

The goal of this project is to build a classifier that can predict whether a customer will soon stop doing business with SyriaTel.

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SYRIATEL CUSTOMER CHURN ANALYSIS

Overview

This project is addressing a major business challenge of Customer Churn at SyriaTel. Since 2000, the Company has been the leading Syrian mobile telecommunication. However, there has been an increased competition in the Syrian Telecommunication market. This has raised a major concern to the management and hence demanding an anlysis of the root causes of teh increased customer churn in the company.

Business Understanding

With the Increased competition in the Syrian mobile telecommunication market, SyriaTel needs to strategically address customer challenges to remain competitive within the market.

SyrialTel has identified an increased customer churn rate which has led in declined sale revenue and brand loyalty.

This project will develop a data-driven strategy to mitigate the risks of customer churn. The project will develop classification models that will predict the likelihood of a customer terminating business with SyriaTel.

Data Understanding

To uncover patterns related to customer churn at SyriaTel, this project has leveraged [Syriatel Dataset](#).

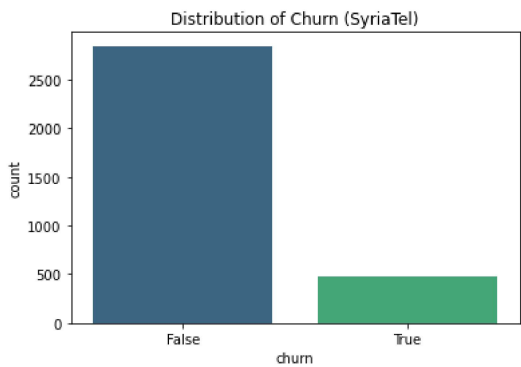
The dataset contains 3333 customer records on 21 features:

The features include:

1. Usage metrics: Day, even,night charges/charges
2. Customer service: the number of customer service calls
3. Service plans: No of Vmail messages, international plan, voicemailplan
4. Demographic and account info: Account length, state and AreaCode
5. Churn as the predictor/ Target variable

Exploratory Data Analysis

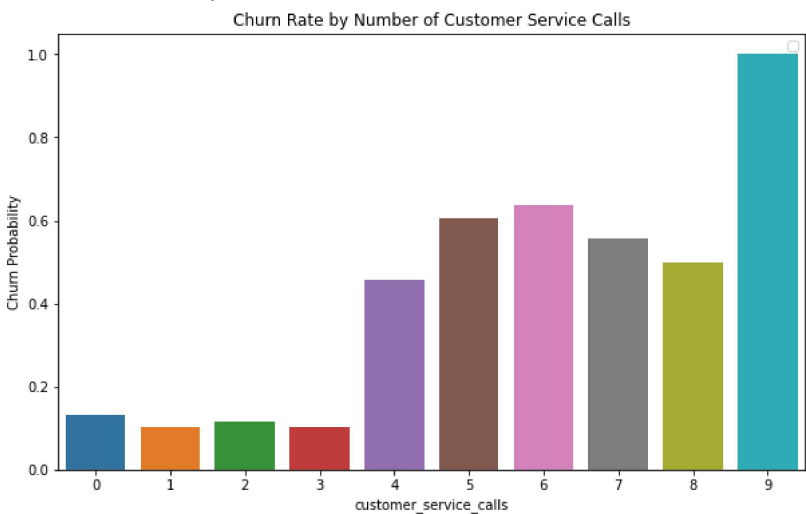
The image below is a illustration of the churn distribution across the dataset.



From the dataset, 85% of the customers stayed with the company however, around 15% of the customers churned SyriaTel. The project will address the challenge of the 15% of the customers who churned.

What is the relationship between Customer service calls and Churn?

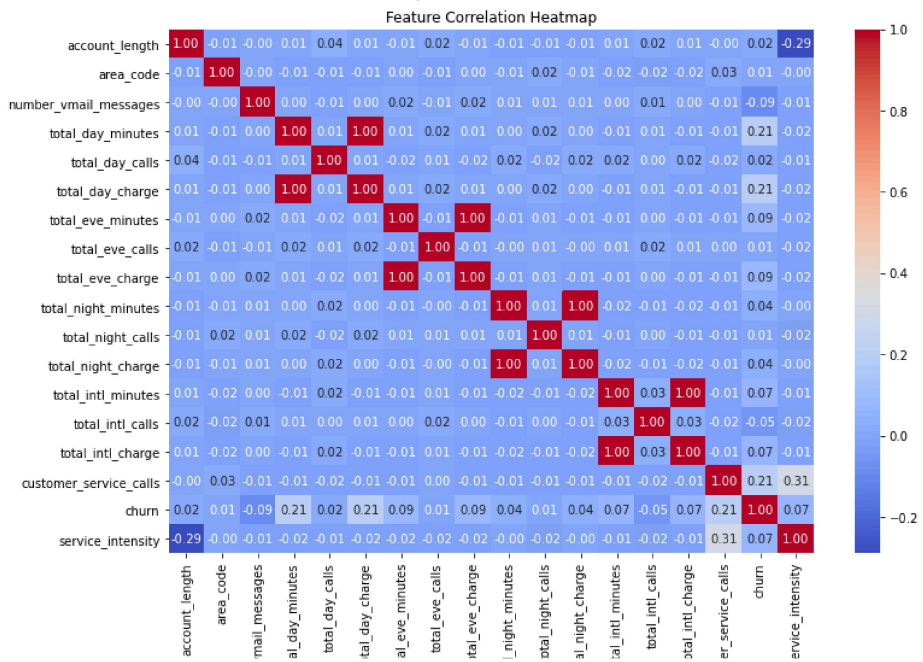
The image below illustrates the count of customer churns based on the number of customer service calls made to SyriaTel



The illustration shows that customers who make more than three customer service calls have an increased rate of customer churn.

Pre_processing for Machine Learning

Since highly correlated features can influence the accuracy of the models. Highly correlated Features had to be dropped before modeling the data.



README

The highly related columns and irrelevant columns were dropped i.e day/eve/night/intl charge, phone number,areacode, day/eve/night calls, account lenth, state

MODELING

This modeling process has followed an iterative approach moving from a simple baseline Logistic Model to complex Random forest models.

Pre_processing

To avoid data leakage, the data tranformations were wrapped in a Pipeline
The Pipeline includes:

- 1. One_Hot encoding: To convert the categorical features
- 2. Standard Scaling: To normarize numerical features.

Model Iterations

Baseline: Logistic Regression : This was the modeling start point. The model had an accuracy of 77% without overfitting . However, the model had a very low precision rate of 36%.

Decision Tree: As compared to the baseline model, Decision Tree, moved the Recall to 81.4%. The model increased precision to 73.4% which is sound and with an increased accuracy of 93.1% .

Tuned Decision Tree: Hyperparameter tuning with GridSearchCV. Tuning the model increased recall to 86.6% however with a significant drop in Precision to 47%.

Random Forest: As compared to the Tuned Decision Tree, for Random Forest, leveraged the F1 scoring metrics.

The table below is a summary of the scoring metrics

Model	Recall	Precision	Accuracy
Logistic Regression	0.75	0.36	0.77
Decision Tree	0.81	0.73	0.93
Decision Tree(Tune	0.87	0.47	0.84

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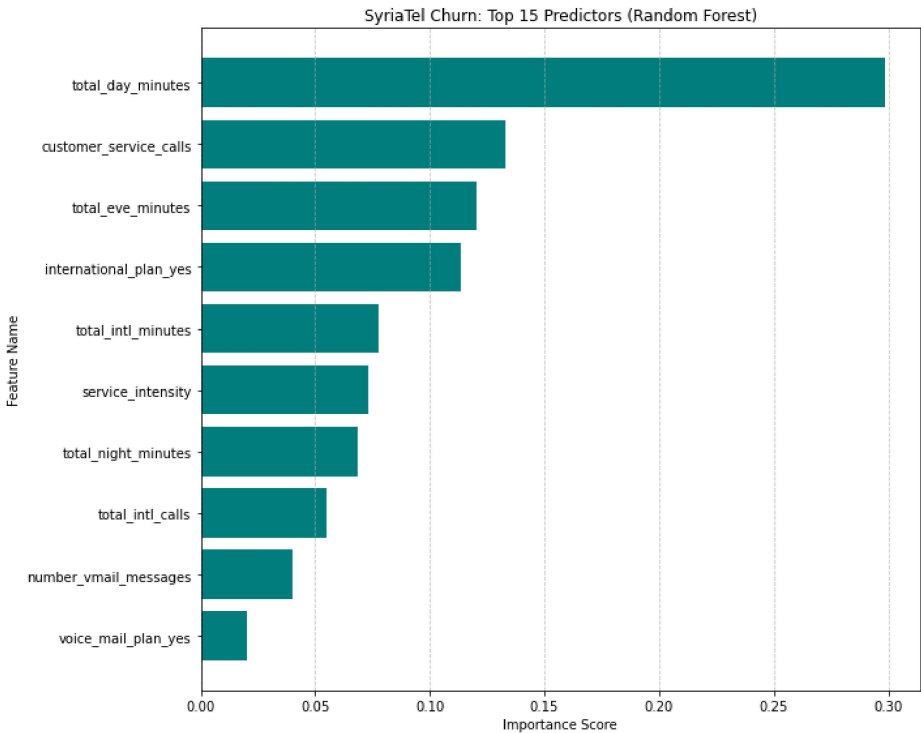
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EVALUATION

The Random Forest Model has the best performance and was able to generalize the dataset. Precision of 93.8% . This implies that out of 100 customers flagged to be at a churn risk 94% of the customers are actually churners. Recall: The model generalizes over 77% of the total as churners The model has an accuracy of 96% as compared to the baseline logistic model (77%) The rationale besides considering Precision as the best metric is that a high precision indicates that SyriaTel will not require to make unnecessary promotions and offers on the non churn customers and hence making the huge investment on churn customers. The model highlights the following as the top 10 indicators of customer churn



Conclusion

Syriatel should enforce a “3rd Call emergency” Rule to address the increased customer churn rate on customers with more than 3 customer service calls.

- Daytime_minutes has the greatest predictor of customer churn. The company can implement a “High volume” specialized bundle to help daytime users reduce their per-minute costs. This could decrease the probability of churn on day_time_minute customers.
- Customers with international plans have a churn risk. Syriatel needs to evaluate the value proposition for international plan customers.

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