- 1. (10 points) Create a Python script to:
- Read a data set from a CSV file (e.g., data.csv).
- Compute the mean, median, mode, standard deviation, and variance for the numerical columns.
- Save the computed statistics to a new CSV file (e.g., statistics\_summary.csv).

data.csv Link - Click here

```
import pandas as pd
import statistics
import csv
url =
'https://raw.githubusercontent.com/gagan-iitb/DataAnalyticsAndVisualiz
ation/main/Lab-W25/data.csv'
df = pd.read csv(url)
stat list = []
numerical columns = df.select dtypes(include=['number']).columns
for column in numerical columns:
    data = df[column].dropna()
    stats = {
        "Column": column,
        "Mean": statistics.mean(data),
        "Median": statistics.median(data),
        "Mode": statistics.mode(data),
        "Standard Deviation": statistics.stdev(data),
        "Variance": statistics.variance(data),
    stat list.append(stats)
stats df = pd.DataFrame(stat list)
stats_df.to_csv("statistics_summary.csv", index=False)
print(f"Statistics summary saved to statistics_summary.csv")
print(stats df)
Statistics summary saved to statistics summary.csv
 Column
            Mean Median Mode Standard Deviation
                                                       Variance
0 Value 77.519
                    77.0
                            45
                                         42.607218 1815.375014
```

- 1. (10 points) Given the data set [2.0, 4.5, 4.5, 5.0, 7.0, 8.0, 10.0]:
- Calculate the mean, median, mode (Traditional Way).
- Verify your results using Python's statistics module.

```
import statistics
data = [2.0, 4.5, 4.5, 5.0, 7.0, 8.0, 10.0]
```

```
# Mean
mean manual = sum(data) / len(data)
# Median
sorted data = sorted(data)
n = len(sorted data)
median_manual = sorted_data[n // 2] if n%2==1 else (sorted_data[n // 2
- 1] + sorted data[n // 2]) / 2
# Mode
frequency = {}
for value in data:
    frequency[value] = frequency.get(value, 0) + 1
mode manual = max(frequency, key=frequency.get)
print("Manual Calculations:")
print(f"Mean: {mean manual}")
print(f"Median: {median manual}")
print(f"Mode: {mode manual}")
# Using statistics module
mean stats = statistics.mean(data)
median stats = statistics.median(data)
mode stats = statistics.mode(data)
print("\nUsing statistics module:")
print(f"Mean: {mean stats}")
print(f"Median: {median stats}")
print(f"Mode: {mode stats}")
Manual Calculations:
Mean: 5.857142857142857
Median: 5.0
Mode: 4.5
Using statistics module:
Mean: 5.857142857142857
Median: 5.0
Mode: 4.5
```

## **Numpy Assignment**

## Part 1: Exploring Dimensions and Shapes (20 points)

- 1. Create an array of shape (4, 3, 2) representing a 3D matrix.
- 2. Print the shape and the number of dimensions (ndim) of the array.
- 3. Reshape the array into a 2D matrix of shape (12, 2) and verify the new shape and number of dimensions.

4. Create a scalar array with the value 7 and print its shape and dimensions.

```
import numpy as np
array 3d = np.arange(24).reshape(4, 3, 2)
print("3D Array:")
print(array 3d)
print(f"Shape: {array 3d.shape}")
print(f"Number of dimensions: {array 3d.ndim}\n")
array_2d = array_3d.reshape(12, 2)
print("Reshaped to 2D Array:")
print(array 2d)
print(f"Shape: {array 2d.shape}")
print(f"Number of dimensions: {array 2d.ndim}\n")
array_1d = np.array(7)
print("Scalar Array:")
print(array_1d)
print(f"Shape: {array 1d.shape}")
print(f"Number of dimensions: {array ld.ndim}")
3D Array:
[[[0 1]
  [23]
  [ 4 5]]
 [[ 6 7]
  [8 9]
[10 11]]
 [[12 13]
  [14 15]
 [16 17]]
 [[18 19]
  [20 21]
  [22 23]]]
Shape: (4, 3, 2)
Number of dimensions: 3
Reshaped to 2D Array:
[[0 1]
 [2 3]
 [ 4 5]
 [67]
 [8 9]
 [10 11]
```

```
[12 13]
[14 15]
[16 17]
[18 19]
[20 21]
[22 23]]
Shape: (12, 2)
Number of dimensions: 2

Scalar Array:
7
Shape: ()
Number of dimensions: 0
```

### Part 2: Matrix Manipulations (15 points)

- 1. Create a random 4x4 matrix and print its values. Then, find the sum of the elements across both axes (i.e., sum along rows and sum along columns).
- 2. Create a 5x5 identity matrix and modify the middle element (at position [2,2]) to 99. Print the modified matrix.
- 3. Create a 3x3 matrix with random values, then flip the matrix left-right to get the opposite diagonal (anti-diagonal).

```
import numpy as np
# Part 1: Create a random 4x4 matrix and find the sum across axes
random matrix = np.random.rand(4, 4)
print("Random 4x4 Matrix:")
print(random matrix)
sum rows = np.sum(random matrix, axis=1)
sum columns = np.sum(random matrix, axis=0)
print("\nSum along rows:", sum rows)
print("Sum along columns:", sum columns)
# Part 2: Create a 5x5 identity matrix and modify the middle element
identity matrix = np.eye(5)
identity_matrix[2, 2] = 99
print("\nModified 5x5 Identity Matrix:")
print(identity matrix)
# Part 3: Create a 3x3 random matrix and flip it left-right
random 3x3 matrix = np.random.rand(3, 3)
print("\nRandom 3x3 Matrix:")
print(random 3x3 matrix)
```

```
# Flip the matrix left-right to get the anti-diagonal
flipped matrix = np.fliplr(random 3x3 matrix)
print("\nFlipped Matrix (Anti-Diagonal):")
print(flipped matrix)
Random 4x4 Matrix:
[[0.61533392 0.88292712 0.47558652 0.11647848]
 [0.98589121 0.12420045 0.58445603 0.00294031]
 [0.2916423  0.76584451  0.62439872  0.49088585]
 [0.4008244 0.91078792 0.25025465 0.02269673]]
Sum along rows: [2.09032604 1.697488 2.17277139 1.58456369]
Sum along columns: [2.29369183 2.68375999 1.93469592 0.63300138]
Modified 5x5 Identity Matrix:
[[1. 0. 0. 0. 0.]
 [ 0. 1. 0. 0.
                  0.1
 [ 0. 0. 99. 0. 0.]
 [0. 0. 0. 1. 0.]
 [0. 0. 0. 0. 1.]
Random 3x3 Matrix:
[[0.52711695 0.08386777 0.37232185]
 [0.44755518 0.80745554 0.85148271]
 [0.04908742 0.17367371 0.60144114]]
Flipped Matrix (Anti-Diagonal):
[[0.37232185 0.08386777 0.52711695]
 [0.85148271 0.80745554 0.44755518]
 [0.60144114 0.17367371 0.04908742]]
```

#### **Data Visualization** (15 points)

```
import pandas as pd
import numpy as np

# Set a seed for reproducibility
np.random.seed(42)

# Generate a random dataset
num_samples = 100

data = {
    "ID": range(1, num_samples + 1),
    "Age": np.random.randint(18, 60, size=num_samples),
    "Height_cm": np.random.normal(165, 10, num_samples).round(1),
    "Weight_kg": np.random.normal(70, 15, num_samples).round(1),
    "City": np.random.choice(["New York", "Los Angeles", "Chicago",
    "Houston", "Phoenix"], size=num_samples),
    "Grade": np.random.choice(["A", "B", "C", "D", "F"],
```

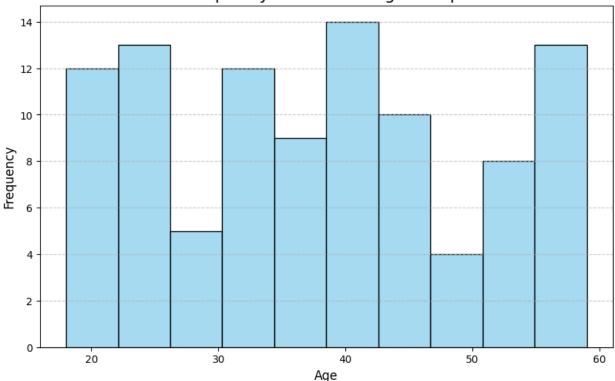
```
size=num samples),
    "Monthly Income": np.random.randint(2000, 10000,
size=num samples),
    "Hours Studied": np.random.exponential(5, num samples).round(1),
    "Passed": np.random.choice(["Yes", "No"], size=num samples,
p=[0.8, 0.2]),
    "Category": np.random.choice(["Category 1", "Category 2",
"Category 3"], size=num samples),
    "Test Score": np.random.uniform(50, 100,
size=num samples).round(2),
    "Exercise Hours": np.random.poisson(3, num samples),
    "Favorite Color": np.random.choice(["Red", "Blue", "Green",
"Yellow", "Purple"], size=num samples),
}
df = pd.DataFrame(data)
# Save to a CSV for reuse
df.to csv("random plotting dataset.csv", index=False)
# Display the first few rows
print(df.head())
            Height cm Weight kg
                                          City Grade
                                                      Monthly Income \
   ID
       Age
    1
        56
                163.3
0
                            67.8
                                       Chicago
                                                                3852
                                                   Α
1
   2
        46
                176.6
                            72.7
                                      Houston
                                                   Α
                                                                6910
2
        32
                162.5
                            84.5
    3
                                  Los Angeles
                                                   Α
                                                                7268
3
    4
        25
                157.3
                            54.0
                                       Chicago
                                                   D
                                                                6175
   5
        38
                177.1
                                       Phoenix
                            71.6
                                                   Α
                                                                4933
   Hours Studied Passed Category Test Score Exercise Hours
Favorite_Color
             9.2
                    Yes Category 2
                                           64.40
                                                               6
Blue
                                                               2
                                           80.75
1
             9.1
                    Yes Category 1
Blue
                                                               3
             3.2
                                           95.59
                    Yes Category 3
Blue
             2.7
                    Yes
                         Category 1
                                           56.96
                                                               7
Blue
                                                               0
             1.6
                    Yes
                         Category 1
                                           55.04
Blue
```

#### Create a histogram to show the frequency of different age groups.

```
import matplotlib.pyplot as plt
import seaborn as sns
plt.figure(figsize=(10, 6))
```

```
sns.histplot(df['Age'], bins=10, color='skyblue', edgecolor='black')
plt.title('Frequency of Different Age Groups', fontsize=16)
plt.xlabel('Age', fontsize=12)
plt.ylabel('Frequency', fontsize=12)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.show()
```

# Frequency of Different Age Groups

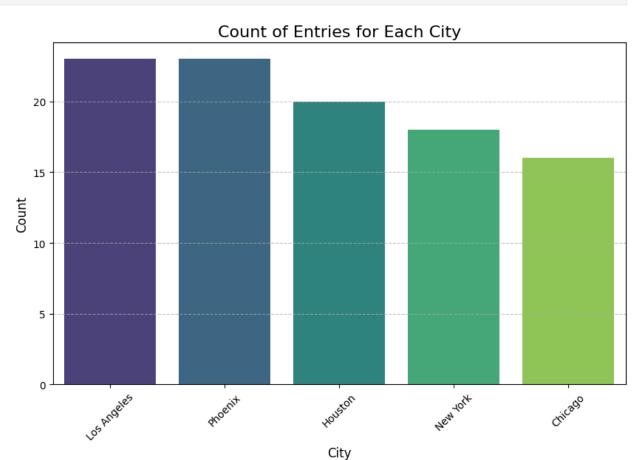


# Use a bar chart to display the count of entries for each City.

```
plt.figure(figsize=(10, 6))
sns.countplot(data=df, x='City', palette='viridis',
order=df['City'].value_counts().index)
plt.title('Count of Entries for Each City', fontsize=16)
plt.xlabel('City', fontsize=12)
plt.ylabel('Count', fontsize=12)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.xticks(rotation=45)
plt.show()
<ipython-input-77-52b2299ea06b>:2: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.
```

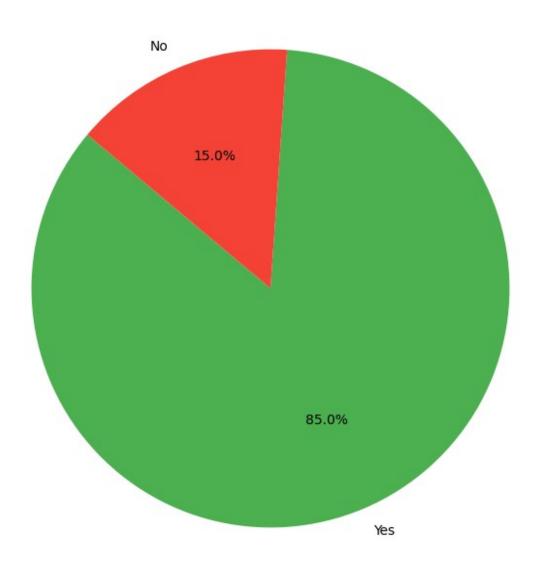
```
sns.countplot(data=df, x='City', palette='viridis',
order=df['City'].value_counts().index)
```



### Create a pie chart to show the proportion of Passed values.

```
passed_counts = df['Passed'].value_counts()
plt.figure(figsize=(8, 8))
plt.pie(
    passed_counts,
    labels=passed_counts.index,
    autopct='%1.1f%%',
    startangle=140,
    colors=['#4caf50', '#f44336']
)
plt.title('Proportion of Passed Values', fontsize=16)
plt.show()
```

# Proportion of Passed Values



# Matrix Problem: (10+40 = 50 points)

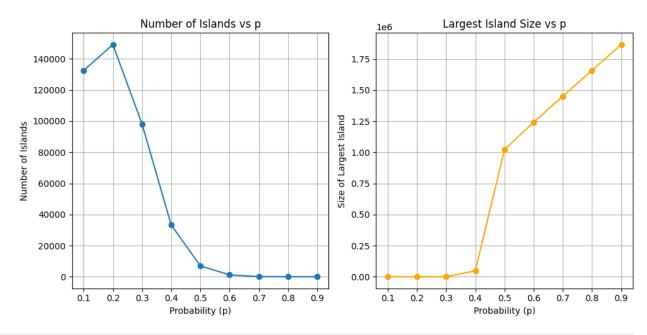
- 1. Calculate the rank of a 15 \* 15 binary random matrix, similar to what we did in the activity. But this time, we will go to each cell and generate a 1 with probability "p".
- 2. Now generate a binary random matrix of size 1920 x 1080. Imagine 1s to be land and 0s to be water. Write a program to count the number of islands in your matrix. Study the following: number of islands, size of the largest island and plot them for varying values of "p".

```
import numpy as np
import matplotlib.pyplot as plt
from scipy.linalg import svd
```

```
def calculate matrix rank(p, size):
    matrix = np.random.choice([0, 1], size=size, p=[1-p, p])
    rank = np.linalg.matrix rank(matrix)
    return matrix, rank
def count islands(matrix):
    rows, cols = matrix.shape
    visited = np.zeros like(matrix, dtype=bool)
    def iterative dfs(start x, start y):
        stack = [\overline{(start x, start_y)}]
        size = 0
        while stack:
            x, y = stack.pop()
            if x < 0 or y < 0 or x >= rows or y >= cols or visited[x,
y] or matrix[x, y] == 0:
                continue
            visited[x, y] = True
            size += 1
            for dx, dy in [(-1, 0), (1, 0), (0, -1), (0, 1), (-1, -1),
(-1, 1), (1, -1), (1, 1):
                stack.append((x + dx, y + dy))
        return size
    num islands = 0
    \max \text{ size } = 0
    for i in range(rows):
        for j in range(cols):
            if matrix[i, j] == 1 and not visited[i, j]:
                num islands += 1
                \max \text{ size} = \max(\max \text{ size}, \text{ iterative dfs}(i, j))
    return num islands, max size
def plot islands(p values, num islands list, max sizes list):
    plt.figure(figsize=(10, 5))
    plt.subplot(1, 2, 1)
    plt.plot(p values, num islands list, marker='o', label="Number of
Islands")
    plt.xlabel("Probability (p)")
    plt.ylabel("Number of Islands")
    plt.title("Number of Islands vs p")
    plt.grid(True)
    plt.subplot(1, 2, 2)
    plt.plot(p values, max sizes list, marker='o', label="Largest
Island Size", color='orange')
```

```
plt.xlabel("Probability (p)")
   plt.ylabel("Size of Largest Island")
   plt.title("Largest Island Size vs p")
   plt.grid(True)
   plt.tight layout()
   plt.show()
p = 0.5
binary matrix, rank = calculate matrix rank(p, (15, 15))
print("Binary Matrix (15x15):\n", binary_matrix)
print("Rank of the matrix:", rank)
Binary Matrix (15x15):
 [[0 1 0 0 0 0 0 1 1 1 0 1 0 0 1]
 [0\ 1\ 1\ 0\ 0\ 0\ 1\ 0\ 0\ 1\ 0\ 0\ 1\ 1]
 [0 0 0 1 0 1 1 1 0 1 1 1 0 1 0]
 [0 0 1 0 0 0 0 1 1 0 0 1 0 0 1]
 [1 1 0 0 1 0 1 1 1 1 1 1 1 0]
 [0\ 1\ 0\ 0\ 0\ 1\ 1\ 1\ 1\ 1\ 0\ 1\ 0\ 1\ 1]
 [0\ 0\ 0\ 0\ 0\ 1\ 1\ 1\ 1\ 1\ 1\ 0\ 1\ 0\ 1]
 [1 1 0 1 1 0 0 1 0 1 0 0 1 0 1]
 [0\ 0\ 1\ 1\ 1\ 1\ 1\ 0\ 0\ 1\ 1\ 1\ 0\ 0\ 1]
 [1\ 1\ 1\ 1\ 0\ 0\ 0\ 0\ 1\ 0\ 1\ 0\ 0\ 0\ 0]
 [1 0 1 0 0 0 1 1 0 0 1 0 0 1 1]]
Rank of the matrix: 15
p values = np.linspace(0.1, 0.9, 9)
num islands list = []
largest islands = []
for p in p values:
 large matrix, rank = calculate matrix rank(p, (1920, 1080))
 num islands, max size = count islands(large matrix)
 num_islands_list.append(num_islands)
 largest islands.append(max size)
 print(f"p = {p:.1f} | Number of Islands: {num islands} | Largest
Island Size: {max size}")
plot islands(p values, num islands list, largest islands)
p = 0.1 | Number of Islands: 132567 | Largest Island Size: 17
p = 0.2
         Number of Islands: 149228 | Largest Island Size: 52
         Number of Islands: 98141 | Largest Island Size: 272
p = 0.3
p = 0.4 | Number of Islands: 33285 | Largest Island Size: 47517
p = 0.5 | Number of Islands: 6968 | Largest Island Size: 1021064
```

```
p = 0.6 | Number of Islands: 1196 | Largest Island Size: 1242581
p = 0.7 | Number of Islands: 96 | Largest Island Size: 1450951
p = 0.8 | Number of Islands: 7 | Largest Island Size: 1658524
p = 0.9 | Number of Islands: 1 | Largest Island Size: 1866582
```



```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
from scipy.stats import linregress, spearmanr, pearsonr
# Statistical summary
def compute statistics(p values, num islands, largest islands):
    num islands stats = {
        "Mean": np.mean(num islands),
        "Standard Deviation": np.std(num_islands),
        "Variance": np.var(num islands)
    largest islands stats = {
        "Mean": np.mean(largest islands),
        "Standard Deviation": np.std(largest islands),
        "Variance": np.var(largest islands)
    }
    print("\nStatistical Summary:")
    print("Number of Islands Statistics:", num_islands_stats)
    print("Largest Island Size Statistics:", largest_islands_stats)
    return num islands stats, largest islands stats
compute_statistics(p_values, num_islands, largest_islands)
```

```
Statistical Summary:
Number of Islands Statistics: {'Mean': 1.0, 'Standard Deviation': 0.0,
'Variance': 0.0}
Largest Island Size Statistics: {'Mean': 809728.8888888889, 'Standard
Deviation': 747438.4657030592, 'Variance': 558664260012.5432}
({'Mean': 1.0, 'Standard Deviation': 0.0, 'Variance': 0.0},
{'Mean': 809728.8888888889,
  'Standard Deviation': 747438.4657030592,
  'Variance': 558664260012.5432})
# Optimal p value
def find optimal p(p values, num islands, largest islands):
    max_islands_p = p_values[np.argmax(num_islands)]
    max islands value = np.max(num islands)
    largest_island_p = p_values[np.argmax(largest islands)]
    largest island value = np.max(largest islands)
    print("\nOptimal Probability Analysis:")
    print(f"p with Maximum Number of Islands: {max islands p} (Value:
{max islands value})")
    print(f"p with Largest Island Size: {largest island p} (Value:
{largest island value})")
find optimal p(p values, num islands, largest islands)
Optimal Probability Analysis:
p with Maximum Number of Islands: 0.1 (Value: 1)
p with Largest Island Size: 0.9 (Value: 1866582)
```