MARCH 07, 2024

#### INTRODUCTION

A Project by Alice Mumbi.

## Overview

Predicting Customer Churn in a Telecommunications Company

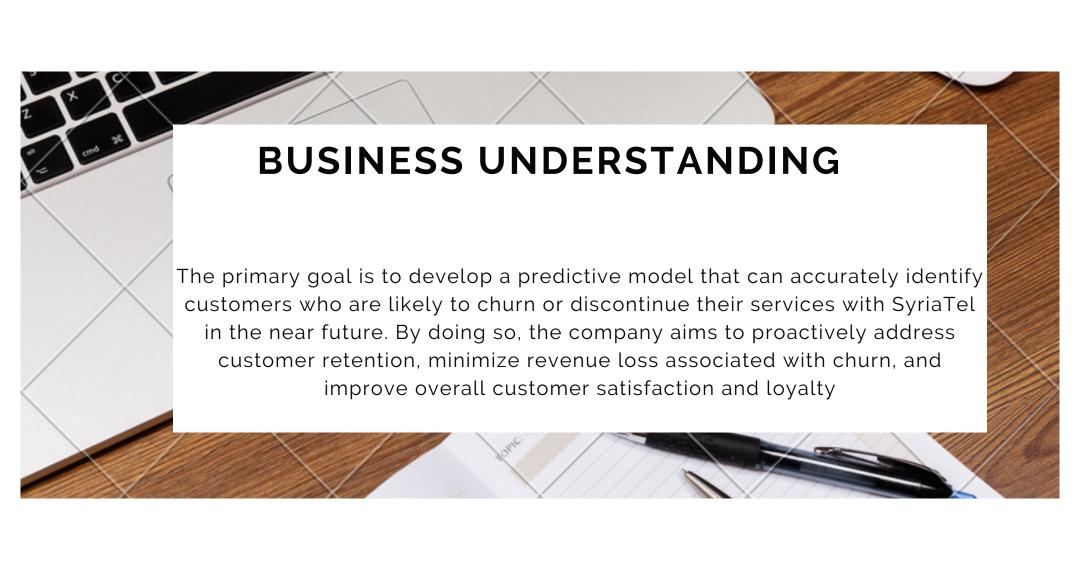


## Dataset Used

**Kaggle Dataset** 



Identify factors influencing churn
Build a classifier to predict customer churn

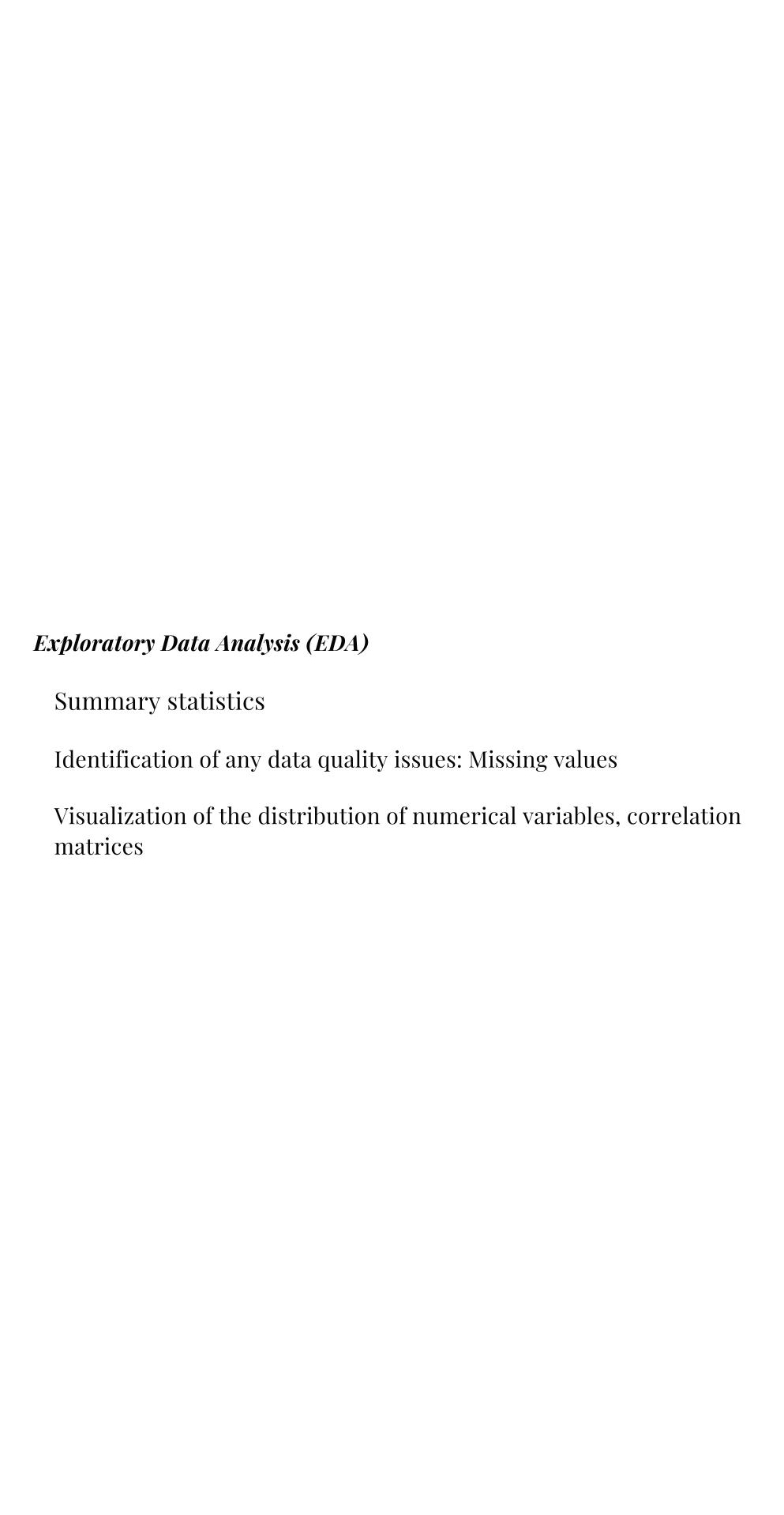


#### DATA UNDERSTANDING



- Overview of the dataset's structure, including the features and target variable.
- Description of the features available in the dataset.
- Importance of understanding the data before proceeding with analysis.





#### **Preprocessing**

Distribution of categorical variables

Handling of categorical variables: One-hot encoding

Feature Scaling

Feature Engineering







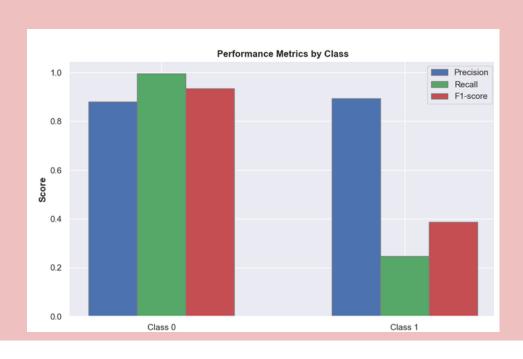
Selection of classification algorithms

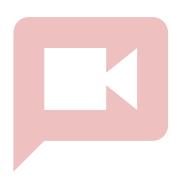
Comparison of model performance

## K-Nearest Neighbors

METRICS PERFORMANCE EVALUATION: ACCURACY, PRECISION, RECALL, AND F1-SCORE

The model correctly predicts the class label for about 88.2% of the instances in the testing dataset



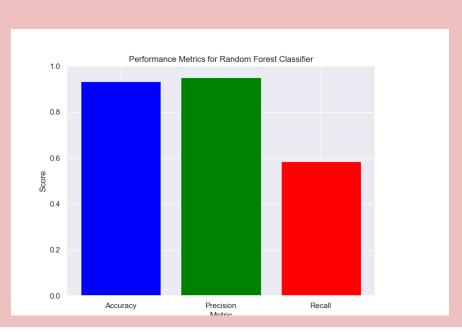


#### Hyperparameter Tuning and Optimization Identification of best hyperparameters

Techniques: GridSearchCV

## $Random\ Forest$ metrics performance evaluation: accuracy, precision, recall

The precision score of 0.9516 indicates that 95.16% of the instances predicted as churn were actually churn.
The recall score of 0.5842 means that the model correctly identified 58.42% of the actual churn instances





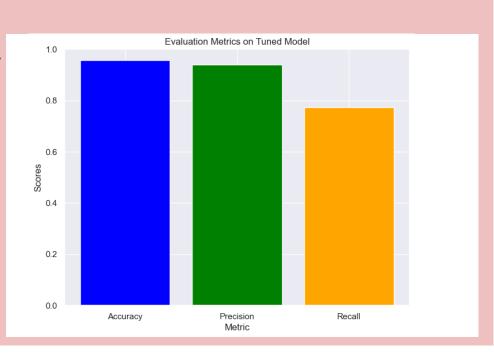
#### Hyperparameter Tuning and Optimization Identification of best hyperparameters

Techniques: Regularization

#### $\chi GBoost$

METRICS PERFORMANCE EVALUATION: ACCURACY, PRECISION, RECALL

We have successfully improved the Recall value therefore an improvement in correctly identifying all churn instances



## Impact on model performance



#### **GridSearchCV**

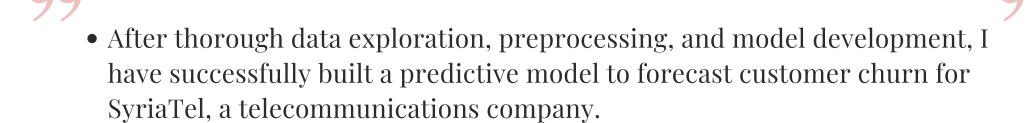
The model achieves an improved accuracy of approximately 88.4% on the training data



#### Regularization

Model improvement with an accuracy of approximately 98.46% on the training dataset and 95.80% on the testing dataset

# Conclusion



- The final model, an optimized XGBoost classifier, achieved an impressive accuracy of 95.8% on the testing dataset, indicating its effectiveness in predicting customer churn.
- Through feature engineering, model tuning, and evaluation, I have identified key factors influencing customer churn, including customer service calls, account length, and international plan subscription.



- Enhanced Service Quality Focus on improving service quality and customer satisfaction, particularly in areas identified as significant predictors of churn, such as customer service interactions and plan subscriptions.
- **Personalized Marketing Strategies** Leverage the predictive model to tailor marketing campaigns and communication strategies based on individual customer profiles and behavior patterns, thereby increasing engagement and loyalty.
- Continuous Model Monitoring and Improvement Establish a framework for ongoing model monitoring and evaluation to ensure its continued effectiveness over time. Regularly update the model with new data and refine its parameters as necessary to adapt to changing customer dynamics and market trends.

99

## THANK YOU

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