# Customer Churn Prediction

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#### The Process

**Project Overview Business Problem Data Understanding Data Preparation** Modelling Conlusions **Recommendations** 

- Data Cleaning
- Explanatory Data Analysis
   Univariate Analysis
   Bivariate Analysis



# Project Overiew

SyriaTel faces a challenge with customer churn, impacting revenue and market standing. This project aims to predict churn using key features like customer service calls, daytime call minutes, and international plan subscriptions



### **Business Problem**

SyriaTel, is a telecommunications company facing a significant challenge: customer churn. They've noticed an increasing number of customers leaving their service, which is not only impacting their revenue but also their reputation in the market

# Data Understanding



Syrial Tel dataset Contains data bout customer usage, calls and charges with churn column as the target variable



The data contains: 3333 rows 21 columns

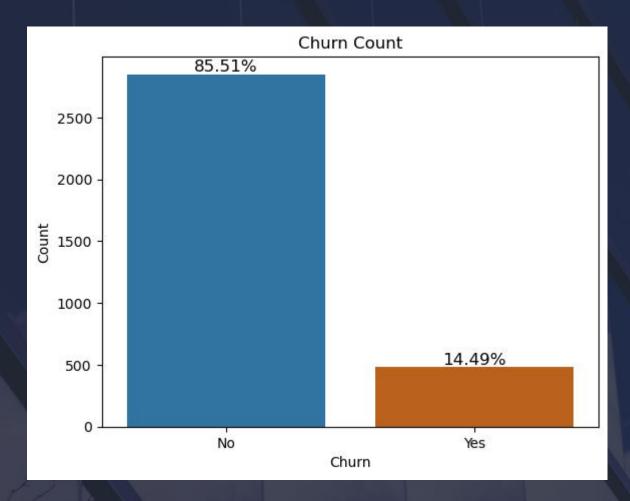


THe datatypes of the column includes:

- int64
- object
- bool
- object
- int32

#### **EXPLANATORY DATA ANALYSIS**

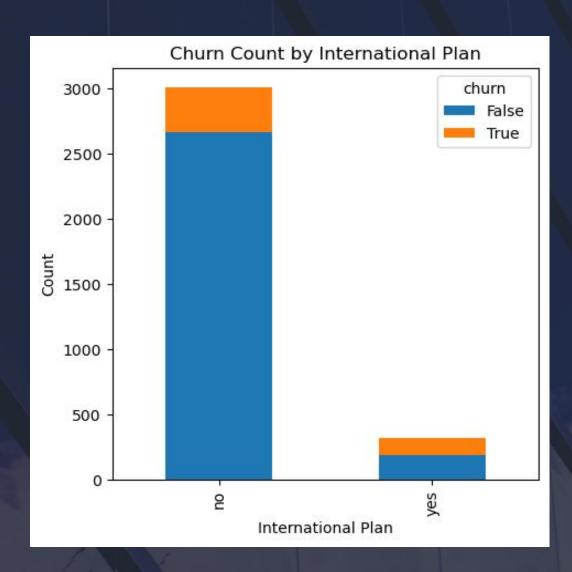
This shows that SyriaTel Telecommunication Comapny has a 14.49% churn rate





Customer chur rate

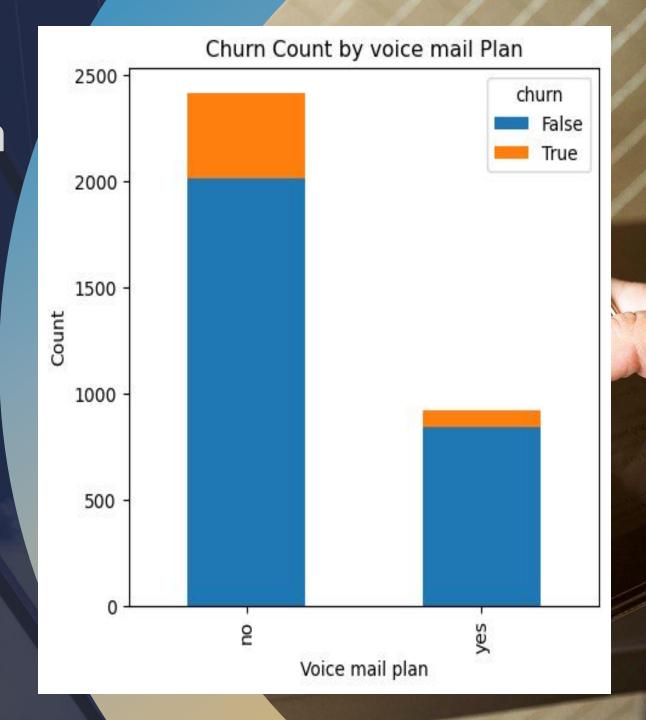
# Churn by international plan

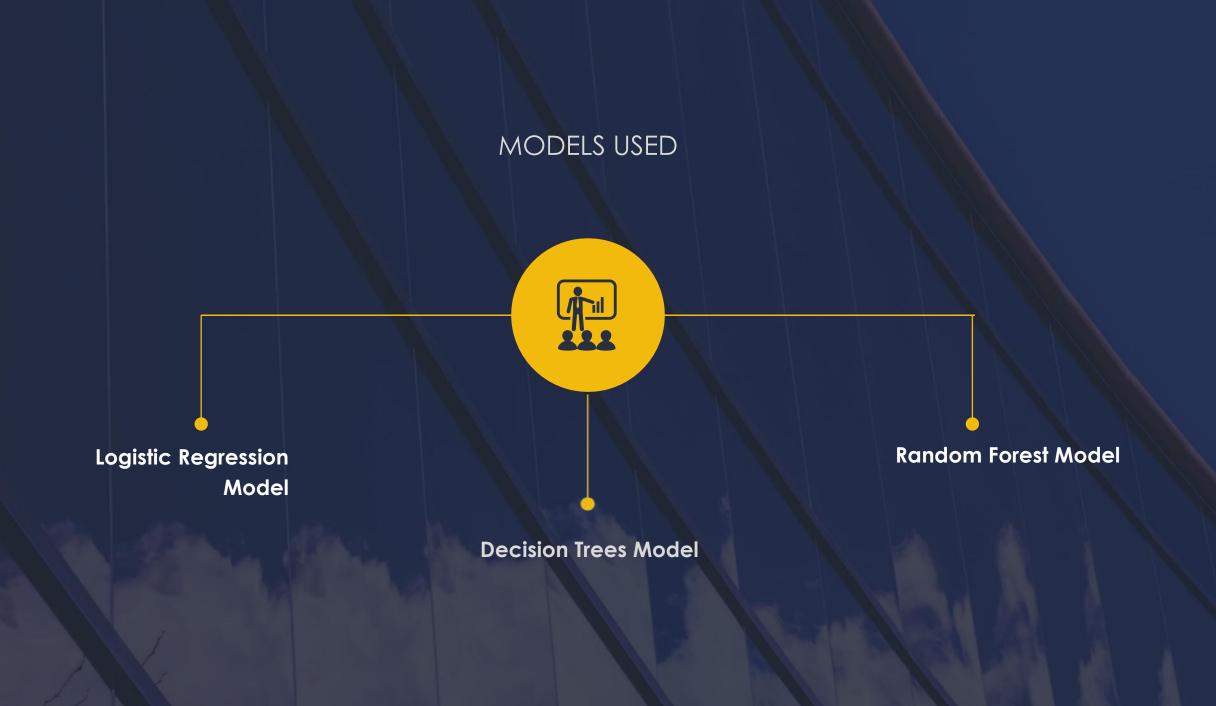


People with an international plan tend to churn more than people with no international plan. It is also evident that most customers in Syria Tel have no internal plan

# Churn by voice mail plan

Customers with no voice mail plan tend to churn more than customers with voice mail plan





#### **Models Evaluation**

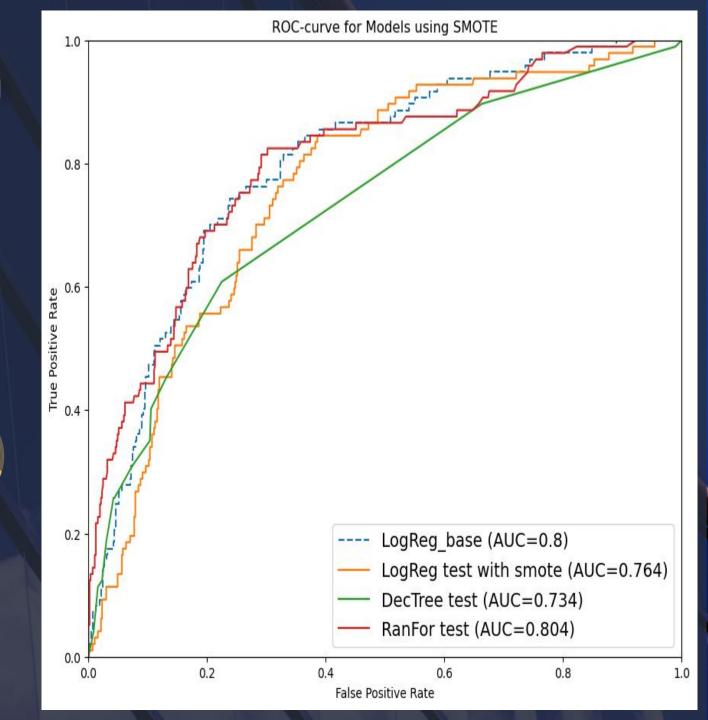
This are the results of my models explained by the ROC-AUC curve

76.4 %

Logistic regression model with smotel 80%

Logistic REgression Baseline model 80.4

Random Forest regressor model



### Conclusion

The best model, The random classifier Model, achieved 80.4% accuracy. The most influential features were whether the client had an internal plan, the total minutes talked during the day, and the number of customer service calls. These features significantly contributed to the model's predictive power and overall performance



#### Recommendations

Recommendation 1

Focus marketing efforts on clients with internal plans to increase retention and satisfaction through targeted incentives.

Recommendation 2

Monitor and optimize daily call minutes, offering tailored plans for heavy users to improve customer experience.

Recommendation 3

Enhance customer service quality, reducing the number of calls by addressing common issues effectively and proactively.

#### **NEXT STEPS**



#### • Step 1

Integrating datetime data enables temporal analysis, tracking trends, and identifying seasonality in customer interactions for actionable insights.

#### • Step 2

Leveraging NLP models extracts sentiment and topics from customer feedback, correlating sentiments with churn, satisfaction, and preferences.

#### • step 3

Continuous monitoring, iteration, and automation ensure effectiveness, enabling refinement and optimization of analysis and modeling processes.

