

S.G.T.B.I.M.I.T**BCA - BCAP - 212 - DATA SCIENCE****PRACTICAL FILE QUESTIONS**

S. No.	Detailed Statement										
1	Create 2 numpy arrays with 5 elements using arange() and linspace() and display the implement the concept of slicing on them.										
2	Create two 2-d arrays and perform addition, subtraction and multiplication on these arrays. Print the value of bordered elements of both matrices.										
3	Create two pandas series from a dictionary of values and an ndarray and display the values from 2 nd to 5 th index. Print the index, minimum and maximum values in the first series.										
4	Create a Series and print all the elements that are above 75th percentile. Print minimum, maximum, and sum of series using aggregate()										
5	Series objects Temp1, temp2, temp3, temp 4 stores the temperature of days of week 1, week 2, week 3, week 4. Write a script to:- a. Print average temperature per week b. Print the average temperature of the entire month										
6	Create a series containing 2 NaN values. Perform check for null values and then replace the null values with 0. Perform any 5 statistical functions on the series.										
7	Two Series object, Population stores the details of four metro cities of India and another object AvgIncome store the total average income reported in four years in these cities. Calculate income per capita for each of these metro cities.										
8	Given two series S1 and S2. <table border="0"> <tr> <td>S1</td> <td>S2</td> </tr> <tr> <td>A 39</td> <td>A 10</td> </tr> <tr> <td>B 41</td> <td>B 10</td> </tr> <tr> <td>C 42</td> <td>D 10</td> </tr> <tr> <td>D 44</td> <td>F 10</td> </tr> </table> Find the output for following python pandas statements? a. S1[: 2]*100 b. S1 * S2 c. S2[: : -1]*10	S1	S2	A 39	A 10	B 41	B 10	C 42	D 10	D 44	F 10
S1	S2										
A 39	A 10										
B 41	B 10										
C 42	D 10										
D 44	F 10										
9	Write a program to create a Series having 10 random numbers in the range of 10 and 20										
10	Consider a series object s10 that stores the number of students in each section of class 12 as shown below. First two sections have been given task for selling tickets @ Rs.100/- per ticket as a part of social experiment. Write code to create the series and display how much section A and B have collected. A-39, B- 31, C- 32, D- 34, E- 35										
11	Write a program to create a DataFrame to store weight, age and name of										

	three people. Print the DataFrame and its transpose. Add 5 rows in the dataframe through code. Rename the Weight column as Wgt. And then print the index names, column names and total amount of data. Sort the dataframe on the basis of age.
12	Create a DataFrame having age, name, weight of five students. Print the dataframe using head(). Modify the weight of student in first and 4 th row. Display only the weight of first and fourth rows before and after modification.
13	Create a DataFrame based on E-Commerce data and generate mean, mode, median.
14	Write a Program to create a CSV file with student data containing 10 rows. create its DataFrame and use describe() to display its statistics. Write all the steps and definition of all the statistical functions displayed.
15	Consider the DataFrame QtrSales where each row contains the item category, item name and expenditure and group the rows by category, and print the average expenditure per category
16	Write a program to implement pivot() and pivot-table() on a DataFrame
17	Write a program to find mean absolute deviation on a DataFrame.
18	Create a DataFrame based on employee data and generate quartile and variance.
19	Program to implement Skewness on Random data.
20	Create a DataFrame on any Data and compute statistical function of Kurtosis.
21	<p>CASE STUDY ON :CRIME_BY_STATE_RT.CSV crime by state rt.zip</p> <p>21.1: Find the 75% percentile for data 21.2: Apply aggregate function to find count, min , max and sum . 21.3: Perform group by on State and Year 21.4: Calculate the number of entries year wise 21.5: Create a pie chart for all states 21.6: Plot the no. of murders state-wise 21.7: Check the data type of all column 21.8: Check the robbery column for NULL values. if exist replace with 0 21.9: Select the data set where robbery greater than 10 and year= 2001 21.10: describe () the dataset 21.11: Perform sorting on Arson and Robbery 21.12: Perform renaming on two columns</p>
22	<p>CASE STUDY ON: investigating clinical data investigating clinical data.csv</p> <p>22.1: describe () the dataset 22.2: Perform renaming on two columns 22.3: Is there any relation between the age of the patient and having a delay on the date of appointment. 22.4: Is there any relation between gender and delaying appointments? (Show by plotting bar graph)</p>

	<p>22.5: Do people delay their appointments when they have no scholarship? (Show by plotting)</p> <p>22.6: Is there any relationship between gender having scholarships and delaying the appointment? (show by plotting)</p> <p>22.7: Which gender has more appointments in which illness? (Show by plotting)</p> <p>22.8: Does a person suffering from alcoholism tend to delay appointments? (Show by plotting)</p> <p>22.9: Perform group by and count on neighborhood</p> <p>22.10: Perform agg() function on dataset.</p>
23	<p>CASE STUDY ON: Credit card transaction</p> <p>Credit card transactions - India - Simple.csv</p> <p>23.1: Show the head and visualize the basic stats for all columns.</p> <p>23.2: Display pivot table to create a summary for the total amount spent</p> <p>a) month b) city</p> <p>23.3: Analyze the impact of gender to study consumer behavior</p> <p>23.4: Check for NULL values in columns.If they exist replace them with appropriate values.</p> <p>23.5: Analyze the relationship type between expense type and amount.</p> <p>23.6: Analyze the spending habits by city and gender.</p> <p>23.7: Implement Skewness and kurtosis on data</p> <p>Draw plots along with an appropriate question</p>
24	CASE STUDY ON PROJECT TOPIC
25	RESEARCH PAPER

Q1

```
import numpy as np
```

```
# Using arange() to create an array with 5 elements
```

```
array_arange = np.arange(1, 10, 2) # Start at 1, stop before 10, step by 2
```

```
print("Array created using arange():", array_arange)
```

```
# Using linspace() to create an array with 5 elements
```

```
array_linspace = np.linspace(1, 10, 5) # Start at 1, stop at 10, with 5 elements
```

```
print("Array created using linspace():", array_linspace)
```

```
# Slicing the arrays
```

```
slice_arange = array_arange[1:4] # Get elements from index 1 to index 3 (exclusive)
```

```
print("Sliced array using arange():", slice_arange)
```

```
slice_linspace = array_linspace[2:] # Get elements from index 2 to the end
```

```
print("S")
```

OUTPUT

```
===== RESTART: C:/Users/natur/OneDrive/Desktop/array.py =
Array created using arange(): [1 3 5 7 9]
Array created using linspace(): [ 1.    3.25  5.5   7.75 10. ]
Sliced array using arange(): [3 5 7]
S
```

Q2

```
import numpy as np
```

```
array1 = np.array([[1, 2, 3], [4, 5, 6], [7, 8, 9]])
```

```
array2 = np.array([[9, 8, 7], [6, 5, 4], [3, 2, 1]])
```

```
addition_result = array1 + array2
```

```
subtraction_result = array1 - array2
```

```
multiplication_result = array1 * array2
```

```
bordered_elements_array1 = array1[[0, -1], :]
```

```
bordered_elements_array2 = array2[:, [0, -1]]
```

```
print("Array 1:")
```

```
print(array1)
```

```
print("Array 2:")
```

```
print(array2)
```

```
print("Addition result:")
```

```
print(addition_result)
```

```
print("Subtraction result:")
```

```
print(subtraction_result)
```

```
print("Multiplication result:")
```

```
print(multiplication_result)
```

```
print("Bordered elements of Array 1:")
```

```
print(bordered_elements_array1)
```

```
print("Bordered elements of Array 2:")
```

```
===== RESTART: C:/Use
```

```
Array 1:
```

```
[[1 2 3]
 [4 5 6]
 [7 8 9]]
```

```
Array 2:
```

```
[[9 8 7]
 [6 5 4]
 [3 2 1]]
```

```
Addition result:
```

```
[[10 10 10]
 [10 10 10]
 [10 10 10]]
```

```
Subtraction result:
```

```
[[ -8  -6  -4]
 [ -2   0   2]
 [  4   6   8]]
```

```
Multiplication result:
```

```
[[ 9 16 21]
 [24 25 24]
 [21 16  9]]
```

```
Bordered elements of Array 1:
```

```
[[1 2 3]
 [7 8 9]]
```

```
Bordered elements of Array 2:
```

```
[[9 7]
 [6 4]
 [3 1]]
```

```
print(bordered_elements_array2)
```

Q3

```
import pandas as pd
```

```
import numpy as np
```

```
dict_data = {'A': 10, 'B': 20, 'C': 30, 'D': 40, 'E': 50}
```

```
series_dict = pd.Series(dict_data)
```

```
print("Series from dictionary:")
```

```
print(series_dict)
```

```
ndarray_data = np.array([1, 2, 3, 4, 5])
```

```
series_ndarray = pd.Series(ndarray_data)
```

```
print("\nSeries from ndarray:")
```

```
print(series_ndarray)
```

```
print("\nValues from 2nd to 5th index of Series from dictionary:")
```

```
print(series_dict[1:5]) # Indexing starts from 0, so 1 corresponds to the 2nd index and 5  
corresponds to the 6th index
```

```
print("Values from 2nd to 5th index of Series from ndarray:")
```

```
print(series_ndarray[1:5])
```

```
print("\nIndex, minimum, and maximum values of the Series from dictionary:")
```

```
print("Index:", series_dict.index.tolist())
```

```
print("Minimum value:", series_dict.min())
```

```
print("Maximum value:", series_dict.max())
```

```
===== RESTART: C:/Users/natur/OneDrive/Desktop/array.py
Series from dictionary:
A    10
B    20
C    30
D    40
E    50
dtype: int64

Series from ndarray:
0     1
1     2
2     3
3     4
4     5
dtype: int32

Values from 2nd to 5th index of Series from dictionary:
B    20
C    30
D    40
E    50
dtype: int64
Values from 2nd to 5th index of Series from ndarray:
1     2
2     3
3     4
4     5
dtype: int32

Index, minimum, and maximum values of the Series from dictionary:
Index: ['A', 'B', 'C', 'D', 'E']
Minimum value: 10
Maximum value: 50
```

Q4

```
import pandas as pd
```

```
import numpy as np
```

```
data = [10, 20, 30, 40, 50, 60, 70, 80, 90, 100]
```

```
series = pd.Series(data)
```

```
print("Original Series:")
```

```
print(series)
```

```
percentile_75 = series.quantile(0.75)
```

```
elements_above_75th_percentile = series[series > percentile_75]
```

```
print("\nElements above the 75th percentile:")
```

```
print(elements_above_75th_percentile)
```

```
aggregates = series.agg(['min', 'max', 'sum'])
```

```
print("\nAggregate results - Minimum, Maximum, Sum:")
```

```
print(aggregates)
```

```
----- RESIARI: C:/Users/nadur/One
```

```
Original Series:
```

```
0    10
1    20
2    30
3    40
4    50
5    60
6    70
7    80
8    90
9   100
```

```
dtype: int64
```

```
Elements above the 75th percentile:
```

```
7    80
8    90
9   100
```

```
dtype: int64
```

```
Aggregate results - Minimum, Maximum, Sum:
```

```
min    10
max   100
sum   550
```

```
dtype: int64
```


Q5

```
import pandas as pd
```

```
temp_data = {  
    'Week 1': [22, 25, 23, 24, 21], # Sample temperatures for Week 1  
    'Week 2': [26, 27, 28, 25, 24], # Sample temperatures for Week 2  
    'Week 3': [20, 22, 21, 23, 24], # Sample temperatures for Week 3  
    'Week 4': [24, 25, 26, 27, 28] # Sample temperatures for Week 4  
}
```

```
Temp1 = pd.Series(temp_data['Week 1'])
```

```
Temp2 = pd.Series(temp_data['Week 2'])
```

```
Temp3 = pd.Series(temp_data['Week 3'])
```

```
Temp4 = pd.Series(temp_data['Week 4'])
```

```
avg_temp_per_week = {
```

```
    'Week 1': Temp1.mean(),
```

```
    'Week 2': Temp2.mean(),
```

```
    'Week 3': Temp3.mean(),
```

```
    'Week 4': Temp4.mean()
```

```
}
```

```
print("Average temperature per week:")
```

```
for week, avg_temp in avg_temp_per_week.items():
```

```
    print(f"{week}: {avg_temp}")
```

```
all_temps = pd.concat([Temp1, Temp2, Temp3, Temp4])
```

```
avg_temp_month = all_temps.mean()
```

```
print("\nAverage temperature of the entire month:", avg_temp_month)
```

```
Average temperature per week:  
Week 1: 23.0  
Week 2: 26.0  
Week 3: 22.0  
Week 4: 26.0  
  
Average temperature of the entire month: 24.25
```

Q6

```
import pandas as pd
```

```
import numpy as np
```

```
data = [10, np.nan, 30, np.nan, 50]
```

```
series = pd.Series(data)
```

```
print("Original Series:")
```

```
print(series)
```

```
null_check = series.isnull()
```

```
print("\nNull check:")
```

```
print(null_check)
```

```
series_filled = series.fillna(0)
```

```
print("\nSeries with null values replaced by 0:")
```

```
print(series_filled)
```

```
mean_value = series_filled.mean()
```

```
median_value = series_filled.median()
```

```
std_deviation = series_filled.std()
```

```
min_value = series_filled.min()
```

```
max_value = series_filled.max()
```

```
print("\nStatistical functions on the Series:")
```

```
print("Mean:", mean_value)
```

```
print("Median:", median_value)
```

```
print("Standard Deviation:", std_deviation)
```

```
print("Minimum Value:", min_value)
```

```
print("Maximum Value:", max_value)
```

```
Original Series:
```

```
0    10.0  
1     NaN  
2    30.0  
3     NaN  
4    50.0  
dtype: float64
```

```
Null check:
```

```
0    False  
1     True  
2    False  
3     True  
4    False  
dtype: bool
```

```
Series with null values replaced by 0:
```

```
0    10.0  
1     0.0  
2    30.0  
3     0.0  
4    50.0  
dtype: float64
```

```
Statistical functions on the Series:
```

```
Mean: 18.0  
Median: 10.0  
Standard Deviation: 21.6794833886788  
Minimum Value: 0.0  
Maximum Value: 50.0
```

Q7

```
import pandas as pd
```

```
population_data = {'Delhi': 20000000, 'Mumbai': 22000000, 'Kolkata': 15000000, 'Chennai':  
12000000}
```

```
avg_income_data = {'Delhi': 800000, 'Mumbai': 900000, 'Kolkata': 700000, 'Chennai': 600000}
```

```
Population = pd.Series(population_data)
```

```
AvgIncome = pd.Series(avg_income_data)
```

```
income_per_capita = AvgIncome / Population
```

```
income_per_capita = income_per_capita.fillna(0) # Replace NaN values with 0 if any
```

```
print("Income per capita for each metro city:")
```

```
print(income_per_capita)
```

```
===== KESIAKI: C:/Users/natur/Or  
Income per capita for each metro city:  
Delhi      0.040000  
Mumbai     0.040909  
Kolkata    0.046667  
Chennai    0.050000  
dtype: float64
```

Q8

```
import pandas as pd
```

```
data_S1 = {'A': 39, 'B': 41, 'C': 42, 'D': 44}
```

```
data_S2 = {'A': 10, 'B': 10, 'D': 10, 'F': 10}
```

```
S1 = pd.Series(data_S1)
```

```
S2 = pd.Series(data_S2)
```

```
output_a = S1[:2] * 100
```

```
output_b = S1 * S2
```

```
output_c = S2[::-1] * 10
```

```
print("Output of S1[:2]*100:")
```

```
print(output_a)
```

```
print("\nOutput of S1 * S2:")
```

```
print(output_b)
```

```
print("\nOutput of S2[::-1]*10:")
```

```
print(output_c)
```

```
Output of S1[:2]*100:
A    3900
B    4100
dtype: int64
```

```
Output of S1 * S2:
A    390.0
B    410.0
C      NaN
D    440.0
F      NaN
dtype: float64
```

```
Output of S2[::-1]*10:
F    100
D    100
B    100
A    100
dtype: int64
```

Q9

```
import pandas as pd
```

```
import numpy as np
```

```
# Generate 10 random numbers in the range of 10 and 20
```

```
random_numbers = np.random.randint(10, 21, size=10)
```

```
# Create a Pandas Series from the random numbers
```

```
random_series = pd.Series(random_numbers)
```

```
print("Random Series:")
```

```
print(random_series)
```

```
Random Series:
```

```
0    17
```

```
1    14
```

```
2    19
```

```
3    12
```

```
4    11
```

```
5    13
```

```
6    12
```

```
7    11
```

```
8    13
```

```
9    17
```

```
dtype: int32
```

Q10

```
import pandas as pd
```

```
# Create the Series with the number of students in each section
```

```
data_s10 = {'A': 39, 'B': 31, 'C': 32, 'D': 34, 'E': 35}
```

```
s10 = pd.Series(data_s10)
```

```
# Calculate the amount collected by sections A and B
```

```
amount_collected_A = s10['A'] * 100
```

```
amount_collected_B = s10['B'] * 100
```

```
print("Amount collected by Section A:", amount_collected_A)
```

```
print("Amount collected by Section B:", amount_collected_B)
```

```
===== RESIARI: C:/Users/natur/
Amount collected by Section A: 3900
Amount collected by Section B: 3100
```

Q11

```
import pandas as pd
```

```
data = {'Name': ['Alice', 'Bob', 'Charlie'], 'Weight': [65, 70, 75], 'Age': [25, 30, 35]}
```

```
df = pd.DataFrame(data)
```

```
print("Original DataFrame:")
```

```
print(df)
```

```
print("\nDataFrame Transpose:")
```

```
print(df.T)
```

```
new_data = {'Name': ['David', 'Emily', 'Frank', 'Grace', 'Henry'], 'Weight': [80, 85, 90, 95, 100],
'Age': [40, 45, 50, 55, 60]}
```

```
df = df.append(pd.DataFrame(new_data), ignore_index=True)
```

```
df.rename(columns={'Weight': 'Wgt'}, inplace=True)
```

```
print("\nIndex Names:")
```

```
print(df.index)
```

```
print("\nColumn Names:")
```

```
print(df.columns)
```

```
print("\nTotal Amount of Data:")
```

```
print(df.size)
```

```
df.sort_values(by='Age', inplace=True)
```

```
print("\nSorted DataFrame based on Age:")
```

```
print(df)
```

```
===== RESIARI: C:/User
```

```
Original DataFrame:
```

	Name	Weight	Age
0	Alice	65	25
1	Bob	70	30
2	Charlie	75	35

```
DataFrame Transpose:
```

	0	1	2
Name	Alice	Bob	Charlie
Weight	65	70	75
Age	25	30	35

Q12

```
import pandas as pd
```

```
data = {'Name': ['Alice', 'Bob', 'Charlie', 'David', 'Emily'],
```

```
        'Age': [20, 21, 22, 23, 24],
```

```
        'Weight': [55, 60, 65, 70, 75]}
```

```
df = pd.DataFrame(data)
```

```
print("Initial DataFrame:")
```

```
print(df.head())
```

```
print("\nWeight of first and fourth rows before modification:")
```

```
print("First Row Weight:", df.loc[0, 'Weight'])
```

```
print("Fourth Row Weight:", df.loc[3, 'Weight'])
```

```
df.loc[0, 'Weight'] = 58 # Modify weight of student in the first row
```

```
df.loc[3, 'Weight'] = 73 # Modify weight of student in the fourth row
```

```
print("\nWeight of first and fourth rows after modification:")
```

```
print("First Row Weight:", df.loc[0, 'Weight'])
```

```
print("Fourth Row Weight:", df.loc[3, 'Weight'])
```

```
Initial DataFrame:
```

	Name	Age	Weight
0	Alice	20	55
1	Bob	21	60
2	Charlie	22	65
3	David	23	70
4	Emily	24	75

```
Weight of first and fourth rows before modification:
```

```
First Row Weight: 55
```

```
Fourth Row Weight: 70
```

```
Weight of first and fourth rows after modification:
```

```
First Row Weight: 58
```

```
Fourth Row Weight: 73
```

Q13

```
import pandas as pd
```

```
import numpy as np
```

```
np.random.seed(0) # For reproducibility
```

```
num_orders = 1000
```

```
order_numbers = np.arange(1, num_orders + 1)
```

```
order_amounts = np.random.randint(10, 500, size=num_orders)
```

```
payment_methods = np.random.choice(['Credit Card', 'Debit Card', 'Net Banking', 'Cash on Delivery'], size=num_orders)
```

```
ecommerce_df = pd.DataFrame({'Order Number': order_numbers, 'Order Amount':  
order_amounts, 'Payment Method': payment_methods})
```

```
print("Sample E-Commerce DataFrame:")
```

```
print(ecommerce_df.head())
```

```
mean_order_amount = ecommerce_df['Order Amount'].mean()
```

```
mode_order_amount = ecommerce_df['Order Amount'].mode()[0] # Mode can have multiple  
values, so we choose the first one
```

```
median_order_amount = ecommerce_df['Order Amount'].median()
```

```
print("\nMean Order Amount:", mean_order_amount)
```

```
print("Mode Order Amount:", mode_order_amount)
```

```
print("Median Order Amount:", median_order_amount)
```

```

Sample E-Commerce DataFrame:
   Order Number  Order Amount Payment Method
0             1           182      Debit Card
1             2            57      Net Banking
2             3           127      Net Banking
3             4           202      Debit Card
4             5           333      Net Banking

Mean Order Amount: 253.067
Mode Order Amount: 207
Median Order Amount: 249.5

```

Q14

```
import pandas as pd
```

```
csv_data = '''StudentID,Name,Age,Grade
```

```
1,Alice,20,A
```

```
2,Bob,21,B
```

```
3,Charlie,22,C
```

```
4,David,23,A
```

```
5,Emily,24,B
```

```
6,Frank,25,C
```

```
7,Grace,26,A
```

```
8,Henry,27,B
```

```
9,Ivy,28,C
```

```
10,Jack,29,A'''
```

```
with open('student_data.csv', 'w') as f:
```

```
    f.write(csv_data)
```

```
df = pd.read_csv('student_data.csv')
```

```
print("Student Data DataFrame:")
```

```
print(df)
```

```
print("\nStatistics of Student Data:")
print(df.describe())
```

```
Student Data DataFrame:
  StudentID  Name  Age  Grade
0         1  Alice   20     A
1         2   Bob   21     B
2         3 Charlie   22     C
3         4  David   23     A
4         5  Emily   24     B
5         6  Frank   25     C
6         7  Grace   26     A
7         8  Henry   27     B
8         9   Ivy   28     C
9        10   Jack   29     A

Statistics of Student Data:
      StudentID      Age
count  10.00000  10.00000
mean     5.50000  24.50000
std     3.02765   3.02765
min     1.00000  20.00000
25%     3.25000  22.25000
50%     5.50000  24.50000
75%     7.75000  26.75000
max    10.00000  29.00000
```

Q15

```
import pandas as pd
```

```
# Sample DataFrame QtrSales (replace this with your actual DataFrame)
```

```
data = {'Category': ['Electronics', 'Clothing', 'Electronics', 'Clothing', 'Grocery', 'Grocery'],
        'Item Name': ['Laptop', 'T-Shirt', 'Smartphone', 'Jeans', 'Fruits', 'Vegetables'],
        'Expenditure': [1500, 200, 1000, 300, 150, 200]}
```

```
QtrSales = pd.DataFrame(data)
```

```
# Group by Category and calculate average expenditure
```

```
avg_expenditure_per_category = QtrSales.groupby('Category')['Expenditure'].mean()
```

```
# Print average expenditure per category
```

```
print("Average Expenditure per Category:")
```

```
print(avg_expenditure_per_category)
```

```
Average Expenditure per Category:
Category
Clothing      250.0
Electronics  1250.0
Grocery       175.0
Name: Expenditure, dtype: float64
```

Q16

```
import pandas as pd
```

```
data = {'Category': ['Electronics', 'Electronics', 'Clothing', 'Clothing', 'Electronics', 'Clothing'],
        'Item Name': ['Laptop', 'Smartphone', 'T-Shirt', 'Jeans', 'Tablet', 'Dress'],
        'Price': [1500, 1000, 200, 300, 800, 400],
        'Quantity': [5, 3, 10, 7, 2, 5]}
```

```
df = pd.DataFrame(data)
```

```
pivot_df = df.pivot(index='Category', columns='Item Name', values='Quantity')
```

```
pivot_table_df = df.pivot_table(index='Category', columns='Item Name', values='Price',
aggfunc='mean')
```

```
print("DataFrame after pivot():")
```

```
print(pivot_df)
```

```
print("\nDataFrame after pivot_table():")  
print(pivot_table_df)
```

```
DataFrame after pivot():
```

Item Name	Dress	Jeans	Laptop	Smartphone	T-Shirt	Tablet
Clothing	5.0	7.0	NaN	NaN	10.0	NaN
Electronics	NaN	NaN	5.0	3.0	NaN	2.0

```
DataFrame after pivot_table():
```

Item Name	Dress	Jeans	Laptop	Smartphone	T-Shirt	Tablet
Clothing	400.0	300.0	NaN	NaN	200.0	NaN
Electronics	NaN	NaN	1500.0	1000.0	NaN	800.0

Q17

```
import pandas as pd
```

Sample DataFrame

```
data = {'Category': ['Electronics', 'Electronics', 'Clothing', 'Clothing', 'Electronics', 'Clothing'],
        'Item Name': ['Laptop', 'Smartphone', 'T-Shirt', 'Jeans', 'Tablet', 'Dress'],
        'Price': [1500, 1000, 200, 300, 800, 400],
        'Quantity': [5, 3, 10, 7, 2, 5]}
```

```
df = pd.DataFrame(data)
```

```
# Calculate mean absolute deviation (MAD) using apply() and lambda function
```

```
mad = df['Price'].apply(lambda x: abs(x - df['Price'].mean())).mean()
```

```
# Display the MAD
```

```
print("Mean Absolute Deviation (MAD):", mad)
```

Mean Absolute Deviation (MAD): 400.0

Q18

```
import pandas as pd
```

```
import numpy as np
```

```
data = {'EmployeeID': [101, 102, 103, 104, 105],
        'Name': ['Alice', 'Bob', 'Charlie', 'David', 'Emily'],
        'Salary': [50000, 60000, 70000, 80000, 90000],
        'Experience': [3, 5, 8, 4, 6]}
df = pd.DataFrame(data)
```

```
quartiles = np.percentile(df['Salary'], [25, 50, 75])
```

```
variance = df['Experience'].var()
```

```
print("Employee Data DataFrame:")
```

```
print(df)
```

```
print("\nQuartiles of Salary:")
```

```
print("25th Percentile (Q1):", quartiles[0])
```

```
print("50th Percentile (Median):",
      quartiles[1])
```

```
print("75th Percentile (Q3):", quartiles[2])
```

```
print("\nVariance of Experience:", variance)
```

```
Employee Data DataFrame:
  EmployeeID  Name  Salary  Experience
0         101  Alice   50000          3
1         102   Bob   60000          5
2         103 Charlie   70000          8
3         104  David   80000          4
4         105  Emily   90000          6
```

```
Quartiles of Salary:
25th Percentile (Q1): 60000.0
50th Percentile (Median): 70000.0
75th Percentile (Q3): 80000.0
```

```
Variance of Experience: 3.7
```

Q19

```
import pandas as pd
```

```
import numpy as np
```

```
# Generate random data
```

```
np.random.seed(0) # For reproducibility
```

```
data = np.random.normal(loc=0, scale=1, size=1000) # Generate 1000 random numbers from a normal distribution
```

```
# Create a Pandas Series from the random data
```

```
random_series = pd.Series(data)
```

```
# Calculate skewness
```

```
skewness = random_series.skew()
```

```
# Display the skewness
```

```
print("Skewness of Random Data:", skewness)
```

```
===== RESIARI: C:/Users/natur/OneDrive
Skewness of Random Data: 0.03390983920295854
```

Q20

```
import pandas as pd
```

```
import numpy as np
```

```
# Generate random data
```

```
np.random.seed(0) # For reproducibility
```



```
data = np.random.normal(loc=0, scale=1, size=1000) # Generate 1000 random numbers from a normal distribution
```

```
# Create a DataFrame from the random data
```

```
df = pd.DataFrame({'RandomData': data})
```

```
# Compute kurtosis
```

```
kurtosis = df['RandomData'].kurtosis()
```

```
# Display the kurtosis
```

```
print("Kurtosis of Random Data:", kurtosis)
```

```
===== RESTART: C:/Users/natur/OneDrive  
Kurtosis of Random Data: -0.04097691643266943
```

Q21

21.2: Apply aggregate function to find count, min , max and sum .

➤ SOURCE CODE:-

```
import pandas as pd
df=pd.read_csv("/content/crime.csv")
#print(df)
aggregate_functions = {
    'Year': ['count'],
    'Murder': ['min', 'max', 'sum'],
    'Robbery': ['min', 'max', 'sum'],
    'Arson': ['min', 'max', 'sum']
}
aggregate_result = df.agg(aggregate_functions)
print("\nAggregate function results:")
print(aggregate_result)
```

➤ OUTPUT

```
Aggregate function results:
      Year  Murder  Robbery  Arson
count  420.0     NaN      NaN    NaN
min      NaN      0.0      0.0      0.0
max      NaN    423.0     83.0   178.0
sum      NaN  7900.0    953.0  2717.0
```

21.2: Apply aggregate function to find count, min , max and sum .**➤ SOURCE CODE:-**

```
import pandas as pd
df=pd.read_csv("/content/crime.csv")
#print(df)
aggregate_functions = {
    'Year': ['count'],
    'Murder': ['min', 'max', 'sum'],
    'Robbery': ['min', 'max', 'sum'],
    'Arson': ['min', 'max', 'sum']
}
aggregate_result = df.agg(aggregate_functions)
print("\nAggregate function results:")
print(aggregate_result)
```

➤ OUTPUT

Aggregate function results:

	Year	Murder	Robbery	Arson
count	420.0	NaN	NaN	NaN
min	NaN	0.0	0.0	0.0
max	NaN	423.0	83.0	178.0
sum	NaN	7900.0	953.0	2717.0

21.5: Create a pie chart for all states

➤ SOURCE CODE:-

```
import pandas as pd
import matplotlib.pyplot as plt
df=pd.read_csv("/content/crime.csv")
#print(df)
state_counts = df['State'].value_counts()
plt.figure(figsize=(10, 8))
state_counts.plot(kind='pie', autopct='%1.1f%%')
plt.title('Crime distribution by state')
plt.ylabel("")
plt.show()
```

➤ OUTPUT

21.6: Plot the no. of murders state-wise

➤ SOURCE CODE:-

```
import pandas as pd
import matplotlib.pyplot as plt
df=pd.read_csv("/content/crime.csv")
#print(df)
murder_by_state = df.groupby('STATE/UT')['Murder'].sum()
murder_by_state.plot(kind='bar', figsize=(10, 6))
plt.title('Number of murders by state')
plt.xlabel('State')
plt.ylabel('Number of murders')
plt.show()
```

➤ OUTPUT

21.7: Check the data type of all column**➤ SOURCE CODE:-**

```
import pandas as pd
import matplotlib.pyplot as plt
df=pd.read_csv("/content/crime.csv")
#print(df)
# Check data type of all columns
print("\nData types of all columns:")
print(df.dtypes)
```

➤ OUTPUT

```
Data types of all columns:
STATE/UT          object
Year             float64
Murder            float64
Assault on women  float64
Kidnapping and Abduction  float64
Dacoity           float64
Robbery           float64
Arson             float64
Hurt              float64
Prevention of atrocities (POA) Act  float64
Protection of Civil Rights (PCR) Act float64
Other Crimes Against SCs  float64
dtype: object
```

21.8: Check the robbery column for NULL values.

if exist replace with 0

21.9: Select the data set where robbery greater than 10 and year= 2001

➤ SOURCE CODE:-

```
import pandas as pd
df=pd.read_csv("/content/crime.csv")
#print(df)
print(df.describe())
```

➤ OUTPUT

	Year	Murder	Assault on women	Kidnapping and Abduction	\
count	420.000000	420.000000	420.000000	420.000000	
mean	2006.500000	18.009524	37.897619	11.138095	
std	3.456160	51.543442	70.435008	34.046327	
min	2001.000000	0.000000	0.000000	0.000000	
25%	2003.750000	0.000000	0.000000	0.000000	
50%	2006.500000	1.000000	1.000000	0.000000	
75%	2009.250000	14.000000	34.250000	8.000000	
max	2012.000000	423.000000	412.000000	363.000000	

	Dacoity	Robbery	Arson	Hurt	\
count	420.000000	420.000000	420.000000	420.000000	
mean	0.040476	2.260048	6.460048	117.033333	
std	2.701976	6.100314	15.771988	206.612233	
min	0.000000	0.000000	0.000000	0.000000	
25%	0.000000	0.000000	0.000000	0.000000	
50%	0.000000	0.000000	0.000000	3.500000	
75%	0.000000	1.000000	5.000000	108.250000	
max	22.000000	83.000000	178.000000	1252.000000	

	Prevention of atrocities (POA) Act	\
count	420.000000	
mean	296.566667	
std	636.605911	
min	0.000000	
25%	0.000000	
50%	0.000000	
75%	260.250000	
max	4005.000000	

	Protection of Civil Rights (PCR) Act	Other Crimes Against SCs	\
count	420.000000	420.000000	
mean	10.166667	380.219048	
std	25.960971	864.165129	
min	0.000000	0.000000	
25%	0.000000	0.000000	
50%	0.000000	6.000000	
75%	2.000000	280.000000	
max	459.000000	4771.000000	

21.11: Perform sorting on Arson and Robbery

21.12: Perform renaming on two columns

Q22

➤ SOURCE CODE:-

```
import pandas as pd
df = pd.read_csv("/content/investigating clinical data.csv")
print(df)
```

➤ OUTPUT:-

	PatientId	AppointmentID	Gender	ScheduledDay
0	2.987250e+13	5642903	F	2016-04-29T18:38:08Z
1	5.589978e+14	5642503	M	2016-04-29T16:08:27Z
2	4.262962e+12	5642549	F	2016-04-29T16:19:04Z
3	8.679512e+11	5642828	F	2016-04-29T17:29:31Z
4	8.841186e+12	5642494	F	2016-04-29T16:07:23Z
...
110522	2.572134e+12	5651768	F	2016-05-03T09:15:35Z
110523	3.596266e+12	5650093	F	2016-05-03T07:27:33Z
110524	1.557663e+13	5630692	F	2016-04-27T16:03:52Z
110525	9.213493e+13	5630323	F	2016-04-27T15:09:23Z
110526	3.775115e+14	5629448	F	2016-04-27T13:30:56Z

	AppointmentDay	Age	Neighbourhood	Scholarship
0	2016-04-29T00:00:00Z	62	JARDIM DA PENHA	0
1	2016-04-29T00:00:00Z	56	JARDIM DA PENHA	0
2	2016-04-29T00:00:00Z	62	MATA DA PRAIA	0
3	2016-04-29T00:00:00Z	8	PONTAL DE CÂMBURI	0
4	2016-04-29T00:00:00Z	56	JARDIM DA PENHA	0
...
110522	2016-06-07T00:00:00Z	56	MARIA ORTIZ	0
110523	2016-06-07T00:00:00Z	51	MARIA ORTIZ	0
110524	2016-06-07T00:00:00Z	21	MARIA ORTIZ	0
110525	2016-06-07T00:00:00Z	38	MARIA ORTIZ	0
110526	2016-06-07T00:00:00Z	54	MARIA ORTIZ	0

	Hipertension	Diabetes	Alcoholism	Handicap	SMS_received	No-show
0	1	0	0	0	0	No
1	0	0	0	0	0	No
2	0	0	0	0	0	No
3	0	0	0	0	0	No
4	1	1	0	0	0	No
...
110522	0	0	0	0	1	No
110523	0	0	0	0	1	No
110524	0	0	0	0	1	No
110525	0	0	0	0	1	No
110526	0	0	0	0	1	No

22.1: describe the dataset

```
import pandas as pd
df = pd.read_csv("/content/investigating clinical data.csv")
df.describe()
```

➤ **OUTPUT:-**

	PatientId	AppointmentID	Age	Scholarship	Hipertension	Diabetes	Alcoholism	Handcap	SMS_received
count	1.105270e+06	1.105270e+06	1.10527.000000	110527.000000	110527.000000	110527.000000	110527.000000	110527.000000	110527.000000
mean	1.474903e+14	5.075305e+06	37.088874	0.098266	0.197246	0.071865	0.093400	0.022248	0.321028
std	2.580640e+14	7.128575e+04	23.110205	0.297875	0.387921	0.258285	0.171688	0.181543	0.488873
min	3.821704e+04	5.030230e+06	-1.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
25%	4.172614e+12	5.840208e+06	18.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
50%	3.175184e+13	5.680573e+06	37.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
75%	9.439172e+13	5.725524e+06	56.000000	0.000000	0.000000	0.000000	0.000000	0.000000	1.000000
max	9.899810e+14	5.790404e+06	115.000000	1.000000	1.000000	1.000000	1.000000	4.000000	1.000000

22.2: Perform renaming on two columns

➤ **SOURCE CODE:-**

```
import pandas as pd
df = pd.read_csv("/content/investigating clinical data.csv")
df.rename(columns={"PatientId": "PatientID_NUM", "Neighbourhood": "Address"},
          inplace=True)
print(df.columns)
```

➤ **OUTPUT:-**

```
Index(['PatientId', 'AppointmentID', 'Gender', 'ScheduledDay',
      'AppointmentDay', 'Age', 'Neighbourhood', 'Scholarship', 'Hipertension',
      'Diabetes', 'Alcoholism', 'Handcap', 'SMS_received', 'No-show'],
      dtype='object')
```

```
Index(['PatientID_NUM', 'AppointmentID', 'Gender', 'ScheduledDay',
      'AppointmentDay', 'Age', 'Address', 'Scholarship', 'Hipertension',
      'Diabetes', 'Alcoholism', 'Handcap', 'SMS_received', 'No-show'],
      dtype='object')
```


➤ SOURCE CODE:-

```
import pandas as pd
import matplotlib.pyplot as plt

df = pd.read_csv("/content/investigating clinical data.csv")

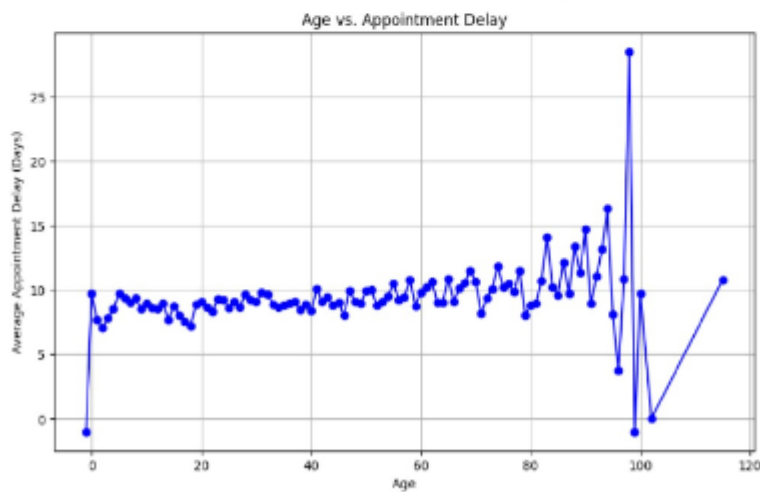
df['AppointmentDay'] = pd.to_datetime(df['AppointmentDay'])
df['ScheduledDay'] = pd.to_datetime(df['ScheduledDay'])

df['AppointmentDelayDays'] = (df['AppointmentDay'] - df['ScheduledDay']).dt.days

age_delay = df.groupby('Age')['AppointmentDelayDays'].mean()

plt.figure(figsize=(10, 6))
plt.plot(age_delay.index, age_delay.values, marker='o', linestyle='-', color='b')
plt.xlabel('Age')
plt.ylabel('Average Appointment Delay (Days)')
plt.title('Age vs. Appointment Delay')
plt.grid(True)
plt.show()
```

➤ OUTPUT:-



➤ SOURCE CODE:-

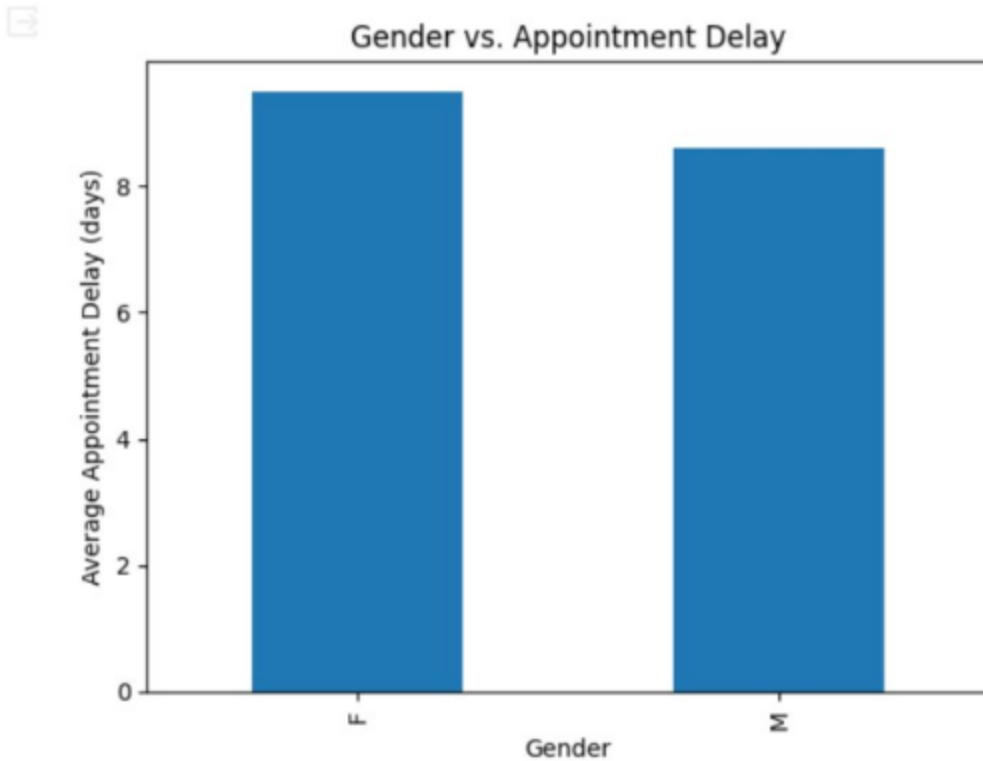
```
import pandas as pd
import matplotlib.pyplot as plt

df = pd.read_csv("/content/investigating clinical data.csv")

df['AppointmentDay'] = pd.to_datetime(df['AppointmentDay'])
df['ScheduledDay'] = pd.to_datetime(df['ScheduledDay'])

df['AppointmentDelayDays'] = (df['AppointmentDay'] - df['ScheduledDay']).dt.days

df.groupby("Gender")["AppointmentDelayDays"].mean().plot(kind="bar")
plt.xlabel("Gender")
plt.ylabel("Average Appointment Delay (days)")
plt.title("Gender vs. Appointment Delay")
plt.show()
```

➤ OUTPUT:-

22.5: Do people delay their appointments when they have no scholarship? (Show by plotting)

➤ SOURCE CODE:-

```
import pandas as pd
import matplotlib.pyplot as plt

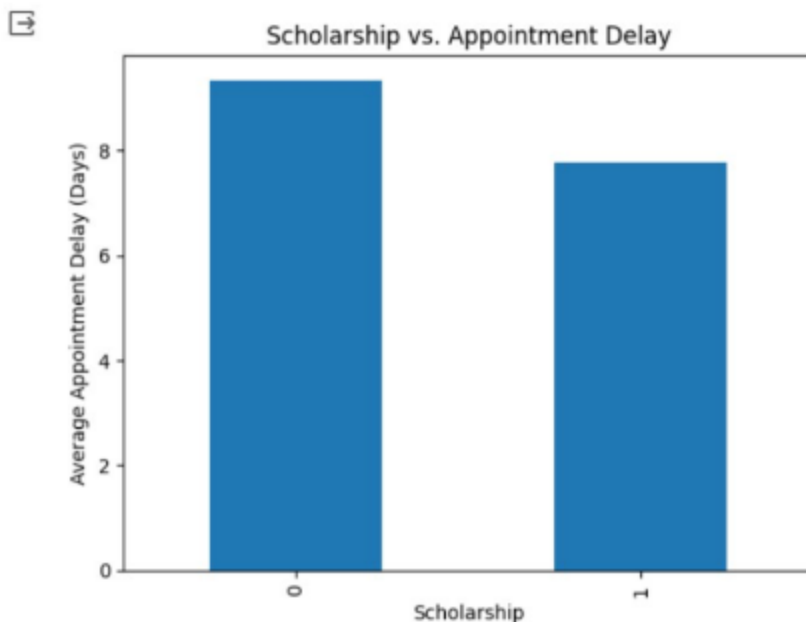
df = pd.read_csv("/content/investigating clinical data.csv")

df['AppointmentDay'] = pd.to_datetime(df['AppointmentDay'])
df['ScheduledDay'] = pd.to_datetime(df['ScheduledDay'])

df['AppointmentDelayDays'] = (df['AppointmentDay'] - df['ScheduledDay']).dt.days

df.groupby("Scholarship")["AppointmentDelayDays"].mean().plot(kind="bar")
plt.xlabel("Scholarship")
plt.ylabel("Average Appointment Delay (Days)")
plt.title("Scholarship vs. Appointment Delay")
plt.show()
```

➤ OUTPUT:-



22.6: Is there any relationship between gender having scholarships and delaying the appointment? (show by plotting)

➤ SOURCE CODE:-

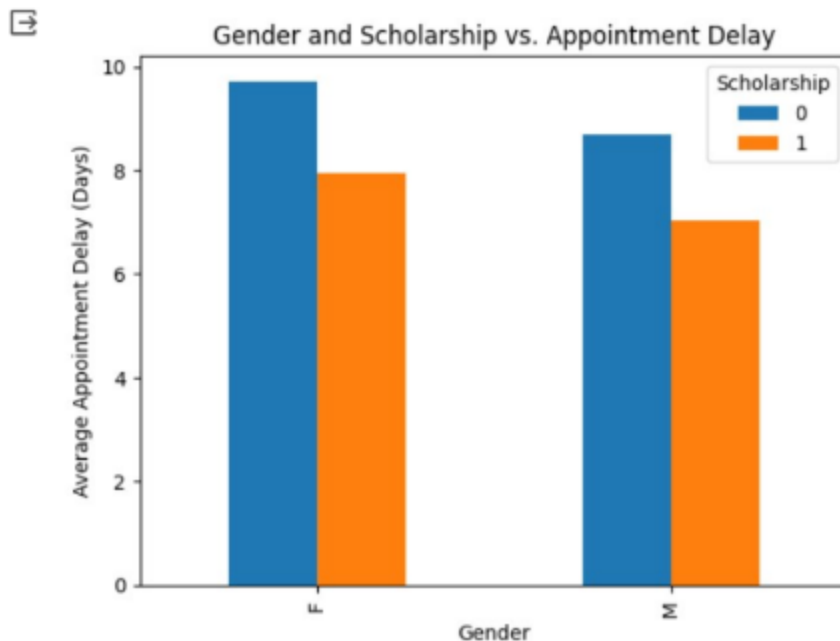
```
import pandas as pd
import matplotlib.pyplot as plt

df = pd.read_csv("/content/investigating clinical data.csv")

df['AppointmentDay'] = pd.to_datetime(df['AppointmentDay'])
df['ScheduledDay'] = pd.to_datetime(df['ScheduledDay'])

df['AppointmentDelayDays'] = (df['AppointmentDay'] - df['ScheduledDay']).dt.days
df.groupby(["Gender", "Scholarship"])
["AppointmentDelayDays"].mean().unstack().plot(kind="bar")
plt.xlabel("Gender")
plt.ylabel("Average Appointment Delay (Days)")
plt.title("Gender and Scholarship vs. Appointment Delay")
plt.show()
```

➤ OUTPUT:-



22.7: Which gender has more appointments in which illness? (Show by plotting)

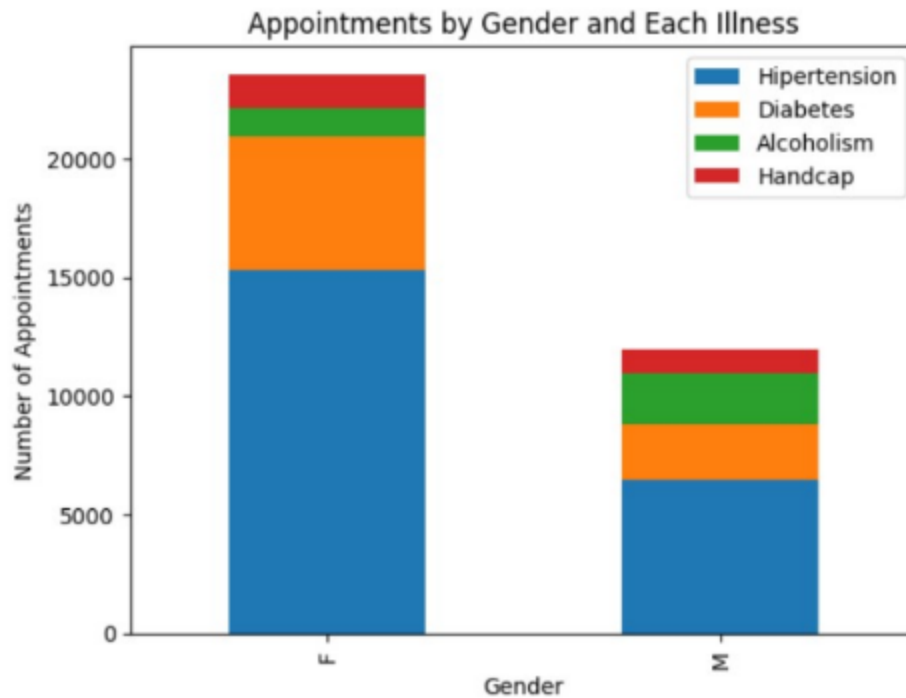
➤ SOURCE CODE:-

```
import pandas as pd
import matplotlib.pyplot as plt
df = pd.read_csv("/content/investigating clinical data.csv")

grouped_df = df.groupby("Gender")[illness_columns].sum()

grouped_df.plot(kind="bar", stacked=True)
plt.xlabel("Gender")
plt.ylabel("Number of Appointments")
plt.title("Appointments by Gender and Each Illness")
plt.show()
```

➤ OUTPUT:-



22.8: Does a person suffering from alcoholism tend to delay appointments? (Show by plotting)

➤ SOURCE CODE:-

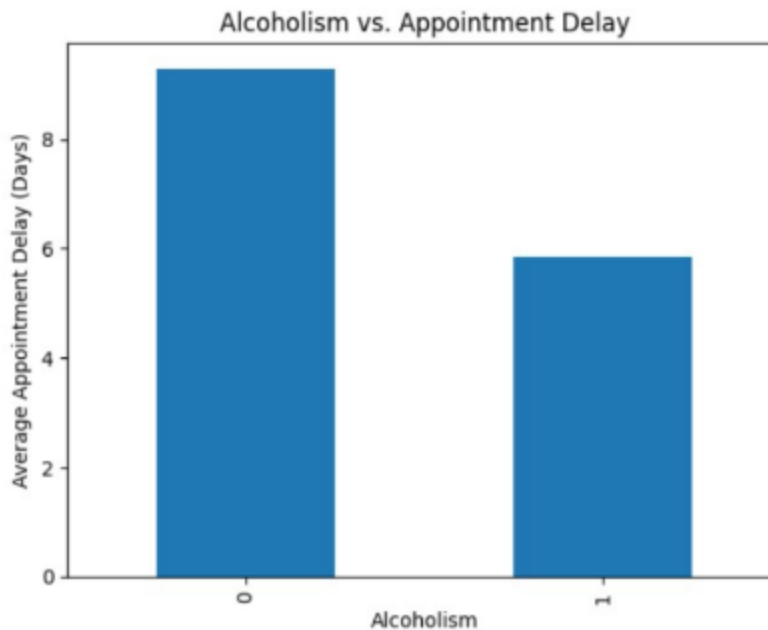
```
import pandas as pd
import matplotlib.pyplot as plt

df = pd.read_csv("/content/investigating clinical data.csv")

df['AppointmentDay'] = pd.to_datetime(df['AppointmentDay'])
df['ScheduledDay'] = pd.to_datetime(df['ScheduledDay'])

df['AppointmentDelayDays'] = (df['AppointmentDay'] - df['ScheduledDay']).dt.days
df.groupby("Alcoholism")["AppointmentDelayDays"].mean().plot(kind="bar")
plt.xlabel("Alcoholism")
plt.ylabel("Average Appointment Delay (Days)")
plt.title("Alcoholism vs. Appointment Delay")
plt.show()
```

➤ OUTPUT:-



22.9: Perform group by and count on neighborhood

➤ SOURCE CODE:-

```
import pandas as pd
df = pd.read_csv("/content/investigating clinical data.csv")
neighborhood_counts = df.groupby("Neighbourhood").size()
print(neighborhood_counts)
```

➤ OUTPUT:-

Neighbourhood	
AEROPORTO	8
ANDORINHAS	2262
ANTÔNIO HONÓRIO	271
ARIOVALDO FAVALESSA	282
BARRO VERMELHO	423
...	
SÃO JOSÉ	1977
SÃO PEDRO	2448
TABUAZEIRO	3132
UNIVERSITÁRIO	152
VILA RUBIM	851
Length: 81, dtype: int64	

22.10: Perform agg() function on dataset.

➤ SOURCE CODE:-

```
import pandas as pd
df = pd.read_csv("/content/investigating clinical data.csv")
df.agg({"PatientId": "median", "Age": "sum", "AppointmentID": "mean", "Scholarship":
"min", "Hipertension": "max"})
```

➤ OUTPUT:-

PatientId	3.173184e+13
Age	4.099322e+06
AppointmentID	5.675305e+06
Scholarship	0.000000e+00
Hipertension	1.000000e+00
dtype: float64	

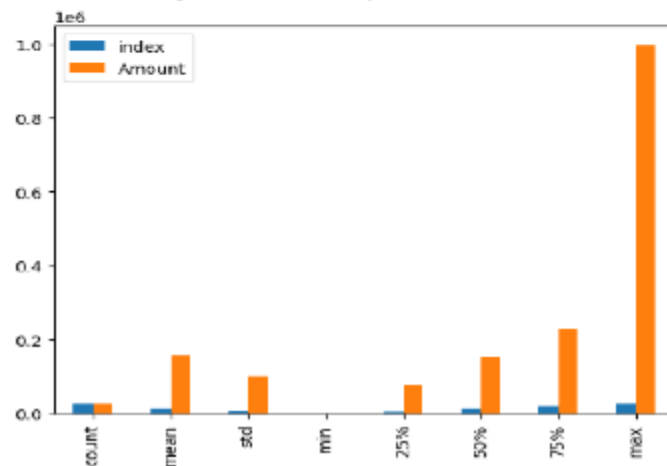
Q23

➤ SOURCE CODE:-

```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
data=pd.read_csv("/content/credit.csv")
#print(data)
print(data.head())
data.describe().plot(kind='bar')
plt.show()
```

➤ OUTPUT

	index	City	Date	Card Type	Exp Type	Gender	Amount
0	0	Delhi, India	29-Oct-14	Gold	Bills	F	82475
1	1	Greater Mumbai, India	22-Aug-14	Platinum	Bills	F	32555
2	2	Bengaluru, India	27-Aug-14	Silver	Bills	F	101738
3	3	Greater Mumbai, India	12-Apr-14	Signature	Bills	F	123424
4	4	Bengaluru, India	05-May-15	Gold	Bills	F	171574



23.2: Display pivot table to create a summary for the total amount spent**a) month b) city****(b) City****➤ SOURCE CODE:-**

```
import pandas as pd
import numpy as np
data=pd.read_csv("/content/credit.csv")
#print(data)
pivot_city = data.pivot_table(index='City', values='Amount', aggfunc='sum')
print("\nTotal amount spent by city:")
print(pivot_city)
```

➤ OUTPUT

```
Total amount spent by city:
      Amount
City
Achalpur, India    1606641
Adilabad, India    1769464
Adityapur, India     963993
Adoni, India       1575355
Adoor, India        647725
...
Zaidpur, India      723818
Zamania, India      865634
Zira, India         1640547
Zirakpur, India     549663
Zunheboto, India    466429

[986 rows x 1 columns]
```

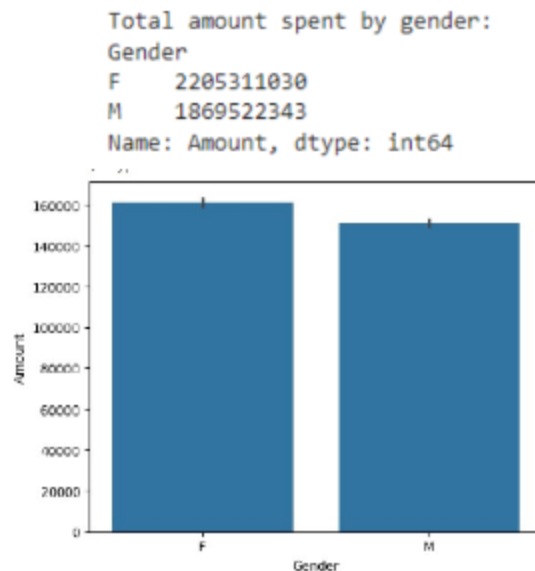
23.3: Analyze the impact of gender to study consumer behavior

➤ SOURCE CODE:-

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
data=pd.read_csv("/content/credit.csv")
#print(data)
gender_analysis = data.groupby('Gender')['Amount'].sum()

print("\nTotal amount spent by gender:")
print(gender_analysis)
sns.barplot(x='Gender', y='Amount', data=data)
plt.show()
```

➤ OUTPUT



23.4: Check for NULL values in columns.If they exist replace them with appropriate values.

➤ **SOURCE CODE:-**

```
import pandas as pd
import numpy as np
data=pd.read_csv("/content/credit.csv")
#print(data)
print("\nNULL values in columns:")
print(data.isnull().sum())
data['Amount'].fillna(0, inplace=True)
print("\nAfter replacing NULL values:")
print(data.isnull().sum())
```

➤ **OUTPUT**

```
NULL values in columns:
index      0
City       0
Date       0
Card Type  0
Exp Type   0
Gender     0
Amount     0
dtype: int64

After replacing NULL values:
index      0
City       0
Date       0
Card Type  0
Exp Type   0
Gender     0
Amount     0
dtype: int64
```

23.5: Analyze the relationship type between expense type and amount.**➤ SOURCE CODE:-**

```

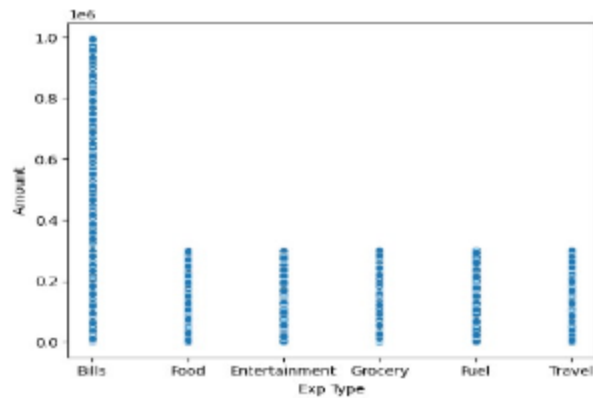
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
data=pd.read_csv("/content/credit.csv")
#print(data)
expense_groups = data.groupby('Exp Type')
print(expense_groups['Amount'].describe())
sns.scatterplot(x='Exp Type', y='Amount', data=data)
plt.show()

```

➤ OUTPUT

	count	mean	std	min	25%
Exp Type					
Bills	5078.0	178627.899370	151893.760199	1026.0	78831.75
Entertainment	4762.0	152548.831583	86628.587025	1061.0	77749.25
Food	5463.0	158965.485272	86492.589170	1018.0	75862.50
Fuel	5257.0	158111.436371	85929.342308	1038.0	77405.00
Grocery	4754.0	151074.447413	86440.356185	1005.0	75648.25
Travel	738.0	148042.833333	86627.487531	1070.0	73909.00

	50%	75%	max
Exp Type			
Bills	160551.0	237895.50	998077.0
Entertainment	153094.5	228414.75	299930.0
Food	151679.0	225387.50	299837.0
Fuel	149629.0	224480.00	299905.0
Grocery	152157.5	225550.50	299920.0
Travel	146947.0	221778.75	299618.0



23.6: Analyze the spending habits by city and gender.**► SOURCE CODE:-**

```
import pandas as pd
import numpy as np
data=pd.read_csv("/content/credit.csv")
#print(data)
city_gender_groups = data.groupby(['City', 'Gender'])
print(city_gender_groups['Amount'].describe())
```

► OUTPUT

		count	mean	std	min	\
City	Gender					
Achalpur, India	F	5.0	172195.200000	66849.453788	55347.0	
	M	4.0	186416.250000	43066.249789	138246.0	
Adilabad, India	F	5.0	232886.000000	65219.130767	122336.0	
	M	5.0	121006.800000	61118.426986	49020.0	
Adityapur, India	F	2.0	253386.500000	61599.607243	209829.0	
...		
Zira, India	M	7.0	178522.714286	81757.430940	48909.0	
Zirakpur, India	F	1.0	289172.000000	NaN	289172.0	
	M	2.0	130245.500000	64092.865754	84925.0	
Zunheboto, India	F	1.0	78591.000000	NaN	78591.0	
	M	2.0	193919.000000	147994.620875	89271.0	
		25%	50%	75%	max	
City	Gender					
Achalpur, India	F	178612.00	201032.0	209805.00	216180.0	
	M	164563.50	182957.5	204810.25	241504.0	
Adilabad, India	F	227325.00	258225.0	278085.00	278459.0	
	M	80193.00	108313.0	178349.00	189159.0	
Adityapur, India	F	231607.75	253386.5	275165.25	296944.0	
...		
Zira, India	M	138824.00	190358.0	219226.50	294291.0	
Zirakpur, India	F	289172.00	289172.0	289172.00	289172.0	
	M	107585.25	130245.5	152905.75	175566.0	
Zunheboto, India	F	78591.00	78591.0	78591.00	78591.0	
	M	141595.00	193919.0	246243.00	298567.0	

[1876 rows x 8 columns]

➤ SOURCE CODE:-

```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
from scipy.stats import skew, kurt
import pandas as pd
import matplotlib.pyplot as plt
df=pd.read_csv("/content/crime.csv")
#print(df)
state_counts = df['STATE/UT'].value_counts()
plt.figure(figsize=(10, 8))
state_counts.plot(kind='pie', autopct='%1.1f%%')
plt.title('Crime distribution by state')
plt.ylabel("")
plt.show()
data=pd.read_csv("/content/credit.csv")
#print(data)
data_skewness = skew(data['Amount'])
data_kurtosis = kurtosis(data['Amount'])
print(f'Skewness: {data_skewness}')
print(f'Kurtosis: {data_kurtosis}')
sns.boxplot(x='Gender', y='Amount', data=data)
plt.title('Spending Habits by Gender')
plt.show()
```

➤ OUTPUT

Skewness: 1.7550705661419843

Kurtosis: 10.312972126257737

