



# Telecom Churn Analysis Capstone Project

By Matthew Smith



## Problem Statement

A california telecom company is faced with customer churn of 26%.

**What are some potential solutions to reduce churn?**



## Findings

- Customers indicated the top 3 reasons they churned were:
  - Competitor has better devices
  - Competitor making a better offer
  - Attitude of Customer Support
- The most predictive feature of a customer's likelihood to churn is tenure



## Suggestions

- Add contract terms
- Offer lower priced plans
- Re-train customer support



# Data Wrangling

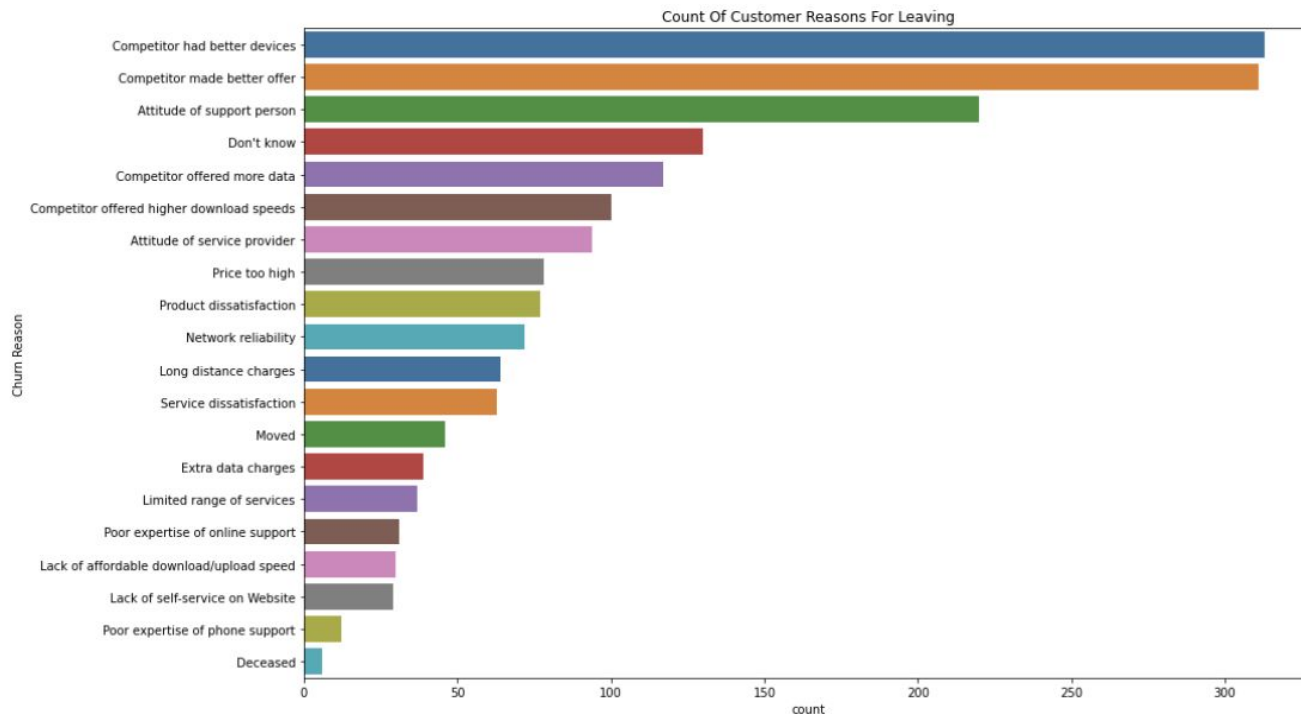
Original Dataset:

<https://www.kaggle.com/datasets/shilongzhuang/telecom-customer-churn-by-maven-analytics?resource=download>

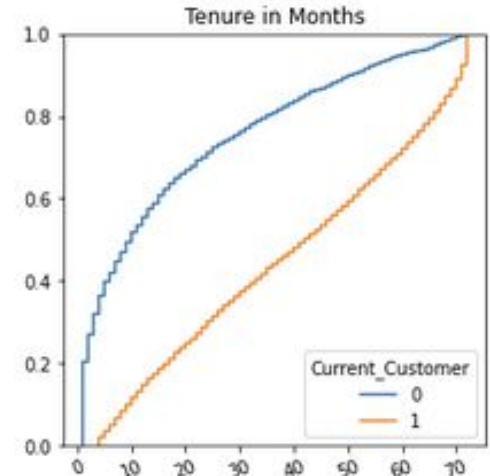
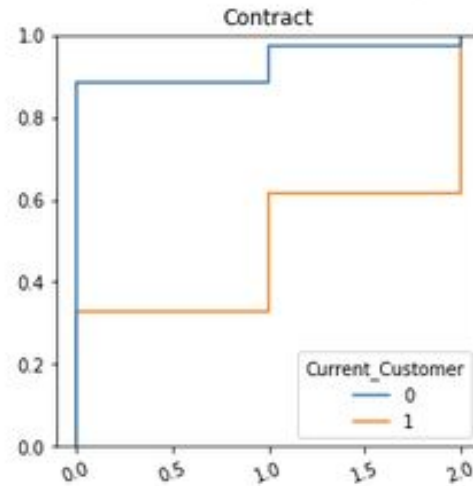
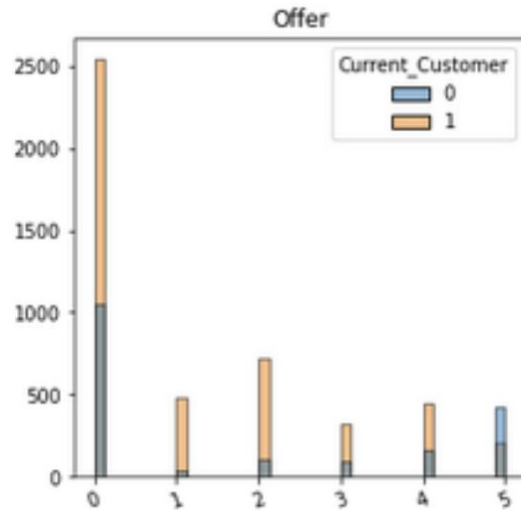
Key steps taken:

- Imputed missing values
- Created numeric categorical variables for appropriate categories

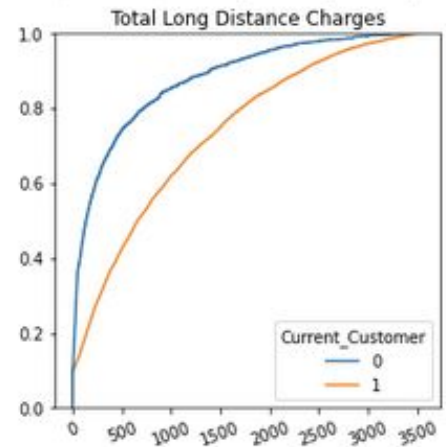
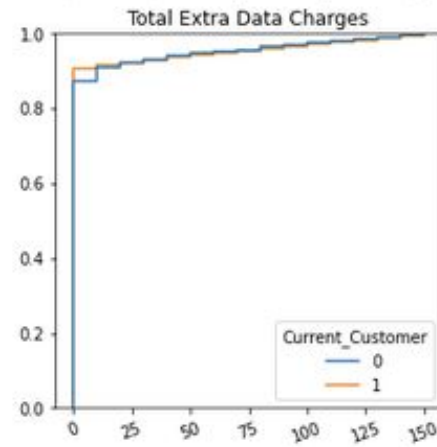
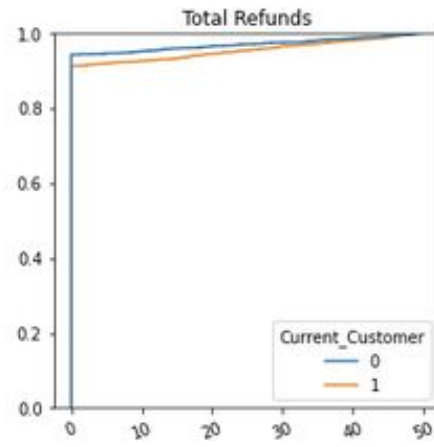
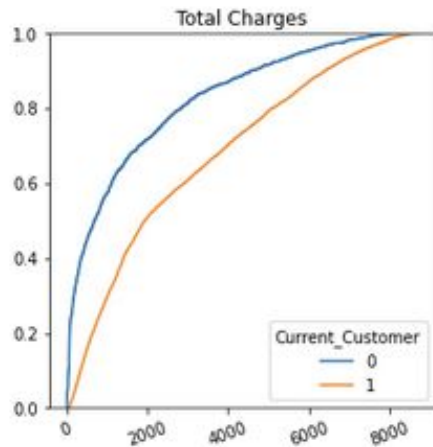
# Reasons Given For Customer Churn



# Exploratory Data Analysis

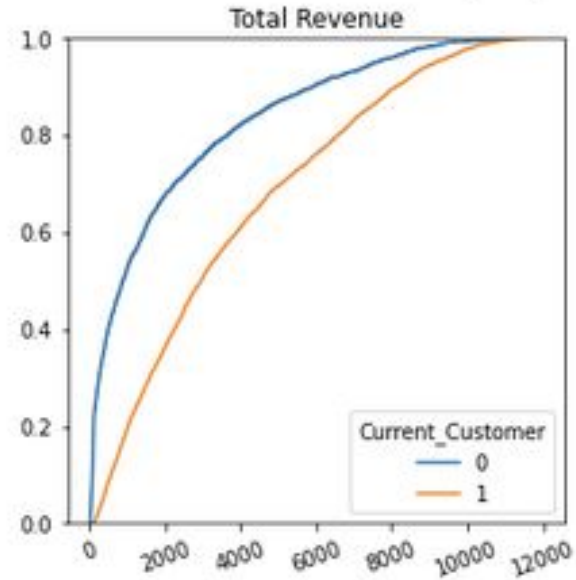
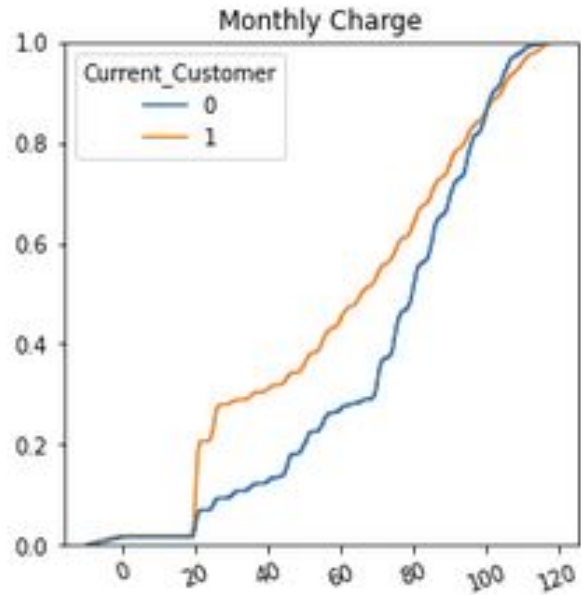


## Exploratory Data Analysis (Continued)

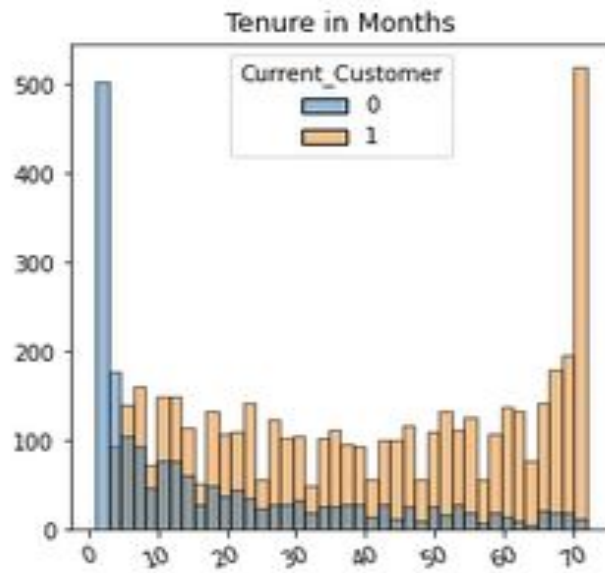
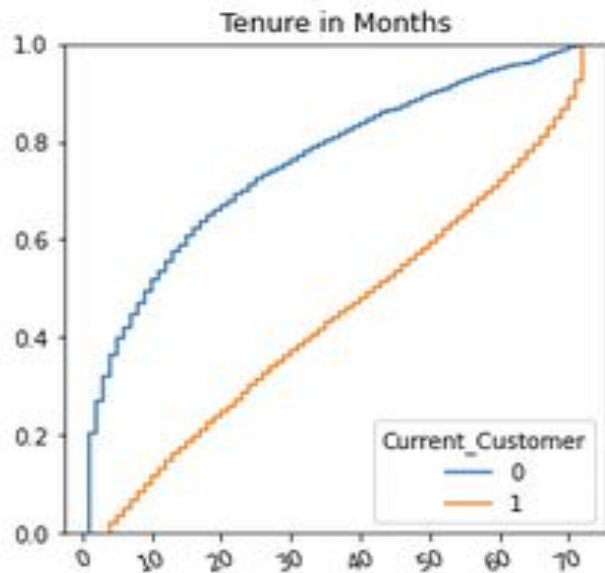




## Exploratory Data Analysis (Continued)



# Customer Tenure

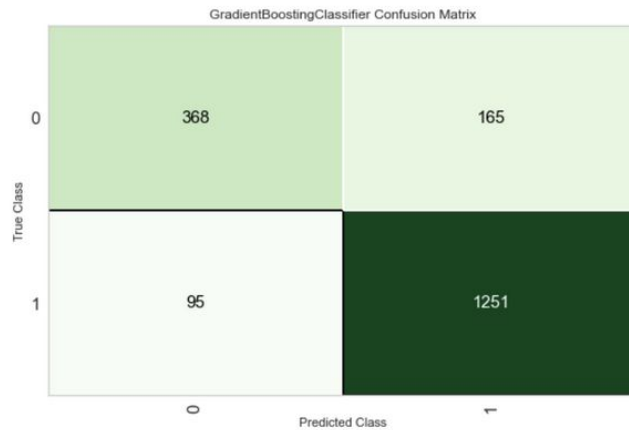
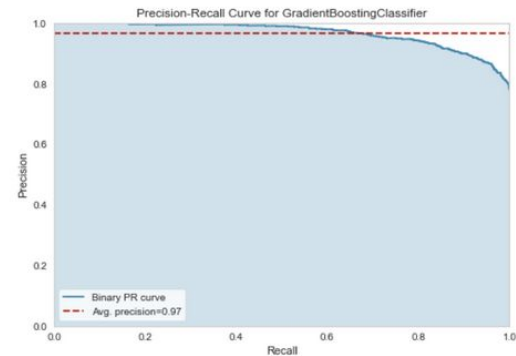
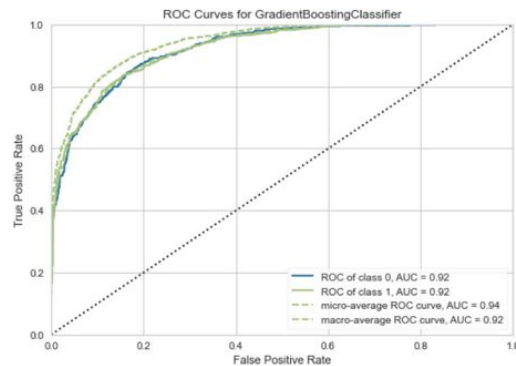




# Model Selection

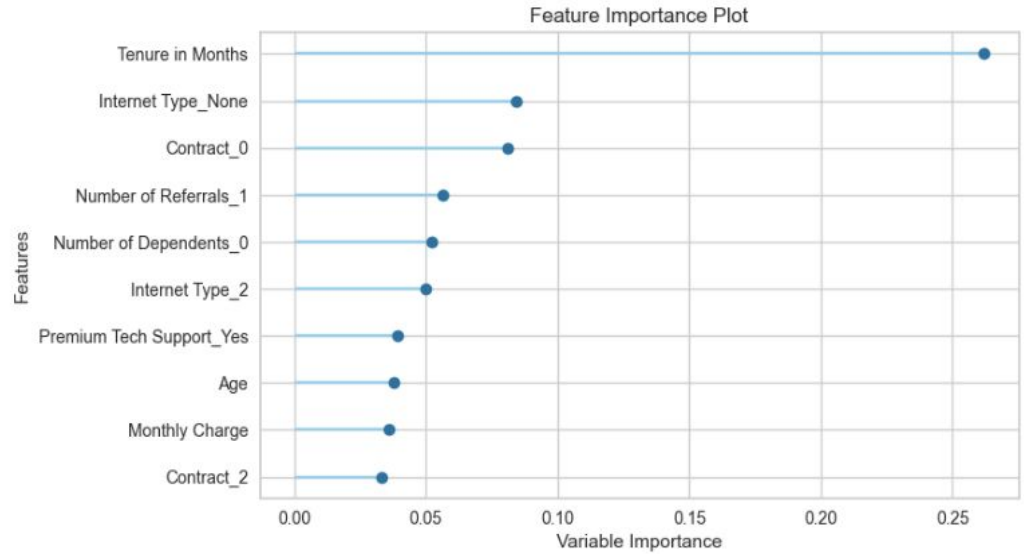
	Model	Accuracy	AUC	Recall	Prec.	F1	Kappa	MCC	TT (Sec)
<b>gbc</b>	Gradient Boosting Classifier	0.8774	0.9377	0.9464	0.8896	0.9171	0.6830	0.6885	0.2520
<b>lightgbm</b>	Light Gradient Boosting Machine	0.8742	0.9329	0.9356	0.8938	0.9141	0.6794	0.6823	0.0530
<b>rf</b>	Random Forest Classifier	0.8717	0.9252	0.9480	0.8819	0.9137	0.6652	0.6723	0.1220
<b>ada</b>	Ada Boost Classifier	0.8678	0.9265	0.9244	0.8946	0.9091	0.6668	0.6689	0.0800
<b>et</b>	Extra Trees Classifier	0.8637	0.9218	0.9298	0.8856	0.9071	0.6516	0.6547	0.1110
<b>lda</b>	Linear Discriminant Analysis	0.8439	0.9075	0.8900	0.8919	0.8908	0.6168	0.6176	0.0310
<b>ridge</b>	Ridge Classifier	0.8432	0.0000	0.9005	0.8830	0.8915	0.6085	0.6096	0.0220
<b>dt</b>	Decision Tree Classifier	0.8206	0.7800	0.8740	0.8752	0.8745	0.5597	0.5602	0.0230
<b>knn</b>	K Neighbors Classifier	0.7466	0.7117	0.8753	0.7927	0.8318	0.3229	0.3303	0.0320
<b>nb</b>	Naive Bayes	0.7188	0.8416	0.6735	0.9106	0.7738	0.4234	0.4584	0.0230
<b>dummy</b>	Dummy Classifier	0.7158	0.5000	1.0000	0.7158	0.8344	0.0000	0.0000	0.0170
<b>svm</b>	SVM - Linear Kernel	0.6743	0.0000	0.7467	0.8131	0.7435	0.2242	0.2674	0.0380
<b>qda</b>	Quadratic Discriminant Analysis	0.6423	0.6401	0.6456	0.8174	0.7054	0.2550	0.2713	0.0340

# Model Performance



# Takeaways

- Customer Tenure is the most important feature for predicting churn
- Customer stated reason for churn was competitor offer & device quality





# Recommendations

- Add contract terms to all plans
  - Reason: Customer tenure is a major factor in churn
    - Pros: Forces customer retention until end of contract, may make turnover more predictable
    - Cons: May be less attractive to potential customers
- Offer lower priced plans to beat competitors
  - Reason: Competitor made better offer is 2nd most common customer reason
  - Pros: More attractive to potential customers
  - Cons: Reduces profit per customer, may still result in churn



## Recommendations (Continued)

- Improve customer support quality
  - Reason: Customer Support attitude was the 3rd most common reason given for customer churn.
  - Pros: More satisfied customers, potentially higher retention rates of customers
  - Cons: Cost of training & cost of monitoring customer support



## Future Research

- Profile & features of high revenue customers
- Profile & features of new customers
- Market analysis of competitor offerings
- Impact of contract terms on customer retention





# Questions & Answers