**Project Name: Credit Risk Analysis** 

# **Project Objective:**

The goal of this project was to identify patterns that indicate whether a client might face difficulty in paying back their installments. These patterns can help guide actions like denying loans, reducing loan amounts, or charging higher interest rates for high-risk applicants.

### Approach/ Steps Followed:

- 1. Data Upload
- 2. **Remove Columns with Excessive Missing Values**: Columns with more than 50% missing data were removed—41 columns from current applications and 4 from previous applications.
- 3. **Data Understanding**: Studied the column descriptions to develop a thorough understanding of the data.
- 4. Individual Column Analysis and Cleaning:
  - o **Parameters Checked**: Null values, unexpected entries, potential inferences, and actions to take for each column.
  - Outcome:
    - Null values for numerical columns were replaced with 0 or the mean, depending on the context.
    - Outliers in numerical columns were adjusted using the IQR method.
- 5. **Merging Dataframes**: Cleaned dataframes were merged after initial cleaning.
- 6. **Identifying Data Types**: Differentiated between numeric and categorical data.
- 7. **Cleaning Categorical Columns**: Focused on columns with 10 or fewer unique values. Replaced nulls with the mode or dropped columns with high null values (~49%).
- 8. **Segmenting Data Based on Target Group**: Created two separate dataframes (Target = 0 or 1) for in-depth analysis.
- 9. Running Univariate, Bivariate, and Multivariate Analysis:
  - o **Tools Used**: Pie charts, heatmaps, scatter plots, and bar plots to explore correlations.
- 10. **Conclusion**: Summarized the findings and identified significant correlations.

#### **Key Findings from the Analysis**

#### **Pie Chart Observations:**

- 1. Loan Type: Revolving loans have fewer repayment issues compared to cash loans.
- 2. **Gender**: Females are generally better at loan repayments than males.
- 3. **Car Ownership**: Car owners have a slightly better repayment rate.
- 4. **Family Accompaniment**: Applicants accompanied by family during loan application have better repayment records.

- 5. **Income Type**: State servants have the best repayment history, followed by working-class individuals.
- 6. **Education**: Applicants with higher education levels are the most reliable in repaying loans.
- 7. Family Status: Married individuals are more reliable in repayments compared to singles.

## **Heatmap Observations:**

Identified strong correlations between:

- AMT\_GOODS\_PRICE, AMT\_CREDIT, and AMT\_ANNUITY
- CNT\_CHILDREN and CNT\_FAM\_MEMBERS
- DAYS\_LAST\_DUE and DAYS\_TERMINATION

#### **Scatter and Count Plot Observations:**

- No specific combinations showed distinct behavior differences between those who struggle with payments and those who do not.
- Some categories like **Cash Loans** and **House/Apartment owners** displayed unique patterns.

## **Conclusion:**

Based on the analysis, the following parameters have the most significant impact on loan repayment:

- Gender
- Education Level
- Family Status

These factors are critical for assessing a client's risk profile and guiding loan-related decisions.

## **Project Outcomes:**

- For the use case, the project provided crucial insights to aid decision-making in the loan approval process.
- Personally, this project has enhanced my ability to manage large datasets and handle a wide range of parameters, even without direct guidance from clients.