



TELCO CUSTOMER CHURN PREDICTION


BY RAKA R.A PRASETYO



BACKGROUND

Churn quantifies the number of customers who have left a brand by cancelling their subscription or stopping paying for the services. This is bad news for any business as it costs five times as much to attract a new customer as it does to keep an existing one. A high customer churn rate will hit a company's finances hard.

<https://www.invespcro.com/blog/customer-acquisition-retention/>

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- Utilize machine learning for predicting customer churn
 - Compare and choose the best machine learning model for this case
 - Model Evaluation using Receiver Operating Characteristic



OBJECTIVE

DATA INTRODUCTION

Source : <https://www.kaggle.com/yeanzc/telco-customer-churn-ibm-dataset>

Data related to a customer who left
churn_label, churn_scores, churn_value, churn reason

Data related to a customer subscribed service
phone_service, multiple_line, internet_service, online_security, online_backup,
device_protection, tech_support, streaming_tv, streaming_movies

Data related to a customer demographic info
country, state, city, zip_code, lat_long, latitude, longitude, gender, senior_citizen,
partner, dependents

Data related to a customer account information
tenure_month, contract, paperless_billing, payment_method, monthly_charges,
total_charges

Data related to IBM company data
customer_id, count, cltv

DATA CLEANSING

ASSIGN DATASET AS DF



CHECK DATA TYPES



ASSIGN NA AS NEW
UNIQUE DATA



DROP UNUSED DATA

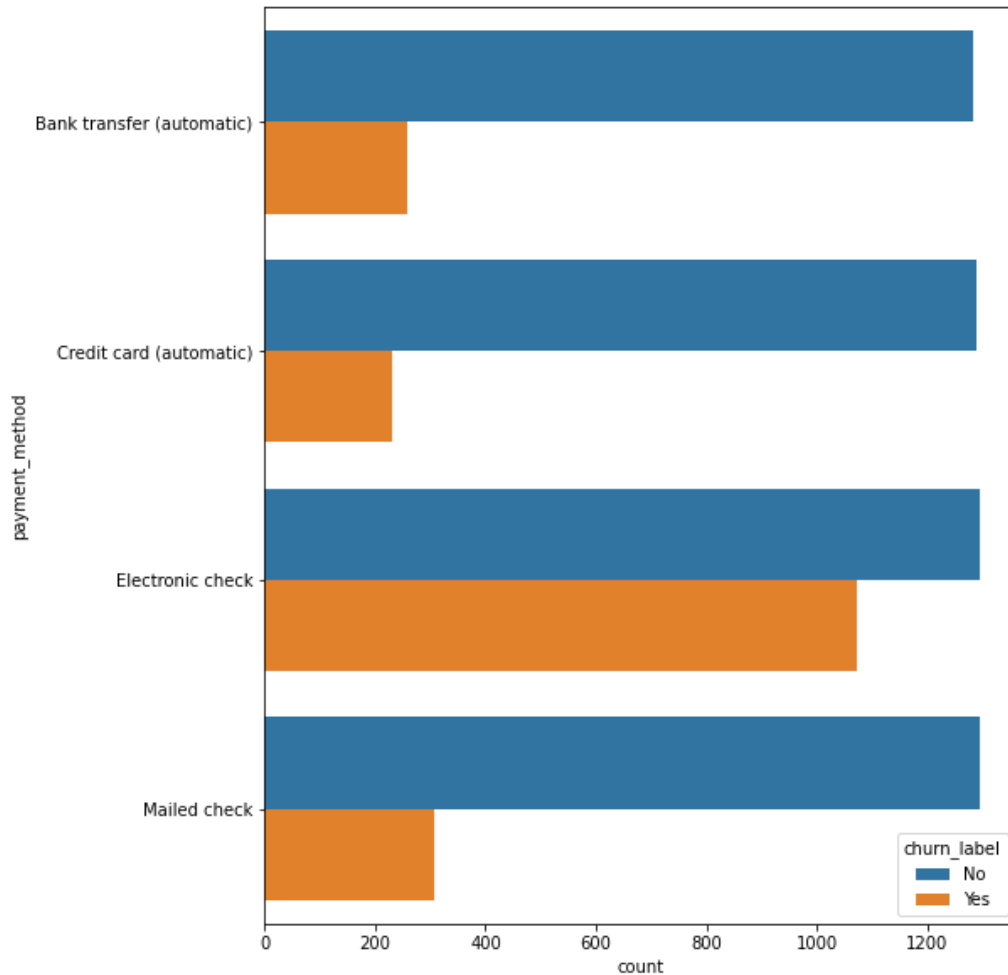


TRANSFORM THE DATA

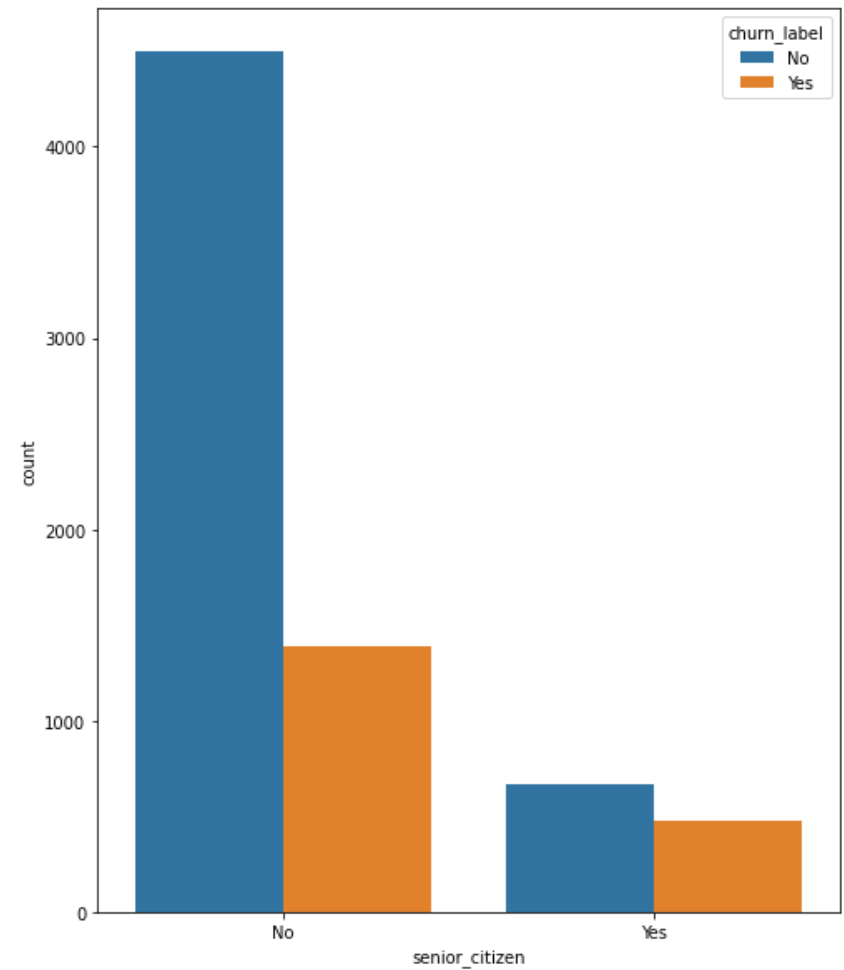
EXPLORATORY DATA ANALYSIS

CATEGORICAL VARIABLE

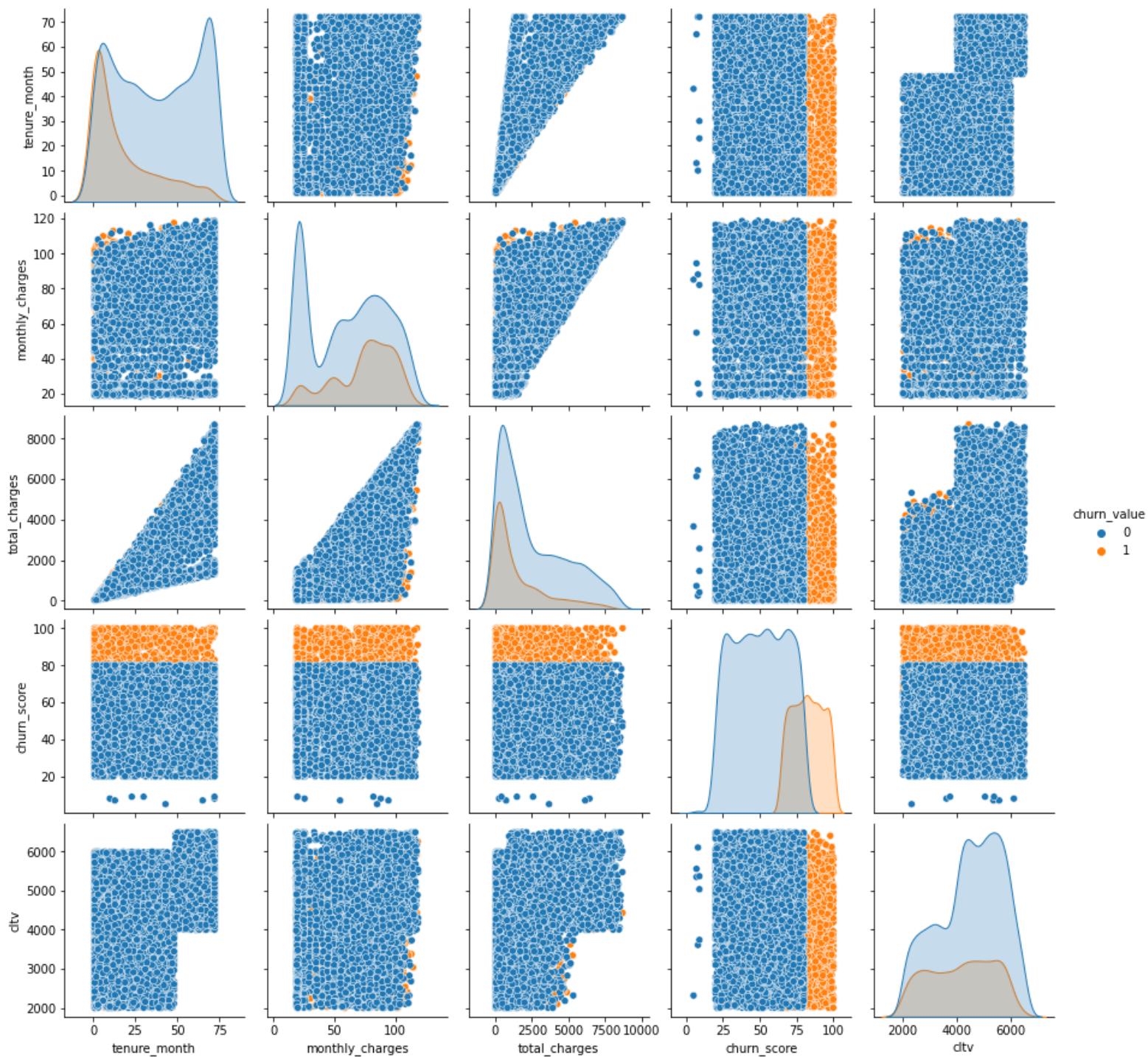
PAYMENT METHOD



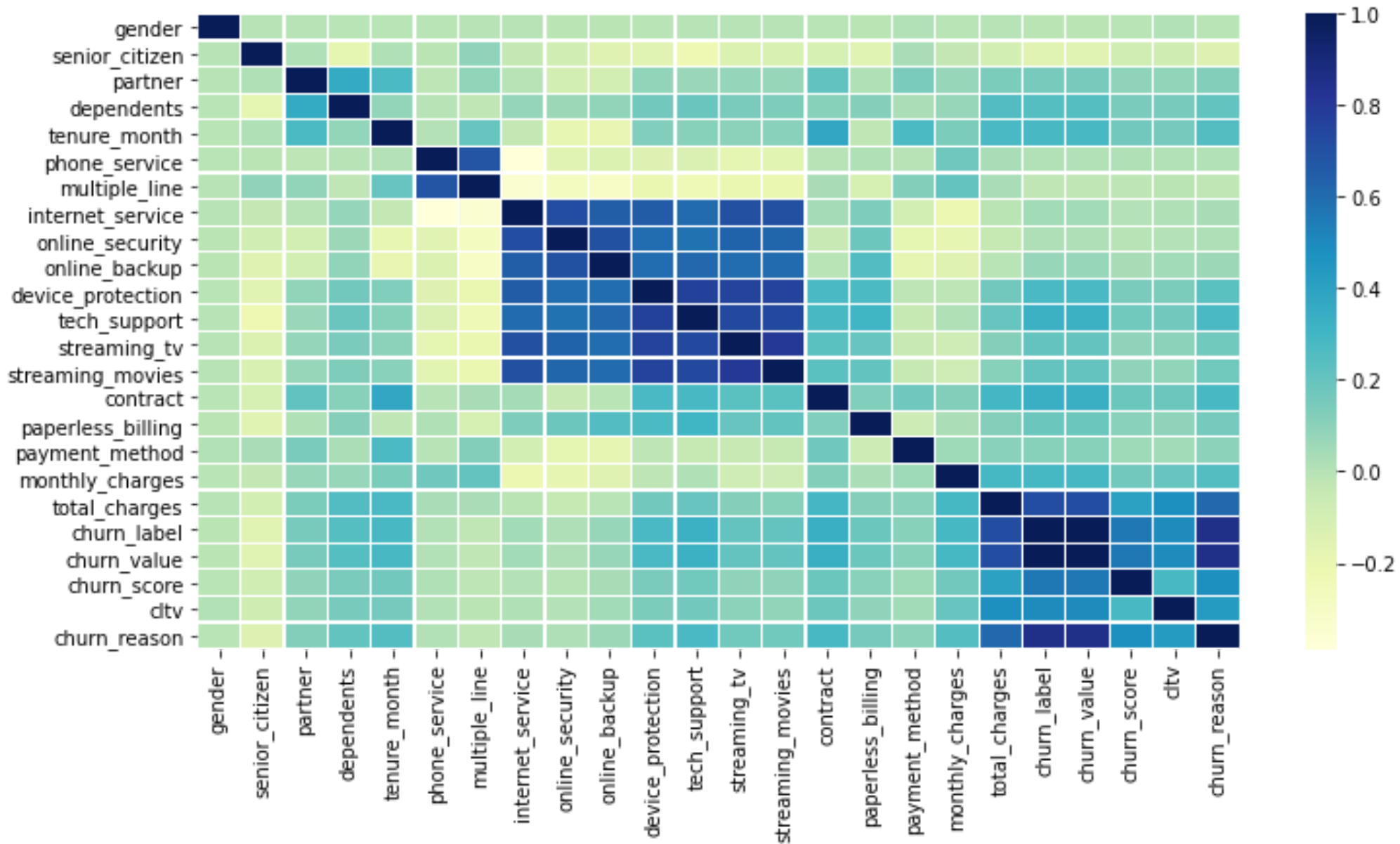
SENIOR CITIZEN



NUMERIC COLUMN



CORRELATION HEATMAP

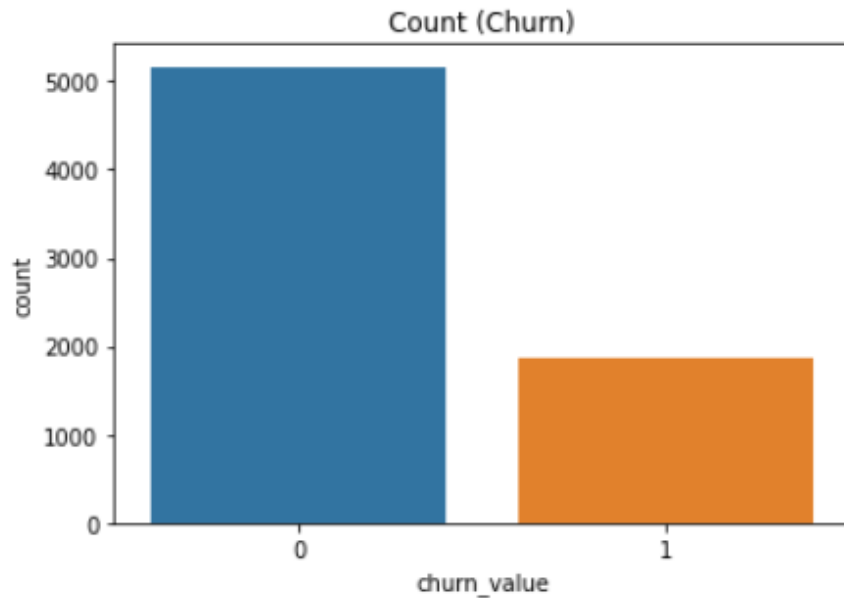


RANDOM UNDERSAMPLING

Random Undersampling involves randomly selecting examples from the majority class to delete from the training dataset.

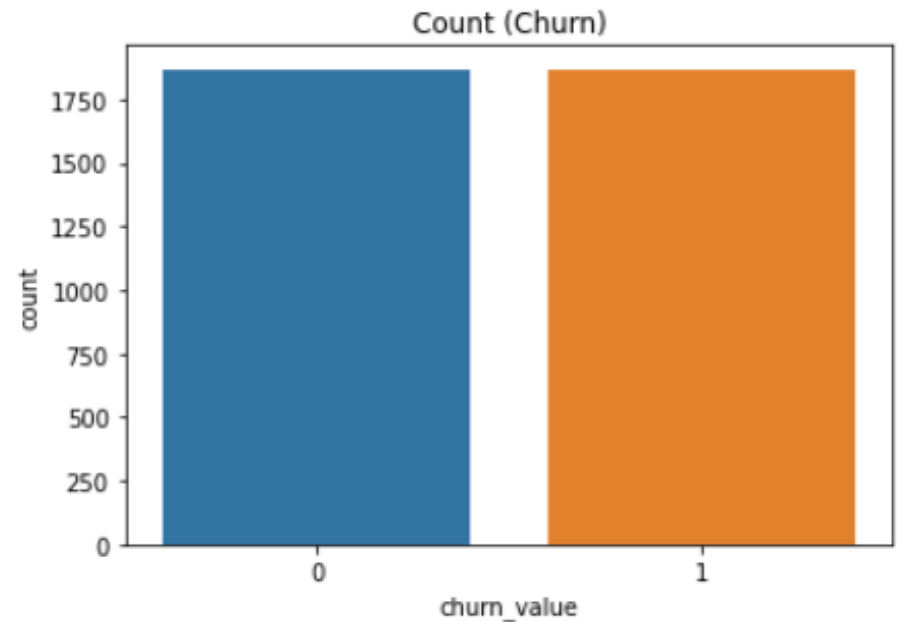
BEFORE

```
Class 0: 5163  
Class 1: 1869  
Proportion: 2.76 : 1  
Text(0.5, 1.0, 'Count (Churn)')
```



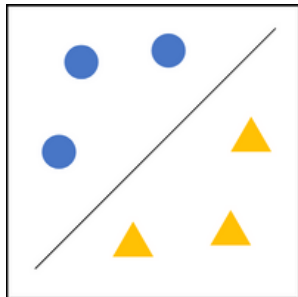
AFTER

```
1 1869  
0 1869  
Name: churn_value, dtype: int64  
Text(0.5, 1.0, 'Count (Churn)')
```

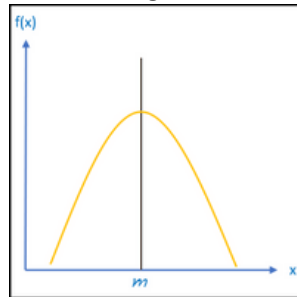


MODELLING

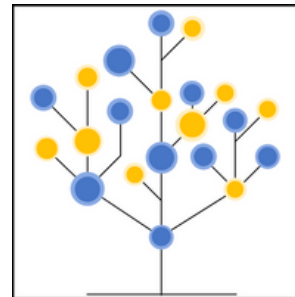
Support Vector
Machine



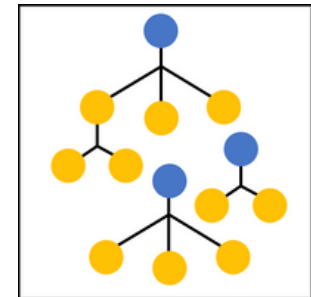
Gaussian Naive
Bayes



Decision Tree
Classifier



Random Forest
Classifier



Train Set

80 % Total Data

$X = 2990, 19$

$y = 2990, 1$

Test Set

20 % Total Data

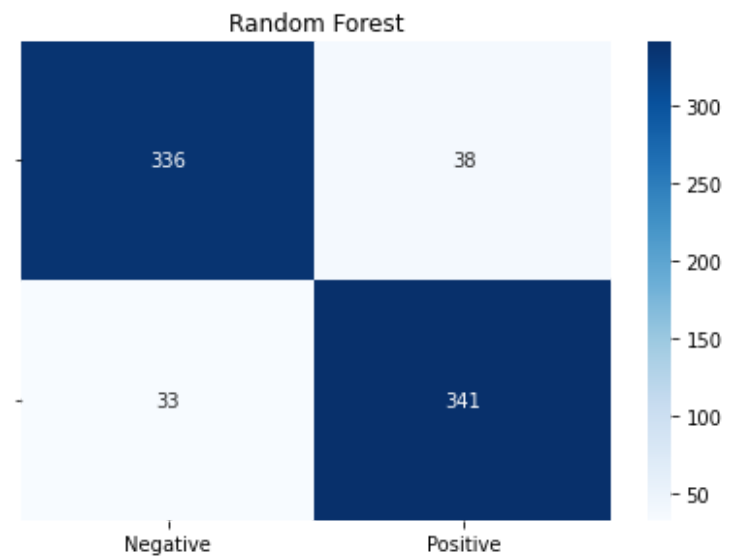
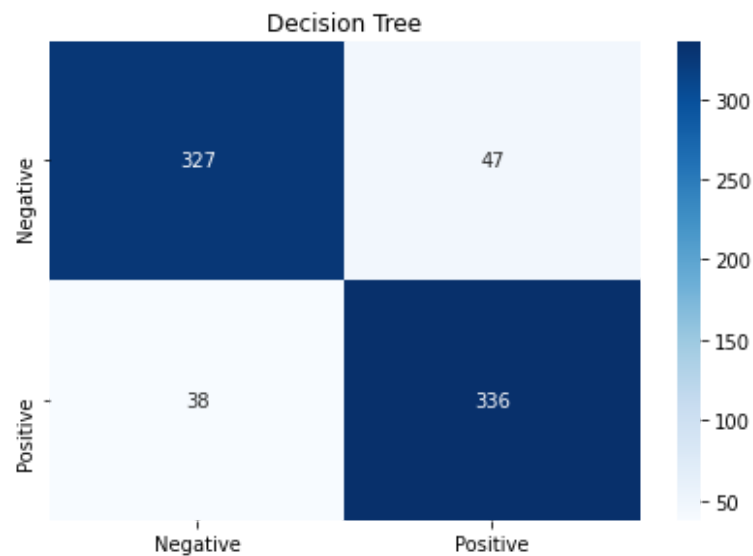
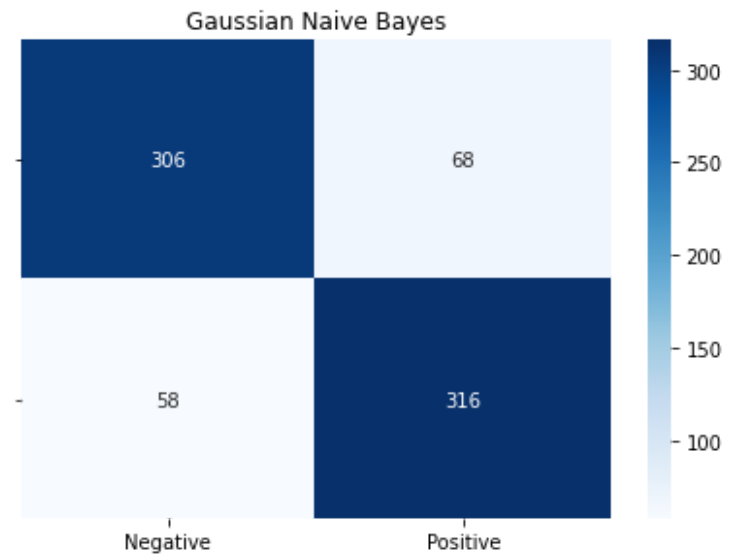
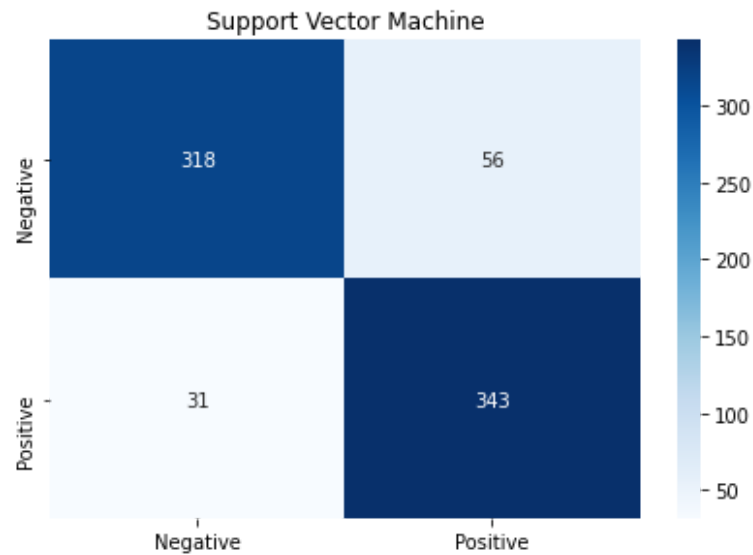
$X = 748, 19$

$y = 748, 1$

MODEL COMPARISON

MODEL	ACCURACY	RECALL	PRECISION	F1 SCORE
Support Vector Machine	0.883	0.917	0.859	0.887
Gaussian Naive Bayes	0.831	0.844	0.822	0.833
Decision Tree Classifier	0.886	0.898	0.877	0.887
Random Forest Classifier	0.905	0.911	0.899	0.905

CONFUSION MATRIX

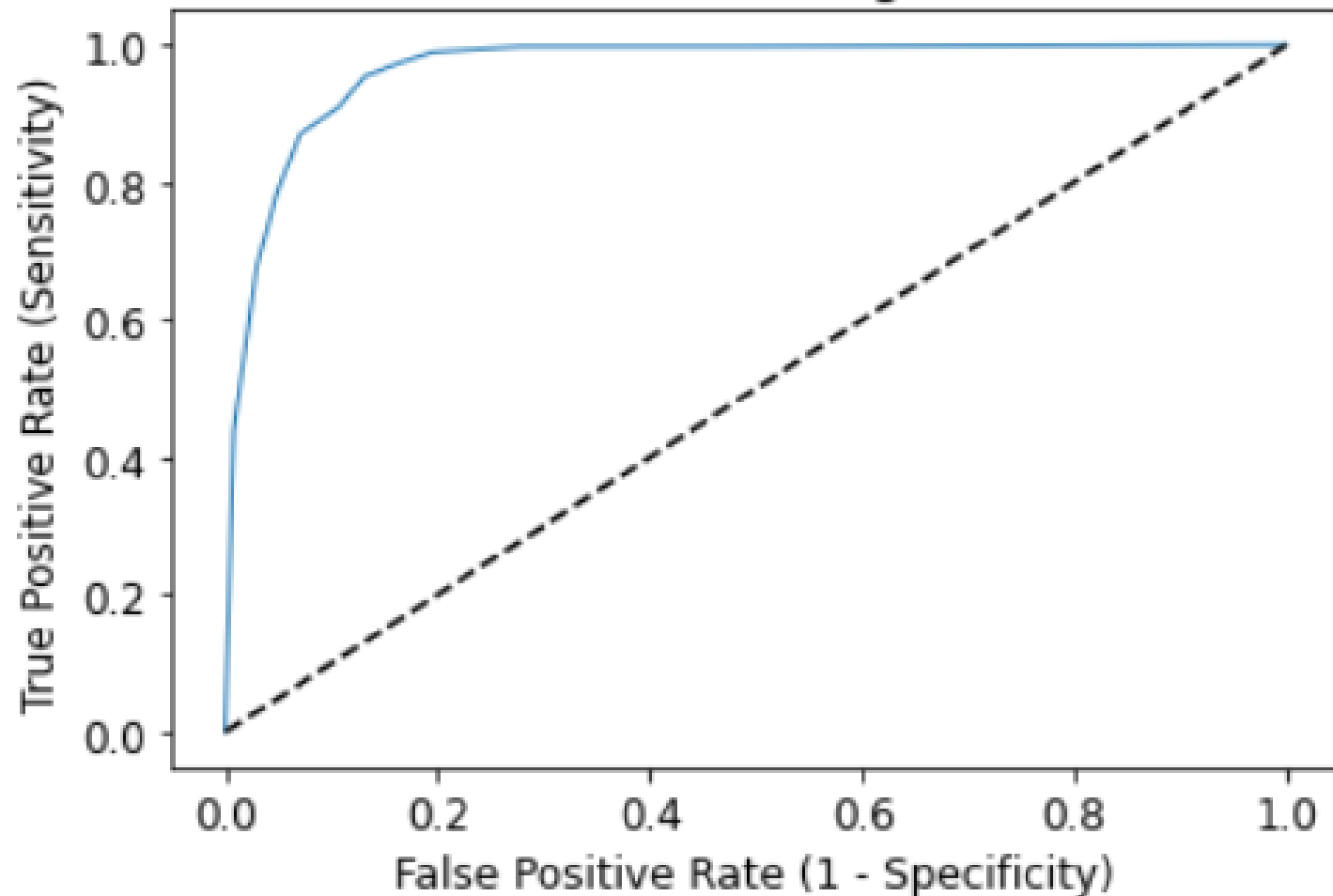


RECEIVER OPERATING CHARACTERISTIC

Cross validated : 0.9672

ROC Score : 0.9051

ROC curve for Churn Prediction using Random Forest Classifier



CONCLUSION

- Some column was dropped because they can cause the model to overfit.
- Random Undersampling was chosen to handle the imbalance data.
- Random Forest Classifier was chosen as it is the best model compared to the other model in this case.
- Model were validated using ROC AUC.

THANK YOU



<https://www.linkedin.com/in/raka-raprast/>



<https://github.com/raka-raprast>