



Customer Churn Prediction

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Agenda

- ❖ Overview
- ❖ Project workflow
- ❖ Business Understanding
- ❖ Analysis
- ❖ Modeling
- ❖ Evaluation
- ❖ Conclusion
- ❖ Recommendations



Overview

- One of the major challenges facing telecommunication providers is customer churn – a scenario where users discontinue their service, mostly due to dissatisfaction from the provider.
- In order to mitigate this challenge, telecom companies are exploring churn prediction mechanisms, as well as the need to understand factors that contribute to customer churn.



Project Workflow

Recommendations

Conclusion

Evaluation



**Business
Understanding**

**Data
Understanding**

Modeling



Business Understanding

Business Problem



SyriaTel, a leading telecom provider, is experiencing a significant loss of customers who are choosing to leave its services for other competitors.

Project Objectives



- To determine the key characteristics and behavior patterns that contribute to customer churn.
- To build a robust predictive model that will identify customers with a high likelihood of discontinuing their service, with a recall of about 80%.
- To provide data-driven insights and recommendations that will proactively engage, and retain high-risk customers.

Data Understanding



- The dataset is a customer churn dataset from the telecom sector, obtained from Kaggle.
- It contains about 3333 records, and 21 columns, of which 4 columns are categorical and 17 columns are numerical.
- Some of the key features of the dataset include: **Call rates, State and Area Code, International and Voice Mail Plans, and Customer Service Calls.**

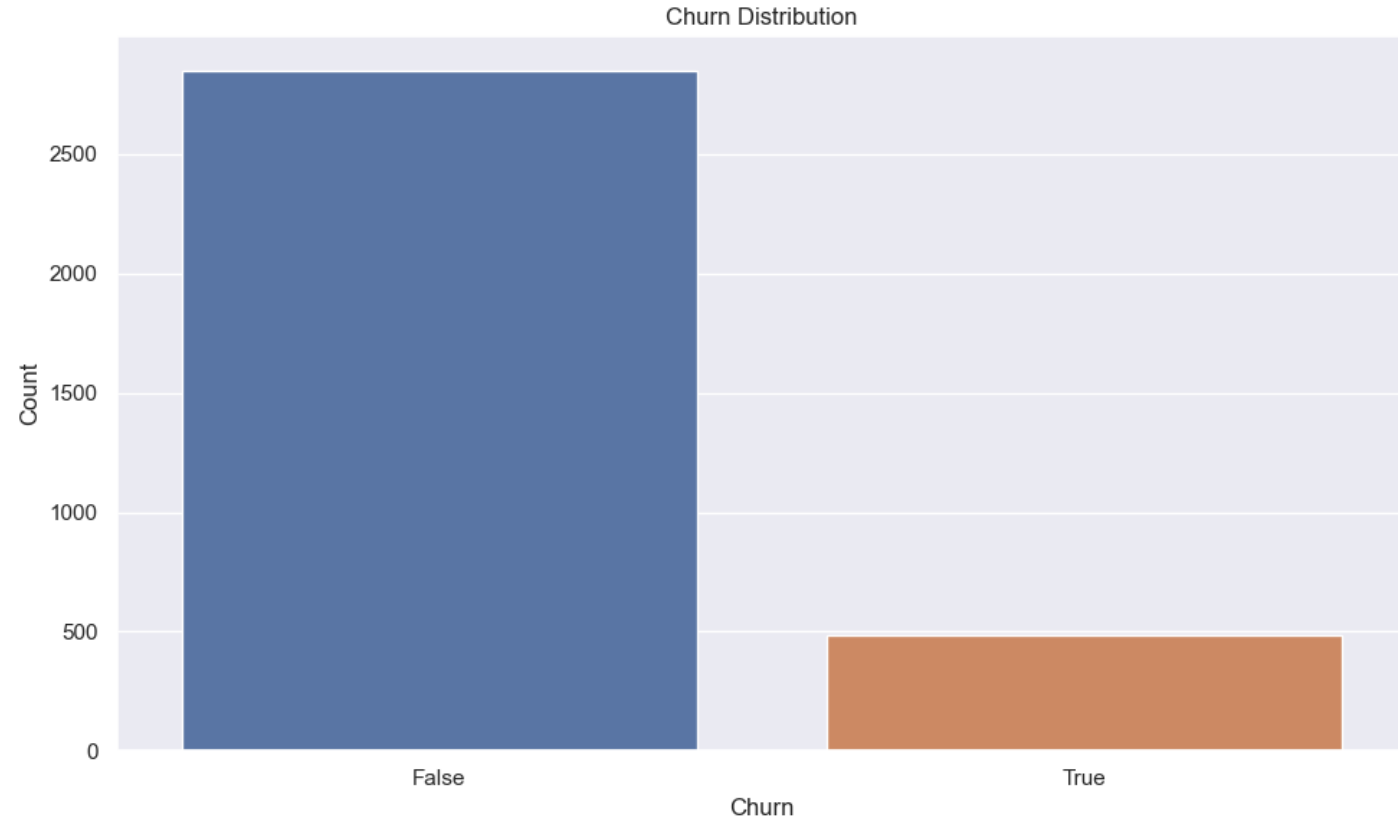


Analysis

Total number of customers
in the dataset; **3333**

Number of customers who
churned from the
company; **483**

Percentage of churned
customers; **14.5%**

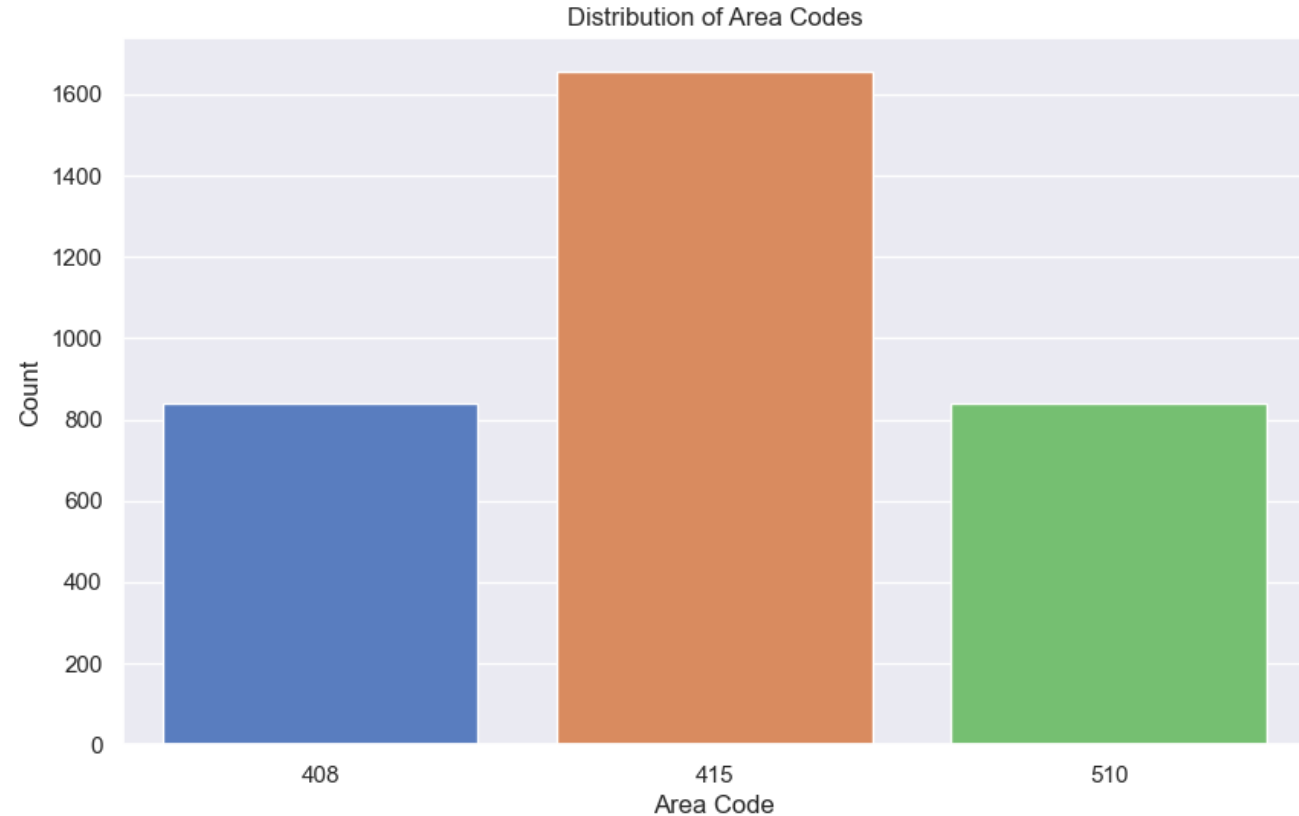


Distribution of the Churn target

- Area code 415 has a higher number of customers with about ***1655 customers***, which accounts for about 49.7%.
- Area codes 408 and 510 have a close number of customers:

Area 408: ***838 customers***

Area 510: ***840 customers***



Distribution of customers by Area Code

- Customers who churn (churn=True) tend to have more customer service calls than those who don't.
- Additionally, majority of the customers who churn come from area codes 415 and 510



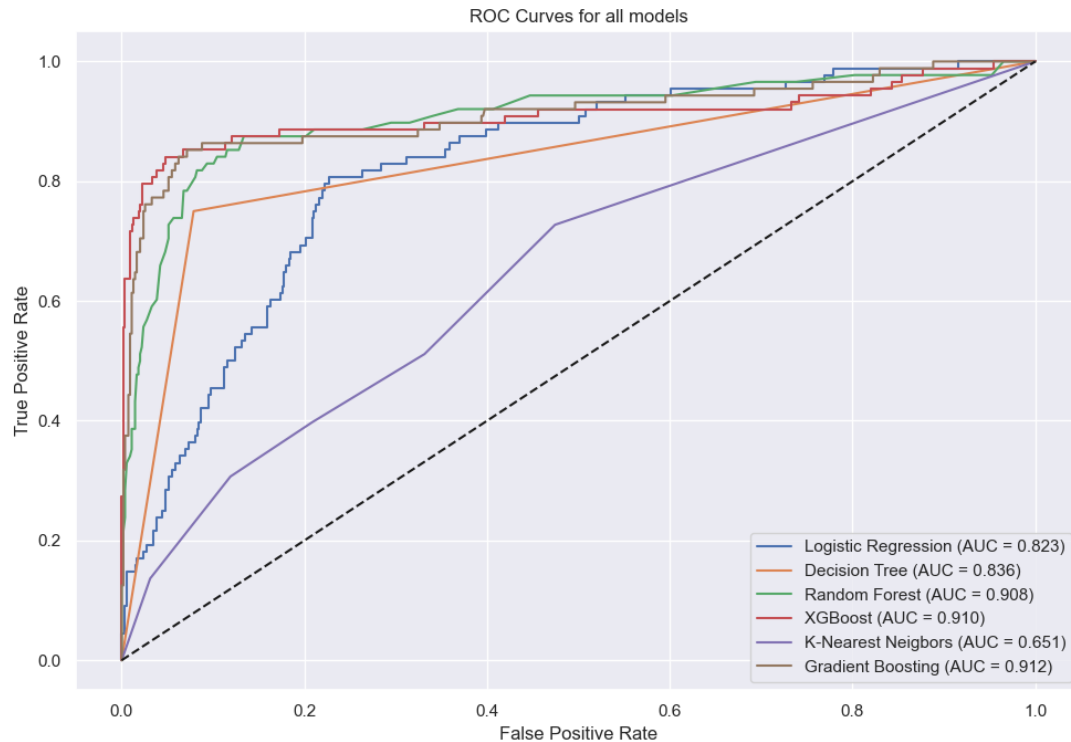
Relationship between customer service calls and the churn rate, by area code

Modeling

Algorithms used:



Evaluation



Top 3 models Based on Recall Score:

- Gradient Boosting - 0.807
- XGBoost – 0.795
- Decision Tree – 0.75

Top 3 models Based on ROC_AUC curve:

- Gradient Boosting – 0.912
- XGBoost – 0.910
- Random Forest – 0.908

Model Tuning Results



- The tuned Gradient Boosting model achieved a recall score of 0.81, which was similar to the untuned model score, and an AUC score of 0.921, which was an improvement from the AUC score of the untuned model.
- The tuned XGBoost model achieved a recall score of 0.82, which was a significant improvement from the recall score of the untuned model, and an AUC score of 0.911, which was a slight improvement from the AUC score of the untuned model.

Conclusion

- The recall score of the XGBoost model was 0.82, which surpassed our initial target of 0.8.
- From my analysis, the most important features that contributed significantly to the model performance are *Customer Service calls, Total Day Charge and International Plan*.
- We achieved our objectives of identifying the key attributes that contribute to customer churn, and developing a customer churn prediction model with a recall score of 0.8 and above.

Recommendations



- Offer specialized discounts, loyalty rewards, or exclusive promotions in high-churn area codes such as 415, which will serve as an effective incentive to retain customers.
- Investing in comprehensive training programs for support staff, and implementing better issue/conflict resolution frameworks in order to enhance customer satisfaction, thus minimizing the rate of churn.
- Developing localized market efforts, personalized engagement strategies, and enhanced customer support in high-churn areas such as Texas and New Jersey, in order to strengthen customer loyalty and retention.



Thank you

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