









Objectives

- •Understanding Factors
 Influencing Loan Default
- Exploratory Data Analysis(EDA) on Banking Data
- •Impact on Lending Practices

Introduction

The analysis delves into loan default risk factors via Exploratory Data Analysis (EDA) on banking data. With a focus on demographic, financial, and occupational attributes, this study aims to unearth patterns influencing loan defaults, providing crucial insights for risk assessment and tailored lending strategies.







Data Processes

Handling Missing Values: Addressing missing data in columns.

Removing Irrelevant Data: Dropping unnecessary columns for analysis.

Dealing with Outliers: Detecting and managing outlier values.

Feature Engineering: Creating new variables or modifying existing ones to enhance analysis.

Exploratory Analysis: Univariate, bivariate analyses to find payment risk indicators.

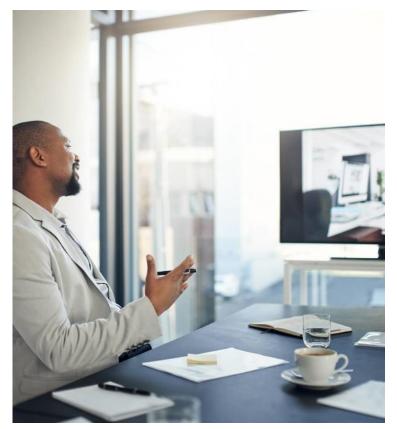
Variable Analysis: Assessing credit, income, education, and contract impacts.

Correlation Study: Examining relationships between variables.

Visualization: Pie, scatter, and correlation plots for insights.

Target Analysis: Identifying demographics more prone to payment issues.

Recommendation & Conclusion: Actionable insights and suggestions based on findings.





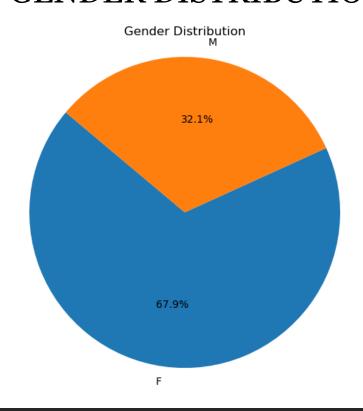


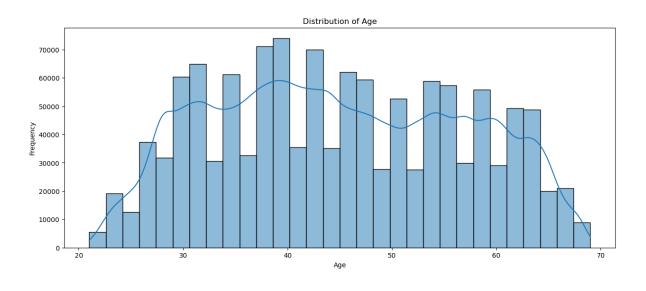
SECTION 1: Demographic Insights

Chart

GENDER DISTRIBUTION

AGE DISTRIBUTION

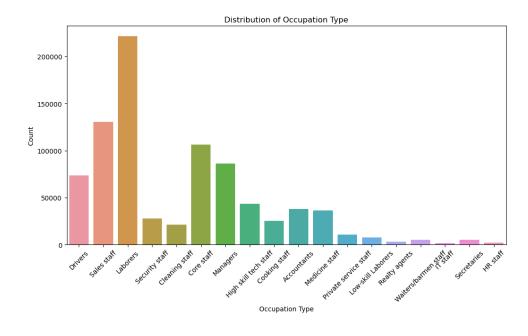




Univariate Analysis

FAMILY STATUS DISTRIBUTION

OCCUPATION DISTRIBUTION

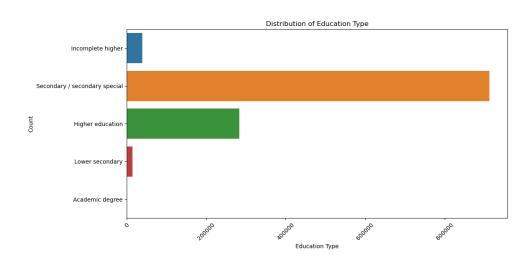


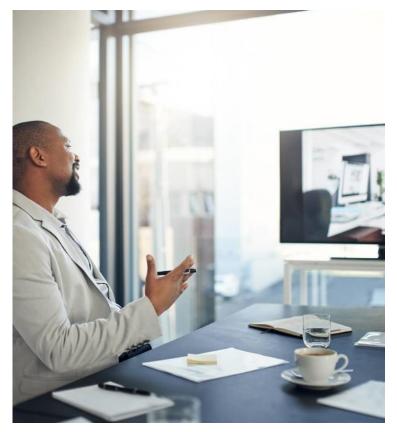
Univariate Analysis

CONTRACT DISTRIBUTION

Distribution of Contract Types 500000 - 400000 - 200000 - 100000 - Consumer loans Cash loans Revolving loans XNA NAME_CONTRACT_TYPE_x

EDUCATION DISTRIBUTION







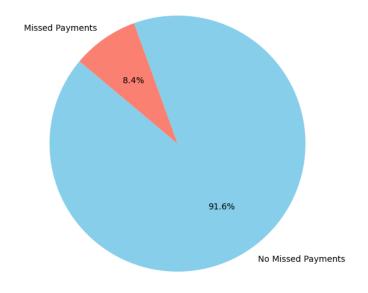


SECTION 2: Missed payments VS No Missed Payment

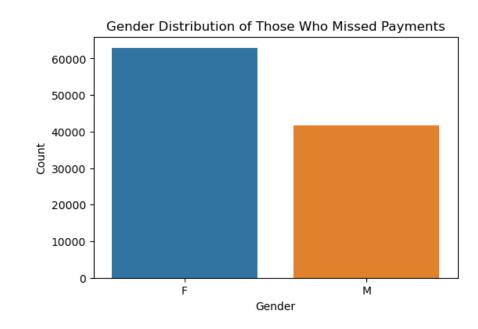
Bivariate Analysis

TARGET DISTRIBUTION

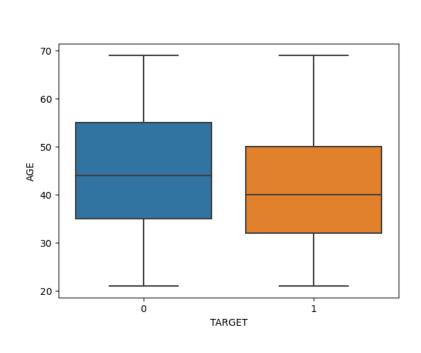
Target Distribution (Missed Payments vs. No Missed Payments)

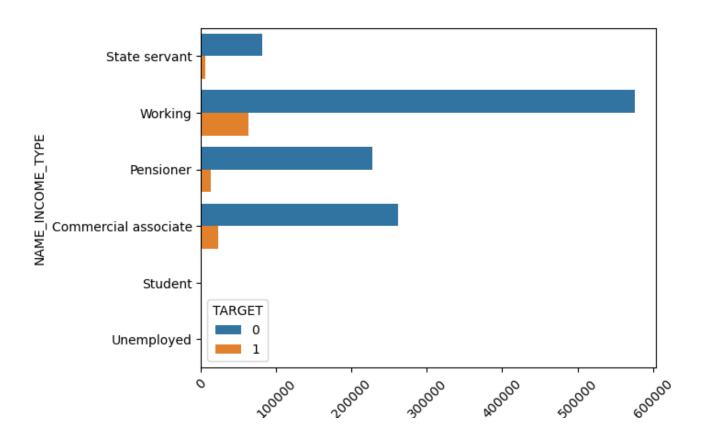


MISSED PAYMENT: GENDER DISTRIBUTION

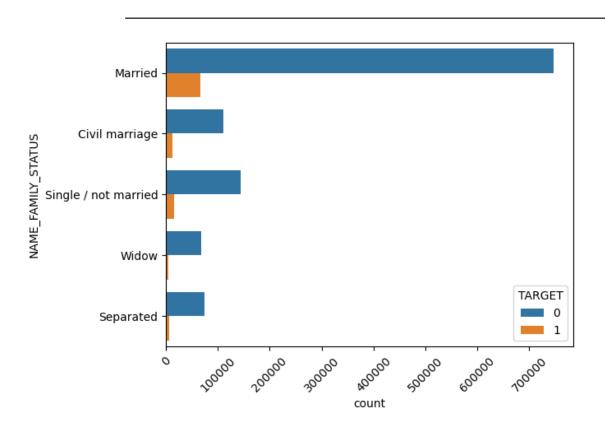


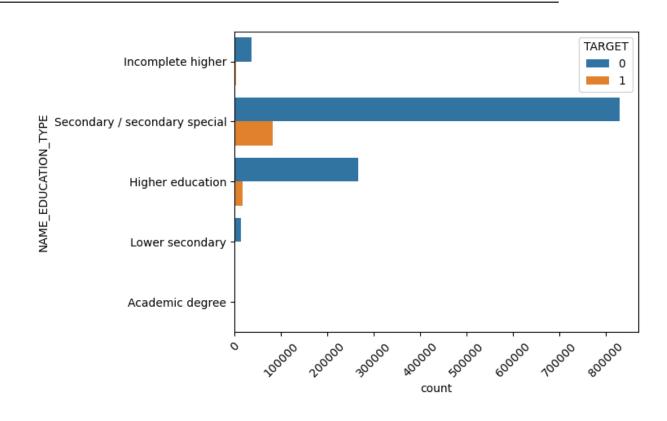
Income Type/ Age Against Target



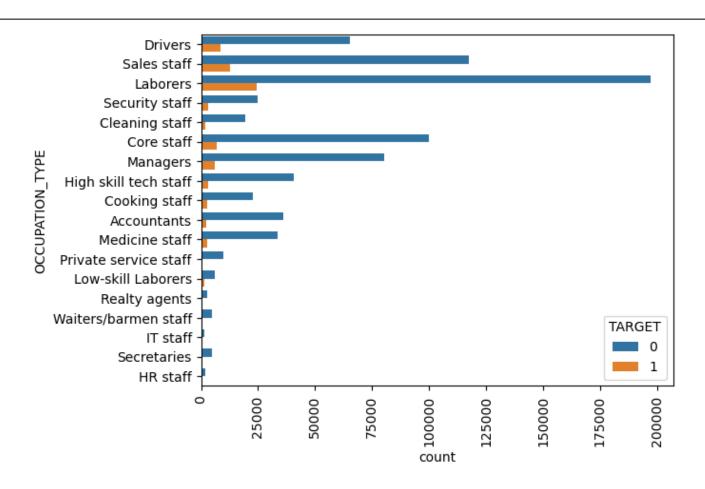


Education Type/Family Status Against Target

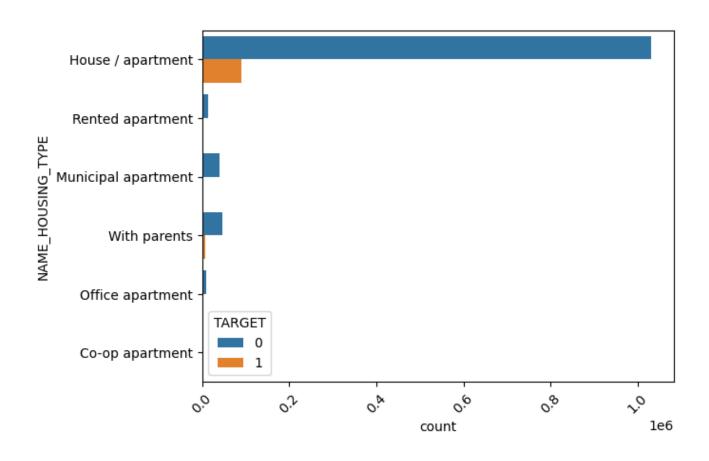


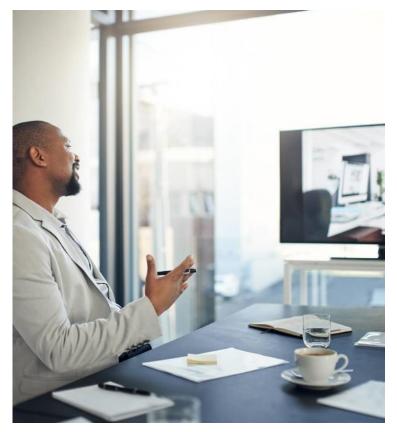


Occupation Type Against Target



Housing Type Against Target

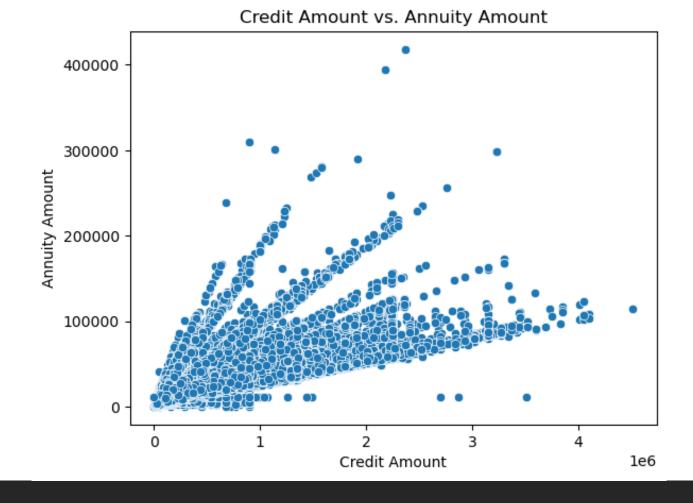








SECTION 3: The Relationship



CREDIT VS ANUITY

The credit amount and annuity exhibit a strong positive correlation of 0.82, implying a robust relationship where higher loan amounts are associated with larger installment payments.

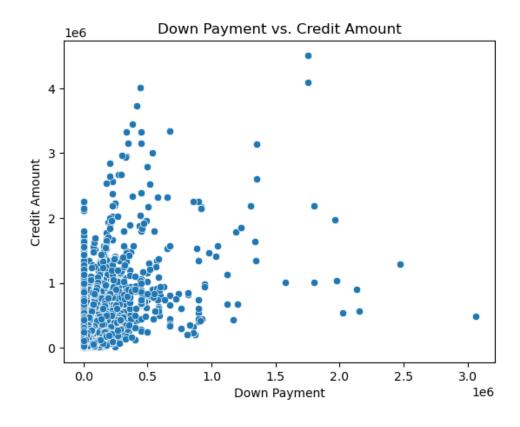


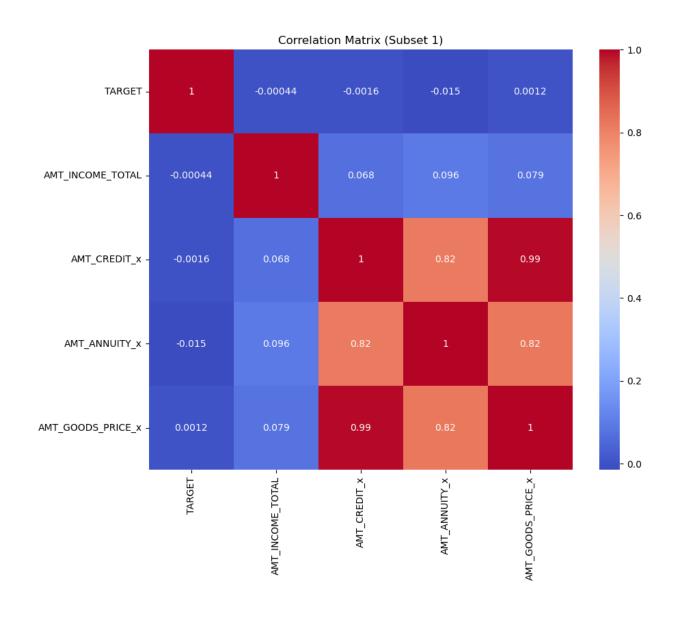
Good Price VS APP Amount

A correlation of 0.99 between the goods price and application amount signifies an extremely strong positive relationship. It indicates that the requested loan amount closely aligns with the price of the goods being financed.

Down Payment VS Credit Amount

A correlation of 0.29 between the down payment and credit amount suggests a moderate positive relationship. While there's some association, it's not as strong as other correlations. This indicates that changes in down payment may somewhat affect the credit amount but not significantly.



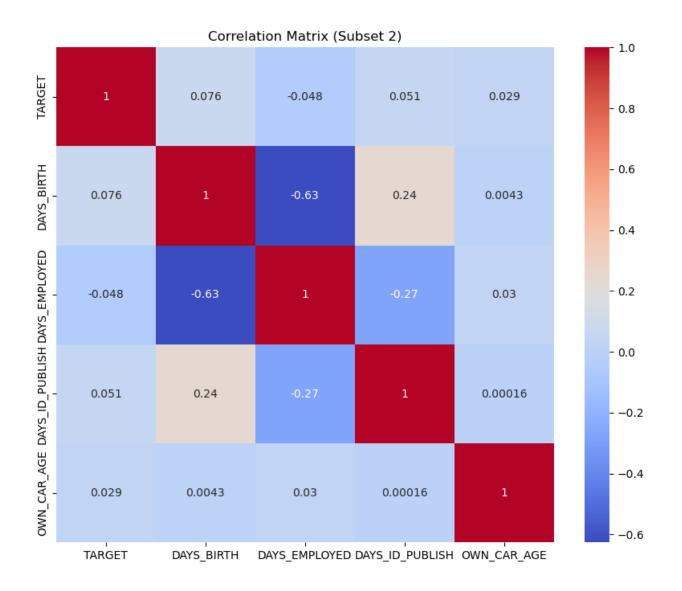


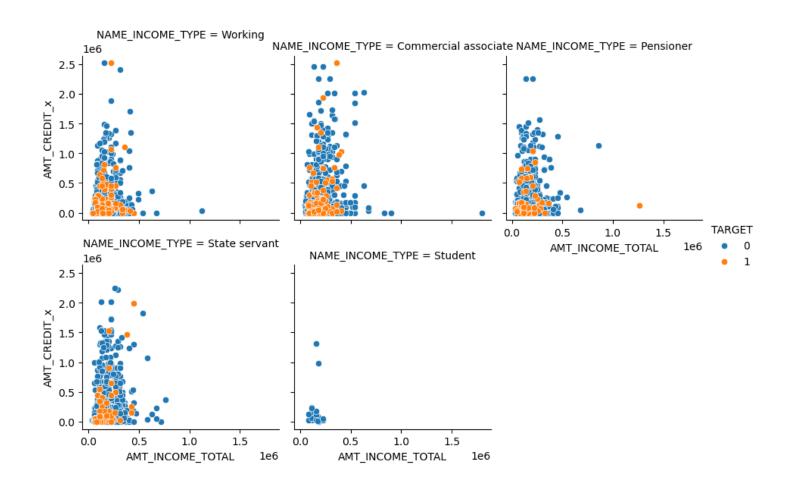
The Correlation Matrix

Showcasing relationship between target, income type, credit amount, good price and annuity amount.

The Correlation Matrix

Showcasing relationship between date of birth, date of car purchase, and ID published date.





SCATTER PLOT

Assessing Multiple Relationship Between Variables

Findings

Payment Status: 8.4% faced payment difficulties.

Household
Composition: Mostly 12 children households.

Gender Dynamics: Females slightly more prone to payment issues.

Age Distribution: Concentration in 30-60 age range. Credit Bureau Activity:
Single inquiry is common.

Occupational
Influence: Laborers,
sales staff, managers at
risk.

Education's Impact: Secondary education linked to missed payments.

Marital Status: Married clients face payment difficulties.

Contract Preferences: Cash, consumer loans are favored. Income, Age, Family
Impact: Laborers in 3049 age group, childless,
face payment
challenges.

Recommendation



Tailored Risk Assessment: Implement targeted risk assessment models considering demographic factors such as age, gender, family size, and education level to determine individual repayment capacities more accurately.

Client-Specific Products: Develop customized financial products and repayment schemes tailored to specific demographics and occupational categories, considering their propensity for payment delinquency.

Financial Literacy
Programs: Offer financial
literacy initiatives,
especially to segments
with higher payment
defaults, to enhance their
understanding of financial
responsibilities and foster
improved repayment
habits.

Marital Dynamics Study: Conduct in-depth studies on marital dynamics impacting financial obligations to identify specific challenges faced by married individuals in meeting payment deadlines. Credit Education
Initiatives: Establish credit
education programs
aimed at clients with
limited credit bureau
activity, providing insights
into responsible credit
usage and building a
positive credit history.



Conclusion

The analysis provides crucial insights into the diverse factors influencing payment defaults among loan applicants. Understanding demographic, occupational, and educational correlations with payment delinquency facilitates the creation of targeted risk mitigation strategies. By tailoring financial products, enhancing financial literacy, and conducting specialized studies, the bank can proactively mitigate payment default risks and ensure more robust lending practices aligned with diverse client segments' financial behaviors.



Thank you

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