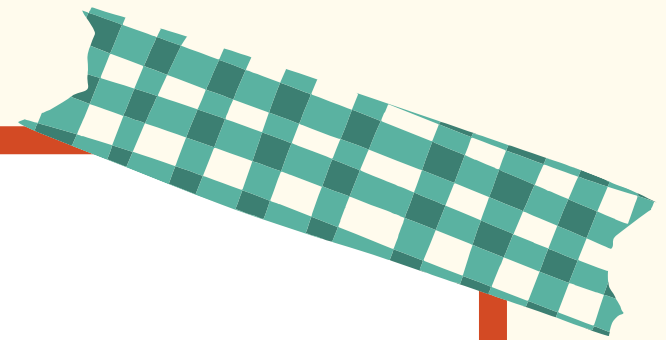


Feature Importance in Gradient Boosting Model



Evaluation Metrics:

- Main: False Negative (FN) & Recall from “0” (I minimized wrong predicted bad loan)
- Additional: ROC-AUC & KolmogorovSmirnov (KS)



Selected model: Gradient Boosting Trees

Model	FN	Recall	ROC-AUC	KS
Random Forest	608	96%	99.41%	94.40%
Gradient Boosting Trees	386	97%	99.48%	94.28%
XGBoost	447	97%	99.43%	93.83%
Voting Classifier	420	97%	99.48%	94.34%

