

Predictive Analytics for Loan Repayment

Date: 2023

\$ 11,017,023
DOLLARS

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Authorized Signature

Agenda

01

Data Overview

02

Summary of Models

03

Final Model Implementation

04

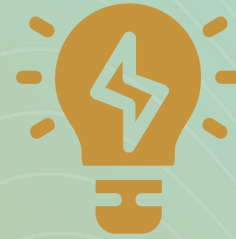
Final Takeaways

Understanding the Task



Goal

Predict whether an applicant will be able to repay their loan



Objective

Analyzing patterns of previous applicants to understand what factors drive whether they will repay their loan

01

Data Overview

Data Overview

Data Source:

- Home Credit Group:
 - https://www.kaggle.com/competitions/home-credit-default-risk/data?select=application_train.csv

Data Preparation + Feature Engineering:

- Collapsed categories of occupation into broader categories
- Transformed variables in days into years

Variables:

Financial Profile



Personal and Demographic Information



Employment and Occupation Details



Asset Ownership



Loan Specifics



Housing Situation



Glimpse of Dataset

Target Variable:

0

Client with no
payment difficulties

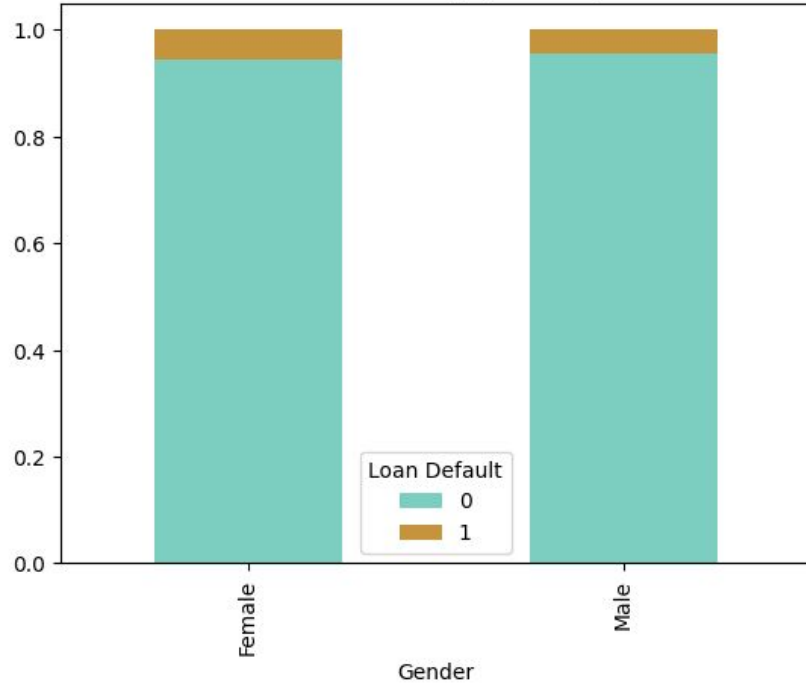
1

Client with payment
difficulties

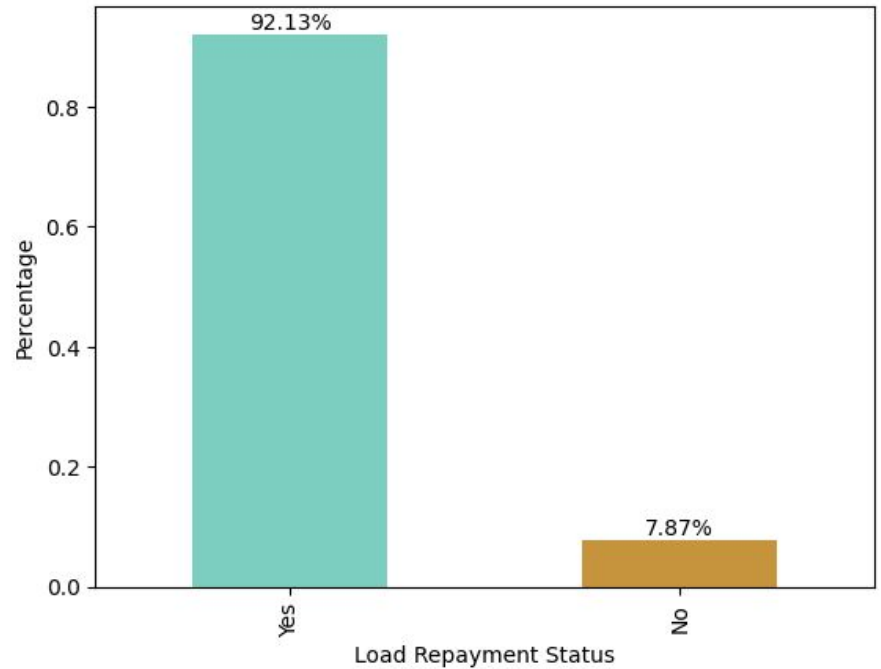
Target	AMT_INCOME_TOTAL	AMT_CREDIT	...	AMT_ANNUITY	CODE_GENDER
1	\$202,500	\$406,597.50	...	\$24,700.50	M
0	\$270,000	\$1,293,502.50	...	\$35,698.50	F
0	\$67,500	\$135,000	...	\$6750	M

Data Exploration

Distribution of Loan Repayment by Gender



Distribution of Whether Loan was Paid Back



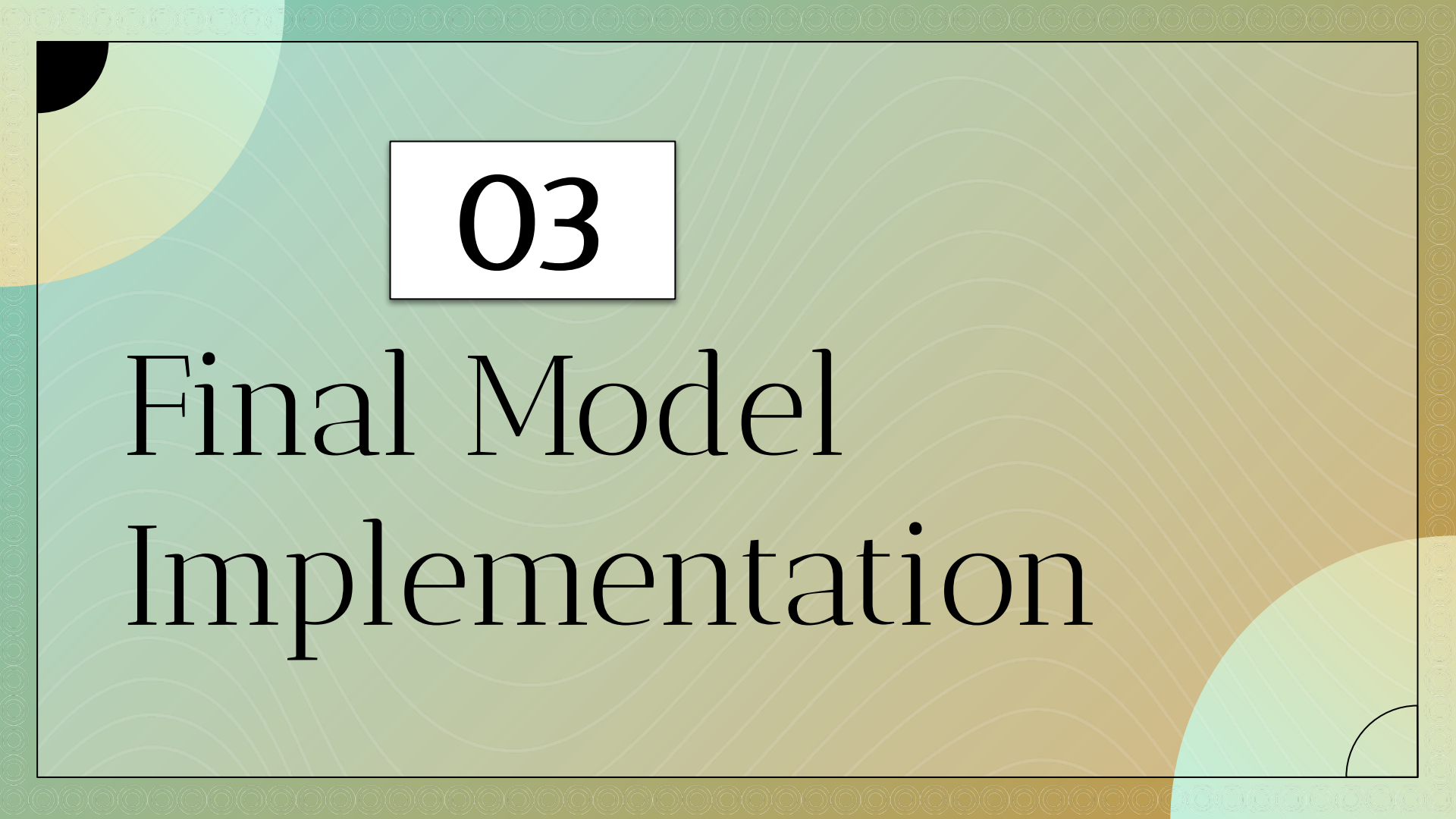
02

Summary of Models

1) Logistic Regression

2) Support Vector Machines

3) Linear Discriminant Analysis



03

Final Model Implementation

Final Model: Logistic Regression

ROC-AUC

Percentage model correctly differentiates between the two classes

59.7%

Accuracy

Percentage correctly classified

75.6%

Recall

Percentage of loan defaults model correctly classified

27.6%

F1 Score

Balance of precision and recall

17.1%

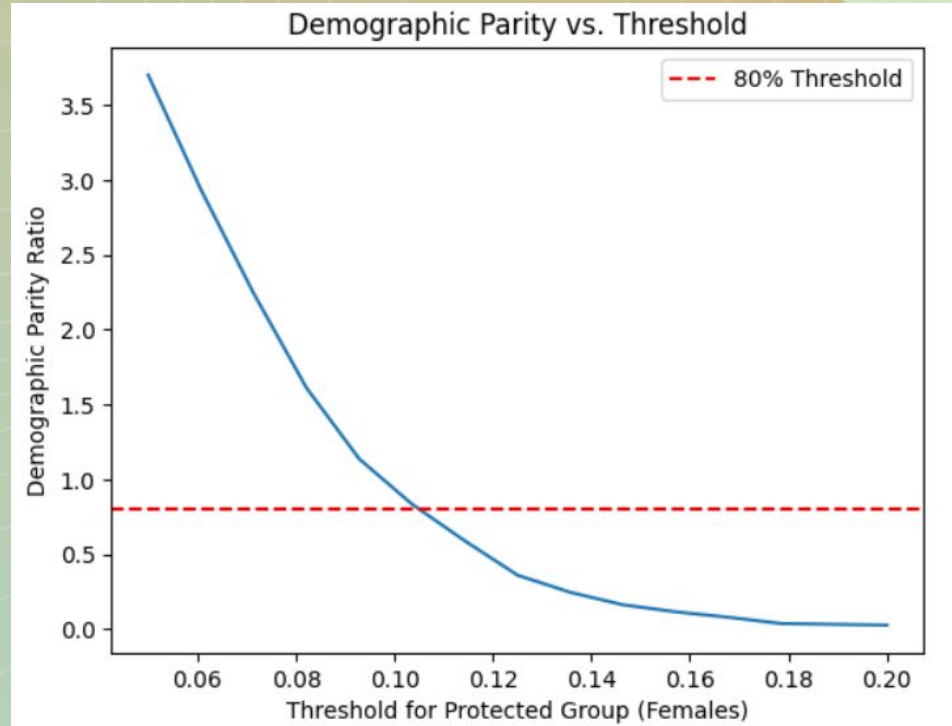
Fairness Metric: Demographic Parity

Gender:

Demographic Parity

The ratio of predicted loan defaults between males and females

1.04



Your Example Customer #1



Gender

Woman

Age

38

Occupation

CEO of Pritchett's
Closets & Blinds

More Information

Education Type

Higher
education

Income

\$176,000
0

Amount of Credit

\$100,000

Number of Children

3

Prediction:

Will repay her loan with no payment difficulties

Your Example Customer #2



Gender

Man

Age

26

Occupation

Temp at Dunder Mifflin

More Information

Education Type

Higher
education

Income

\$21,000

Amount of Credit

\$200,000

Number of Children

0

Prediction:

Will not repay his loan and will have payment difficulties



04

Final Takeaways

Conclusions + Ethical Considerations

Our model demonstrates a reasonable capability in classifying loan defaults



Our model ensures that the loan approval process is equitable across different genders



Our model should be treated as an aid



Thanks!

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Do you have any questions?

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