



# CREDIT CARD DEFAULT prediction: A Data-Driven Approach

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# AGENDA

what is the Credit Card  
Default System?

01

02

Objectives of the Analysis

Methodology and Visualisation

03

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Key Insights on Biases

Conclusions

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Recommendations



# Objective

The goal of this analysis is to predict the likelihood of credit card default by clients on their demographic and financial features. Additionally, we assess model fairness and interpretability to ensure transparency and ethical outcomes



# Metholodology

Data Preprocessing:

- Handled missing values.
- Scaled features ( eg: credit utilization)

Feature Engineering: Added derived features like CREDIT\_UTILIZATION, PAYMENT\_TO\_BILL\_RATIO

Balancing Dataset: Oversampled minority class to handle imbalance.

Model Training: Logistic Regression, Random Forest, XGBoost, Neural Networks

# Why these Models?

Logistic Regression: Simple,  
interpretable baseline

Random Forest: Handles complex patterns,  
reduces overfitting.

XGBoost: High performance with large  
datasets

Neural Networks: Superior accuracy for  
non - linear relationships

# models

## RANDOM FOREST

- ACCURACY: 81%
- PRECISION: 72%
- RECALL: 67%
- F1- SCORE: 80%

MODEL 2

## XGBOOST

- ACCURACY: 77%
- PRECISION: 78%
- RECALL: 77%
- F1- SCORE: 77%

MODEL 4

## LOGISTIC REGRESSION

- ACCURACY: 70%
- PRECISION: 76%
- RECALL: 70%
- F1- SCORE: 72%

MODEL 3

## NEURAL NETWORK

- ACCURACY: 77%

MODEL 1

# Bias ANALYSIS

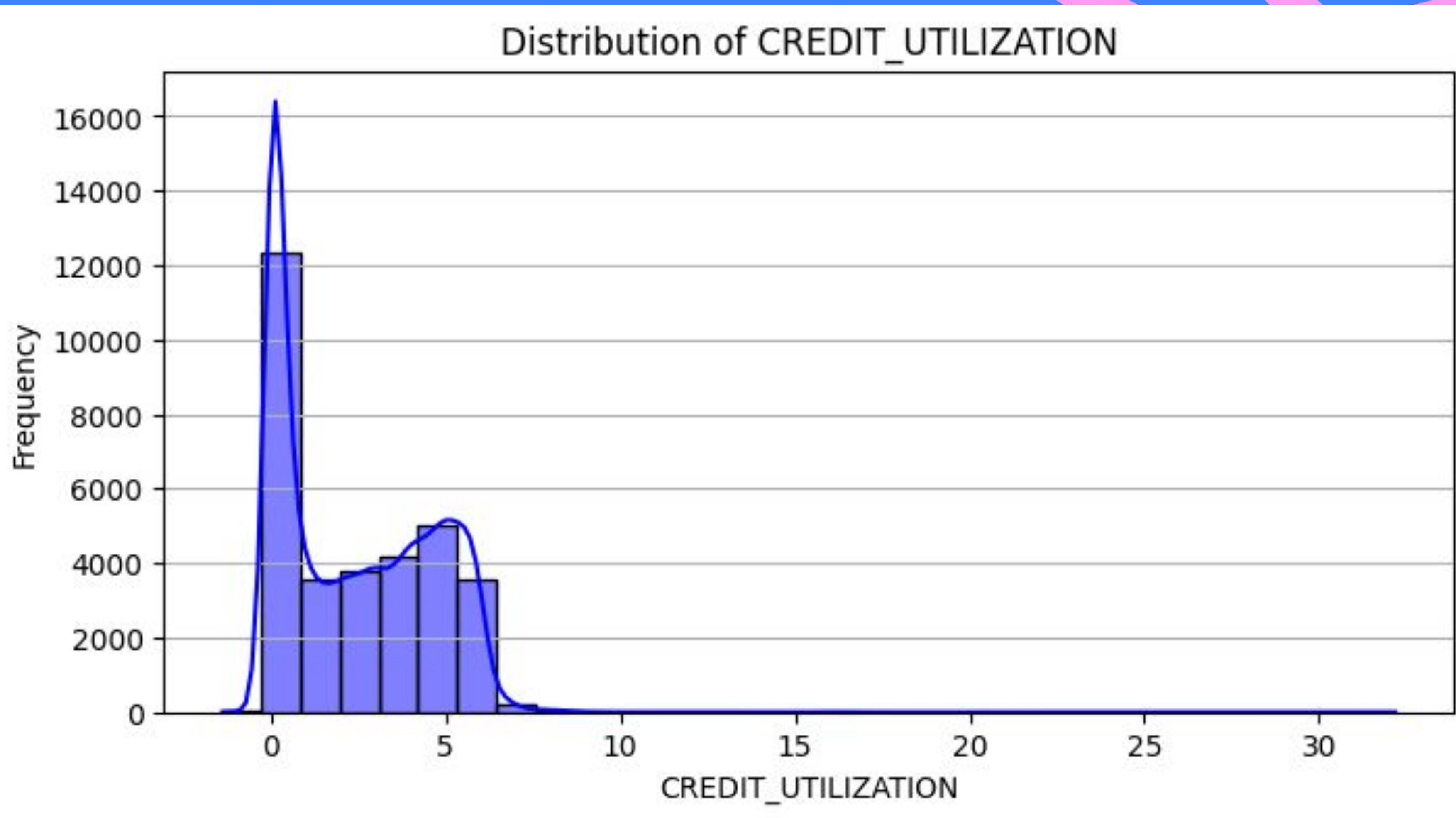
Demographic Factors: Disparities in predictions based on gender, education and age

Models may inherit societal biases from historical data

Example: Women predicted more likely to default due to systemic biases in credit systems

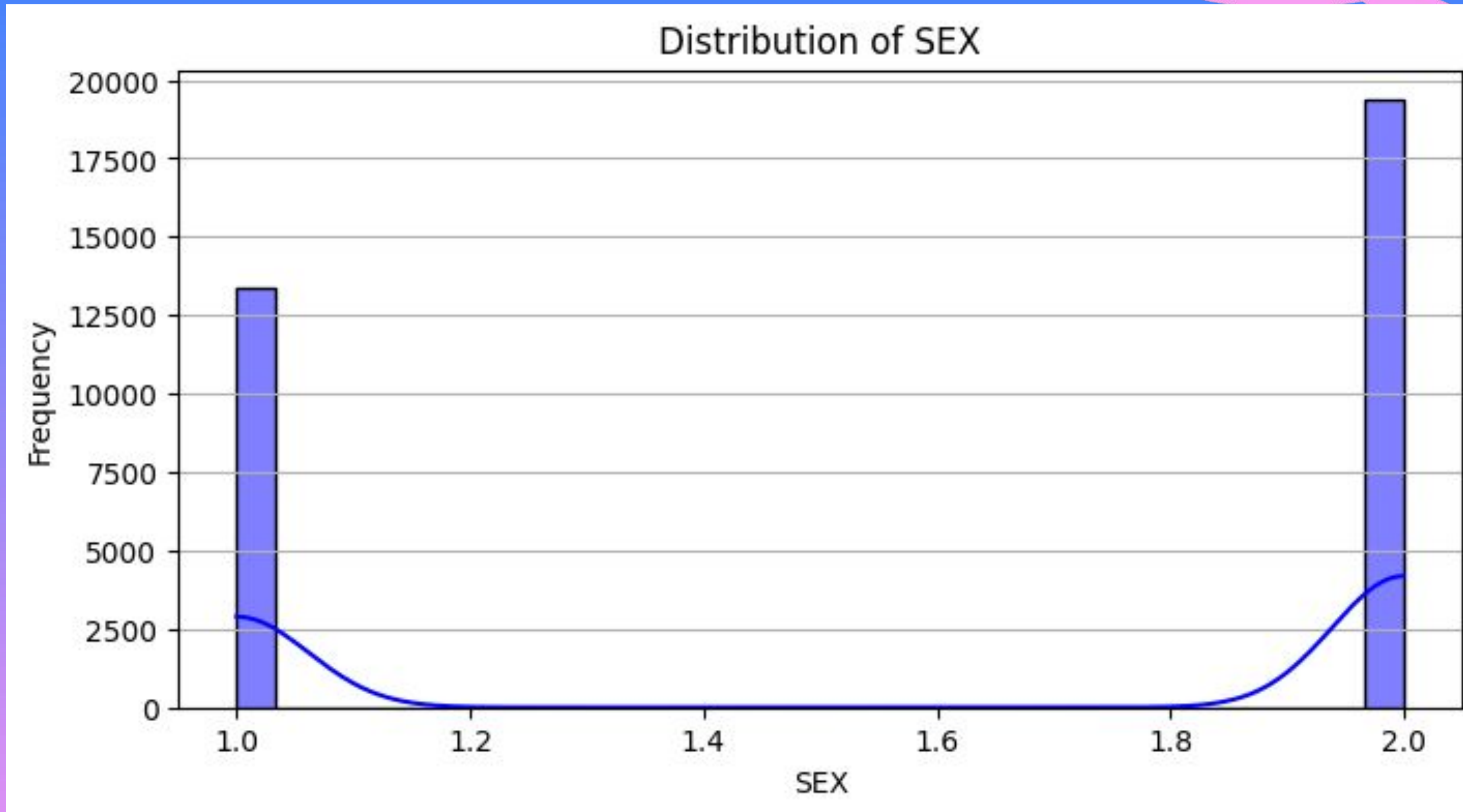


# Distribution of CREDIT UTILIZATION

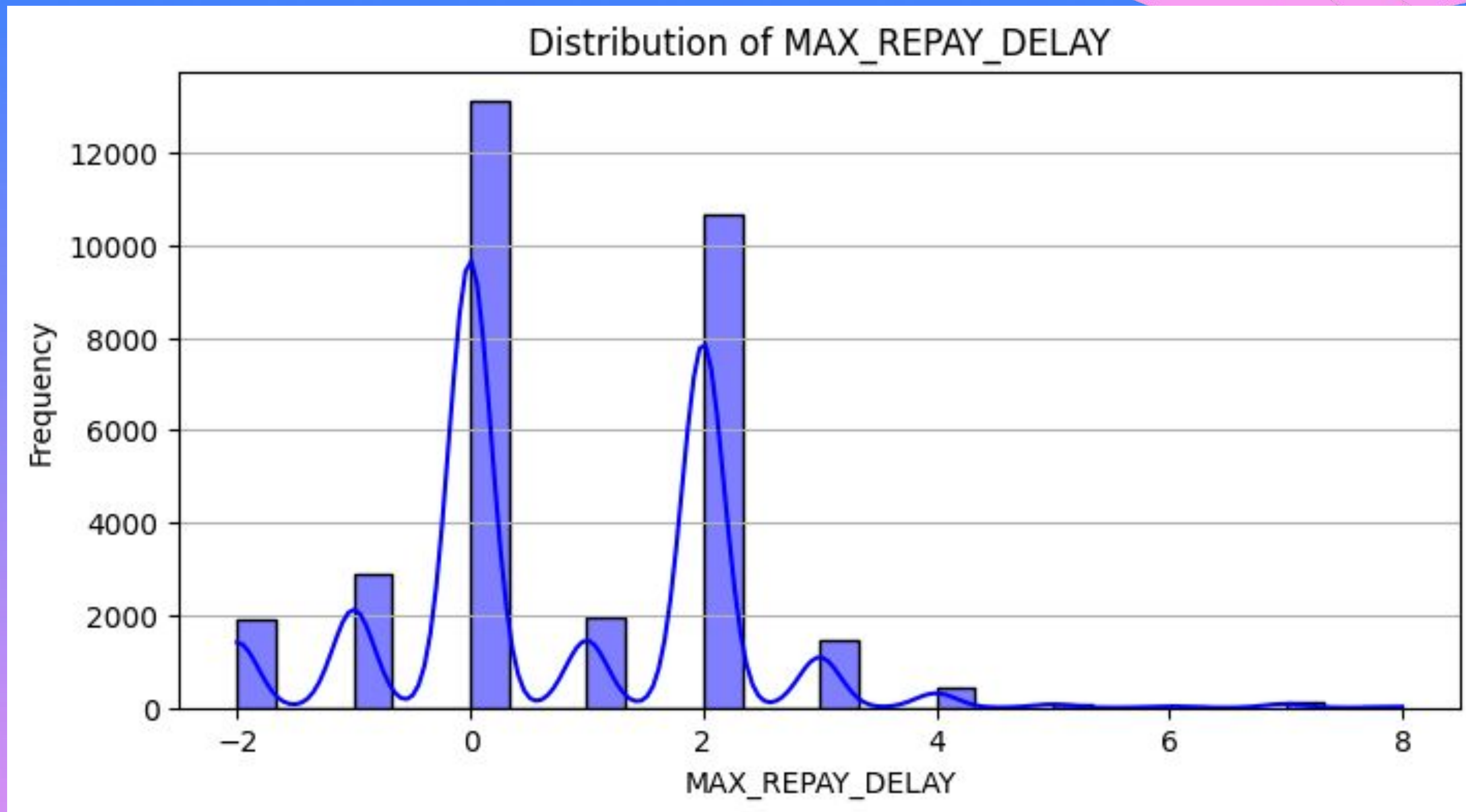




# Distribution of sex



# Distribution of max REPAY DELAY

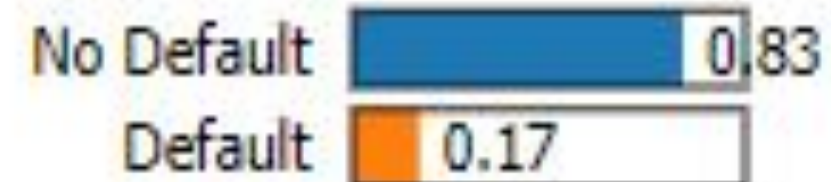




# LIME EXPLANATION FOR SAMPLE INSTANCE

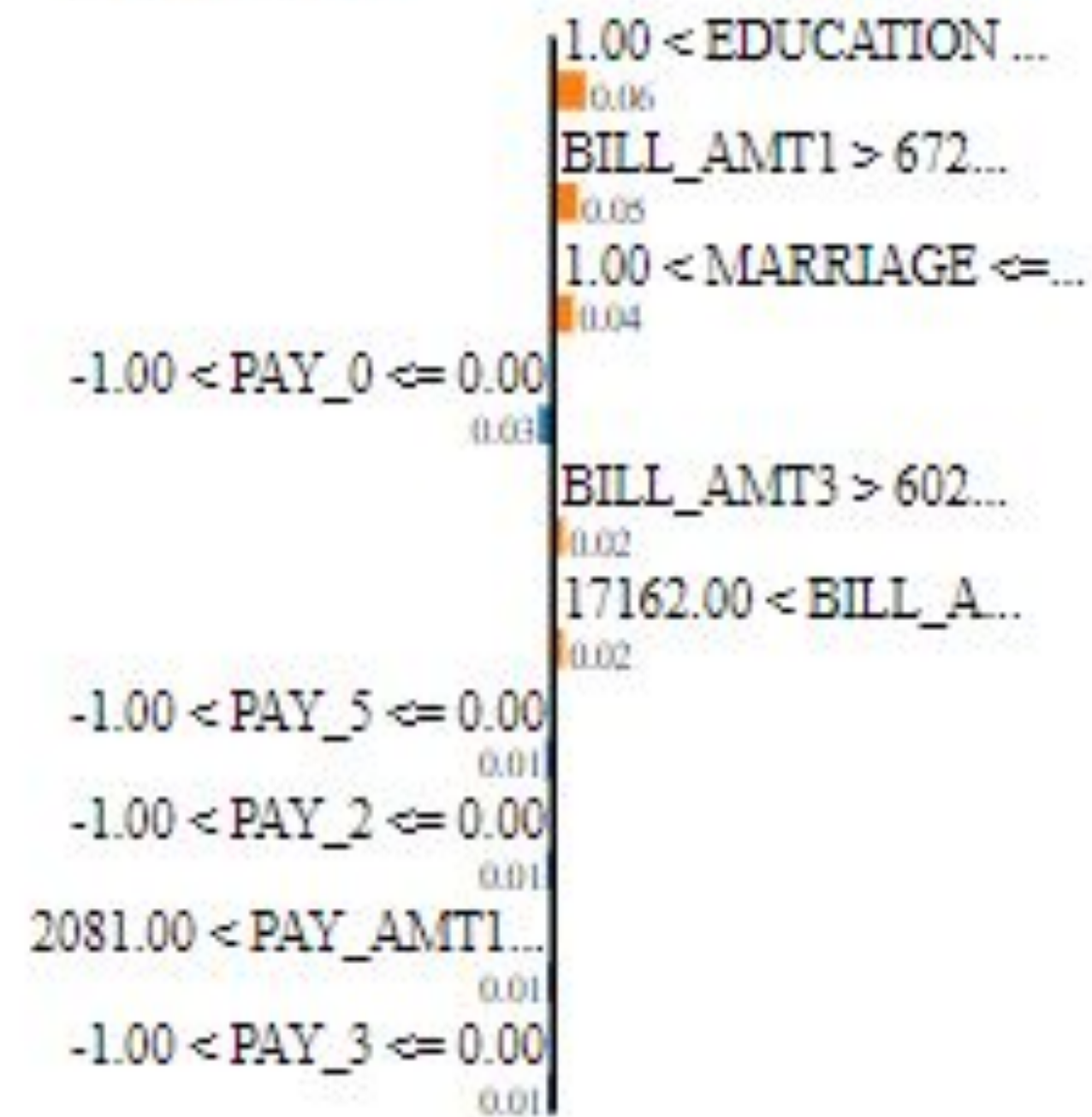
LIME Explanation for Sample Instance:

Prediction probabilities



No Default

Default



Feature	Value
EDUCATION	2.00
BILL_AMT1	101832.00
MARRIAGE	2.00
PAY_0	0.00
BILL_AMT3	84297.00
BILL_AMT6	44800.00
PAY_5	0.00
PAY_2	0.00
PAY_AMT1	5000.00
PAY_3	0.00

Fairness Metrics:

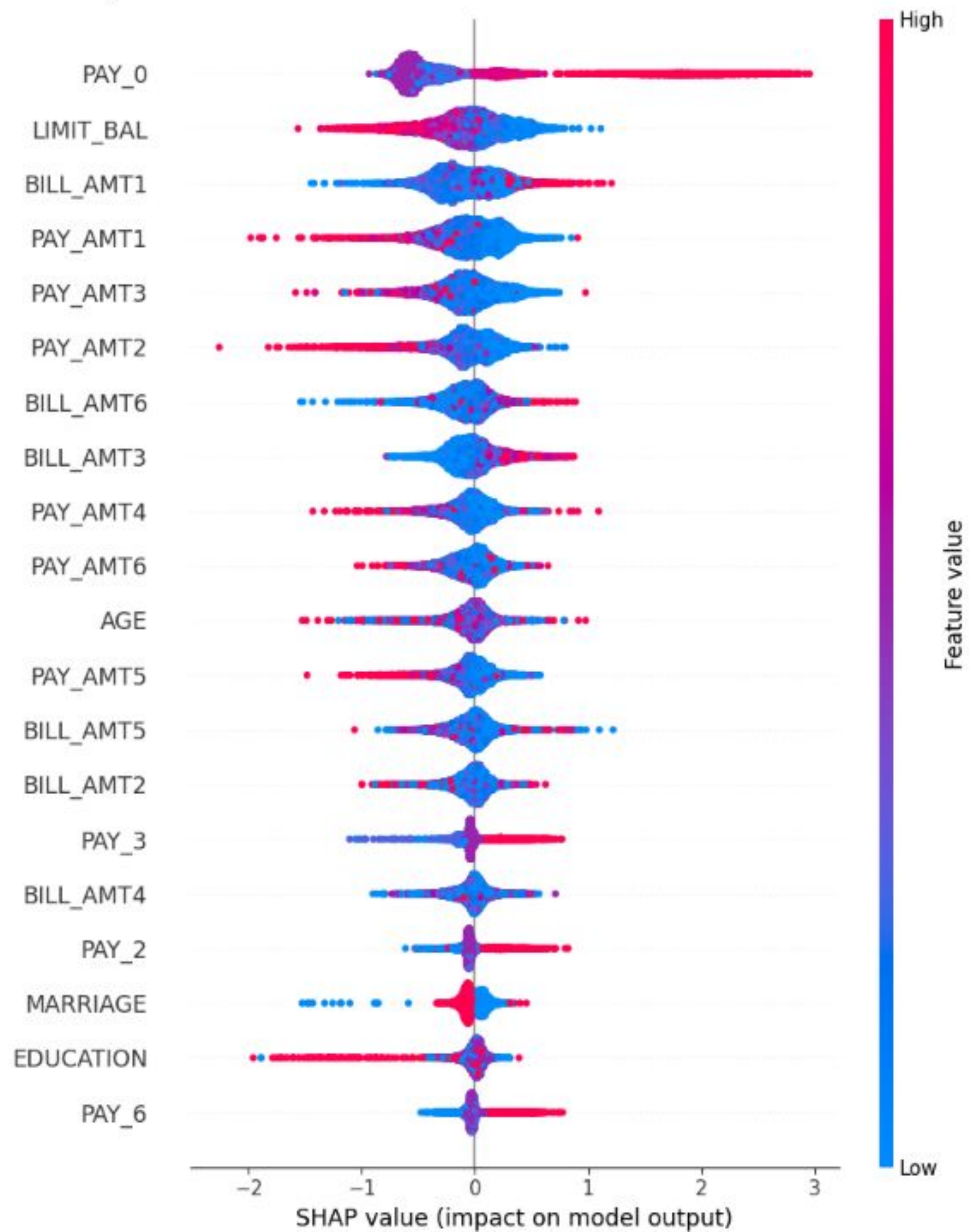
Demographic Parity Difference: 0.03

Equalized Odds Difference: 0.02





SHAP Summary Plot:



# SNAP SUMMARY PLOT





# CONCLUSIONS AND RECOMMENDATIONS