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5.	Write a Python program that calculates and displays a frequency distribution of a given dataset.		
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10.	Create the dashboard using Power BI.		
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Practical -1 :- Python Program to perform Array operation using Numpy package.

Array Operations.

```
a = np.arange(15).reshape(3, 5)
a
```

```
↳ array([[ 0,  1,  2,  3,  4],
       [ 5,  6,  7,  8,  9],
       [10, 11, 12, 13, 14]])
```

```
a.shape
```

```
↳ (3, 5)
```

```
a.ndim
```

```
↳ 2
```

```
a.dtype.name
type(a)
```

```
↳ numpy.ndarray
```

```
print(np.__version__)
```

```
↳ 1.26.4
```

```
a.itemsize
```

```
↳ 8
```

```
a.size
```

```
↳ 15
```

```
a.tolist()
```

```
↳ [0, 1, 2, 3, 4, 5, 6, 7, 8, 9]
```

```
import numpy as np
a = np.array([2, 3, 4])
a
```

```
↳ array([2, 3, 4])

a.dtype
↳ dtype('int64')

np.ones((2, 3, 4), dtype=np.int16)
↳ array([[[1, 1, 1],  
          [1, 1, 1],  
          [1, 1, 1]],  
  
        [[[1, 1, 1],  
          [1, 1, 1],  
          [1, 1, 1]]], dtype=int16)

np.zeros((3, 3))
↳ array([[0., 0., 0.],  
        [0., 0., 0.],  
        [0., 0., 0.]])  
  
a = np.arange(10)                      # 1d array  
print(a)  
  
↳ [0 1 2 3 4 5 6 7 8 9]  
  
b = np.arange(12).reshape(4, 3)        # 2d array  
print(b)  
  
↳ [[ 0  1  2]  
  [ 3  4  5]  
  [ 6  7  8]  
  [ 9 10 11]]  
  
c = np.arange(24).reshape(2,3,4)    # 3d array  
print(c)  
  
↳ [[[ 0  1  2  3]  
  [ 4  5  6  7]  
  [ 8  9 10 11]]  
  
  [[12 13 14 15]  
   [16 17 18 19]  
   [20 21 22 23]]]  
  
print(np.arange(10000))
↳ [ 0    1    2 ... 9997 9998 9999]
```

```
print(np.arange(10000).reshape(100, 100))

[[    0     1     2 ...    97    98    99]
 [ 100   101   102 ...   197   198   199]
 [ 200   201   202 ...   297   298   299]
 ...
 [9700  9701  9702 ...  9797  9798  9799]
 [9800  9801  9802 ...  9897  9898  9899]
 [9900  9901  9902 ...  9997  9998  9999]]


b = np.arange(12).reshape(3, 4)
b

array([[ 0,  1,  2,  3],
       [ 4,  5,  6,  7],
       [ 8,  9, 10, 11]])


b.sum(axis=0)

array([12, 15, 18, 21])


b.min(axis=1)

array([0, 4, 8])


b.cumsum(axis=1)

array([[ 0,  1,  3,  6],
       [ 4,  9, 15, 22],
       [ 8, 17, 27, 38]])


a = np.matrix('1 2; 3 4')
a

matrix([[1, 2],
       [3, 4]])


np.matrix([[1, 2], [3, 4]])


matrix([[1, 2],
       [3, 4]])


x = np.matrix(np.arange(12).reshape((3,4))); x

matrix([[ 0,  1,  2,  3],
       [ 4,  5,  6,  7],
       [ 8,  9, 10, 11]]]
```

```
[ 8,  9, 10, 11])  
  
#Max and Min in array  
print("Maximum in Array",x.max())  
print("Minimum in Array",x.min())  
print("Mean in Array",x.mean())  
  
Maximum in Array 11  
Minimum in Array 0  
Mean in Array 5.5  
  
# Creating arrays  
arr1 = np.array([1, 2, 3, 4, 5])  
arr2 = np.array([6, 7, 8, 9, 10])  
  
print("Array 1:", arr1)  
print("Array 2:", arr2)  
  
Array 1: [1 2 3 4 5]  
Array 2: [ 6  7  8  9 10]  
  
#Sorting Array  
arr1.sort()  
print(arr1)  
  
[21 58 73 84 93]  
  
arr = np.array([1, 2, 3, 4])  
print("array + 5:", arr + 5)  
print("array * 2:", arr * 2)  
print("array squared:", arr ** 2)  
  
array + 5: [6 7 8 9]  
array * 2: [2 4 6 8]  
array squared: [ 1  4  9 16]  
  
# Array addition  
sum_arr = np.add(arr1, arr2)  
print("Sum of arrays:", sum_arr)  
  
Sum of arrays: [ 27  65  81  93 103]  
  
# Array subtraction  
diff_arr = np.subtract(arr1, arr2)  
print("Difference of arrays:", diff_arr)  
  
Difference of arrays: [15 51 65 75 83]  
  
# Array multiplication  
prod_arr = np.multiply(arr1, arr2)
```

```
print("Product of arrays:", prod_arr)
Product of arrays: [126 406 584 756 930]

# Array division
div_arr = np.divide(arr1, arr2)
print("Division of arrays:", div_arr)

Division of arrays: [3.5          8.28571429  9.125        9.33333333  9.3

# Reshaping an array
reshaped_arr = np.arange(1, 10).reshape(3, 3)
print("Reshaped 3x3 array:")
print(reshaped_arr)

Reshaped 3x3 array:
[[1 2 3]
 [4 5 6]
 [7 8 9]]

# Transposing an array
transposed_arr = reshaped_arr.T
print("Transposed array:")
print(transposed_arr)

Transposed array:
[[1 4 7]
 [2 5 8]
 [3 6 9]]

# Concatenation of arrays
concatenated_arr = np.concatenate((arr1, arr2))
print("Concatenated array:", concatenated_arr)

Concatenated array: [21 58 73 84 93  6  7  8  9 10]

# Indexing and Slicing
print("First element of arr1:", arr1[0])
print("Last element of arr1:", arr1[-1])
print("Elements from index 1 to 3 of arr1:", arr1[1:4])
print("Every second element of arr1:", arr1[::2])

First element of arr1: 21
Last element of arr1: 93
Elements from index 1 to 3 of arr1: [58 73 84]
Every second element of arr1: [21 73 93]

# Slicing in a 2D array
print("First row of reshaped array:", reshaped_arr[0])
print("2*2 matrix array:", reshaped_arr[0:2, 0:2])
```

```
print("First column of reshaped array: ", reshaped_arr[:, 0])
print("Submatrix from reshaped array: ", reshaped_arr[0:2, 1:3])

First row of reshaped array: [1 2 3]
First column of reshaped array: [1 4 7]
Submatrix from reshaped array: [[2 3]
 [5 6]]
```

```
arr = np.array([1,2,3,4,5,6,9,7,5])
newarr = np.array_split(arr, 3)
print(newarr)
```

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Practical -2 :- Python Program to perform Data Manipulation operations using Pandas package.

```
import pandas as pd
import numpy as np

df = pd.DataFrame(data=np.random.randint(0,10,(5,3)),
                   columns=['c1','c2','c3'],
                   index = ['A','B','C','D','E'])
df
```

	c1	c2	c3
A	3	5	3
B	9	6	9
C	9	3	5
D	8	2	3
E	4	9	8

```
# Properties
df.shape
```

```
(5, 3)
```

```
pip install pandas
```

```
df.dtypes
```

```
c1    int32
c2    int32
c3    int32
dtype: object
```

```
list(df.columns)
```

```
['c1', 'c2', 'c3']
```

```
df.index
```

```
Index(['A', 'B', 'C', 'D', 'E'], dtype='object')
```

```
df
```

```
- - - - -
```

	c1	c2	c3
A	3	5	3
B	9	6	9
C	9	3	5
D	8	2	3
E	4	9	8

```
# Selection of columns  
df['c1'] # 1D
```

```
↳ A    3  
B    9  
C    9  
D    8  
E    4  
Name: c1, dtype: int32
```

```
df[['c1','c2']] # list of columns
```

```
↳  
    c1  c2  
_____  
A  3  5  
B  9  6  
C  9  3  
D  8  2  
E  4  9
```

```
df
```

```
↳  
    c1  c2  c3  
_____  
A  3  5  3  
B  9  6  9  
C  9  3  5  
D  8  2  3  
E  4  9  8
```

```
df.loc['A']
```

```
↳  c1    3  
  c2    5  
  c3    3  
Name: A, dtype: int32
```

```
df.loc[['A','D']]
```

	c1	c2	c3
A	3	5	3
D	8	2	3

```
df
```

	c1	c2	c3
A	3	5	3
B	9	6	9
C	9	3	5
D	8	2	3
E	4	9	8

```
df.iloc[[2,3]]
```

	c1	c2	c3
C	6	10	3
D	14	15	4

```
df.iloc[::2]
```

	c1	c2	c3
A	3	5	3
C	9	3	5
E	4	9	8

```
df
```

```
df
df.iloc[1:2][0:1:]
```

	c1	c2	c3
B	9	6	9

```
df.iloc[1:2][0:1:]
```

```
        c1  c2  c3  c4
```

	c2	c3
B	6	9
D	2	3

```
# Adding new columns  
df['c4'] = [11,12,13,14,15] # Same as of Dict column inserted  
df
```

	c1	c2	c3	c4
A	3	5	3	11
B	9	6	9	12
C	9	3	5	13
D	8	2	3	14
E	4	9	8	15

```
df.loc['F'] = [1,2,3,4] #row inserted
```

```
df
```

	c1	c2	c3	c4
A	3	5	3	11
B	9	6	9	12
C	9	3	5	13
D	8	2	3	14
E	4	9	8	15
F	1	2	3	4

```
# Delete row/record or column  
df.drop('c4', axis=1)
```

	c1	c2	c3
A	3	5	3
B	9	6	9
C	9	3	5
D	8	2	3
E	4	9	8

```
F 1 2 3
```

```
df
```

	c1	c2	c3	c4
A	3	5	3	11
B	9	6	9	12
C	9	3	5	13
D	8	2	3	14
E	4	9	8	15
F	1	2	3	4

```
df.drop('c4', axis=1, inplace=True) # Permanent delete
```

```
df
```

	c1	c2	c3
A	3	5	3
B	9	6	9
C	9	3	5
D	8	2	3
E	4	9	8
F	1	2	3

```
df.drop('F', axis=0)
```

	c1	c2	c3
A	3	5	3
B	9	6	9
C	9	3	5
D	8	2	3
E	4	9	8

```
df
```

	c1	c2	c3
--	----	----	----

A	3	5	3
B	9	6	9
C	9	3	5
D	8	2	3
E	4	9	8
F	1	2	3

```
df.drop('F', axis=0, inplace=True)
```

```
df
```

	c1	c2	c3
A	3	5	3
B	9	6	9
C	9	3	5
D	8	2	3
E	4	9	8

```
df
```

	c1	c2	c3
A	3	5	3
B	9	6	9
C	9	3	5
D	8	2	3
E	4	9	8

```
df.drop(index=df.index[2])
```

	c1	c2	c3
A	3	5	3
B	9	6	9
D	8	2	3
E	4	9	8

```
at.index[4]
```

```
'C'
```

```
df.iloc[[0,1,3,4]]
```

	c1	c2	c3
A	8	7	10
B	3	10	0
D	14	15	4
E	13	2	16

▼ Reading Data From external files

```
df = pd.read_excel('stud_db.xlsx', sheet_name='XII')  
df
```

	Roll	Name	Marks	Address	Dept
0	1	qw	76	Pune	1
1	2	df	45	Mumbai	1
2	3	vc	56	Nagpur	1
3	4	bn	78	Pune	2
4	5	hj	74	Mumbai	2
5	6	yu	49	Nagpur	2
6	7	fg	68	Pune	2
7	8	tr	84	Mumbai	1

```
df.shape
```

```
(8, 5)
```

```
df.columns
```

```
Index(['Roll', 'Name', 'Marks', 'Address', 'Dept'], dtype='object')
```

```
df['Marks'].sum()
```

```
530
```

```
df['Marks'].mean()
```

```
df['Marks'].mean()
```

66.25

```
df['Marks']
```

```
0    76  
1    45  
2    56  
3    78  
4    74  
5    49  
6    68  
7    84  
Name: Marks, dtype: int64
```

```
df['Marks']<60 # Broadcasting
```

	Roll	Name	Marks	Address	Dept
1	2	df	45	Mumbai	1
2	3	vc	56	Nagpur	1
5	6	yu	49	Nagpur	2

```
df[df['Marks']<60] # Selection
```

	Roll	Name	Marks	Address	Dept
1	2	df	45	Mumbai	1
2	3	vc	56	Nagpur	1
5	6	yu	49	Nagpur	2

```
df_pune = df[df['Address']=='Pune']  
df_pune  
df_pune['Marks'].mean()
```

74.0

```
df_mumbai = df[df['Address']=='Mumbai']  
df_mumbai['Marks'].mean()
```

67.66666666666667

```
df_nagpur = df[df['Address']=='Nagpur']  
df_nagpur['Marks'].mean()
```

52.5

df

	Roll	Name	Marks	Address	Dept
0	1	qw	76	Pune	1
1	2	df	45	Mumbai	1
2	3	vc	56	Nagpur	1
3	4	bn	78	Pune	2
4	5	hj	74	Mumbai	2
5	6	yu	49	Nagpur	2
6	7	fg	68	Pune	2
7	8	tr	84	Mumbai	1

df['Address'].unique()

array(['Pune', 'Mumbai', 'Nagpur'], dtype=object)

df['Address'].nunique()

3

df['Address'].value_counts()

Address

Pune 3

Mumbai 3

Nagpur 2

Name: count, dtype: int64

df

	Roll	Name	Marks	Address	Dept
0	1	qw	76	Pune	1
1	2	df	45	Mumbai	1
2	3	vc	56	Nagpur	1
3	4	bn	78	Pune	2
4	5	hj	74	Mumbai	2
5	6	yu	49	Nagpur	2
6	7	fg	68	Pune	2
7	8	tr	84	Mumbai	1

```
df['Dept'].value_counts()

1    4
2    4
Name: Dept, dtype: int64

# To find name of topper
df[df['Marks'].max()==df['Marks']]['Name']

7    tr
Name: Name, dtype: object

# To find city of topper
df[df['Marks'].max()==df['Marks']]['Address'].iloc[0]

'Mumbai'

# To find Name second topper
temp = df[df['Marks'] < df['Marks'].max()] # All students except topper
marks_2nd_topper = temp['Marks'].max()
df[df['Marks']==marks_2nd_topper]['Name'].iloc[0]

'bn'

df.sort_values('Marks', ascending=False)['Name'].iloc[3] # 4th topper
```

	Roll	Name	Marks	Address	Dept
7	8	tr	84	Mumbai	1
3	4	bn	78	Pune	2
0	1	qw	76	Pune	1
4	5	hj	74	Mumbai	2
6	7	fg	68	Pune	2
2	3	vc	56	Nagpur	1
5	6	yu	49	Nagpur	2
1	2	df	45	Mumbai	1

df

	Roll	Name	Marks	Address	Dept
0	1	qw	76	Pune	1
1	2	df	45	Mumbai	1
2	3	vc	56	Nagpur	1
-	-	-	--	-	-

3	4	bn	78	Pune	2
4	5	hj	74	Mumbai	2
5	6	yu	49	Nagpur	2
6	7	fg	68	Pune	2
7	8	tr	84	Mumbai	1

```
df['City_Code'] = df['Address'].apply(lambda x:x[:2])
```

df

	Roll	Name	Marks	Address	Dept	City_Code
0	1	qw	76	Pune	1	Pu
1	2	df	45	Mumbai	1	Mu
2	3	vc	56	Nagpur	1	Na
3	4	bn	78	Pune	2	Pu
4	5	hj	74	Mumbai	2	Mu
5	6	yu	49	Nagpur	2	Na
6	7	fg	68	Pune	2	Pu
7	8	tr	84	Mumbai	1	Mu

```
df['Marks'].apply(lambda x:x/10)
```

0	7.6
1	4.5
2	5.6
3	7.8
4	7.4
5	4.9
6	6.8
7	8.4

Name: Marks, dtype: float64

▼ GROUPBY

df

	Roll	Name	Marks	Address	Dept	City_Code
0	1	qw	76	Pune	1	Pu
1	2	df	45	Mumbai	1	Mu

```

2   3    vc     56   Nagpur    1      Na
3   4    bn     78   Pune      2      Pu
4   5    hj     74   Mumbai     2      Mu
5   6    yu     49   Nagpur     2      Na
6   7    fg     68   Pune      2      Pu
7   8    tr     84   Mumbai     1      Mu

```

```
# City-wise Avg marks
df.groupby('Address').mean()['Marks']
```

```

Address
Mumbai      67.666667
Nagpur      52.500000
Pune        74.000000
Name: Marks, dtype: float64

```

```
df.groupby('Dept').max()['Roll']
```

```

Dept
1      8
2      7
Name: Roll, dtype: int64

```

▼ Merging of DF's

```
df_student = pd.read_excel('stud_db.xlsx', sheet_name='XII')
df_dept = pd.read_excel('stud_db.xlsx', sheet_name='Depts')
```

```
df_student
```

	Roll	Name	Marks	Address	Dept
0	1	qw	76	Pune	1
1	2	df	45	Mumbai	1
2	3	vc	56	Nagpur	1
3	4	bn	78	Pune	2
4	5	hj	74	Mumbai	2
5	6	yu	49	Nagpur	2
6	7	fg	68	Pune	2
7	8	tr	84	Mumbai	1

```
df_dept
```

	dept_id	Name	HOD
0	1	Comp Science	XYZ
1	2	Statistics	PQR
2	3	Maths	MNO

```
pd.merge(df_student,df_dept, left_on='Dept', right_on='dept_id')
```

	Roll	Name_x	Marks	Address	Dept	dept_id	Name_y	HOD
0	1	qw	76	Pune	1	1	Comp Science	XYZ
1	2	df	45	Mumbai	1	1	Comp Science	XYZ
2	3	vc	56	Nagpur	1	1	Comp Science	XYZ
3	8	tr	84	Mumbai	1	1	Comp Science	XYZ
4	4	bn	78	Pune	2	2	Statistics	PQR
5	5	hj	74	Mumbai	2	2	Statistics	PQR
6	6	yu	49	Nagpur	2	2	Statistics	PQR
7	7	fg	68	Pune	2	2	Statistics	PQR

```
pd.merge(df_student,df_dept, left_on='Dept', right_on='dept_id', how='right')
```

	Roll	Name_x	Marks	Address	Dept	dept_id	Name_y	HOD
0	1.0	qw	76.0	Pune	1.0	1	Comp Science	XYZ
1	2.0	df	45.0	Mumbai	1.0	1	Comp Science	XYZ
2	3.0	vc	56.0	Nagpur	1.0	1	Comp Science	XYZ
3	8.0	tr	84.0	Mumbai	1.0	1	Comp Science	XYZ
4	4.0	bn	78.0	Pune	2.0	2	Statistics	PQR
5	5.0	hj	74.0	Mumbai	2.0	2	Statistics	PQR
6	6.0	yu	49.0	Nagpur	2.0	2	Statistics	PQR
7	7.0	fg	68.0	Pune	2.0	2	Statistics	PQR
8	NaN	NaN	NaN	NaN	NaN	3	Maths	MNO

```
import pandas as pd
data = {
    'ID': [101, 102, 103, 104, 105],
    'Name': ['Sonal', 'Tanvi', 'Anuja', 'Rutuja', 'Eve'],
    'Age': [20, 22, 23, 20, 15],
```

```
'Salary': [50000, 60000, 70000, 80000, 90000],  
'Department': ['HR' , 'IT', 'IT', 'Finance', 'HR']  
}  
  
df = pd.DataFrame(data)  
print("Original DataFrame:")  
print(df)  
  
print("\nPrint Data (Name):")  
print(df['Age'])  
  
print("\nFiltering employees with Salary < 70000:")  
filtered_df = df[df['Salary'] <70000]  
print(filtered_df)  
  
print("\nSorting employees by Age:")  
sorted_df = df.sort_values(by='Age', ascending=False)  
print(sorted_df)  
  
print("\nAverage Salary by Department:")  
grouped_df = df.groupby('Department')['Salary'].mean()  
print(grouped_df)  
  
# Updating Values  
df.loc[df['Department'] == 'IT', 'Salary'] += 5000  
print("\nDataFrame after Salary increment for IT Department:\n ")  
print(df)  
  
# Dropping a Column  
df.drop(columns=['Bonus'], inplace=True)  
print("\nDataFrame after dropping Bonus column:")  
print(df)  
  
# Reset Index  
df.reset_index(drop=True, inplace=True)  
print("\nDataFrame after resetting index:")  
print(df)  
  
# Sample data  
data = {  
    'Name': ['Alice', 'Bob', 'Charlie', 'David'],  
    'Age': [25, 30, 35, 40],  
    'City': ['New York', 'Los Angeles', 'Chicago', 'Houston']  
}  
  
# Create DataFrame  
df = pd.DataFrame(data)
```

```
# Save to CSV  
df.to_csv('dataset.csv', index=False)  
  
print("Dataset saved as dataset.csv")  
  
ds_loaded = pd.read_csv('dataset.csv')  
print(ds_loaded)
```

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```
pip install matplotlib
```

```
Requirement already satisfied: matplotlib in /usr/local/lib/python3.11/dist-  
Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.1  
Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.11/di  
Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3.  
Requirement already satisfied: kiwisolver>=1.3.1 in /usr/local/lib/python3.  
Requirement already satisfied: numpy>=1.23 in /usr/local/lib/python3.11/dis  
Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.11  
Requirement already satisfied: pillow>=8 in /usr/local/lib/python3.11/dist-  
Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.1  
Requirement already satisfied: python-dateutil>=2.7 in /usr/local/lib/pytho  
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.11/dist-p
```

```
import matplotlib.pyplot as plt
```

```
x = [10, 20, 30, 40, 55]  
y = [20, 25, 35, 55, 70]
```

```
plt.plot(x, y)
```

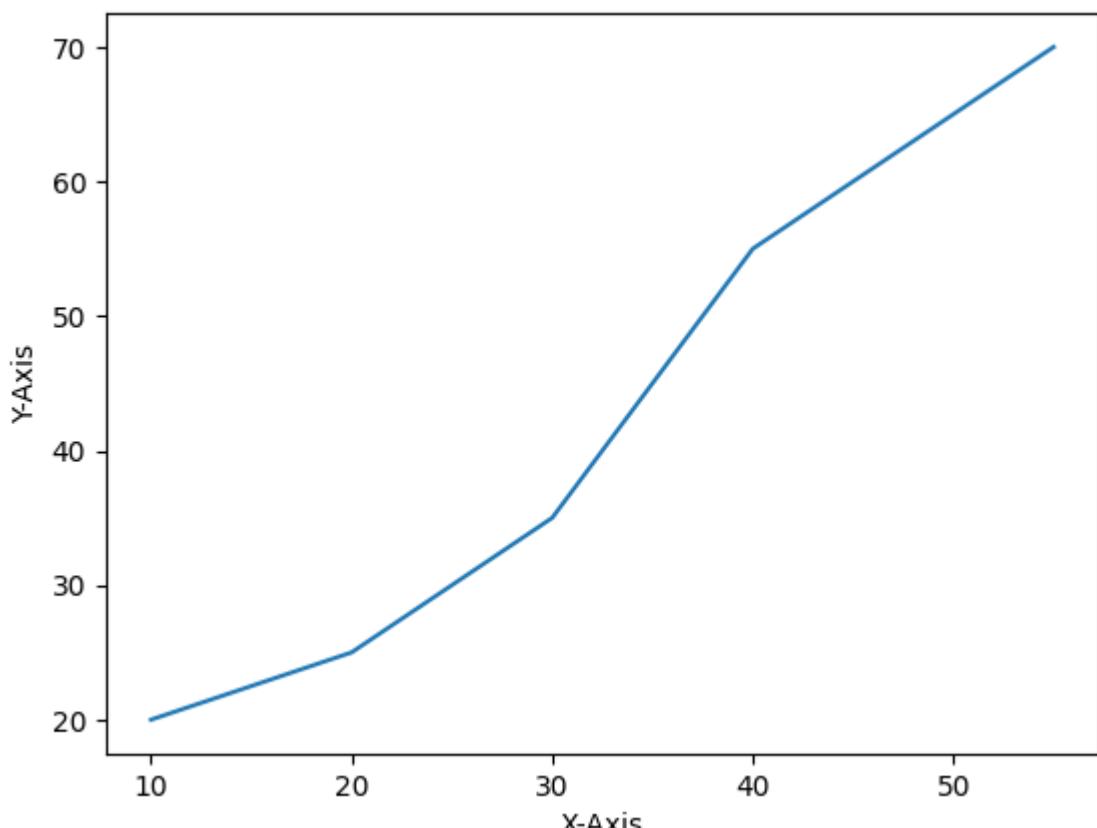
```
plt.title("Line Chart")
```

```
plt.ylabel('Y-Axis')
```

```
plt.xlabel('X-Axis')  
plt.show()
```



Line Chart



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

data = pd.read_csv("tip.csv")

x = data['day']
y = data['total_bill']

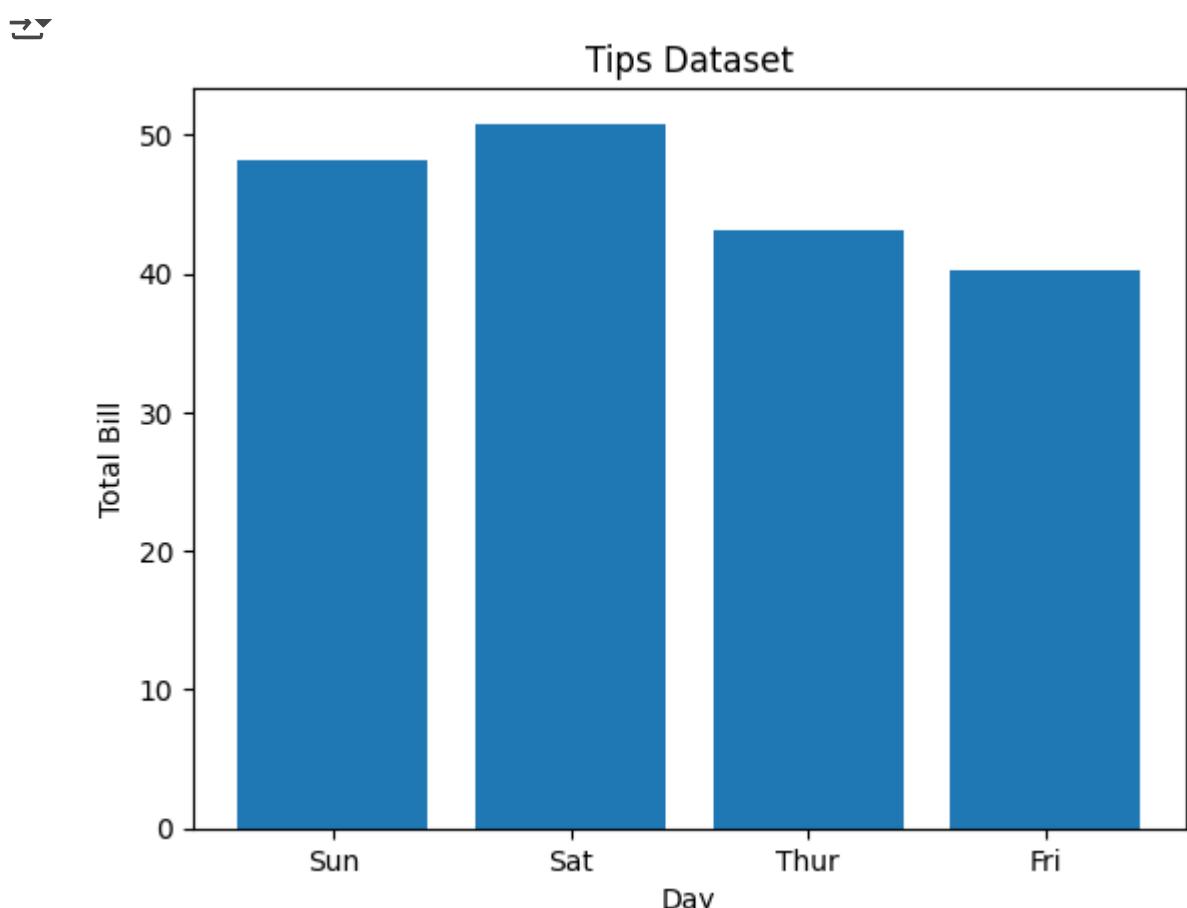
plt.bar(x, y)

plt.title("Tips Dataset")

plt.ylabel('Total Bill')

plt.xlabel('Day')

plt.show()
```



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

data = pd.read_csv('tip.csv')

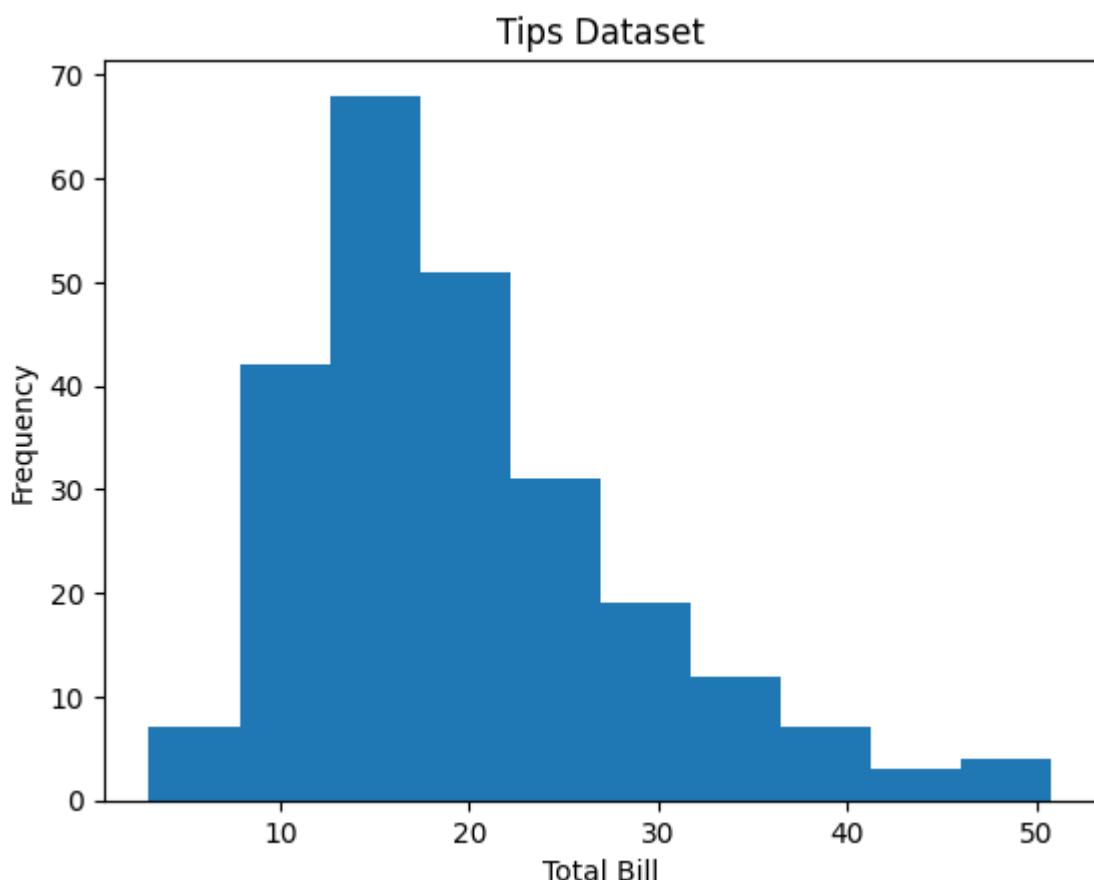
x = data['total_bill']
```

```
plt.hist(x)

plt.title("Tips Dataset")
2)
plt.ylabel('Frequency')

plt.xlabel('Total Bill')

plt.show()
```



New Section

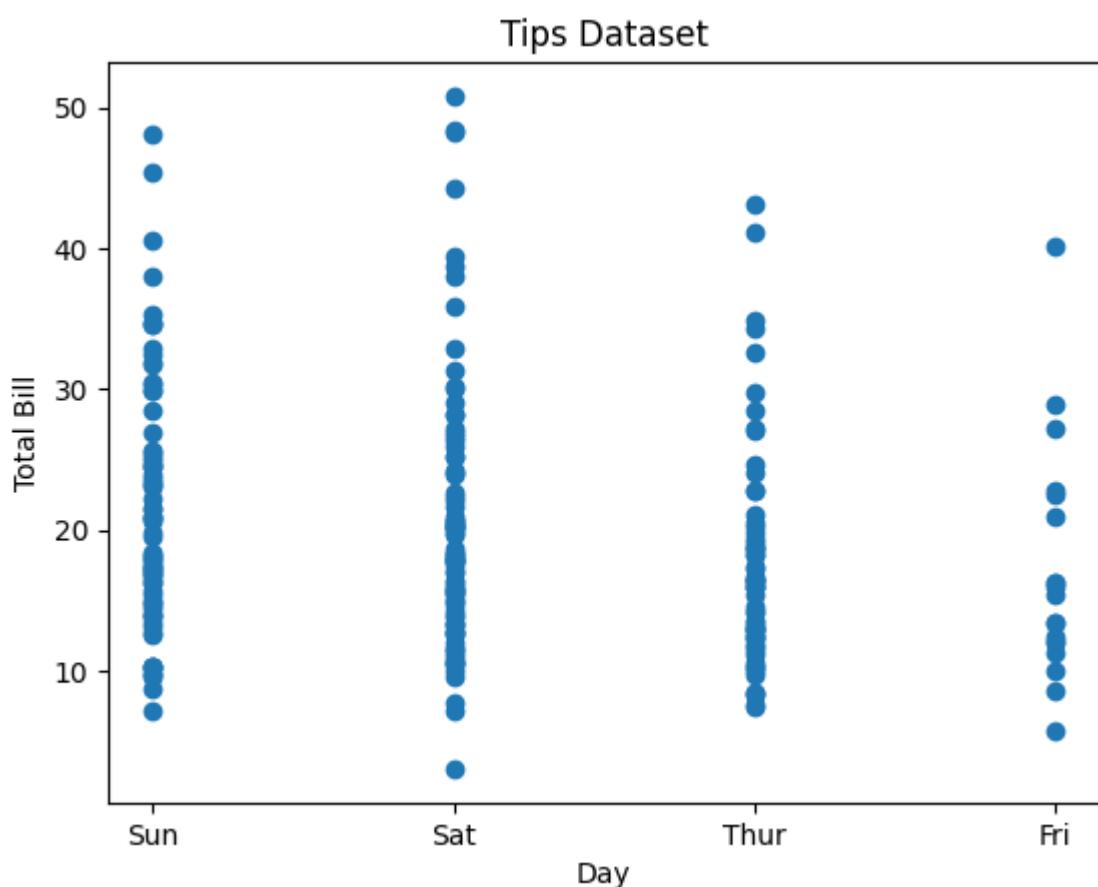
▼ New Section

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

data = pd.read_csv('tip.csv')

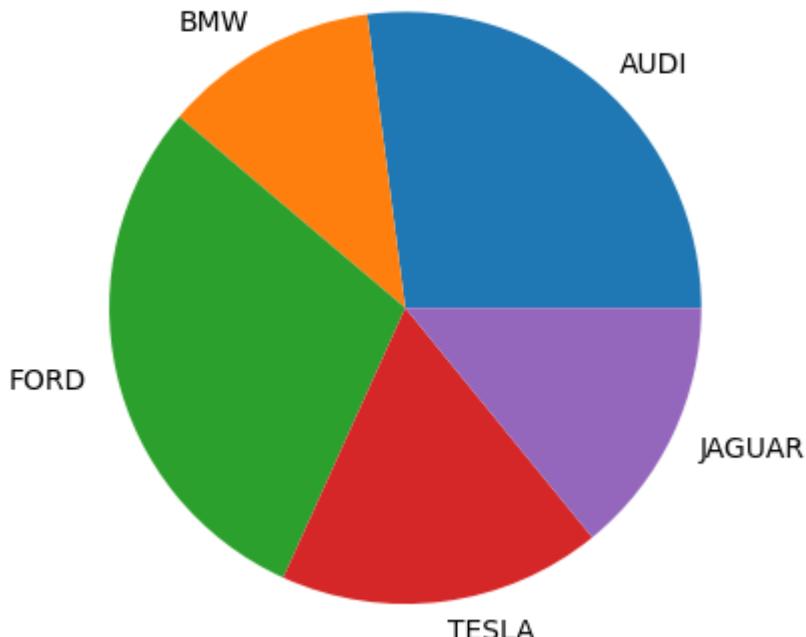
x = data['day']
y = data['total_bill']
```

```
plt.scatter(x, y)  
plt.title("Tips Dataset")  
plt.ylabel('Total Bill')  
plt.xlabel('Day')  
plt.show()
```



```
import matplotlib.pyplot as plt  
import pandas as pd  
  
data = pd.read_csv('tip.csv')  
  
cars = ['AUDI', 'BMW', 'FORD',  
        'TESLA', 'JAGUAR',]  
data = [23, 10, 25, 15, 12]  
  
plt.pie(data, labels=cars)  
  
plt.title("Car data")  
plt.show()
```

Car data

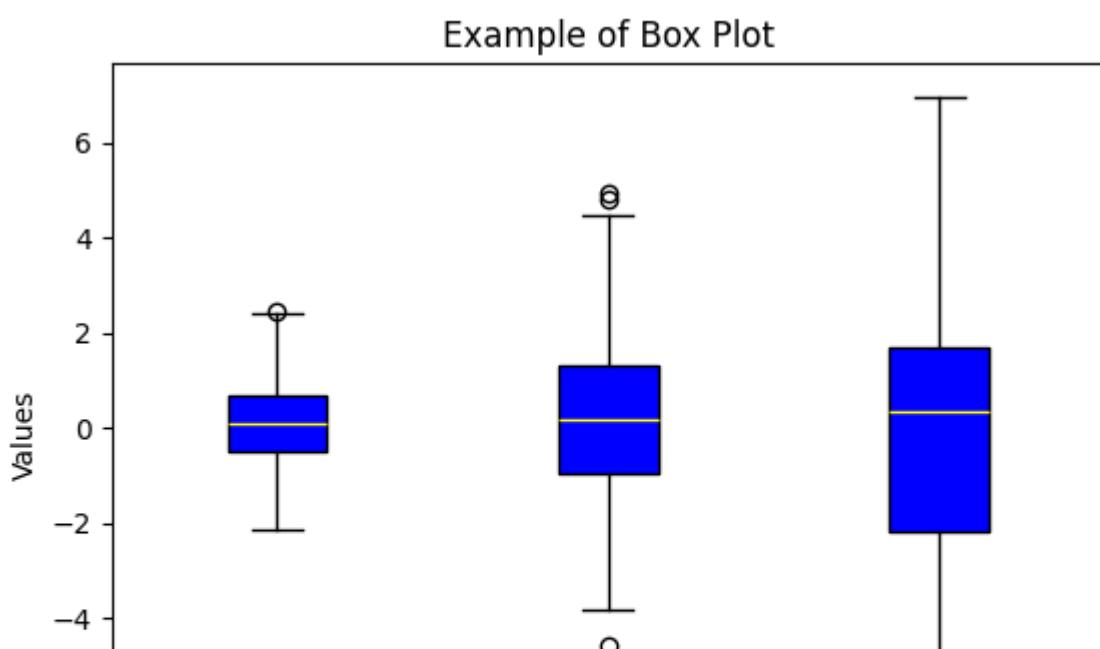


```
import matplotlib.pyplot as plt
import numpy as np

np.random.seed(10)
data = [np.random.normal(0, std, 100) for std in range(1, 4)]

# Create a box plot
plt.boxplot(data, vert=True, patch_artist=True,
            boxprops=dict(facecolor='blue'),
            medianprops=dict(color='yellow'))

plt.xlabel('Data Set')
plt.ylabel('Values')
plt.title('Example of Box Plot')
plt.show()
```





```
import matplotlib.pyplot as plt
import numpy as np

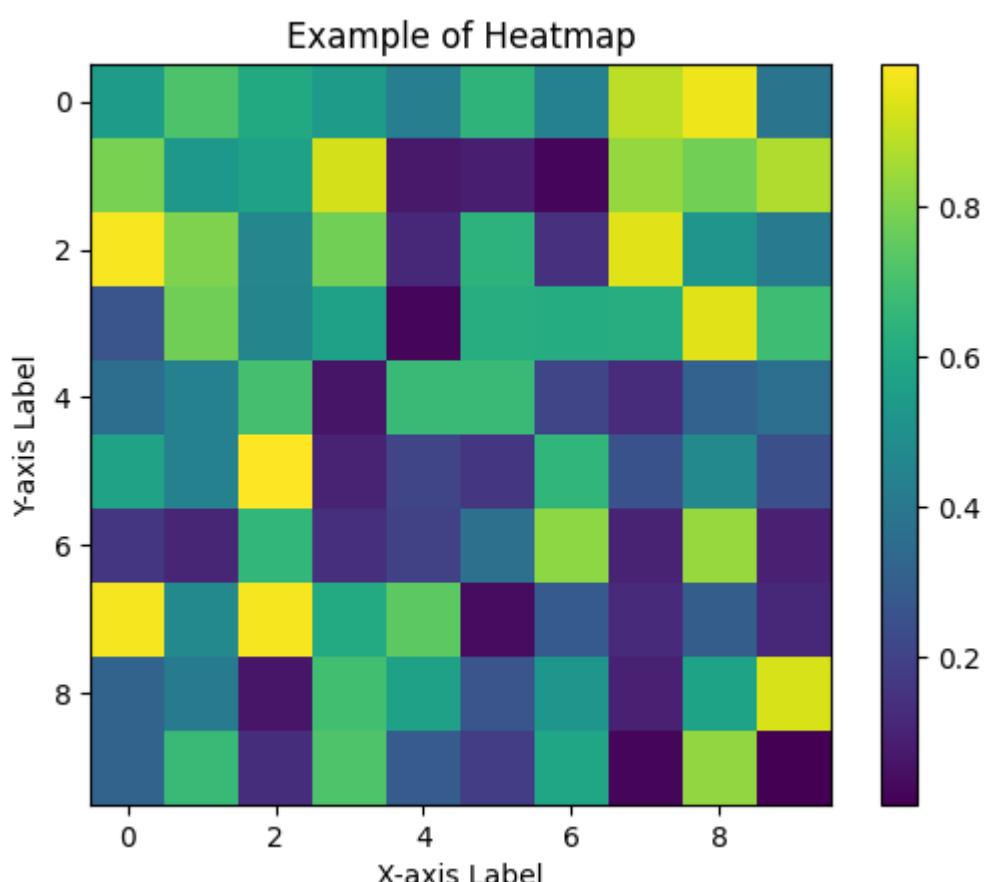
np.random.seed(0)
data = np.random.rand(10, 10)

plt.imshow(data, cmap='viridis', interpolation='nearest')

plt.colorbar()

plt.xlabel('X-axis Label')
plt.ylabel('Y-axis Label')
plt.title('Example of Heatmap')

plt.show()
```



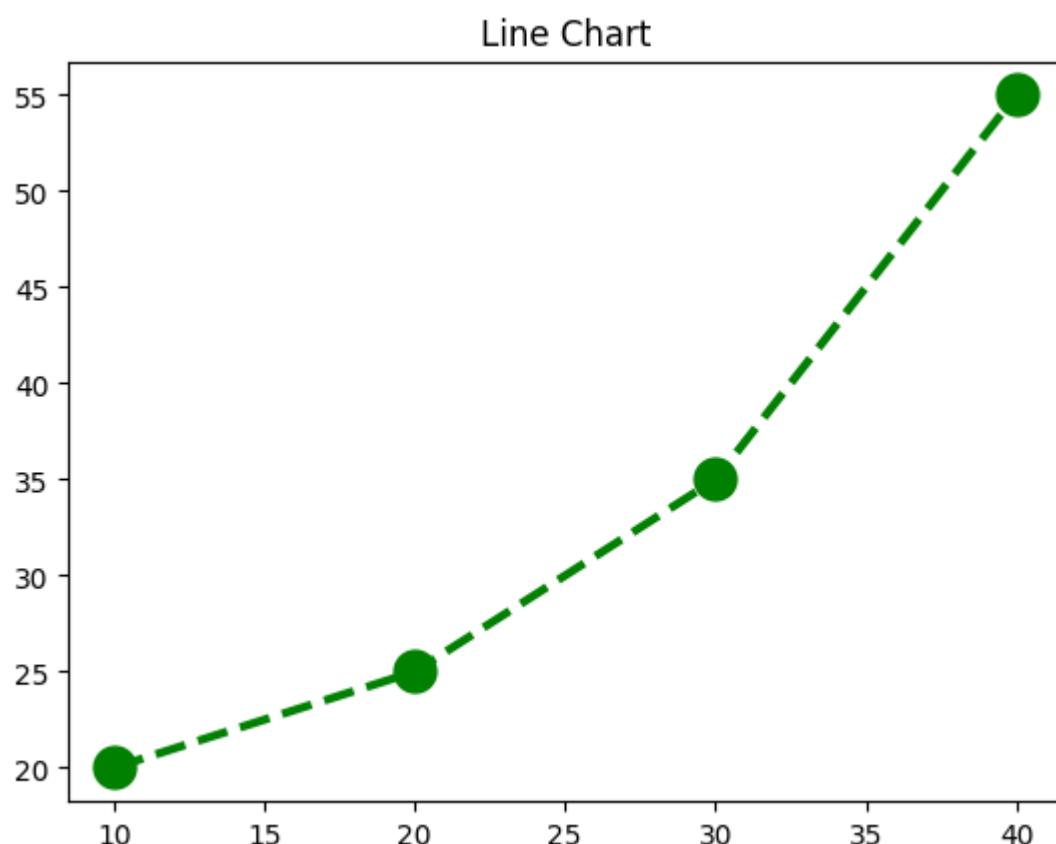
```
import matplotlib.pyplot as plt

x = [10, 20, 30, 40]
y = [20, 25, 35, 55]
```

```
plt.plot(x, y, color='green', linewidth=3, marker='o',
          markersize=15, linestyle='--')

plt.title("Line Chart")
```

Text(0.5, 1.0, 'Line Chart')



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Practical 3:- Python program to display multiple types of charts using Matplotlib package

```
pip install matplotlib
```

```
→ Requirement already satisfied: matplotlib in /usr/local/lib/python3.11/dist-  
Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.1  
Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.11/di  
Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3.  
Requirement already satisfied: kiwisolver>=1.3.1 in /usr/local/lib/python3.  
Requirement already satisfied: numpy>=1.23 in /usr/local/lib/python3.11/dis  
Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.11  
Requirement already satisfied: pillow>=8 in /usr/local/lib/python3.11/dist-  
Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.1  
Requirement already satisfied: python-dateutil>=2.7 in /usr/local/lib/pytho  
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.11/dist-p
```

```
import matplotlib  
print(matplotlib.__version__)
```

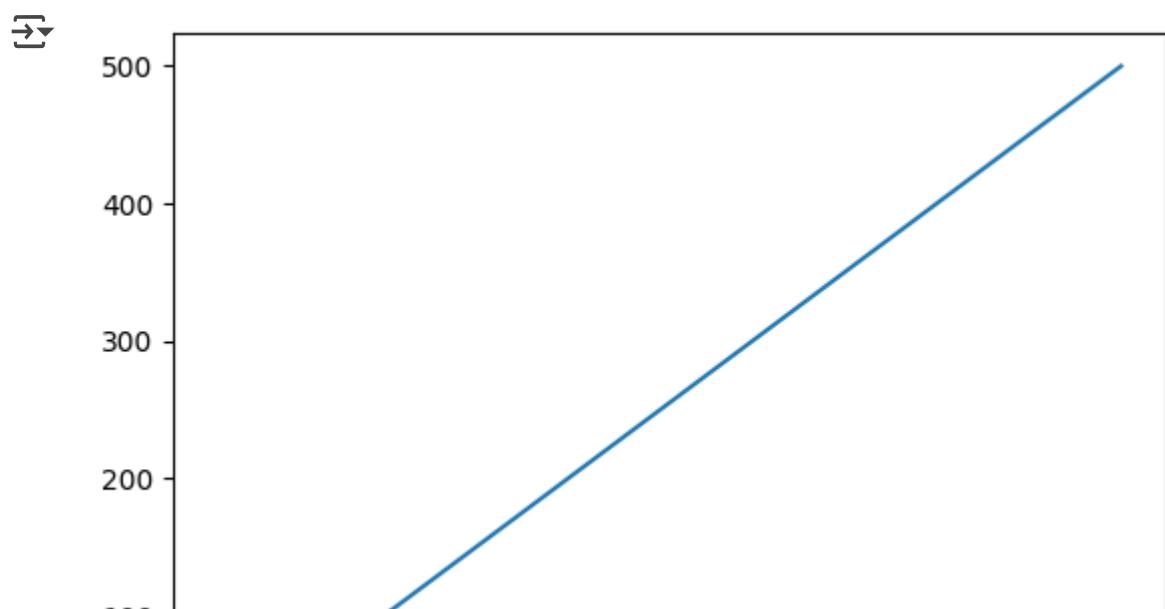
```
→ 3.10.0
```

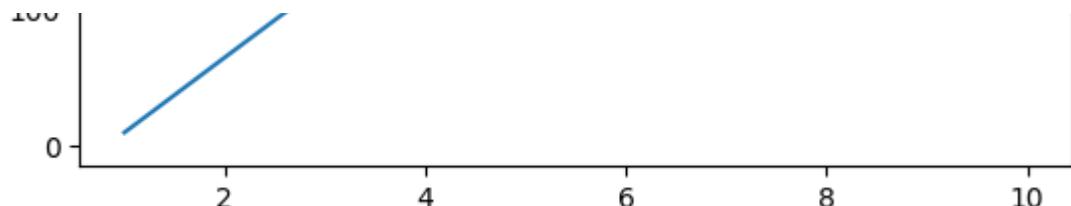
```
import matplotlib.pyplot as plt  
import numpy as np
```

```
import matplotlib.pyplot as plt  
import numpy as np
```

```
xpoints = np.array([1, 10])  
y whole points = np.array([10, 500])
```

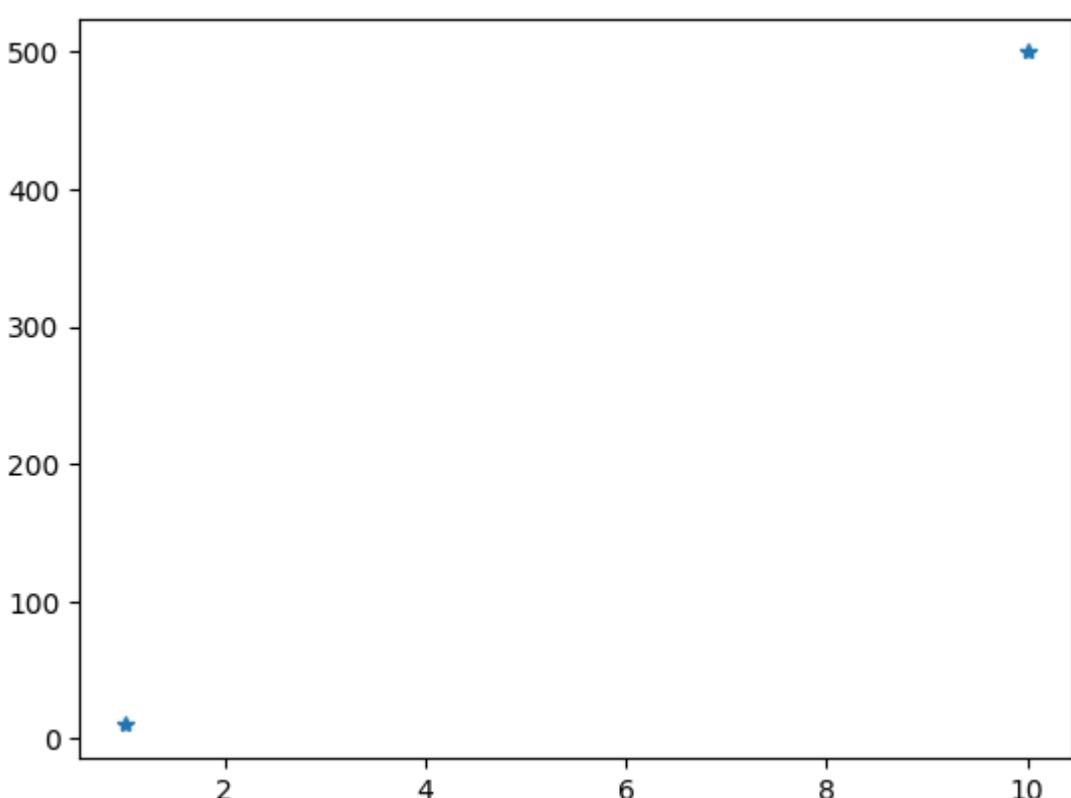
```
plt.plot(xpoints, y whole points)  
plt.show()
```





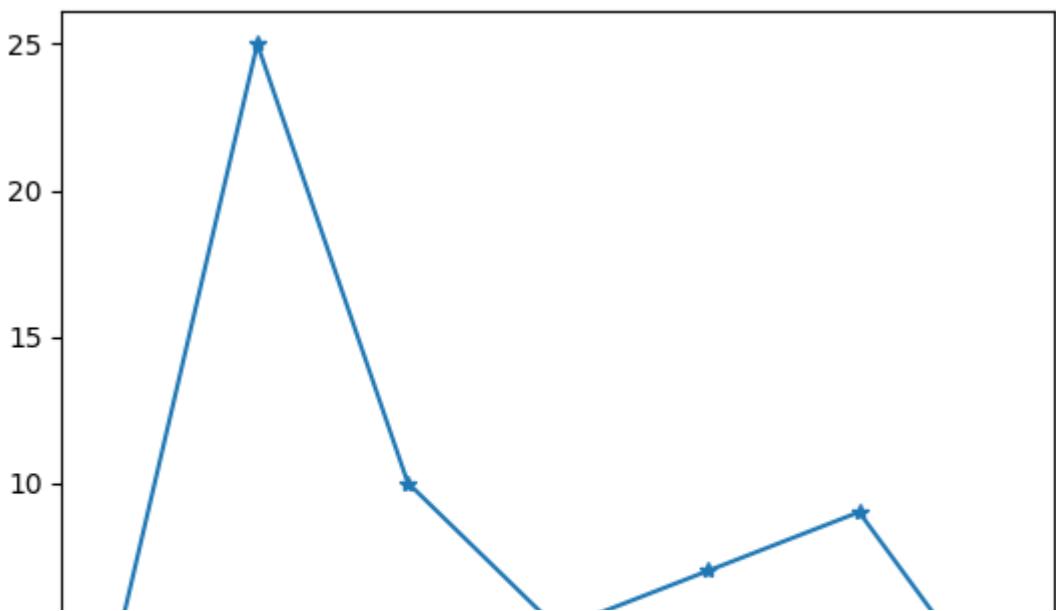
```
plt.plot(xpoints, ypoints, '*')
```

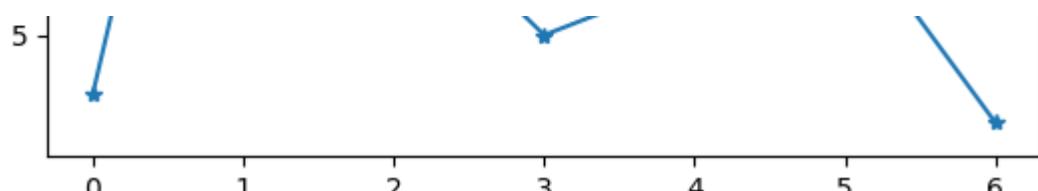
```
↳ [〈matplotlib.lines.Line2D at 0x7ff308774690〉]
```



```
ypoints = np.array([3, 25, 10, 5, 7, 9, 2])
plt.plot(ypoints, marker = '*')
plt.show()
```

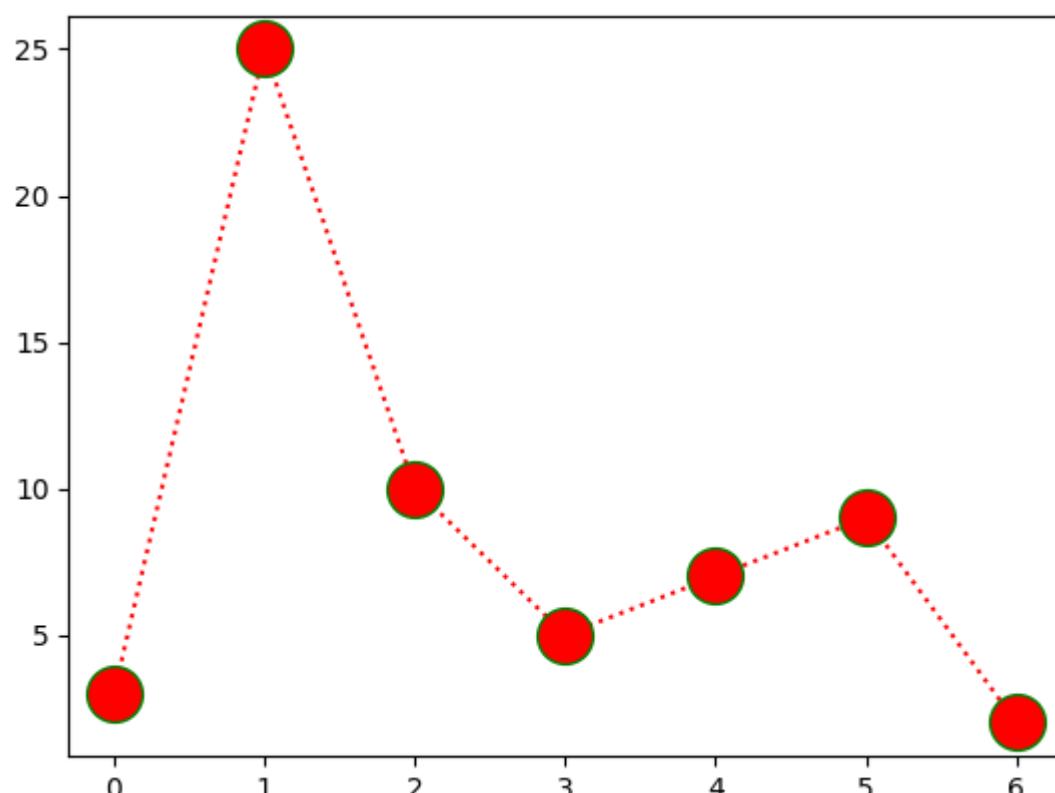
```
↳
```





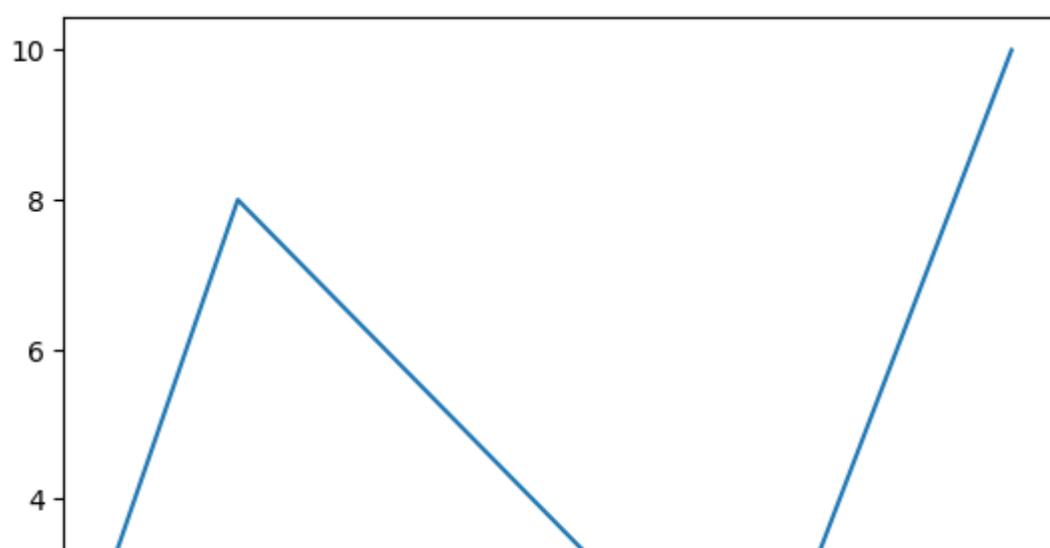
```
plt.plot(ypoints, 'o:r' ,ms = 20, mec = 'g')
```

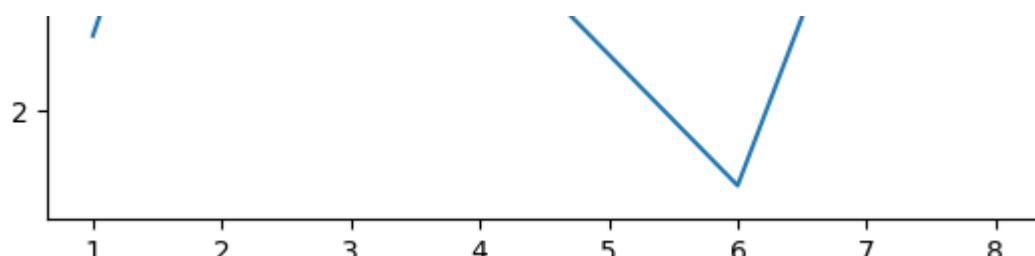
```
[<matplotlib.lines.Line2D at 0x7ff325206150>]
```



```
xpoints = np.array([1, 2, 6, 8])  
ypoints = np.array([3, 8, 1, 10])
```

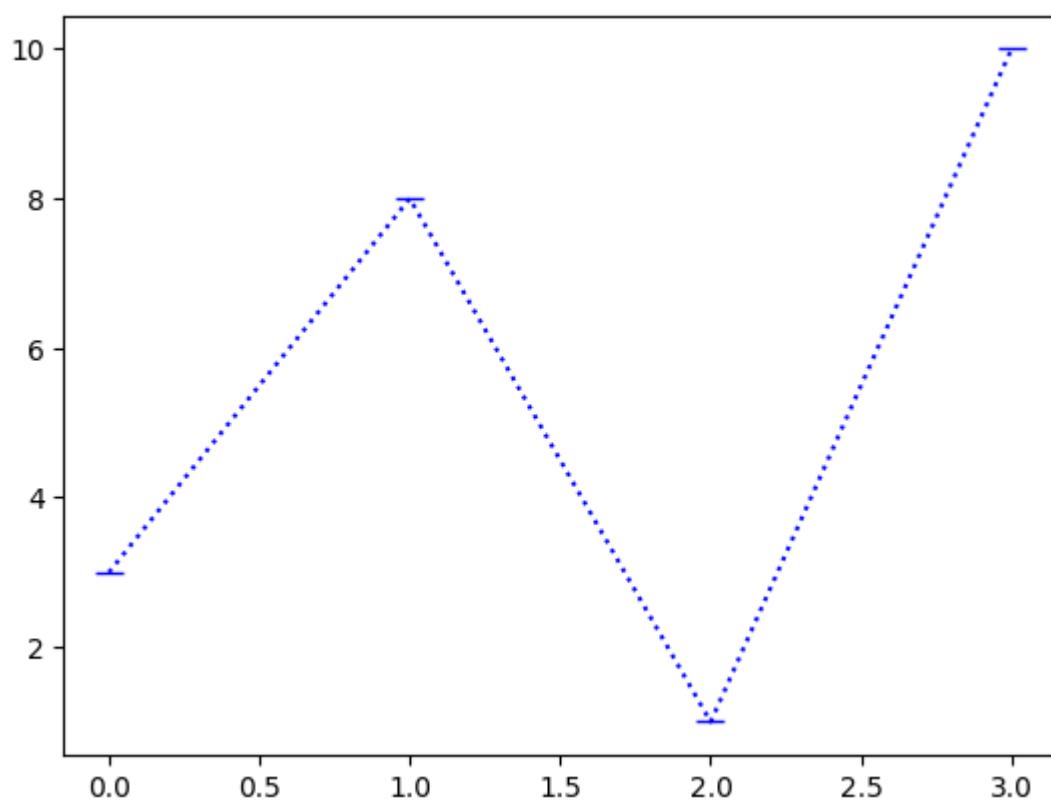
```
plt.plot(xpoints, ypoints)  
plt.show()
```





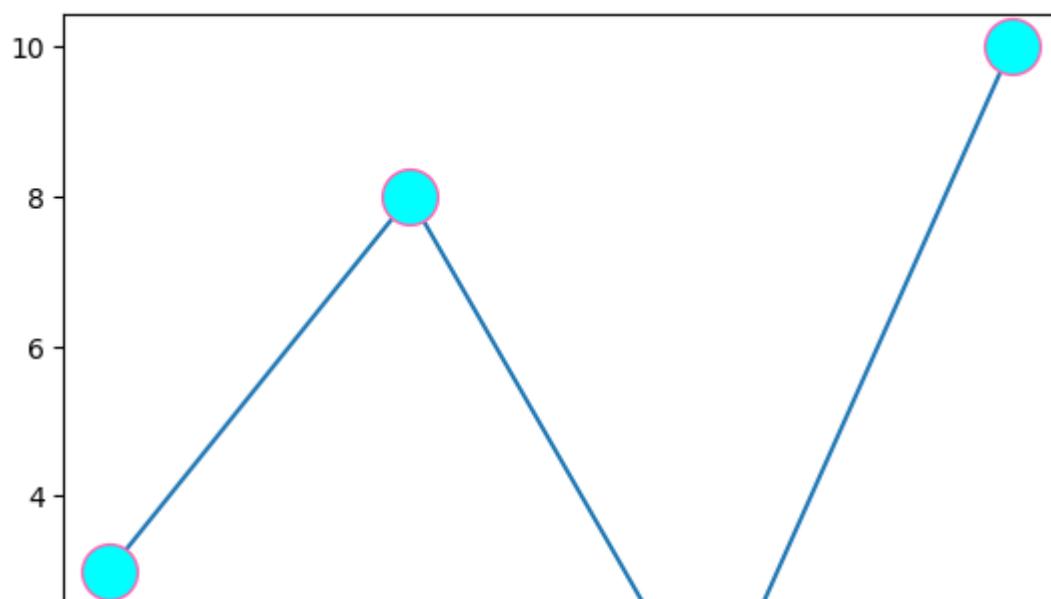
```
plt.plot(ypoints, '_:b' , ms = 10)
```

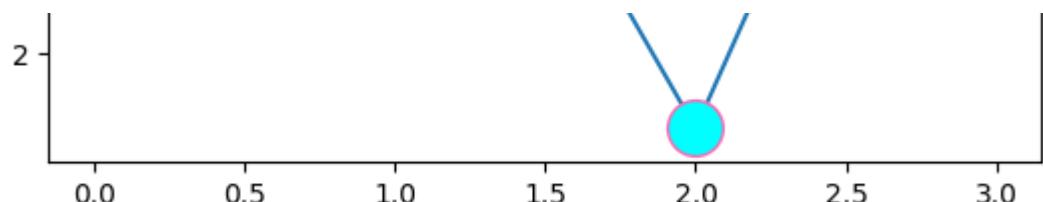
```
[<matplotlib.lines.Line2D at 0x7ff308661ad0>]
```



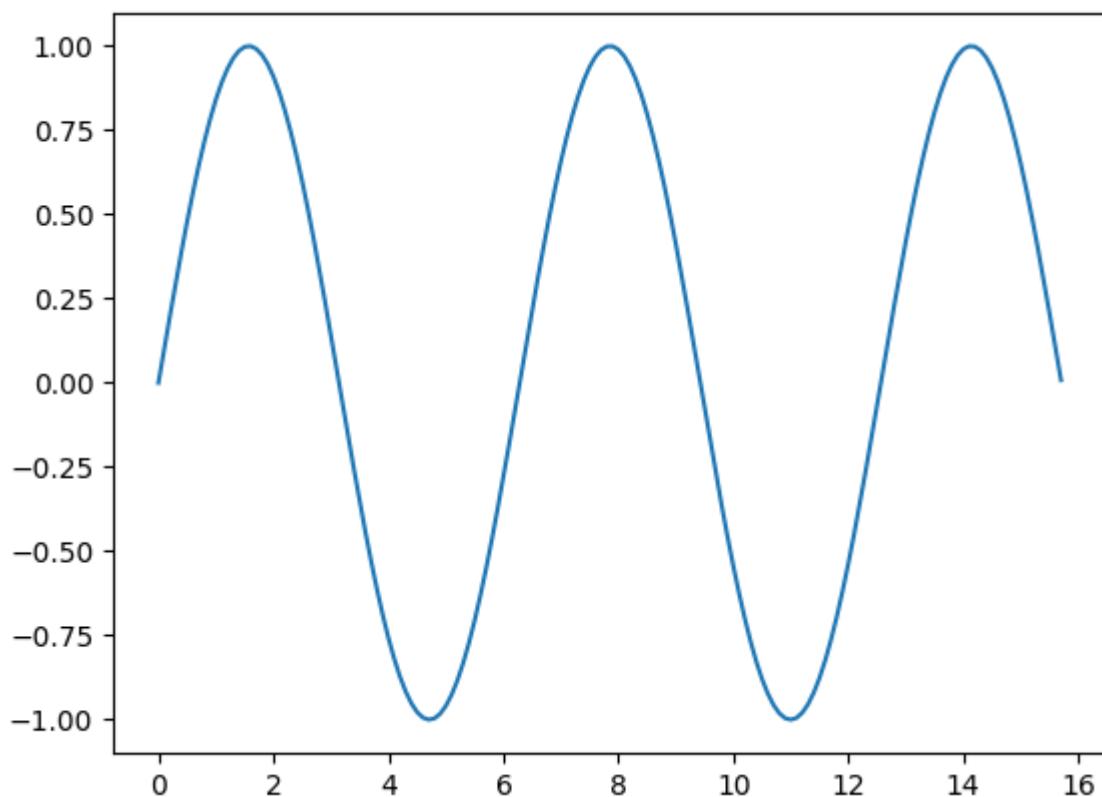
```
plt.plot(ypoints, marker = 'o', ms = 20, mec = 'hotpink', mfc = 'cyan')
```

```
[<matplotlib.lines.Line2D at 0x7ff3086efe10>]
```



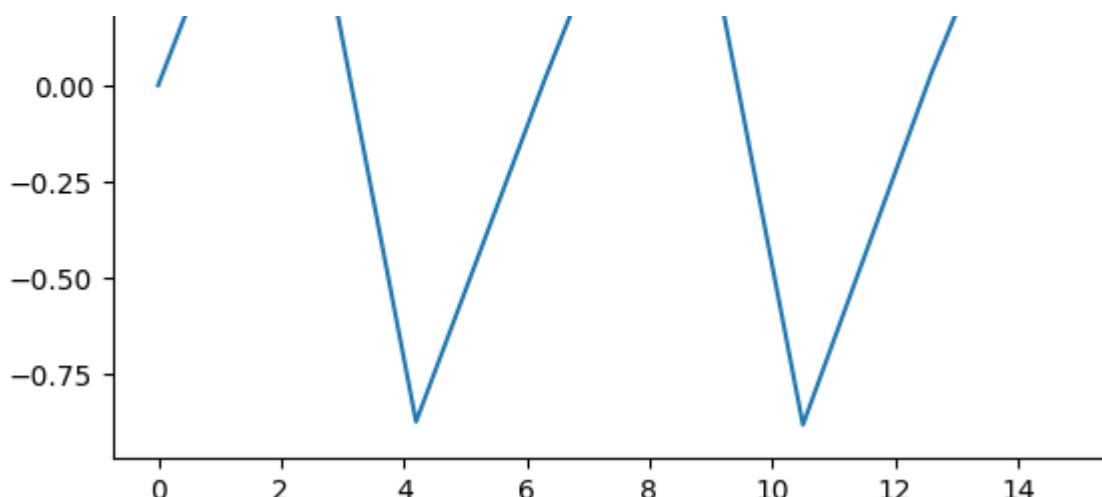


```
import matplotlib.pyplot as plt
import numpy as np
x=np.arange(0,5*np.pi,0.1);
y=np.sin(x);
plt.plot(x,y);
plt.show;
```

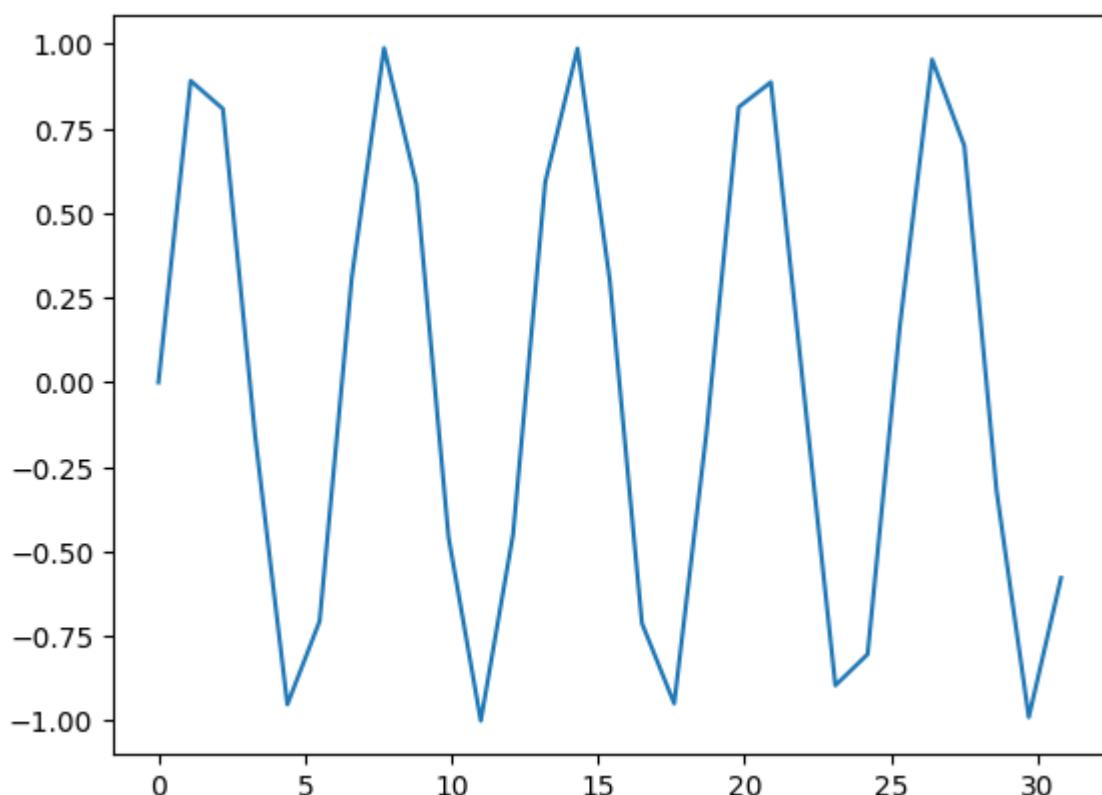


```
x=np.arange(0,5*np.pi,2.1);
y=np.sin(x);
plt.plot(x,y);
plt.show;
```



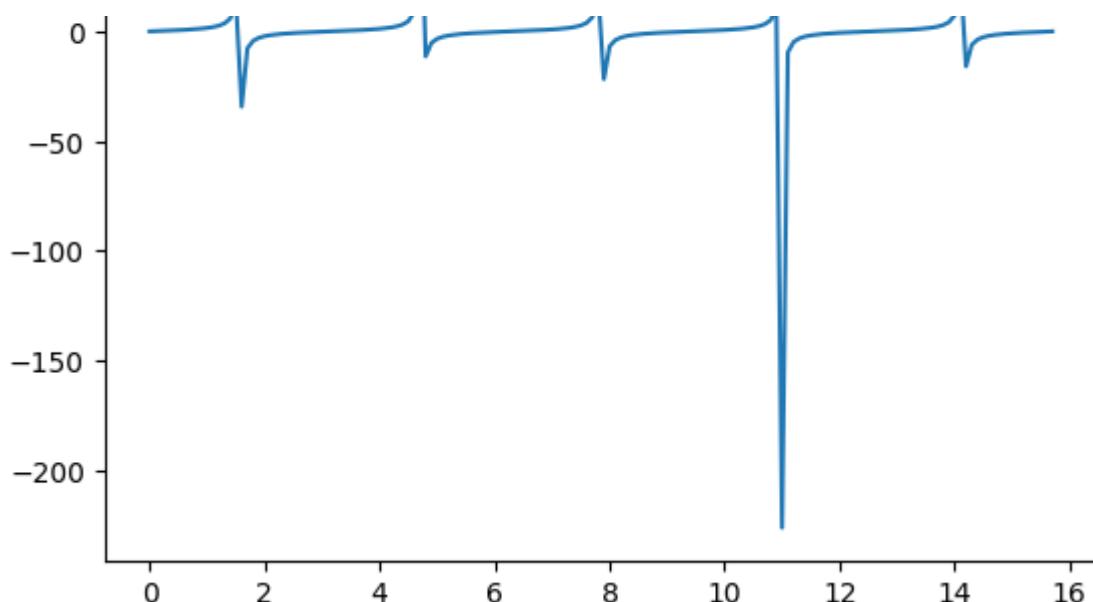


```
x=np.arange(0,10*np.pi,1.1);
y=np.sin(x);
plt.plot(x,y);
plt.show;
```



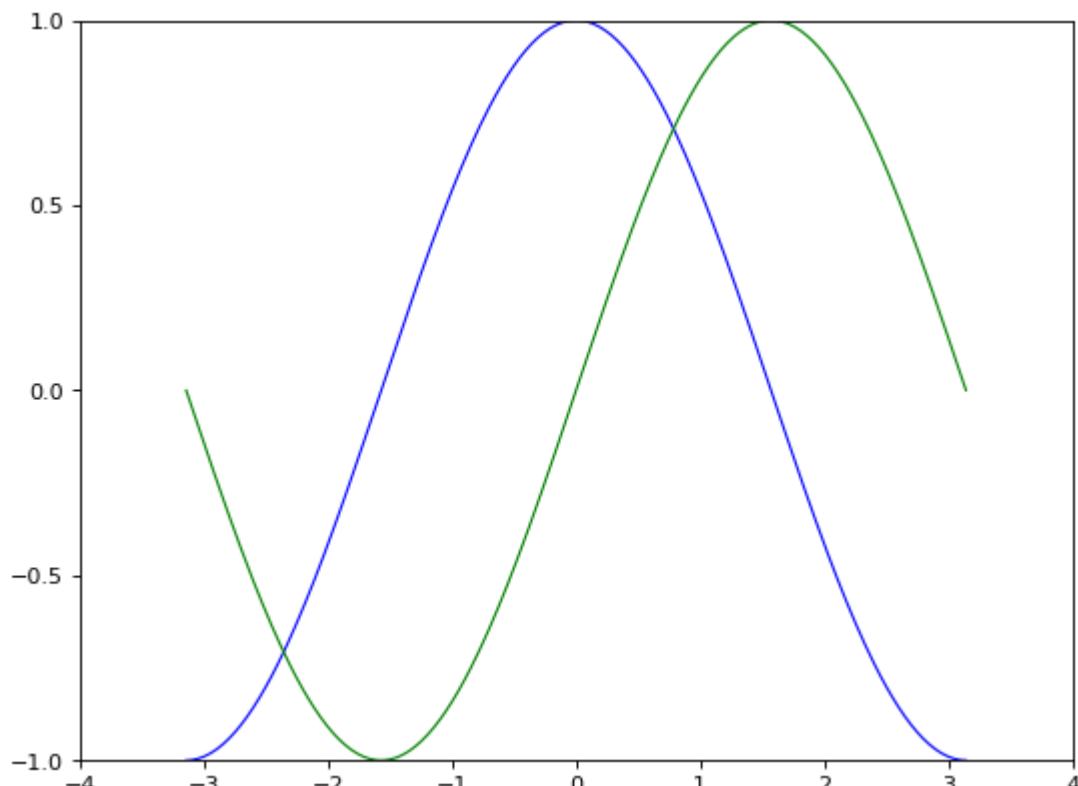
```
x=np.arange(0,5*np.pi,0.1);
y=np.tan(x);
plt.plot(x,y);
plt.show;
```





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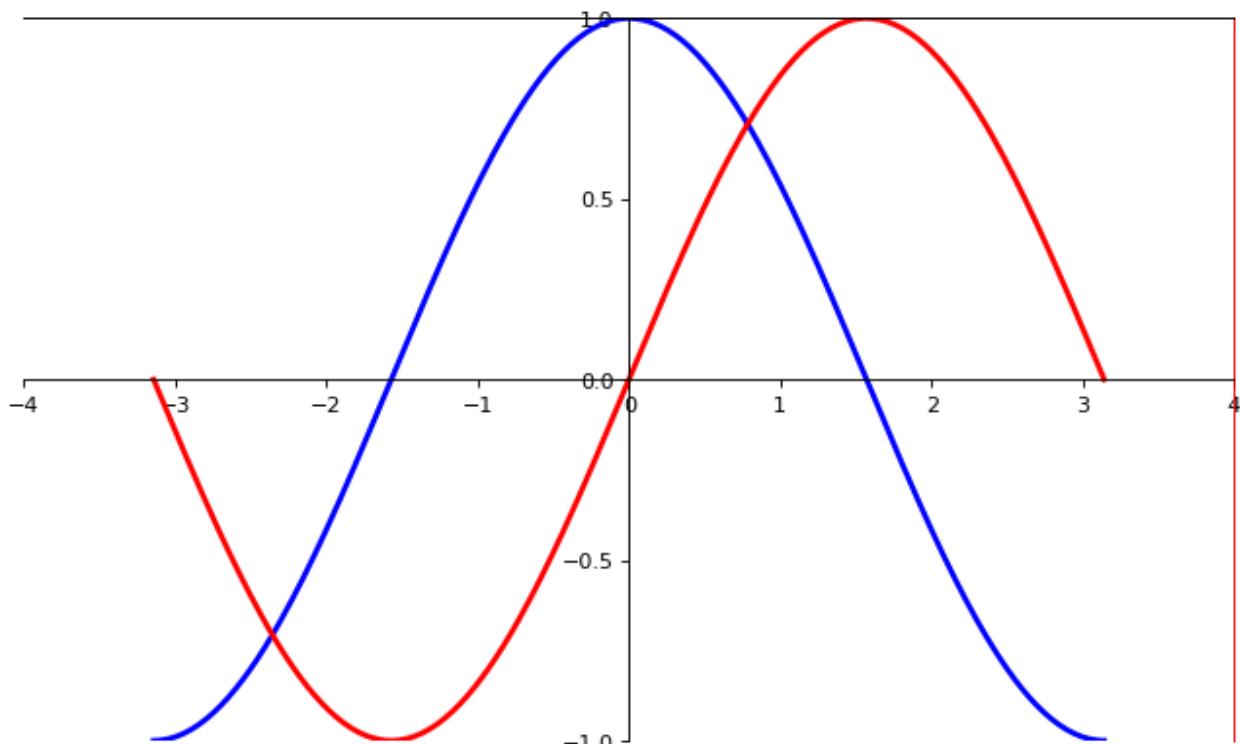
```
plt.figure(figsize=(8,6), dpi=80)
plt.subplot(1, 1, 1)
X = np.linspace(-np.pi, np.pi, 256)
C, S = np.cos(X), np.sin(X)
plt.plot(X, C, color="blue", linewidth=1.0, linestyle="--")
plt.plot(X, S, color="green", linewidth=1.0, linestyle="--")
plt.xlim(-4.0, 4.0)
plt.xticks(np.linspace(-4, 4, 9))
plt.ylim(-1.0, 1.0)
plt.yticks(np.linspace(-1, 1, 5))
plt.show()
```



```
plt.figure(figsize=(10, 6), dpi=80)
plt.subplot(1, 1, 1)
X = np.linspace(-np.pi, np.pi, 256)
C, S = np.cos(X), np.sin(X)
plt.plot(X, C, color="blue", linewidth=2.5, linestyle="--")
plt.plot(X, S, color="red", linewidth=2.5, linestyle="--")
plt.xlim(-4.0, 4.0)
plt.xticks(np.linspace(-4, 4, 9))
plt.ylim(-1.0, 1.0)
plt.yticks(np.linspace(-1, 1, 5))

ax = plt.gca() # gca stands for get current axis;
ax.spines['right'].set_color('red')
ax.spines['top'].set_color('black')
ax.xaxis.set_ticks_position('bottom')
ax.spines['bottom'].set_position(('data',0))
ax.yaxis.set_ticks_position('left')
ax.spines['left'].set_position(('data',0))

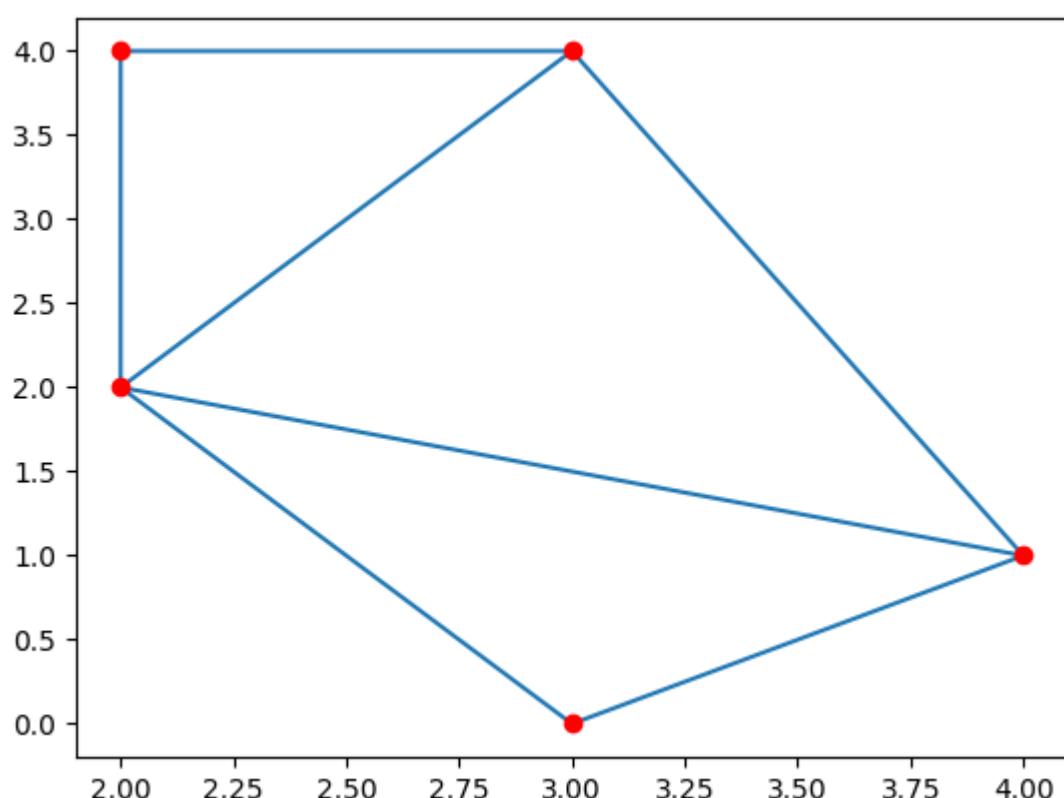
plt.show()
```



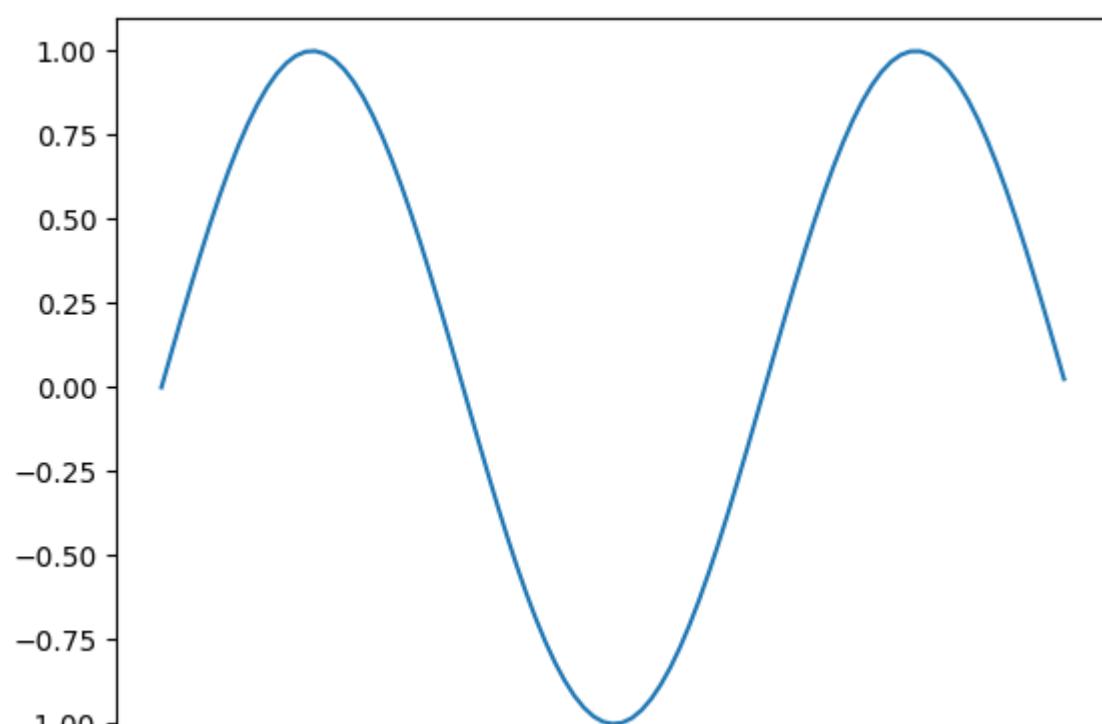
```
import numpy as np
from scipy.spatial import Delaunay
import matplotlib.pyplot as plt

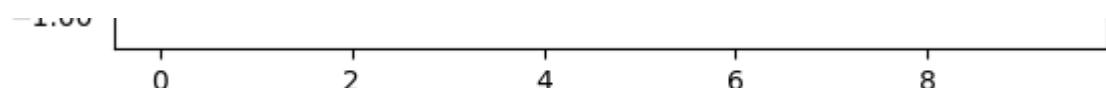
points= np.array([
    [0, 1], [0, -1], [1, 0], [-1, 0], [0.5, 0.866], [-0.5, 0.866], [-0.866, 0.5], [0.866, 0.5], [0.5, -0.866], [-0.5, -0.866], [-0.866, -0.5], [0.866, -0.5]
])
```

```
[12, 4], [13, 4], [13, 5], [12, 2], [14, 1]]  
])  
simplices = Delaunay(points).simplices  
plt.triplot(points[:,0],points[:,1],simplices)  
plt.scatter(points[:,0],points[:,1],color='r')  
plt.show()
```

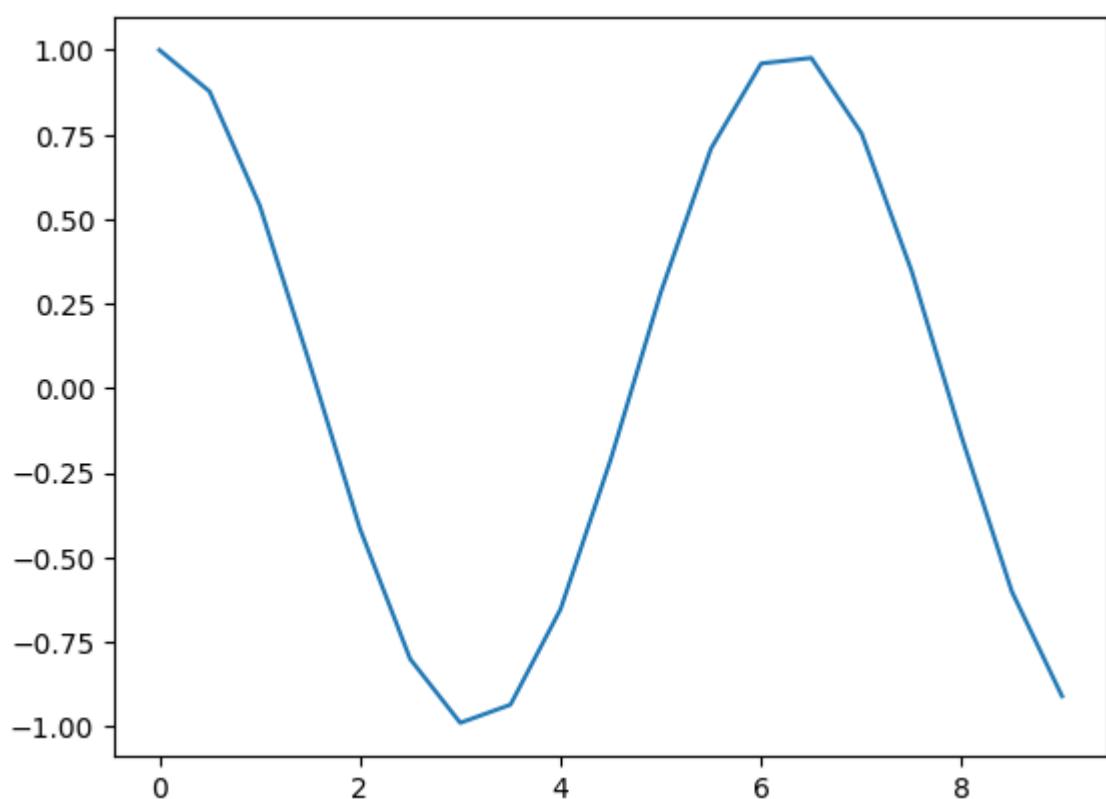


```
x=np.arange(0,3* np.pi, 0.1)  
y=np.sin(x)  
plt.plot(x,y)  
plt.show;
```

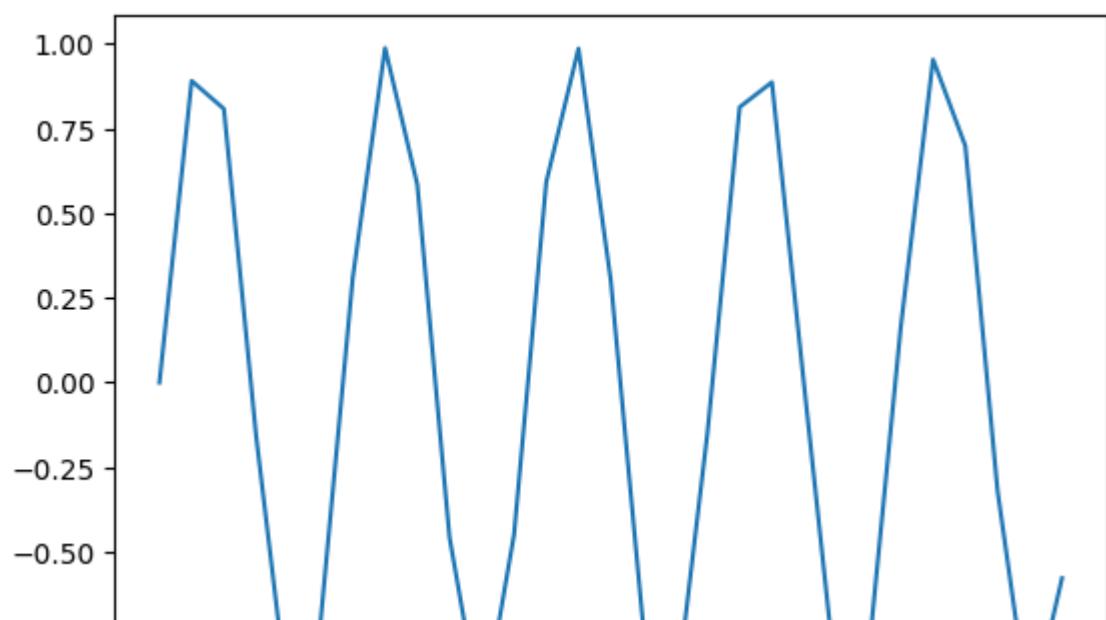


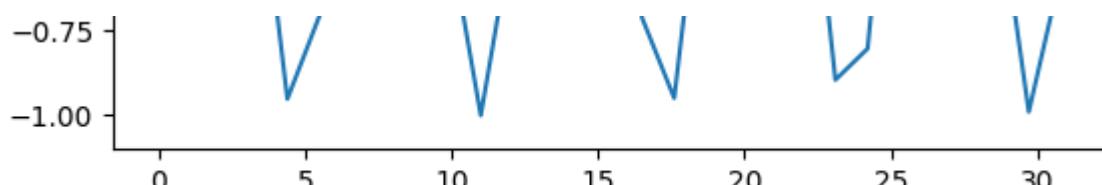


```
x=np.arange(0,3* np.pi, 0.5)
y=np.cos(x)
plt.plot(x,y)
plt.show()
```

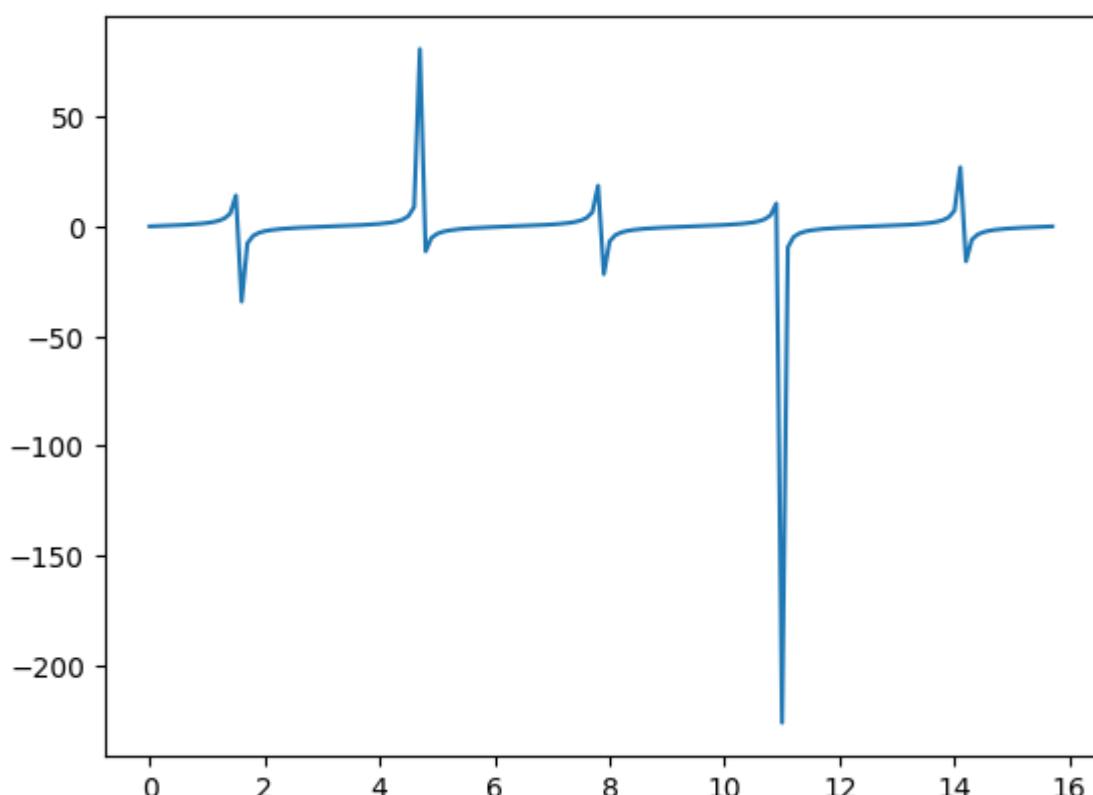


```
x=np.arange(0,10*np.pi,1.1);
y=np.sin(x);
plt.plot(x,y);
plt.show;
```

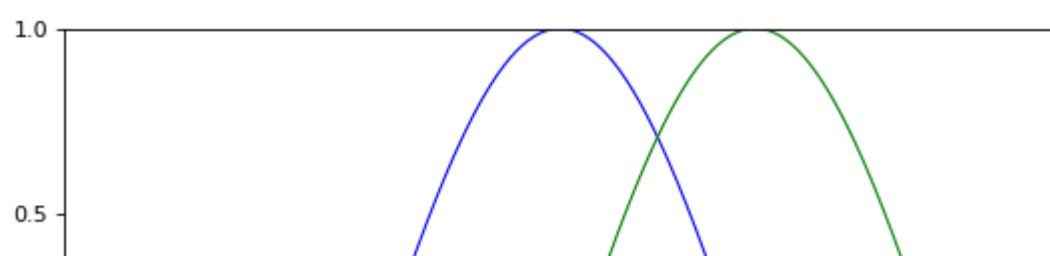


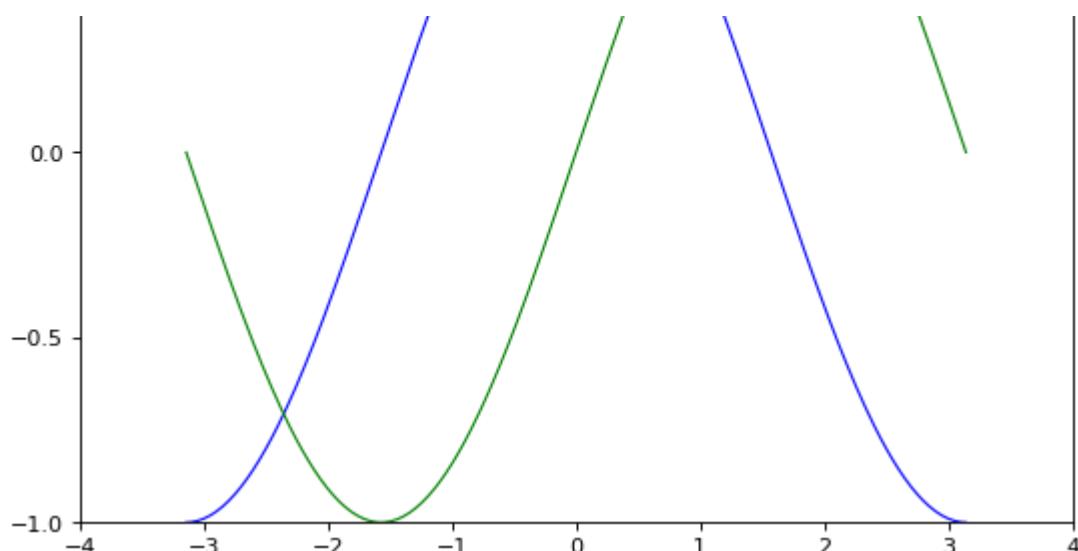


```
x=np.arange(0,5*np.pi, 0.1)
y=np.tan(x)
plt.plot(x,y)
plt.show()
```



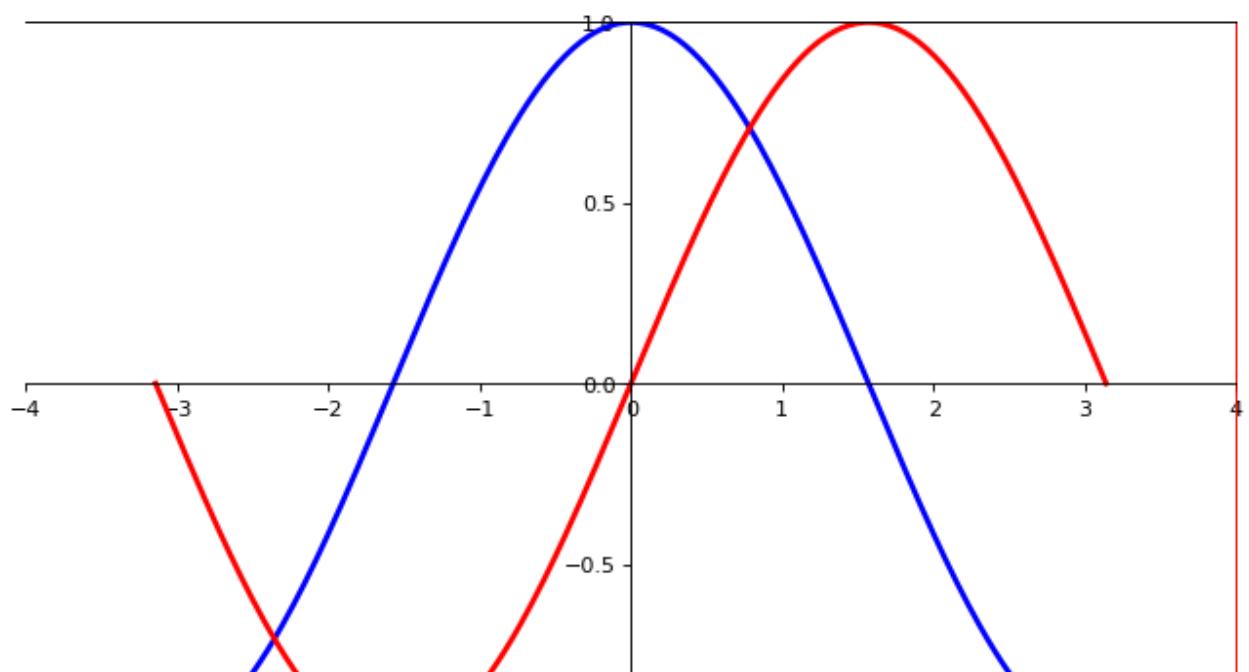
```
plt.figure(figsize=(8,6), dpi=80)
plt.subplot(1,1,1)
X= np.linspace(-np.pi, np.pi,256)
C,S= np.cos(X),np.sin(X)
plt.plot(X,C, color='blue', linewidth=1.0, linestyle='--')
plt.plot(X,S,color='green', linewidth=1.0, linestyle='--')
plt.xlim(-4.0,4.0)
plt.xticks(np.linspace(-4,4,9))
plt.ylim(-1.0,1.0)
plt.yticks(np.linspace(-1,1,5))
plt.show()
```





```
plt.figure(figsize=(10, 6), dpi=80)
plt.subplot(1, 1, 1)
X = np.linspace(-np.pi, np.pi, 256)
C, S = np.cos(X), np.sin(X)
plt.plot(X, C, color="blue", linewidth=2.5, linestyle="--")
plt.plot(X, S, color="red", linewidth=2.5, linestyle="--")
plt.xlim(-4.0, 4.0)
plt.xticks(np.linspace(-4, 4, 9))
plt.ylim(-1.0, 1.0)
plt.yticks(np.linspace(-1, 1, 5))

ax = plt.gca() # gca stands for get current axis;
ax.spines['right'].set_color('red')
ax.spines['top'].set_color('black')
ax.xaxis.set_ticks_position('bottom')
ax.spines['bottom'].set_position(('data',0))
ax.yaxis.set_ticks_position('left')
ax.spines['left'].set_position(('data',0))
plt.show()
```



Practical - 4

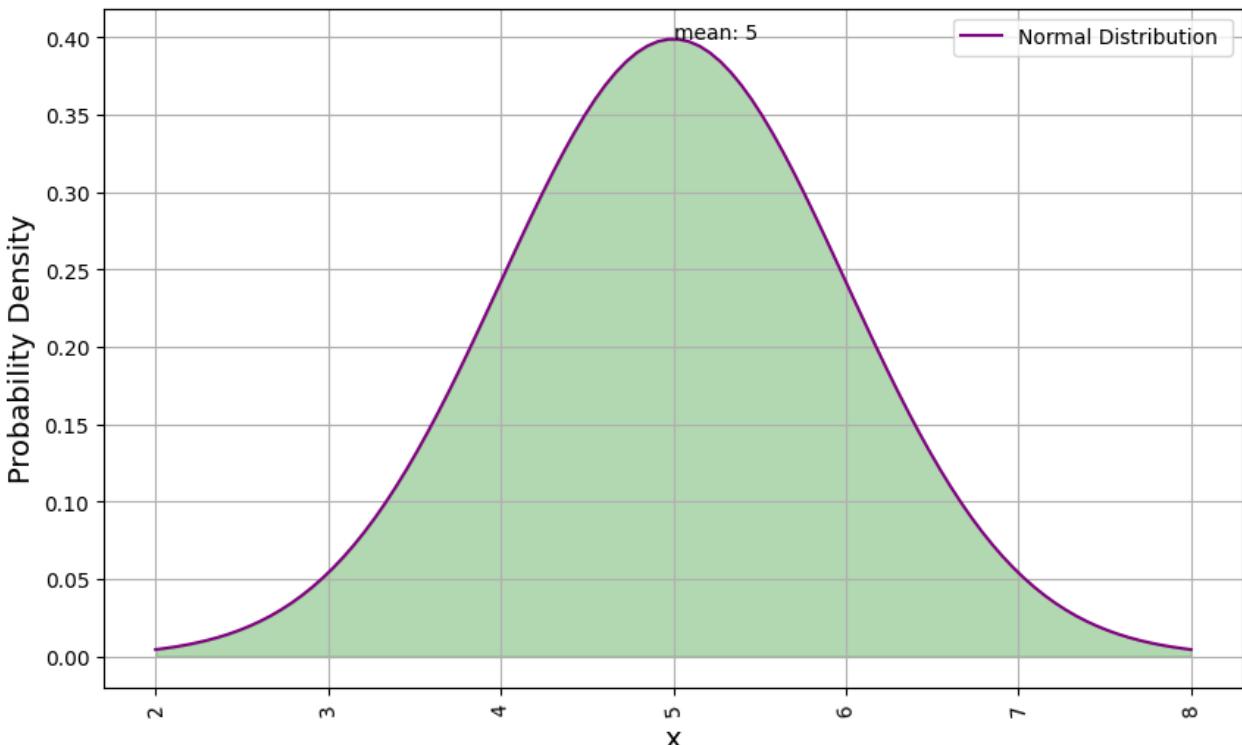
Aim: Create a python program that generates and display a normal distribution curve.

```
mean = 5
sigma = 1
x = np.linspace(mean - 3*sigma, mean + 3*sigma, 100)
y = norm.pdf(x, mean , sigma)
plt.figure(figsize=(10, 6))
plt.plot(x, y, color='purple', label='Normal Distribution Curve')
plt.title('Normal Distribution Curve', fontsize=16)
plt.xlabel('x', fontsize=14)
plt.ylabel('Probability Density', fontsize=14)
plt.grid(True)
plt.legend(['Normal Distribution'])

plt.fill_between(x, y, alpha=0.3,color='green')
plt.text(mean , max(y), f'mean: {mean}')
#plt.axvline(x=0, color='r', linestyle='--')
#plt.axhline(y=0.5, color='r', linestyle='--')
plt.xticks(rotation=95)
plt.show()
```



Normal Distribution Curve

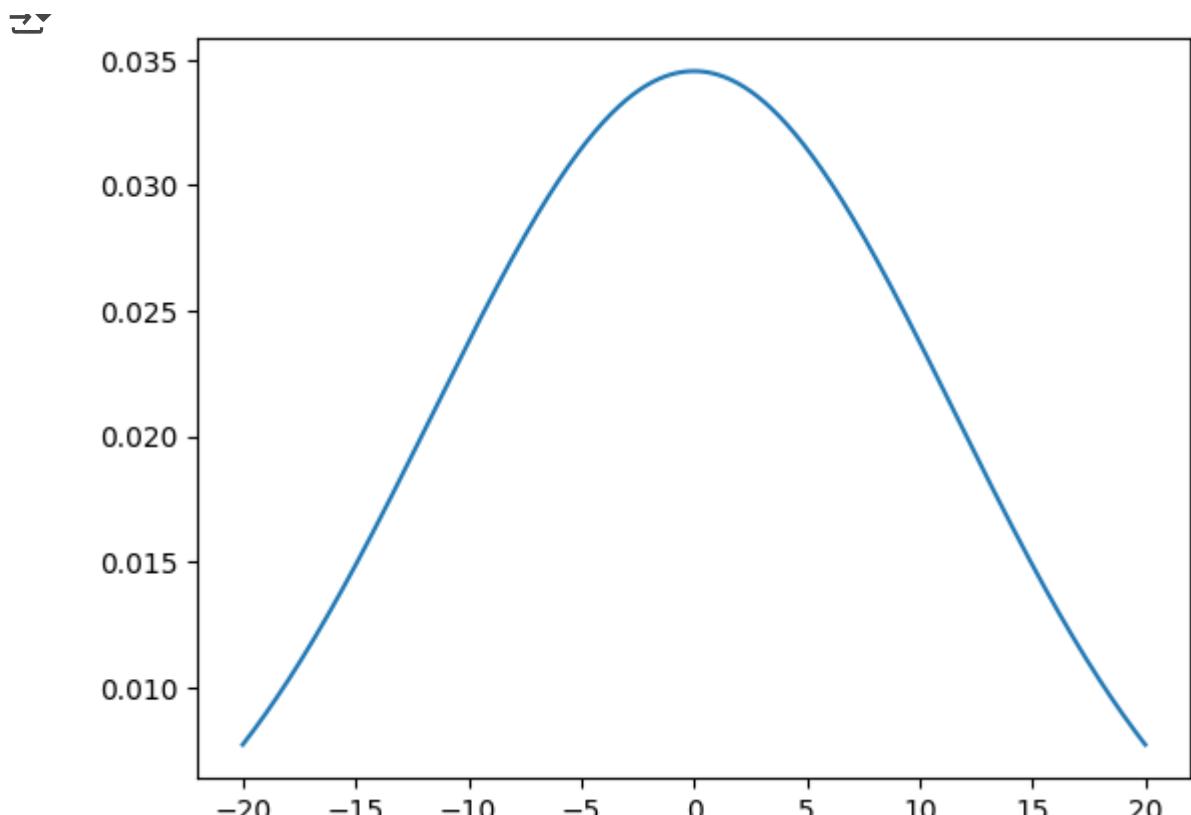


```
x = random.normal(size=(2,3))
print(x)
```

```
[[ 0.60454964  0.45127951  0.64501106]
 [-0.33336061  0.42675764  0.45312146]]
```

```
from numpy import random
import numpy as np
import matplotlib.pyplot as plt
from scipy.stats import norm
import statistics
import seaborn as sns

x_axis = np.arange(-20, 20, 0.01)
mean = statistics.mean(x_axis)
sd = statistics.stdev(x_axis)
plt.plot(x_axis, norm.pdf(x_axis, mean, sd))
plt.show()
```

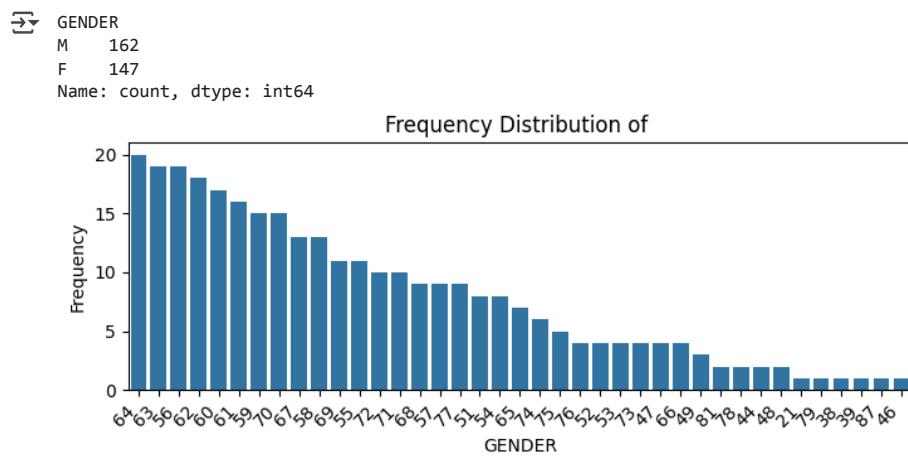


Practical 5: Write a Python program that calculates and displays a frequency distribution of a given dataset.

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

df = pd.read_csv('lung_cancer.csv')
frequency_distribution = df['GENDER'].value_counts()
print(frequency_distribution)

plt.figure(figsize=(7, 3))
sns.countplot(x='AGE', data=df, order=df['AGE'].value_counts().index)
plt.xticks(rotation=45, ha='right')
plt.title('Frequency Distribution of ')
plt.xlabel('GENDER')
plt.ylabel('Frequency')
plt.tight_layout()
plt.show()
```



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Practical 6: Write a Python program that computes the correlation between two sets of data and visualizes the relationship using a scatter plot.

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
df = pd.read_csv('Iris.csv')
print(df.head())

# Calculate the Pearson correlation coefficient between 'num_voted_users' and 'gross'
correlation = df['SepalLengthCm'].corr(df['PetalLengthCm'])

print(f'Pearson correlation coefficient: {correlation}')

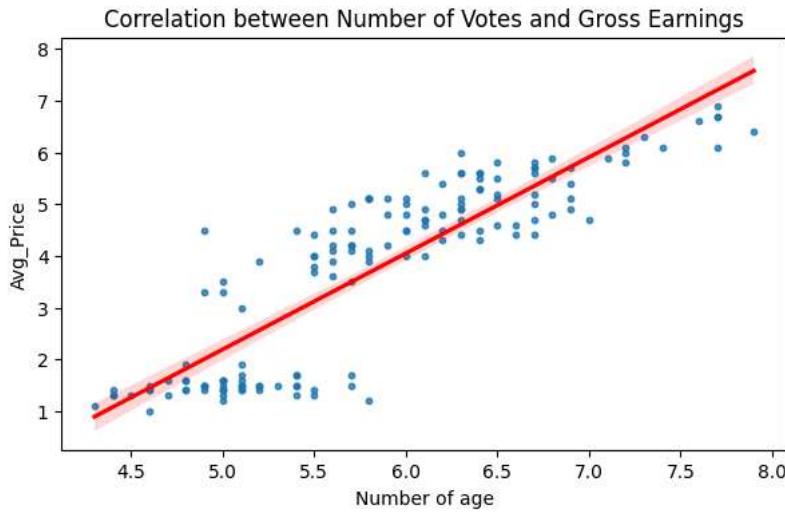
plt.figure(figsize=(7, 4))
sns.regplot(x='SepalLengthCm', y='PetalLengthCm', data=df, scatter_kws={'s': 10}, line_kws={'color': 'red'})

plt.title('Correlation between Number of Votes and Gross Earnings')
plt.xlabel('Number of age')
plt.ylabel('Avg_Price')

plt.show()
```

	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
0	1	5.1	3.5	1.4	0.2	Iris-setosa
1	2	4.9	3.0	1.4	0.2	Iris-setosa
2	3	4.7	3.2	1.3	0.2	Iris-setosa
3	4	4.6	3.1	1.5	0.2	Iris-setosa
4	5	5.0	3.6	1.4	0.2	Iris-setosa

Pearson correlation coefficient: 0.8717541573048718



Practical 7: Develop a Python program to calculate the correlation coefficient between two sets of data.

```
import numpy as np

def calculate_correlation(x, y):
    if len(x) != len(y):
        raise ValueError("Both datasets must have the same length")

    # Calculate means
    mean_x = np.mean(x)
    mean_y = np.mean(y)

    # Calculate covariance
    covariance = np.sum((x - mean_x) * (y - mean_y))

    # Calculate standard deviations
    std_x = np.sqrt(np.sum((x - mean_x) ** 2)) # Corrected: squared differences of x
    std_y = np.sqrt(np.sum((y - mean_y) ** 2)) # Corrected: squared differences of y

    # Calculate correlation coefficient
    correlation = covariance / (std_x * std_y)

    return correlation

# New example datasets
x_data = [10, 15, 20, 25, 30]
y_data = [1, 2, 3, 4, 5]

# Calculate and print correlation
correlation_coefficient = calculate_correlation(np.array(x_data), np.array(y_data))
print(f"Correlation Coefficient: {correlation_coefficient:.2f}")
```

Correlation Coefficient: 1.00

Practical 8: Develop a Python program that performs Simple Linear Regression to model the relationship between two variables.

```
import numpy as np
import matplotlib.pyplot as plt

# Function to calculate the slope (m) and intercept (b):
def simple_linear_regression(x, y):
    # Calculate the means of x and y
    mean_x = np.mean(x)
    mean_y = np.mean(y)

    # Calculate the slope (m)
    m = np.sum((x - mean_x) * (y - mean_y)) / np.sum((x - mean_x) ** 2)

    # Calculate the intercept (b)
    b = mean_y - m * mean_x

    return m, b

# Function to predict y values using the regression line
def predict(x, m, b):
    return m * x + b

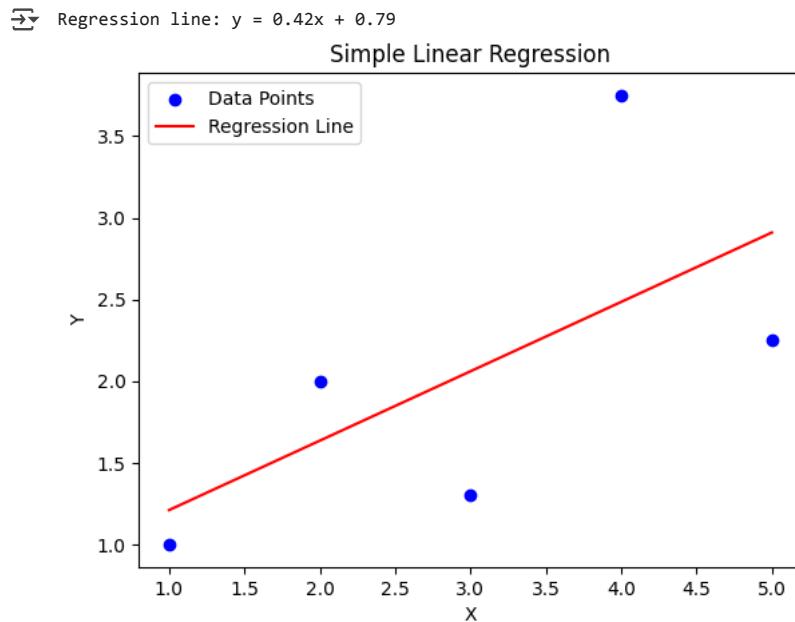
# Example datasets (x - independent variable, y - dependent variable)
x_data = np.array([1, 2, 3, 4, 5])
y_data = np.array([1, 2, 1.3, 3.75, 2.25])

# Perform Simple Linear Regression
m, b = simple_linear_regression(x_data, y_data)

# Print the regression line equation
print(f'Regression line: y = {m:.2f}x + {b:.2f}')

# Predict the y values using the regression line
y_pred = predict(x_data, m, b)

# Plot the original data points and the regression line
plt.scatter(x_data, y_data, color='blue', label='Data Points')
plt.plot(x_data, y_pred, color='red', label='Regression Line')
plt.xlabel('X')
plt.ylabel('Y')
plt.title('Simple Linear Regression')
plt.legend()
plt.show()
```



9. Write a Python program to convert data (e.g., JSON, CSV, XML, etc.) from one format to another.

```

import json
import csv
import xml.etree.ElementTree as ET

# Convert JSON to CSV
def json_to_csv(json_data, csv_filename):
    # Convert JSON data to Python list of dictionaries
    data = json.loads(json_data)

    # Get the keys from the first item in the list (to use as header)
    keys = data[0].keys()

    # Write to CSV file
    with open(csv_filename, mode='w', newline='') as file:
        writer = csv.DictWriter(file, fieldnames=keys)
        writer.writeheader()
        writer.writerows(data)
    print(f"Data successfully written to {csv_filename}")

# Convert CSV to JSON
def csv_to_json(csv_filename):
    # Read CSV file and convert to list of dictionaries
    with open(csv_filename, mode='r') as file:
        reader = csv.DictReader(file)
        data = list(reader)

    # Convert to JSON
    json_data = json.dumps(data, indent=4)
    return json_data

# Convert JSON to XML
def json_to_xml(json_data):
    data = json.loads(json_data)

    # Create the root element of XML
    root = ET.Element("root")

    # Add child elements to root
    for entry in data:
        entry_elem = ET.SubElement(root, "entry")
        for key, value in entry.items():
            child_elem = ET.SubElement(entry_elem, key)
            child_elem.text = str(value)

    # Convert tree to string
    xml_data = ET.tostring(root, encoding='unicode', method='xml')
    return xml_data

# Convert XML to JSON
def xml_to_json(xml_data):
    root = ET.fromstring(xml_data)

    data = []
    for entry_elem in root.findall('entry'):
        entry = {}
        for child_elem in entry_elem:
            entry[child_elem.tag] = child_elem.text
        data.append(entry)

    # Convert to JSON
    json_data = json.dumps(data, indent=4)
    return json_data

# Example usage:

# JSON data as a string
json_data = '''[
    {"name": "Alice", "age": 25, "city": "New York"},
    {"name": "Bob", "age": 30, "city": "Los Angeles"},
    {"name": "Charlie", "age": 35, "city": "Chicago"}
]'''

```

```
# 1. Convert JSON to CSV
json_to_csv(json_data, "output.csv")

# 2. Convert CSV back to JSON
csv_json = csv_to_json("output.csv")
print("CSV to JSON:\n", csv_json)

# 3. Convert JSON to XML
xml_data = json_to_xml(json_data)
print("JSON to XML:\n", xml_data)

# 4. Convert XML back to JSON
xml_json = xml_to_json(xml_data)
print("XML to JSON:\n", xml_json)
```

→ Data successfully written to output.csv
CSV to JSON:

```
[  
  {  
    "name": "Alice",  
    "age": "25",  
    "city": "New York"  
  },  
  {  
    "name": "Bob",  
    "age": "30",  
    "city": "Los Angeles"  
  },  
  {  
    "name": "Charlie",  
    "age": "35",  
    "city": "Chicago"  
  }  
]
```

JSON to XML:

```
<root><entry><name>Alice</name><age>25</age><city>New York</city></entry><entry><name>Bob</name><age>30</age><city>Los Angeles</city></entry>
```

XML to JSON:

```
[  
  {  
    "name": "Alice",  
    "age": "25",  
    "city": "New York"  
  },  
  {  
    "name": "Bob",  
    "age": "30",  
    "city": "Los Angeles"  
  },  
  {  
    "name": "Charlie",  
    "age": "35",  
    "city": "Chicago"  
  }  
]
```

Practical No. 10

Create a Dashboard using Power BI

Pokemon Detailed

This dataset shows all 151 Pokemons in Kanto Region. Check other properties and attributes in this dataset.

type1	name	hp	attack	defense	speed	sp_attack	sp_defense	capture_rate
bug	Abra	64.34	74.53	70.08	70.15	69.40	67.74	106.19
dragon	Aerodactyl							
electric	Alakazam							
fairy	Arbok							
fighting	Arcanine							
fire	Articuno							
type2	Beedrill							
dark	Bellsprout							
electric	Blastoise							
fairy	Bulbasaur							
fighting	Butterfree							
fire	Caterpie							
type1	Chansey							
classification	abilities							
Atrocious Pokémon	[Blaze, Solar Power]							
Ball Pokémon	[Chlorophyll, Effect Spore]							
Balloon Pokémon	[Chlorophyll, Gluttony]							
Barrier Pokémon	[Chlorophyll, Harvest, Frisk...]							
Bat Pokémon	[Chlorophyll, Harvest]							
Beak Pokémon	[Chlorophyll, Leaf Guard, R...]							
Bird Pokémon	[Chlorophyll, Run Away]							
Bivalve Pokémon	[Chlorophyll, Stench]							

Pokemon Detailed

This dataset shows all 151 Pokemons in Kanto Region. Check other properties and attributes in this dataset.

type1	name	hp	attack	defense	speed	sp_attack	sp_defense	capture_rate
electric	bulbasaur	64.34	74.53	70.08	70.15	69.40	67.74	106.19
electric	Ivysaur	74.53	85.00	90.00	85.00	85.00	80.00	110.00
electric	venusaur	142.00	100.00	130.00	120.00	120.00	110.00	130.00
electric	charmander	64.34	74.53	70.08	70.15	69.40	67.74	106.19
electric	charizard	142.00	100.00	130.00	120.00	120.00	110.00	130.00

Pokemon Detailed

This dataset shows all 151 Pokemons in Kanto Region. Check other properties and attributes in this dataset.

type1	name	hp	attack	defense	speed	sp_attack	sp_defense	capture_rate
electric	pikachu	54.44	61.44	64.11	100.00	91.67	73.89	95.33
electric	raichu	61.44	74.00	70.00	100.00	91.67	73.89	95.33
electric	magnemite	64.34	74.53	70.08	70.15	69.40	67.74	106.19
electric	magnezone	142.00	100.00	130.00	120.00	120.00	110.00	130.00

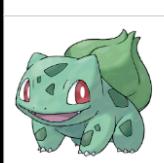
←

Pokemon Name	HEIGHT 70	WEIGHT 69	EXPERIENCE 64	49 attack	49 defense	45 speed	65 sp_attack	65 sp_defense
---------------------	------------------	------------------	----------------------	------------------	-------------------	-----------------	---------------------	----------------------

beedrill
 bellsprout
 blastoise
 bulbasaur
 butterfree
 caterpie
 chansey
 charizard
 charmander
 charmeleon
 clefable
 defairy
 dodster
 cubone
 dewgong

Bulbasaur

[*'Overgrow', 'Chlorophyll'*]



There is a plant seed on its back right from the day this Pokemon is born. The seed slowly grows larger. While it is young, it uses the nutrients that are stored in the seed on its back in order to grow.

Pokemon Classification

Is This a Legendary Pokemon?

False

TYPE 1
grass

TYPE 2
poison

HEALTH (HP)
31.47% → 34.27%
34.27% → 50%
● attack ● defense ● speed
● sp_attack ● sp_defense

CAPTURE RATE
50% → 50%
● attack ● defense ● speed
● sp_attack ● sp_defense

Top 5 Pokemon
Bulbasaur
318

←

Pokemon Name	HEIGHT 60	WEIGHT 85	EXPERIENCE 62	52 attack	43 defense	65 speed	60 sp_attack	50 sp_defense
---------------------	------------------	------------------	----------------------	------------------	-------------------	-----------------	---------------------	----------------------

beedrill
 bellsprout
 blastoise
 bulbasaur
 butterfree
 caterpie
 chansey
 charizard
 charmander
 charmeleon
 clefable
 defairy
 dodster
 cubone
 dewgong

Charmander

[*'Blaze', 'Solar Power'*]



It has a preference for hot things. When it rains, steam is said to spout from the tip of its tail. From the time it is born, a flame burns at the tip of its tail. Its life would end if the flame were to go out.

Pokemon Classification

Is This a Legendary Pokemon?

False

TYPE 1
fire

TYPE 2

HEALTH (HP)
40.63% → 32.5%
26.88% → 54.55%
● attack ● defense ● speed
● sp_attack ● sp_defense

CAPTURE RATE
45.45% → 54.55%
● attack ● defense ● speed
● sp_attack ● sp_defense

Top 5 Pokemon
Charmander
309

Pokemon Name

HEIGHT 220 WEIGHT 21... EXPERIENCE 270

Dragonite

[**'Inner Focus'**, **'Multiscale'**]



It's a kindhearted Pokemon. If it spots a drowning person or Pokemon, Dragonite simply must help them. This Pokemon is known as the Sea Incarnate. Figureheads that resemble Dragonite decorate the bows of many

Pokemon Classification

Is This a Legendary Pokemon?

False

134 95 80
attack defense speed

100 100
sp_attack sp_defense

25.89% 43.37%
30.74%
● attack ● defense ● speed

50% 50%
● sp_attack ● sp_defense

HEALTH (HP)

0 255 91

CAPTURE RATE

0 255 45

Top 5 Pokemon

Dragonite 600

TYPE 1 dragon

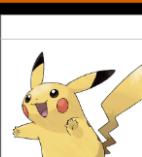
TYPE 2 flying

Pokemon Name

HEIGHT 40 WEIGHT 60 EXPERIENCE 112

Pikachu

[**'Static'**, **'Lightningrod'**]



Pikachu that can generate powerful electricity have cheek sacs that are extra soft and super stretchy. When Pikachu meet, they'll touch their tails together and exchange electricity through them as a form of greeting.

Pokemon Classification

Is This a Legendary Pokemon?

False

55 40 90
attack defense speed

50 50
sp_attack sp_defense

48.65% 29.73%
21.62%
● attack ● defense ● speed

50% 50%
● sp_attack ● sp_defense

HEALTH (HP)

0 255 35

CAPTURE RATE

0 255 190

Top 5 Pokemon

Pikachu 320

TYPE 1 electric

TYPE 2

← HEIGHT 200 WEIGHT 12... EXPERIENCE 306
Pokemon Name

- mewtwo*
- machamp*
- machoke*
- machop*
- magikarp*
- magmar*
- magnemite*
- magneton*
- mankey*
- marowak*
- meowth*
- metapod*
- mew*
- mewtwo*
- moltrix*
- mr-mime*

Mewtwo

[**'Pressure', 'Unnerve'**]



Its DNA is almost the same as Mews. However, its size and disposition are vastly different. .

Pokemon Classification

Is This a Legendary Pokemon?

True

150 70 140
attack defense speed

194 120
sp_attack sp_defense

38.89% 41.67%
● attack ● defense ● speed

38.22% 61.78%
● sp_attack ● sp_defense

HEALTH (HP)
0 255 106

CAPTURE RATE
0 255 3

Top 5 Pokemon
Mewtwo 780

Practical No. 11

Create a Dashboard using Tableau

