

Documentation for Bank Customer Churn Prediction Project

➤ Overview

- This project focuses on predicting customer churn in a banking environment using machine learning. The goal is to identify customers likely to leave the bank, enabling proactive measures to retain them.

➤ Dataset Description

- **Target Variable:**
 - **Exited:** Indicates whether the customer has churned (1) or stayed (0).
- **Features:**
 - **RowNumber:** Row number in the dataset.
 - **CustomerId:** Unique identifier for each customer.
 - **Surname:** Customer's last name.
 - **CreditScore:** Customer's credit score.
 - **Geography:** Country of residence.
 - **Gender:** Gender of the customer.
 - **Age:** Customer's age.
 - **Tenure:** Number of years the customer has been with the bank.
 - **Balance:** Customer's account balance.
 - **NumOfProducts:** Number of products used by the customer.
 - **HasCrCard:** Whether the customer owns a credit card (1: Yes, 0: No).
 - **IsActiveMember:** Indicates if the customer is an active member (1: Yes, 0: No).
 - **EstimatedSalary:** Estimated annual salary of the customer.

➤ Libraries and Tools Used

- **Libraries:**
 - NumPy, Pandas: For data manipulation and analysis.
 - Matplotlib, Seaborn: For visualization.
 - Scikit-learn: For preprocessing and modeling.

➤ Steps in the Notebook

1. **Data Loading:**
 - The dataset is loaded into a Pandas DataFrame from a CSV file.
2. **Exploratory Data Analysis (EDA):**
 - Includes visualizations and statistical summaries of key features.
3. **Preprocessing:**
 - Encoding categorical variables, handling missing values, and scaling numeric features.
4. **Model Building:**
 - Selection and training of machine learning models.
5. **Evaluation:**
 - Performance metrics to evaluate the predictive power of the model.

➤ Visualizations

- **Analysis of the relationship between the credit score (CreditScore) and the exited**
 - **Note:** customers with low credit scores (less than 600) leave with a higher percentage compared to those in the higher category (700+)
 - **Resolution:** focus on improving the experience of customers with medium and low credit scores, such as providing financial advice or incentive programs.
- **Analysis of the relationship between age and exited.**
 - **Note:** customers with a high balance tend to leave the bank more often than customers with a low balance
 - **Resolution:** improve investment offers or offer higher benefits to clients with high balances
- **Analysis of the relationship between the number of products (NumOfProducts) and exited**
 - **Note:** customers who own one product are most likely to leave.
 - **Resolution:** encourage customers to increase the use of products by offering incentives when using more than one product.
- **Analysis of the relationship between the activity (IsActiveMember) and exited**
 - **Note:** inactive clients leave at a higher percentage compared to active ones.
 - **Resolution:** implementation of interactive programs to motivate inactive customers to participate more with the bank.
- **Analysis of the relationship between estimated salary (EstimatedSalary) and exited**
 - **Note:** there is no obvious effect of the estimated salary level on departure, as the departure percentages are distributed almost equally at all salary levels.
 - **Resolution:** focus on other characteristics, such as activity and number of products, when developing customer retention strategies.
- **Analysis of the impact of Geography on Exited**
 - **Note:**
 - **In France:** most customers stayed with the bank, the percentage of departures is low compared to other regions.
 - **In Spain:** the percentage of customers leaving is average, but it is less pronounced compared to Germany.
 - **In Germany:** the percentage of departures is the highest compared to other regions, which indicates a possible problem with services or interaction with customers.
 - **Resolution:**
 - **In France:** maintaining the quality of services and the continuity of Special Offers to maintain customer loyalty.
 - **In Spain:** improving customer interaction through targeted support programs to prevent departures.
 - **In Germany:** improving the services and offers offered to customers to reduce departure rates, such as offering competitive advantages or loyalty programs.

- **The relationship between gender and the client's exit status**
 - **Note:**
 - The percentage of departures is higher among female customers compared to male ones.
 - Male clients show a higher retention rate with the bank.
 - **Resolution:**
 - Improve interaction with female clients by customizing services and programs that suit their needs.
 - Study the reasons that lead to females leaving more and work on addressing them through questionnaires or direct interaction.
 - Promote retention strategies for female clients by offering offers or incentives specifically targeted at them.

➤ **Instructions for Running the Notebook**

1. **Dependencies:**
 - Ensure Python 3.x is installed.
 - Install required libraries using `pip install -r requirements.txt`.
2. **Dataset:**
 - Place the `Churn_Modelling.csv` file in the same directory as the notebook.
3. **Execution:**
 - Run each cell sequentially to process the data, build the model, and visualize results.

➤ **Key Insights**

- The importance of age, credit score, and activity status in predicting churn.
- Customers with higher engagement tend to stay longer.