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# Introductio

Customer churn refers to the loss of customers or subscribers for any reason at all.

- The telecommunications industry is competitive and characterized by very high churn rates.
- Attracting new customers is both difficult and expensive hence avoiding churn, should be given a higher priority than trying to attract new customers.





#### Problem Statement

- Syriatel has a significant problem with customer churn.
- To solve this problem, it needs to identify the customers who are likely to churn and try to keep them.
- Machine learning models can help telco's in predicting the customers who are most likely to churn, based on various factors such as customer usage patterns, payment history, and demographics.
- The goal is to build a classifier to predict whether a customer will ("soon") stop doing business with Syriatel so that the company can take action to retain them.



#### Data used

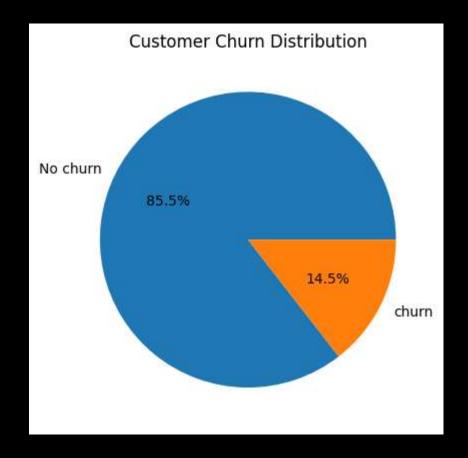
- The data set was obtained from Kaggle.
- It contained features such as customer demographics, their usage patterns and subscriptions and payments which can be used to predict customers who are most likely to churn.
- The dataset comprises 21 columns with a total of 3333 entries.





### Customer churn distribution

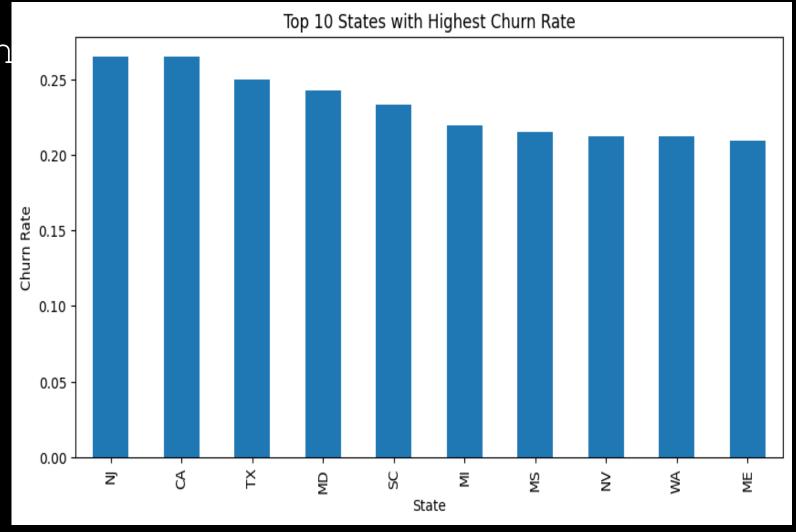
From the chart: 85.5% of customers did not churn (represented by the blue section), meaning they continued to use the product or service. 14.5% of customers did churn (represented by the orange section), meaning they stopped using the product or service



State and churn rates

The top 5 states with the highest churn rate are:

- NJ (New Jersey)
- CA (California)
- TX (Texas)
- MD (Maryland)
- SC (South Carolina)





The data was trained using the following machine learning models:

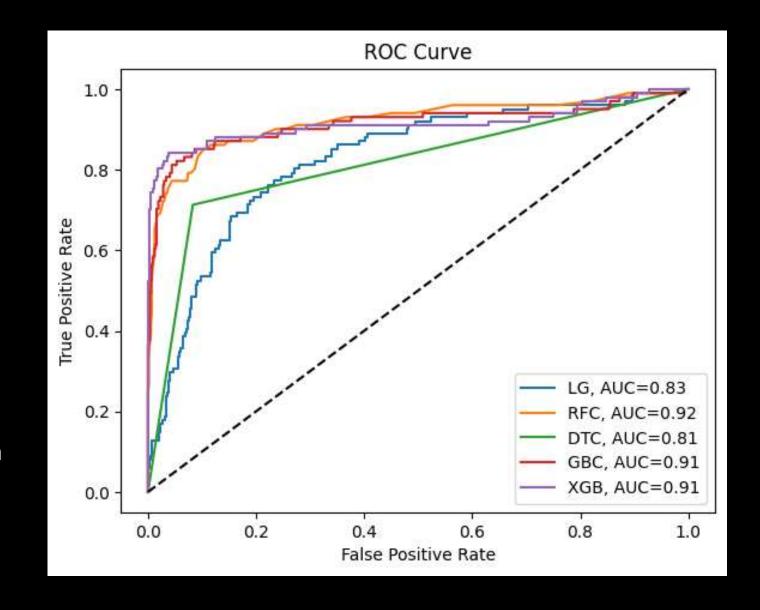
- Logistic Regression.
- Random Forest Classifier.
- Decision Tree Classifier.
- XGB Classifier.



### Model evaluation

- •Logistic Regression (LG): 0.83
- •Random Forest Classifier (RFC): 0.92
- •Decision Tree Classifier (DTC): 0.81
- •Gradient Boosting Classifier (GBC): 0.91
- •XGBoost Classifier (XGB): 0.91

Random Forest Classifier (RFC) with an AUC of 0.92 is the best model to predict customer churn among the compared models





#### Model Evaluation

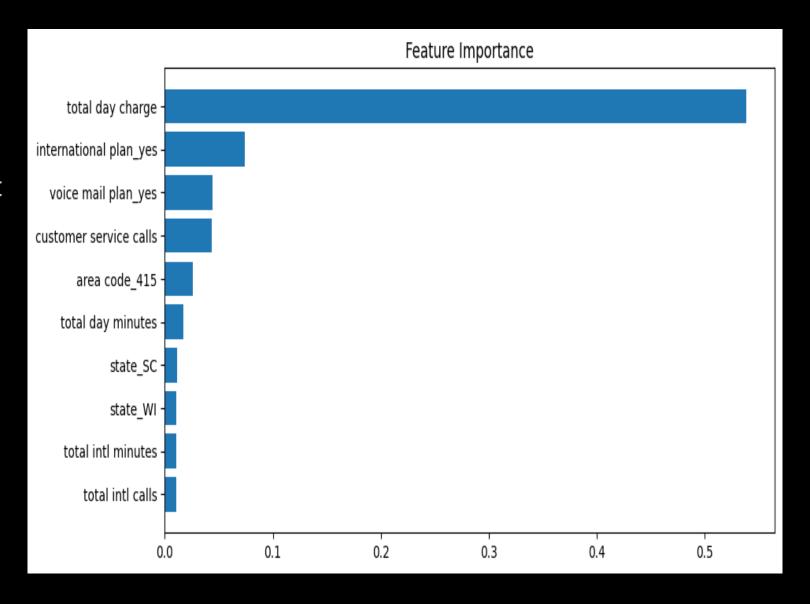
• The XGBoost model achieved a recall of 0.79 for churned customers. This means it correctly identified 79% of the actual churn cases.

 The XGBoost Classifier ROC score of 0.91 indicates a high level of accuracy in predicting customer churn. This means that the model is able to correctly identify 91% of the cases where customers are likely to leave the service

## Top five features

The top 5 features that affect customer churn include:

- Total Day Charge
- International Plan
- Voice Mail Plan
- Area Code (415)
- Customer Service Calls





Recommendati ons

- XGBoost for predicting customer churn
- Enhance international service offerings and review pricing.
- Improve voice mail services and offer tutorials or assistance.
- Customer care team should strengthen customer support, resolve issues effectively, and enhance service quality.
- Improve network quality

