Customer Behavior Analysis

- 1 import pandas as pd
- 2 import matplotlib.pyplot as plt
- 3 import seaborn as sns

1 df = pd.read_csv('/content/drive/MyDrive/CustomerBehavior.csv')

1 df.head(10)

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Đ	Order Date Time	Order ID	Purchased Product ID	Quantity of SKU in the order	Category ID	Category alias	Brand ID	Price	User ID	Gender	Color	Metal	Main gem
	2018-12-01 11:40:29 UTC	1.920000e+18	1.840000e+18	1	1.810000e+18	jewelry.earring	0.0	561.51	1.520000e+18	NaN	red	gold	diamond
	2018-12-01 1 17:38:31 UTC	1.920000e+18	1.810000e+18	1	1.810000e+18	NaN	NaN	212.14	1.520000e+18	NaN	yellow	gold	NaN
	2018-12-02 13:53:42 UTC	1.930000e+18	1.840000e+18	1	1.810000e+18	jewelry.pendant	1.0	54.66	1.520000e+18	f	white	gold	sapphire
	2018-12-02 3 17:44:02 UTC	1.930000e+18	1.840000e+18	1	1.810000e+18	jewelry.pendant	0.0	88.90	1.520000e+18	f	red	gold	diamond
	2018-12-02 4 21:30:19 UTC	1.930000e+18	1.870000e+18	1	1.810000e+18	jewelry.necklace	0.0	417.67	1.520000e+18	NaN	red	gold	amethyst
	2018-12-02 5 22:09:34 UTC	1.930000e+18	1.840000e+18	1	1.810000e+18	jewelry.earring	1.0	102.27	1.520000e+18	NaN	red	gold	NaN
	2018-12-02 6 22:09:34 UTC	1.930000e+18	1.600000e+18	1	1.810000e+18	jewelry.earring	0.0	150.55	1.520000e+18	NaN	red	gold	diamond

Data Cleaning

1 df.dtypes

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	0
Order Date Time	object
Order ID	float64
Purchased Product ID	float64
Quantity of SKU in the order	int64
Category ID	float64
Category alias	object
Brand ID	float64
Price	float64
User ID	float64
Gender	object
Color	object
Metal	object
Main gem	object

dtype: object

1 #Detect Type

```
2 df['Category ID'] =df['Category ID'].astype(str)
   3 df['Order ID'] =df['Order ID'].astype(str)
  4 df['Purchased Product ID'] =df['Purchased Product ID'].astype(str)
  5 df['Brand ID'] =df['Brand ID'].astype(str)
  6 df['User ID'] =df['User ID'].astype(str)
  7 df['Order Date Time'] =pd.to_datetime(df['Order Date Time'])
  1 df.dtypes
<del>_</del>
                            datetime64[ns, UTC]
         Order Date Time
            Order ID
                                       object
       Purchased Product ID
                                       object
    Quantity of SKU in the order
                                        int64
           Category ID
                                       object
          Category alias
                                       object
            Brand ID
                                       object
             Price
                                       float64
             User ID
                                       object
             Gender
                                       object
             Color
                                       object
             Metal
                                       object
            Main gem
                                       object
    dtype: object
  1 df.isnull().sum()
<del>_</del>
                                0
         Order Date Time
                                0
            Order ID
                                0
       Purchased Product ID
                                0
    Quantity of SKU in the order
                                0
           Category ID
                                0
          Category alias
                             9933
            Brand ID
                                0
             Price
                             5352
             User ID
                                0
             Gender
                            48168
             Color
                             7660
             Metal
                             5462
            Main gem
                            34058
    dtype: int64
  1 df.shape[0]
→ 95911
  1 # Visualize the distribution of the 'Price' column
  2 plt.figure(figsize=(10, 6))
  3 sns.histplot(df['Price'].dropna(), kde=True)
  4 plt.title('Distribution of Price')
  5 plt.xlabel('Price')
  6 plt.ylabel('Frequency')
```

7 plt.show()

```
9 # Calculate descriptive statistics for the 'Price' column
 10 print("\nDescriptive Statistics for Price:")
 11 print(df['Price'].describe())
                                              Distribution of Price
       4000
       3500
       3000
       2500
    Frequency
       2000
       1500
       1000
        500
          0
                         5000
                                    10000
                                                15000
                                                                      25000
                                                                                  30000
                                                           20000
                                                                                             35000
               0
                                                     Price
   Descriptive Statistics for Price:
   count
           90559.000000
             362.215217
   mean
             444,155706
   std
   min
              0.990000
   25%
             145.620000
   50%
             258.770000
             431.370000
   75%
           34448.600000
   Name: Price, dtype: float64
  1 #Data is Skewed ->replace with median
  2 df['Price']=df['Price'].fillna(df['Price'].median())
  3 df['Price'].isnull().sum()
→ np.int64(0)
  1 df['Category alias'] =df['Category alias'].fillna('UnKown')
  2 df['Gender'] =df['Gender'].fillna('UnKown')
```

3 df['Color'] = df['Color'].fillna('UnKown')
4 df['Metal'] = df['Metal'].fillna('UnKown')
5 df['Main gem']= df['Main gem'].fillna('UnKown')

7 df.isnull().sum()

```
<del>_</del>__
                           a
         Order Date Time
                           0
            Order ID
                           0
       Purchased Product ID
                           0
    Quantity of SKU in the order 0
          Category ID
                           0
         Category alias
                           0
            Brand ID
                           0
             Price
                           0
            User ID
            Gender
             Color
             Metal
           Main gem
   dtype: int64
  1 df.duplicated().sum()
→ np.int64(2956)
  1 df.drop_duplicates(inplace=True)
  2 df.duplicated().sum()
→ np.int64(0)
 RFM Analysis & Customer Segmentation
  1 df = df[df['Quantity of SKU in the order'] > 0]
  2 df['Revenue'] = df['Price']*df['Quantity of SKU in the order']
  3 df[['Quantity of SKU in the order', 'Price', 'Revenue']]
₹
          Quantity of SKU in the order Price Revenue
      0
                                  1 561.51
                                            561.51
      1
                                  1 212.14
                                            212.14
      2
                                     54.66
                                             54.66
      3
                                     88.90
                                             88.90
                                  1 417.67
                                            417.67
    95906
                                  1 258.77
                                            258.77
    95907
                                  1 229.99
                                            229.99
    95908
                                            630.00
                                    630.00
    95909
                                     83.42
                                             83.42
    95910
                                     97.12
                                             97.12
   95911 rows × 3 columns
  1 #RFM Analysis
  2 reference_date = df['Order Date Time'].max() + pd.Timedelta(days=1)
  3
  4 rfm = df.groupby('User ID').agg({
         'Order Date Time' : lambda x : reference_date - x.max()
  5
                                                                               #R
                                                                               #F
         ,'Order ID' : 'nunique'
        ,'Revenue' : 'sum'
  7
                                                                               #M
  8 }).reset index()
```

```
9
10 rfm.columns = ['User ID', 'Recency', 'Frequency', 'Monetary']
11
12 rfm['R_Score'] = pd.qcut(rfm['Recency'],5,labels=[5,4,3,2,1])
13 rfm['F_Score'] = pd.qcut(rfm['Frequency'].rank(method='first'),5,labels=[1,2,3,4,5])
14 rfm['M_Score'] = pd.qcut(rfm['Monetary'],5,labels=[1,2,3,4,5])
16 rfm['RFM_Score'] = rfm['R_Score'].astype(str) + rfm['F_Score'].astype(str) + rfm['M_Score'].astype(str)
17 rfm.head()
₹
      User ID
                   Recency Frequency
                                   Monetary R_Score F_Score M_Score RFM_Score
    0 1.31e+18 21 days 05:05:56
                                58 101895.68
                                                 3
                                                        4
                                                               4
                                                                      344
    1 1.32e+18 1 days 02:00:29
                                73 150777.29
                                                        5
                                                               5
                                                                      555
                                                5
    2 1.33e+18 10 days 23:53:27
                                68 118626.97
                                                               4
                                                3
                                                        4
                                                                      344
    3 1.34e+18 45 days 08:31:30
                                38
                                    47488.03
                                                2
                                                        3
                                                               3
                                                                      233
                                                        2
    4 1.35e+18 22 days 17:11:37
                                34 19918.57
                                                3
                                                                      321
                                                               1
  1 #Customer Segmentation
  2 def Segment(row):
  3
  4
     if row['RFM Score'] =='555':
  5
        return 'Champion'
     elif row['R\_Score'] in ['4','5'] and row['F\_Score'] in ['4','5'] :
  6
        return 'Loyal'
  7
     elif row['R Score'] in ['1','2'] and row['F Score'] in ['4','5'] :
  8
  9
        return 'At Risk'
 10
     elif row['R Score'] in ['1','2'] and row['F Score'] in ['1','2'] :
        return 'Lost_Customer'
 11
 12
     else:
 13
        return 'Others'
15 rfm['Customer_Segment'] = rfm.apply(Segment,axis=1)
16
17 rfm
```

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	User ID	Recency	Frequency	Monetary	R_Score	F_Score	M_Score	RFM_Score	Customer_Segment
0	1.31e+18	21 days 05:05:56	58	101895.68	3	4	4	344	Others
1	1.32e+18	1 days 02:00:29	73	150777.29	5	5	5	555	Champion
2	1.33e+18	10 days 23:53:27	68	118626.97	3	4	4	344	Others
3	1.34e+18	45 days 08:31:30	38	47488.03	2	3	3	233	Others
4	1.35e+18	22 days 17:11:37	34	19918.57	3	2	1	321	Others
5	1.36e+18	110 days 19:52:10	18	6973.04	1	1	1	111	Others
6	1.37e+18	8 days 14:03:42	39	40287.62	4	3	3	433	Others
7	1.38e+18	42 days 02:00:05	11	4435.59	2	1	1	211	Others
8	1.45e+18	3 days 20:12:06	54	75359.78	4	4	4	444	Others
9	1.46e+18	1 days 16:49:06	71	134631.57	4	5	5	455	Others
10	1.47e+18	42 days 15:41:23	47	62212.01	2	3	3	233	Others
11	1.48e+18	128 days 12:28:17	34	23874.24	1	2	2	122	Others
12	1.49e+18	2 days 22:04:28	37	34199.30	4	3	2	432	Others
13	1.51e+18	137 days 11:27:40	24	17117.58	1	1	1	111	Others
14	1.52e+18	1 days 00:00:00	81	30957222.30	5	5	5	555	Champion
15	1.53e+18	23 days 00:17:04	23	33798.78	2	1	2	212	Others
16	1.54e+18	401 days 15:04:57	27	45027.17	1	2	3	123	Others
17	1.55e+18	1 days 04:26:38	69	116981.43	5	4	4	544	Others
18	1.5e+18	20 days 19:26:28	31	22546.58	3	2	2	322	Others
19	nan	1 days 00:11:30	70	1278582.57	5	5	5	555	Champion

Sales

1 df

→*

	Order Date Time	Order ID	Purchased Product ID	Quantity of SKU in the order	Category ID	Category alias	Brand ID	Price	User ID	Gender	Color	Metal	Main gem
0	2018-12-01 11:40:29+00:00	1.92e+18	1.84e+18	1	1.81e+18	jewelry.earring	0.0	561.51	1.52e+18	UnKown	red	gold	diamond
1	2018-12-01 17:38:31+00:00	1.92e+18	1.81e+18	1	1.81e+18	UnKown	nan	212.14	1.52e+18	UnKown	yellow	gold	UnKown
2	2018-12-02 13:53:42+00:00	1.93e+18	1.84e+18	1	1.81e+18	jewelry.pendant	1.0	54.66	1.52e+18	f	white	gold	sapphire
3	2018-12-02 17:44:02+00:00	1.93e+18	1.84e+18	1	1.81e+18	jewelry.pendant	0.0	88.90	1.52e+18	f	red	gold	diamond
4	2018-12-02 21:30:19+00:00	1.93e+18	1.87e+18	1	1.81e+18	jewelry.necklace	0.0	417.67	1.52e+18	UnKown	red	gold	amethyst
95906	2021-12-01 09:47:37+00:00	2.72e+18	1.52e+18	1	nan	0	1.52e+18	258.77	nan	UnKown	UnKown	UnKown	UnKown
95907	2021-12-01 09:50:37+00:00	2.72e+18	1.96e+18	1	1.81e+18	UnKown	0.0	229.99	1.52e+18	UnKown	red	gold	UnKown
95908	2021-12-01 09:52:06+00:00	2.72e+18	1.96e+18	1	1.81e+18	jewelry.earring	1.0	630.00	1.52e+18	f	white	gold	diamond
95909	2021-12-01 09:55:35+00:00	2.72e+18	1.84e+18	1	1.81e+18	jewelry.earring	0.0	83.42	1.52e+18	UnKown	white	gold	fianit
95910	2021-12-01 09:59:07+00:00	2.72e+18	1.96e+18	1	1.81e+18	jewelry.pendant	1.0	97.12	1.52e+18	UnKown	red	gold	UnKown

92955 rows × 13 columns

¹ count_of_orders = df['Order ID'].nunique()

² total_quantity = df['Quantity of SKU in the order'].sum()
3 avg_quantity = df['Quantity of SKU in the order'].mean()

```
4 total_revenue = df['Revenue'].sum()
  5 avg_total_price = df['Price'].mean()
  7 print("Number of Orders:", count_of_orders)
  8 print("Total Quantity:", total_quantity)
  9 print("Average Quantity:", avg_quantity)
 10 print("Total Revenue:", total_revenue)
 11 print("Average Total Price:", avg_total_price)
Number of Orders: 81
   Total Quantity: 92955
   Average Quantity: 1.0
   Total Revenue: 33291956.099999998
   Average Total Price: 358.1513216072293
  1 df['Year'] = df['Order Date Time'].dt.year
  2 sales_per_year = df.groupby('Year')['Revenue'].sum()
  3 sales_per_year
<del>_</del>
            Revenue
    Year
    2018
            33143.57
          4136225.75
    2019
          9042261.20
    2020
    2021 20080325.58
   dtype: float64
  1 df['Month'] = df['Order Date Time'].dt.month
  2 sales_per_month = df.groupby('Month')['Revenue'].sum()
  3 sales_per_month
₹
            Revenue
    Month
          1888248.28
      1
          1963446.45
      2
      3
          2247752.74
          1698547.76
      5
          1738255.59
          2020141.14
          2675748.23
      8
          4337431.41
      9
          3069369.17
      10
          2987954.16
          5596249.47
      11
          3068811.70
      12
   dtype: float64
  1 df['Hour'] = df['Order Date Time'].dt.hour
  2 most_hour_in_sales = df['Hour'].mode()[0]
  3 int(most_hour_in_sales)
```

Product

```
1 num_of_products=df['Purchased Product ID'].nunique()
  2 num_of_categories=df['Category ID'].nunique()
  3 most_color = df['Color'].mode()[0]
  4 most_metal = df['Metal'].mode()[0]
  5 most_main_gem = df['Main gem'].mode()[0]
  7
  8 print("Number of Products:", num_of_products)
  9 print("Number of Categories:", num_of_categories)
 10 print("Most Color:", most_color)
 11 print("Most Metal:", most_metal)
12 print("Most Main Gem:", most_main_gem)
Number of Products: 103
   Number of Categories: 3
   Most Color: red
   Most Metal: gold
   Most Main Gem: UnKown
  1 most_product_selling = df['Purchased Product ID'].mode()[0]
  2 product details = df[df['Purchased Product ID'] == most product selling]
  3 product_details.head(1)
<del>_</del>
                                     Quantity
                            Purchased
          Order Date
                       Order
                                       of SKU Category
                                                          Category Brand
                                                                                                           Main
                              Product
                                                                        Price User ID Gender Color Metal
                                                                                                                Revenue Year Month H
               Time
                                       in the
                         TD
                                                   TD
                                                             alias
                                                                     TD
                                                                                                            gem
                                  ID
                                        order
           2019-01-13
    148
                    1.96e+18
                             1.96e+18
                                           1 1.81e+18 jewelry.pendant
                                                                     1.0 123.15 1.52e+18
                                                                                                    gold diamond
                                                                                                                 123.15 2019
                                                                                               red
  1 Category_sales = df.groupby('Category alias')['Revenue'].sum()
  2 Category_sales.sort_values(ascending=False).head(5)
₹
                    Revenue
    Category alias
                  11503531.02
    jewelry.earring
     jewelry.ring
                  9679520.82
       UnKown
                  4336725.60
```

jewelry.bracelet 3020484.70 jewelry.pendant 2198554.30

dtype: float64

Customer

1 df

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	→	▼	

4)

6 print("Average revenue per customer by Gender:")

7 nrint(avg revenue gender)

₹		Order Date Time	Order ID	Purchased Product ID	Quantity of SKU in the order	Category ID	Category alias	Brand ID	Price	User ID	Gender	Color	Metal	Main gem	Revenue	Yea
	0	2018-12-01 11:40:29+00:00	1.92e+18	1.84e+18	1	1.81e+18	jewelry.earring	0.0	561.51	1.52e+18	UnKown	red	gold	diamond	561.51	201
	1	2018-12-01 17:38:31+00:00	1.92e+18	1.81e+18	1	1.81e+18	UnKown	nan	212.14	1.52e+18	UnKown	yellow	gold	UnKown	212.14	201
	2	2018-12-02 13:53:42+00:00	1.93e+18	1.84e+18	1	1.81e+18	jewelry.pendant	1.0	54.66	1.52e+18	f	white	gold	sapphire	54.66	201
	3	2018-12-02 17:44:02+00:00	1.93e+18	1.84e+18	1	1.81e+18	jewelry.pendant	0.0	88.90	1.52e+18	f	red	gold	diamond	88.90	201
	num_of_customers=df['User ID'].nunique() num_of_customers															
_ →	20															
	05006	2021-12-01	ე 7ეი⊥10	1 520±10	1	non	Λ	1 520±10	25 <u>0</u> 77	nan	Hakowa	Linkown	Linkown	Linkown	250 77	200
2	<pre>gender_values = df['Gender'].value_counts() gender_values</pre>															
_ _ *	95908	c802t -12-01 09:52:06+00:00	2.72e+18	1.96e+18	1	1.81e+18	jewelry.earring	1.0	630.00	1.52e+18	f	white	gold	diamond	630.00	202
	Gende f	2021-12-01 46655 00:00	2.72e+18	1.84e+18	1	1.81e+18	jewelry.earring	0.0	83.42	1.52e+18	UnKown	white	gold	fianit	83.42	202
	m	48921 -12-01 00-50-07 : 00:00 359 00-50-07 : 00:00	2.72e+18	1.96e+18	1	1.81e+18	jewelry.pendant	1.0	97.12	1.52e+18	UnKown	red	gold	UnKown	97.12	202
	dtype: ir	nt64														
1 2 3		revenue_gend df.groupby(' .agg(avg_r	Gender	•	tomer=('	Revenue	', 'mean'))									