



Dataset

The dataset used is Churn_Modelling.csv, containing customer details such as:

- Credit Score
- Age
- Balance
- Gender
- Geography
- Exited (whether the customer churned)





Project Workflow

1 Data Loading & Exploration



-  Load the dataset using pandas.
-  Explore for missing values, inconsistencies, and general data insights.

2 Data Visualization

Generate visualizations to understand:

-  Distribution of numerical features (Age, Credit Score, Balance, Estimated Salary).
-  Distribution of categorical features (Gender, Geography).
-  Correlation heatmap to identify relationships between features.
-  Target variable (Exited) distribution to check for class imbalance.


3 Feature Engineering

-  Encode categorical variables (Geography, Gender) using One-Hot Encoding or Label Encoding.
-  Ensure feature cleanliness, relevance, and proper formatting for modeling.


4 Data Splitting

-  Split data into 80% training and 20% testing sets using `train_test_split` from scikit-learn.

5 Handling Class Imbalance





-  If class imbalance is detected in the Exited column, apply SMOTE (Synthetic Minority Oversampling Technique) to balance the dataset.

Feature Standardization





-  Normalize numerical features to ensure they are on the same scale (mean = 0, standard deviation = 1), which is crucial for models like SVM and KNN.


Model Training & Evaluation

Train the following machine learning models:




-  K-Nearest Neighbors (KNN)
-  Naive Bayes
-  Support Vector Machine (SVM)
-  Decision Tree (DT)

Evaluate models using the following performance metrics:

-  Accuracy
-  Precision & Recall
-  F1-Score
-  ROC-AUC Score

 **Compare all models based on performance metrics and identify the best-performing model.**

Deliverables

-  A trained model capable of predicting customer churn.
-  A comparative analysis of different machine learning models.
-  Visual representations of results and insights from the dataset.

Repository Structure

```
|-- Customer_Churn_Detection/ |-- data/ |-- Churn_Modelling.csv |-- file.ipynb |-- README.md
```

How to Run the Project

Clone the repository:

- git clone <https://github.com/ayaatef11/Churn-Modeling.git>
- cd Customer_Churn_Detection

Run the preprocessing and training scripts:

```
python src/data_processing.py python src/model_training.py
```


View results and model evaluation:

```
python src/evaluation.py
```

Dependencies

 Python 3.x


 Pandas


 NumPy


 Scikit-learn

 Matplotlib


 Seaborn

 Imbalanced-learn (for SMOTE, if needed)

 Results and Findings

 The project evaluates four models to determine the best one for predicting customer churn.

 The best-performing model is selected based on accuracy, precision, recall, F1-score, and ROC-AUC score.

 Visualizations and charts are provided to illustrate key findings.

Contributors

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