

DMBI Lab

EXPERIMENT NO. 4

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AIM : To perform exploratory data analysis and data visualization using python.

1. Descriptive analysis - statistical measures of data (Central tendency) 2.

Descriptive analysis - statistical measures of data (Dispersion)

3. Correlation between attributes

4. Different Visualization techniques and use of it.

Inferences derived after every analysis.

Theory :

Exploratory Data Analysis (EDA) is a pivotal initial step in data analysis, aimed at comprehensively understanding dataset characteristics through statistical measures and visualizations. Central tendency metrics like mean, median, and mode offer insights into typical data values, while dispersion metrics such as standard deviation and variance indicate data variability. Correlation analysis, utilizing Pearson and Spearman coefficients, reveals associations between variables. Visualization techniques such as histograms, box plots, scatter plots, and heatmaps provide intuitive representations of data distributions, outliers, and relationships. Through EDA, analysts derive valuable insights into dataset structures, trends, and potential anomalies, guiding subsequent analytical decisions and modeling processes.

Link to the dataset : <https://www.kaggle.com/datasets/aungpyaeap/supermarket-sales>

Steps:

1. Import necessary libraries and load the dataset.

```
✓ [1] import pandas as pd  
0s      import numpy as np  
      import seaborn as sns  
      import matplotlib.pyplot as plt
```

```
✓ [2] data = pd.read_csv('supermarket_sales.csv')
```


✓ 0s  `print(data.head())`

```
Invoice ID Branch      City Customer type Gender \
0 750-67-8428      A      Yangon      Member  Female
1 226-31-3081      C  Naypyitaw      Normal  Female
2 631-41-3108      A      Yangon      Normal  Male
3 123-19-1176      A      Yangon      Member  Male
4 373-73-7910      A      Yangon      Normal  Male

Product line  Unit price  Quantity  Tax 5%  Total  Date \
0 Health and beauty      74.69         7  26.1415  548.9715  1/5/2019
1 Electronic accessories  15.28         5   3.8200   80.2200  3/8/2019
2 Home and lifestyle      46.33         7  16.2155  340.5255  3/3/2019
3 Health and beauty      58.22         8  23.2880  489.0480  1/27/2019
4 Sports and travel      86.31         7  30.2085  634.3785  2/8/2019

Time      Payment  cogs  gross margin percentage  gross income  Rating
0 13:08      Ewallet  522.83      4.761905      26.1415      9.1
1 10:29      Cash     76.40      4.761905      3.8200      9.6
2 13:23  Credit card  324.31      4.761905      16.2155      7.4
3 20:33      Ewallet  465.76      4.761905      23.2880      8.4
4 10:37      Ewallet  604.17      4.761905      30.2085      5.3
```

2. Descriptive analysis - Central tendency: Mean, Median, and Mode.

✓ 0s  `mean_values = data.mean()`
`median_values = data.median()`
`mode_values = data.mode().iloc[0]`

`print("Mean Values:")`
`print(mean_values)`
`print("\nMedian Values:")`
`print(median_values)`
`print("\nMode Values:")`
`print(mode_values)`

Median Values:		Mean Values:	
Unit price	55.230000	Unit price	55.672130
Quantity	5.000000	Quantity	5.510000
Tax 5%	12.088000	Tax 5%	15.379369
Total	253.848000	Total	322.966749
cogs	241.760000	cogs	307.587380
gross margin percentage	4.761905	gross margin percentage	4.761905
gross income	12.088000	gross income	15.379369
Rating	7.000000	Rating	6.972700
dtype: float64		dtype: float64	

```

Mode Values:
Invoice ID          101-17-6199
Branch              A
City                Yangon
Customer type       Member
Gender              Female
Product line        Fashion accessories
Unit price          83.77
Quantity            10.0
Tax 5%              4.154
Total               87.234
Date                2/7/2019
Time                14:42
Payment              Ewallet
cogs                83.08
gross margin percentage  4.761905
gross income         4.154
Rating              6.0
Name: 0, dtype: object

```

3. Descriptive analysis - Dispersion : Standard deviation and Variance

```

✓ 0s ▶ std_deviation = data.std()
      variance = data.var()

      print("\nStandard Deviation:")
      print(std_deviation)
      print("\nVariance:")
      print(variance)

```

```

Standard Deviation:
Unit price          26.494628
Quantity            2.923431
Tax 5%              11.708825
Total               245.885335
cogs                234.176510
gross margin percentage  0.000000
gross income         11.708825
Rating              1.718580
dtype: float64

```

```

Variance:
Unit price          701.965331
Quantity            8.546446
Tax 5%              137.096594
Total               60459.598018
cogs                54838.637658
gross margin percentage  0.000000
gross income        137.096594
Rating              2.953518
dtype: float64

```

Inference :

Standard Deviation: Indicates the amount of variation or dispersion from the mean. Here, Total price values are spread out over a wider range while Rating value shows little spreading.

Variance: Represents the average squared deviation from the mean. Here, gross margin percentage has same value as mean, while Total value has large deviation from mean.

4. Correlation between attributes : Pearson and Spearman correlation

```

0s ▶ subset_data = data[['Unit price', 'Quantity', 'gross income', 'Total']]
    pearson_corr = subset_data.corr(method='pearson')
    spearman_corr = subset_data.corr(method='spearman')

    print("\nPearson Correlation:")
    print(pearson_corr)
    print("\nSpearman Correlation:")
    print(spearman_corr)

```

Pearson Correlation:

	Unit price	Quantity	gross income	Total
Unit price	1.000000	0.010778	0.633962	0.633962
Quantity	0.010778	1.000000	0.705510	0.705510
gross income	0.633962	0.705510	1.000000	1.000000
Total	0.633962	0.705510	1.000000	1.000000

Spearman Correlation:

	Unit price	Quantity	gross income	Total
Unit price	1.000000	0.011167	0.630054	0.630054
Quantity	0.011167	1.000000	0.735265	0.735265
gross income	0.630054	0.735265	1.000000	1.000000
Total	0.630054	0.735265	1.000000	1.000000

Inference :

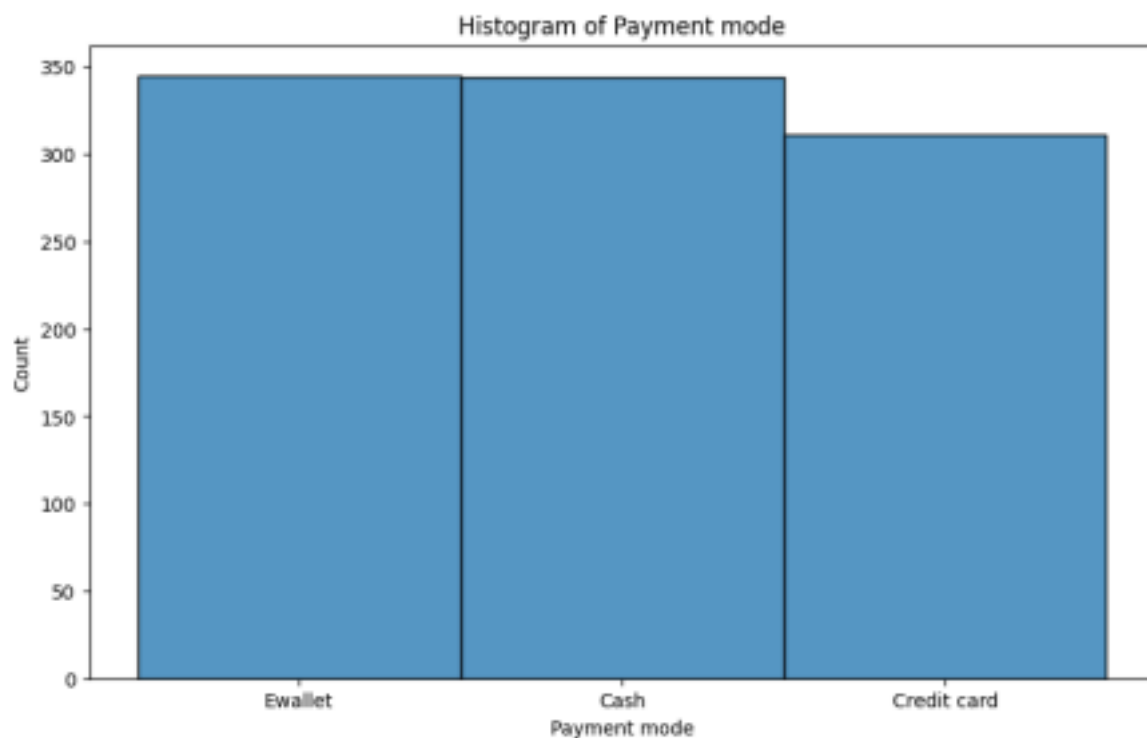
Pearson Correlation: Measures the linear correlation between two continuous variables. Unit price and gross income are highly correlated while unit price and quantity are less correlated.

Spearman Correlation: Measures the strength and direction of association between two ranked variables. Here, gross income and total are closely related while unit price and quantity are not.

5. Visualization techniques

1. Histogram :

```
✓ [12] plt.figure(figsize=(10,6))  
0s sns.histplot(data['Payment'])  
plt.title('Histogram of Column')  
plt.xlabel('Payment mode')  
plt.ylabel('Count')  
plt.show()
```

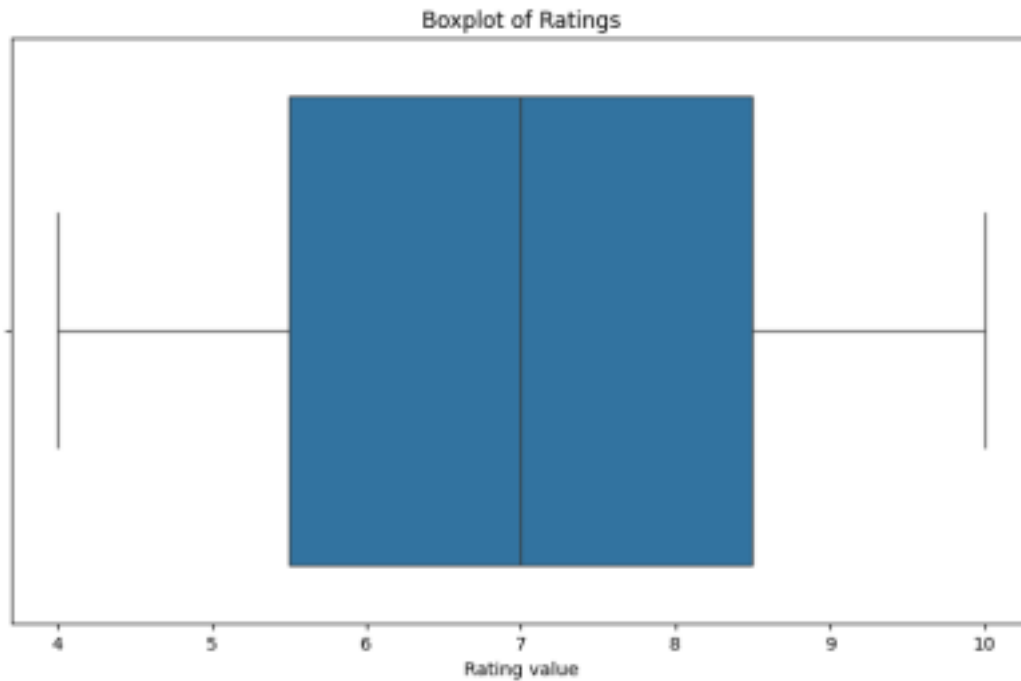


2. Box plot :

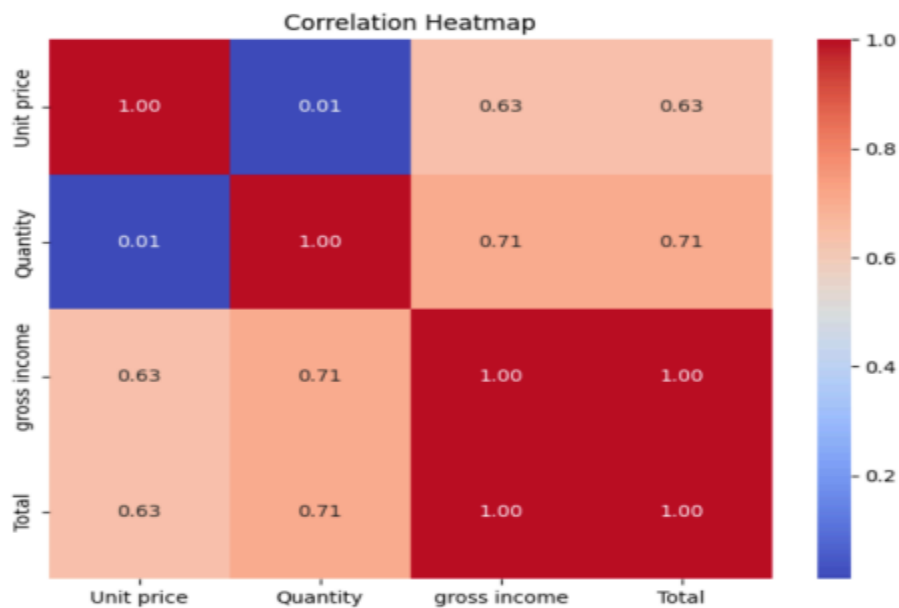
✓
08



```
plt.figure(figsize=(10,6))
sns.boxplot(x=data['Rating'])
plt.title('Boxplot of Column')
plt.xlabel('Values')
plt.show()
```



3. Heatmap:



Inference :

Histogram: Provides a graphical representation of the distribution of numerical data. Here, It helps to understand the frequency distribution of Payment mode type.

Box plot: Displays the distribution of Rating data through quartiles. It's useful for detecting outliers and comparing distributions between different groups.

Heatmap: Visualizes the correlation matrix between variables. Here, It identifies the correlation between unit price, gross income, total, and quantity values.

CONCLUSION : Hence we have performed Exploratory data analysis (EDA) on our chosen dataset of Supermarket Sales, and also performed Data visualization using Python on the dataset.