



Telecom-Churn Prediction

**A Machine Learning Study to Identify
Customers at Risk of Leaving**



Outline

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Overview

Description

- This project focuses on predicting customer churn for a telecom company, DIGICEL Haiti for example.
- The goal is to build a machine learning classification model that identifies customers at high risk of leaving the company's services.



Overview

Goals

- Identifies customers at high risk of leaving
- Identify and Predicting churn-prone customers
- Uncover patterns that indicate churn
- Provide **data-driven recommendations** for strategic decision-making.

Business Understanding

Business Problem :

A company is losing a significant number of customers.

1. High customer churn bring about lost revenue & higher acquisition costs.
2. No reliable way to identify at-risk customers proactively.



Data Understanding

Data Sources :

Kaggle - Churn in Telecoms Dataset

Data format : CSV

Records : 3 333

Data Understanding

Methods

- **Data Preprocessing:** Clean and prepare Data for modeling Build the Baseline Decision Tree to predict churn
- **Model Building:** Explore multiple models and select the best performer
- **Model Evaluation :** Measure performance using key metrics
- **Recommendations:** Translate findings into actionable strategies for the company

Data Understanding

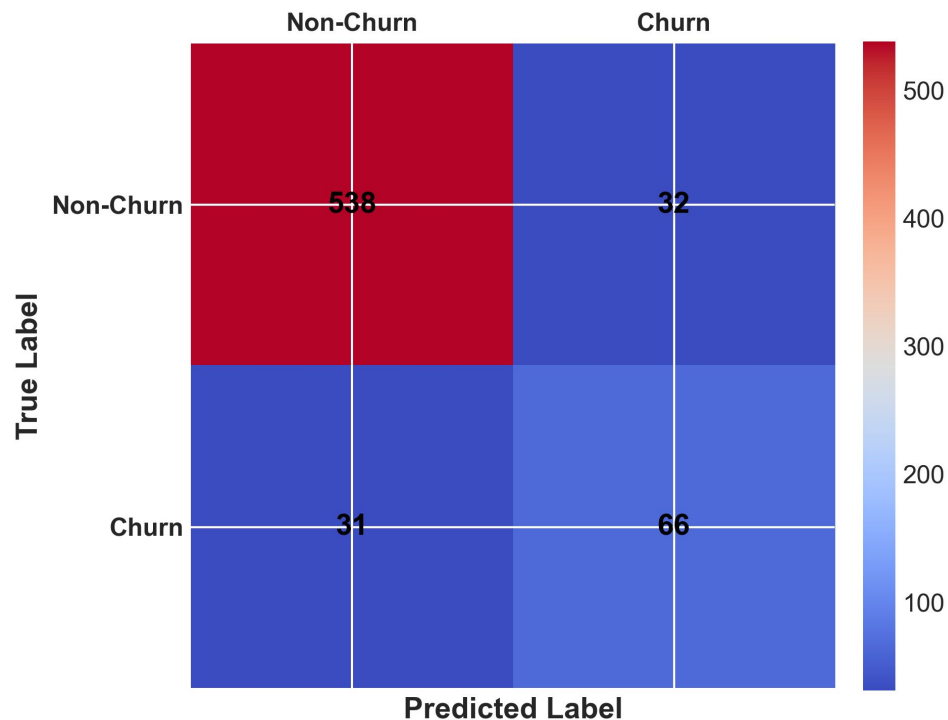
Key Variables :

- **Movie Attributes** : Age, Gender, Service Plan
- **Financial Metrics** : Total Charges, Payment Method
- **Usage Metrics** : Call Minutes, Data Usage, SMS
- **Churn Indicator** : Whether the customer left the service

Model Performance & Results

Confusion Matrix

- **Loyal Customers:** 538 correctly identified → saves retention costs At-Risk Customers:
- **At-Risk Customers:** 66 flagged → proactive actions can prevent churn



Model Performance & Results

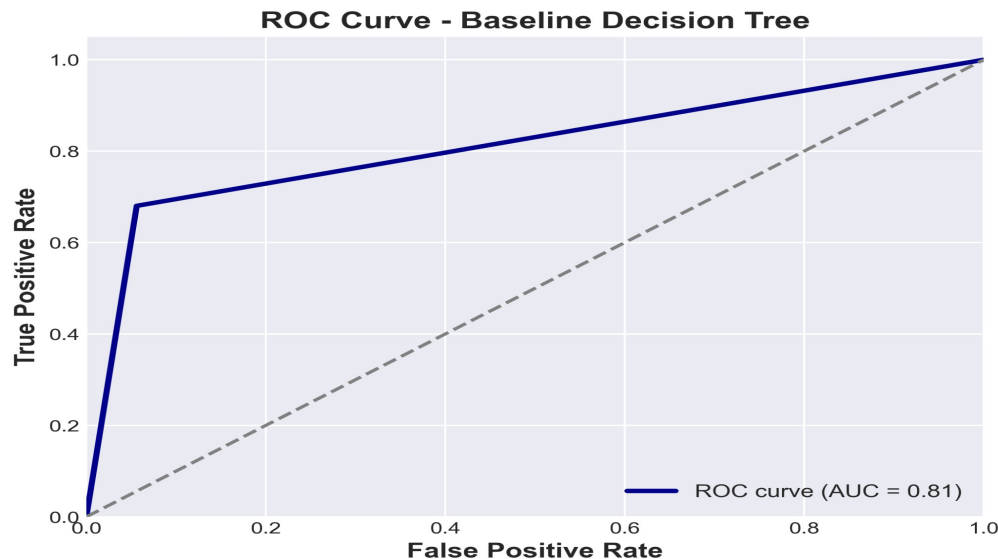
Key metrics

- Correctly predicts **91% of customers**
- Accurately identifies **most loyal clients**
- Captures **68% of churners** for retention focus
- Balance ensures **revenue protection with limited extra costs**

Model Performance & Results

ROC Curve

- Distinguishes churners from loyal customers with **81% accuracy**
- It shows performance **well above random chance**
- Helps **focus retention on at-risk customers**



Recommendations

1. **Proactive Retention Programs :**

Target high-risk customers with offers and loyalty programs

2. **Prioritize High-Value Customers :**

Combine churn risk with CLV to focus on most valuable clients

3. **Enhance Service & Marketing Strategies :**

Use model insights to improve products & campaigns

4. **Monitor & Update Models :**

Integrate model in operations and retrain with new data.

Next Steps

- **Continuously monitor churn patterns:**
Track revenue, profit, and audience engagement continuously.
- **Refine predictive models ::**
Retrain with new data to improve accuracy and recall.
- **Leverage Partnerships :**
Personalize offers and outreach for at-risk customers.
- **Implement targeted retention strategies :**
Measure churn reduction, revenue gains, and ROI.
- **Integrate into CRM systems :**
Embed churn scores to guide real-time sales and support actions.

END !

- Questions?



END !

Thank you for your attention!

Feel free to reach out!

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