



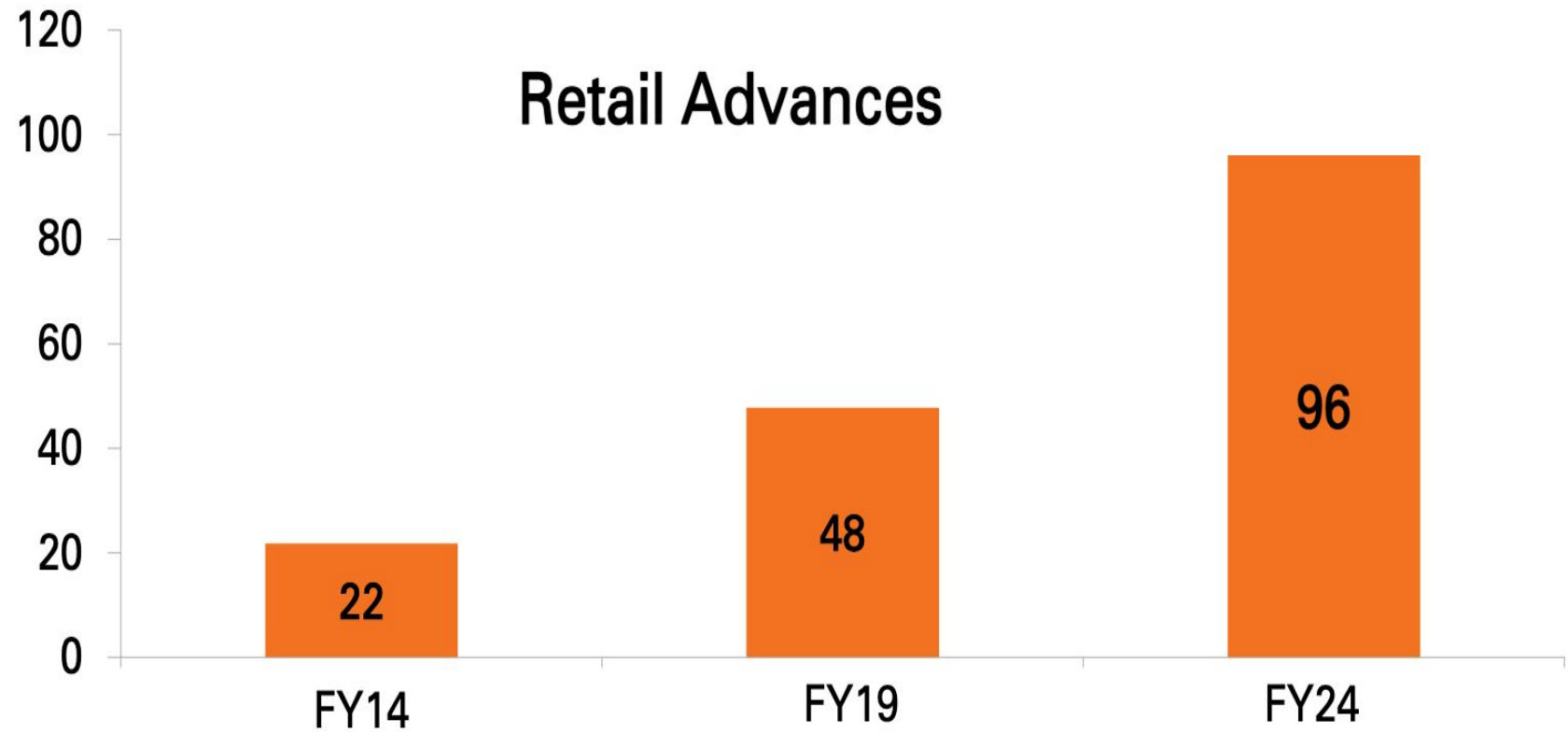
# PREDICTING CONSUMER LOAN DEFAULTS IN INDIA



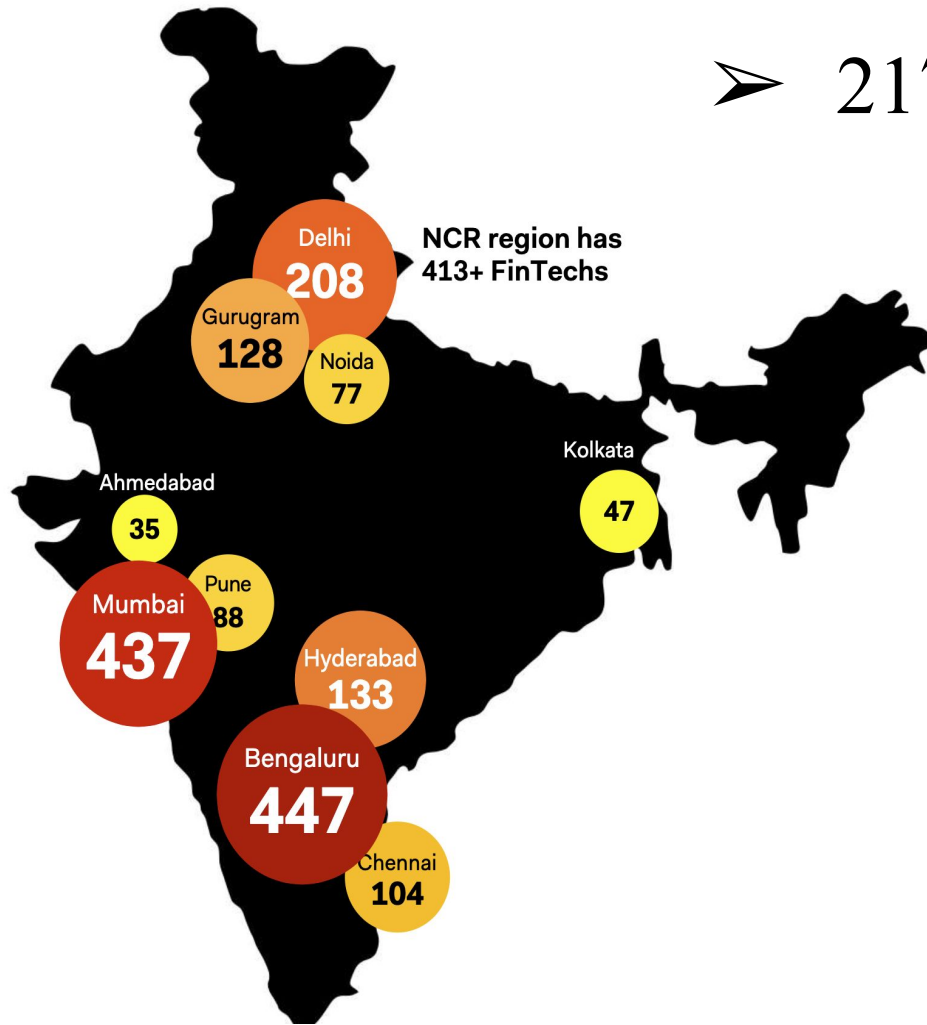


Rs. Trillion

## Retail Advances



## ➤ 2174 Fin-tech Start-ups



### TYPES

- Online Marketplaces
- Peer-to-Peer Lending
- Instant Loans
- Non-Bank Financing Corps
- Micro-Loan Crowd Funding

# DATA



```
graph LR; DATA[DATA] --> B1[University of Alabama<br/>Kaggle]; DATA --> G1[Observations 252,000]; DATA --> B2[Features: 10<br/>Quantitative: 5<br/>Categorical: 5]; DATA --> G2[Data Cleaning/EDA<br/>pandas Python];
```

- University of Alabama
- Kaggle

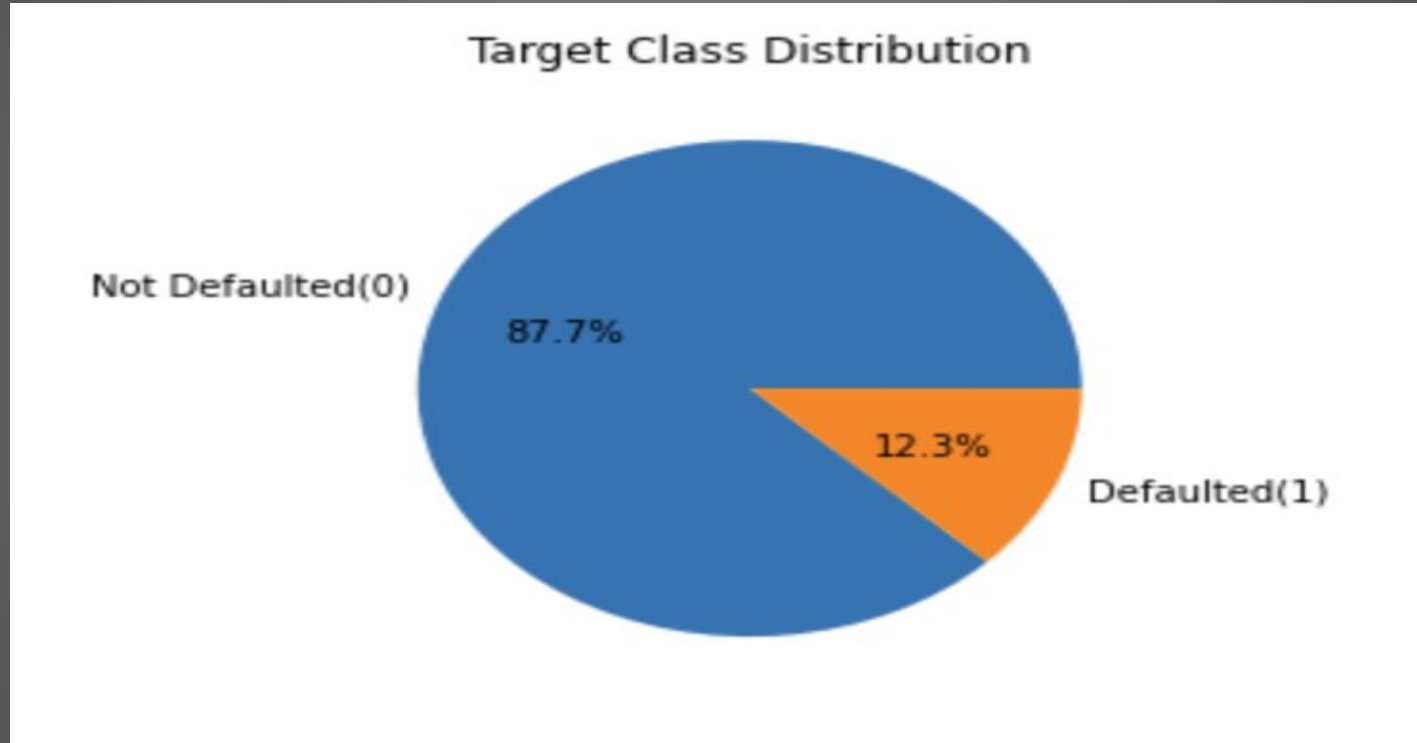
- Observations 252,000

- Features: 10
  - Quantitative: 5
  - Categorical: 5

- Data Cleaning/EDA



# CLASS IMBALANCE





# EVALUATION METRICS

- False Negatives(Recall)
  - Don't approve those highly likely to default
    - Principle is a priority
    - Low opportunity cost of False Positive(Precision)
- F-Beta Score
  - Scale: 0 to 1
  - Beta Hyperparameter: Change relative importance of precision v. recall



# Modeling Choices





# RANDOM FOREST

Bootstrapping

Decreases Variance/Limited Bias Increase

Corrects Overfitting

Decorrelate Trees/Feature Subsampling

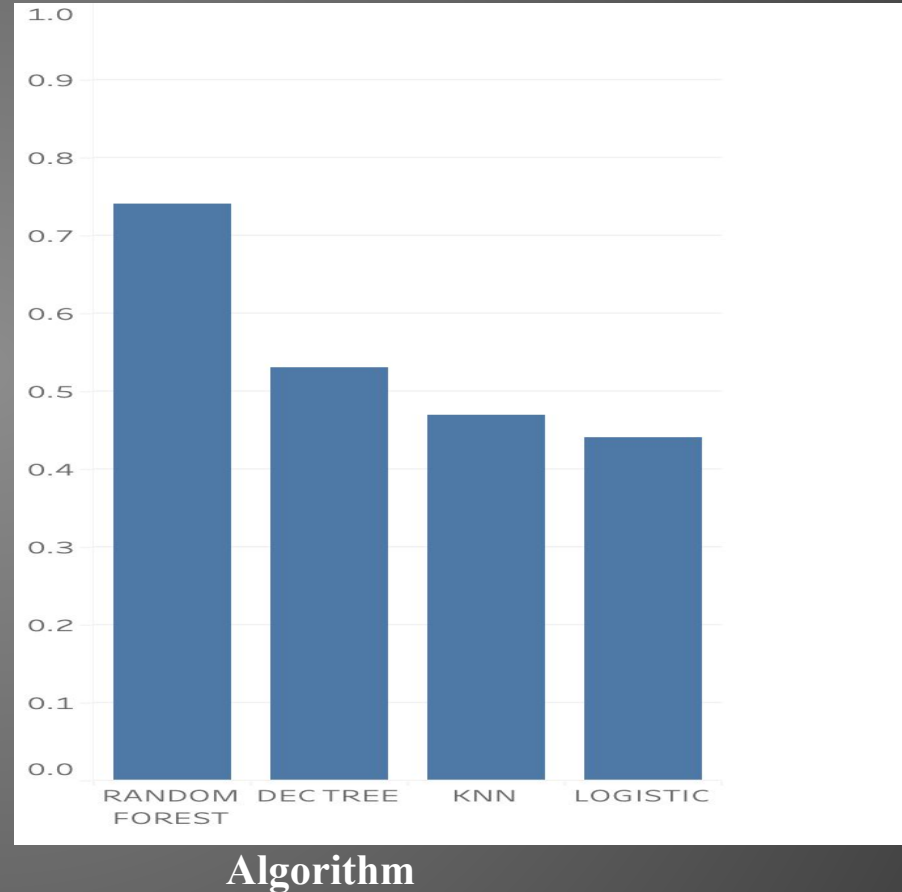
# COMPARATIVE PERFORMANCE

F-Beta Score

0.74

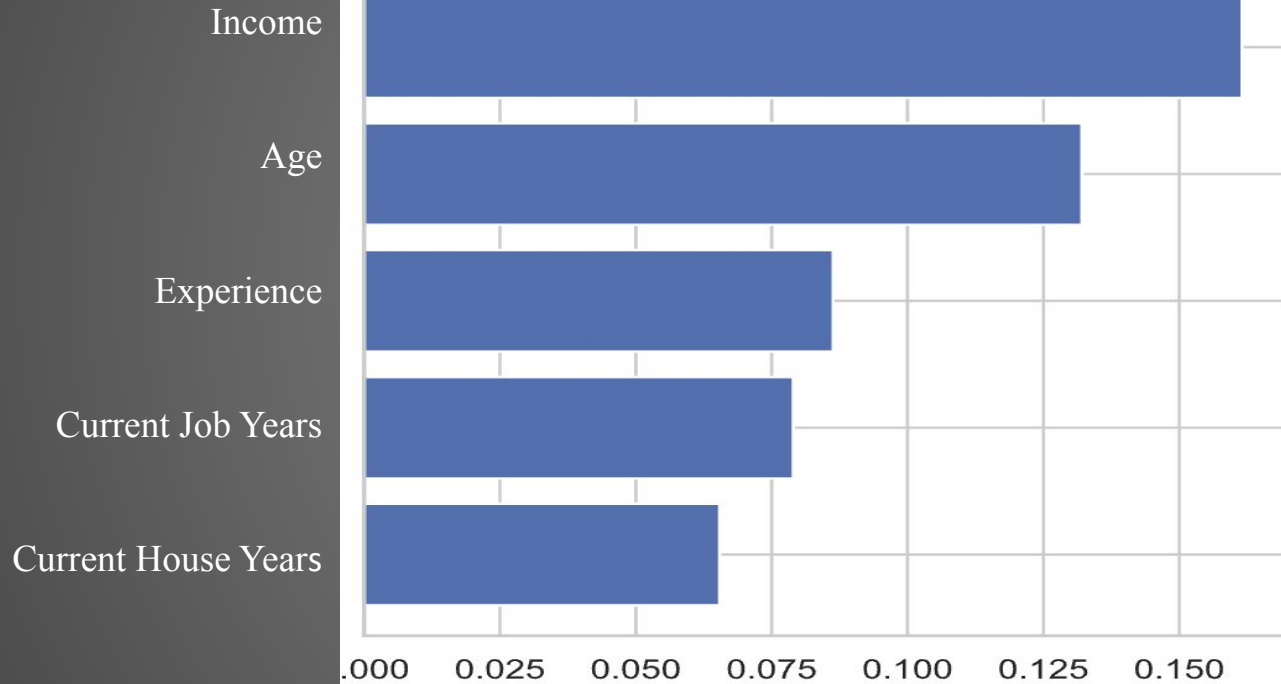
F-Beta  
Score

Beta = 3



# Feature Importance

Features



Importance

# Future Work

- Add Rural v. Urban Feature
- Include dollar value of loans
- XGBoost

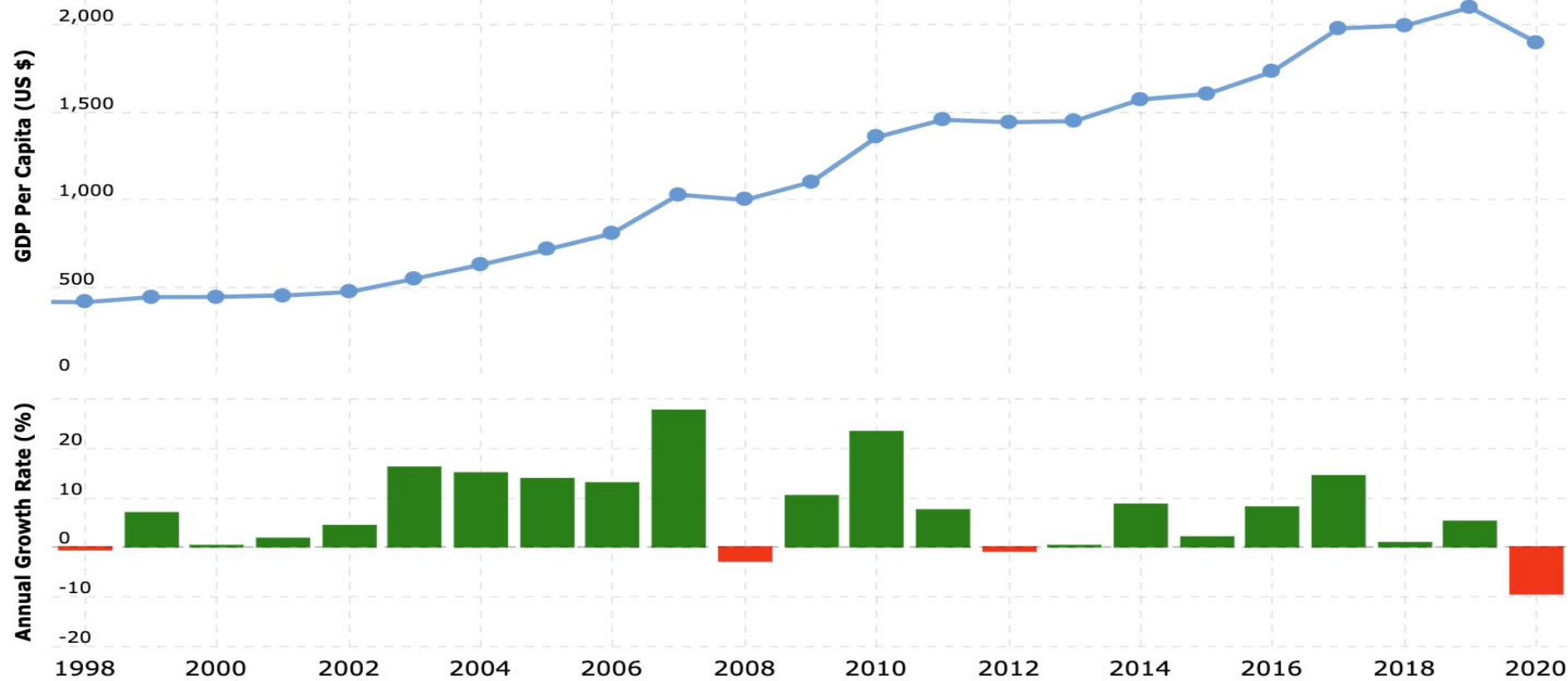


Thank you!



# APPENDIX

# Indian Per Capita GDP GROWTH





# PAIR/HUE PLOT

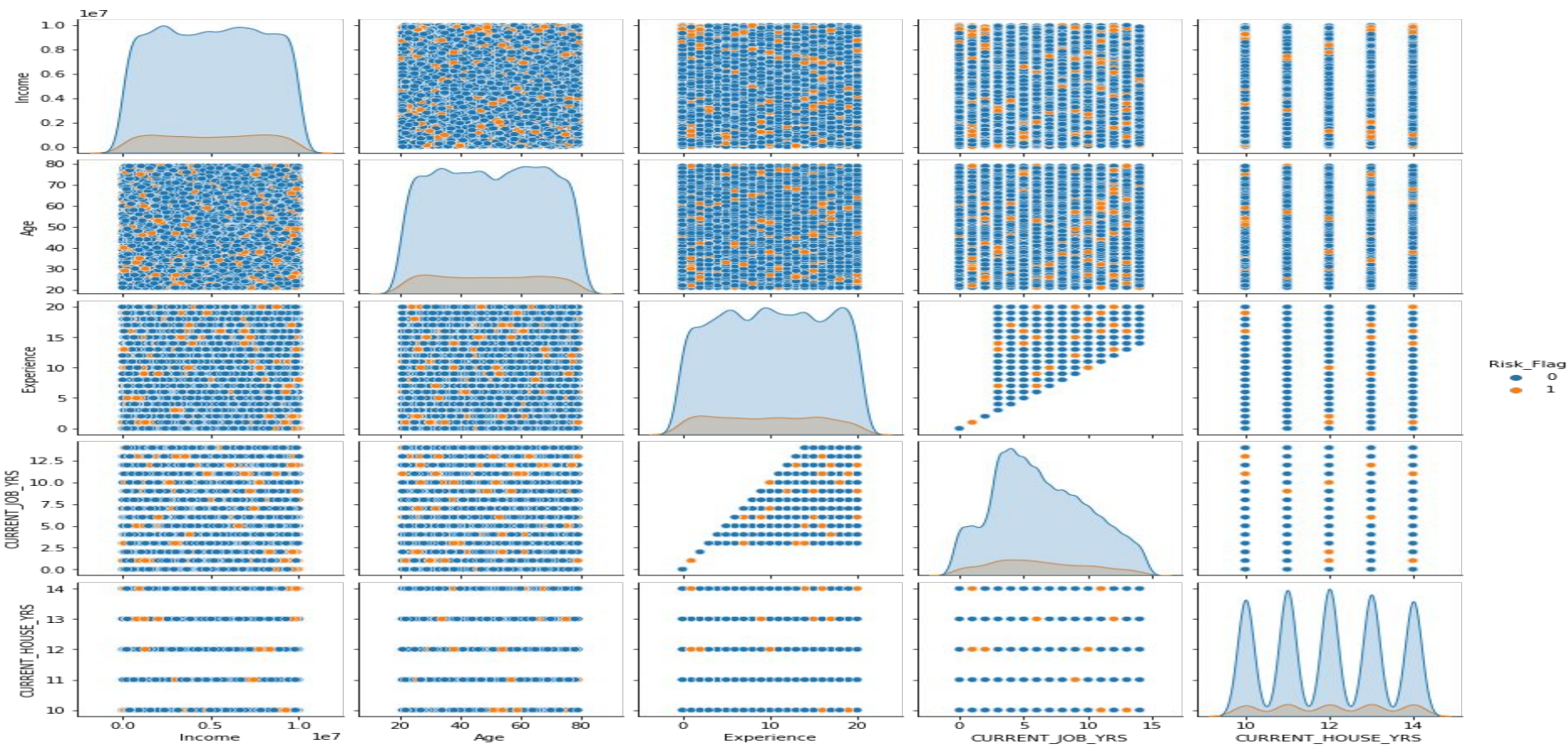


Figure 1. Retail financial services industry loans - India

