Predicting Customer Churn

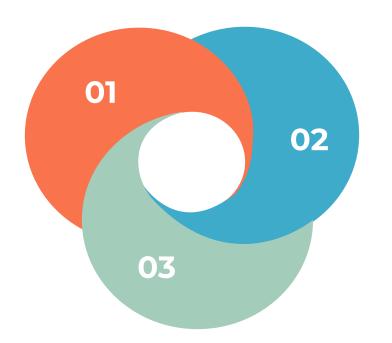
This presentation provides an overview of a machine learning model developed to predict customer churn and its business implications.

Introduction

Project Goal: To develop a machine learning model that predicts customer churn with high accuracy.

Why It Matters:

Reducing churn improves customer retention, lowers acquisition costs, and increases revenue stability.



Business Value:

Helps the company identify at-risk customers and take proactive measures to reduce churn rates.



Data Overview

 Dataset Source: The data is sourced from Kaggle's Telecom Customer Churn Dataset.

 Key Features: Customer demographics, account information, and service usage.

 Target Variable: Whether a customer churned (Yes/No).

Methodology

O1 Evaluation Metrics: Accuracy, Precision, Recall, and F1-Score.

Data Preprocessing: Handled missing values, encoded categorical variables, and normalized numerical features.

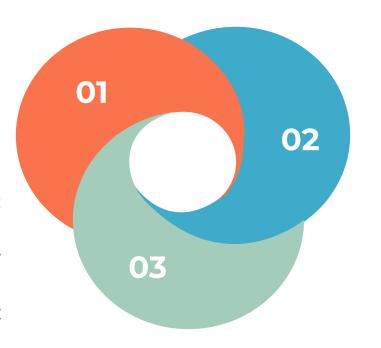
Modeling Approach: Compared multiple models including Logistic Regression, Random Forest, and Decision Tree.

Key Findings

Best Performing
Model: The Decision
Tree model
outperformed others
with 93.9% accuracy.

Business Implication:

The model can be used to prioritize customer outreach efforts, focusing on those most likely to churn.

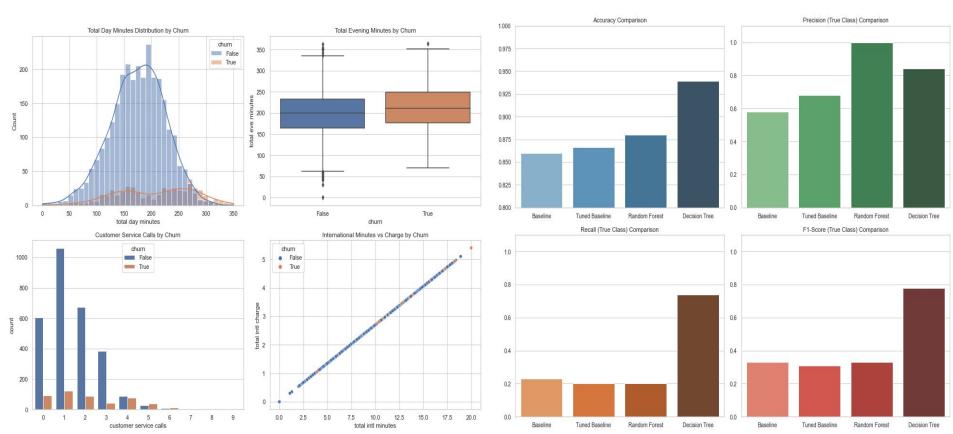


Insights: Precision and recall indicate the model is effective at identifying both churned and retained customers.

Visualizations

Visualization of Key Features

Model Comparison Plot



Recommendations

 Short Term Actions: Implement the Decision Tree model into the customer management system.

 Long Term Strategy: Regularly retrain the model with updated data to maintain its predictive accuracy.

 Stakeholder Engagement: Use model outputs to guide marketing and customer service teams in retention efforts.

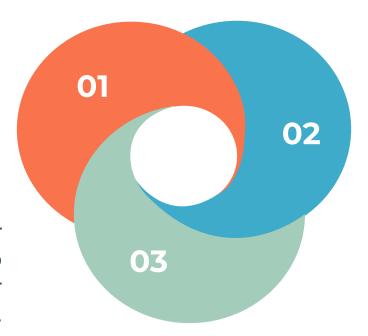


Next Steps

Potential Improvements:

Explore additional features like customer satisfaction scores for better predictions.

Scaling: Consider extending the model to other customer segments or regions.



Further
Development: Test
model performance
in real-time with live
customer data.

Conclusion

Final Thought: Ongoing refinement and stakeholder collaboration will ensure the model continues to add value.

Recap: Successfully built a predictive model that identifies customers at risk of churning.

Business Value: The model supports targeted retention strategies, helping reduce churn and increase profitability.

Thank you for your time 😊