

# SUBMITTED BY GROUP 4

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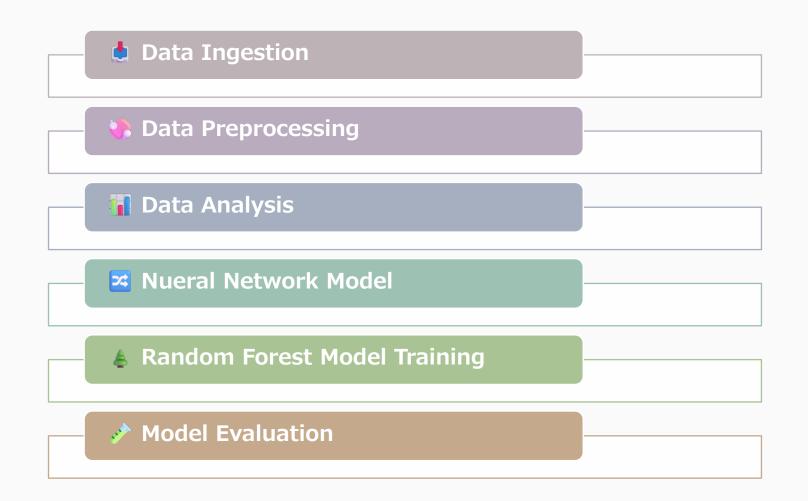
# Default Risk Debt-to-Income Ratio Credit Score



# Project Overview

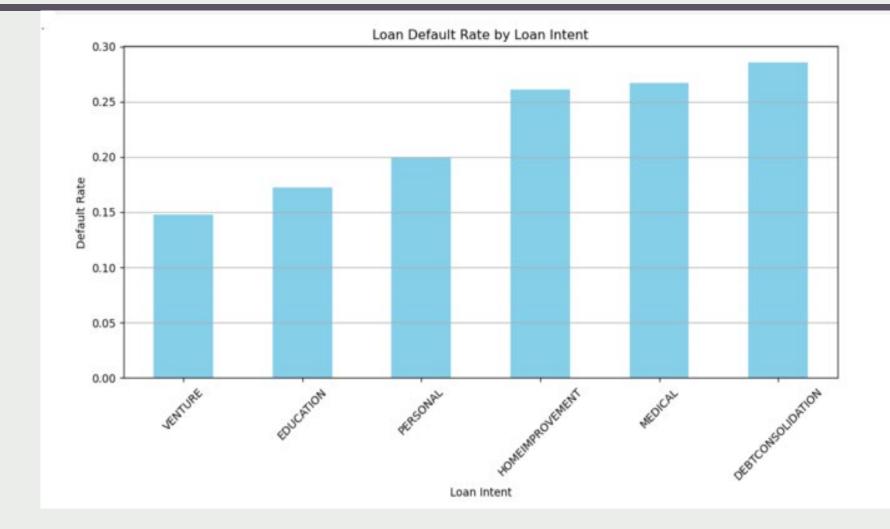
This project investigates the effectiveness of machine learning algorithms in predicting credit risk using dataset from Kaggle. The goal is to build a model that improves predictive accuracy while maintaining transparency and fairness, thereby supporting responsible and inclusive lending decisions.

# Workflow

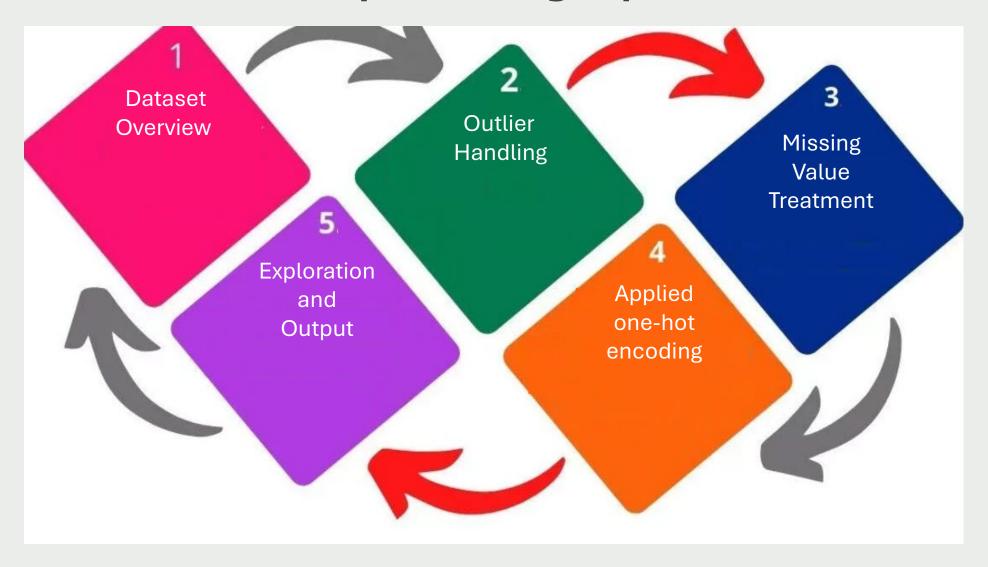


## **Key Factors**

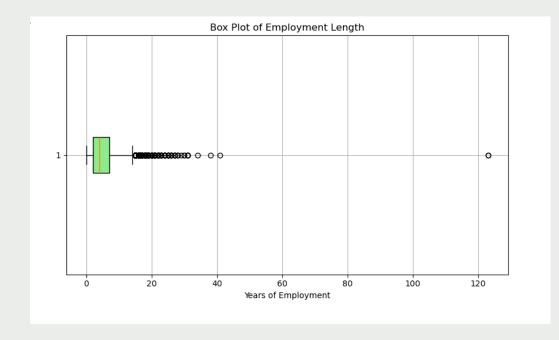
- Age
- Income
- Home Ownership
- Length of Employment
- Loan Intent
- Loan Grade
- Loan Amount
- Loan Interest Rate
- Loan Status
- Loan Percent Income
- Default on file
- Length of Credit History

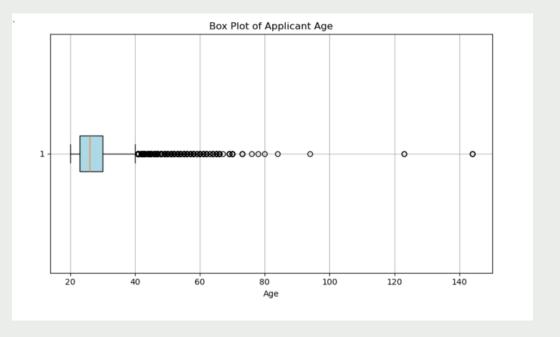


## **Data Preprocessing Pipeline**



### **Outliers**





# Model Development

#### Model 1: Baseline Model

Used the **Adam** optimizer

Simple architecture with 1–2 dense layers

Each Layer had 16 –lowed for **deeper networks** with 2–4 layers and up to **256 units** per layer. 128 units

#### Model 2: Enhanced Model

Introduced **AUC** as a secondary metric to better handle potential class imbalance.

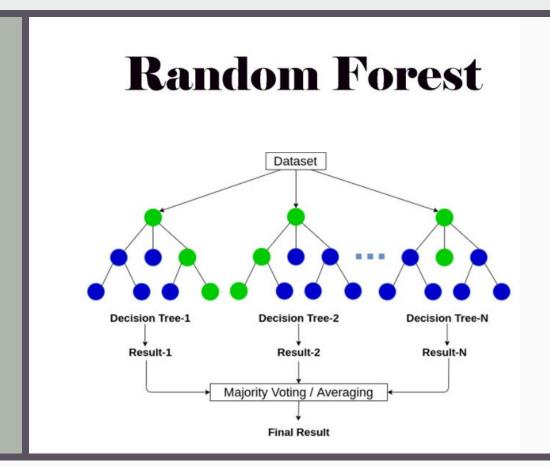
Has **deeper networks** with 2–4 layers and up to **256 units** per layer

### Model Performance

Both models achieved **85% accuracy** on test data.

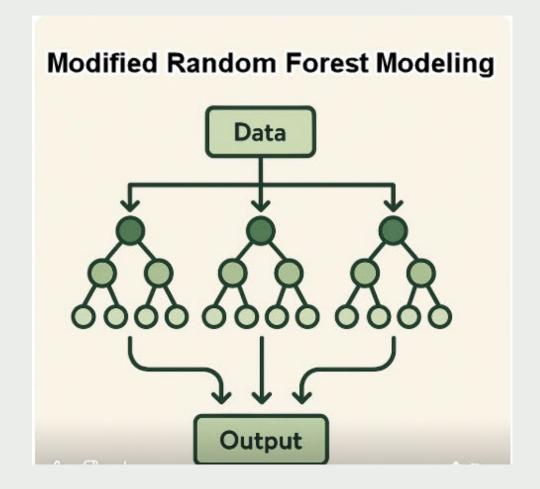
**Model 1** lacked overfitting protection and thus had lower generalizability.

Model 2 incorporated regularization and early stopping, making it more robust and production-ready.

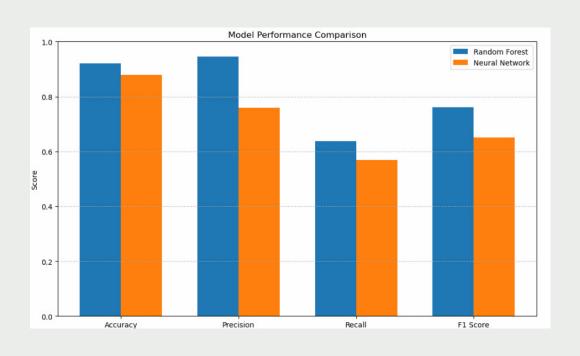


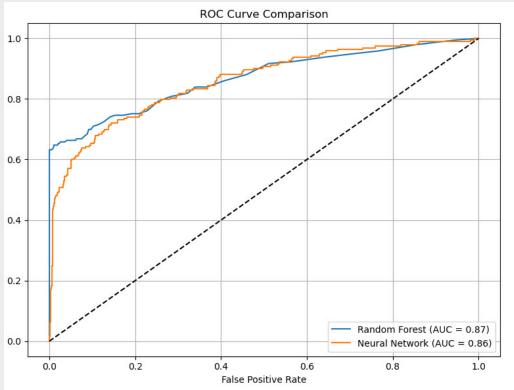
Random Forest Model Evaluation	Test 1	Test 2	Test 3
Accuracy	92.01%	92.32%	92.62%
Precision	91.89%	94.07%	98.59%
Recall	62.50%	65.46%	66.67%
F1 Score	74.40%	77.20%	79.55%

Random Forest Model Evaluation	Results 89.96%	
Accuracy		
Precision	86.84%	
Recall	62.86%	
F1 Score	72.93%	



### **Model Comparison Graph**





# THANK YOU

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