



```
In [1]: # Loading the dataset using pandas

import pandas as pd
d = pd.read_csv('Customer_shopping_data.csv')
```

```
In [2]: # Display the first 5 rows
d.head()
```

```
Out[2]:
```

	Customer ID	Age	Gender	Item Purchased	Category	Purchase Amount (USD)	Location	Size
0	1	55	Male	Blouse	Clothing	53	Kentucky	L
1	2	19	Male	Sweater	Clothing	64	Maine	L
2	3	50	Male	Jeans	Clothing	73	Massachusetts	S
3	4	21	Male	Sandals	Footwear	90	Rhode Island	M
4	5	45	Male	Blouse	Clothing	49	Oregon	M

```
In [3]: # Display the last 5 rows
d.tail()
```

```
Out[3]:
```

	Customer ID	Age	Gender	Item Purchased	Category	Purchase Amount (USD)	Location	Size
3495	3496	24	Female	Shorts	Clothing	33	Georgia	XL
3496	3497	39	Female	Shoes	Footwear	38	Idaho	M
3497	3498	35	Female	Handbag	Accessories	48	Colorado	M
3498	3499	47	Female	Sandals	Footwear	35	Kentucky	M
3499	3500	22	Female	Socks	Clothing	29	New Jersey	L

```
In [4]: d.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 3500 entries, 0 to 3499
Data columns (total 18 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Customer ID                          3500 non-null   int64
1   Age                                  3500 non-null   int64
2   Gender                              3500 non-null   object
3   Item Purchased                       3500 non-null   object
4   Category                            3500 non-null   object
5   Purchase Amount (USD)                3500 non-null   int64
6   Location                             3500 non-null   object
7   Size                                 3500 non-null   object
8   Color                                3500 non-null   object
9   Season                               3500 non-null   object
10  Review Rating                        3463 non-null   float64
11  Subscription Status                  3500 non-null   object
12  Shipping Type                       3500 non-null   object
13  Discount Applied                     3500 non-null   object
14  Promo Code Used                      3500 non-null   object
15  Previous Purchases                    3500 non-null   int64
16  Payment Method                       3500 non-null   object
17  Frequency of Purchases                3500 non-null   object
dtypes: float64(1), int64(4), object(13)
memory usage: 492.3+ KB

```

```

In [5]: # Summerize statistics for numerical columns
d.describe()

```

```

Out[5]:

```

	Customer ID	Age	Purchase Amount (USD)	Review Rating	Previous Purchases
<b>count</b>	3500.000000	3500.000000	3500.000000	3463.000000	3500.000000
<b>mean</b>	1750.500000	44.032571	59.712857	3.748859	25.373143
<b>std</b>	1010.507298	15.233519	23.713364	0.716415	14.441421
<b>min</b>	1.000000	18.000000	20.000000	2.500000	1.000000
<b>25%</b>	875.750000	31.000000	38.000000	3.100000	13.000000
<b>50%</b>	1750.500000	44.000000	60.000000	3.800000	25.000000
<b>75%</b>	2625.250000	57.000000	80.000000	4.400000	38.000000
<b>max</b>	3500.000000	70.000000	100.000000	5.000000	50.000000

```

In [6]: # Descriptive statistics for all columns
d.describe(include='all')

```

Out[6]:

	Customer ID	Age	Gender	Item Purchased	Category	Purchase Amount (USD)	Location
<b>count</b>	3500.000000	3500.000000	3500	3500	3500	3500.000000	3500
<b>unique</b>	NaN	NaN	2	25	4	NaN	NaN
<b>top</b>	NaN	NaN	Male	Pants	Clothing	NaN	Indoor
<b>freq</b>	NaN	NaN	2652	156	1560	NaN	NaN
<b>mean</b>	1750.500000	44.032571	NaN	NaN	NaN	59.712857	NaN
<b>std</b>	1010.507298	15.233519	NaN	NaN	NaN	23.713364	NaN
<b>min</b>	1.000000	18.000000	NaN	NaN	NaN	20.000000	NaN
<b>25%</b>	875.750000	31.000000	NaN	NaN	NaN	38.000000	NaN
<b>50%</b>	1750.500000	44.000000	NaN	NaN	NaN	60.000000	NaN
<b>75%</b>	2625.250000	57.000000	NaN	NaN	NaN	80.000000	NaN
<b>max</b>	3500.000000	70.000000	NaN	NaN	NaN	100.000000	NaN

```
In [7]: # Count the number of missing (null or NaN) values in each column
d.isnull().sum()
```

```
Out[7]: Customer ID          0
Age              0
Gender           0
Item Purchased   0
Category         0
Purchase Amount (USD)  0
Location         0
Size            0
Color           0
Season          0
Review Rating    37
Subscription Status  0
Shipping Type    0
Discount Applied  0
Promo Code Used  0
Previous Purchases  0
Payment Method   0
Frequency of Purchases  0
dtype: int64
```

```
In [8]: # Filling missing values in 'Review Rating' by the median value of each 'Category'
d['Review Rating'] = d['Review Rating'].fillna(d.groupby('Category')['Review Rating'].median())
```

```
In [9]: d.isnull().sum()
```

```
Out[9]: Customer ID      0
        Age             0
        Gender          0
        Item Purchased   0
        Category        0
        Purchase Amount (USD) 0
        Location         0
        Size            0
        Color           0
        Season          0
        Review Rating    0
        Subscription Status 0
        Shipping Type    0
        Discount Applied 0
        Promo Code Used  0
        Previous Purchases 0
        Payment Method   0
        Frequency of Purchases 0
        dtype: int64
```

```
In [14]: # Renaming columns according to snake casing for better readability and documentation
        # Converting all column names to lowercase
        d.columns = d.columns.str.lower()

        # Replacing spaces with underscores
        d.columns = d.columns.str.replace(' ', '_')

        # Renaming a 'purchase_amount_(usd)' column manually
        d = d.rename( columns = {'purchase_amount_(usd)': 'purchase_amount'} )
```

```
In [16]: d.columns
```

```
Out[16]: Index(['customer_id', 'age', 'gender', 'item_purchased', 'category',
               'purchase_amount', 'location', 'size', 'color', 'season',
               'review_rating', 'subscription_status', 'shipping_type',
               'discount_applied', 'promo_code_used', 'previous_purchases',
               'payment_method', 'frequency_of_purchases'],
              dtype='object')
```

```
In [23]: # Creating New Column age_group
        # Manually define age boundaries (bins) and labels for analysis using pd.cut.
        # Young Adult: 18–25 | Adult: 26–35 | Middle-aged: 36–60 | Senior: 61–

        bins = [17, 25, 35, 60, 100]
        labels = ['Young Adult', 'Adult', 'Middle-aged', 'Senior']

        d['age_group'] = pd.cut(
            d['age'],
            bins = bins,
            labels = labels,
            right = True,
            include_lowest = True,
            # Ensures the right edge of the bin is included
            # Ensures the lowest value (18) is included
```

```
)
```

```
In [24]: d[['age', 'age_group']].head(10)
```

```
Out[24]:
```

	age	age_group
--	-----	-----------

0	55	Middle-aged
---	----	-------------

1	19	Young Adult
---	----	-------------

2	50	Middle-aged
---	----	-------------

3	21	Young Adult
---	----	-------------

4	45	Middle-aged
---	----	-------------

5	46	Middle-aged
---	----	-------------

6	63	Senior
---	----	--------

7	27	Adult
---	----	-------

8	26	Adult
---	----	-------

9	57	Middle-aged
---	----	-------------

```
In [29]: # Creating New Column purchase_frequency_days by mapping text labels to numeri
```

```
frequency_mapping = {  
    'Fortnightly' : 14,  
    'Weekly' : 7,  
    'Monthly' : 30,  
    'Quarterly' : 90,  
    'Bi-Weekly' : 14,  
    'Annually' : 365,  
    'Every 3 Months' : 90  
}
```

```
d['purchase_frequency_days'] = d['frequency_of_purchases'].map(frequency_mappi
```

```
In [31]: d[['purchase_frequency_days', 'frequency_of_purchases']].head(10)
```

Out[31]:

	<b>purchase_frequency_days</b>	<b>frequency_of_purchases</b>
--	--------------------------------	-------------------------------

<b>0</b>	14	Fortnightly
<b>1</b>	14	Fortnightly
<b>2</b>	7	Weekly
<b>3</b>	7	Weekly
<b>4</b>	365	Annually
<b>5</b>	7	Weekly
<b>6</b>	90	Quarterly
<b>7</b>	7	Weekly
<b>8</b>	365	Annually
<b>9</b>	90	Quarterly

```
In [33]: d[['discount_applied', 'promo_code_used']].head(10)
```

Out[33]:

	<b>discount_applied</b>	<b>promo_code_used</b>
--	-------------------------	------------------------

<b>0</b>	Yes	Yes
<b>1</b>	Yes	Yes
<b>2</b>	Yes	Yes
<b>3</b>	Yes	Yes
<b>4</b>	Yes	Yes
<b>5</b>	Yes	Yes
<b>6</b>	Yes	Yes
<b>7</b>	Yes	Yes
<b>8</b>	Yes	Yes
<b>9</b>	Yes	Yes

```
In [37]: (d['discount_applied'] == d['promo_code_used']).all()
```

Out[37]: np.True\_

```
In [38]: # Dropping promo_code_used column  
  
d = d.drop('promo_code_used', axis = 1)
```

```
In [39]: d.columns
```

```
Out[39]: Index(['customer_id', 'age', 'gender', 'item_purchased', 'category',  
              'purchase_amount', 'location', 'size', 'color', 'season',  
              'review_rating', 'subscription_status', 'shipping_type',  
              'discount_applied', 'previous_purchases', 'payment_method',  
              'frequency_of_purchases', 'age_group', 'purchase_frequency_days'],  
              dtype='object')
```

### Connecting Python script to PostgreSQL

```
In [42]: !pip install psycopg2-binary sqlalchemy
```

```

Downloading psycpg2_binary-2.9.11-cp310-cp310-win_amd64.whl.metadata (5.1 k
B)

```

Downloading sqlalchemy-2.0.44-cp310-cp310-win\_amd64.whl.metadata (9.8 kB)

Downloading greenlet-3.2.4-cp310-cp310-win\_amd64.whl.metadata (4.2 kB)

Downloading psycpg2\_binary-2.9.11-cp310-cp310-win\_amd64.whl (2.7 MB)

```
----- 0.0/2.7 MB ? eta -:--:--
----- 0.3/2.7 MB ? eta -:--:--
----- 0.8/2.7 MB 3.4 MB/s eta 0:00:01
----- 0.8/2.7 MB 3.4 MB/s eta 0:00:01
----- 1.0/2.7 MB 1.5 MB/s eta 0:00:02
----- 1.0/2.7 MB 1.5 MB/s eta 0:00:02
----- 1.0/2.7 MB 1.5 MB/s eta 0:00:02
----- 1.3/2.7 MB 895.5 kB/s eta 0:00:02
----- 1.6/2.7 MB 976.0 kB/s eta 0:00:02
----- 1.8/2.7 MB 1.0 MB/s eta 0:00:01
----- 2.1/2.7 MB 1.0 MB/s eta 0:00:01
----- 2.4/2.7 MB 1.1 MB/s eta 0:00:01
----- 2.7/2.7 MB 1.1 MB/s 0:00:02
```

```
----- 0.0/2.1 MB ? eta -:-:--
----- 0.3/2.1 MB ? eta -:-:--
----- 0.5/2.1 MB 1.3 MB/s eta 0:00:02
----- 1.0/2.1 MB 1.6 MB/s eta 0:00:01
----- 1.3/2.1 MB 1.7 MB/s eta 0:00:01
----- 1.6/2.1 MB 1.7 MB/s eta 0:00:01
----- 1.8/2.1 MB 1.6 MB/s eta 0:00:01
----- 2.1/2.1 MB 1.7 MB/s 0:00:01
```

```
Installing collected packages: psycpg2-binary, greenlet, sqlalchemy
```

[illegible]



```

----- 2/3 [sqlalchemy]
----- 2/3 [sqlalchemy]
----- 2/3 [sqlalchemy]
----- 2/3 [sqlalchemy]
----- 2/3 [sqlalchemy]
----- 2/3 [sqlalchemy]
----- 2/3 [sqlalchemy]
----- 2/3 [sqlalchemy]
----- 2/3 [sqlalchemy]
----- 2/3 [sqlalchemy]
----- 2/3 [sqlalchemy]
----- 2/3 [sqlalchemy]
----- 2/3 [sqlalchemy]
----- 2/3 [sqlalchemy]
----- 2/3 [sqlalchemy]
----- 2/3 [sqlalchemy]
----- 2/3 [sqlalchemy]
----- 2/3 [sqlalchemy]
----- 2/3 [sqlalchemy]
----- 3/3 [sqlalchemy]

```

Successfully installed greenlet-3.2.4 psycpg2-binary-2.9.11 sqlalchemy-2.0.44

In [43]: `!pip install psycpg2-binary sqlalchemy`

```

Requirement already satisfied: psycpg2-binary in c:\users\soleman\appdata\local\programs\python\python310\lib\site-packages (2.9.11)
Requirement already satisfied: sqlalchemy in c:\users\soleman\appdata\local\programs\python\python310\lib\site-packages (2.0.44)
Requirement already satisfied: greenlet>=1 in c:\users\soleman\appdata\local\programs\python\python310\lib\site-packages (from sqlalchemy) (3.2.4)
Requirement already satisfied: typing-extensions>=4.6.0 in c:\users\soleman\appdata\local\programs\python\python310\lib\site-packages (from sqlalchemy) (4.12.2)

```

In [45]: `from sqlalchemy import create_engine`

```

# Step 1: Connect to PostgreSQL
# Replace placeholders with your actual details
username = "postgres"      # default user
password = "123456789"     # the password you set during installation
host = "localhost"        # if running locally
port = "5432"             # default PostgreSQL port
database = "customer_behavior" # the database you created in pgAdmin

engine = create_engine(f"postgresql+psycpg2://{username}:{password}@{host}:{port}/{database}")

# Step 2: Load DataFrame into PostgreSQL
table_name = "customer"    # choose any table name
d.to_sql(table_name, engine, if_exists="replace", index=False)
print(f>Data successfully loaded into table '{table_name}' in database '{database}'<

```

Data successfully loaded into table 'customer' in database 'customer\_behavior'.

In [ ]: