CUSTOMER CHURN ANALYSIS AND SERVICE OPTIMIZATION



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Machine Learning Project
Phase 3
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○VERVIEW TELECOM COMPANIES FACE REVENUE LOSSES FROM CUSTOMER CHURN.

UNDERSTANDING CHURN FACTORS IS CRUCIAL FOR PROFITABILITY.

This project's goals:

Identify Customer Churn rate

Evaluate the current call charges
for different plans

Choose appropriate models for predicting churn

Segment Customers Based on Service Usage Patterns, Satisfaction Levels, and Likelihood of Churn.

DATA UNDERSTANDING



- state 3333 non-null object
- account length 3333 non-null int64
- area code 3333 non-null int64
- phone number 3333 non-null object
- international plan 3333 non-null object
- voice mail plan 3333 non-null object
- number voicemail messages 3333non-null int64
- total day minutes 3333 non-null float64
- total day calls 3333 non-null int64
- total day charge 3333 non-null float64
- ♦ total eve minutes + ওপ্ত non-null

- total eve calls 3333 non-null int64
- total eve charge 3333 non-null float64
- total night minutes 3333 non-null float64
- total night calls 3333 non-null int64
- total night charge 3333 non-null float64
- total intl minutes 3333 non-null float64
- total intl calls 3333 non-null int64
- total intl charge 3333 non-null float64
- **
- customer service calls 3333 nonnull int64
- churn

STATISTICAL SUMMARY

Descriptive statatistics.

Account Length: Mean - 101, Min - 1, Max - 243

Voicemail Messages: Mean - 8, Min - 0, Max - 51

Day Minutes: Mean - 179, Min - 0, Max - 350

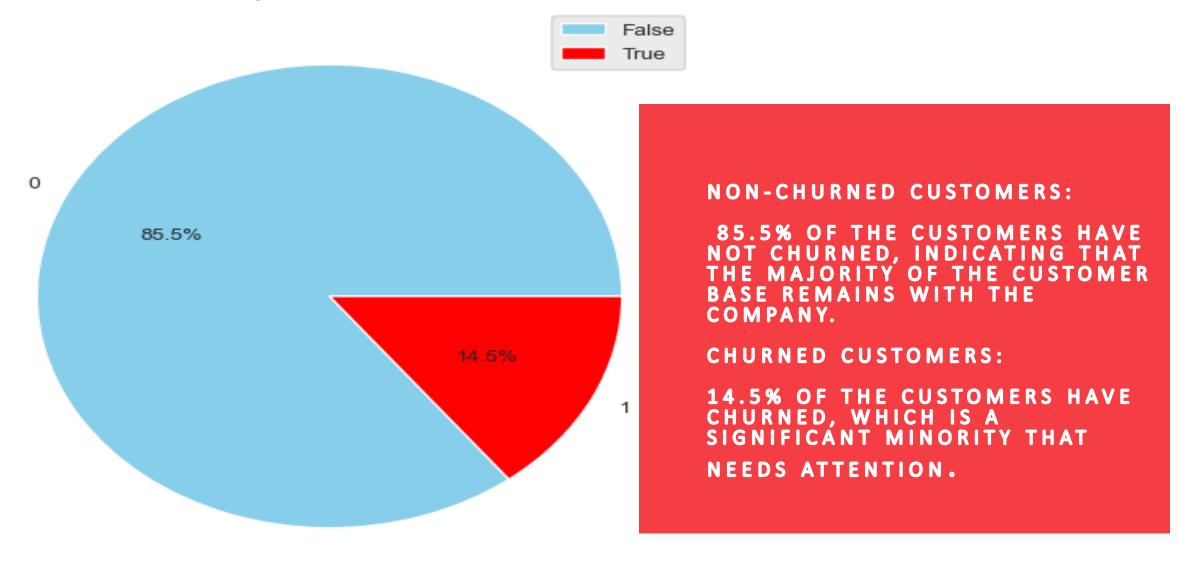
Evening Minutes: Mean - 200, Min - 0, Max - 363

Night Minutes: Mean - 200, Min - 23, Max - 395

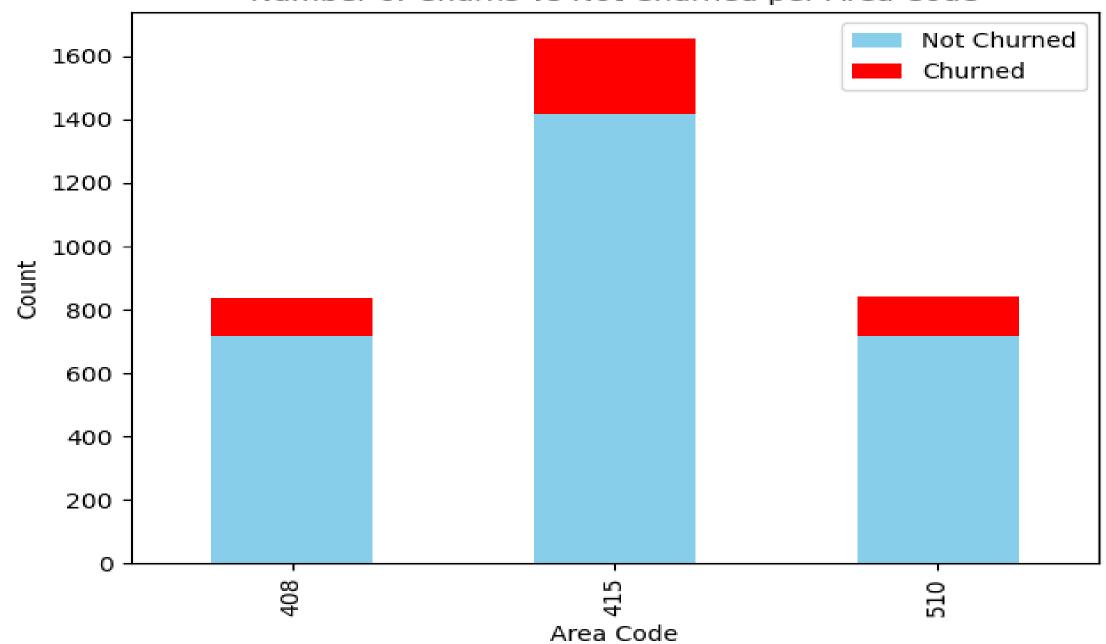
International Minutes: Mean - 10, Min - 0, Max - 20

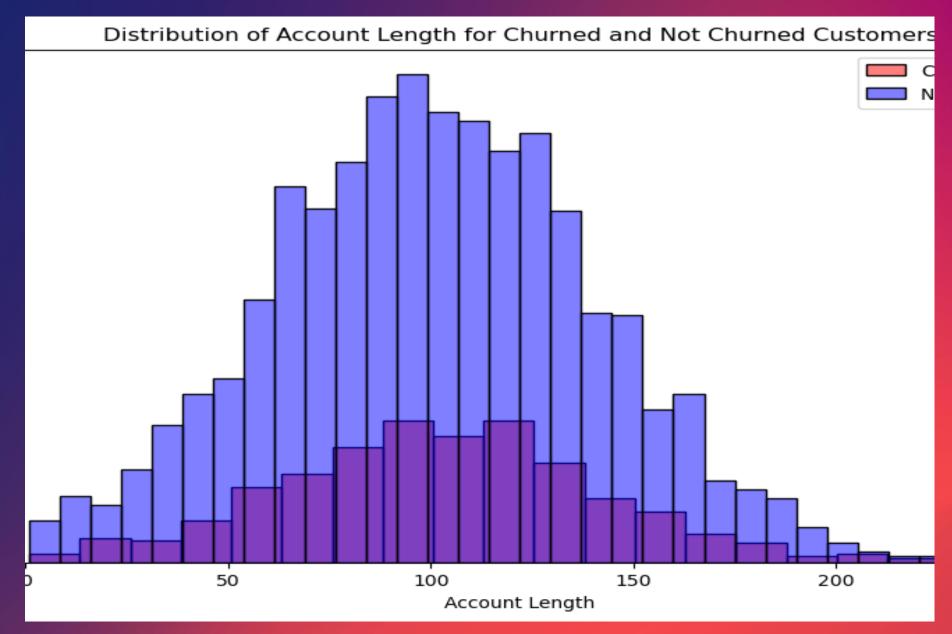
Customer Service Calls: Mean - 1.6, Min - 0, Max - 9

Proportion of Churn

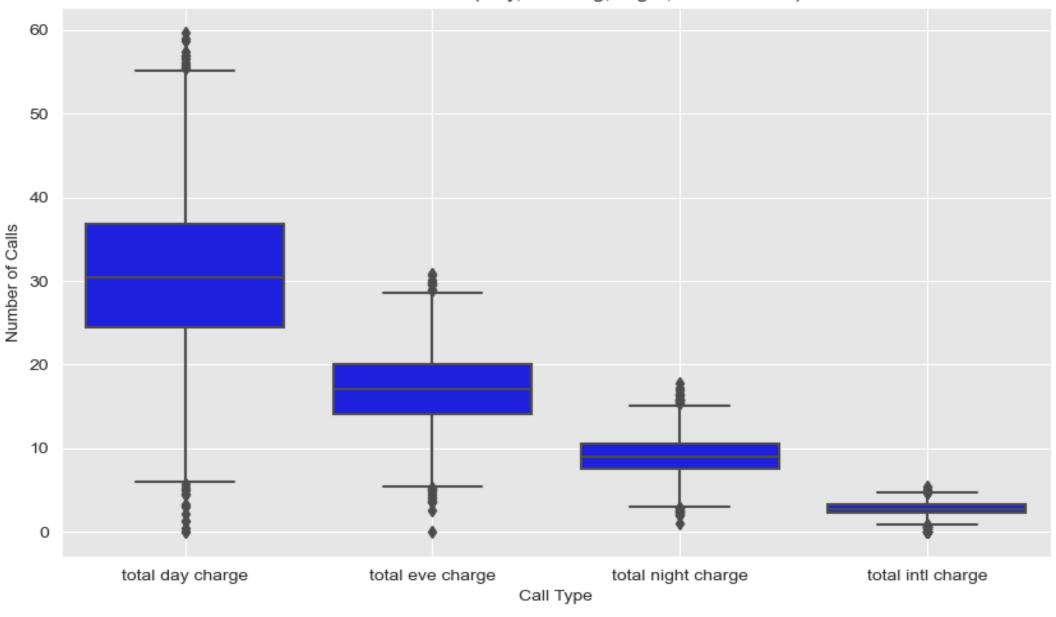


Number of Churns vs Not Churned per Area Code





Box Plot of Calls (Day, Evening, Night, International)



MODELS

LOGISTIC LINEAR

A baseline model to establish a benchmark for other models.

KNN MODEL

providing additional insights into customer
behavior and further validating the model's
effectiveness

DECISION TREE

Used for their interpretability and ability to capture non-linear relationships

RANDOM FOREST

An ensemble method that improves the performance and accuracy over single decision trees.



MODEL RESULTS

KEY METRICS

	PRECISION	RECALL	F1 SCORE	ACCURACY
Logistic	0.95	0.72	0.82	0.74
Regression	0.23	0.69	0.35	
KNN Model	0.96	0.79	087	0.78
	0.31	0.74	0.43	
Decision	0.97	0.89	0.93	0.88
Tree Tuned	0.49	0.81	0.61	0.00
Random	0.98	0.96	0.97	0.94
Forest	0.70	0.82	0.76	
8/06/20XX		PITCH DECK		10

MODEL EVALUATION.

For class label 0, all models perform relatively similarly across precision, recall, and F1-score, with scores close to or at 1.0, indicating very high performance.

For class label 1, there is more variation in performance. The Random Forest and Decision Tree models appear to have the highest precision, recall, and F1-score for this class, while the Untuned Decision Tree and Logistic Regression models have the lowest scores.

The KNN model has moderate performance for class label 1, with scores that are lower than the Random Forest and Decision Tree but higher than the Untuned Decision Tree and Logistic Regression

OBSERVATIONS

- Overall, the Random Forest and Decision Tree models seem to be the best performers for both classes, with the Untuned Decision Tree and Logistic Regression models being the least effective, particularly for class label 1. The KNN model falls in between these two extremes.
- The significant drop in performance for the Untuned Decision Tree compared to the tuned Decision Tree suggests that tuning hyperparameters is important for achieving good performance.
 - However between the Random Forest and Decision Tree, Checking the accurracy and ROC curve scores:Random Forest accurracy 94%,Decision tree accurracy 88%, The random Forest performs way better on test unseen or new data. Also, The consistent cross validation score performance across folds suggests that the model complexity is well balanced, allowing it to learn the patterns in the data without being overly complex.
- The best model to go by is Random Forest.

RECOMMENDATIONS

1. Conduct detailed analyses to understand the primary factors driving customer churn, this could include surveys and customer feedback. Since Area Code 415 has the highest number of churned customers, it would be beneficial to conduct a detailed analysis to understand why customers are leaving.

2.Reevaluate the voicemail plan's importance and consider repositioning it as an add-on service rather than a standard offering.

3.Implement loyalty programs to reward long-term customers and accounts above a length of 80. incentivize them to stay.



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PITCH DECK

RECOMMENDATIONS

The charges for day, evening, night, and international usage are relatively stable. Regularly review and adjust pricing to remain competitive. Especially Optimized Call Charges for each plan and time of the call.

Evening charges showed the widest range, suggesting potential for targeted plans to optimize revenue.

.Based on the evaluation metrics , the best performing model is Random Forest thus use it to make customer churn predictions.

