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QUANTITATIVE AND SOFTWARE  
PROGRAMMING (21MEA753) LA - 01

UNIT-01

1) LINE PLOT

Write a Python script to plot the population growth of a country over the years using a line plot. The years are from 2010 to 2020 and the population data is given as:

Years : [2010, 2011, 2012, 2013, 2014, 2015, 2016, 2017, 2018, 2019, 2020]

Population (in millions) : [2.5, 2.6, 2.7, 2.8, 2.85, 2.9, 3.0, 3.1, 3.2, 3.25, 3.3]

Include appropriate labels for the X & Y axis, title, & grid lines for better readability.

→

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

```
Years = [2010, 2011, 2012, 2013, 2014, 2015, 2016, 2017, 2018, 2019, 2020]
```

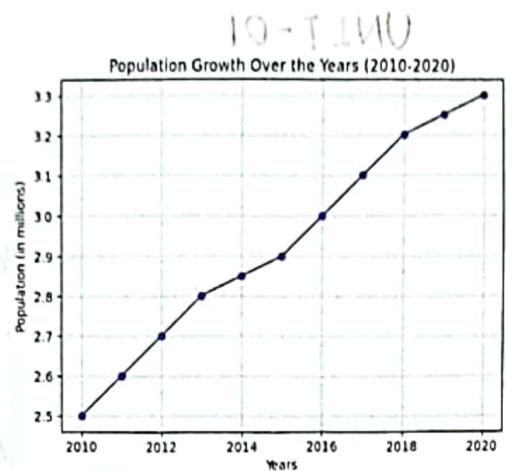
```
Population = [2.5, 2.6, 2.7, 2.8, 2.85, 2.9, 3.0, 3.1, 3.2, 3.25, 3.3]
```

```
plt.plot(Years, Population, marker='o', linestyle='-', color='b')
```

```
plt.xlabel('Years')
```

```
plt.ylabel('Population (in millions)')
```

plt.title('Population Growth over the Years (2010-2020)')  
plt.grid(True)  
plt.show()



## 2) Bar Chart

Write a Python script to create a bar chart showing the sales of different products in a store. The products and their corresponding sales are as follows:

Products : ['A', 'B', 'C', 'D', 'E']

Sales : [100, 150, 90, 200, 130]

Ensure that each bar has a unique color, label X-axis as 'products', the Y-axis as 'Sales' & give the chart a title.

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

```
Products = ['A', 'B', 'C', 'D', 'E']
```

```
Sales = [100, 150, 90, 200, 130]
```

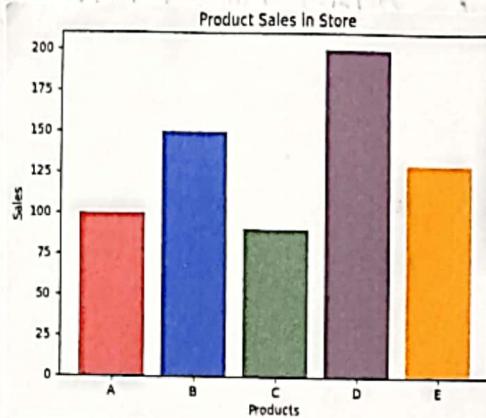
```
plt.bar(Products, Sales, color = ['red', 'blue', 'green', 'purple',  
'orange'])
```

```
plt.xlabel('Products')
```

```
plt.ylabel('Sales')
```

```
plt.title('Product Sales in Store')
```

```
plt.show()
```



### 3) HISTOGRAM

Write a Python script to plot a histogram of student scores in a test. The scores are as follow:

Scores : [88, 92, 100, 67, 85, 79, 95, 80, 78, 90, 85, 99, 76, 65, 70, 91, 87, 82]

Set the number of bins to 5 and make sure to label the X-axis as 'Score' & the Y axis as 'Number of Students'. Give the plot an appropriate title.

import matplotlib.pyplot as plt.

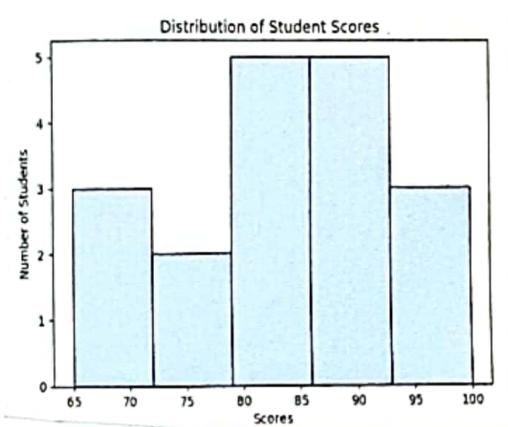
Score = [88, 92, 100, 67, 85, 79, 95, 80, 78, 90, 85, 99, 76, 65, 70, 91, 87, 82]

plt.hist(Score, bins=5, edgecolor='black', color='skyblue')

plt.xlabel('Score')

plt.ylabel('Number of Student Score')

plt.show()

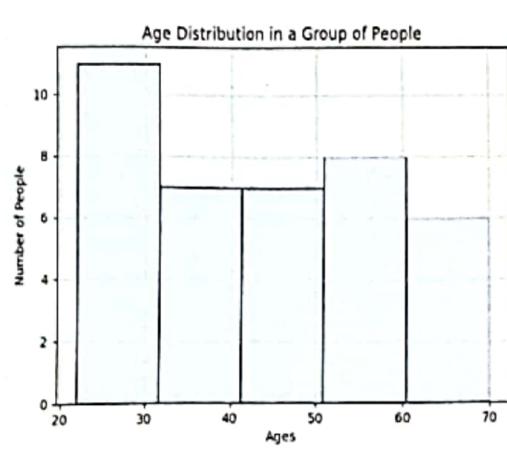


Write a Python Script to create a histogram that shows the distribution of ages in a group of 40 people.  
The age data is as follow:

Ages : [22, 25, 30, 35, 40, 45, 50, 55, 60, 65, 70, 22, 25, 30, 35, 40, 45, 50, 55, 60, 65, 70, 22, 25, 30, 35, 40, 45, 50, 55, 60, 65, 70, 22, 30, 40, 45, 55, 60]

Create a histogram with 5 bins.

```
import matplotlib.pyplot as plt  
Ages = [22, 25, 30, 35, 40, 45, 50, 55, 60, 65, 70, 22, 25, 30, 35, 40, 45, 50, 55, 60, 65, 70, 22, 25, 30, 35, 40, 45, 50, 55, 60, 65, 70, 22, 30, 40, 45, 55, 60]  
plt.hist(Ages, bins=5, edgecolor='black', color='lightblue')  
plt.xlabel('Ages')  
plt.ylabel('Number of People')  
plt.title('Age Distribution in a Group of People')  
plt.grid(True)  
plt.show()
```



## 5) SCATTER PLOT

Write a Python script to create a scatter plot for the relationship between the age of employees and their salaries in a company. The data is given as:

Age : [22, 25, 26, 28, 30, 32, 34, 36, 40, 45, 50]

Salaries (in \$1000s) : [35, 38, 40, 50, 60, 70, 65, 75, 90, 100, 110]

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

```
Age = [22, 25, 26, 28, 30, 32, 34, 36, 40, 45, 50]
```

```
Salaries = [35, 38, 40, 50, 60, 70, 65, 75, 90, 100, 110]
```

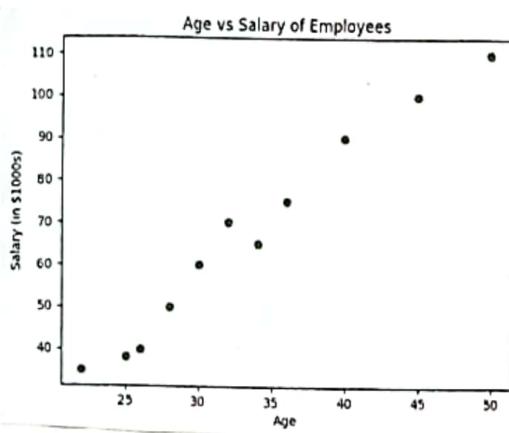
```
plt.scatter(Age, Salaries, color = 'green')
```

```
plt.xlabel('Age')
```

```
plt.ylabel('Salary (in $1000s)')
```

```
plt.title('Age VS Salary of Employees')
```

```
plt.show()
```



## 6) PIE CHART

Write a Python Program to create a pie chart representing the market share of different companies in a specific industry. The Companies and their market share are given. Make sure to label each slice with the percentage & give the chart a title.

→ import matplotlib.pyplot as plt

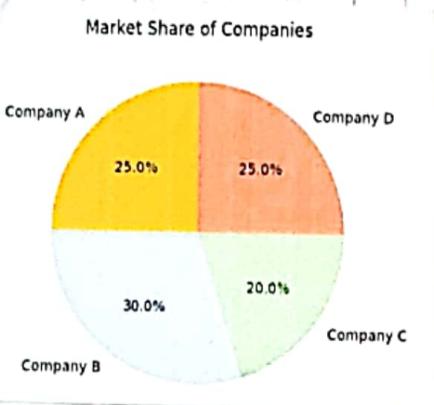
Companies = ['Company A', 'Company B', 'Company C', 'Company D']

Market\_Share = [25, 30, 20, 25]

plt.pie(Market\_Share, labels=Companies, autopct='%.1f %%',  
startangle=90, colors=['gold', 'lightblue', 'lightgreen',  
'coral'])

plt.title('Market Share of Companies')

plt.show()



## 7) BOX PLOT

Write a Python script to create a box plot for the test scores of students in three different subjects: Math, Science and English. Plot a box plot for each subject & label the X-axis of 'Subjects' and the Y-axis of 'Scores'. Give the plot a title and display the outliers, if any.

→ import matplotlib.pyplot as plt

math\_scores = [88, 92, 75, 85, 90, 95, 89, 91, 76, 85]

science\_scores = [80, 85, 88, 92, 91, 87, 90, 95, 89, 88]

english\_scores = [78, 82, 80, 85, 88, 90, 91, 86, 89, 92]

data = [math\_scores, science\_scores, english\_scores]

plt.boxplot(data, patch\_artist=True, labels=['Math', 'Science', 'English'])

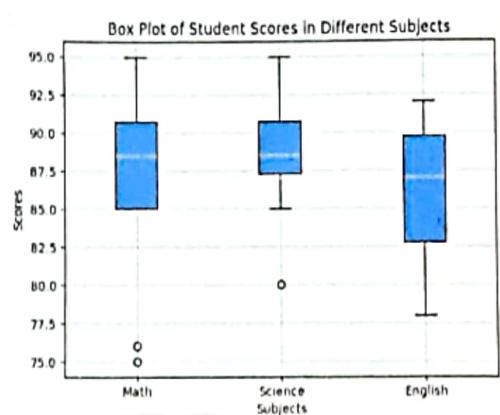
plt.xlabel('Subjects')

plt.ylabel('Scores')

plt.title('Box plot of Student Scores in Different Subjects')

plt.grid(True)

plt.show()



## 8) HEAT MAP

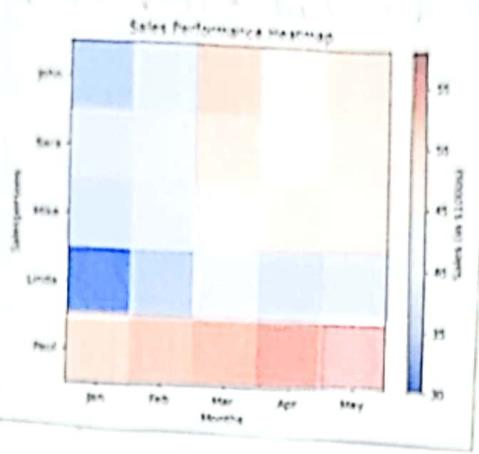
Write a Python script to create a heatmap representing the performance of different salespersons over five months. The sale data is given as.

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

salespersons = ['John', 'Sara', 'Mike', 'Linda', 'Paul']
months = ['Jan', 'Feb', 'Mar', 'Apr', 'May']

sales_data = [
    [35, 40, 50, 45, 48], # John
    [40, 42, 48, 44, 47], # Sara
    [38, 41, 45, 47, 46], # Mike
    [30, 35, 42, 38, 40], # Linda
    [50, 52, 53, 55, 58], # Paul
]

plt.imshow(sales_data, cmap='coolwarm', interpolation='nearest')
plt.colorbar(label='Sales (in $1000s)')
plt.xticks(ticks=np.arange(len(months)), labels=months)
plt.yticks(ticks=np.arange(len(salespersons)), labels=salespersons)
plt.xlabel('Month')
plt.ylabel('Sales Performance Heatmap')
plt.title('Salespersons')
plt.show()
```



## 9) CONTOUR PLOT

Write a Python script to create a contour plot of a function

$Z = f(X, Y)$ , where  $Z = X^2 + Y^2$ . The  $X$  &  $Y$  values should range from -5 to 5. Generate a contour plot with 10 contour levels & label both the axes and the contour lines.

```
import matplotlib.pyplot as plt
import numpy as np
x = np.linspace(-5, 5, 100)
y = np.linspace(-5, 5, 100)
X, Y = np.meshgrid(x, y)
Z = X**2 + Y**2
plt.contour(X, Y, Z, levels=10, cmap='viridis')
plt.colorbar(label='Z = X^2 + Y^2')
```

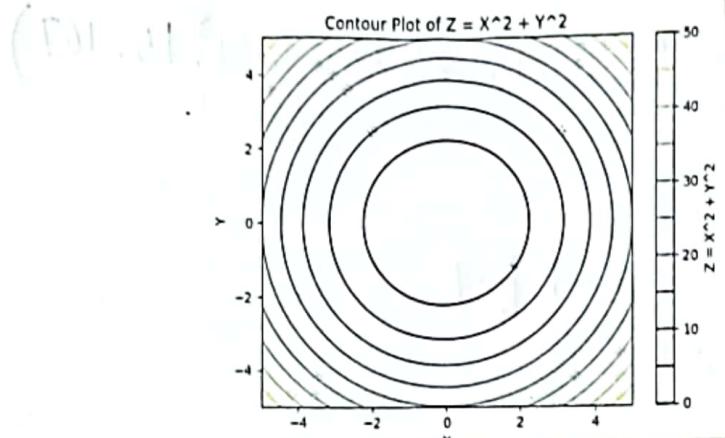
```
plt.xlabel('x')
```

```
plt.ylabel('y')
```

```
plt.title('Contour Plot of Z = X^2 + Y^2')
```

```
plt.contourf(x,y,z, levels=10, cmap='viridis'),  
            inline=True, font_size=8)
```

```
plt.show()
```



## UNIT - 02

Q) Calculate the mean, median, mode & standard deviation for the following array using Numpy & Scipy:

arr = np.array([15, 18, 20, 18, 21, 17, 20, 19, 15, 16])

→ import numpy as np  
from scipy import stats.

arr = np.array([15, 18, 20, 18, 21, 17, 20, 19, 15, 16])

mean = np.mean(arr)

median = np.median(arr)

mode = stats.mode(arr).mode[0]

std\_dev = np.std(arr)

print('Mean = {mean}')

print('Median = {median}')

print('Mode = {mode}')

print('Standard Deviation = {std-dev}')

Output:

Mean = 17.9

Median = 18.0

Mode = 15

Standard Deviation = 2.0223

2) Create a 2D Numpy array as follows:

$\text{arr\_2d} = \text{np.array}([[1, 2, 3], [4, 5, 6], [7, 8, 9]])$

Calculate the mean, median, mode & standard deviation of the array along both axes.

```
→ import numpy as np
```

```
from numpy import stats
```

```
arr_2d = np.array([[1, 2, 3], [4, 5, 6], [7, 8, 9]])
```

```
mean_rows = np.mean(arr_2d, axis=1)
```

```
mean_columns = np.mean(arr_2d, axis=0)
```

```
median_rows = np.median(arr_2d, axis=1)
```

```
median_columns = np.median(arr_2d, axis=0)
```

```
mode_rows = stats.mode(arr_2d, axis=1)
```

```
mode_columns = stats.mode(arr_2d, axis=0)
```

```
std_rows = np.std(arr_2d, axis=1)
```

```
std_columns = np.std(arr_2d, axis=0)
```

```
print("Mean Rows:", mean_rows)
```

```
print("Mean Columns:", mean_columns)
```

```
print("Median Rows:", Median_rows)
```

```
print("Median Columns:", Median_columns)
```

```
print("Mode Rows:", Mode_rows)
```

```
print("Mode Columns:", mode_columns)
```

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

`print("Standard Deviation Columns", std[columns]).`

Output:

Mean Rows : [2. 5. 8.]

Mean Columns : [4. 5. 6.]

Median Rows : [2. 5. 8]

Median Columns : [4. 5. 6]

Mode Rows : [1. 4. 7]

Mode Columns : [1. 2. 3]

Standard Deviation Rows : [0.8164 0.8164 0.8164]

Standard Deviation Columns : [2.449 2.449 2.449]

Suppose you have the following 2D array where some data points are missing (np.nan).

```
data = np.array([[1, 2, np.nan], [4, np.nan, 6], [7, 8, 9]])
```

Calculate the mean, median, mode & SD. Explain how missing data can affect statistical analysis.

- Missing values are handled with `np.nanmean`, `np.nanmedian`, etc. Missing data can skew results; it's crucial to address it appropriately by imputing, removing or using functions that ignore NaN.

```
import numpy as np  
from scipy import stats  
data = np.array ([[1, 2, np.nan], [4, np.nan, 6],  
[7, 8, 9]])
```

mean = np.nanmean (data)

Median = np.nanmedian (data)

Mode = stats.mode (data [-np.isnan (data)]). mode [0]

Std-dev = np.nanstd (data)

Print (Mean : {mean})

Print (Median : {Median})

Print (Mode : {mode})

Print (Standard Deviation : {Std-dev})

Output :

Mean : 5.28571

Median : 6.0

Mode : ~~NaN~~

Standard Deviation : 2.8139

Two Groups of Students took the same test, & their scores given by the following 1D arrays.

group\_A = np.array ([85, 87, 90, 92, 88, 86, 84, 91, 89, 93])

group\_B = np.array ([78, 85, 92, 90, 83, 87, 95, 70, 91, 88])

→ import numpy as np

group\_A = np.array ([85, 87, 90, 92, 88, 86, 84, 91, 89, 93])

group\_B = np.array ([78, 85, 92, 90, 83, 87, 95, 70, 91, 88])

mean\_A = np.mean (group\_A)

median\_A = np.median (group\_A)

mode\_A = stats.mode (group\_A)

std\_dev\_A = np.std (group\_A)

mean\_B = np.mean (group\_B)

median\_B = np.median (group\_B)

mode\_B = stats.mode (group\_B)

std\_dev\_B = np.std (group\_B).

print ("Mean\_A : {mean\_A} , Mean\_B : {mean\_B}")

print ("Median\_A : {median\_A} , Median\_B : {median\_B}")

print ("Mode\_A : {mode\_A} , Mode\_B : {mode\_B}")

print ("Standard\_Deviation\_A : {std\_dev\_A} , Standard\_Deviation\_B : {std\_dev\_B}")

Output:

Mean - A : 88.5 , Mean - B : 85.9

Median - A : 88.5 , Median - B : 87.5

Mode - A : 84 , Mode - B : 70

Standard Deviation - A : 2.8722 , Standard Deviation - B : 7.0206

Q2 Create a 1D Numpy array with 50 random integers between 10 & 100 using Python code

arr = np.random.randint(10, 100, size=50).

Using Python code calculate the mean, median, mode & standard deviation of the array. Additionally, Plot a histogram of the array to visualize its distribution.

arr = np.random.randint(10, 100, size=50)

mean = np.mean(arr)

median = np.median(arr)

mode = stats.mode(arr).mode

std-dev = np.std(arr)

Point 6 "Mean : {mean}, Median : {median}, Mode : {mode},

Standard Deviation : {std-dev})"

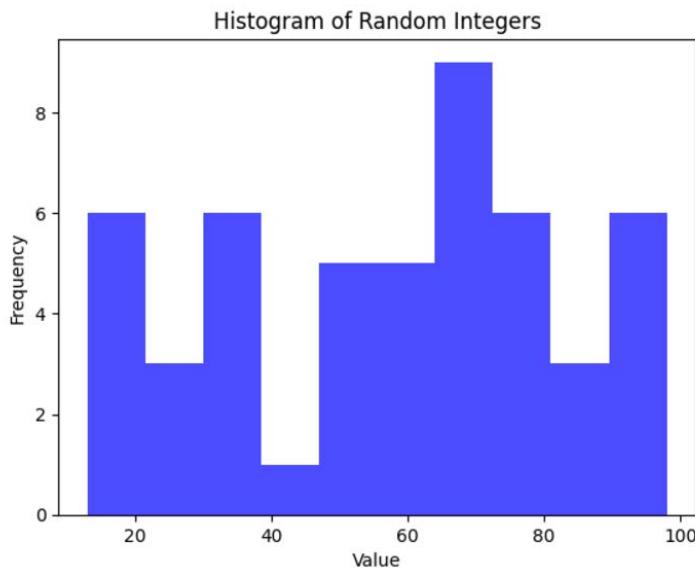
plt.hist(arr, bins=10, alpha=0.7, color='blue')

plt.title("Histogram of Random Integers")

plt.xlabel("Value")

plt.ylabel("Frequency")

plt.show()



## Pandas Question

Create a Pandas Series of follows.

```
data = pd.Series([10, 12, 14, 13, 15, 16, 14, 18, 19, 20])
```

Write a Python program that calculates the mean, median, mode & standard deviation for this series.

```
import pandas as pd
```

```
data = pd.Series([10, 12, 14, 13, 15, 16, 14, 18, 19, 20])
```

```
mean = data.mean()
```

```
median = data.median()
```

```
mode = data.mode()
```

```
std_dev = data.std()
```

```
Variance = data.var()
```

```
print(f"Mean: {mean}, Median: {median}, Mode: {mode}, Standard Deviation: {std_dev}, Variance: {Variance}")
```

Output :

Mean : 15.1, Median : 14.5, Standard Deviation : 3.17804 ,  
Variance : 10.100

Basic Statistical Measures on a pandas DataFrame :

Given the following pandas DataFrame representing the ages and heights of 5 individuals:

data = { 'Age' : [23, 25, 31, 22, 35], 'Height' : [167, 175, 180, 165, 170] }

import Pandas as pd

data = { 'Age' : [23, 25, 31, 22, 35], 'Height' : [167, 175, 180, 165, 170] }

df = pd.DataFrame(data)

mean = df.mean()

median = df.median()

mode = df.mode().iloc[0]

std-dev = df.std()

print ("Mean : ", mean)

print ()

print ("Median : ", median)

print ()

print ("Mode : ", mode)

print ()

print ("Standard Deviation : ", std-dev)

Output

Mean:

Age 27.2

Height 171.6

dtype float64

Median:

Age 25.0

Height 170.0

dtype float64

Mode:

Age 22

Height 165

dtype int64

Standard Deviation

Age 5.585696

Height 6.1073

dtype float64

## Handling Missing Data in Pandas Data Frame.

Consider the following pandas Data Frame with some missing data (NaN values):

```
data = {'Age': [23, 25, None, 22, 35], 'Height': [167, 175, 180, None, 170], 'Weight': [65, 70, None, 68, 72]}
```

```
→ data = {'Age': [23, 25, None, 22, 35], 'Height': [167, 175, 180, None, 170], 'Weight': [65, 70, None, 68, 72]}
```

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

```
mean = df.mean()
```

```
median = df.median()
```

```
std_dev = df.std()
```

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

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

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

Output

Mean : 26.25

Height 173.00

dtype: float64

Median :

Age 24.0

Height 172.5

Weight 69.0

dtype : float64

Standard Deviation.

Age 5.9651

Height 5.7154

Weight 2.9860

dtype : float64

Column - Wise & Row - Wise Statistical Calculations :

Consider the following Dataframe representing the marks scored by 3 students in 4 subjects.

data = {'Math': [85, 92, 78], 'English': [88, 79, 85], 'Science': [90, 87, 92], 'History': [80, 70, 88]}

df = pd.DataFrame(data, index = ['Student 1', 'Student 2', 'Student 3'])

```
import Pandas as pd
```

```
data = {
```

```
'Math': [85, 92, 78],
```

```
'English': [88, 87, 85],
```

```
'Science': [90, 87, 91],
```

```
'History': [80, 70, 88]
```

```
}
```

```
df = pd.DataFrame(data, index = ['Student 1', 'Student 2',  
                                 'Student 3'])
```

```
print(df)
```

```
print()
```

```
col_mean = df.mean()
```

```
col_median = df.median()
```

```
col_std = df.std()
```

```
row_mean = df.mean(axis = 1)
```

```
row_Median = df.median(axis = 1)
```

```
row_std = df.std(axis = 1)
```

```
print("Column-Wise Statistics :")
```

```
print()
```

```
print("Column Mean:", col_mean)
```

```
print("Column Median:", col_median)
```

```
Print ("Column Standard Deviation : ", col_std)
```

```
Print ()
```

```
Print ("Row-Wise Statistics : ")
```

```
Print ("Row Mean : ", row_mean)
```

```
Print ("Row Median : ", row_median)
```

```
Print ("Row Standard Deviation : ", row_std)
```

Output:

	Math	English	Science	History
Student 1	85	88	90	80
Student 2	92	79	87	70
Student 3	78	85	92	88

Column-Wise Statistics:

Mean:

Math 85

English 84

Science 89.667

History 79.333

dtype: float64

Median

Math 85

English 85

Science 90

History 80

dtype: float64

Standard Deviation

Math 7.00

English 4.5825

Science 2.5166

History 9.0185

dtype: float64

Row-Wise Statistics:

Mean:

Student 1 85.75

Student 2 82.00

Student 3 85.75

dtype: float64

Median:

Student 1 86.5

Student 2 83.0

Student 3 86.5

dtype: float64

~~Student~~

Standard Deviation:

Student 1 4.349

Student 2 9.626

Student 3 5.909

dtype: float 64

Understanding Bernoulli Trials:

Define a Bernoulli trial and give an example of a real world scenario where Bernoulli trials can be applied.

Write a Python program that simulates 10 independent Bernoulli trials. With a success probability of 0.6 to compute the mean & variance of the outcome.

→ import numpy as np

$$P_{\text{Success}} = 0.6$$

$$n_{\text{trials}} = 10$$

trials = np.random.binomial(1, P\_Success, n\_trials)

mean\_outcome = np.mean(trials)

Variance\_outcome = np.var(trials)

print(trials)

print(mean\_outcome)

print(Variance\_outcome)

[1 1 0 1 001 000]

0.4

0.24005

Write a Python Program that simulates 1000 Bernoulli trials with a success probability of 0.4 using Numpy.

Calculate the proportion of successes and compare it with the theoretical probability. Plot a bar chart to visualize the number of successes & failures.

Explain the relationship b/w the simulated results & the theoretical probability of success.

```
import numpy as np
```

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

$$P_{\text{Success}} = 0.4$$

$$n_{\text{trials}} = 1000$$

```
trials = np.random.binomial(1, P_Success, n_trials)
```

```
proportion_success = np.mean(trials)
```

```
theoretical_Prob = p_Success
```

```
n_Success = np.sum(trials)
```

```
n_failure = n_trials - n_success
```

```
print(f"Number of successes: {n_success}")
```

```
print(f"Number of failures: {n_failure}")
```

~~```
print(f"Number of
```~~

```
print(f"Proportion of success (simulated): {proportion_success}")
```

```
print(f"Theoretical probability of success:  
{theoretical_prob}")
```

```
# Plot the bar chart
```

```
labels = ['Success', 'Failure']
```

```
counts = [n_success, n_failure]
```

```
plt.bar(labels, counts, color=['green', 'red'])
```

```
plt.ylabel('Count')
```

```
plt.title('Bernoulli Trials: Success vs Failure')
```

```
plt.show()
```

Output

Number of successes: 368

Number of failures: 632

Proportion of success (simulated): 0.368

Theoretical probability of success: 0.4

