N VAMSI

AP21110010844

BATCH 99

Netflix Userbase Data Analysis



Abstract

- This Project focuses on analyzing Netflix user data and subscription patterns to gain insights into customer behavior
- The dataset includes information on user ID, subscription type, monthly revenue, join date, last payment date, country, age, gender, device, and plan duration.
- The main objectives of this analysis are to understand user preferences, identify trends in subscription adoption, and visualize key metrics for better decision-making.

Introduction

- Netflix, a leading online streaming platform, has revolutionized the way people consume entertainment content. With millions of subscribers worldwide, it has become essential to understand user behavior and subscription patterns to optimize service offerings and enhance customer satisfaction.
- This Project aims to analyze a comprehensive dataset containing key attributes such as user ID, subscription type, monthly revenue, join date, last payment date, country, age, gender, device, and plan duration.
- By conducting in-depth analysis and visualizations, we seek to uncover valuable insights that can assist Netflix in making informed decisions for its business growth.

System Requirements

Operating System:

Windows 10, macOS, or Linux (Ubuntu, CentOS, etc.)

Software:

- Python (3.7 or higher) with necessary libraries: Pandas, NumPy, Matplotlib, Seaborn, Plotly, etc.
- Jupyter Notebook or any preferred integrated development environment (IDE) for Python.
- SQL database management system (DBMS) for data storage and retrieval.

Hardware:

- Processor: Intel Core i5 or equivalent AMD processor (or higher).
- RAM: 8GB (or higher) for smooth data handling and analysis.
- Storage: At least 50GB of free space for storing the dataset and analysis results.
- Graphics Card: A dedicated graphics card is not strictly required for this analysis, but it can enhance the performance of data visualization, especially for large datasets.

Uses of Data Analysis Library

• The combination of NumPy, pandas, scikit-learn, Matplotlib, and Seaborn can provide a comprehensive toolkit for data analysis project. Here are some specific uses of each library within a project:

1.NumPy:

- NumPy provides powerful numerical operations and arrays, making it useful for mathematical computations and manipulation of large datasets.
- It offers essential functions for array manipulation, such as reshaping, slicing, indexing, and aggregations.

2.Pandas:

- It enables efficient handling and preprocessing of structured data, including CSV, Excel, SQL databases, and other formats.
- pandas allows data cleaning, filtering, sorting, merging, and aggregating, making it suitable for data wrangling tasks.

3.scikit-learn:

• scikit-learn offers various preprocessing methods, feature selection techniques, Scaling techniques.

 scikit-learn integrates seamlessly with NumPy and pandas, allowing easy data preparation and model training.

4. Matplotlib:

- It provides a wide range of plot types, including line plots, scatter plots, bar plots, histograms, heatmaps, and more.
- Matplotlib is highly compatible with NumPy and pandas, enabling visualization of data from these libraries.

5.Seaborn:

- Seaborn is a statistical data visualization library that works closely with Matplotlib and enhances its capabilities.
- Seaborn includes functions for creating visually appealing plots like distribution plots, box plots, violin

Importing libraries

In [1]:

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

Read Dataset

In [2]:

```
# To read the .csv file
nf = pd.read_csv('C:\\Users\\91827\\Downloads\\Netflix userbase.csv')
nf
```

Out[2]:

	User ID	Subscription Type	Monthly Revenue	Join Date	Last Payment Date	Country	Age	Gender	Device	Dι
0	1	Basic	10.0	15- 01- 2022	10-06- 2023	United States	28.0	Male	Smartphone	1
1	2	Premium	15.0	05- 09- 2021	22-06- 2023	Canada	35.0	Female	Tablet	1
2	3	Standard	12.0	28- 02- 2023	27-06- 2023	United Kingdom	42.0	Male	NaN	1
3	4	Standard	NaN	10- 07- 2022	26-06- 2023	Australia	51.0	Female	Laptop	1
4	5	Basic	10.0	01- 05- 2023	28-06- 2023	Germany	33.0	Male	Smartphone	1
2495	2496	Premium	14.0	25- 07- 2022	12-07- 2023	Spain	28.0	Female	Smart TV	1
2496	2497	Basic	15.0	04- 08- 2022	14-07- 2023	Spain	33.0	Female	Smart TV	1
2497	2498	Standard	12.0	09- 08- 2022	15-07- 2023	United States	38.0	Male	Laptop	1
2498	2499	Standard	13.0	12- 08- 2022	12-07- 2023	Canada	48.0	Female	Tablet	1
2499	2500	Basic	15.0	13- 08- 2022	12-07- 2023	United States	35.0	Female	Smart TV	1

2500 rows × 10 columns

◀

Applying Basic Functions

First 2 values

```
In [63]:
```

```
nf.head(2)
```

Out[63]:

	User ID	Subscription Type	Monthly Revenue	Join Date	Last Payment Date	Country	Age	Gender	Device	Pla Duratio
0	1	Basic	10.0	15- 01- 2022	10-06- 2023	United States	28.0	Male	Smartphone	1 Mor
1	2	Premium	15.0	05- 09- 2021	22-06- 2023	Canada	35.0	Female	Tablet	1 Mor
4										•

Shape of the dataset

```
In [55]:
```

```
# checking the no. of rows and columns
nf.shape
```

Out[55]:

(2500, 10)

Print the name of columns

```
In [51]:
```

```
# List down all the column names
nf.columns
```

Out[51]:

Check for NULL Values

In [12]:

```
nf.isnull().sum()
```

Out[12]:

User ID 0 Subscription Type 1 Monthly Revenue 1 Join Date 0 Last Payment Date 0 Country 2 1 Age Gender Device 1 Plan Duration dtype: int64

Check unique values

In [49]:

```
nf.nunique()
```

Out[49]:

2493
3
6
297
26
10
26
2
4
1

Check for Duplicate values

In [13]:

```
nf.duplicated().sum()
```

Out[13]:

Make a copy of the dataset

```
In [47]:
```

```
cd = nf.copy()
```

In [46]:

cd.shape

Out[46]:

(2500, 10)

Drop NULL values

In [57]:

```
nf=nf.dropna()
nf.shape
```

Out[57]:

(2493, 10)

Last 2 values

In [58]:

```
nf.tail(2)
```

Out[58]:

		User ID		Monthly Revenue	Join Date	Last Payment Date	Country	Age	Gender	Device	Pla Duratio
2	498	2499	Standard	13.0	12- 08- 2022	12-07- 2023	Canada	48.0	Female	Tablet	1 Mont
24	499	2500	Basic	15.0	13- 08- 2022	12-07- 2023	United States	35.0	Female	Smart TV	1 Mont
4											•

checking datatype

```
In [59]:
nf.info()
<class 'pandas.core.frame.DataFrame'>
Int64Index: 2493 entries, 0 to 2499
Data columns (total 10 columns):
#
    Column
                       Non-Null Count Dtype
    _____
                       -----
    User ID
                                       int64
0
                       2493 non-null
    Subscription Type 2493 non-null
                                       object
 1
 2
    Monthly Revenue
                       2493 non-null
                                       float64
    Join Date
 3
                       2493 non-null
                                       object
 4
    Last Payment Date 2493 non-null
                                       object
 5
    Country
                       2493 non-null
                                       object
 6
                                       float64
    Age
                       2493 non-null
 7
    Gender
                       2493 non-null
                                       object
 8
    Device
                       2493 non-null
                                       object
 9
    Plan Duration
                     2493 non-null
                                       object
dtypes: float64(2), int64(1), object(7)
memory usage: 214.2+ KB
In [14]:
# To count the number of females and males?
nf['Gender'].value_counts()
Out[14]:
Basic
           998
Standard
           768
Premium
           733
Name: Subscription Type, dtype: int64
In [ ]:
# To print only tablet device users
nf[nf["Device"] == "Tablet"]
print(nf[nf["Device"] == "Tablet"])
```

Data cleaning:

- 1.isnull(): To check the data having null values or not.
- 2.notnull(): To get the not null values.
- 3.dropna(): To remove the null values.
- 4.fillna(): To fill the null values with meaningfull data.
- · 5.replace(): To replace the data
- . 6.sort values(): To sort the data
- NaN : Not a number.

In [65]:

1.To count the number of null values nf.isnull().sum()

Out[65]:

User ID 0 Subscription Type Monthly Revenue 1 Join Date 0 Last Payment Date 0 Country 2 Age 1 Gender 0 Device 1 Plan Duration 1 dtype: int64

In [67]:

1.2 To print the only null values records
nf[nf.isnull().any(axis=1)]

Out[67]:

	User ID	Subscription Type	Monthly Revenue	Join Date	Last Payment Date	Country	Age	Gender	Device	F Dura
2	3	Standard	12.0	28- 02- 2023	27-06- 2023	United Kingdom	42.0	Male	NaN	1 Mc
3	4	Standard	NaN	10- 07- 2022	26-06- 2023	Australia	51.0	Female	Laptop	1 Mc
5	6	Premium	15.0	18- 03- 2022	27-06- 2023	France	NaN	Female	Smart TV	1 Mc
7	8	NaN	10.0	02- 04- 2023	24-06- 2023	Mexico	39.0	Female	Laptop	1 Ma
9	10	Premium	15.0	07- 01- 2023	22-06- 2023	NaN	44.0	Female	Smart TV	1 Ma
12	13	Standard	12.0	30- 11- 2021	27-06- 2023	United Kingdom	48.0	Female	Laptop	1
13	14	Basic	10.0	01- 08- 2022	26-06- 2023	NaN	27.0	Male	Smartphone	1 Mc
4										•

In [7]:

```
# 2.To count the not null records?
nf.notnull().sum()
```

Out[7]:

User ID	2500
Subscription Type	2499
Monthly Revenue	2499
Join Date	2500
Last Payment Date	2500
Country	2498
Age	2499
Gender	2500
Device	2499
Plan Duration	2499

dtype: int64

In [9]:

2.2 To print the not null records nf[nf.notnull().all(axis=1)]

Out[9]:

	User ID	Subscription Type	Monthly Revenue	Join Date	Last Payment Date	Country	Age	Gender	Device	Dι
0	1	Basic	10.0	15- 01- 2022	10-06- 2023	United States	28.0	Male	Smartphone	1
1	2	Premium	15.0	05- 09- 2021	22-06- 2023	Canada	35.0	Female	Tablet	1
4	5	Basic	10.0	01- 05- 2023	28-06- 2023	Germany	33.0	Male	Smartphone	1
6	7	Standard	12.0	09- 12- 2021	25-06- 2023	Brazil	46.0	Male	Tablet	1
8	9	Standard	12.0	20- 10- 2022	23-06- 2023	Spain	37.0	Male	Smartphone	1
2495	2496	Premium	14.0	25- 07- 2022	12-07- 2023	Spain	28.0	Female	Smart TV	1
2496	2497	Basic	15.0	04- 08- 2022	14-07- 2023	Spain	33.0	Female	Smart TV	1
2497	2498	Standard	12.0	09- 08- 2022	15-07- 2023	United States	38.0	Male	Laptop	1
2498	2499	Standard	13.0	12- 08- 2022	12-07- 2023	Canada	48.0	Female	Tablet	1
2499	2500	Basic	15.0	13- 08- 2022	12-07- 2023	United States	35.0	Female	Smart TV	1

2493 rows × 10 columns

In [89]:

3.dropna(): To remove the null values..
nf.dropna()

Out[89]:

	User ID	Subscription Type	Monthly Revenue	Join Date	Last Payment Date	Country	Age	Gender	Device	Dι
0	1	Basic	10.0	15- 01- 2022	10-06- 2023	United States	28.0	Male	Smartphone	1
1	2	Premium	15.0	05- 09- 2021	22-06- 2023	Canada	35.0	Female	Tablet	1
4	5	Basic	10.0	01- 05- 2023	28-06- 2023	Germany	33.0	Male	Smartphone	1
6	7	Standard	12.0	09- 12- 2021	25-06- 2023	Brazil	46.0	Male	Tablet	1
8	9	Standard	12.0	20- 10- 2022	23-06- 2023	Spain	37.0	Male	Smartphone	1
2495	2496	Premium	14.0	25- 07- 2022	12-07- 2023	Spain	28.0	Female	Smart TV	1
2496	2497	Basic	15.0	04- 08- 2022	14-07- 2023	Spain	33.0	Female	Smart TV	1
2497	2498	Standard	12.0	09- 08- 2022	15-07- 2023	United States	38.0	Male	Laptop	1
2498	2499	Standard	13.0	12- 08- 2022	12-07- 2023	Canada	48.0	Female	Tablet	1
2499	2500	Basic	15.0	13- 08- 2022	12-07- 2023	United States	35.0	Female	Smart TV	1

2493 rows × 10 columns

In [14]:

4.fillna(): To fill the null values with meaningfull data.
nf.fillna('missing value',inplace=True)
nf

Out[14]:

	User ID	Subscription Type	Monthly Revenue	Join Date	Last Payment Date	Country	Age	Gender	Device	Dι
0	1	Basic	10.0	15- 01- 2022	10-06- 2023	United States	28.0	Male	Smartphone	1
1	2	Premium	15.0	05- 09- 2021	22-06- 2023	Canada	35.0	Female	Tablet	1
2	3	Standard	12.0	28- 02- 2023	27-06- 2023	United Kingdom	42.0	Male	missing value	1
3	4	Standard	missing value	10- 07- 2022	26-06- 2023	Australia	51.0	Female	Laptop	1
4	5	Basic	10.0	01- 05- 2023	28-06- 2023	Germany	33.0	Male	Smartphone	1
2495	2496	Premium	14.0	25- 07- 2022	12-07- 2023	Spain	28.0	Female	Smart TV	1
2496	2497	Basic	15.0	04- 08- 2022	14-07- 2023	Spain	33.0	Female	Smart TV	1
2497	2498	Standard	12.0	09- 08- 2022	15-07- 2023	United States	38.0	Male	Laptop	1
2498	2499	Standard	13.0	12- 08- 2022	12-07- 2023	Canada	48.0	Female	Tablet	1
2499	2500	Basic	15.0	13- 08- 2022	12-07- 2023	United States	35.0	Female	Smart TV	1

2500 rows × 10 columns

◀

In [125]:

```
dc = pd.read_csv('Netflix userbase.csv')
dc
```

Out[125]:

	User ID	Subscription Type	Monthly Revenue	Join Date	Last Payment Date	Country	Age	Gender	Device	Dι
0	1	Basic	10.0	15- 01- 2022	10-06- 2023	United States	28.0	Male	Smartphone	1
1	2	Premium	15.0	05- 09- 2021	22-06- 2023	Canada	35.0	Female	Tablet	1
2	3	Standard	12.0	28- 02- 2023	27-06- 2023	United Kingdom	42.0	Male	NaN	1
3	4	Standard	NaN	10- 07- 2022	26-06- 2023	Australia	51.0	Female	Laptop	
4	5	Basic	10.0	01- 05- 2023	28-06- 2023	Germany	33.0	Male	Smartphone	1
2495	2496	Premium	14.0	25- 07- 2022	12-07- 2023	Spain	28.0	Female	Smart TV	1
2496	2497	Basic	15.0	04- 08- 2022	14-07- 2023	Spain	33.0	Female	Smart TV	1
2497	2498	Standard	12.0	09- 08- 2022	15-07- 2023	United States	38.0	Male	Laptop	1
2498	2499	Standard	13.0	12- 08- 2022	12-07- 2023	Canada	48.0	Female	Tablet	1
2499	2500	Basic	15.0	13- 08- 2022	12-07- 2023	United States	35.0	Female	Smart TV	1

2500 rows × 10 columns

In [126]:

4.2 To fill the specific column name null values
dc['Device'].fillna('Computer',inplace=True)
dc

Out[126]:

	User ID	Subscription Type	Monthly Revenue	Join Date	Last Payment Date	Country	Age	Gender	Device	Dι
0	1	Basic	10.0	15- 01- 2022	10-06- 2023	United States	28.0	Male	Smartphone	1
1	2	Premium	15.0	05- 09- 2021	22-06- 2023	Canada	35.0	Female	Tablet	1
2	3	Standard	12.0	28- 02- 2023	27-06- 2023	United Kingdom	42.0	Male	Computer	1
3	4	Standard	NaN	10- 07- 2022	26-06- 2023	Australia	51.0	Female	Laptop	
4	5	Basic	10.0	01- 05- 2023	28-06- 2023	Germany	33.0	Male	Smartphone	1
2495	2496	Premium	14.0	25- 07- 2022	12-07- 2023	Spain	28.0	Female	Smart TV	1
2496	2497	Basic	15.0	04- 08- 2022	14-07- 2023	Spain	33.0	Female	Smart TV	1
2497	2498	Standard	12.0	09- 08- 2022	15-07- 2023	United States	38.0	Male	Laptop	1
2498	2499	Standard	13.0	12- 08- 2022	12-07- 2023	Canada	48.0	Female	Tablet	1
2499	2500	Basic	15.0	13- 08- 2022	12-07- 2023	United States	35.0	Female	Smart TV	1

2500 rows × 10 columns

◀ |

In [127]:

```
# 5. replace(): To replace the data
dc['Gender'].replace(to_replace='Male',value='M',inplace=True)
dc
```

Out[127]:

	User ID	Subscription Type	Monthly Revenue	Join Date	Last Payment Date	Country	Age	Gender	Device	Dι
0	1	Basic	10.0	15- 01- 2022	10-06- 2023	United States	28.0	М	Smartphone	1
1	2	Premium	15.0	05- 09- 2021	22-06- 2023	Canada	35.0	Female	Tablet	1
2	3	Standard	12.0	28- 02- 2023	27-06- 2023	United Kingdom	42.0	М	Computer	1
3	4	Standard	NaN	10- 07- 2022	26-06- 2023	Australia	51.0	Female	Laptop	
4	5	Basic	10.0	01- 05- 2023	28-06- 2023	Germany	33.0	М	Smartphone	1
2495	2496	Premium	14.0	25- 07- 2022	12-07- 2023	Spain	28.0	Female	Smart TV	1
2496	2497	Basic	15.0	04- 08- 2022	14-07- 2023	Spain	33.0	Female	Smart TV	1
2497	2498	Standard	12.0	09- 08- 2022	15-07- 2023	United States	38.0	М	Laptop	1
2498	2499	Standard	13.0	12- 08- 2022	12-07- 2023	Canada	48.0	Female	Tablet	1
2499	2500	Basic	15.0	13- 08- 2022	12-07- 2023	United States	35.0	Female	Smart TV	1

2500 rows × 10 columns

◀ |

In [134]:

```
# 6.Sort_values(): Sorting by column 'Country'
nf.sort_values(by=['Country'])
```

Out[134]:

	User ID	Subscription Type	Monthly Revenue	Join Date	Last Payment Date	Country	Age	Gender	Device	Du
273	274	Premium	10.0	01- 08- 2022	24-06- 2023	Australia	27.0	Female	Tablet	1
988	989	Basic	12.0	08- 11- 2022	01-07- 2023	Australia	34.0	Male	Smartphone	1
2143	2144	Standard	14.0	15- 07- 2022	10-07- 2023	Australia	47.0	Female	Smartphone	1
1258	1259	Premium	13.0	11- 10- 2022	03-07- 2023	Australia	41.0	Female	Laptop	1
413	414	Premium	14.0	21- 08- 2022	27-06- 2023	Australia	47.0	Female	Laptop	1
1524	1525	Basic	14.0	24- 08- 2022	04-07- 2023	United States	50.0	Female	Tablet	1
1479	1480	Basic	13.0	03- 11- 2022	07-07- 2023	United States	49.0	Female	Laptop	1
2499	2500	Basic	15.0	13- 08- 2022	12-07- 2023	United States	35.0	Female	Smart TV	1
13	14	Basic	10.0	01- 08- 2022	26-06- 2023	missing value	27.0	Male	Smartphone	1
9	10	Premium	15.0	07- 01- 2023	22-06- 2023	missing value	44.0	Female	Smart TV	1

2500 rows × 10 columns

Data Filtering

In [150]:

```
df= pd.read_csv('Netflix userbase.csv')
df
```

Out[150]:

	User ID	Subscription Type	Monthly Revenue	Join Date	Last Payment Date	Country	Age	Gender	Device	Dι
() 1	Basic	10.0	15- 01- 2022	10-06- 2023	United States	28.0	Male	Smartphone	1
,	1 2	Premium	15.0	05- 09- 2021	22-06- 2023	Canada	35.0	Female	Tablet	1
:	2 3	Standard	12.0	28- 02- 2023	27-06- 2023	United Kingdom	42.0	Male	NaN	1
;	3 4	Standard	NaN	10- 07- 2022	26-06- 2023	Australia	51.0	Female	Laptop	
•	4 5	Basic	10.0	01- 05- 2023	28-06- 2023	Germany	33.0	Male	Smartphone	1
249	5 2496	Premium	14.0	25- 07- 2022	12-07- 2023	Spain	28.0	Female	Smart TV	1
249	3 2497	Basic	15.0	04- 08- 2022	14-07- 2023	Spain	33.0	Female	Smart TV	1
249	7 2498	Standard	12.0	09- 08- 2022	15-07- 2023	United States	38.0	Male	Laptop	1
249	3 2499	Standard	13.0	12- 08- 2022	12-07- 2023	Canada	48.0	Female	Tablet	1
249	9 2500	Basic	15.0	13- 08- 2022	12-07- 2023	United States	35.0	Female	Smart TV	1

2500 rows × 10 columns

In [158]:

1.Select particular columns using column values in a dataframe method
df[["Monthly Revenue","Plan Duration"]]

Out[158]:

	Monthly Revenue	Plan Duration
0	10.0	1 Month
1	15.0	1 Month
2	12.0	1 Month
3	NaN	NaN
4	10.0	1 Month
2495	14.0	1 Month
2496	15.0	1 Month
2497	12.0	1 Month
2498	13.0	1 Month
2499	15.0	1 Month

2500 rows × 2 columns

In [159]:

```
# 2.Slicing using index to filter rows
df[1:3]
```

Out[159]:

	User ID	•	Monthly Revenue	Join Date	Payment Date	Country	Age	Gender	Device	Plan Duration
1	2	Premium	15.0	05- 09- 2021	22-06- 2023	Canada	35.0	Female	Tablet	1 Month
2	3	Standard	12.0	28- 02- 2023	27-06- 2023	United Kingdom	42.0	Male	NaN	1 Month

In [160]:

```
# 3.Filter rows and columns using iloc() function
# iloc() method selects rows and columns based on the index/position values.
df.iloc[ 0:3 , 0:2]
```

Out[160]:

	User ID	Subscription Type
0	1	Basic
1	2	Premium
2	3	Standard

In [161]:

```
# 4.Filter rows and columns using loc() method(slicing using labels)
df.loc[ [3,4] , ['Country', 'Device'] ]
```

Out[161]:

	Country	Device
3	Australia	Laptop

4 Germany Smartphone

In [162]:

```
# 5.Filter Devices by Tablet
df[nf['Device']=='Tablet']
```

Out[162]:

	User ID	Subscription Type	Monthly Revenue	Join Date	Last Payment Date	Country	Age	Gender	Device	Pla Duratic
1	2	Premium	15.0	05- 09- 2021	22-06- 2023	Canada	35.0	Female	Tablet	1 Mon
6	7	Standard	12.0	09- 12- 2021	25-06- 2023	Brazil	46.0	Male	Tablet	1 Mon
11	12	Premium	15.0	23- 03- 2023	28-06- 2023	Canada	45.0	Male	Tablet	1 Mon
15	16	Premium	15.0	07- 04- 2022	27-06- 2023	France	36.0	NaN	Tablet	1 Mon
19	20	Basic	10.0	27- 05- 2023	22-06- 2023	Italy	41.0	Male	Tablet	1 Mon
2474	2475	Basic	12.0	05- 09- 2022	11-07- 2023	Germany	37.0	Female	Tablet	1 Mon
2475	2476	Basic	13.0	31- 08- 2022	15-07- 2023	France	28.0	Male	Tablet	1 Mon
2477	2478	Standard	10.0	16- 08- 2022	13-07- 2023	Mexico	47.0	Male	Tablet	1 Mon
2484	2485	Premium	14.0	25- 07- 2022	12-07- 2023	United States	47.0	Female	Tablet	1 Mon
2498	2499	Standard	13.0	12- 08- 2022	12-07- 2023	Canada	48.0	Female	Tablet	1 Mon

633 rows × 10 columns

```
In [163]:
```

```
# 6.Filter single item from the dataset
df.iat[1,1]
```

Out[163]:

'Premium'

In [164]:

```
df.at[1, 'Country']
```

Out[164]:

'Canada'

In [165]:

7.Using regular expression to extract all columns which has letter 'a' or 'A' in its r = (aA)')

Out[165]:

	Join Date	Last Payment Date	Age	Plan Duration
0	15-01-2022	10-06-2023	28.0	1 Month
1	05-09-2021	22-06-2023	35.0	1 Month
2	28-02-2023	27-06-2023	42.0	1 Month
3	10-07-2022	26-06-2023	51.0	NaN
4	01-05-2023	28-06-2023	33.0	1 Month
2495	25-07-2022	12-07-2023	28.0	1 Month
2496	04-08-2022	14-07-2023	33.0	1 Month
2497	09-08-2022	15-07-2023	38.0	1 Month
2498	12-08-2022	12-07-2023	48.0	1 Month
2499	13-08-2022	12-07-2023	35.0	1 Month

2500 rows × 4 columns

In [166]:

```
# 8.count various Subscription Types
df['Subscription Type'].value_counts()
```

Out[166]:

Basic 998 Standard 768 Premium 733

Name: Subscription Type, dtype: int64

In [169]:

9.TO print the least Monthly Revenue records from a df?
df[df['Monthly Revenue']==df['Monthly Revenue'].min()]

Out[169]:

	User ID	Subscription Type	Monthly Revenue	Join Date	Last Payment Date	Country	Age	Gender	Device	Dι
0	1	Basic	10.0	15- 01- 2022	10-06- 2023	United States	28.0	Male	Smartphone	1
4	5	Basic	10.0	01- 05- 2023	28-06- 2023	Germany	33.0	Male	Smartphone	1
7	8	NaN	10.0	02- 04- 2023	24-06- 2023	Mexico	39.0	Female	Laptop	1
10	11	Basic	10.0	16- 05- 2022	22-06- 2023	United States	31.0	Female	Smartphone	1
13	14	Basic	10.0	01- 08- 2022	26-06- 2023	NaN	27.0	Male	Smartphone	1
2473	2474	Premium	10.0	06- 09- 2022	12-07- 2023	Australia	49.0	Male	Smart TV	1
2477	2478	Standard	10.0	16- 08- 2022	13-07- 2023	Mexico	47.0	Male	Tablet	1
2480	2481	Premium	10.0	05- 08- 2022	12-07- 2023	Spain	33.0	Female	Laptop	1
2485	2486	Basic	10.0	25- 07- 2022	14-07- 2023	United States	40.0	Female	Smartphone	1
2489	2490	Standard	10.0	17- 07- 2022	15-07- 2023	Germany	35.0	Male	Smart TV	1

409 rows × 10 columns

In [187]:

10. To print the Country name starting with 'F' from dataset?
nf[nf['Country'].str.startswith('F')] # taking nf dataset becoz of nan values in df

Out[187]:

	User ID	Subscription Type	Monthly Revenue	Join Date	Last Payment Date	Country	Age Gender		Device
5	6	Premium	15.0	18- 03- 2022	27-06- 2023	France	missing value	Female	Smart TV
15	16	Premium	15.0	07- 04- 2022	27-06- 2023	France	36.0	Male	Tablet
25	26	Premium	15.0	12- 01- 2022	27-06- 2023	France	29.0	Male	Smartphone
35	36	Premium	15.0	01- 03- 2022	27-06- 2023	France	35.0	Male	Tablet
45	46	Premium	15.0	23- 02- 2022	27-06- 2023	France	34.0	Male	Smartphone
2430	2431	Premium	12.0	15- 08- 2022	14-07- 2023	France	29.0	Female	Smart TV
2445	2446	Premium	11.0	18- 10- 2022	12-07- 2023	France	46.0	Male	Tablet
2460	2461	Premium	13.0	13- 11- 2022	12-07- 2023	France	41.0	Male	Laptop
2475	2476	Basic	13.0	31- 08- 2022	15-07- 2023	France	28.0	Male	Tablet
2490	2491	Premium	13.0	18- 07- 2022	11-07- 2023	France	41.0	Female	Smartphone

183 rows × 10 columns

Grouping

In [10]:

```
gp= pd.read_csv('Netflix userbase.csv')
gp
```

Out[10]:

	User ID	Subscription Type	Monthly Revenue	Join Date	Last Payment Date	Country	Age	Gender	Device	Dι
0	1	Basic	10.0	15- 01- 2022	10-06- 2023	United States	28.0	Male	Smartphone	1
1	2	Premium	15.0	05- 09- 2021	22-06- 2023	Canada	35.0	Female	Tablet	1
2	3	Standard	12.0	28- 02- 2023	27-06- 2023	United Kingdom	42.0	Male	NaN	1
3	4	Standard	NaN	10- 07- 2022	26-06- 2023	Australia	51.0	Female	Laptop	
4	5	Basic	10.0	01- 05- 2023	28-06- 2023	Germany	33.0	Male	Smartphone	1
2495	2496	Premium	14.0	25- 07- 2022	12-07- 2023	Spain	28.0	Female	Smart TV	1
2496	2497	Basic	15.0	04- 08- 2022	14-07- 2023	Spain	33.0	Female	Smart TV	1
2497	2498	Standard	12.0	09- 08- 2022	15-07- 2023	United States	38.0	Male	Laptop	1
2498	2499	Standard	13.0	12- 08- 2022	12-07- 2023	Canada	48.0	Female	Tablet	1
2499	2500	Basic	15.0	13- 08- 2022	12-07- 2023	United States	35.0	Female	Smart TV	1

2500 rows × 10 columns

In [11]:

```
# 1.applying groupby() function to group the data on Device value.
gp = nf.groupby('Device')
gp.first()
```

Out[11]:

	User ID	Subscription Type	Monthly Revenue	Join Date	Last Payment Date	Country	Age	Gender	Plan Duration
Device									
Laptop	4	Standard	10.0	10- 07- 2022	26-06- 2023	Australia	51.0	Female	1 Month
Smart TV	6	Premium	15.0	18- 03- 2022	27-06- 2023	France	44.0	Female	1 Month
Smartphone	1	Basic	10.0	15- 01- 2022	10-06- 2023	United States	28.0	Male	1 Month
Tablet	2	Premium	15.0	05- 09- 2021	22-06- 2023	Canada	35.0	Female	1 Month
4									•

In [12]:

2.We use the function $get_group()$ to find the entries contained in any of the groups. $gp.get_group('Smartphone')$

Out[12]:

	User ID	Subscription Type	Monthly Revenue	Join Date	Last Payment Date	Country	Age	Gender	Device	Dι
0	1	Basic	10.0	15- 01- 2022	10-06- 2023	United States	28.0	Male	Smartphone	1
4	5	Basic	10.0	01- 05- 2023	28-06- 2023	Germany	33.0	Male	Smartphone	1
8	9	Standard	12.0	20- 10- 2022	23-06- 2023	Spain	37.0	Male	Smartphone	1
10	11	Basic	10.0	16- 05- 2022	22-06- 2023	United States	31.0	Female	Smartphone	1
13	14	Basic	10.0	01- 08- 2022	26-06- 2023	NaN	27.0	Male	Smartphone	1
2470	2471	Basic	10.0	17- 10- 2022	11-07- 2023	United States	46.0	Male	Smartphone	1
2485	2486	Basic	10.0	25- 07- 2022	14-07- 2023	United States	40.0	Female	Smartphone	1
2487	2488	Standard	11.0	18- 07- 2022	13-07- 2023	United Kingdom	29.0	Female	Smartphone	1
2488	2489	Basic	11.0	17- 07- 2022	12-07- 2023	Australia	48.0	Female	Smartphone	1
2490	2491	Premium	13.0	18- 07- 2022	11-07- 2023	France	41.0	Female	Smartphone	1

621 rows × 10 columns

Sorting

In [19]:

```
sg= pd.read_csv('Netflix userbase.csv')
sg
```

Out[19]:

0
1
2
3
4
2495
2496
2497
2498
2497

2500 rows × 10 columns

•

In [24]:

```
# 1.Sorting by column "Age"
sg.sort_values(by=['Age'], ascending=True)
```

Out[24]:

	User ID	Subscription Type	Monthly Revenue	Join Date	Last Payment Date	Country	Age	Gender	Device	Du
20	21	Premium	15.0	10- 06- 2023	22-06- 2023	United States	26.0	Female	Laptop	1
1297	1298	Standard	14.0	26- 06- 2022	03-07- 2023	United States	27.0	Male	Laptop	1
1045	1046	Basic	13.0	06- 10- 2022	03-07- 2023	United States	27.0	Male	Laptop	1
1077	1078	Standard	14.0	17- 07- 2022	02-07- 2023	United Kingdom	27.0	Male	Laptop	1
1081	1082	Basic	13.0	28- 06- 2022	02-07- 2023	Brazil	27.0	Male	Tablet	1
849	850	Basic	13.0	13- 11- 2022	30-06- 2023	United States	51.0	Female	Smart TV	1
1110	1111	Premium	13.0	10- 09- 2022	01-07- 2023	France	51.0	Male	Smart TV	1
1758	1759	Standard	10.0	06- 11- 2022	07-07- 2023	Spain	51.0	Male	Smart TV	1
1119	1120	Basic	13.0	10- 10- 2022	04-07- 2023	United States	51.0	Female	Smartphone	1
5	6	Premium	15.0	18- 03- 2022	27-06- 2023	France	NaN	Female	Smart TV	1

2500 rows × 10 columns

In [25]:

2.Sorting Pandas Data frame by putting missing values first
sg.sort_values(by=['Age'], na_position='first')

Out[25]:

	User ID	Subscription Type	Monthly Revenue	Join Date	Last Payment Date	Country	Age	Gender	Device	Du
5	6	Premium	15.0	18- 03- 2022	27-06- 2023	France	NaN	Female	Smart TV	1
20	21	Premium	15.0	10- 06- 2023	22-06- 2023	United States	26.0	Female	Laptop	1
1297	1298	Standard	14.0	26- 06- 2022	03-07- 2023	United States	27.0	Male	Laptop	1
1045	1046	Basic	13.0	06- 10- 2022	03-07- 2023	United States	27.0	Male	Laptop	1
1077	1078	Standard	14.0	17- 07- 2022	02-07- 2023	United Kingdom	27.0	Male	Laptop	1
215	216	Premium	14.0	08- 10- 2022	24-06- 2023	France	51.0	Female	Smart TV	1
849	850	Basic	13.0	13- 11- 2022	30-06- 2023	United States	51.0	Female	Smart TV	1
1110	1111	Premium	13.0	10- 09- 2022	01-07- 2023	France	51.0	Male	Smart TV	1
1758	1759	Standard	10.0	06- 11- 2022	07-07- 2023	Spain	51.0	Male	Smart TV	1
1119	1120	Basic	13.0	10- 10- 2022	04-07- 2023	United States	51.0	Female	Smartphone	1

2500 rows × 10 columns

In [26]:

```
# 3.Sorting Data frames by multiple columns
# Sorting by columns "Country" and then "Director"
sg.sort_values(by=['Monthly Revenue', 'Age'])
```

Out[26]:

	User ID	Subscription Type	Monthly Revenue	Join Date	Last Payment Date	Country	Age	Gender	Device	Dι
13	14	Basic	10.0	01- 08- 2022	26-06- 2023	NaN	27.0	Male	Smartphone	1
70	71	Basic	10.0	26- 01- 2022	20-06- 2023	United States	27.0	Male	Smartphone	1
259	260	Basic	10.0	17- 07- 2022	26-06- 2023	Italy	27.0	Female	Smartphone	1
273	274	Premium	10.0	01- 08- 2022	24-06- 2023	Australia	27.0	Female	Tablet	1
298	299	Premium	10.0	30- 10- 2022	27-06- 2023	Spain	27.0	Male	Smartphone	1
2039	2040	Standard	15.0	08- 11- 2022	08-07- 2023	Germany	51.0	Male	Laptop	1
2227	2228	Standard	15.0	16- 09- 2022	10-07- 2023	United States	51.0	Male	Tablet	1
2425	2426	Premium	15.0	23- 07- 2022	12-07- 2023	United States	51.0	Male	Smartphone	1
5	6	Premium	15.0	18- 03- 2022	27-06- 2023	France	NaN	Female	Smart TV	1
3	4	Standard	NaN	10- 07- 2022	26-06- 2023	Australia	51.0	Female	Laptop	

2500 rows × 10 columns

Aggregations

• To work on numerical data

In [6]:

```
ag= pd.read_csv('Netflix userbase.csv')
ag
```

Out[6]:

Dι	Device	Gender	Age	Country	Last Payment Date	Join Date	Monthly Revenue	Subscription Type	User ID	
1	Smartphone	Male	28.0	United States	10-06- 2023	15- 01- 2022	10.0	Basic	1	0
1	Tablet	Female	35.0	Canada	22-06- 2023	05- 09- 2021	15.0	Premium	2	1
1	NaN	Male	42.0	United Kingdom	27-06- 2023	28- 02- 2023	12.0	Standard	3	2
	Laptop	Female	51.0	Australia	26-06- 2023	10- 07- 2022	NaN	Standard	4	3
1	Smartphone	Male	33.0	Germany	28-06- 2023	01- 05- 2023	10.0	Basic	5	4
1	Smart TV	Female	28.0	Spain	12-07- 2023	25- 07- 2022	14.0	Premium	2496	2495
1	Smart TV	Female	33.0	Spain	14-07- 2023	04- 08- 2022	15.0	Basic	2497	2496
1	Laptop	Male	38.0	United States	15-07- 2023	09- 08- 2022	12.0	Standard	2498	2497
1	Tablet	Female	48.0	Canada	12-07- 2023	12- 08- 2022	13.0	Standard	2499	2498
1	Smart TV	Female	35.0	United States	12-07- 2023	13- 08- 2022	15.0	Basic	2500	2499

2500 rows × 10 columns

```
In [28]:
```

```
# 1.min
ag.min()
```

C:\Users\91827\AppData\Local\Temp\ipykernel_16172\1429976717.py:1: Future
Warning: Dropping of nuisance columns in DataFrame reductions (with 'nume
ric_only=None') is deprecated; in a future version this will raise TypeEr
ror. Select only valid columns before calling the reduction.
 ag.min()

Out[28]:

User ID 1
Monthly Revenue 10.0
Join Date 01-03-2022
Last Payment Date 01-07-2023
Age 26.0

dtype: object

In [29]:

```
# 2.max
ag.max()
```

C:\Users\91827\AppData\Local\Temp\ipykernel_16172\2130797385.py:1: Future
Warning: Dropping of nuisance columns in DataFrame reductions (with 'nume
ric_only=None') is deprecated; in a future version this will raise TypeEr
ror. Select only valid columns before calling the reduction.
 ag.max()

Out[29]:

User ID 2500
Monthly Revenue 15.0
Join Date 31-10-2022
Last Payment Date 30-06-2023
Age 51.0

dtype: object

In [30]:

```
# 3.mean
ag.mean()
```

C:\Users\91827\AppData\Local\Temp\ipykernel_16172\2231397629.py:1: Future Warning: Dropping of nuisance columns in DataFrame reductions (with 'nume ric_only=None') is deprecated; in a future version this will raise TypeEr ror. Select only valid columns before calling the reduction.

ag.mean()

Out[30]:

User ID 1250.500000 Monthly Revenue 12.508603 Age 38.799520

dtype: float64

```
In [32]:
```

```
# 4.std
ag.std()
```

C:\Users\91827\AppData\Local\Temp\ipykernel_16172\1087247426.py:1: Future
Warning: Dropping of nuisance columns in DataFrame reductions (with 'nume
ric_only=None') is deprecated; in a future version this will raise TypeEr
ror. Select only valid columns before calling the reduction.
 ag.std()

Out[32]:

User ID 721.832160 Monthly Revenue 1.687158 Age 7.170534

dtype: float64

In [33]:

```
# 5.var ag.var()
```

C:\Users\91827\AppData\Local\Temp\ipykernel_16172\3862806996.py:1: Future
Warning: Dropping of nuisance columns in DataFrame reductions (with 'nume
ric_only=None') is deprecated; in a future version this will raise TypeEr
ror. Select only valid columns before calling the reduction.
ag.var()

Out[33]:

User ID 521041.666667 Monthly Revenue 2.846503 Age 51.416557

dtype: float64

In [7]:

```
# Statistics of data set
ag.describe().T.style.background_gradient(cmap='turbo')
```

Out[7]:

	count	mean	std	min	25%	50%	
User ID	2500.000000	1250.500000	721.832160	1.000000	625.750000	1250.500000	1875.2
Monthly Revenue	2499.000000	12.508603	1.687158	10.000000	11.000000	12.000000	14.00
Age	2499.000000	38.799520	7.170534	26.000000	32.000000	39.000000	45.00
4							•

Data Visualization

installing matplotlib

```
In [42]:
```

```
pip install matplotlib
Requirement already satisfied: matplotlib in c:\users\91827\anaconda3\lib
\site-packages (3.5.1)
Requirement already satisfied: cycler>=0.10 in c:\users\91827\anaconda3\1
ib\site-packages (from matplotlib) (0.11.0)
Requirement already satisfied: kiwisolver>=1.0.1 in c:\users\91827\anacon
da3\lib\site-packages (from matplotlib) (1.3.2)
Requirement already satisfied: python-dateutil>=2.7 in c:\users\91827\ana
conda3\lib\site-packages (from matplotlib) (2.8.2)
Requirement already satisfied: pyparsing>=2.2.1 in c:\users\91827\anacond
a3\lib\site-packages (from matplotlib) (3.0.4)
Requirement already satisfied: numpy>=1.17 in c:\users\91827\anaconda3\li
b\site-packages (from matplotlib) (1.21.5)
Requirement already satisfied: pillow>=6.2.0 in c:\users\91827\anaconda3
\lib\site-packages (from matplotlib) (9.0.1)
Requirement already satisfied: packaging>=20.0 in c:\users\91827\anaconda
3\lib\site-packages (from matplotlib) (21.3)
Requirement already satisfied: fonttools>=4.22.0 in c:\users\91827\anacon
da3\lib\site-packages (from matplotlib) (4.25.0)
Requirement already satisfied: six>=1.5 in c:\users\91827\anaconda3\lib\s
ite-packages (from python-dateutil>=2.7->matplotlib) (1.16.0)
```

importing matplotlib

```
In [43]:
```

```
import matplotlib as mpl
```

Note: you may need to restart the kernel to use updated packages.

In [2]:

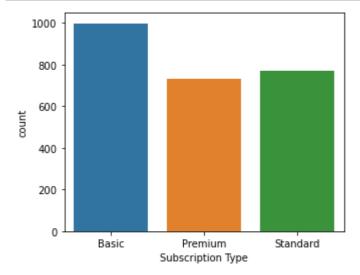
```
import matplotlib.pyplot as plt
```

1.Subscription Type

Which subscription plan are most people using?

In [6]:

```
plt.figure(figsize=(5,4))
sns.countplot(data=nf,x='Subscription Type')
plt.show()
```

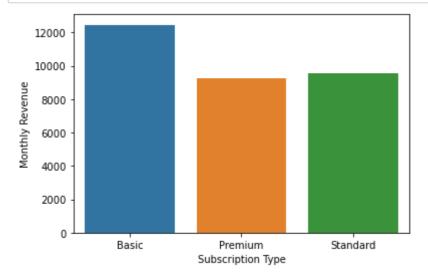


· Basic subscription plan are most people using

Which subscription type generate the most revenue?

In [12]:

```
sm = nf.groupby('Subscription Type').sum().reset_index()
sns.barplot(data=sm,x='Subscription Type', y='Monthly Revenue')
plt.show()
```

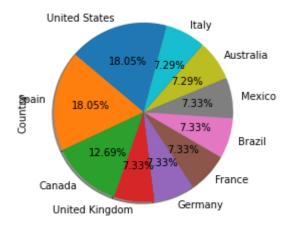


· Basic plan generates maximum revenue

Country

In [8]:

```
nf.Country.value_counts().plot(kind='pie',shadow='True' ,autopct='%1.2f%%', startangle
plt.show()
```

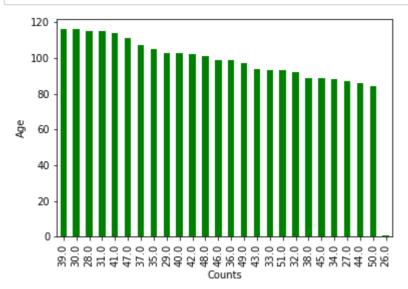


· Spain and United States have more customers

Age

In [12]:

```
nf.Age.value_counts().plot(kind='bar',color='g',figsize=(6,4))
plt.ylabel("Age")
plt.xlabel("Counts")
plt.show()
```

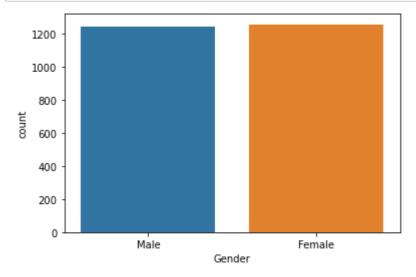


Maximum users are aged 39 followed by 30,28,31,41

Gender

In [22]:

```
sns.countplot(data=nf,x='Gender')
plt.show()
```

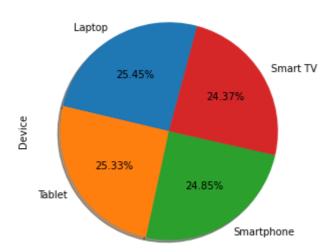


• Male & Female has equal counts

Device

In [13]:

```
nf.Device.value_counts().plot(kind='pie',shadow='True' ,autopct='%1.2f%%', startangle =
plt.show()
```

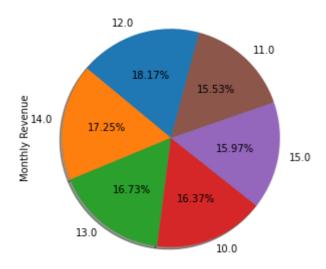


MY SUGGESTION

Monthly Revenue

In [14]:

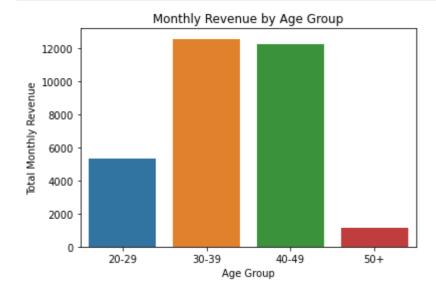
```
nf['Monthly Revenue'].value_counts().plot(kind='pie',shadow='True' ,autopct='%1.2f%%', s
plt.show()
```



Monthly Revenue by Age Group

In [15]:

```
age_bins = [20, 30, 40, 50, 60]
age_labels = ['20-29', '30-39', '40-49', '50+']
nf['age group'] = pd.cut(nf['Age'], bins=age_bins, labels=age_labels)
revenue_by_age = nf.groupby('age group')['Monthly Revenue'].sum().reset_index()
sns.barplot(data=revenue_by_age, x='age group',y='Monthly Revenue')
plt.xlabel('Age Group')
plt.ylabel('Total Monthly Revenue')
plt.title('Monthly Revenue by Age Group')
plt.show()
```

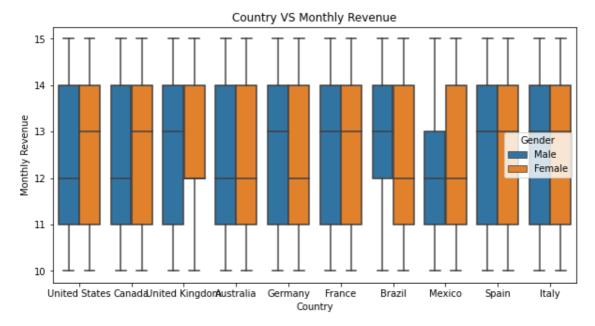


• Most of the users who are purchasing subscription has aged 30 to 49

Country VS Monthly Revenue

In [24]:

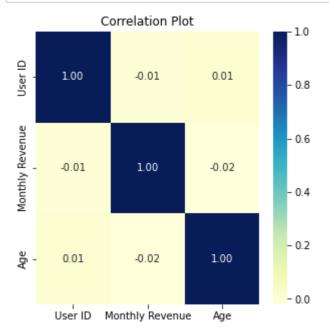
```
plt.figure(figsize=(10,5))
sns.boxplot(data=nf,x='Country', y='Monthly Revenue',hue='Gender',)
plt.title("Country VS Monthly Revenue")
plt.xlabel("Country")
plt.ylabel("Monthly Revenue")
plt.show()
```



Heatmap

In [17]:

```
corr_matrix = nf.corr()
plt.figure(figsize=(5, 5))
sns.heatmap(corr_matrix, annot=True, cmap="YlGnBu", fmt=".2f")
plt.title("Correlation Plot")
plt.show()
```



Final Insights

- Age group 20-29 spend more time on Netflix
- · All countries have similar rate of spending time
- Age group 30-49 doing the most subscription and giving more revenue
- · Basic type subscription has maximum purchasing

Future Scope

 The future scope could involve predictive modeling for churn prevention, content recommendation enhancements, and refining subscription plans based on user segments. Additionally, using AI to personalize user experiences and refining content delivery strategies could also be part of the future scope.

Advantages

- 1. Understanding User Preferences: By analyzing the subscription types chosen by users, Netflix can gain insights into which plans are most popular among their customer base. This knowledge can guide the platform in designing attractive subscription bundles and pricing models.
- 2. Revenue Trends and Growth: Monthly revenue analysis provides a clear view of Netflix's financial performance over time. Identifying revenue trends can help the company track its growth and

- assess the effectiveness of marketing campaigns or promotional strategies.
- 3. Demographic Profiling: Analyzing user age, gender, and country data allows Netflix to better
 understand the diversity of its user base. Tailoring content to different demographics can increase
 engagement and retention.
- 4. Device Usage Insights: By examining the devices through which users access the platform, Netflix
 can optimize its app and website experiences for various platforms, leading to improved user
 satisfaction.
- 5. Plan Duration and User Retention: Calculating the average plan duration offers valuable insights into user retention. This information can help Netflix devise strategies to enhance customer loyalty and reduce churn rates.

Conclusion:

• In conclusion, this analysis and visualization of Netflix user data offer invaluable insights into user behavior, subscription trends, and revenue growth. By understanding user preferences, demographic profiles, and device usage patterns, Netflix can optimize its content offerings and user experience to better cater to its diverse customer base. The study also sheds light on user retention, helping Netflix focus on strategies to increase customer loyalty and reduce subscription cancellations. The data-driven approach presented in this analysis can serve as a foundation for Netflix's future business decisions, ensuring continuous improvement and innovation in the highly competitive streaming industry. As the digital landscape evolves, such analyses will remain crucial in maintaining Netflix's position as a market leader and providing its global audience with exceptional streaming experiences.

References websites

https://www.kaggle.com/ (https://www.kaggle.com/)

· Greeks for greeks

FINAL REVIEW QNS

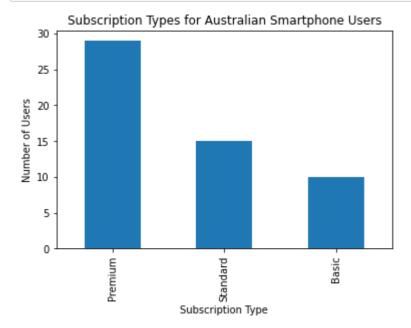
```
In [23]:
```

```
# Filter for Australian users with smartphone subscription
australian_smartphone_users = nf[(nf['Country'] == 'Australia') & (nf['Device'] == 'Smar
len(australian_smartphone_users)
```

Out[23]:

In [14]:

```
subscription_counts = australian_smartphone_users['Subscription Type'].value_counts()
subscription_counts.plot(kind='bar', title='Subscription Types for Australian Smartphone
plt.xlabel('Subscription Type')
plt.ylabel('Number of Users')
plt.show()
```



In [27]:

```
# Group data by country and calculate mean monthly revenue for each country
country_revenue = nf.groupby('Country')['Monthly Revenue'].mean()
country_revenue
```

Out[27]:

Country

Australia 12.425414 Brazil 12.486339 Canada 12.460568 France 12.606557 Germany 12.349727 Italy 12.648352 Mexico 12.224044 12.554324 Spain United Kingdom 12.666667 United States 12.558758

Name: Monthly Revenue, dtype: float64

In [29]:

```
# Find the country with the highest mean monthly revenue
highest_revenue_country = country_revenue.idxmax()
highest_revenue_value = country_revenue.max()
print("Country with the highest mean monthly revenue:", highest_revenue_country)
print("Highest mean monthly revenue:", highest_revenue_value)
```

Country with the highest mean monthly revenue: United Kingdom Highest mean monthly revenue: 12.666666666666

