#### **Customer Churn Prediction Report**

### 1. Executive Summary

This report presents the development and evaluation of a Churn Prediction System that identifies customers likely to stop using a service.

Retaining existing customers is significantly more cost-effective than acquiring new ones.

Our best models achieved up to 87% accuracy, with key churn drivers including short contract lengths, payment method types, and early customer tenure.

This predictive tool enables proactive retention strategies.

# 2. Dataset Overview

• Dataset: Churn\_Modelling.csv

• Total Customers: 10,000

• Churned (Exited = 1):  $2,037 (\sim 20.4\%)$ 

• Retained (Exited = 0):  $7.963 (\sim 79.6\%)$ 

## 3. Data Preparation and Feature Engineering

- Cleaned missing or inconsistent values
- One Hot encoded categorical features
- Engineered features such as:
  - o Tenure
  - o Total Charges(Balance, Credit Scores)
  - Contract Type
  - o Payment Method
  - Service Subscriptions
  - Active Membership
  - o Gender
  - o Age
  - Number of products

### 4. Machine Learning Models and Evaluation

Three classification models were trained and evaluated:

# • Logistic Regression:

o Cross-validation accuracy: 86%

o Final test accuracy: 84.25%

#### • Random Forest:

o Cross-validation accuracy: 87%

o Final test accuracy: 84.25%

#### • XGBoost:

o Cross-validation accuracy: 87%

o Final test accuracy: 84.25%

#### 5. Final Evaluation Metrics

• Accuracy Score: 0.8425

• Confusion Matrix:

[1491 116]

[199 194]]

# **Classification Report:**

Class	Precision	Recall	F1-Score	Support
0	0.88	0.93	0.90	1607
1	0.63	0.49	0.55	393

• Overall Accuracy: 84%

• Macro Average F1-Score: 0.73

# 6. Key Insights: Churn Drivers

Top predictors of churn:

- Month-to-month contracts
- Electronic check payment method
- Low tenure
- Not subscribing to multiple services

# Loyal customers typically:

- Have long-term contracts
- Use bank transfer or card payments
- Subscribe to bundled services