

A background image showing a group of business professionals in an office setting. A woman in a dark blazer is holding a smartphone, while others are looking at documents and tablets. There are coffee cups visible, suggesting a meeting or collaborative work environment.

Subscription Churn Team 5

Javeed Ahmed

Afreen

Sakina

Tariq

Hanna

Uzair

Humza

Sep 1st, 2022

Team - Background

Sakina Igaadi - From Morocco, from Beni Mellal and she work as a teacher, and is interested in data science. She holds a license in human resource management.

Batool Abidi – Grasduated from undergrad and want to expand her skills.

Uzair Khalil – Comes formCivil Engineering background and is planning to enter data science field.

Afreen Nikhat - Computer science Engineer. Experienced in Web Designing and website Mgmt.

Hanna Khan – Worked as BI & Reporting Analyst, Finance & Corp. Mgmt.

Javeed Ahmed – Service Delivery Manager at an IT Software & Service Organization



Project - Background



Analyze the Churn Rate for a Subscription-based personal finance business and make projections about the future.



Customer Churn is important in any business. The ability to predict that a particular customer might churn could impact revenue, market share, and reputation



A minimum churn rate and higher retention rate bring stability to any business; therefore, we aim to help the business minimize the churn rate.



Predictable, recurring revenue



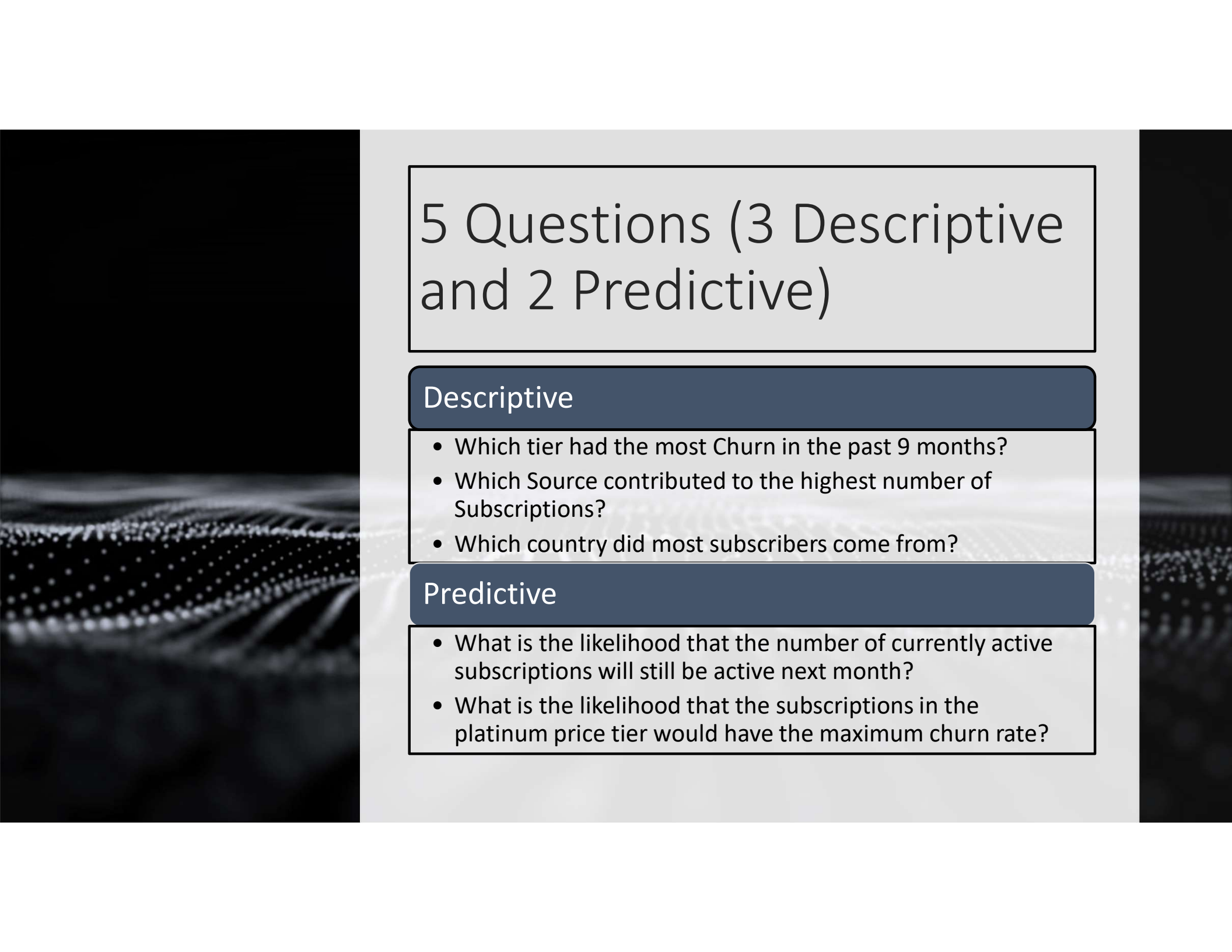
Higher customer lifetime value



Build strong customer relationship

Objectives

Calculate	Predict	Predict	Build
<ul style="list-style-type: none">• Calculate average churn rate over the last 9 months for each price tier.	<ul style="list-style-type: none">• Predict the number of currently active subscriptions that will still be active next month.	<ul style="list-style-type: none">• Predict the number of currently active subscriptions that will still be active in 3 months.	<ul style="list-style-type: none">• Build a separate model that predicts tenure based on price tier, source, and country. Provide actionable insights to the business.



5 Questions (3 Descriptive and 2 Predictive)

Descriptive

- Which tier had the most Churn in the past 9 months?
- Which Source contributed to the highest number of Subscriptions?
- Which country did most subscribers come from?

Predictive

- What is the likelihood that the number of currently active subscriptions will still be active next month?
- What is the likelihood that the subscriptions in the platinum price tier would have the maximum churn rate?

Analytical Base Table – Subscription Churn

Subscription Churn Rate Prediction

Target
Variable
(Will the
subscriber
Churn?)
Y/N

Tenure

1 Month

3 Months

9 Months

Tier

Platinum

Gold

Silver

Location

UK

France

Germany

Source

Organic

Referral

As-is-state & To-be-state



As Is State - Currently, the business doesn't have a system in place to determine the subscription churn rate.

Loosing customer is a concern for businesses as they do not want to lose the revenue



To Be State - Create a model to determine/predict the subscription churn rate for the business and avoid revenue loss

Communication Channel



Whatsapp

Slack

Google Meet

Google Drive

Subscription Churn – Presentation – 2

Team Members

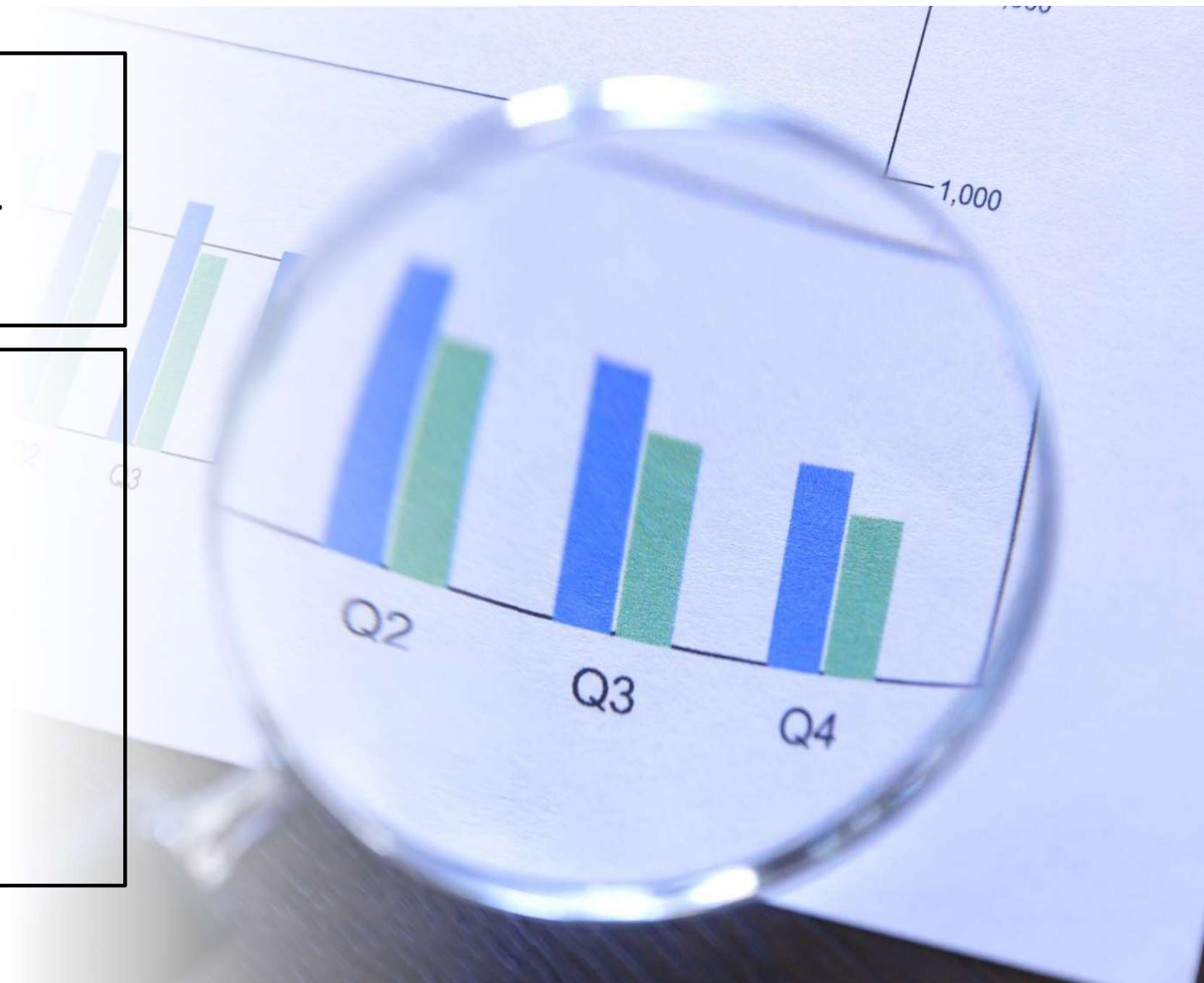
- Javeed Ahmed
- Afreen
- Sakina
- Hanna

Oct 6th, 2022

Presentation 2

Agenda

- EDA
- Data Viz
- Data Quality Report
- Data Understanding
- Appendix



Exploratory Data Analysis

Dataset Statistics







Number of Variables	6
Number of Observations	99,944
Missing Cells	None
Duplicate Rows	None
Variables Types	
Categorical	3
Numeric	3

Data Dictionary

- member_id - Unique ID of the user.
- tier - Price tier (Silver, Gold, or Platinum).
- country - Member country.
- source - Original acquisition channel.
- tenure - Number of cycles billed. Min is 1. Max is 9.
- active - Is the subscription still active?

DATA QUALITY REPORT

Salient Features

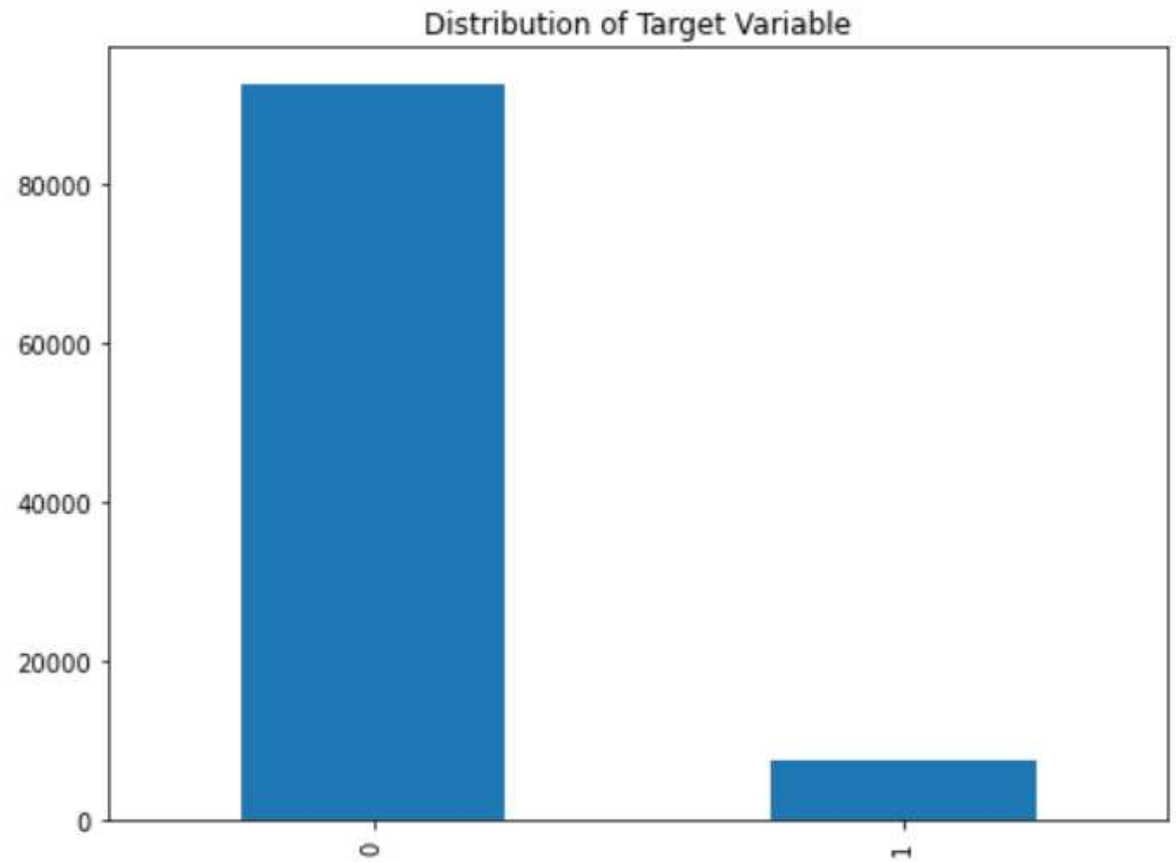
	Data Type	Categorical , Numeric,
	3	Categorical Source (3) – Organic, Referral, Partnership Country (8) – US,UK, Germany, Netherlands, France, Spain, Italy, Canada Tier (3) – Gold, Silver, Platinum
	3	Numeric Member ID, Tenure (9 Months) Active – Active or Not active
	6	Total Features
	TV	Active (Churn) – Binary 0/1
	TV(D)	Churn Rate

Data Observations/Description

Record Count	99,944
Skewness	Active, Tenure- Positively skewed to the right.
Outliers	Tenure, Active – Outliers to the right of the distribution
Duplicates	None
Missing Values	None
Special Characters	None

Data Understanding

- Target Variable (TV) is Imbalanced
- Churn subscribers(0) = 92.5%
- Active (1) = 7.4%

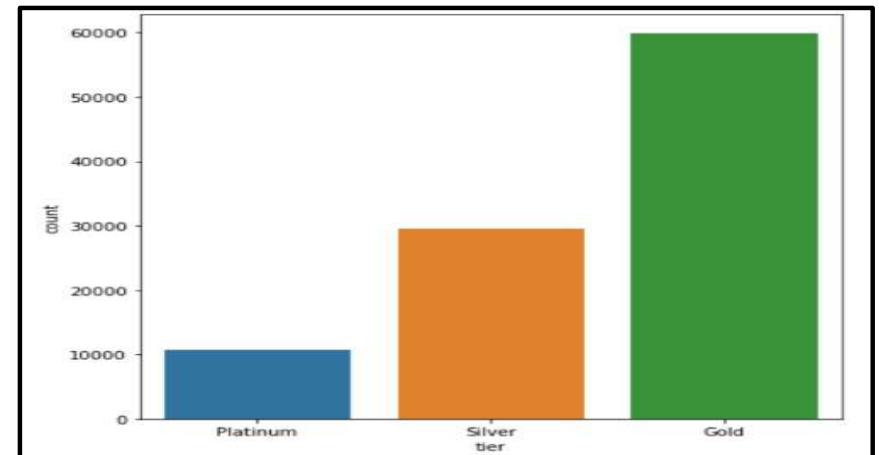
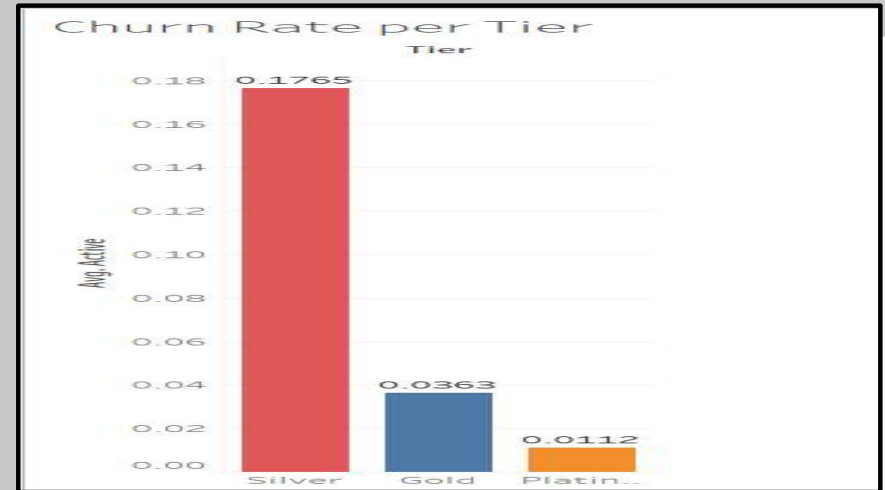


Data Understanding – Churn Rate per Tier

Which tier had the most Churn in the past 9 months?

- Silver tier had the highest churn rate of 17.65%
- Gold had the highest subscriber count.

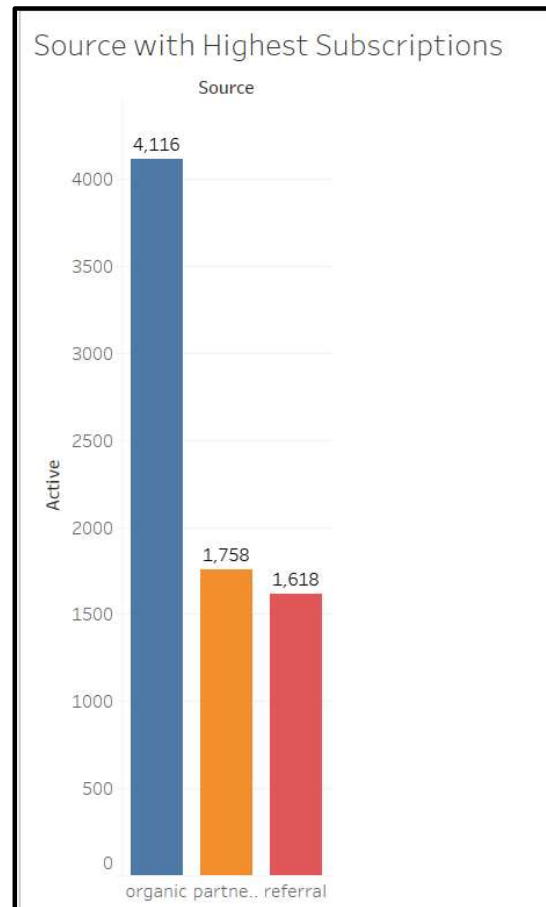
Details	Gold	Platinum	Silver	Grand Total
Churn Count	57669	10516	24270	92455
Active Count	2171	120	5204	7495
Active Rate (%)	3.62	1.12	17.65	22.39
Grand Total	59840	10636	29474	99950



Questions from Presentation 1

Which Source contributed to the highest number of Subscriptions?

- Organic source had 82,873 Subscribers and 4116 Subscribers are active within the 9-month period.



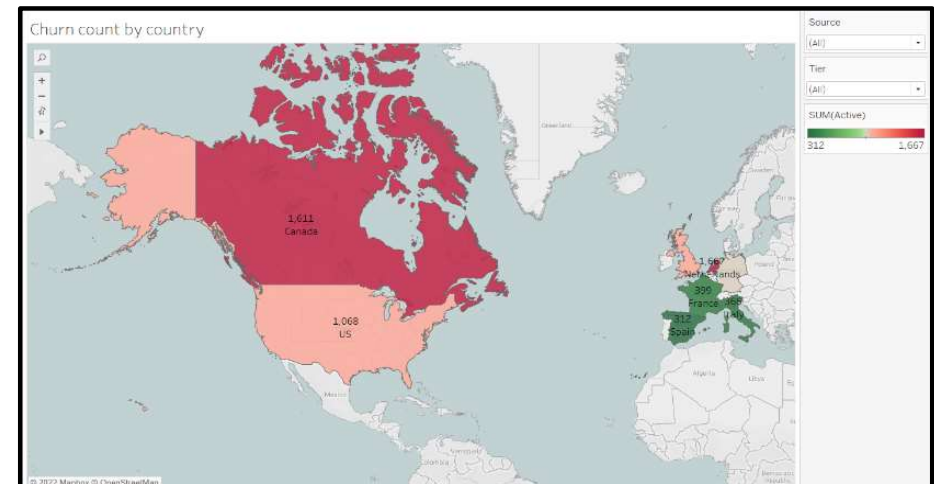
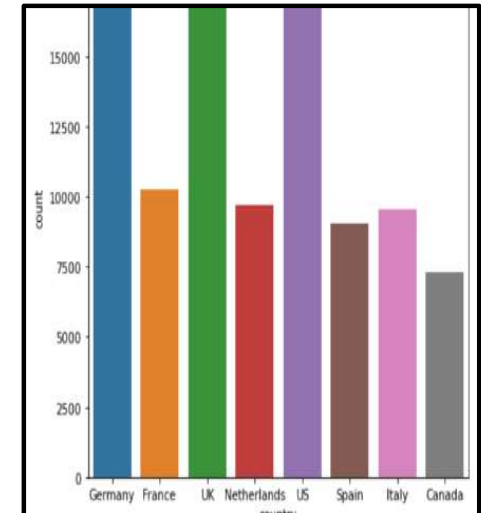
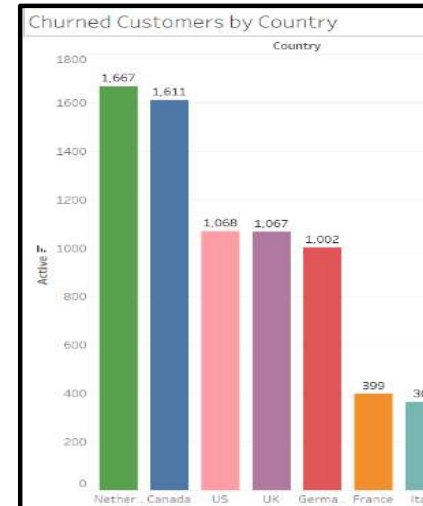
Questions Answered - Data Visualization

Which country did most subscribers come from?

- Most Subscribers came from the US
- Netherlands had the highest churn 9694 subscribed and 1667 are active

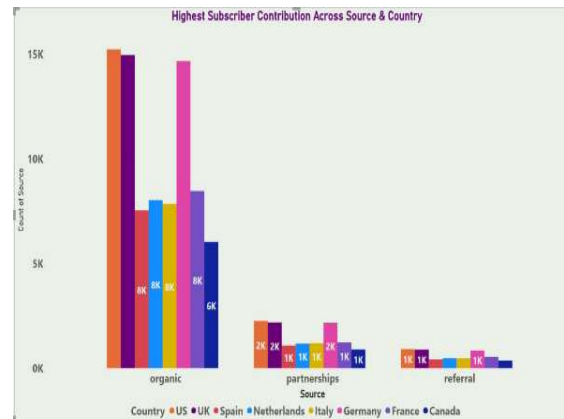
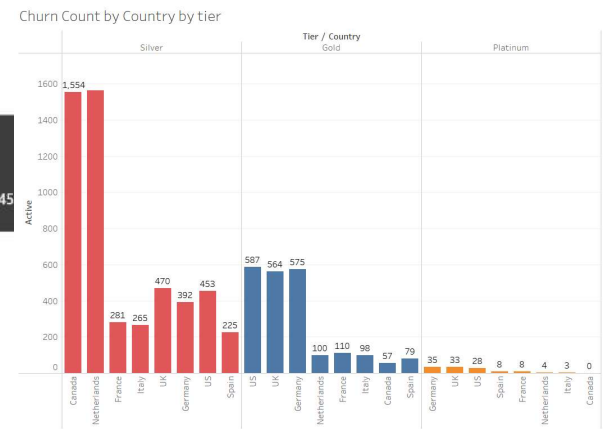
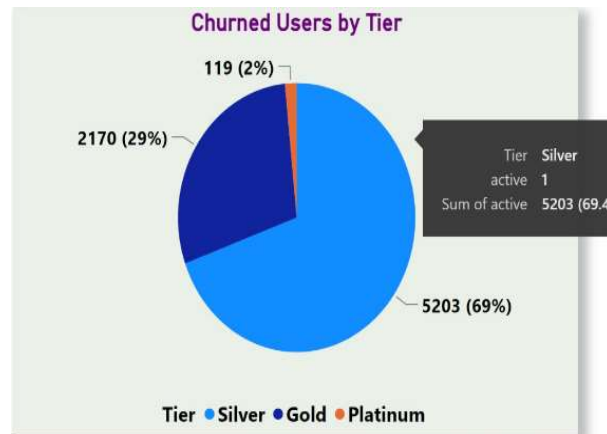
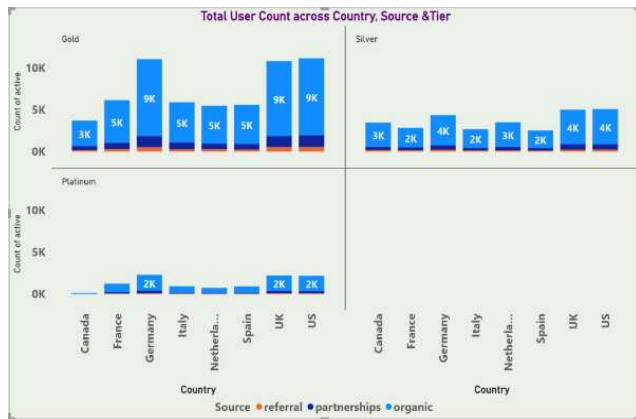
Which country had the most subscribers (count)?

Countries	Total Count	Active Count
Germany	17689	1002
Netherlands	9693	1667
UK	18033	1067
US	18425	1068
Grand Total	63840	4804



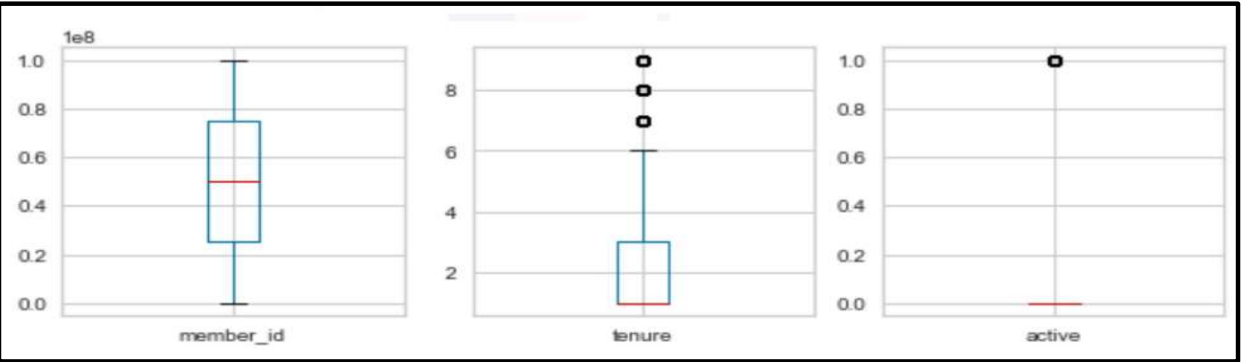


Appendix

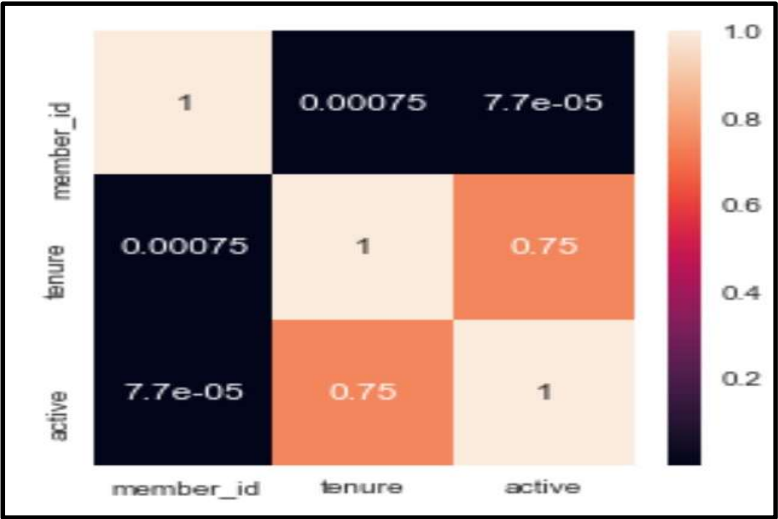
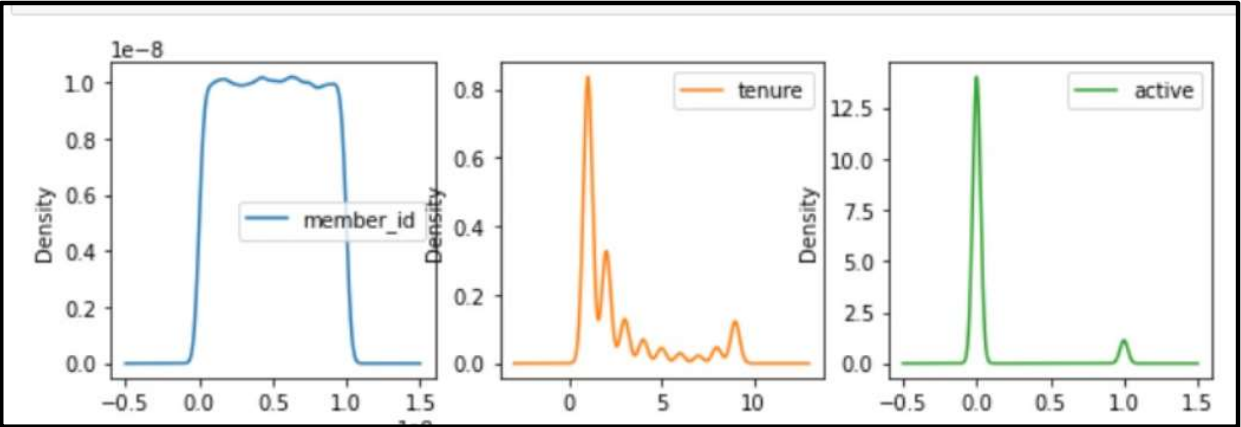


Appendix – Data Visualization

Outliers in Tenure depicts that most customers started churning in later months



Positively skewed to the right depicting outliers



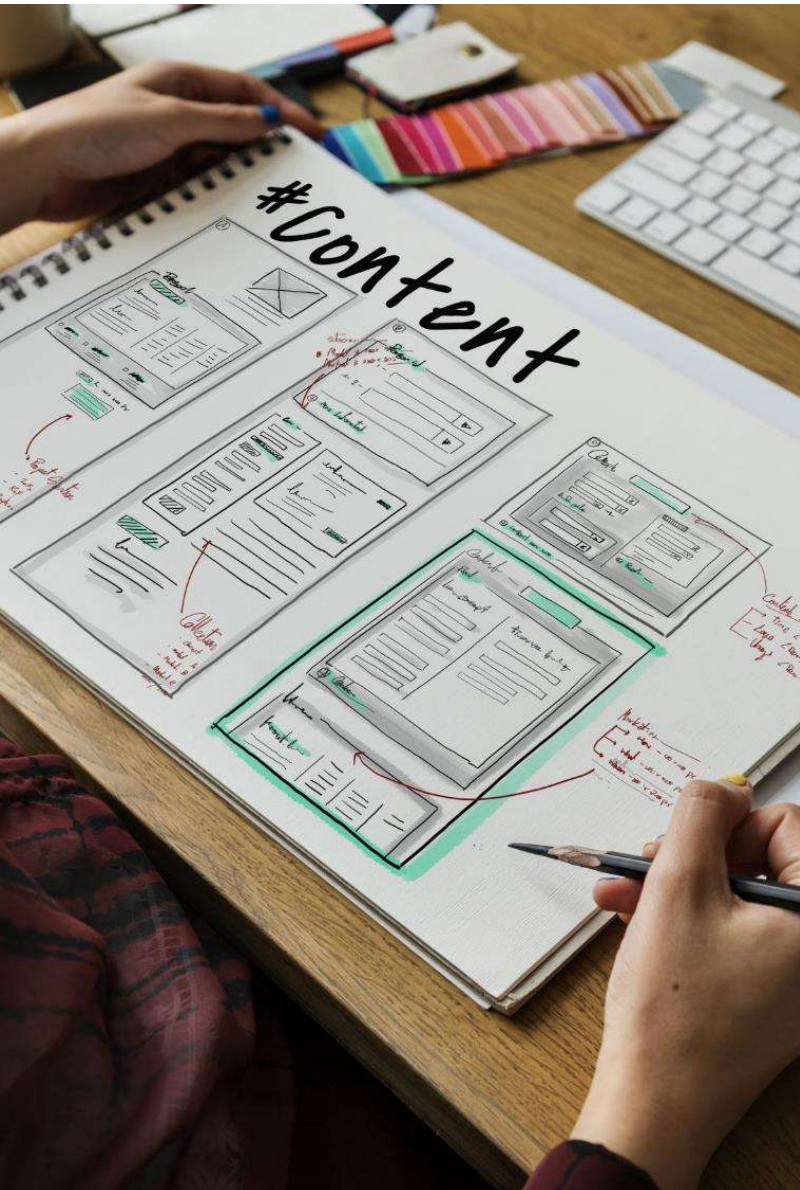
Tenure and Active features are highly correlated which depicts that most customers churned during the later months.

Subscription Churn – Presentation – 3

Team Members

- Javeed Ahmed
- Afreen
- Hanna

Nov 7th, 2022



Agenda

- Data Preparation
- Modelling Technique
- Data Ready for Modelling
- Appendix



Data Preparation

- Total Number of Features = 6
- Numeric Features = 3
- Categorical Features = 3
- Replace method (mapping) to transform the ordinal column (tier)
- Use one hot encoding (Dummies) to transform the nominal features.
- Dropped member ID

Data Preparation

Scale our Data except not required for Decision Tree

Use Precision and Recall as evaluation metric before balancing the data

Use SMOTE (Synthetic Minority Over sampling technique) to Balance out TV and check accuracy.

Use Binary Classification Algorithms like DT, NB, SVM, LR to predict our TV.

Subscription Churn – Presentation – 4

Team Members

- Javeed Ahmed
- Afreen
- Hanna

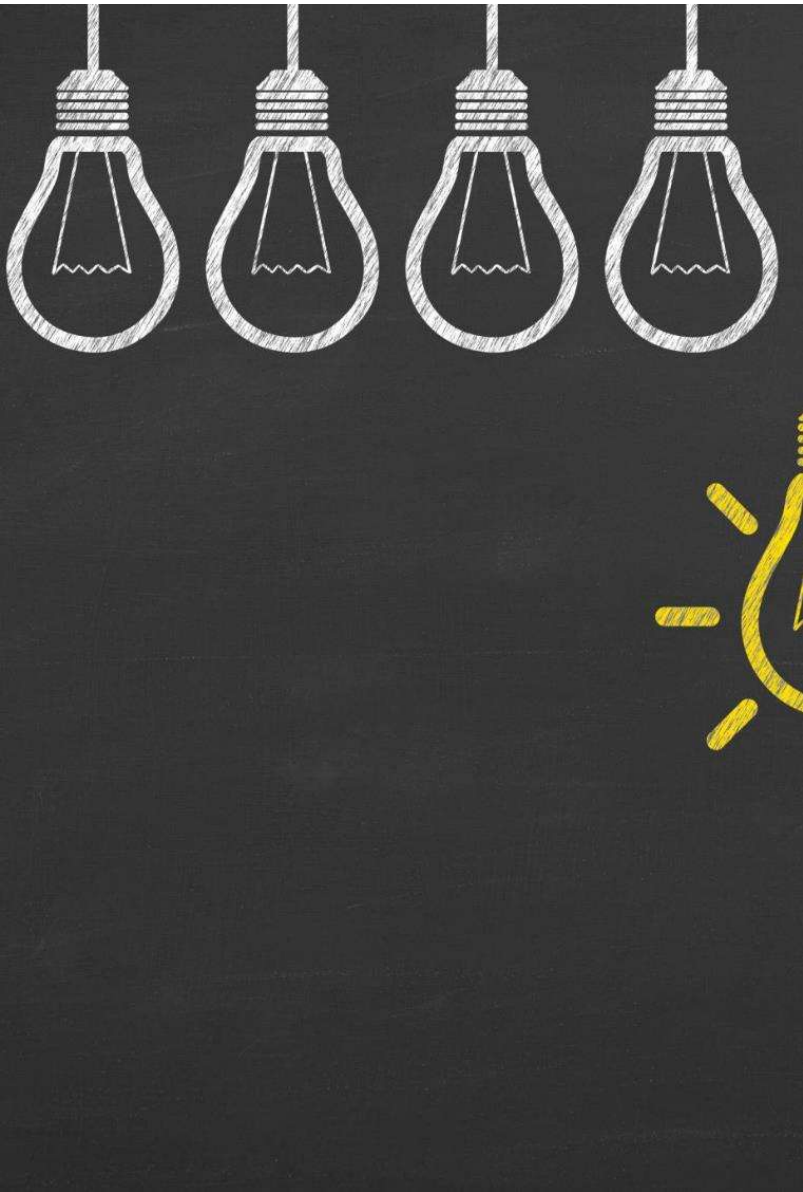
Dec 9th, 2022

A background image showing a group of business professionals in an office setting. A man in a suit and tie is on the left, gesturing with his hand. A woman in a dark blazer is in the center, holding a smartphone. Another person is on the right, partially visible. There are coffee cups and a tablet with a chart on the table.

Subscription Churn Team 5 – Presentation 4

Javeed Ahmed
Afreen
Hanna

Dec 9th, 2022



Agenda

- Data Modelling
- Model Comparison
- Model Implementation
- Appendix

Decision Tree



Used DT Classifier to train our Data without scaling our Data
Balancing our Data



Achieved an accuracy of 100% as expected because data is imbalanced as it might have captured the same class only.



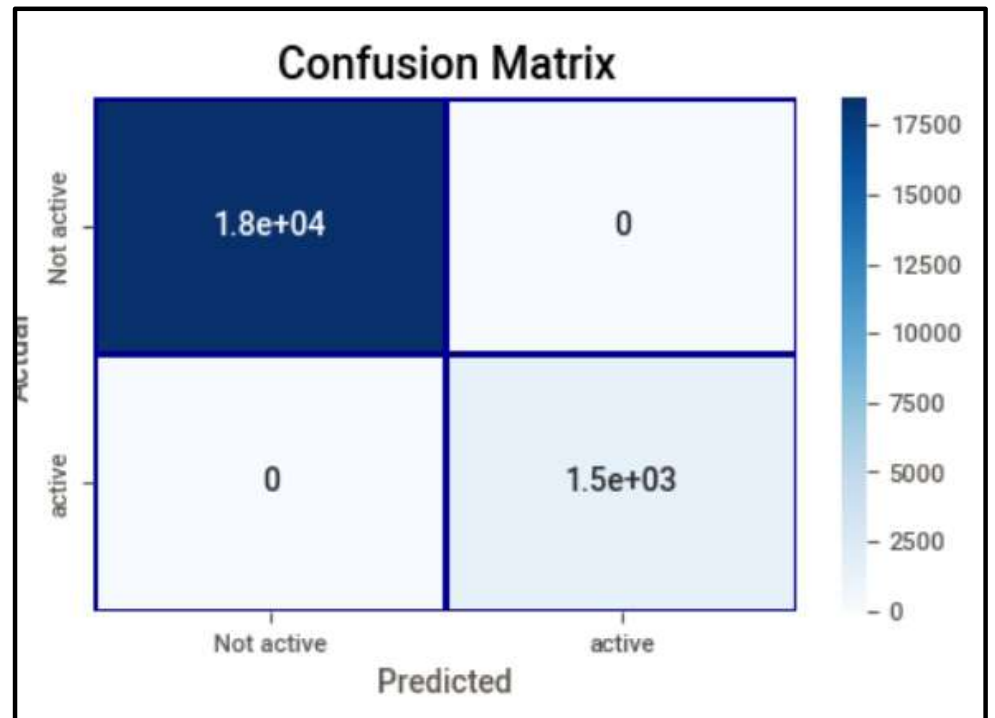
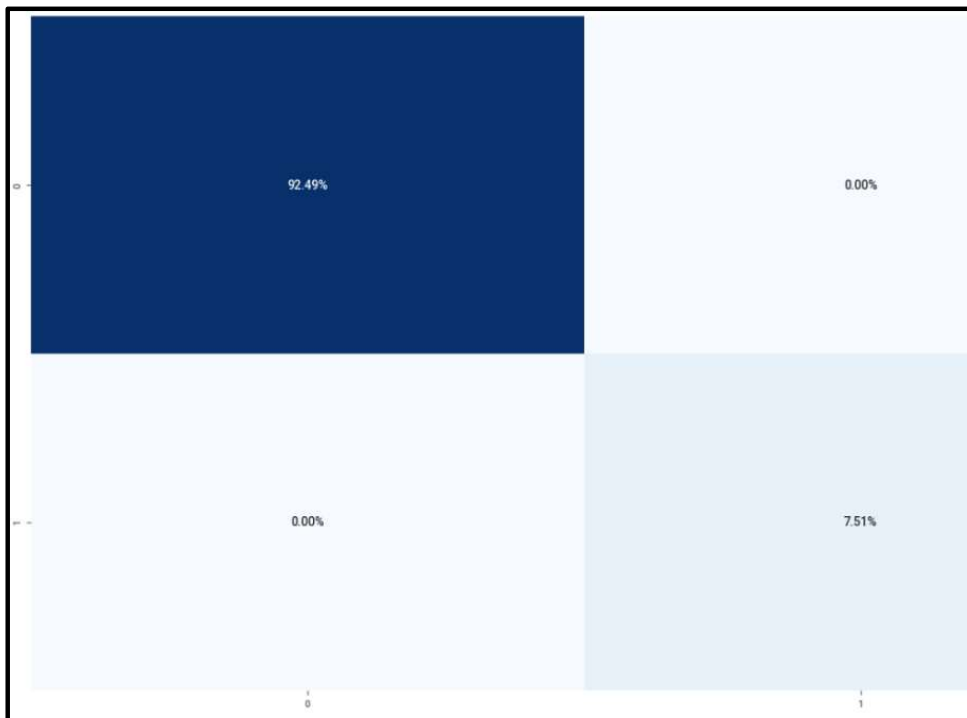
Precision and Recall to be determined, also 100%



Will check for accuracy once again after balancing and scaling the data

Naïve Bayes

Accuracy = 100%





Model Comparison

- Run another Classification Models and pick the best model
- Optimize the best model by using Hyper parameter tuning

Subscription Churn – Presentation – 5

Team Members

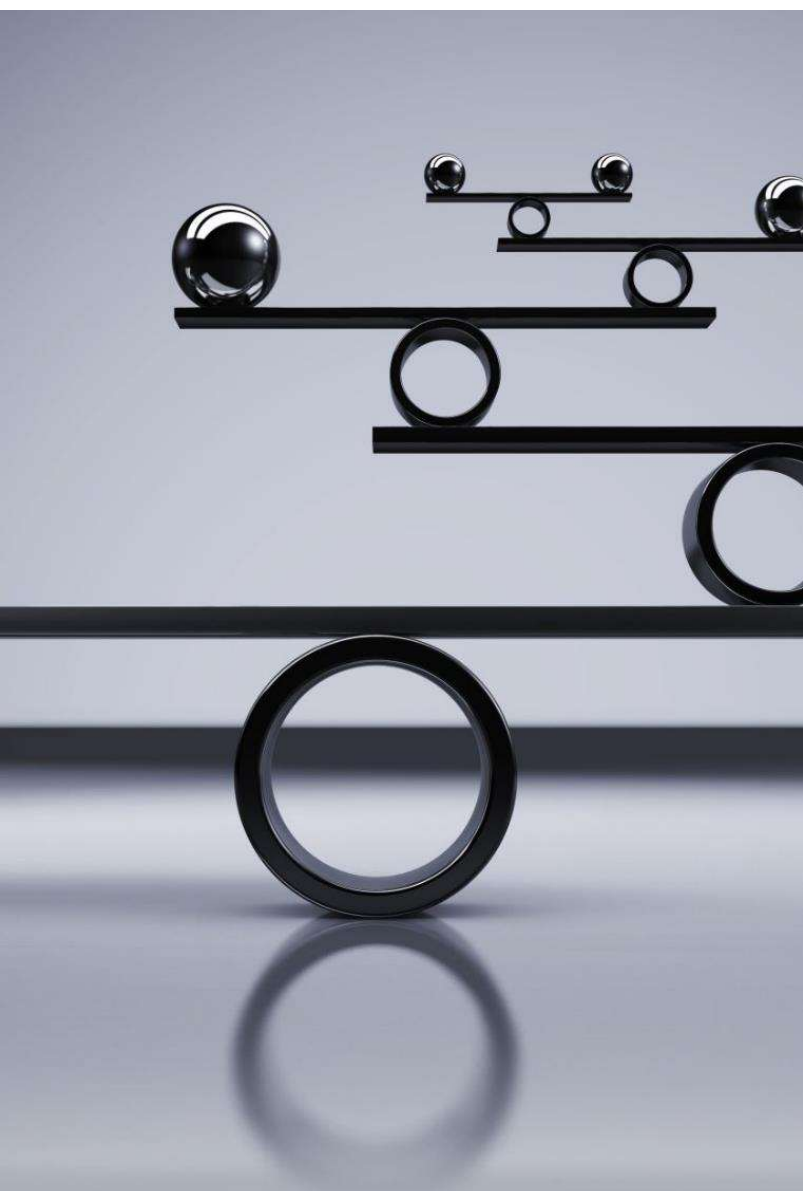
- Javeed Ahmed
- Afreen
- Hanna

Jan 6th, 2023



Agenda

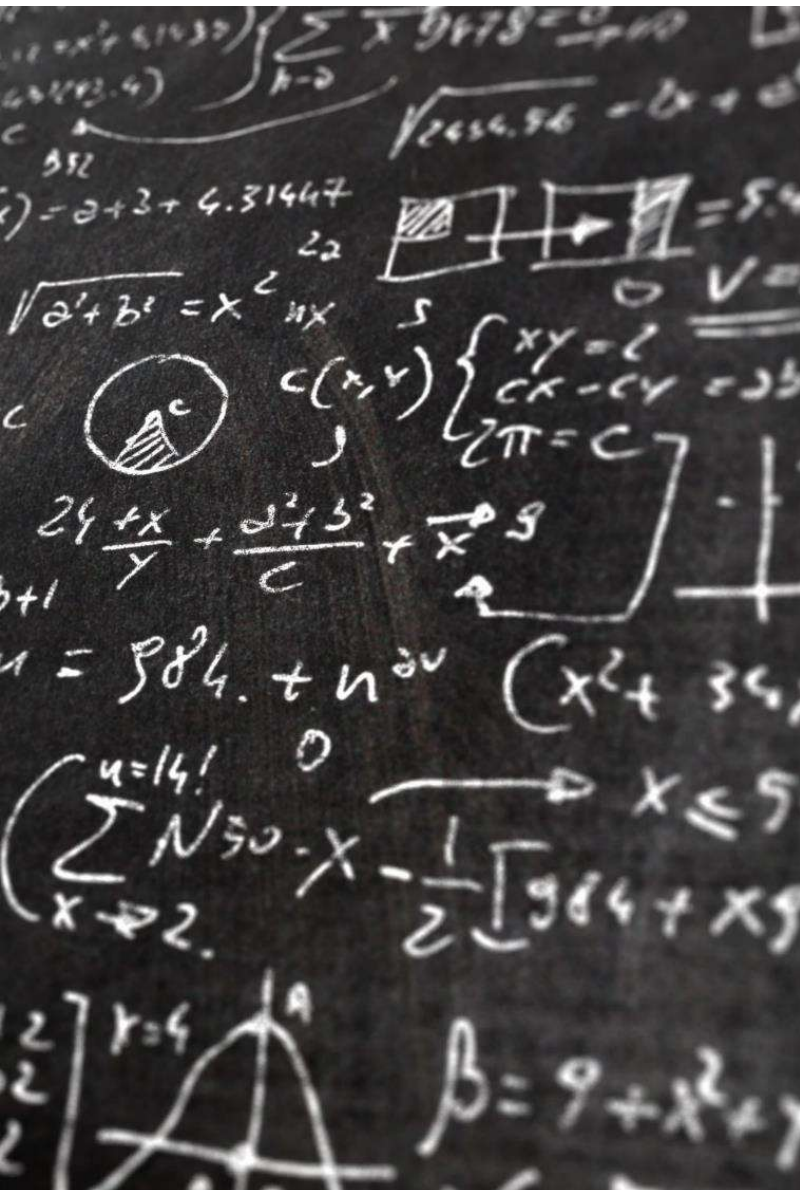
- End to End Model Delivery
- Action Plan
- Client Recommendation
- Appendix



Model Delivery

- End to End Binary Classification problem where we predicted if a customer would churn in 3 months or 9 months. The model prediction was 100% because we had class imbalance, so we looked at Precision which is True Positive / True Positive + False positive got a value we got a value of 1.
- Only 1 Feature Tenure has the highest impact on our TV that is Churn.

```
***** Accuracy Score Prediction *****
*
*      Dummy Classifier - Baseline           : 92.51%      *
*      Decision Tree Accuracy               : 100.00%      *
*      Naive Bayes Accuracy                 : 100.00%      *
*      Logistic Regression Accuracy          : 92.62%      *
*      Random Forest Accuracy               : 100.00%      *
*      SVM Prediction Accuracy              : 100.00%      *
*      Kernalized SVM Accuracy              : 100.00%      *
*      KNeighbors Accuracy                  : 99.80%      *
*
*****
```



Action Plan

- In our use case in which we observe more data points of one class than of another, the accuracy is no longer a useful metric. Using a different sample size and balancing the data using SMOTE achieved 100% accuracy, meaning that 92% of customers will churn.
- Extract More Features or look for revised data
- Look into different features that is impacting churn rate.
- Churners are generally more recent subscribers, with a median subscription of approximately two months. This information can be used to pinpoint milestones during the customer subscription journey. For example, once a user reaches two months into their subscription, businesses should consider providing promotions, incentives, and features that will encourage them to stay.

Appendices

	Tier	Count	Churn Rate against Tier	Overall Churn Rate against Tier
Active	Gold	57668		
	Platinum	10515		
	Silver	24269		
	Total	92452		
Churned	Gold	2170	3.76%	2.2%
	Platinum	119	1.13%	0.1%
	Silver	5203	21.44%	5.2%
	Total	7492		
	Grand Total	99944		7.5%

Churn Count by Country by tier

Country	Tier		
	Silver	Gold	Platin..
Canada	1,554	57	0
France	281	110	8
Germany	392	575	35
Italy	265	98	3
Netherlands	1,563	100	4
Spain	225	79	8
UK	470	564	33
US	453	587	28

