

```
In [1]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt

#Reading the data
customers_df= pd.read_csv("customers.csv")
customers_df.head()
```

```
Out[1]:
```

	CustomerID	FirstName	LastName	Email	PhoneNumber
0	1	John	Doe	john.doe@example.com	123-456-7890
1	2	Jane	Smith	jane.smith@example.com	098-765-4321
2	3	Alice	Johnson	alice.j@example.com	567-890-1234
3	4	Bob	Brown	bob.brown@example.com	234-567-8901
4	5	Charlie	Davis	charlie.d@example.com	345-678-9012

```
In [2]: customers_df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 300 entries, 0 to 299
Data columns (total 8 columns):
#   Column          Non-Null Count  Dtype
---  -
0   CustomerID      300 non-null   int64
1   FirstName       299 non-null   object
2   LastName        299 non-null   object
3   Email           298 non-null   object
4   PhoneNumber     298 non-null   object
5   JoinDate        298 non-null   object
6   Status          299 non-null   object
7   Region          299 non-null   object
dtypes: int64(1), object(7)
memory usage: 18.9+ KB
```

```
In [3]: customers_df.describe()
```

Out[3]:

CustomerID	
count	300.000000
mean	150.500000
std	86.746758
min	1.000000
25%	75.750000
50%	150.500000
75%	225.250000
max	300.000000

```
In [14]: churn_df= pd.read_csv("churn.csv")
churn_df.head()
```

Out[14]:

	ChurnID	CustomerID	ChurnDate	Reason
0	1	22	02-01-2024	Poor Customer Service
1	2	56	06-06-2024	High Prices
2	3	45	10-10-2024	Product Quality Issues
3	4	54	12-12-2024	Lack of Product Features
4	5	82	14-02-2024	Competitor Offerings

```
In [9]: Subscriptions_df= pd.read_csv("Subscriptions.csv")
Subscriptions_df.head()
```

Out[9]:

	SubscriptionID	CustomerID	StartDate	EndDate	PlanType
0	1	1	10-01-2022	09-01-2023	Annual
1	2	2	15-12-2021	14-12-2022	Monthly
2	3	3	20-03-2020	19-03-2021	Annual
3	4	4	25-06-2019	24-06-2020	Annual
4	5	5	14-07-2021	13-07-2022	Monthly

```
In [12]: Transactions_Data= pd.read_csv("Transactions.csv")
Transactions_Data.head()
```

Out[12]:

	TransactionID	CustomerID	TransactionDate	Amount	TransactionType
0	1	15	01-01-2024	32	Purchase
1	2	22	02-01-2024	52	Refund
2	3	43	03-01-2024	165	Purchase
3	4	87	04-01-2024	134	Purchase
4	5	34	05-01-2024	158	Purchase

In [16]: *#1. Print first 5 customers FirstNames and LastNames*

```
Customer_names= customers_df[['FirstName','LastName']]
Customer_names.head()
```

Out[16]:

	FirstName	LastName
0	John	Doe
1	Jane	Smith
2	Alice	Johnson
3	Bob	Brown
4	Charlie	Davis

In [23]: *#2. Find out customers from specific Region (North America)*

```
Customer_Region=customers_df[customers_df['Region']== 'North America']
Customer_Region.head()
print(Customer_Region)
```

	CustomerID	FirstName	LastName	Email	PhoneNumber
\					
0	1	John	Doe	john.doe@example.com	123-456-7890
3	4	Bob	Brown	bob.brown@example.com	234-567-8901
6	7	Eva	Harris	eva.h@example.com	567-890-1234
9	10	Hank	Wilson	hank.w@example.com	890-123-4567
12	13	Kelly	Lewis	kelly.l@example.com	123-456-7890
..
287	288	Rachel	Harris	rachel.harris@example.com	678-901-2345
290	291	Lucas	Taylor	lucas.taylor@example.com	901-234-5678
293	294	Sophia	Morris	sophia.morris@example.com	234-567-8901
296	297	Liam	Wilson	liam.wilson@example.com	567-890-1234
299	300	Lucas	Lee	lucas.lee@example.com	890-123-4567

	JoinDate	Status	Region
0	10-01-2022	Active	North America
3	25-06-2019	Inactive	North America
6	22-12-2020	Inactive	North America
9	27-05-2020	Active	North America
12	15-04-2018	Active	North America
..
287	05-03-2019	Inactive	North America
290	19-08-2019	Active	North America
293	14-12-2019	Inactive	North America
296	10-08-2020	Active	North America
299	01-11-2021	Inactive	North America

[103 rows x 8 columns]

In [26]: *#3. Find customers in specific Regions (Europe or Asia)*

```
Customer_region=customers_df[customers_df['Region'].isin(['Europe','Asia'])]
print(Customer_region)
```

\	CustomerID	FirstName	LastName	Email	PhoneNumber
1	2	Jane	Smith	jane.smith@example.com	098-765-4321
2	3	Alice	Johnson	alice.j@example.com	567-890-1234
4	5	Charlie	Davis	charlie.d@example.com	345-678-9012
5	6	Diana	Clark	diana.c@example.com	456-789-0123
7	8	Frank	Garcia	frank.g@example.com	678-901-2345
..
292	293	Jake	Young	jake.young@example.com	123-456-7890
294	295	Aiden	Davis	aiden.davis@example.com	345-678-9012
295	296	Emily	Brown	emily.brown@example.com	456-789-0123
297	298	Olivia	Perry	olivia.perry@example.com	678-901-2345
298	299	Ella	Johnson	ella.johnson@example.com	789-012-3456

	JoinDate	Status	Region
1	15-12-2021	Inactive	Europe
2	20-03-2020	Active	Asia
4	14-07-2021	Active	Europe
5	30-08-2018	Active	Asia
7	11-03-2019	Active	Europe
..
292	29-05-2021	Active	Asia
294	03-11-2022	Active	Europe
295	19-03-2021	Inactive	Asia
297	12-06-2019	Inactive	Europe
298	18-05-2022	Active	Asia

[196 rows x 8 columns]

In [29]: *#4. Find Customers with Active Status*

```
status= customers_df[customers_df['Status']=='Active']
status.head()
print(status)
```

	CustomerID	FirstName	LastName	Email	PhoneNumber
\					
0	1	John	Doe	john.doe@example.com	123-456-7890
2	3	Alice	Johnson	alice.j@example.com	567-890-1234
4	5	Charlie	Davis	charlie.d@example.com	345-678-9012
5	6	Diana	Clark	diana.c@example.com	456-789-0123
7	8	Frank	Garcia	frank.g@example.com	678-901-2345
..
290	291	Lucas	Taylor	lucas.taylor@example.com	901-234-5678
292	293	Jake	Young	jake.young@example.com	123-456-7890
294	295	Aiden	Davis	aiden.davis@example.com	345-678-9012
296	297	Liam	Wilson	liam.wilson@example.com	567-890-1234
298	299	Ella	Johnson	ella.johnson@example.com	789-012-3456

	JoinDate	Status	Region
0	10-01-2022	Active	North America
2	20-03-2020	Active	Asia
4	14-07-2021	Active	Europe
5	30-08-2018	Active	Asia
7	11-03-2019	Active	Europe
..
290	19-08-2019	Active	North America
292	29-05-2021	Active	Asia
294	03-11-2022	Active	Europe
296	10-08-2020	Active	North America
298	18-05-2022	Active	Asia

[152 rows x 8 columns]

```
In [31]: #5. Visualization of Active & Inactive Customers

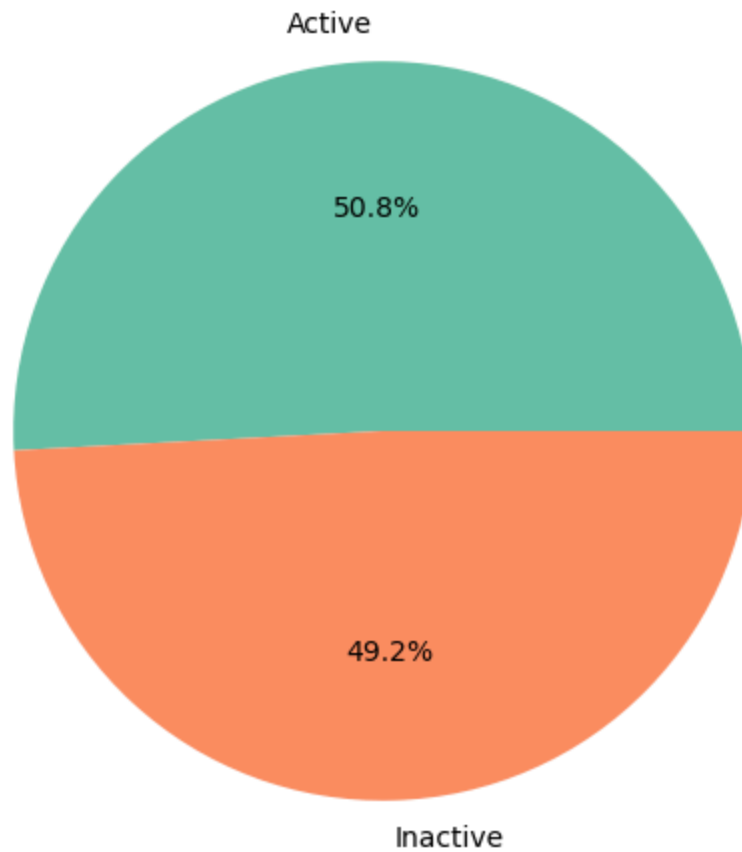
#Calculate the number of active / Inactive customers

status_count=customers_df['Status'].value_counts()

#Create a pie chart to show the distribution

plt.figure(figsize=(8,6))
plt.pie(status_count, labels=status_count.index, autopct='%1.1f%%', colors=[
plt.title('Distribution of Customer Status')
plt.show()
```

Distribution of Customer Status



```
In [36]: #6. List down the customers who joined after '2020-01-01'  
#convert the 'JoinDate' column to date time format  
customer_joined=customers_df[pd.to_datetime(customers_df['JoinDate'],format=  
print(customer_joined)
```

	CustomerID	FirstName	LastName	Email	PhoneNumber
\					
0	1	John	Doe	john.doe@example.com	123-456-7890
1	2	Jane	Smith	jane.smith@example.com	098-765-4321
2	3	Alice	Johnson	alice.j@example.com	567-890-1234
4	5	Charlie	Davis	charlie.d@example.com	345-678-9012
6	7	Eva	Harris	eva.h@example.com	567-890-1234
..
294	295	Aiden	Davis	aiden.davis@example.com	345-678-9012
295	296	Emily	Brown	emily.brown@example.com	456-789-0123
296	297	Liam	Wilson	liam.wilson@example.com	567-890-1234
298	299	Ella	Johnson	ella.johnson@example.com	789-012-3456
299	300	Lucas	Lee	lucas.lee@example.com	890-123-4567

	JoinDate	Status	Region
0	10-01-2022	Active	North America
1	15-12-2021	Inactive	Europe
2	20-03-2020	Active	Asia
4	14-07-2021	Active	Europe
6	22-12-2020	Inactive	North America
..
294	03-11-2022	Active	Europe
295	19-03-2021	Inactive	Asia
296	10-08-2020	Active	North America
298	18-05-2022	Active	Asia
299	01-11-2021	Inactive	North America

[228 rows x 8 columns]

```
In [39]: #7. Create a visualisation(bar chart) who joined after and before 2021
#converting JoinDate into DATE TIME FORMAT
customers_df['JoinDate']=pd.to_datetime(customers_df['JoinDate'],format='%d-

# Filtering the customers who joined before and after 2021

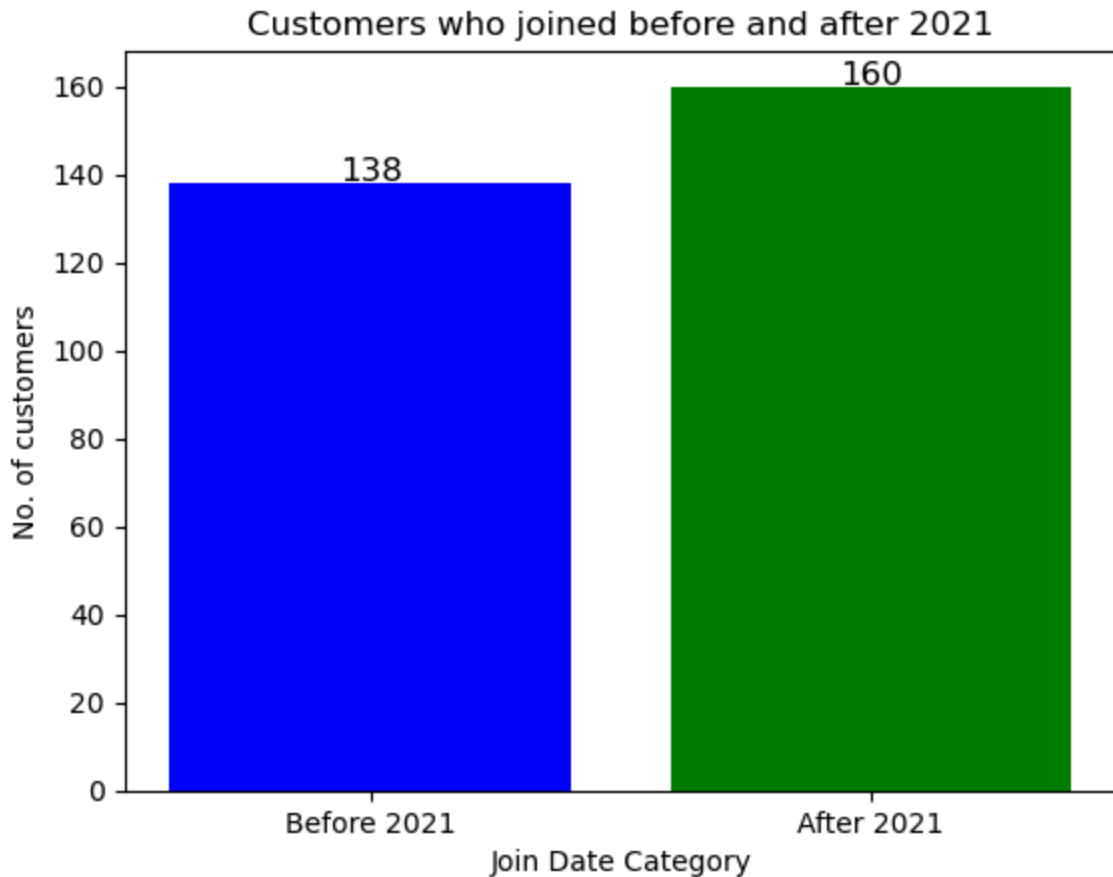
before_2021=customers_df[customers_df['JoinDate']<'2021-01-01']
after_2021=customers_df[customers_df['JoinDate']>'2021-01-01']

# 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
#counts.items(): This gives you both the key and the values from dictionary
```

In []:

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

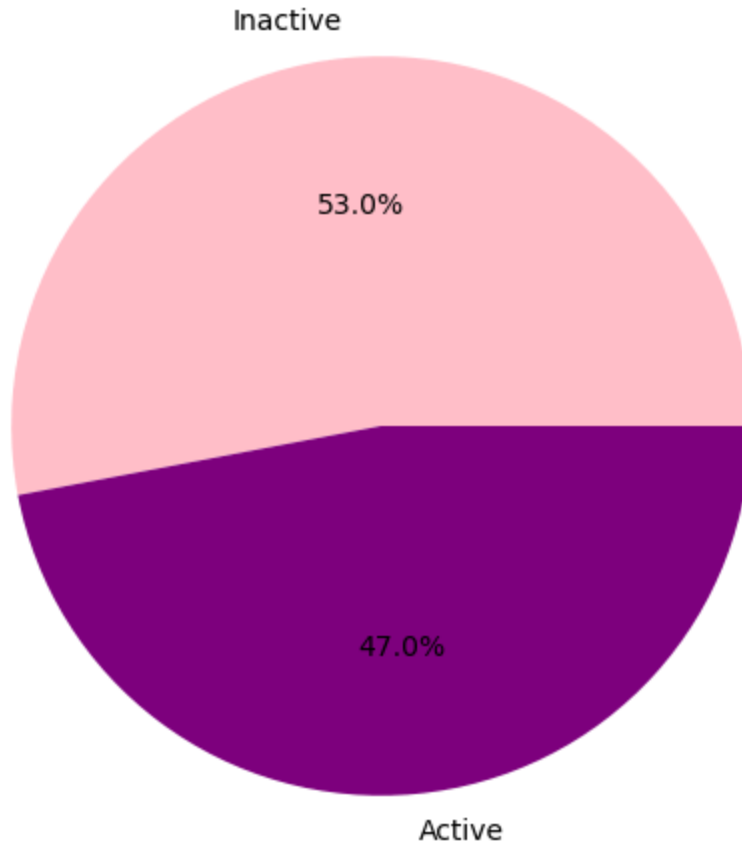
47

```
In [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=
plt.title('Distribution of Customers')
plt.show()
```

Distribution of Customers

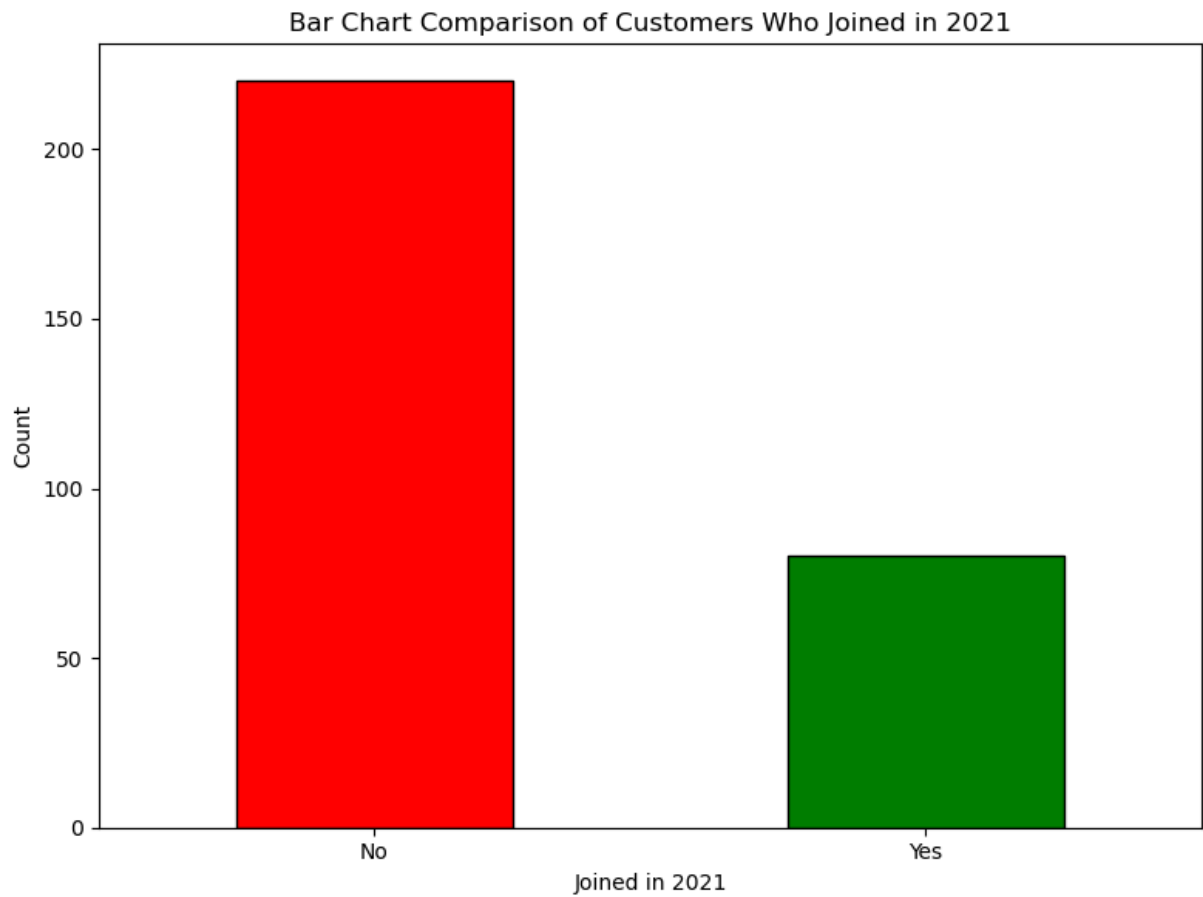


```
In [101... customers_df['JoinDate']=pd.to_datetime(customers_df['JoinDate'],format='%d-
customers_df['Joined2021'] = pd.to_datetime(customers_df['JoinDate']).dt.yea

# Count the number of customers who are in yes or no
joined_2021_counts = customers_df['Joined2021'].value_counts()

# Create a bar chart
plt.figure(figsize=(8, 6))
joined_2021_counts.plot(kind='bar', color=['red', 'green'], edgecolor='black')
plt.xlabel('Joined in 2021')
plt.ylabel('Count')
plt.title('Bar Chart Comparison of Customers Who Joined in 2021')
plt.xticks(ticks=[0, 1], labels=['No', 'Yes'], rotation=0)
plt.tight_layout()

plt.show()
```



```
In [95]: customers_df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 300 entries, 0 to 299
Data columns (total 8 columns):
#   Column          Non-Null Count  Dtype
---  -
0   CustomerID      300 non-null   int64
1   FirstName       299 non-null   object
2   LastName        299 non-null   object
3   Email           298 non-null   object
4   PhoneNumber      298 non-null   object
5   JoinDate        298 non-null   object
6   Status          299 non-null   object
7   Region          299 non-null   object
dtypes: int64(1), object(7)
memory usage: 18.9+ KB
```

```
In [98]: customers_df.describe()
```

Out[98]:

CustomerID	
count	300.000000
mean	150.500000
std	86.746758
min	1.000000
25%	75.750000
50%	150.500000
75%	225.250000
max	300.000000

In [102]:

```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt

#Reading the data
Transactions_df= pd.read_csv("Transactions.csv")
Transactions_df.head()
```

Out[102]:

	TransactionID	CustomerID	TransactionDate	Amount	TransactionType
0	1	15	01-01-2024	32	Purchase
1	2	22	02-01-2024	52	Refund
2	3	43	03-01-2024	165	Purchase
3	4	87	04-01-2024	134	Purchase
4	5	34	05-01-2024	158	Purchase

In [103]:

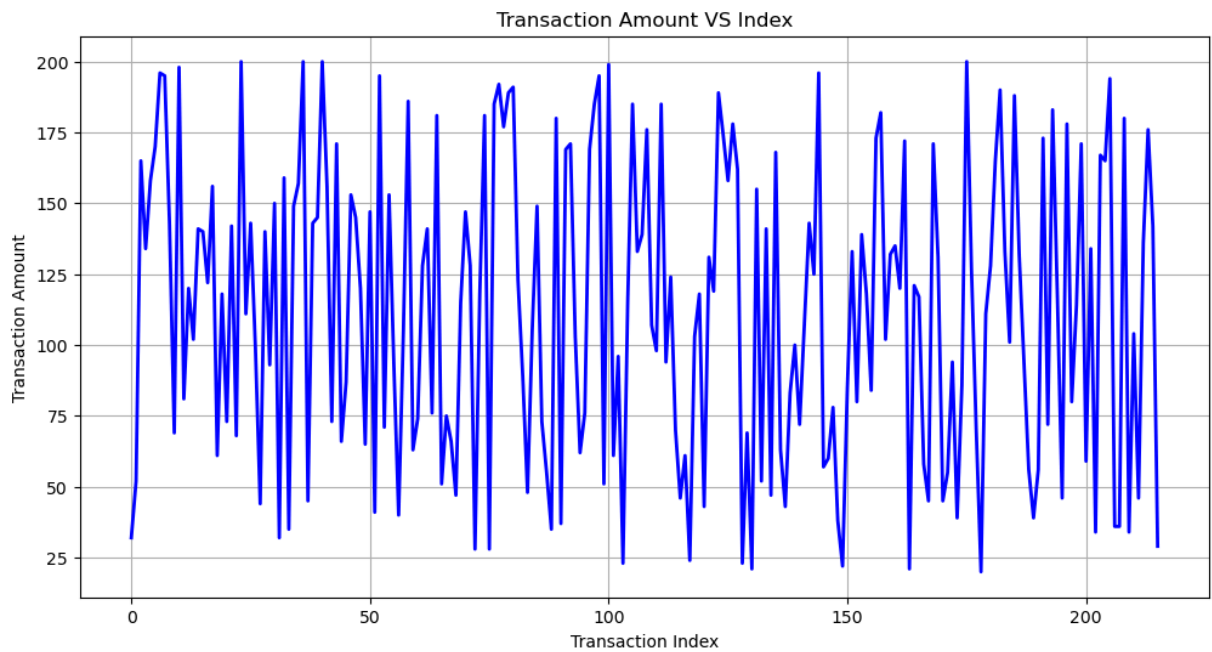
```
#calculate the average transaction amount
Average_amount= Transactions_df['Amount'].mean()
print(f"Average Transaction Amount: {Average_amount}")
```

Average Transaction Amount: 112.19907407407408

In [111]:

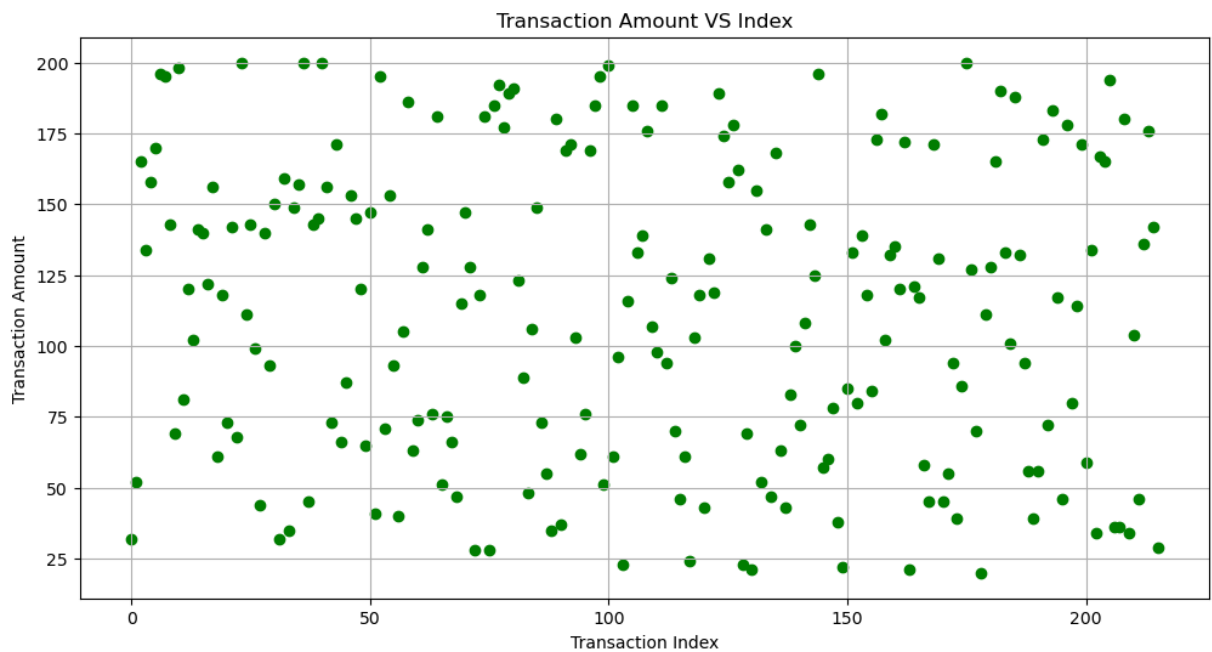
```
# Visualize / Analysis for Transaction Amount w.r.t to Index - Line graph
plt.figure(figsize=(12, 6))
plt.plot(Transactions_df.index, Transactions_df['Amount'],linestyle='-', lir

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



```
In [116... # 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()
```



```
In [118... # Transactions which are more than $100
Max_amount= Transactions_df[Transactions_df['Amount']>100]
print(Max_amount)
```

	TransactionID	CustomerID	TransactionDate	Amount	TransactionType
2	3	43	03-01-2024	165	Purchase
3	4	87	04-01-2024	134	Purchase
4	5	34	05-01-2024	158	Purchase
5	6	56	06-01-2024	170	Refund
6	7	23	07-01-2024	196	Purchase
..
208	209	263	01-09-2024	180	Refund
210	211	265	03-09-2024	104	Purchase
212	213	267	05-09-2024	136	Refund
213	214	268	06-09-2024	176	Purchase
214	215	269	07-09-2024	142	Purchase

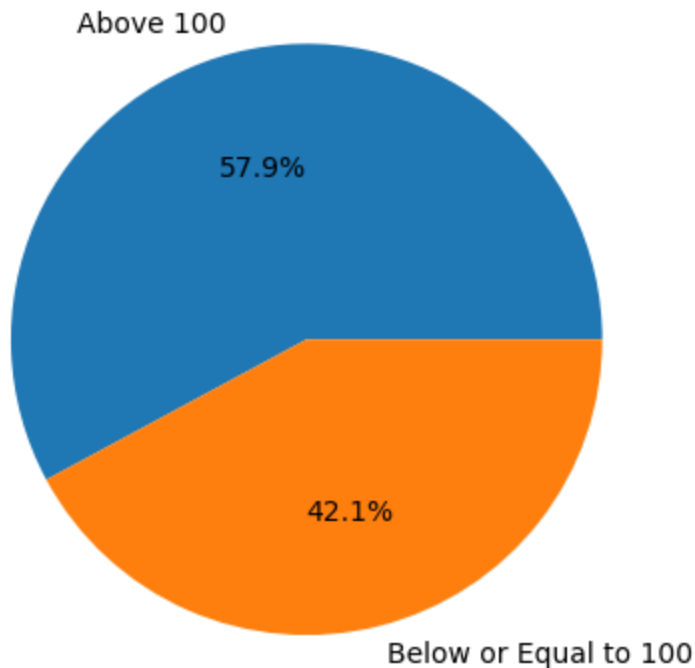
[125 rows x 5 columns]

```
In [121]: # Using a pie chart .. show me the % breakdown of "Above $100 and Below / E
above_100= Transactions_df[Transactions_df['Amount']>100].shape[0]
Below_or_equal_100 = Transactions_df[Transactions_df['Amount']<=100].shape[0]

labels= ['Above 100', 'Below or Equal to 100']
sizes= [above_100, Below_or_equal_100]

plt.pie(sizes, labels=labels, autopct='%1.1f%%')
```

```
Out[121]: ([<matplotlib.patches.Wedge at 0x1af3199f110>,
<matplotlib.patches.Wedge at 0x1af3199f4d0>],
[Text(-0.26921771464971617, 1.0665466807030923, 'Above 100'),
Text(0.26921755360564475, -1.0665467213538242, 'Below or Equal to 100')],
[Text(-0.14684602617257242, 0.5817527349289594, '57.9%'),
Text(0.14684593833035167, -0.5817527571020858, '42.1%')])
```



```
In [122]: # Bar chart .. "Transactions Categories" Vs "No. of Transactions"
```

```

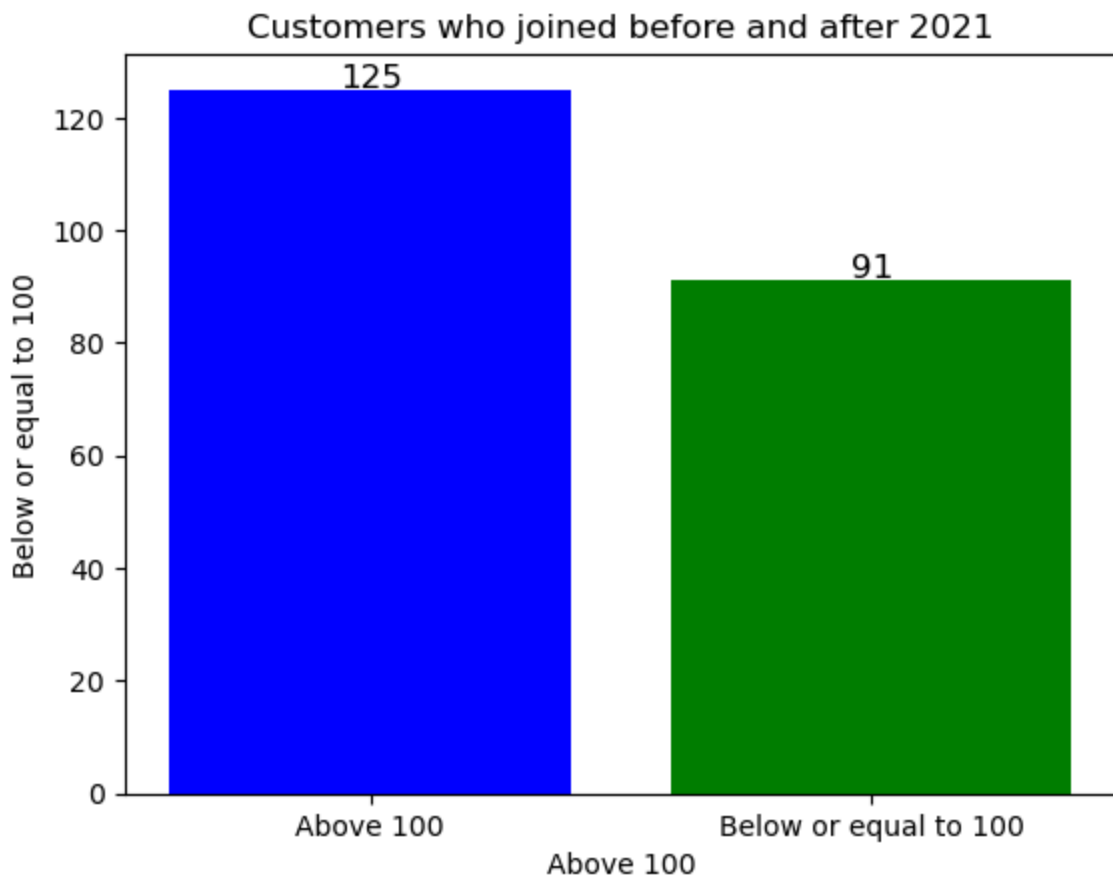
above_100= Transactions_df[Transactions_df['Amount']>100].shape[0]
Below_or_equal_100 = Transactions_df[Transactions_df['Amount']<=100].shape[0]

# Count the number of customers in each of the category
counts={'Above 100': above_100,
        'Below or equal to 100' : Below_or_equal_100}

# create a bar chart
plt.bar(counts.keys(),counts.values(), color=['blue','green'])
plt.xlabel('Above 100')
plt.ylabel('Below or equal to 100')
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()

```



```

In [126... import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt

#Reading the data
Churn_df= pd.read_csv("Churn.csv")
Churn_df.head()

```

Out[126...	ChurnID	CustomerID	ChurnDate	Reason
0	1	22	02-01-2024	Poor Customer Service
1	2	56	06-06-2024	High Prices
2	3	45	10-10-2024	Product Quality Issues
3	4	54	12-12-2024	Lack of Product Features
4	5	82	14-02-2024	Competitor Offerings

```
In [127... #Find the number of customers who left the company/ churned for each reason
reason_counts=Churn_df['Reason'].value_counts
print(reason_counts)
```

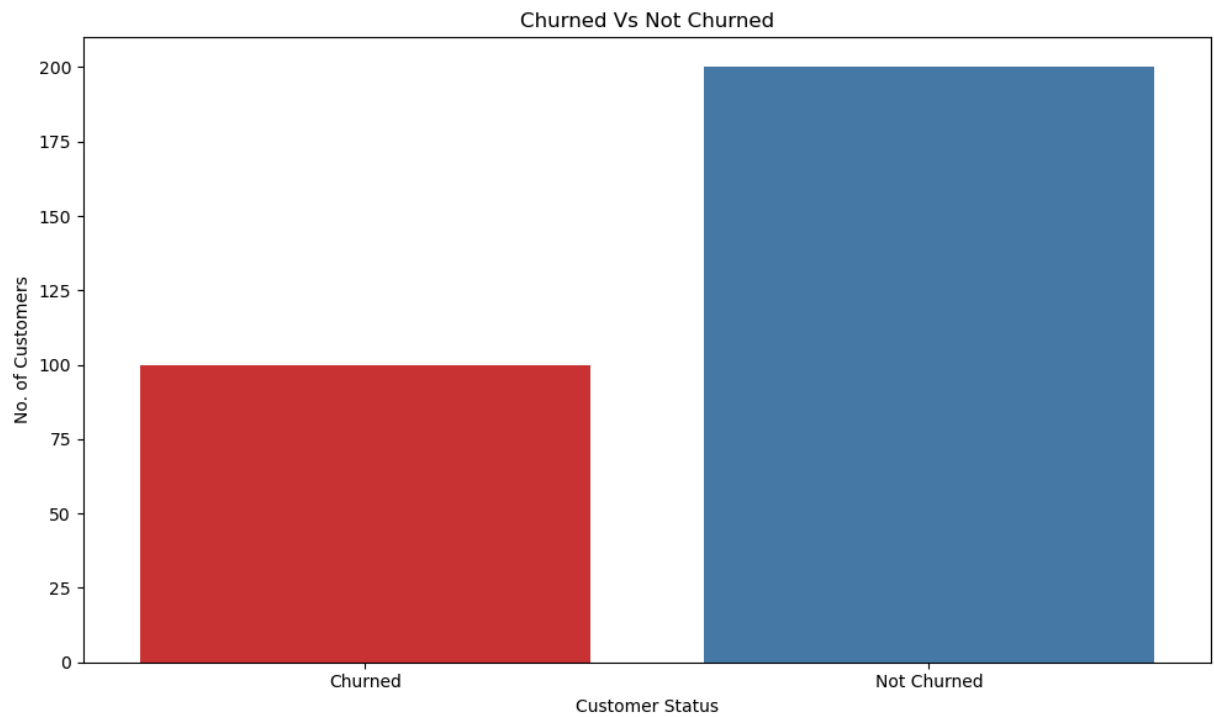
```
<bound method IndexOpsMixin.value_counts of 0      Poor Customer Service
1      High Prices
2      Product Quality Issues
3      Lack of Product Features
4      Competitor Offerings
...
95     Inconvenience
96     Lack of Engagement
97     Unmet Expectations
98     Price Increases
99     Personal Circumstances
Name: Reason, Length: 100, dtype: object>
```

```
In [134... # Visualization of Churned vs Not Churned customers
#Assuming we have a "Status" column indicating "Churn" and "Not Churned"

total_customers=customers_df.shape[0]
total_churned_customers=Churn_df.shape[0]
total_not_churned_customers= total_customers - total_churned_customers

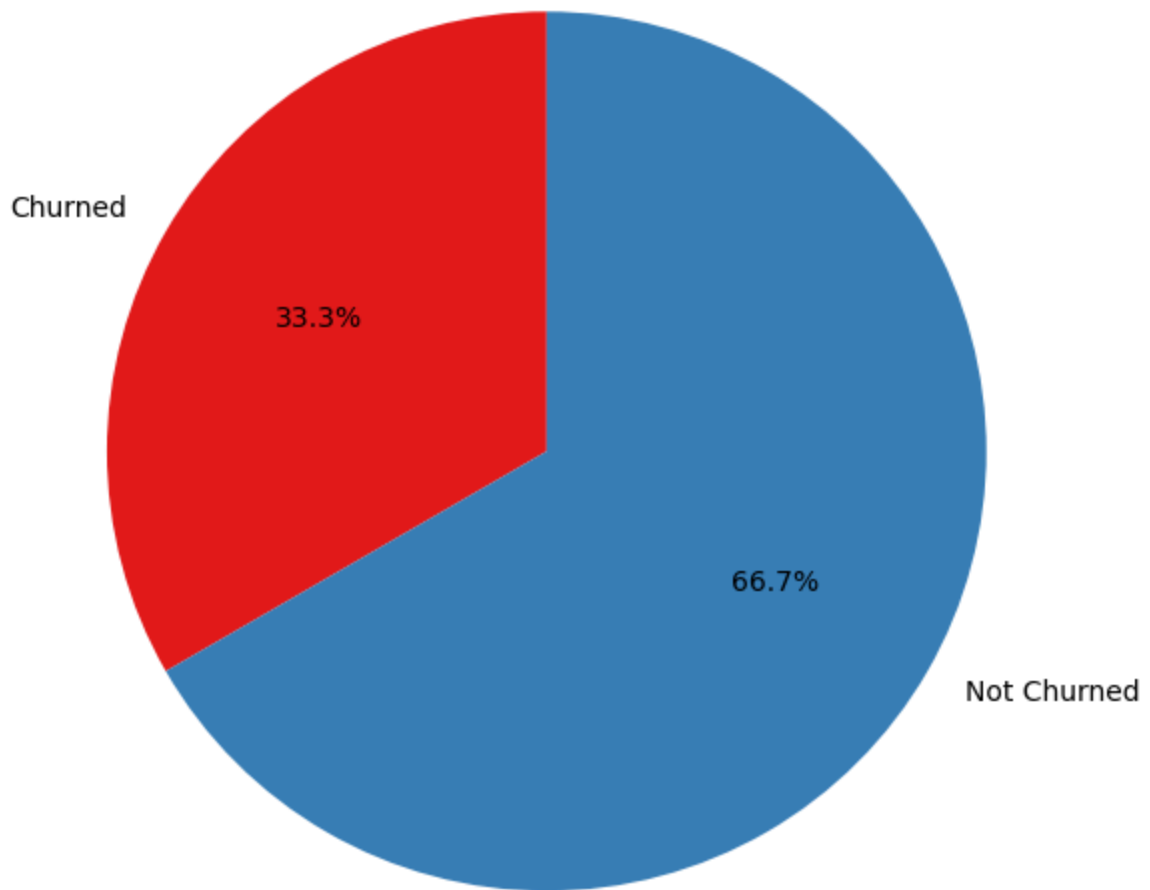
#going to represent the two categories
status= pd.DataFrame({
    'Status': ['Churned', 'Not Churned'],
    'Count' : [total_churned_customers, total_not_churned_customers]
})

#Bar Chart
plt.figure(figsize=(10,6))
sns.barplot(x='Status',y='Count', data= status, palette='Set1', hue='Status')
plt.xlabel('Customer Status')
plt.ylabel('No. of Customers')
plt.title('Churned Vs Not Churned')
plt.tight_layout()
plt.show()
```

```
In [136... plt.figure(figsize=(10,6))
plt.pie(status['Count'],labels=status['Status'],autopct='%1.1f%%',startangle
plt.title('Churned Vs Not Churned')
plt.tight_layout()
plt.show()
```

Churned Vs Not Churned



```
In [139... #List Churn Reasons with Service"
#Filter the 'Churn_df' dataframe to include only rows where the "reason" col

service_reasons= Churn_df[Churn_df['Reason'].str.contains('service', case=False)]
print(service_reasons[['CustomerID', 'Reason']])
```

	CustomerID	Reason
0	22	Poor Customer Service
10	27	Poor Customer Service
20	72	Poor Customer Service
30	137	Poor Customer Service
40	81	Poor Customer Service
50	34	Poor Customer Service
60	102	Poor Customer Service
70	43	Poor Customer Service
80	104	Poor Customer Service
90	132	Poor Customer Service

```
In [140... import pandas as pd
import numpy as np
import seaborn as sns
```

```
import matplotlib.pyplot as plt

#Reading the data
Subscriptions_df=pd.read_csv("Subscriptions.csv")
Subscriptions_df.head()
```

Out[140...

	SubscriptionID	CustomerID	StartDate	EndDate	PlanType
0	1	1	10-01-2022	09-01-2023	Annual
1	2	2	15-12-2021	14-12-2022	Monthly
2	3	3	20-03-2020	19-03-2021	Annual
3	4	4	25-06-2019	24-06-2020	Annual
4	5	5	14-07-2021	13-07-2022	Monthly

In [144...

```
# Join Customers and Subscriptions
join_df= pd.merge(customers_df, Subscriptions_df, on='CustomerID', how='inne
print(join_df)
```

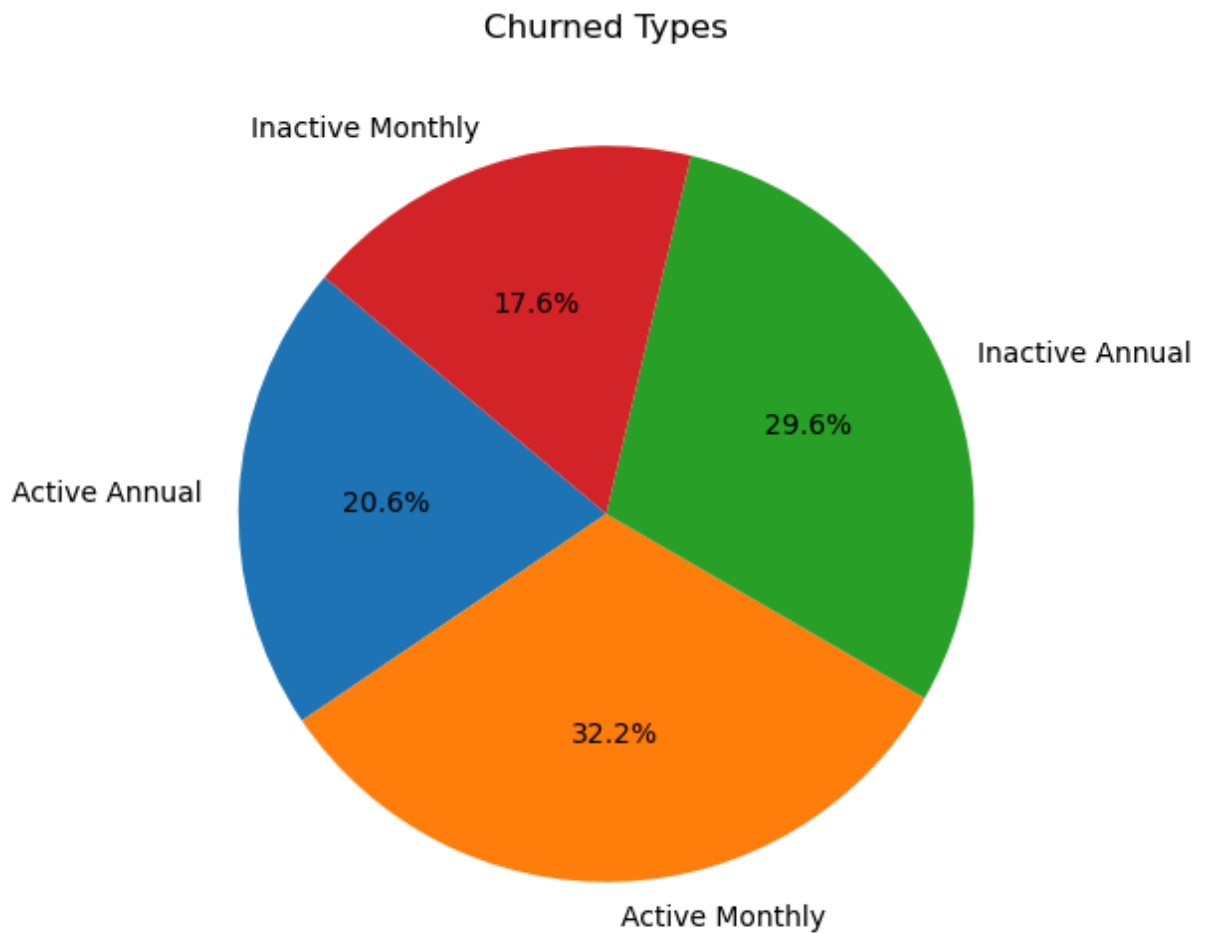
\	CustomerID	FirstName	LastName	Email	PhoneNumber
0	1	John	Doe	john.doe@example.com	123-456-7890
1	1	John	Doe	john.doe@example.com	123-456-7890
2	1	John	Doe	john.doe@example.com	123-456-7890
3	2	Jane	Smith	jane.smith@example.com	098-765-4321
4	2	Jane	Smith	jane.smith@example.com	098-765-4321
..
394	296	Emily	Brown	emily.brown@example.com	456-789-0123
395	297	Liam	Wilson	liam.wilson@example.com	567-890-1234
396	298	Olivia	Perry	olivia.perry@example.com	678-901-2345
397	299	Ella	Johnson	ella.johnson@example.com	789-012-3456
398	300	Lucas	Lee	lucas.lee@example.com	890-123-4567

	JoinDate	Status	Region	SubscriptionID	StartDate	\
0	10-01-2022	Active	North America	1	10-01-2022	
1	10-01-2022	Active	North America	44	10-01-2023	
2	10-01-2022	Active	North America	101	01-07-2023	
3	15-12-2021	Inactive	Europe	2	15-12-2021	
4	15-12-2021	Inactive	Europe	45	15-12-2022	
..	
394	19-03-2021	Inactive	Asia	395	01-01-2023	
395	10-08-2020	Active	North America	396	01-02-2023	
396	12-06-2019	Inactive	Europe	397	01-03-2023	
397	18-05-2022	Active	Asia	398	01-04-2023	
398	01-11-2021	Inactive	North America	399	01-05-2023	

	EndDate	PlanType
0	09-01-2023	Annual
1	09-01-2024	Monthly
2	30-06-2024	Monthly
3	14-12-2022	Monthly
4	14-12-2023	Annual
..
394	01-01-2024	Annual
395	01-02-2024	Monthly
396	01-03-2024	Annual
397	01-04-2024	Monthly
398	01-05-2024	Annual

[399 rows x 12 columns]

```
In [146... #visualization for different Customer Groups [Active-Annual, Active-monthly,
result= customers_df.merge(Subscriptions_df,on='CustomerID')
result['flag'] = result['Status']+" "+result['PlanType']
data=result.groupby('flag', as_index=False).agg(total_cnt=('CustomerID', 'count'))
plt.figure(figsize=(6,6))
plt.pie(data['total_cnt'],labels=data['flag'],autopct="%1.1f%%", startangle=0)
plt.title("Churned Types")
plt.show()
```



```
In [10]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt

Transactions_df= pd.read_csv("Transactions.csv")
Transactions_df.head()
```

```
Out[10]:
```

	TransactionID	CustomerID	TransactionDate	Amount	TransactionType
0	1	15	01-01-2024	32	Purchase
1	2	22	02-01-2024	52	Refund
2	3	43	03-01-2024	165	Purchase
3	4	87	04-01-2024	134	Purchase
4	5	34	05-01-2024	158	Purchase

```
In [11]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
```

```
customers_df= pd.read_csv("customers.csv")
customers_df.head()
```

Out[11]:

	CustomerID	FirstName	LastName	Email	PhoneNumber
0	1	John	Doe	john.doe@example.com	123-456-7890
1	2	Jane	Smith	jane.smith@example.com	098-765-4321
2	3	Alice	Johnson	alice.j@example.com	567-890-1234
3	4	Bob	Brown	bob.brown@example.com	234-567-8901
4	5	Charlie	Davis	charlie.d@example.com	345-678-9012

```
In [10]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt

Transactions_df= pd.read_csv("Transactions.csv")
Transactions_df.head()
```

Out[10]:

	TransactionID	CustomerID	TransactionDate	Amount	TransactionType
0	1	15	01-01-2024	32	Purchase
1	2	22	02-01-2024	52	Refund
2	3	43	03-01-2024	165	Purchase
3	4	87	04-01-2024	134	Purchase
4	5	34	05-01-2024	158	Purchase

```
In [12]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt

Subscriptions_df= pd.read_csv("Subscriptions.csv")
Subscriptions_df.head()
```

```
Out[12]:
```

	SubscriptionID	CustomerID	StartDate	EndDate	PlanType
0	1	1	10-01-2022	09-01-2023	Annual
1	2	2	15-12-2021	14-12-2022	Monthly
2	3	3	20-03-2020	19-03-2021	Annual
3	4	4	25-06-2019	24-06-2020	Annual
4	5	5	14-07-2021	13-07-2022	Monthly

```
In [13]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt

Churn_df= pd.read_csv("Churn.csv")
Churn_df.head()
```

```
Out[13]:
```

	ChurnID	CustomerID	ChurnDate	Reason
0	1	22	02-01-2024	Poor Customer Service
1	2	56	06-06-2024	High Prices
2	3	45	10-10-2024	Product Quality Issues
3	4	54	12-12-2024	Lack of Product Features
4	5	82	14-02-2024	Competitor Offerings

```
In [14]: #EDA - Exploratory Data Analysis
#Data Cleaning
# Merge All Datasets

merged_df= pd.merge(customers_df, Subscriptions_df,on='CustomerID', how='left')
merged_df=pd.merge(merged_df,Transactions_df,on='CustomerID',how='left')
final_df=pd.merge(merged_df,Churn_df,on='CustomerID',how='left')
#Display the result
final_df.head()
```

```
Out[14]:
```

	CustomerID	FirstName	LastName	Email	PhoneNumber
0	1	John	Doe	john.doe@example.com	123-456-7890
1	1	John	Doe	john.doe@example.com	123-456-7890
2	1	John	Doe	john.doe@example.com	123-456-7890
3	2	Jane	Smith	jane.smith@example.com	098-765-4321
4	2	Jane	Smith	jane.smith@example.com	098-765-4321

```
In [15]: final_df.shape
```

```
Out[15]: (443, 19)
```

```
In [16]: #Add a column named as "churned" based on the presence of "churnID"  
final_df['Churned']=final_df['ChurnID'].apply(lambda x:1 if pd.notna(x)else  
final_df.head()
```

```
Out[16]:
```

	CustomerID	FirstName	LastName	Email	PhoneNumber
0	1	John	Doe	john.doe@example.com	123-456-7890
1	1	John	Doe	john.doe@example.com	123-456-7890
2	1	John	Doe	john.doe@example.com	123-456-7890
3	2	Jane	Smith	jane.smith@example.com	098-765-4321
4	2	Jane	Smith	jane.smith@example.com	098-765-4321

```
In [156... #Delete Missing values  
missing_values=final_df.isnull().sum()  
print("Missing values in each column:")  
print(missing_values)
```

Missing values in each column:

```
CustomerID      0  
FirstName       1  
LastName        1  
Email           2  
PhoneNumber     2  
JoinDate        2  
Status          1  
Region          1  
SubscriptionID  1  
StartDate       1  
EndDate         1  
PlanType        1  
TransactionID   141  
TransactionDate 141  
Amount         141  
TransactionType 141  
ChurnID         286  
ChurnDate       286  
Reason          286  
Churned         0  
dtype: int64
```

```
In [18]: #Fill missing values in "FirstName" and "LastName"  
final_df['FirstName']=final_df['FirstName'].fillna('xyz')  
final_df['LastName']=final_df['LastName'].fillna('xyz')
```



```
print("Missing values after filling:")
print(final_df[['FirstName', 'LastName']].isnull().sum())
```

Missing values after filling:

```
FirstName      0
LastName      0
dtype: int64
```

```
In [19]: #Filter rows where 'Email' or 'PhoneNumber' is missing
missing_email_phone=final_df[final_df['Email'].isnull() | final_df['PhoneNum

print("Rows with missing phone number & email:")
print(missing_email_phone)
```

Rows with missing phone number & email:

	CustomerID	FirstName	LastName	Email	PhoneNumber
\					
395	253	Robert	Jones	robert.jones@example.com	NaN
398	256	Sarah	Davis	NaN	456-789-0123
408	266	Sophia	Johnson	NaN	567-890-1234
410	268	Ethan	Davis	ethan.davis@example.com	NaN

	JoinDate	Status	Region	SubscriptionID	StartDate	EndDate	\
395	01-05-2021	Active	Asia	352.0	01-06-2023	01-06-2024	
398	30-03-2022	Inactive	Europe	355.0	01-09-2023	01-09-2024	
408	11-11-2021	Inactive	Asia	365.0	01-07-2022	01-07-2023	
410	12-12-2019	Inactive	Europe	367.0	01-09-2022	01-09-2023	

	PlanType	TransactionID	TransactionDate	Amount	TransactionType	ChurnID
\						
395	Monthly	199.0	22-08-2024	114.0	Purchase	NaN
398	Annual	202.0	25-08-2024	134.0	Refund	NaN
408	Annual	212.0	04-09-2024	46.0	Purchase	NaN
410	Annual	214.0	06-09-2024	176.0	Purchase	NaN

	ChurnDate	Reason	Churned
395	NaN	NaN	0
398	NaN	NaN	0
408	NaN	NaN	0
410	NaN	NaN	0

```
In [20]: #Drop rows where Churned=0
final_df.dropna(subset=['Email', 'PhoneNumber'], how='any', inplace=True)

print(final_df[['Email', 'PhoneNumber']].isnull().sum())
```

```
Email      0
PhoneNumber 0
dtype: int64
```

```
In [21]: #Handle JoinDate , Status and Region
missing_values_row=final_df[final_df[['JoinDate', 'Status', 'Region']].isnull(
print("Rows with missing JoinDate, Status & Region:")
missing_values_row
```

Rows with missing JoinDate, Status & Region:

```
Out[21]:
```

	CustomerID	FirstName	LastName	Email	PhoneNui
396	254	Emily	Taylor	emily.taylor@example.com	234-567
397	255	Michael	Brown	michael.brown@example.com	345-678
407	265	Hannah	Moore	hannah.moore@example.com	456-789

```
In [24]: #drop rows where both JoinDate, Status, and Region is missing:
final_df.dropna(subset=['JoinDate','Status','Region'], how='any',inplace=True)

print(final_df[['JoinDate','Status','Region']].isnull().sum())

JoinDate    0
Status      0
Region      0
dtype: int64
```

```
In [25]: final_df.shape
```

```
Out[25]: (436, 20)
```

```
In [26]: #Handle SubscriptionID, StartDate,EndDate,and PlanType
missing_values_row=final_df[final_df[['SubscriptionID','StartDate','EndDate','PlanType']].isnull().any()]
print("Rows with missing SubscriptionID, StartDate,EndDate,PlanType:")
missing_values_row

Rows with missing SubscriptionID, StartDate,EndDate,PlanType:
```

```
Out[26]:
```

	CustomerID	FirstName	LastName	Email	PhoneNumber
343	201	Ethan	Brown	ethan.b@example.com	123-456-7906

```
In [28]: #drop rows where SubscriptionID, StartDate,EndDate,and PlanType are missing:
final_df.dropna(subset=['SubscriptionID','StartDate','EndDate','PlanType'], how='any',inplace=True)

print(final_df[['SubscriptionID','StartDate','EndDate','PlanType']].isnull().sum())

SubscriptionID    0
StartDate         0
EndDate           0
PlanType          0
dtype: int64
```

```
In [29]: final_df.shape
```

```
Out[29]: (435, 20)
```

```
In [31]: #Set "TransactionID" to one less than the minimum existing transaction ID
#Set "TransactionDate" to a date 10 years before the minimum existing transaction date
#Set Amount to 0
#Set Transaction Type to 'No Transaction'
```

```

import pandas as pd
from datetime import timedelta

#Ensure TransactionDate is in datetime format,
Transactions_df['TransactionDate']=pd.to_datetime(Transactions_df['Transacti

#Find the minimum TransactionDate and TransactionDate
min_transaction_id= Transactions_df['TransactionID'].min()
min_transaction_date=Transactions_df['TransactionDate'].min()

#Define the date 10 years before the minimum transaction Date
ten_years_prior= min_transaction_date - timedelta(days=365*10)

#Fill missing values for customers with no transactions
final_df.loc[final_df['TransactionID'].isnull(),'TransactionID']=min_transac
final_df.loc[final_df['TransactionDate'].isnull(),'TransactionDate']=ten_yea
final_df.loc[final_df['Amount'].isnull(),'Amount']=0
final_df.loc[final_df['TransactionType'].isnull(),'TransactionType']='No Tra

print(final_df[['TransactionID','TransactionDate','Amount','TransactionType']

```

```

TransactionID      0
TransactionDate    0
Amount             0
TransactionType    0
dtype: int64

```

```

In [32]: missing_values=final_df.isnull().sum()
print("Missing values in each column:")
print(missing_values)

```

```

Missing values in each column:
CustomerID      0
FirstName       0
LastName        0
Email           0
PhoneNumber     0
JoinDate        0
Status          0
Region          0
SubscriptionID  0
StartDate       0
EndDate         0
PlanType        0
TransactionID   0
TransactionDate 0
Amount          0
TransactionType 0
ChurnID         278
ChurnDate       278
Reason          278
Churned         0
dtype: int64

```

```

In [33]: final_df.shape

```

Out[33]: (435, 20)

```
In [38]: import pandas as pd
         from datetime import timedelta

         #Ensure TransactionDate is in datetime format,
         Churn_df['ChurnDate']=pd.to_datetime(Churn_df['ChurnDate'],errors='coerce')

         #Find the minimum TransactionDate and TransactionDate
         min_churn_id= Churn_df['ChurnID'].min()
         min_churn_date=Churn_df['ChurnDate'].min()

         #Define the date 10 years before the minimum transaction Date
         ten_years_prior= min_churn_date - timedelta(days=365*10)

         #Fill missing values for customers with no transactions
         final_df.loc[final_df['ChurnID'].isnull(),'ChurnID']=min_churn_id-1
         final_df.loc[final_df['ChurnDate'].isnull(),'ChurnDate']=ten_years_prior.str
         final_df.loc[final_df['Reason'].isnull(),'Reason']='Unknown Reason'

         print(final_df[['ChurnID','ChurnDate','Reason']].isnull().sum())
```

```
ChurnID      0
ChurnDate    0
Reason       0
dtype: int64
```

```
In [39]: missing_values=final_df.isnull().sum()
         print("Missing values in each column:")
         print(missing_values)
```

Missing values in each column:

```
CustomerID      0
FirstName       0
LastName        0
Email           0
PhoneNumber     0
JoinDate        0
Status          0
Region          0
SubscriptionID  0
StartDate       0
EndDate         0
PlanType        0
TransactionID   0
TransactionDate 0
Amount          0
TransactionType 0
ChurnID         0
ChurnDate       0
Reason          0
Churned         0
dtype: int64
```

```
In [40]: final_df.shape
```

Out[40]: (435, 20)

In []:

This notebook was converted with convert.ploomber.io