# Reducing Churn to Increase Life Time Value

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# **Agenda**

- 1. Problem
- 2. Business and data understanding
- 3. Modeling
- 4. Results
- 5. Recommendations
- 6. Limitations

#### **Problem**

Are there any predictable patterns that determine churn for the Telecom Company?

This analysis seeks to *classify* customers who leave the service (churn) and determine which levers can be pulled to *increase customer retention*.

## **Business and Data Understanding**

#### The data:

- Telecom Data
- Includes 20 customer variables ranging from state to account length
- Omitted irrelevant variables such as phone number

Final variables include →

account length area code international plan voice mail plan number vmail messages total day minutes total day calls total day charge total eve minutes total eve calls total eve charge total night minutes total night calls total night charge total intl minutes total intl calls total intl charge customer service calls churn

# **Modeling**

Approached the data with 3 modeling types:

- Logistic Regression, Decision Forest Classifier, and XGBoost Classifier
- Created baseline models, then tuned parameters through Grid Search
- Evaluated feature importances of best performing model: XGBoost

### Recall

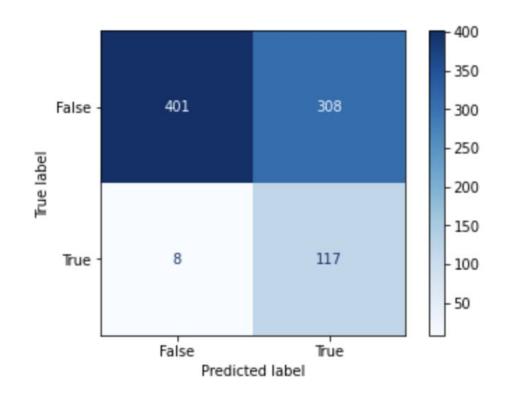
Optimized models for recall. This leaves us susceptible to false positives, but missing potential churners is a greater concern than misclassifying non-churners.

Recall = 
$$\frac{\text{Number of True Positives}}{\text{Number of Actual Total Positives}}$$

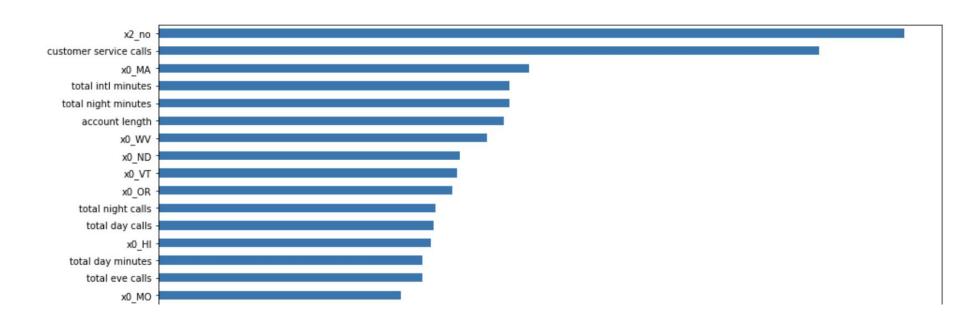
## **Results**

Recall: 0.936

Accuracy: 0.62



# **Feature Importance**



#### **Recommendations**

Our best performing model, XGBoost, identified the following features as most important to predicting churn:

- 1. Not having an international plan
- 2. Number of customer service calls
- 3. Being a MA customer

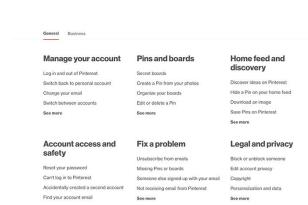
A tailored, automated email flow (Klaviyo) addressing common pain points to customers with more than 2 customer service calls.

- Assuage the most common concerns that customers have
- Likelihood of churn decreases



Melp Center

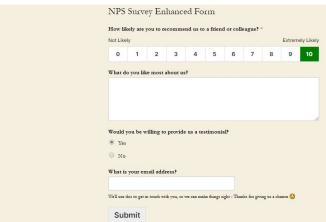
Q Log in Pinterest home



#### A comprehensive investigation of the MA market.

- 1. First, customers in MA should be sent a survey assessing their level of satisfaction with the service, as well as assessing their pain points and complaints.
- 2. These survey results should be analyzed to find why MA customers are more likely to churn.

  NPS Survey Enhanced Form



#### A renewed focus on customers in the international plan segment.

- Target international customer segment with a new marketing campaign designed to bring these higher lifetime value customers into our network.
- Drive revenue growth with better quality customers who spend more and churn less.

Customer Lifetime Value = Average Revenue per User / Churn





#### Limitations

Our final algorithm optimized for recall produces false positives at a higher rate than the baseline.

- False positives may produce inefficiencies at scale, since the model may produce many false positives when used on large data sets.
- Problems may include wasted marketing communications and customer annoyance.

# **Questions**