```
[4]: #Write a Python program to create a Pie plot to get the frequency
#of the three species of the Iris data (Use 1ris.csv)

# (slip1 and slip11 are same)

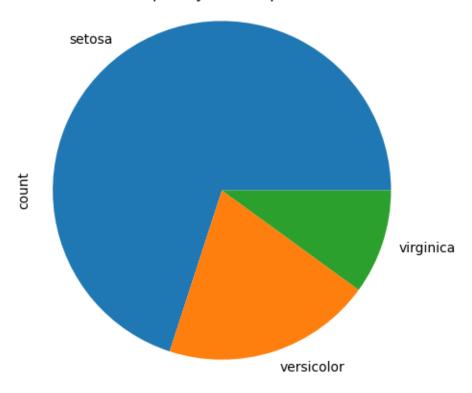
import pandas as pd
import matplotlib.pyplot as plt

d = pd.read_csv('Iris.csv')

ax = plt.figure(figsize=(5, 5))
d['Species'].value_counts().plot.pie()

plt.title('Frequency of Iris Species')
plt.axis('equal')
plt.show()
```

## Frequency of Iris Species



```
[3]: # write a Python program to view basic statistical details of the data
#.(Use wineequality-red.csv)
import pandas as pd

wine_data = pd.read_csv('winequality-red.csv')
print(wine_data.describe())
```

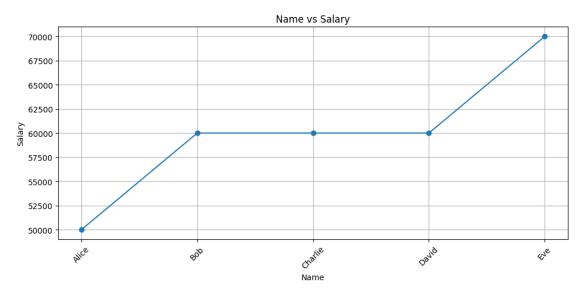
	fixed acidity	volatile acidity	citric acid	residual sugar	\
count	5.000000	5.000000	5.000000	5.000000	
mean	8.420000	0.492000	0.128000	1.940000	
std	1.565886	0.350885	0.242322	0.427785	
min	7.400000	0.000000	0.000000	1.400000	
25%	7.800000	0.280000	0.000000	1.900000	
50%	7.800000	0.600000	0.040000	1.900000	
75%	7.900000	0.700000	0.040000	1.900000	
max	11.200000	0.880000	0.560000	2.600000	
	chlorides fr	ee sulfur dioxide	total sulfur	dioxide density	r \
count	5.000000	5.00000		5.00000 5.000000	)
mean	0.074800	17.00000	5	5.00000 0.997280	)
std	0.003633	5.09902	1	2.60952 0.000576	;
min	0.069000	11.00000	3	4.00000 0.996800	)

```
25%
        0.075000
                              15.00000
                                                    54.00000 0.996800
50%
        0.075000
                             17.00000
                                                    60.00000
                                                              0.997000
75%
        0.076000
                              17.00000
                                                    60.00000 0.997800
        0.079000
                             25.00000
                                                    67.00000 0.998000
max
                 sulphates
                             alcohol
                                        quality
             pН
                    5.0000
                                       5.000000
count
       5.000000
                            5.000000
       3.258000
                    0.6220
                            9.720000
                                       5.200000
mean
std
       0.146697
                    0.0502 0.178885
                                      0.447214
min
       3.160000
                    0.5600
                            9.400000
                                      5.000000
25%
                            9.800000
       3.160000
                    0.5800
                                       5.000000
50%
       3.200000
                    0.6400
                            9.800000
                                       5.000000
75%
       3.260000
                    0.6500
                                       5.000000
                            9.800000
       3.510000
                    0.6800
                            9.800000
                                       6.000000
max
```

[]:

```
[12]: # write a Python program for Handling Missing Value. Replace missing
      # value of salary, age column with mean of that column.
      # (Use Data.csv file)
      # slip 2 and slip6 are same
      import pandas as pd
      df = pd.read_csv('Data.csv')
      print("Original DataFrame:")
      print(df)
      df['salary'] = df['salary'].fillna(df['salary'].mean())
      df['age'] = df['age'].fillna(df['age'].mean())
      print("\nDataFrame after handling missing values:")
      print(df)
     Original DataFrame:
           name
                  salary
                           age
          Alice 50000.0 30.0
     0
     1
            Bob
                     NaN 25.0
     2 Charlie 60000.0
                         {\tt NaN}
     3
          David
                     NaN 35.0
            Eve 70000.0 28.0
     4
     DataFrame after handling missing values:
           name salary
                           age
     0
          Alice 50000.0 30.0
            Bob 60000.0 25.0
     1
     2 Charlie 60000.0 29.5
          David 60000.0 35.0
     3
            Eve 70000.0 28.0
 [4]: # Write a Python program to generate a line plot of name Vs salary
      import pandas as pd
      import matplotlib.pyplot as plt
      df = pd.read_csv('Data_filled.csv')
```

```
# Generate a line plot for name vs salary
plt.figure(figsize=(10, 5))
plt.plot(df['name'], df['salary'], marker='o')
plt.title('Name vs Salary')
plt.xlabel('Name')
plt.ylabel('Salary')
plt.xticks(rotation=45)
plt.grid()
plt.tight_layout()
plt.show()
```



```
[10]: # Download the heights and weights dataset and load the dataset from
# a given csv file into a dataframe. Print the first, last 10 rows
# and random 20 rows also display shape of the dataset.
import pandas as pd

# Load the dataset from the CSV file
df = pd.read_csv('short_heights_weights.csv')

# Print the first 10 rows
print("First 10 rows:")
print(df.head(10))

# Print the last 10 rows
print("\nLast 10 rows:")
print(df.tail(10))
```

```
# Print 20 random rows
print("\nRandom 20 rows:")
print(df.sample(20))

# Display the shape of the dataset
print("\nShape of the dataset:", df.shape)
```

#### First 10 rows:

	Gender	<pre>Height(Inches)</pre>	Weight(Pounds)
0	Male	65	150
1	Female	62	120
2	Male	70	180
3	Female	58	110
4	Male	66	160
5	Female	64	130
6	Male	65	150
7	Female	62	120
8	Male	70	180
9	Female	58	110

#### Last 10 rows:

	Gender	Height(Inches)	Weight(Pounds)
33	Female	62	120
34	Male	70	180
35	Female	58	110
36	Male	66	160
37	Female	64	130
38	Male	65	150
39	Female	62	120
40	Male	70	180
41	Female	58	110
42	Male	66	160

#### Random 20 rows:

	Gender	<pre>Height(Inches)</pre>	Weight(Pounds)
3	Female	58	110
8	Male	70	180
1	Female	62	120
5	Female	64	130
9	Female	58	110
0	Male	65	150
19	Female	62	120
13	Female	62	120
37	Female	64	130
10	Male	66	160
23	Female	64	130
42	Male	66	160

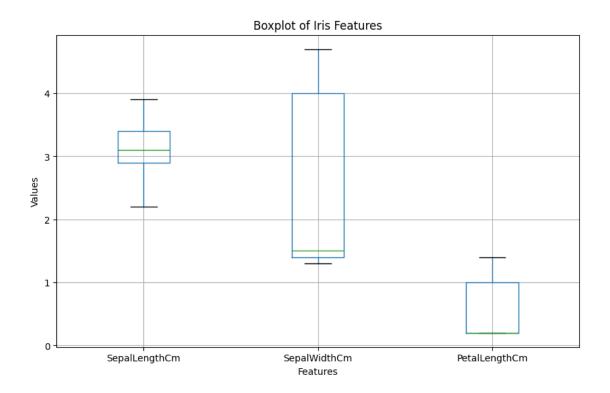
33	Female	62	120
17	Female	64	130
31	Female	64	130
38	Male	65	150
20	Male	70	180
12	Male	65	150
28	Male	66	160
25	Female	62	120

Shape of the dataset: (43, 3)

```
[4]: # Write a Python program to create box plots to see how each feature i.e. Sepalual Length,
# Sepal Width, Petal Length, Petal Width are distributed across the three appeared import pandas as pd import matplotlib.pyplot as plt

d = pd.read_csv('iris3.csv')
new_data = d[["SepalLengthCm", "SepalWidthCm", "PetalLengthCm", "PetalWidthCm"]]

plt.figure(figsize=(10, 6))
new_data.boxplot()
plt.title("Boxplot of Iris Features")
plt.xlabel("Features")
plt.ylabel("Values")
plt.show()
```



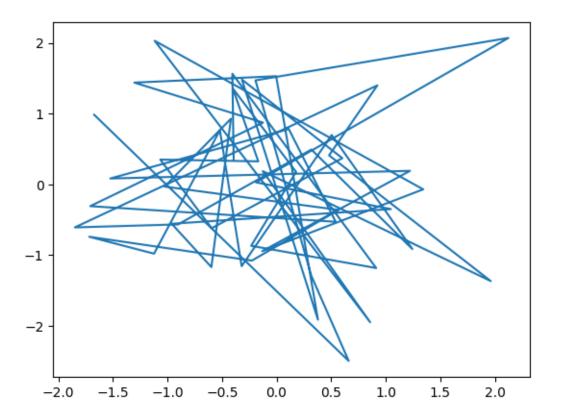
```
[5]: # Write a Python program to view basic statistical details of the data (Use_\)
Heights and
# Weights Dataset
# (similer as slip1 of 2)

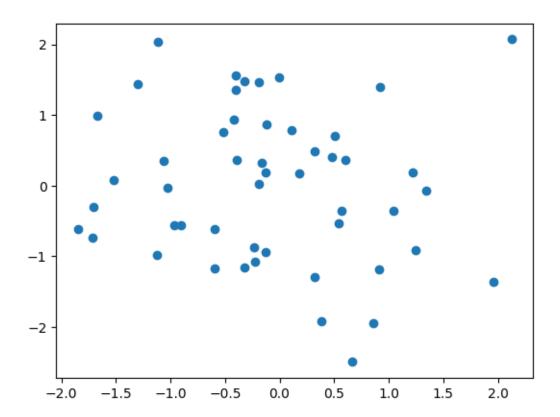
import pandas as pd

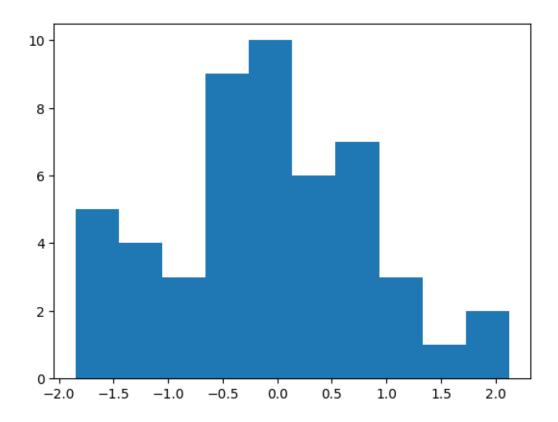
df = pd.read_csv('hw3.csv')
stats = df.describe()
print(stats)
```

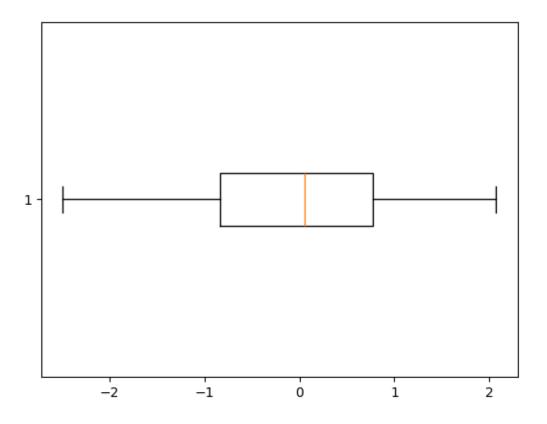
```
HeightCm
                    WeightKg
         5.000000
                    5.000000
count
       170.000000 70.000000
mean
        15.811388
                  15.811388
std
\min
       150.000000
                   50.000000
25%
       160.000000
                   60.000000
50%
       170.000000
                  70.000000
75%
       180.000000
                   80.000000
       190.000000 90.000000
max
```

```
[1]: # Generate a random array of 50 integers and display them using a line chart,
     ⇔scatter plot,
     # histogram and box plot. Apply appropriate color, labels and styling options
     # slip4 and slip 5 is same
     import matplotlib.pyplot as plt
    import numpy as np
     x=np.random.randn(50)
     y=np.random.randn(50)
    plt.plot(x,y)
     plt.show()
    plt.scatter(x,y)
    plt.show()
    plt.hist(x)
     plt.show()
    plt.boxplot(y,vert=False)
    plt.show()
```









```
[2]: # Write a Python program to print the shape, number of rows-columns, data__
     ⇔types, feature
     # names and the description of the data(Use User_Data.csv)
     # (this program also in slip19 of 2 )
    import pandas as pd
    df = pd.read_csv('Data_filled.csv')
    print("Shape of the dataset:", df.shape)
    print("features :",df.info())
    print("\nDescription of the Data:\n", df.describe())
    print("\nData Types:\n", df.dtypes)
    Shape of the dataset: (5, 3)
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 5 entries, 0 to 4
    Data columns (total 3 columns):
         Column Non-Null Count Dtype
                 _____
                 5 non-null
         name
                                 object
     1
         salary 5 non-null
                                 float64
                 5 non-null
                                 float64
     2
         age
    dtypes: float64(2), object(1)
```

memory usage: 252.0+ bytes

features : None

### Description of the Data:

	salary	age
count	5.000000	5.000000
mean	60000.000000	29.500000
std	7071.067812	3.640055
min	50000.000000	25.000000
25%	60000.000000	28.000000
50%	60000.000000	29.500000
75%	60000.000000	30.000000
max	70000.000000	35.000000

### Data Types:

name object salary float64 age float64 dtype: object

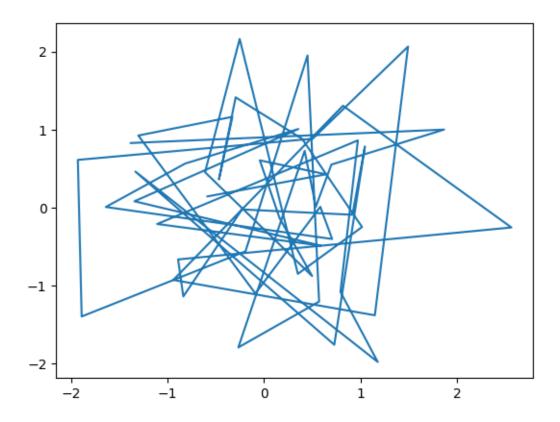
	Purchased	Country_Canada	Country_UK	Country_USA
0	1	False	False	True
1	0	True	False	False
2	1	False	False	True
3	0	False	True	False

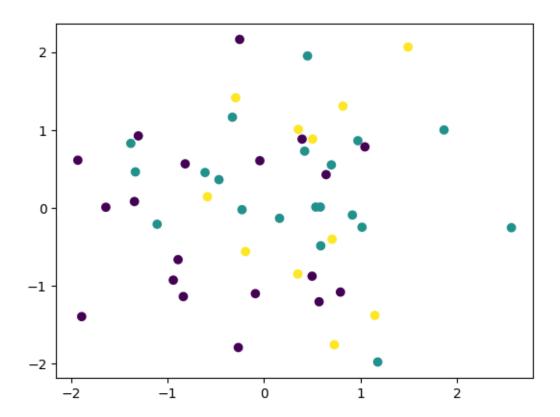
#### October 15, 2024

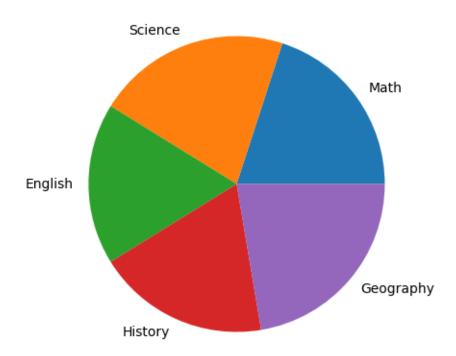
[1]: # Write a program in python to perform following task : [15] Standardizing Data

```
\hookrightarrow (transform
# them into a standard Gaussian distribution with a mean of O and a standard \Box
 ⇔deviation of 1) (Use
# winequality-red.csv)
import pandas as pd
from sklearn.preprocessing import StandardScaler
df = pd.read_csv('winequality-red.csv')
num_cols = df.select_dtypes(include=['float64', 'int64']).columns
scaler = StandardScaler()
std_data = scaler.fit_transform(df[num_cols])
std_df = pd.DataFrame(std_data, columns=num_cols)
print(std_df.head())
   fixed acidity volatile acidity citric acid residual sugar chlorides \
0
       -0.728274
                          0.662757
                                      -0.590571
                                                      -0.104542
                                                                  0.369274
1
      -0.442676
                          1.236296
                                      -0.590571
                                                       1.724938
                                                                  0.061546
2
      -0.442676
                         -1.567674
                                      -0.406017
                                                      -0.104542 -1.784827
3
                         -0.675502
                                                      -0.104542 0.061546
       1.984904
                                      1.993176
       -0.371277
                          0.344124
                                      -0.406017
                                                      -1.411313
                                                                  1.292461
  free sulfur dioxide total sulfur dioxide
                                               density
                                                              pH sulphates \
0
             -1.315587
                                   -1.861983 1.008996 1.920589 -1.380850
1
              1.754116
                                    1.063990 -0.931381 -0.442040
                                                                    1.291763
2
             -0.438529
                                   -0.088666 -0.543305 0.015243
                                                                    0.623610
3
              0.000000
                                    0.443329 1.397071 -0.746896 -0.935414
4
              0.000000
                                    0.443329 -0.931381 -0.746896
                                                                    0.400892
   alcohol quality
0
      -2.0
               -0.5
1
       0.5
               -0.5
2
       0.5
               -0.5
       0.5
3
               2.0
               -0.5
       0.5
```

```
[1]: # Generate a random array of 50 integers and display them using a line chart,
     ⇔scatter plot.
     # Apply appropriate color, labels and styling options
     # slip 9 and slip 15 are same
     import matplotlib.pyplot as plt
     import numpy as np
     from matplotlib import colors
     from matplotlib.ticker import PercentFormatter
    no_of_balls = 50
     x = np.random.randn(50)
     y= np.random.randn(50)
     colors = [np.random.randint(1,4) for i in range(no_of_balls)]
     plt.plot(x,y)
     plt.show()
     plt.scatter(x,y,c=colors)
    plt.show()
```







```
# (similer slip4 of 2)
import pandas as pd
df = pd.read_csv('Data_filled.csv')
print("Dataset Description:")
print(df.describe())
print("\nShape of the dataset:", df.shape)
print("\nFirst 3 rows of the dataset:")
print(df.head(3))
Dataset Description:
            salary
                          age
          5.000000
                     5.000000
      60000.000000 29.500000
       7071.067812 3.640055
      50000.000000 25.000000
```

count mean std min 25% 60000.000000 28.000000 50% 60000.000000 29.500000 75% 60000.000000 30.000000 max 70000.000000 35.000000

Shape of the dataset: (5, 3)

First 3 rows of the dataset:

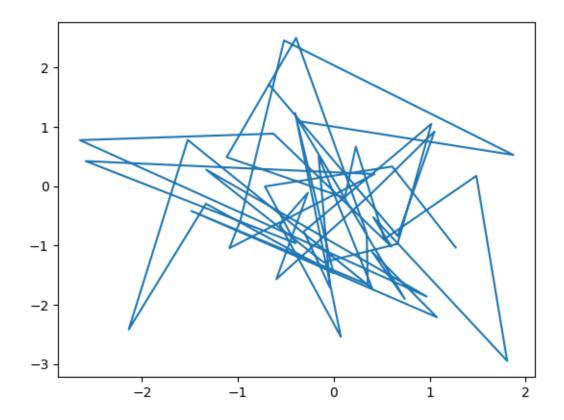
name salary age Alice 50000.0 30.0 0 Bob 60000.0 25.0 2 Charlie 60000.0 29.5

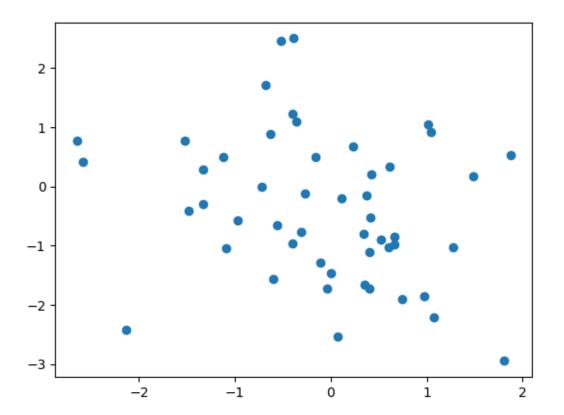
#### October 15, 2024

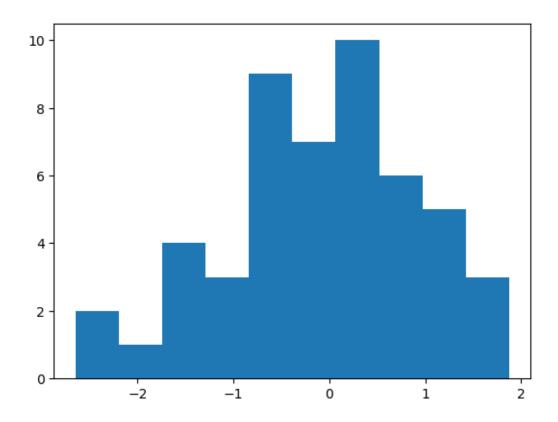
```
[2]: # Write a python program to Display column-wise mean, and median for the
     # SOCR HeightWeight dataset.
     import pandas as pd
     df = pd.read_csv('hw3.csv')
     print("mean is \n", df.mean())
     print("median is \n", df.median())
    mean is
     HeightCm
                 170.0
    WeightKg
                 70.0
    dtype: float64
    median is
     HeightCm
                 170.0
    WeightKg
                 70.0
    dtype: float64
[3]: # Write a python program to compute sum of Manhattan distance between all pairs
      \hookrightarrow of
     # points.
     points = [(1, 2), (3, 4), (5, 6)]
     total_distance = 0
     for i in range(len(points)):
         for j in range(i + 1, len(points)):
             total_distance += abs(points[i][0] - points[j][0]) + abs(points[i][1] -__
      →points[j][1])
     print("Sum of Manhattan distances:", total_distance)
```

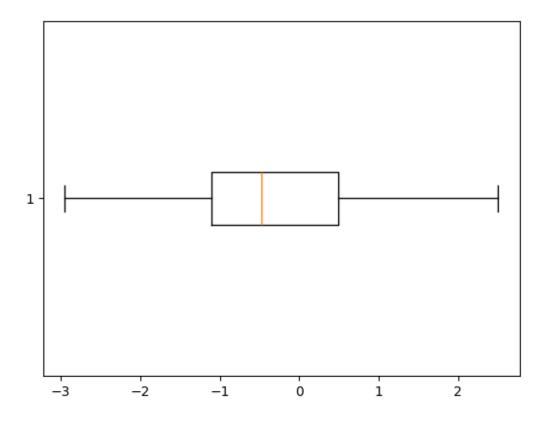
Sum of Manhattan distances: 16

```
[1]: # Generate a random array of 50 integers and display them using a line chart,
     ⇔scatter plot,
     # histogram and box plot. Apply appropriate color, labels and styling options.
     # (this program is same as slip4 of 1 same code)
     import matplotlib.pyplot as plt
     import numpy as np
     x=np.random.randn(50)
     y=np.random.randn(50)
    plt.plot(x,y)
     plt.show()
    plt.scatter(x,y)
    plt.show()
    plt.hist(x)
     plt.show()
    plt.boxplot(y,vert=False)
    plt.show()
```









```
[2]: # Write a Python program to create data frame containing column name, salary,
    # department add 10 rows with some missing and duplicate values to the data \Box
     ⇔frame. Also drop all
    # null and empty values. Print the modified data frame
    import pandas as pd
    import numpy as np
    # Create a DataFrame with some missing and duplicate values
    data = {
        'Name': ['Alice', 'Bob', 'Charlie', 'David', 'Eve', 'Frank', 'Grace', |
     'Salary': [70000, 80000, None, 60000, 90000, 75000, None, 80000, 95000, II
     ⇒50000],
        'Department': ['HR', 'IT', 'Finance', None, 'Marketing', 'IT', 'HR', 'IT',
     }
    df = pd.DataFrame(data)
    # Display the original DataFrame
    print("Original DataFrame:")
```

```
print(df)

# Drop all null and empty values

df_cleaned = df.dropna()

# Display the modified DataFrame
print("\nModified DataFrame (after dropping nulls):")
print(df_cleaned)
```

#### Original DataFrame:

```
Name
             Salary Department
     Alice 70000.0
0
                            HR
1
      Bob 80000.0
                            IT
2 Charlie
                {\tt NaN}
                      Finance
     David 60000.0
                          None
3
4
      Eve 90000.0 Marketing
5
    Frank 75000.0
                            IT
     Grace
6
                {\tt NaN}
                            HR
7
    Alice 80000.0
                            ΙT
8
     None 95000.0
                       Finance
9
   Hannah 50000.0
                          None
```

#### Modified DataFrame (after dropping nulls):

```
        Name
        Salary
        Department

        0
        Alice
        70000.0
        HR

        1
        Bob
        80000.0
        IT

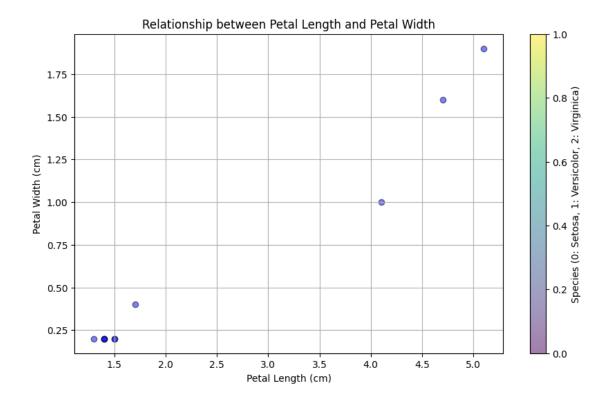
        4
        Eve
        90000.0
        Marketing

        5
        Frank
        75000.0
        IT

        7
        Alice
        80000.0
        IT
```

```
[4]: # Write a Python program to create a graph to find relationship between the
     ⇔petal length
     # and petal width.(Use iris.csv dataset)
     import pandas as pd
     import matplotlib.pyplot as plt
     df = pd.read_csv('Iris.csv')
     print(df.head())
     plt.figure(figsize=(10, 6))
    plt.scatter(df['PetalLength'], df['PetalWidth'], c='blue', alpha=0.5, u
     →edgecolor='k')
     plt.colorbar(label='Species (0: Setosa, 1: Versicolor, 2: Virginica)')
     plt.title('Relationship between Petal Length and Petal Width')
     plt.xlabel('Petal Length (cm)')
    plt.ylabel('Petal Width (cm)')
     plt.grid()
    plt.show()
```

	SepalLength	SepalWidth	PetalLength	PetalWidth	Species
0	5.1	3.5	1.4	0.2	setosa
1	4.9	3.0	1.4	0.2	setosa
2	4.7	3.2	1.3	0.2	setosa
3	4.6	3.1	1.5	0.2	setosa
4	5.0	3.6	1.4	0.2	setosa

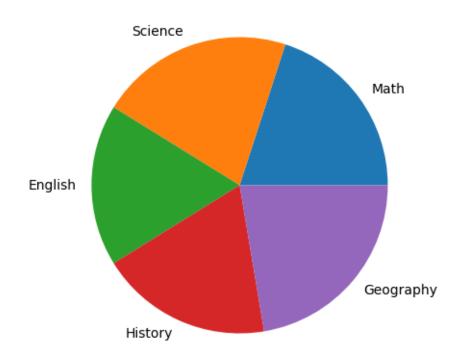


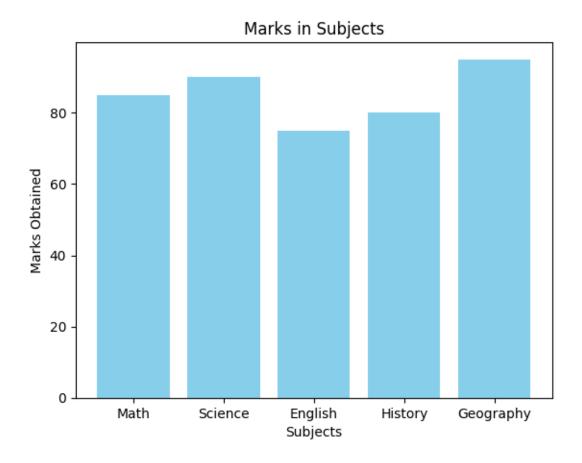
Flattened Array: [3 1 4 1 5 9 2 6 5 3 5]

Maximum Value: 9 Minimum Value: 1

```
[1]: # Write a Python NumPy program to compute the weighted average along the
      \hookrightarrow specified
     # axis of a given flattened array.
     import numpy as np
     data = np.array([10, 20, 30, 40, 50])
     weights = np.array([1, 2, 3, 4, 5])
     weighted_average = np.average(data, weights=weights)
     print(f"Flattened Array: {data}")
     print(f"Weights: {weights}")
     print(f"Weighted Average: {weighted_average}")
    Flattened Array: [10 20 30 40 50]
    Weights: [1 2 3 4 5]
    Weighted Average: 36.6666666666664
[1]: # Write a Python program to view basic statistical details of the data (Use
     \hookrightarrow advertising.csv)
     # (similer as slip1 of 2)
     import pandas as pd
     df = pd.read_csv('s7.csv')
     basic_stats = df.describe()
     print("Basic Statistical Details:")
     print(basic_stats)
    Basic Statistical Details:
           Country Purchased
    count
                 4
                 3
                            2
    unique
               USA
                          Yes
    top
    freq
                 2
```

```
[3]: # Write a python program to create two lists, one representing subject names
     ⇔and the other
     # representing marks obtained in those subjects. Display the data in a pie
     ⇔chart and bar chart
     # (similer program as slip 9 of 2)
     import matplotlib.pyplot as plt
     import numpy as np
     subjects = ['Math', 'Science', 'English', 'History', 'Geography']
     marks = [85, 90, 75, 80, 95]
     fig = plt.figure(figsize=(5, 5))
     plt.pie(marks,labels=subjects)
    plt.show()
     plt.subplot(1, 1, 1)
    plt.bar(subjects, marks, color='skyblue')
    plt.xlabel('Subjects')
     plt.ylabel('Marks Obtained')
     plt.title('Marks in Subjects')
    plt.show()
```





```
[4]: # Write a python program to create a data frame for students' information suchulas name,
# graduation percentage and age. Display average age of students, average oful agraduation
# percentage

import pandas as pd

data = {
    'Name': ['Alice', 'Bob', 'Charlie', 'David', 'Eva'],
    'Graduation Percentage': [85.5, 78.0, 92.5, 88.0, 80.0],
    'Age': [21, 22, 20, 23, 21]
}

# Create a DataFrame from the dictionary
df = pd.DataFrame(data)

# Calculate the average age of students
average_age = df['Age'].mean()
```

```
# Calculate the average graduation percentage
average_graduation_percentage = df['Graduation Percentage'].mean()

print("Student Information:")
print(df)
print(f"\nAverage Age of Students: {average_age:.2f}")
print(f"Average Graduation Percentage: {average_graduation_percentage:.2f}")
```

#### Student Information:

	Name	Graduation	Percentage	Age
0	Alice		85.5	21
1	Bob		78.0	22
2	Charlie		92.5	20
3	David		88.0	23
4	Eva		80.0	21

Average Age of Students: 21.40

Average Graduation Percentage: 84.80

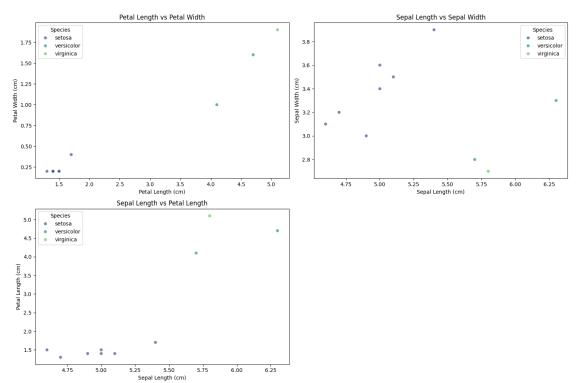
```
[8]: # Write a Python program to draw scatter plots to compare two features of the
     ⇔iris dataset
     import pandas as pd
     import matplotlib.pyplot as plt
     import seaborn as sns
     # Load the iris dataset
     iris = pd.read_csv("Iris.csv")
     # Display the first few rows of the dataset
     print(iris.head())
     # Set up the matplotlib figure
     plt.figure(figsize=(15, 10))
     # Create scatter plot for Petal Length vs Petal Width
     plt.subplot(2, 2, 1)
     sns.scatterplot(data=iris, x='PetalLength', y='PetalWidth', hue='Species', u
      →palette='viridis', alpha=0.7)
     plt.title("Petal Length vs Petal Width")
     plt.xlabel("Petal Length (cm)")
     plt.ylabel("Petal Width (cm)")
     # Create scatter plot for Sepal Length vs Sepal Width
     plt.subplot(2, 2, 2)
     sns.scatterplot(data=iris, x='SepalLength', y='SepalWidth', hue='Species',
      ⇒palette='viridis', alpha=0.7)
     plt.title("Sepal Length vs Sepal Width")
     plt.xlabel("Sepal Length (cm)")
     plt.ylabel("Sepal Width (cm)")
     # Create scatter plot for Sepal Length vs Petal Length
     plt.subplot(2, 2, 3)
     sns.scatterplot(data=iris, x='SepalLength', y='PetalLength', hue='Species', u
      →palette='viridis', alpha=0.7)
     plt.title("Sepal Length vs Petal Length")
     plt.xlabel("Sepal Length (cm)")
```

```
plt.ylabel("Petal Length (cm)")

# Adjust layout
plt.tight_layout()

# Show the plot
plt.show()
```

	SepalLength	SepalWidth	PetalLength	PetalWidth	Species
0	5.1	3.5	1.4	0.2	setosa
1	4.9	3.0	1.4	0.2	setosa
2	4.7	3.2	1.3	0.2	setosa
3	4.6	3.1	1.5	0.2	setosa
4	5.0	3.6	1.4	0.2	setosa



```
[9]: # Write a Python program to create a data frame containing columns name, age ,□ ⇒salary,
# department . Add 10 rows to the data frame. View the data frame

import pandas as pd

data = {
```

```
'name': ['Alice', 'Bob', 'Charlie', 'David', 'Eva', 'Frank', 'Grace',
'Hannah', 'Ian', 'Judy'],
    'age': [25, 30, 22, 28, 26, 35, 29, 31, 24, 27],
    'salary': [50000, 60000, 45000, 70000, 52000, 65000, 58000, 62000, 48000,
$\delta \text{55000}],
    'department': ['HR', 'IT', 'Finance', 'IT', 'HR', 'Sales', 'Marketing',
$\delta 'Finance', 'Sales', 'Marketing']
}

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

```
name age salary department
    Alice
                 50000
0
            25
      Bob
                 60000
                               ΙT
1
            30
2 Charlie
            22
                 45000
                          Finance
3
    David
            28
                 70000
                               ΙT
4
      Eva
            26
                 52000
                               HR
5
    Frank
             35
                 65000
                             Sales
6
    Grace
             29
                 58000 Marketing
7
   Hannah
             31
                 62000
                           Finance
       Ian
                 48000
                             Sales
8
             24
9
      Judy
             27
                 55000 Marketing
```

```
[4]: # Write a Python program To create a dataframe containing columns name, age and
    # percentage. Add 10 rows to the dataframe. View the dataframe.
    import pandas as pd
    data = {
        'name': ['Alice', 'Bob', 'Charlie', 'David', 'Eva', 'Frank', 'Grace', |
     'age': [20, 22, 21, 23, 24, 25, 22, 26, 27, 19],
         'percentage': [85.5, 90.2, 78.0, 88.5, 92.0, 80.0, 84.5, 75.5, 89.0, 91.0]
    }
    df = pd.DataFrame(data)
    print(df)
    # # below code for slip19 of 2
    # print("Shape of the dataset:", df.shape)
    # print("features :", df. info())
    # print("\nDescription of the Data:\n", df.describe())
    # print("\nData Types:\n", df.dtypes)
```

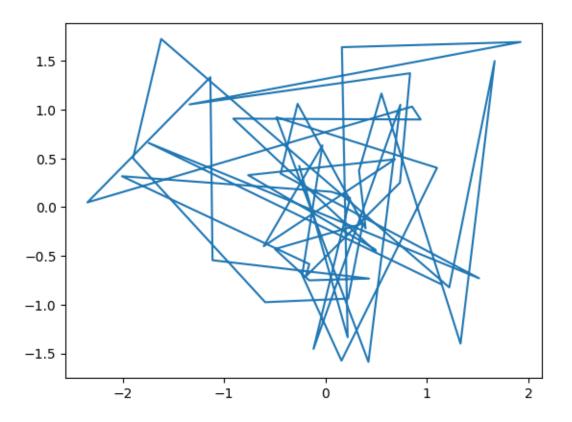
```
age percentage
      name
0
     Alice
             20
                       85.5
       Bob
             22
                       90.2
1
                       78.0
2
  Charlie
             21
3
     David
             23
                       88.5
4
             24
                       92.0
       Eva
5
     Frank
             25
                       80.0
                       84.5
6
     Grace
             22
                       75.5
7
   Hannah
             26
8
       Tan
             27
                       89.0
9
      Judy
             19
                       91.0
```

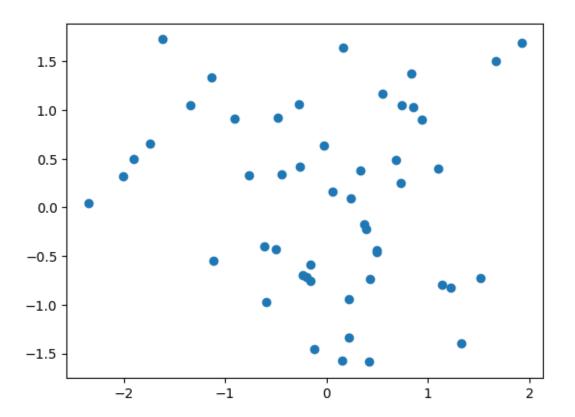
```
[2]: # slip 19 of 2 program in slip 4 of 2 (program is same )
import pandas as pd
import numpy as np
```

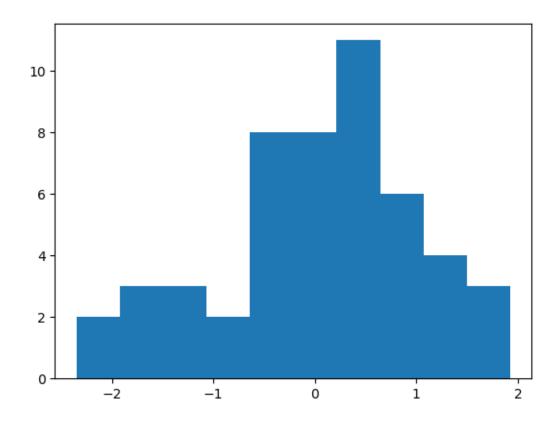
```
# Initial data
data = {
    'name': ['Alice', 'Bob', 'Charlie', 'David', 'Eva', 'Frank', 'Grace', |
 'age': [20, 22, 21, 23, 24, 25, 22, 26, 27, 19],
    'percentage': [85.5, 90.2, 78.0, 88.5, 92.0, 80.0, 84.5, 75.5, 89.0, 91.0]
}
# Create DataFrame
df = pd.DataFrame(data)
# Add duplicate rows
duplicates = pd.DataFrame({
    'name': ['Alice', 'Bob', 'Charlie', 'Charlie', 'Eva'],
    'age': [20, 22, 21, np.nan, 24],
    'percentage': [85.5, 90.2, 78.0, 88.5, 92.0]
})
# Concatenate the original DataFrame and the duplicates
df = pd.concat([df, duplicates], ignore_index=True)
# Add a new column 'remarks' with empty values
df['remarks'] = ''
# Display the DataFrame
print(df)
```

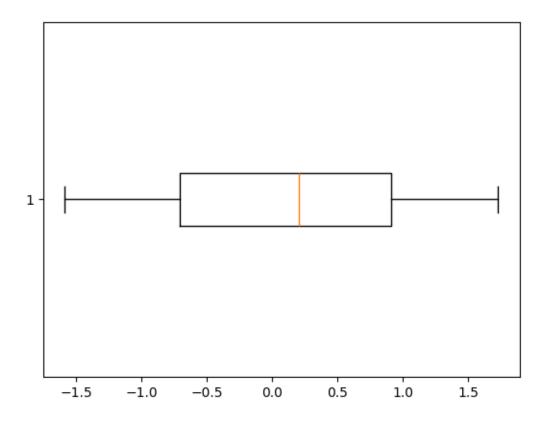
```
name
             age percentage remarks
0
     Alice 20.0
                        85.5
1
       Bob 22.0
                        90.2
2
   Charlie 21.0
                        78.0
3
     David 23.0
                        88.5
4
       Eva 24.0
                        92.0
5
     Frank 25.0
                        80.0
6
     Grace 22.0
                        84.5
7
    Hannah 26.0
                        75.5
8
       Ian 27.0
                        89.0
9
      Judy 19.0
                        91.0
     Alice 20.0
10
                        85.5
       Bob 22.0
                        90.2
11
12 Charlie 21.0
                        78.0
13 Charlie NaN
                        88.5
14
       Eva 24.0
                        92.0
```

```
[2]: \# slip20-1 : Generate a random array of 50 integers and display them using a_{\square}
     ⇒line chart, scatter plot,
     # histogram and box plot. Apply appropriate color, labels and styling options.
     import matplotlib.pyplot as plt
     import numpy as np
     x=np.random.randn(50)
     y=np.random.randn(50)
    plt.plot(x,y)
     plt.show()
    plt.scatter(x,y)
     plt.show()
    plt.hist(x)
     plt.show()
     plt.boxplot(y,vert=False)
     plt.show()
     # slip20-2: Add two outliers to the above data and display the box plot.
     outliers = np.append(y, [10, -10]) # Adding two outliers
     # Display the box plot with outliers
     plt.boxplot(outliers, vert=False)
     plt.title("Box Plot with Outliers")
     plt.show()
```

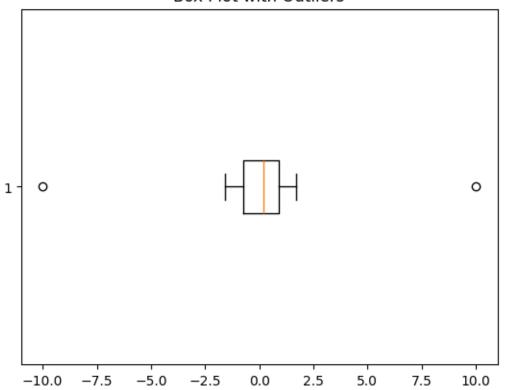




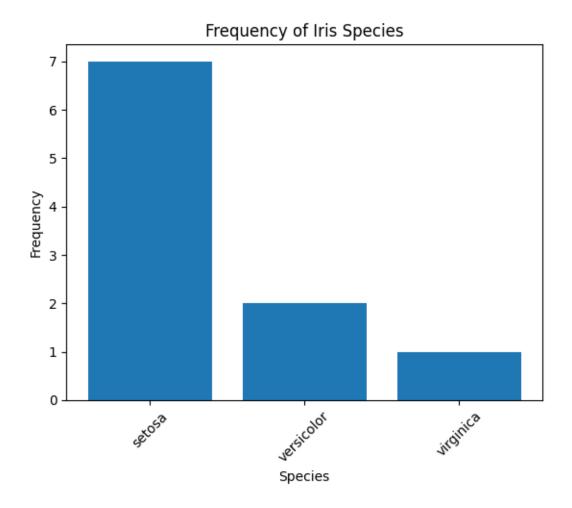




# Box Plot with Outliers



```
[2]: # Import dataset "iris.csv". Write a Python program to create a Bar plot to
     ⇔get the frequency
     # of the three species of the Iris data.
     import pandas as pd
     import matplotlib.pyplot as plt
     # Import the dataset
     df = pd.read_csv('Iris.csv')
     # Count the frequency of each species
     species_counts = df['Species'].value_counts()
     # Create a bar plot
     plt.bar(species_counts.index, species_counts.values)
     plt.title('Frequency of Iris Species')
     plt.xlabel('Species')
     plt.ylabel('Frequency')
     plt.xticks(rotation=45)
     plt.show()
```



```
plt.xlabel("Petal Length (cm)")
plt.ylabel("Frequency")
plt.title("Histogram of Petal Length by Species")
plt.legend()
plt.show()
```

