CUSTOMER CHURN ANALYSIS

CONTENT SLIDE

- Introduction to Customer Churn.
- Data Collection and Preparation
- SQL for Data Extraction
- Python for Data Analysis
- Tableau for Data Visualization



UNDERSTANDING CUSTOMER CHURN



Definition of Churn

Churn refers to the loss of customers or clients within a specific timeframe, often measured as a percentage of total customers at the start of that period.



Importance of Churn Analysis

Analyzing churn is crucial for businesses as it helps identify factors leading to customer loss and informs strategies to improve customer retention and engagement.



Industry Impact

Different industries experience varying rates of churn, which influences strategic decisions regarding customer acquisition, service improvements, and overall market competitiveness.

KEY METRICS IN CHURN ANALYSIS



Churn Rate Calculation

Churn rate is calculated by dividing the number of customers lost during a certain period by the number of customers at the beginning of that period, multiplied by 100.



Customer Lifetime Value (CLV)

CLV is a predictive metric that estimates the total revenue a business can expect from a single customer account over the entire duration of their relationship.



Retention Rate

Retention rate measures the percentage of customers who remain with a company over a specified timeframe, highlighting the effectiveness of customer engagement initiatives.



DATA SOURCES



Internal Databases

Internal databases are repositories within an organization that store data gathered from its operations, providing valuable insights for analysis and decision- making.



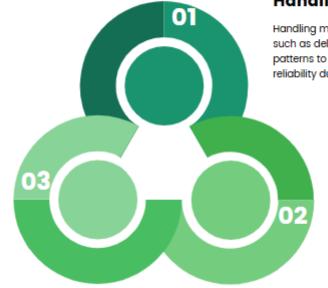
External Data Providers

External data providers offer datasets from outside the organization, including market research firms or public data sets, enhancing the internal data with broader context and trends.

DATA CLEANING

Data Transformation

Data transformation involves modifying data formats or structures to enhance usability, including scaling, aggregating, or encoding, thus facilitating more effective data analysis and interpretation.



Handling Missing Values

Handling missing values involves techniques such as deletion, imputation, or analysis of patterns to ensure dataset integrity and reliability during the analysis phase.

Data Normalization

Data normalization is the process of adjusting values in a dataset to a common scale without distorting differences, ensuring more accurate comparisons and analysis results.

SQL FOR DATA EXTRACTION

SQL BASICS



■ Select Statements

Select statements are fundamental SQL commands used to query data from a database, allowing users to specify columns, filter records, and order results effectively.

Aggregations and Joins

Aggregations in SQL enable the summarization of data using functions like COUNT, SUM, AVG, while joins combine records from two or more tables based on related columns.

SELECT, FROM, WHERE, JOIN, GROUP BY

```
-- Reterive all customers from north

america region

select FirstName, LastName, customerid, region

from customers

where region='North America';
```

- Retrieve the major reason fot the
customer churn
select Reason, count(customerID) from churn
group by reason;

```
-- Number of customers who have taken
annual subscrpitions
select * from subscriptions;
select * from customers;
select count(*)as Total_subscriptions
from subscriptions
where PlanType='Annual';
```

- Retrieve the customer details along with
subscription details
select
c.CustomerID,c.FirstName,c.LastName,s.Subscript
ionID,s.PlanType
from customers c
Join subscriptions s on c.CustomerID =
s.CustomerID;

PYTHONFOR DATA ANALYSIS

PYTHON LIBRARIES



Pandas

Pandas is an essential library for data manipulation and analysis, providing data structures like DataFrames for handling structured data efficiently.



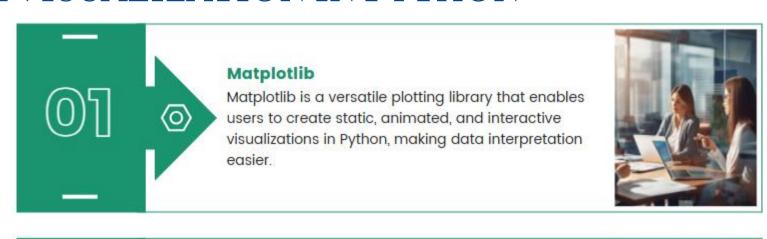
NumPy

NumPy is the foundational package for numerical computing in Python, offering support for multi- dimensional arrays and a variety of mathematical functions.



SciPy

sciPy builds on NumPy and provides additional functionality for scientific computing, including modules for optimization, integration, and statistical analysis.



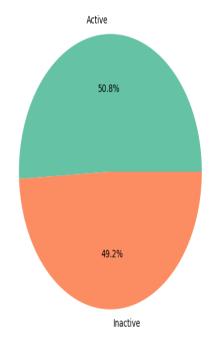


Seaborn

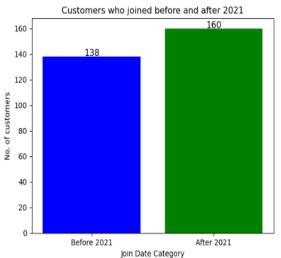
Seaborn is a statistical data visualization library based on Matplotlib, offering a higher-level interface for drawing attractive and informative graphics.



Distribution of Customer Status



```
[39]: #7. Create a visualisation(bar chart) who joined after and befoe 2021
                                                                                                                            ★ □ □ ↑ ↓ 占 〒 ii
       #converting JoinDate into DATE TIME FORMAT
     customers_df['JoinDate']=pd.to_datetime(customers_df['JoinDate'],format='%d-%m-%Y')
     # Filtering the customers who joined before and after 2021
      before_2021=customers_df[customers_df['JoinDate']<'2021-01-01']
     after_2021=customers_df[customers_df['JoinDate']>'2021-01-01']
     # Count the number of customers in each of the category
     counts={'Before 2021': len(before_2021),'After 2021': len(after_2021)}
     # create a bar chart
     plt.bar(counts.keys(),counts.values(), color=['blue','green'])
     plt.xlabel('Join Date Category')
     plt.ylabel('No. of customers')
      plt.title('Customers who joined before and after 2021')
     #Add the exact numbers on the top of the bars
      for i, (key,value) in enumerate(counts.items()):
         plt.text(i,value+0.5, str(value), ha='center', fontsize=12)
     plt.show()
     # Using for Loop, the Loop goes through each bar (Before 2021 and After 2021)
      #For each bar, it places the customer count(value) as a label slightly above the bar, position it at centre
      #counts.items(): This gives you both the key and the values from dictionary
```



```
[44]: #8. List down the active customers from Europe
Active_cx=customers_df[(customers_df['Region']=='Europe') & (customers_df['Status']=='Active')]
result=len(Active_cx)
print(result)

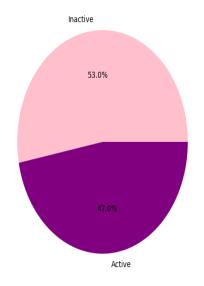
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[49]: #Visualisation of active and Inactive Customers in Europe.
# Filtering the european customers
customers_europe=customers_df[customers_df['Region']=='Europe']

#Count the number of active and Inactive
status_counts= customers_europe['Status'].value_counts()

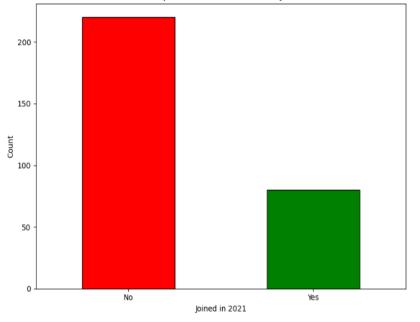
plt.figure(figsize=(8,6))
plt.pie(status_counts, labels=status_counts.index, autopct='%1.1f%%',colors=['pink','purple'])
plt.title('Distribution of Customers')
plt.show()
```

Distribution of Customers





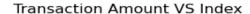


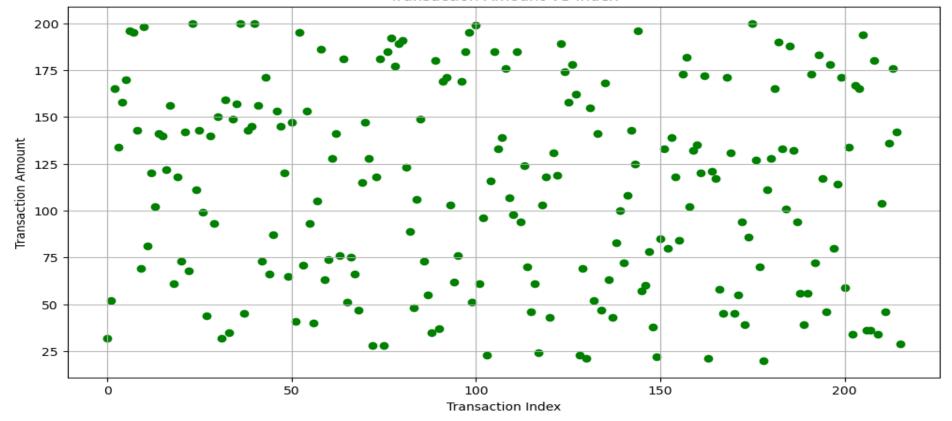


```
#calculate the average transaction amount
Average_amount= Transactions_df['Amount'].mean()
print(f"Average Transaction Amount: {Average_amount}")
Average Transaction Amount: 112.19907407407408
# Visualize / Analysis for Transaction Amount w.r.t to Index - Line graph
plt.figure(figsize=(12, 6))
plt.plot(Transactions_df.index, Transactions_df['Amount'],linestyle='-', linewidth=2, color='blue')
plt.title('Transaction Amount VS Index')
plt.xlabel('Transaction Index')
plt.ylabel('Transaction Amount')
plt.grid(True)
plt.show()
                                                         Transaction Amount VS Index
   200
   175
   150
Transaction Amount
   125
   100
    75
    50
    25
                                         50
                                                                    100
                                                                                                150
                                                                                                                            200
                                                                  Transaction Index
```

```
# Visualize / Analysis for Transaction Amount w.r.t to Index - scatter plot
plt.figure(figsize=(12, 6))
plt.scatter(Transactions_df.index, Transactions_df['Amount'],color='Green')

plt.title('Transaction Amount VS Index')
plt.xlabel('Transaction Index')
plt.ylabel('Transaction Amount')
plt.grid(True)
plt.show()
```







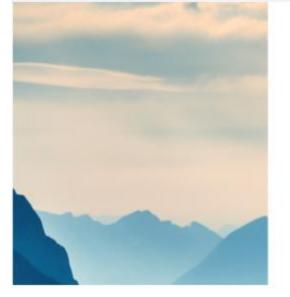
INTRODUCTION TO TABLEAU

Connecting Data Sources

Tableau allows users to connect to various data sources like Excel, SQL databases, and cloud data, enabling seamless integration and analysis across multiple platforms.

Data Preparation in Tableau

Data preparation involves cleaning, transforming, and organizing data within Tableau to ensure accurate visualizations, facilitating effective analysis and decision-making processes.





CREATING DASHBOARDS



Interactive Elements

Interactive elements such as filters, drop- downs, and buttons enhance user engagement, allowing viewers to manipulate visualizations for personalized insights and deeper data exploration.

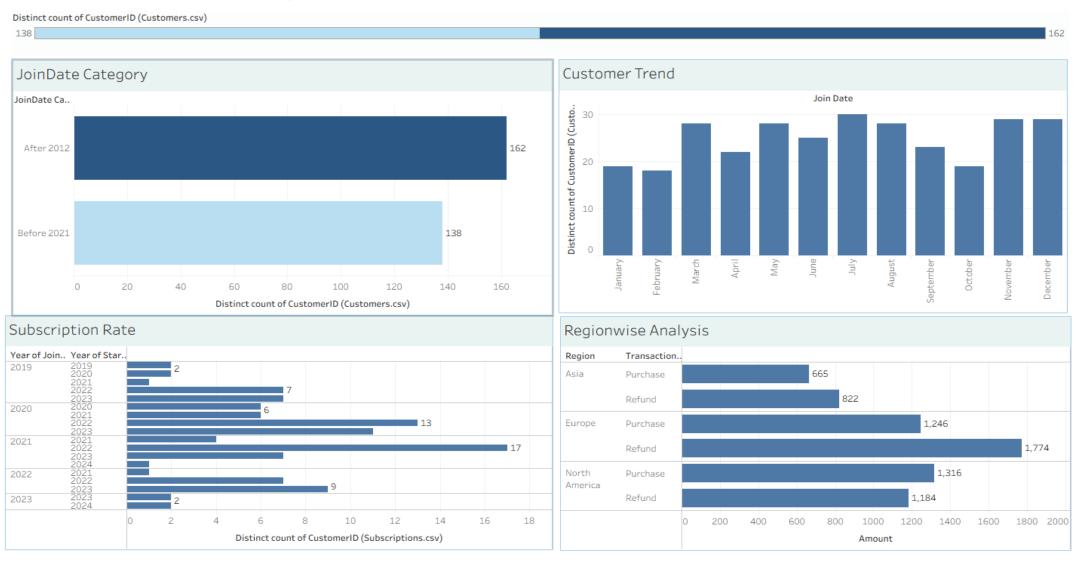


Real-time Updates

Real- time updates enable dashboards to present current data dynamically, ensuring that users have access to the latest information for timely decision- making and strategic initiatives.

VISUALIZING CHURN ANALYSIS REPORT

Customers Churn Analysis



THANKS

