



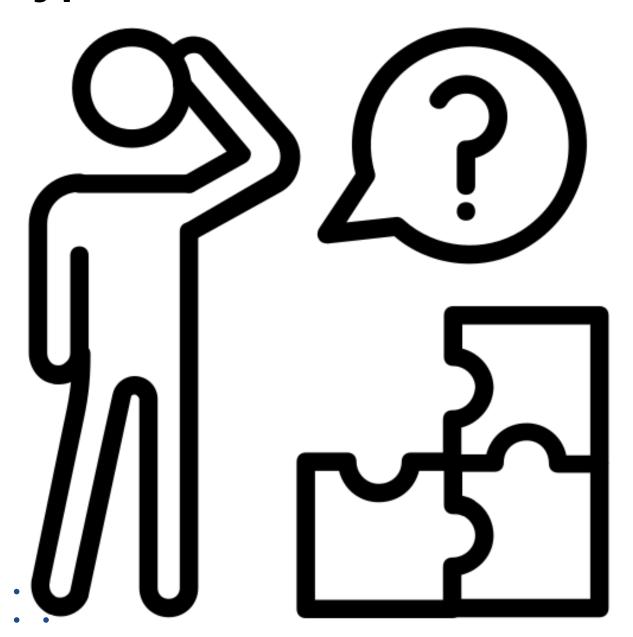
# Credit Risk Prediction

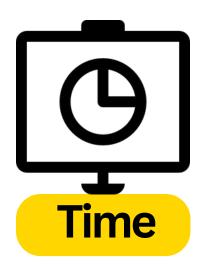
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## **Problems**

Credit risk refers to the risk of loss that a lender faces due to a borrower's failure to repay any type of loan or debt





It will takes a long time if we do the assessment manually



It can be happen if we make wrong decisions & will cause considerable losses



Credit loss is the amount of money lost by the lender when the applicant refuses to pay or runs away with the money owed.

## Goal, Business Matrix

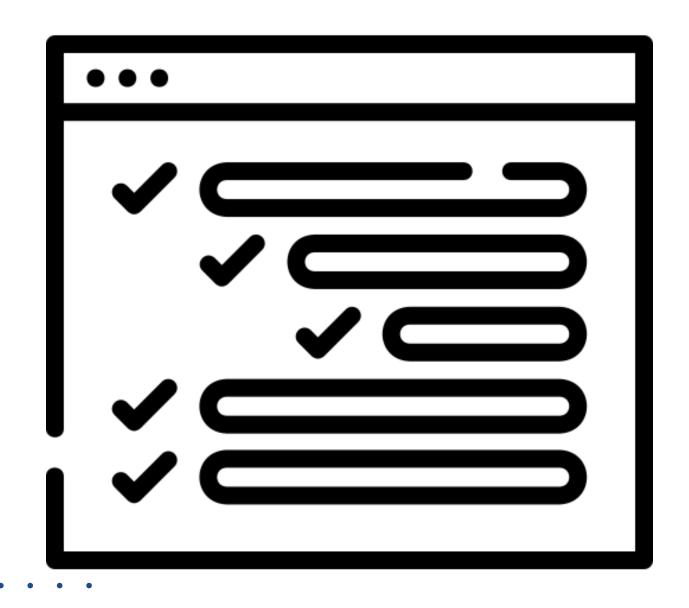




We aim to predict the secure loan before application is approved by implement machine learning algorithms. In addition, this model can be considered by lending companies to provide the amount of interest, the loan term and the loan period

**Loan Status** 

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## Customer Classification Status

#### **Good Borrower**

Fully paid

#### **Bad Borrower**

Charge Off, Default, Late (16-120 days)

## **Ambigus**

Current, In grace period

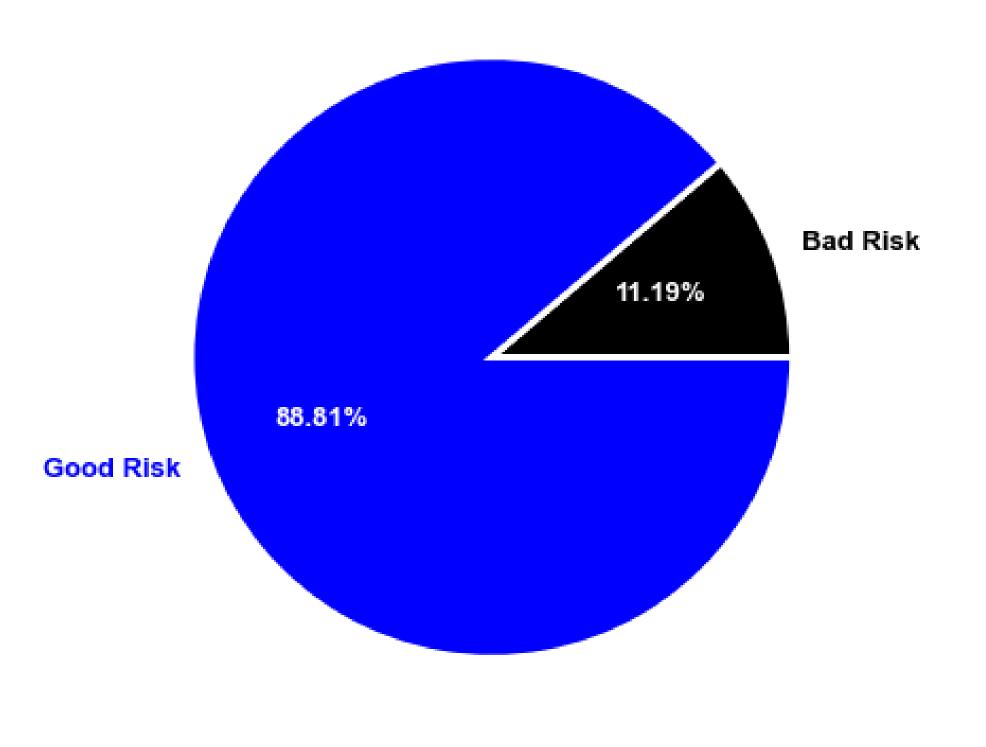


## The loan's problem

We get the imbalance data which have 88,81% good borrower and 11,19% bad borrower from 2007 until 2014

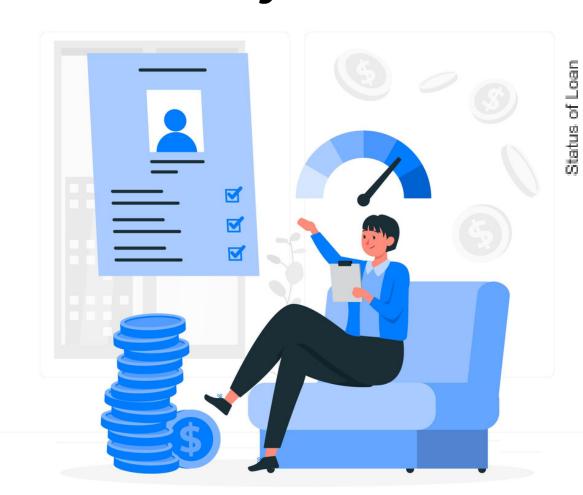


#### Loan Status

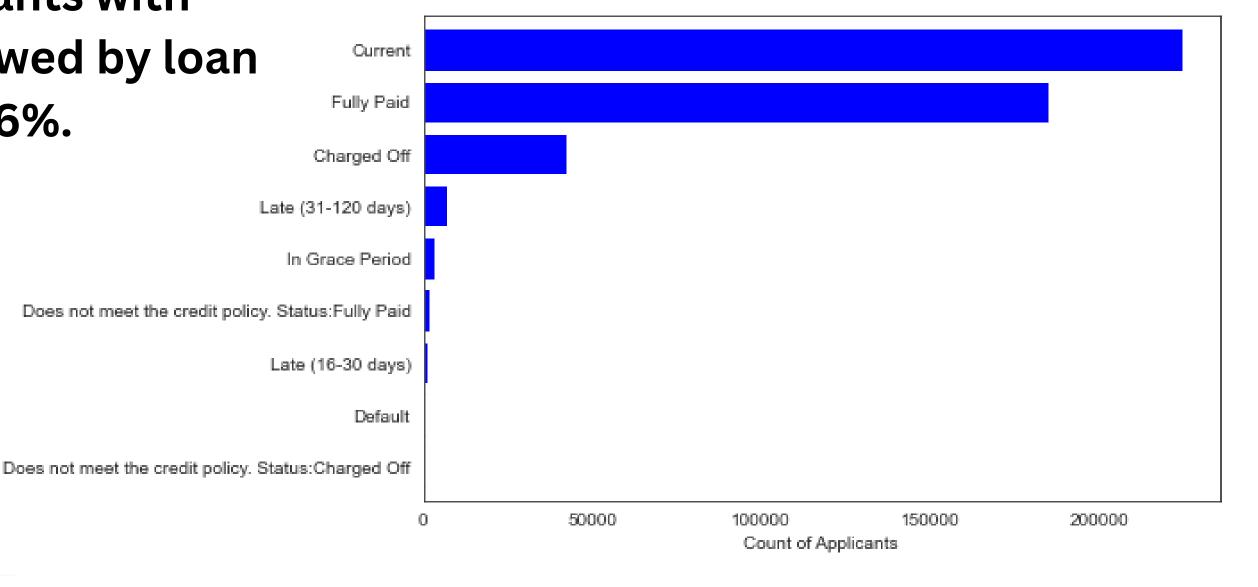


## The Number of Applicants by Loan Status

There are about 48% applicants with loan status of Current, followed by loan status of Fully Paid with 39.6%.



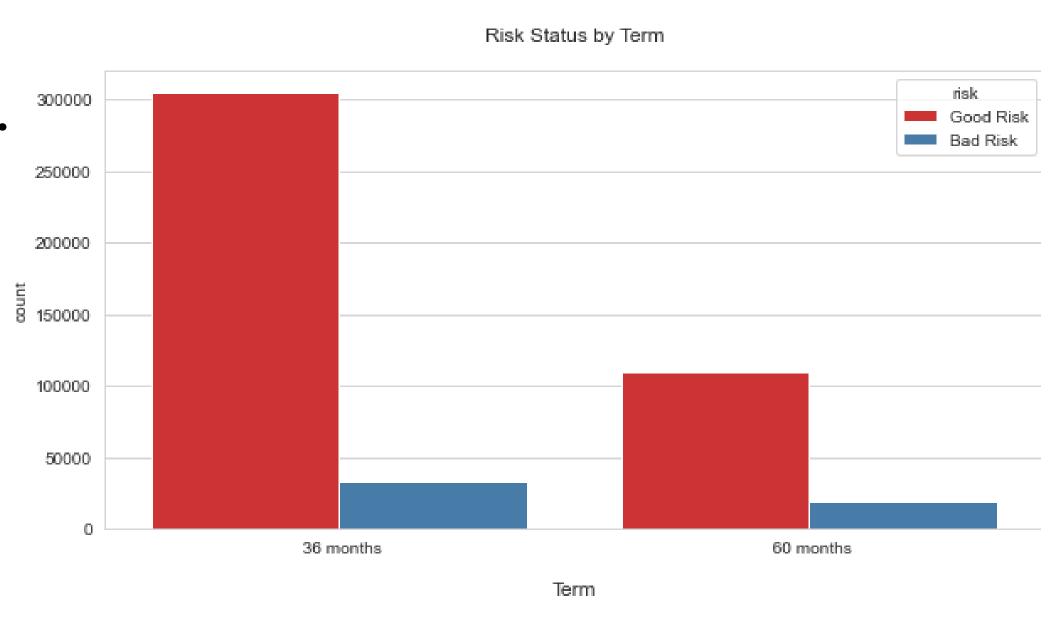
The Number of Applicants based on Loan Status





## Risk Status by Term

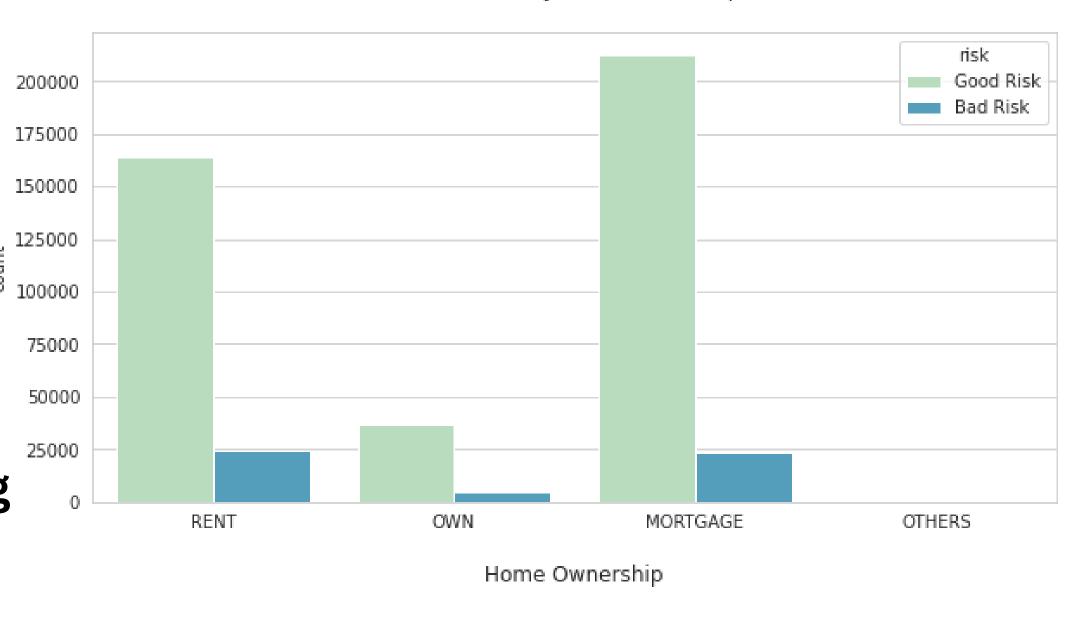
- Loan term tell us about the number of payments on the loan.
- There are only two types of loan terms, either 36 months or 60 months. Most of the loans (73%) are shorter, with a term of 36 months.
- Loans with 36 months period are almost twice as likely to bad risk as loans with 60 months period.



## Risk Status by Home Ownership

- The home ownership feature is category provided by the applicant's during registration.
- Most applicants have an existing 125000 mortgage (50%) or are currently 100000 renting a home (40%).
- Applicants who have an existing mortgage or are currently renting a home have a higher probability of bad risk.

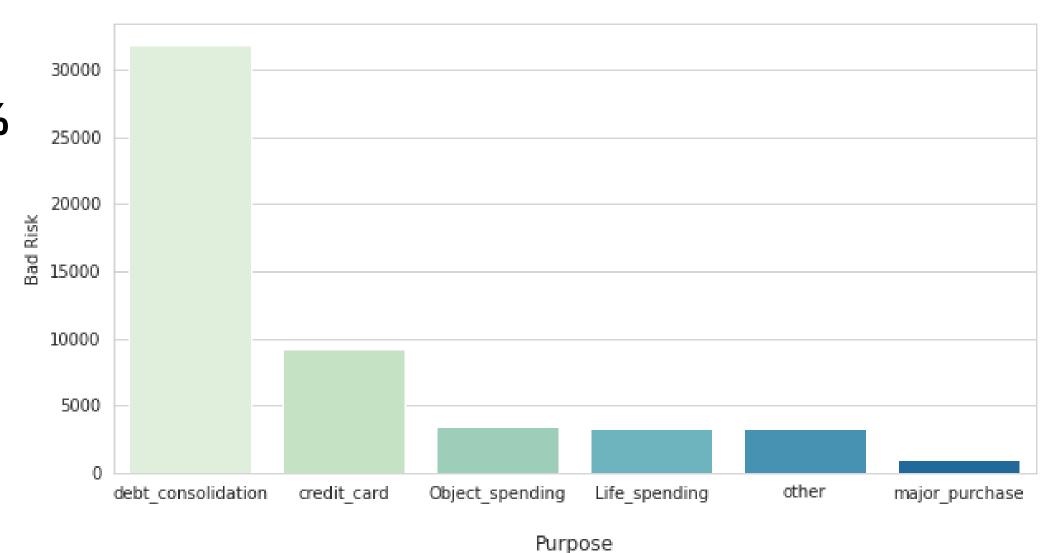
Risk Status by Home Ownership



## Loan Status by Purpose

- Most of Customer that type Bad
   Risk is in debt\_consolidation 6.8%
- For the second is credit\_card
- This shows that for debt reasons, the customer is more likely to be the type of customer that is Bad

#### Bad Risk by Purpose



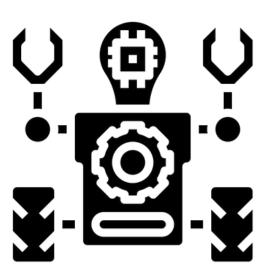


## Data Preprocessing



## **Data Cleansing**

- Drop feature with null value >50%
- Impute feature with missing value
   <20% with mode & median</li>
- Drop unnecessary feature & feature with 1 unique value
- Drop high correlation feature (>0.7)
  with target



## Feature Engineering

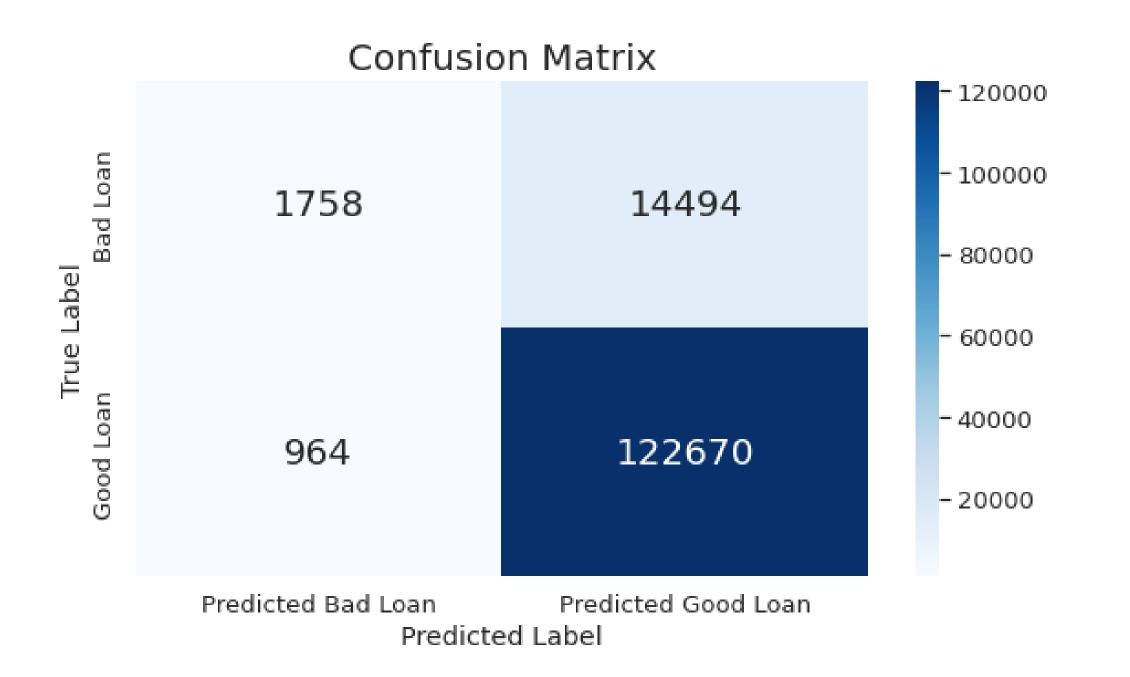
- Change datatype for some feature with incorrect datatype
- Feature selection using weight of evidence and information value
- Encode all features for modeling with label encoding and one hot encoding

## Modeling & Evaluation

Model	Accuracy	Recall	Precision	AUC
LightGBM	0.889	0.992	0.894	0.859
XGBoost	0.888	0.993	0.982	0.855
Gradient Boosting	0.888	0.992	0.983	0.855
RandomF orest	0.887	0.986	0.897	0.847
KNN	0.873	0.983	0.885	0.551

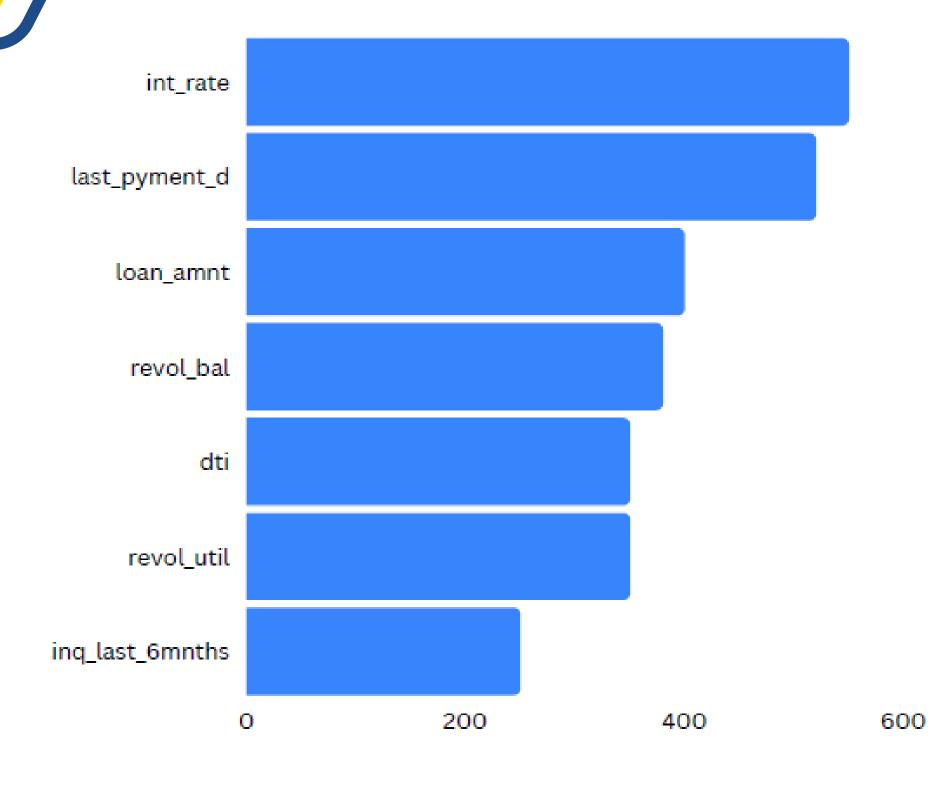
- Metrics evaluation that important for this model is AUC Score
- Best model is LightGBM with the best score & short duration

## Modeling & Evaluation





### **Feature Importence**



#### int\_rate

Interest Rate on the loan

#### revol\_bal

**Total credit revolving balance** 

#### loan\_amnt

The listed amount of the loan applied for by the borrower

#### revol\_util

Revolving line utilization rate, or the amount of credit the borrower is using relative to all available revolving credit.

#### last\_pyment\_d

Last month payment was received

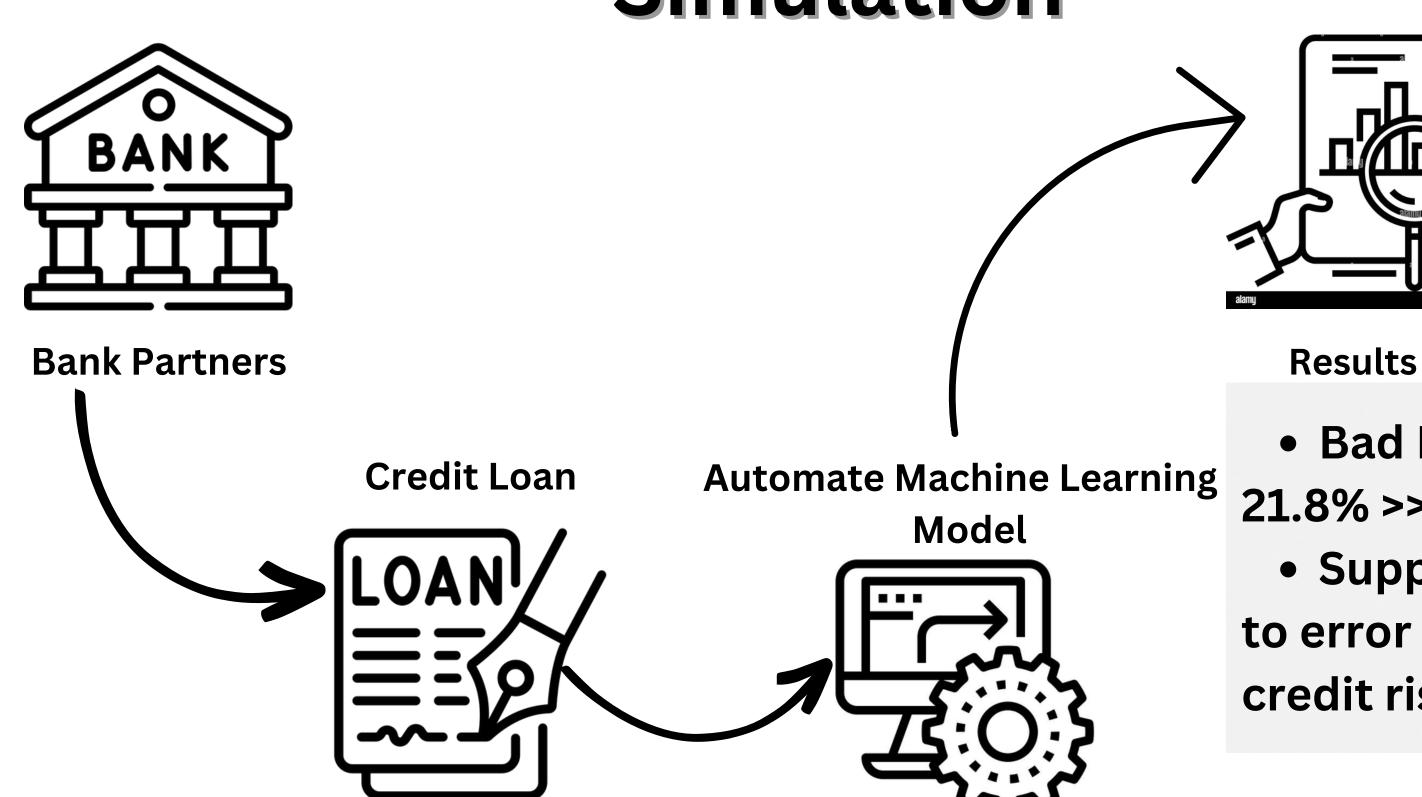
#### dti

A ratio calculated using the borrower's total monthly debt payments on the total debt obligations

#### inq\_last\_6mnths

The number of inquiries in past 6 months (excluding auto and mortgage inquiries)





Bad Loan decrease
 21.8% >> 2.4%

• Suppress the loss due to error determining credit risk

## **Business Recommendation**



### **Loan Amount**

It is important to carefully evaluate the borrower's creditworthiness and the potential risks associated with the loan amount to ensure that the loan is viable and sustainable for both the lender and the borrower. This may include conducting a thorough credit risk assessment, implementing a risk-based pricing strategy.

### **Feature Importence**

Focusing on feature importance could be a valuable business opportunity. By identifying the most important features and building a focused and effective model, this could help to reduce credit risk and make more informed lending decisions

#### **Interest Rate**

It would be beneficial to offer flexible interest rates that can be adjusted based on the borrower's creditworthiness and payment history. This allows for a more personalized approach to lending and can help the lender manage their credit risk more effectively. . .



## Thank you

Portfolio link

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