# PREDICTING CUSTOMER CHURN

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## **OVERVIEW**



Business and Data Understanding



Modeling



**Evaluation** 



Recommendations



Next Steps

# BUSINESS AND DATA UNDERSTANDING



Problem: Better predicting when SyriaTel's customers will soon churn.



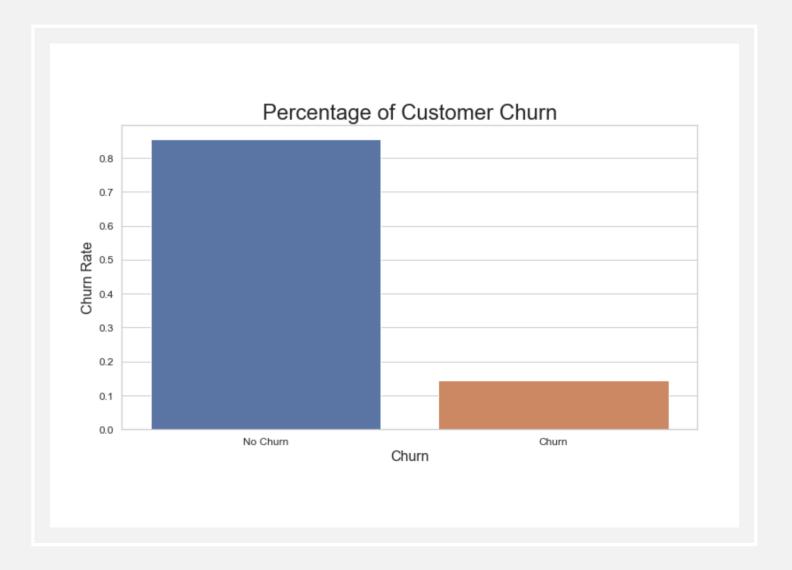
Solution: Finding predictable patterns using a classification model will benefit SyriaTel's business practices to minimize customer churn.

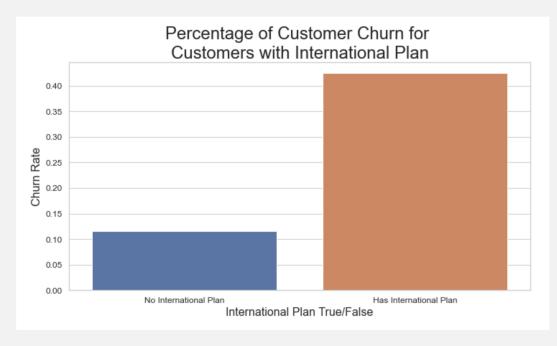


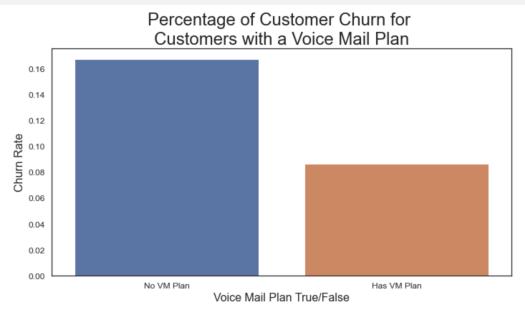
Data: Includes 21 features (both categorical and continuous and 3,333 data points from SyriaTel related to its customers and their accounts and churn information

# % OF CUSTOMER CHURN

 14.5% of the customers in this dataset have churned







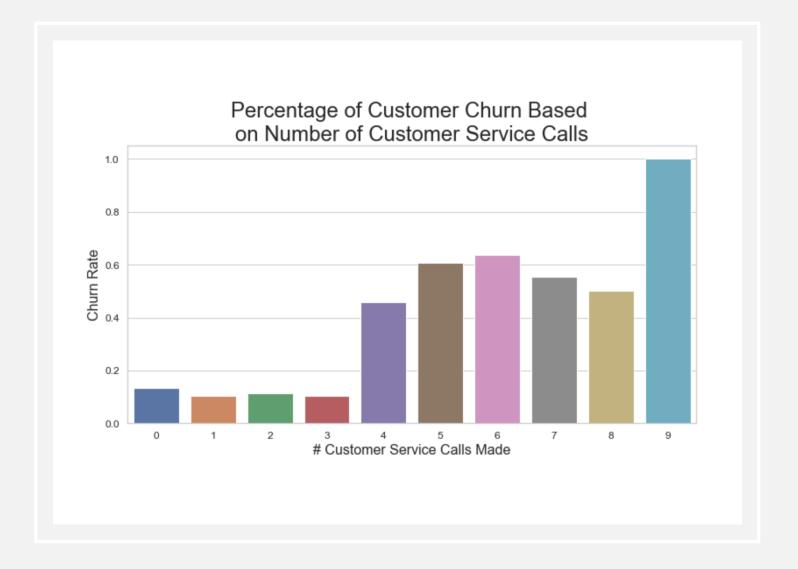
#### **FEATURES**

• 42% of customers with an international plan churn

17% of customers without a voice mail plan churn

# CUSTOMER SERVICE

 Customers who call customer services four or more times have a higher churn rate than those who call fewer than four times



#### MODELING

- Used a type of machine learning algorithm called Classification, which is the process of predicting the class of given data points
  - In this case, the "class" is customer churn whether a customer will leave SyriaTel
- Out of the five different algorithms evaluated in this project, the best performing model used an eXtreme Gradient Boosting (XGBoost) algorithm, which provides best-in-class performance among other classifiers



- Performance metric: F1-score, which combines two classifier metrics:
  - Precision measures what percent of the model's predictions were correct
  - Recall measures what percent of the positive cases were caught correctly
  - FI is the harmonic mean of the model's precision and recall scores
- F1-score values range 0-1, with 0 as the worst value and 1 as the best value
  - The closer the F1-score is to 1, the more perfect the model is classifying samples



- The final tuned XGBoost model achieved an FI-score of 0.94
- The most important features influencing the model:
  - the number of calls the customer made to customer service\*
  - whether the customer has a voice mail plan
  - the total number of minutes used per day
  - whether the customer has an international plan\*
  - the total number of international calls made
- Both bolded features were also influential in other models evaluated

#### RECOMMENDATIONS FOR FUTURE WORK



Provide a larger dataset with greater number of customer information and dates of data collection (to ascertain timeframes)



Further Analysis of Feature Importance: International plans, voice mail plans, calls to customer service



Evaluate Customer Service: Conduct customer service surveys for more information on why customers are calling



Conduct industry benchmarking to determine how voice mail and international plans compare to competitors

## THANK YOU





Questions?

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