# **EDA Lending Case Study**

• Prepared by: Sandeep B, Nishanth A.V:



Photo by Tierra Mallorca on Unsplash

#### Introduction

- Definition and Importance of EDA in Lending: EDA lending, or Exploratory Data Analysis lending, involves analyzing customer and loan data to make informed approval decisions.
- Objectives of the Case Study: Understand patterns in borrower attributes and loan performance, and develop predictive insights to improve loan approval decisions.



Photo by Stephen Dawson on Unsplash

# Background

- Evolution and Challenges in the Lending Industry: The lending industry faces challenges like increasing default rates and changing customer preferences, which impact loan approval processes.
- Key Factors Influencing Loan Defaults: Factors such as credit history, debt-to-income ratio, loan amount, interest rate, and loan terms significantly influence loan defaults.



Photo by Melinda Gimpel on Unsplash

#### **Problem Statement**

- Identifying Patterns in Consumer and Loan Attributes: Explore specific characteristics of borrowers and loan terms linked to defaults to improve loan approval decisions.
- Developing Predictive Models for Loan
   Defaults: Utilize insights from EDA to create
   models that assess new loan applicants'
   likelihood of defaulting, enabling informed
   decisions.



Photo by Jason Briscoe on Unsplash

### **Data Extraction & Cleaning**

- Data Sources and Extraction Process: Data extracted from loan records (2007-2011), converted to CSV format, and imported for EDA.
- Cleaning Procedures: Handled missing values, standardized numerical variables, checked for outliers, and removed irrelevant columns to ensure data quality.

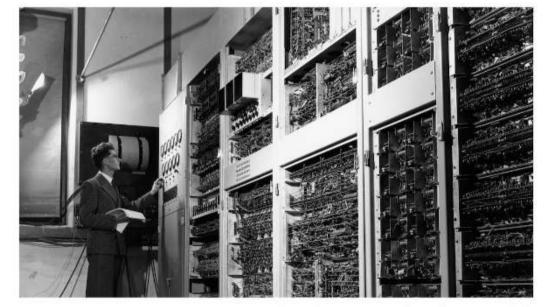


Photo by Museums Victoria on Unsplash

# Exploratory Data Analysis (EDA)

- Univariate Analysis: Examines single variables to identify patterns and trends. Key variables analyzed include loan status, grade, and verification status.
- Segmented Univariate Analysis: Divides data into segments based on categorical variables, allowing for comparison across different groups, such as interest rates by grade.
- Bivariate Analysis: Explores relationships between two variables, such as loan status and verification status, to understand correlations and patterns.
- Derived Metrics: New metrics created from existing data to uncover hidden patterns, such as extracting year and month from issue dates.



Photo by Luke Chesser on Unsplash

#### Results

- Key Findings from EDA: Identified patterns in borrower attributes and loan terms associated with defaults. Developed predictive models for assessing new loan applicants.
- Implications for Loan Approval Decisions:
   Insights can help financial institutions make more informed loan approval decisions, reducing risk and enhancing profitability.



Photo by Sigmund on Unsplash

#### Conclusion

- Summary of Insights: EDA lending provides critical insights into borrower behavior and loan performance, enabling better loan approval decisions.
- Recommendations for Financial Institutions:
   Implement EDA techniques to identify high-risk borrowers, refine loan approval criteria, and guide marketing campaigns.



Photo by Sincerely Media on Unsplash

### **Future Implementations**

- Machine Learning for Loan Default Prediction:
   Implementing machine learning models to analyze borrower and loan data. Algorithms such as logistic regression, decision trees, and random forests will be used to predict the likelihood of loan default.
- Data Preprocessing and Feature Engineering: Cleaning and preparing data, extracting relevant features, and engineering new features to improve model performance.
- Model Training and Validation: Using training and validation datasets to develop, test, and refine models. Employing crossvalidation techniques to ensure robustness.
- Deployment and Monitoring: Deploying the model into the loan approval process. Continuously monitoring performance and updating the model with new data to maintain accuracy.

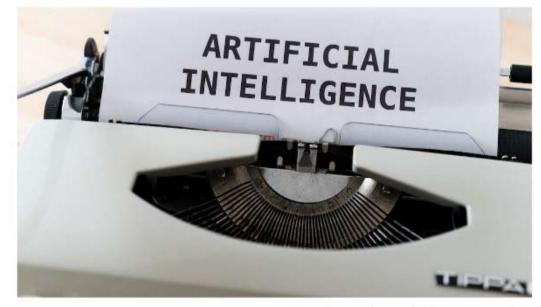


Photo by Markus Winkler on Unsplash