Lighthouse Bank Loan Program Evaluation

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Project Goals



Who are the loan applicants?

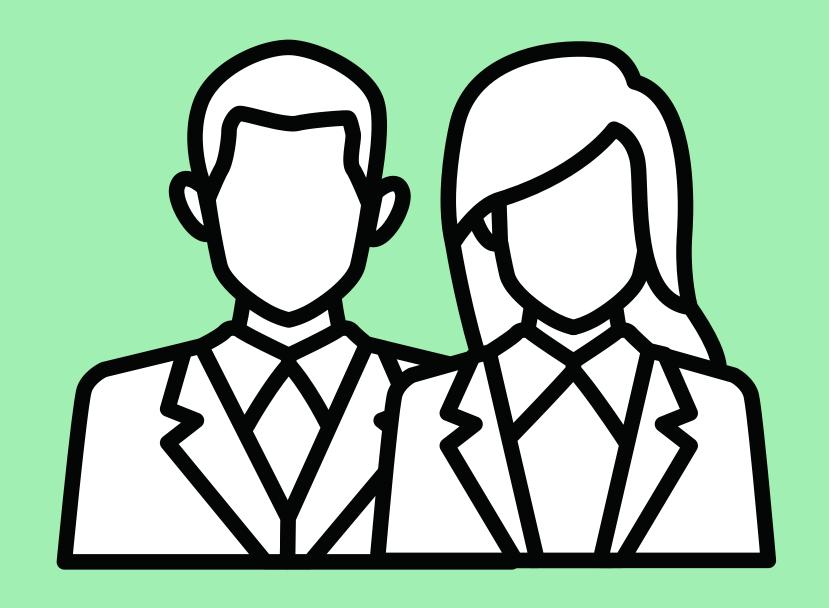
Which applicants get loans?

Can we predict who Lighthouse gives loans?

What determines loan eligibility?

What opportunities are there to reach more customers?

Who are the loan applicants?



male

married

no dependents

highly educated

credit experience



Applicant Income (Median)

\$3,800

Coapplicant Income (Median)

\$1,000

Total Income (Median)

\$5,300

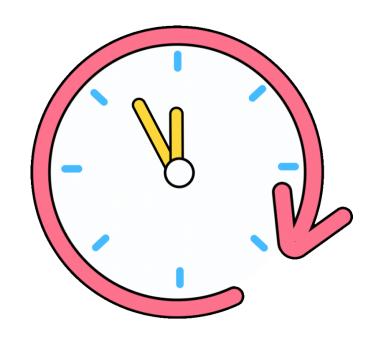
What is the most common loan type?





Loan Amount (Mean)

\$ 140,000

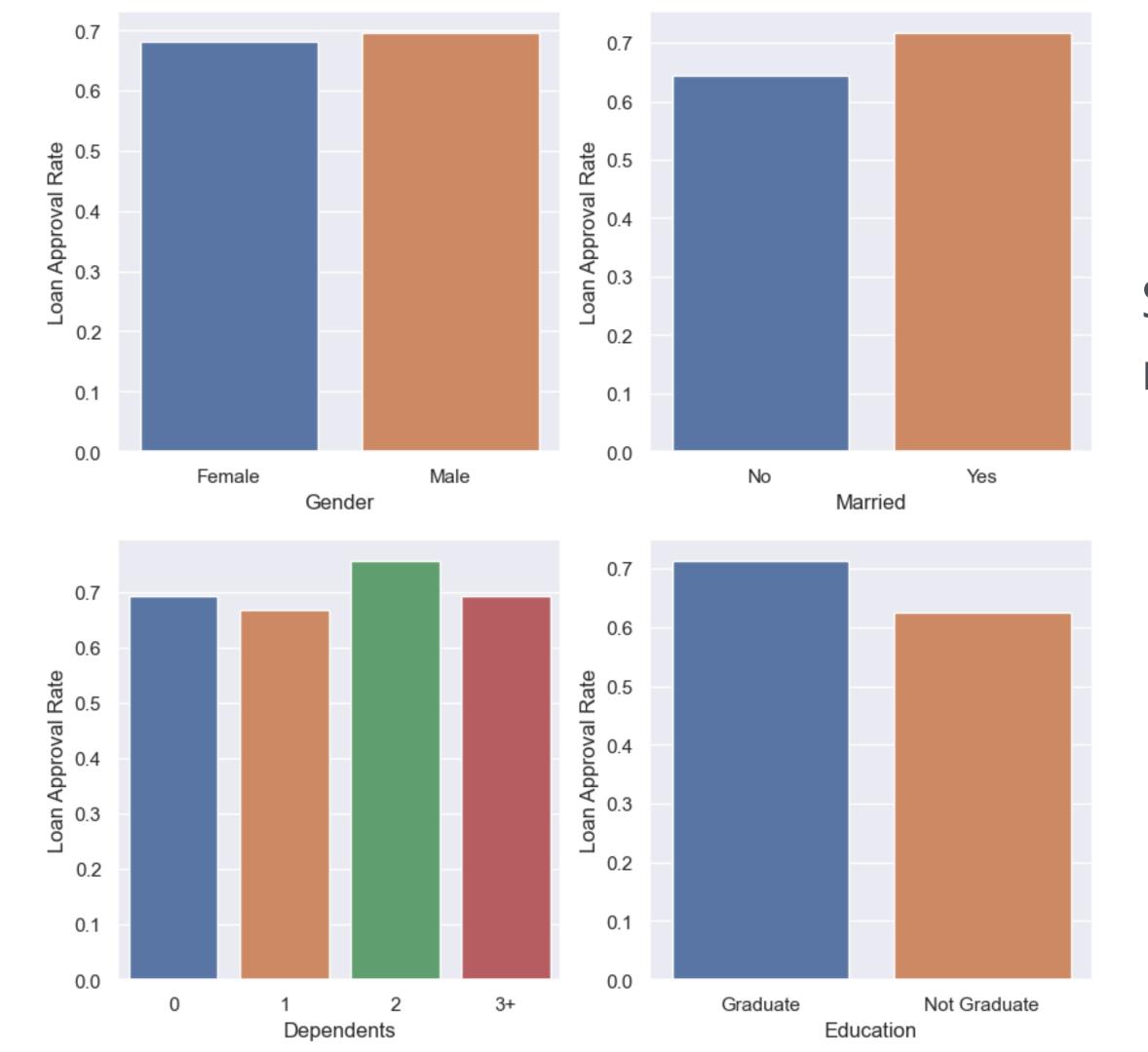


Loan Term (Mean)

30 years

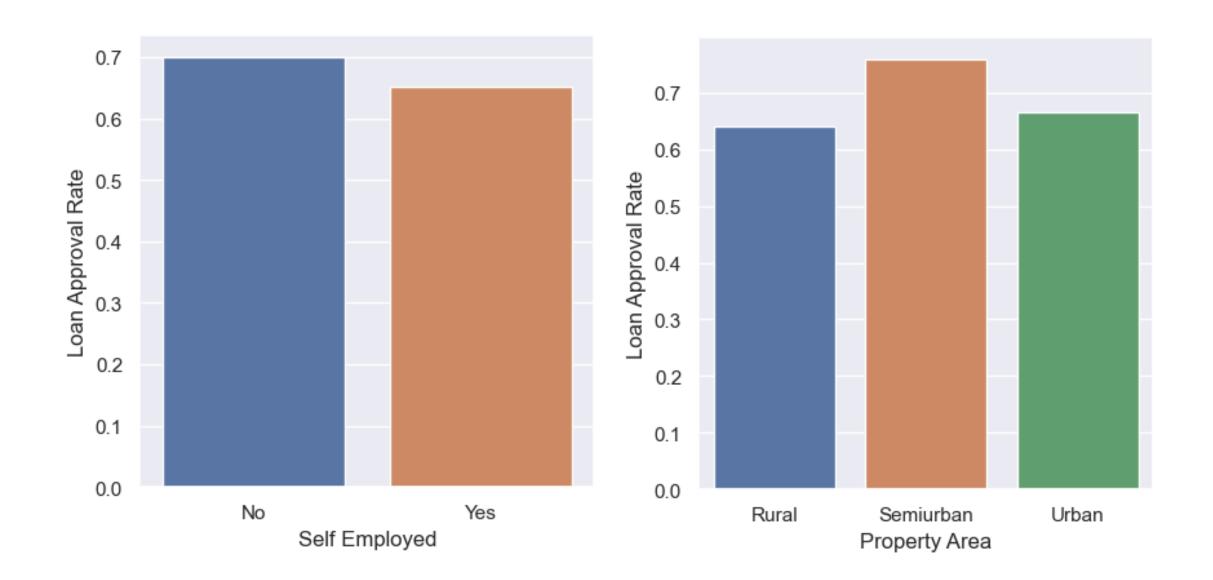
Which applicants get loans?





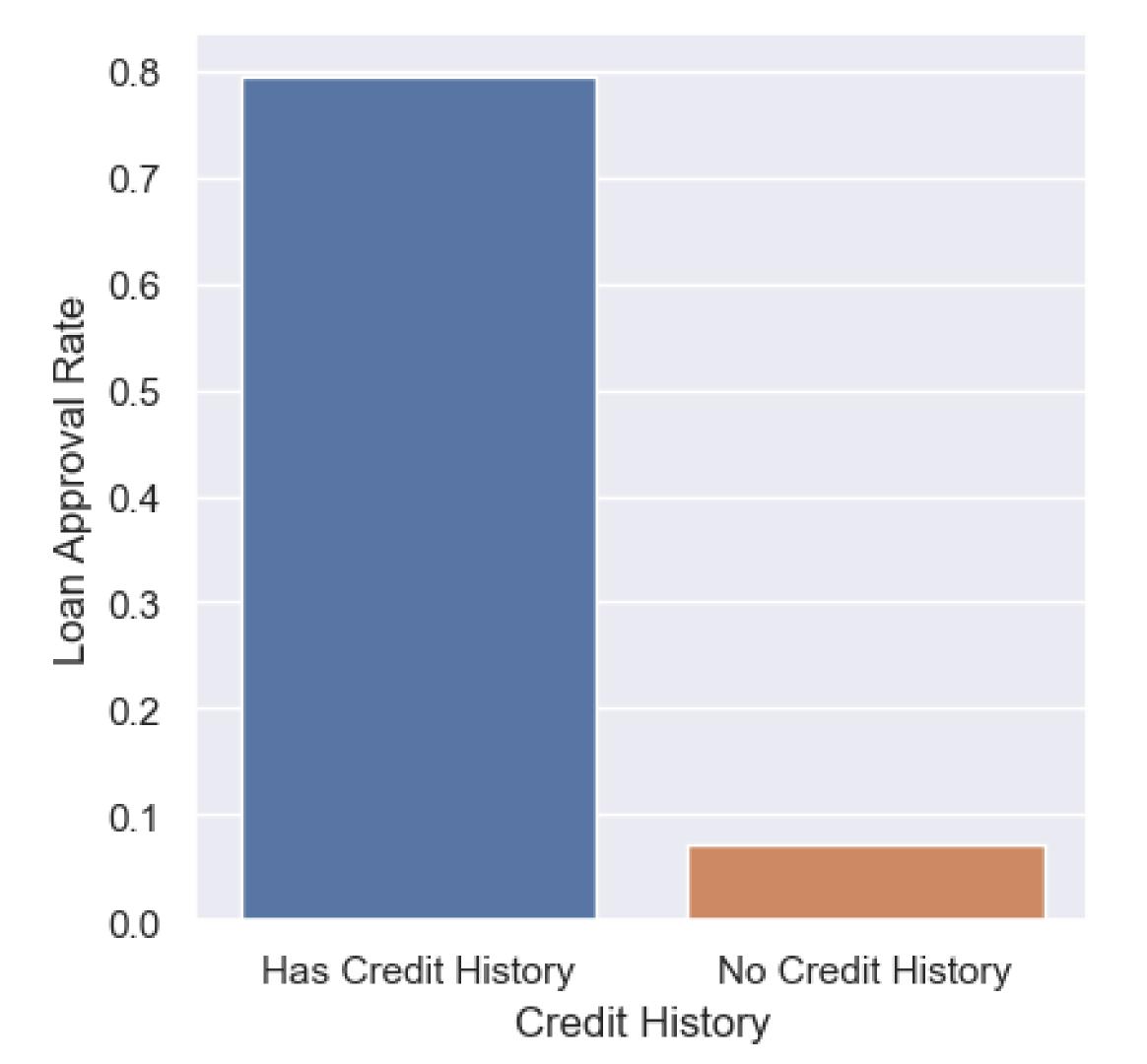
Slightly higher approval rate for applicants who:

- are Men
- are Married
- have 2 Dependents
- have a Graduate
 Education



Slightly higher approval rate for applicants who:

- are not self employed
- have a semiurban property



Much higher approval rate for applicants with a credit history

Machine Learning to Predict Loan Approvals



Pipeline

- add total income: Pipeline
 - ► TotalIncomeAdder
- preprocessing: ColumnTransformer
- numeric
- categorical
- SimpleImputer
- ► SimpleImputer
- StandardScaler
- ▶ OneHotEncoder

RandomForestClassifier

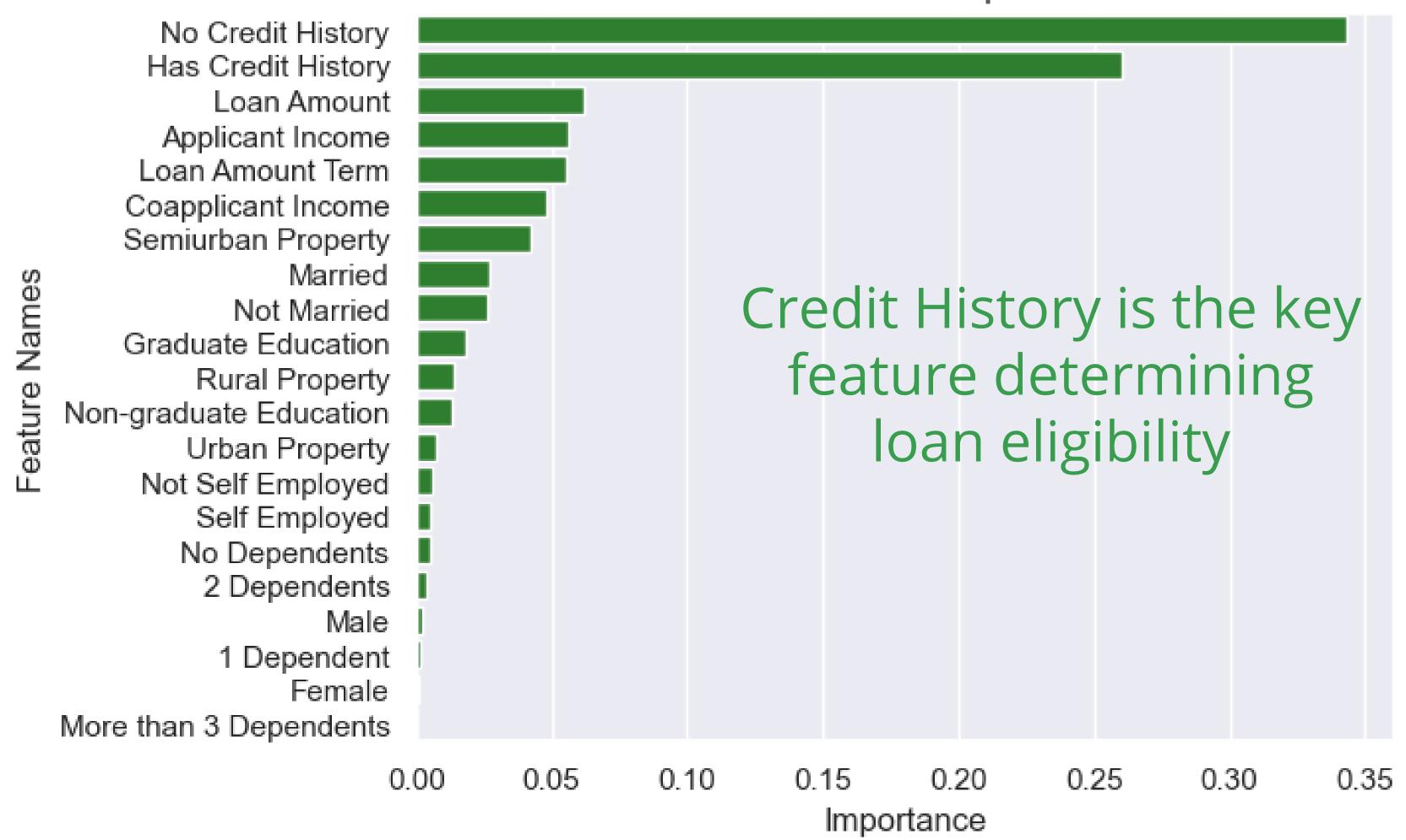
Input

```
"Gender": "Female",
"Married": "No",
"Dependents": "2",
"Education": "Graduate",
"Self Employed": "No",
"Applicant Income": 37719,
"Coapplicant Income": 0.0,
"Loan Amount": 152.0,
"Loan Amount Term": 360.0,
"Credit History": "0.0",
"Property Area": "Semiurban"
```

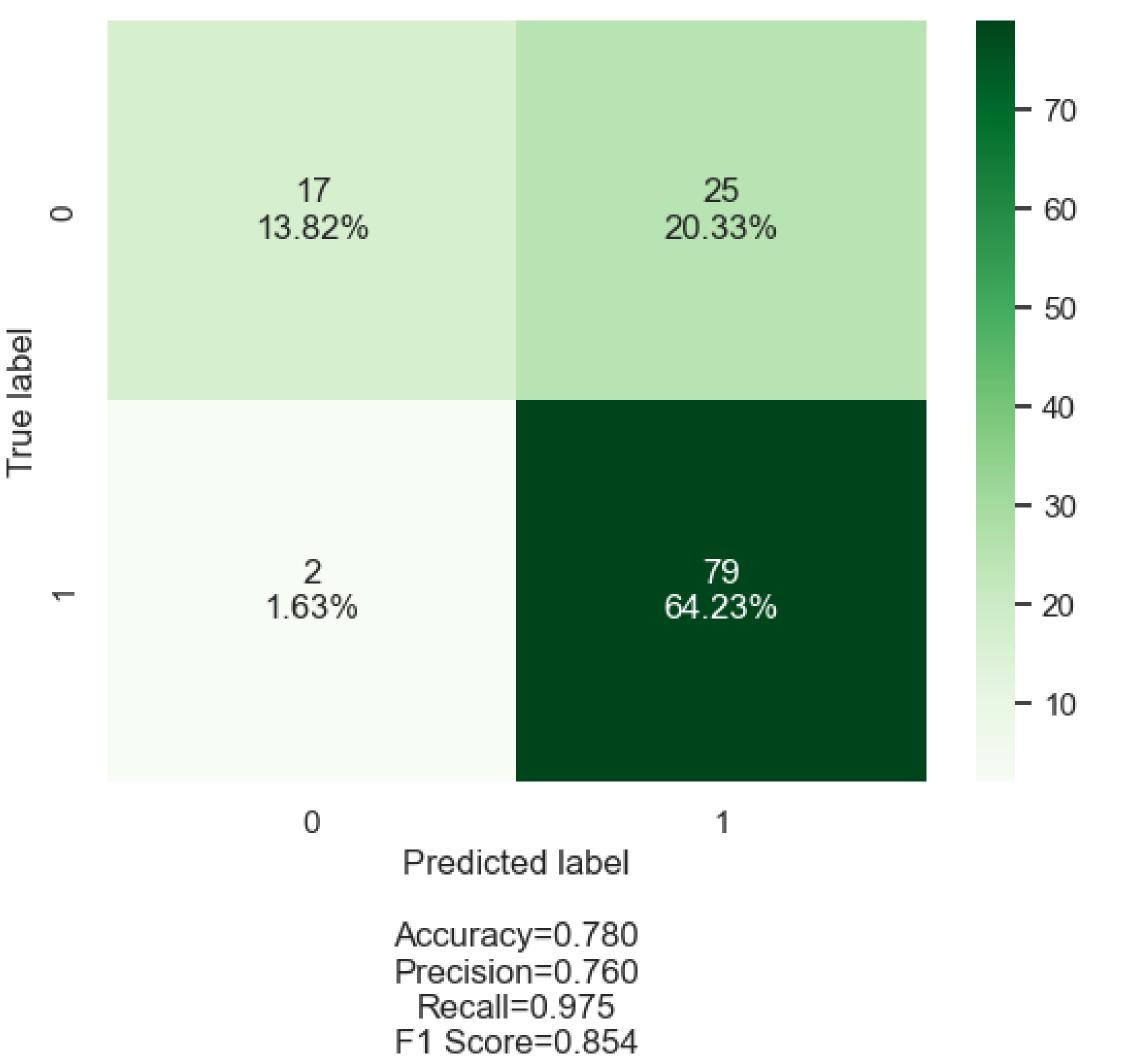
Output

"There is a 60.2% probability this loan application will be rejected."

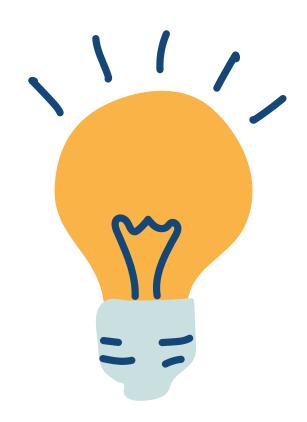
Relative Feature Importance



How good is the model?



Opportunities



Targeted marketing for underrepresented groups

Evaluate emphasis on credit history

Work to eliminate any discriminatory practices

