

## Importing Libraries

```
In [2]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score, confusion_matrix, classification_rep
```

## Load Dataset

```
In [3]: df = pd.read_excel("capstone dataset.xlsx")
print(df.head())
print(df.info())
print("\nMissing values:\n", df.isnull().sum())
```

|   | City Name   | Country | Urban Population (Millions) | \ |
|---|-------------|---------|-----------------------------|---|
| 0 | Chennai     | India   | 1.4                         |   |
| 1 | Delhi       | India   | 10.8                        |   |
| 2 | Mumbai      | India   | 11.5                        |   |
| 3 | New York    | USA     | 13.6                        |   |
| 4 | Los Angeles | USA     | 13.8                        |   |

|   | Population Growth Rate (%) | Green Space (%) | Air Quality Index (PM2.5) | \ |
|---|----------------------------|-----------------|---------------------------|---|
| 0 | 0.8                        | 44              | 99                        |   |
| 1 | 1.2                        | 24              | 64                        |   |
| 2 | 2.7                        | 13              | 50                        |   |
| 3 | 0.8                        | 27              | 30                        |   |
| 4 | 2.8                        | 41              | 85                        |   |

|   | CO2 Emissions (per capita) | Access to Clean Water (%) | \ |
|---|----------------------------|---------------------------|---|
| 0 | 4.9                        | 98                        |   |
| 1 | 5.6                        | 93                        |   |
| 2 | 1.8                        | 96                        |   |
| 3 | 4.9                        | 99                        |   |
| 4 | 0.9                        | 98                        |   |

|   | Waste Recycling Rate (%) | Public Transport Usage (%) | \ |
|---|--------------------------|----------------------------|---|
| 0 | 41                       | 59                         |   |
| 1 | 59                       | 64                         |   |
| 2 | 46                       | 71                         |   |
| 3 | 42                       | 75                         |   |
| 4 | 52                       | 51                         |   |

|   | Energy Consumption (kWh per capita) |
|---|-------------------------------------|
| 0 | 1594                                |
| 1 | 1006                                |
| 2 | 1352                                |
| 3 | 1652                                |
| 4 | 927                                 |

```
<class 'pandas.DataFrame'>
```

```
RangeIndex: 82 entries, 0 to 81
```

```
Data columns (total 11 columns):
```

| #  | Column                              | Non-Null Count | Dtype   |
|----|-------------------------------------|----------------|---------|
| 0  | City Name                           | 82 non-null    | str     |
| 1  | Country                             | 82 non-null    | str     |
| 2  | Urban Population (Millions)         | 82 non-null    | float64 |
| 3  | Population Growth Rate (%)          | 82 non-null    | float64 |
| 4  | Green Space (%)                     | 82 non-null    | int64   |
| 5  | Air Quality Index (PM2.5)           | 82 non-null    | int64   |
| 6  | CO2 Emissions (per capita)          | 82 non-null    | float64 |
| 7  | Access to Clean Water (%)           | 82 non-null    | int64   |
| 8  | Waste Recycling Rate (%)            | 82 non-null    | int64   |
| 9  | Public Transport Usage (%)          | 82 non-null    | int64   |
| 10 | Energy Consumption (kWh per capita) | 82 non-null    | int64   |

```
dtypes: float64(3), int64(6), str(2)
```

```
memory usage: 7.2 KB
```

```
None
```

```
Missing values:
```

|                             |   |
|-----------------------------|---|
| City Name                   | 0 |
| Country                     | 0 |
| Urban Population (Millions) | 0 |
| Population Growth Rate (%)  | 0 |
| Green Space (%)             | 0 |

```
Air Quality Index (PM2.5)          0
CO2 Emissions (per capita)         0
Access to Clean Water (%)          0
Waste Recycling Rate (%)           0
Public Transport Usage (%)         0
Energy Consumption (kWh per capita) 0
dtype: int64
```

### Data Cleaning

```
In [4]: df = df.drop_duplicates()
```

### Encoding Categorical Data

```
In [5]: le_city = LabelEncoder()
le_country = LabelEncoder()

df["City Name"] = le_city.fit_transform(df["City Name"])
df["Country"] = le_country.fit_transform(df["Country"])
```

### Feature Engineering

```
In [6]: df["Sustainability_Score"] = (
    df["Green Space (%)"] +
    df["Access to Clean Water (%)"] +
    df["Waste Recycling Rate (%)"] +
    df["Public Transport Usage (%)"]
) - (
    df["Air Quality Index (PM2.5)"] +
    df["CO2 Emissions (per capita)"]
)

df["Sustainable"] = df["Sustainability_Score"].apply(
    lambda x: 1 if x > df["Sustainability_Score"].median() else 0
)
```

### Feature Selection

```
In [7]: X = df.drop(["Sustainability_Score", "Sustainable"], axis=1)
y = df["Sustainable"]
```

### Train Test Split

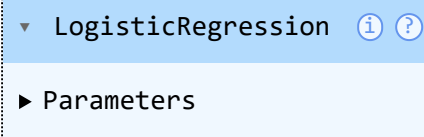
```
In [8]: X_train, X_test, y_train, y_test = train_test_split(
    X, y, test_size=0.2, random_state=42
)
```

### Feature Scaling

```
In [9]: scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)
```

### Training ML Model

```
In [10]: model = LogisticRegression()  
model.fit(X_train, y_train)
```

```
Out[10]: 
```

Predictions

```
In [11]: y_pred = model.predict(X_test)
```

## Evaluation

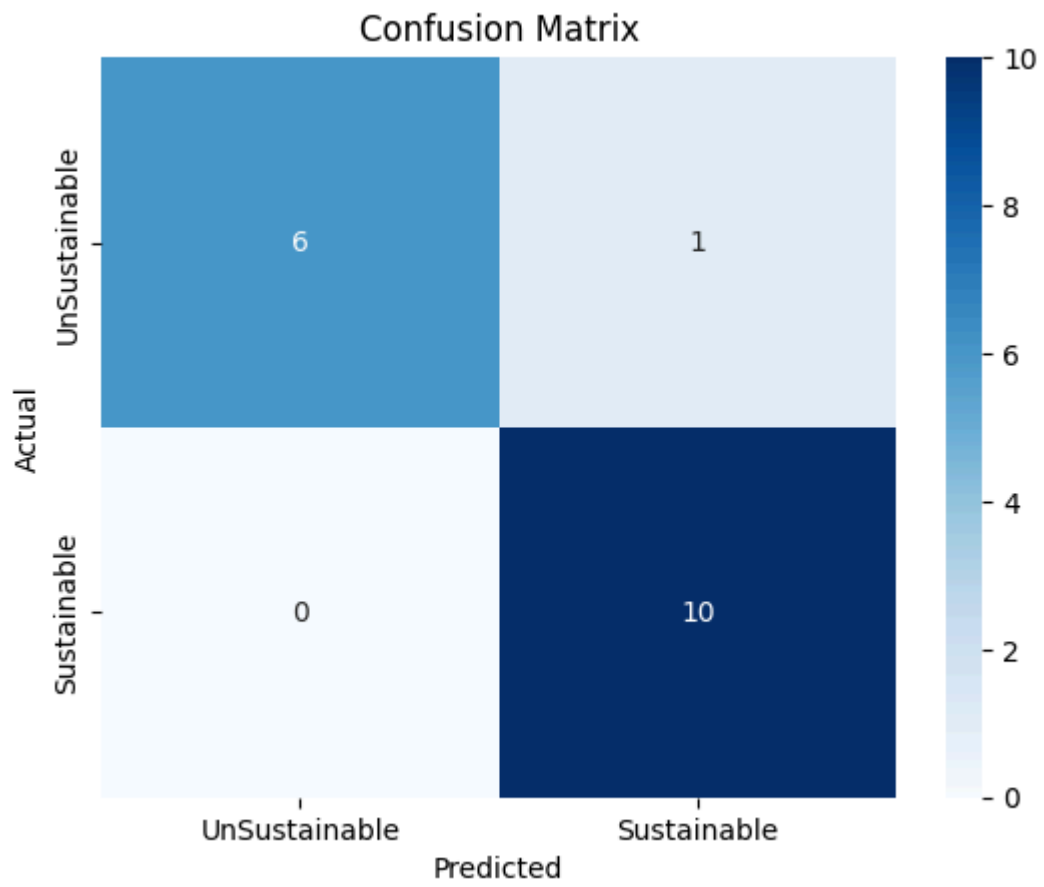
```
In [12]: print("\nAccuracy:", accuracy_score(y_test, y_pred))  
print("\nClassification Report:\n", classification_report(y_test, y_pred))
```

Accuracy: 0.9411764705882353

Classification Report:

|              | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0            | 1.00      | 0.86   | 0.92     | 7       |
| 1            | 0.91      | 1.00   | 0.95     | 10      |
| accuracy     |           |        | 0.94     | 17      |
| macro avg    | 0.95      | 0.93   | 0.94     | 17      |
| weighted avg | 0.95      | 0.94   | 0.94     | 17      |

```
In [13]: conf_matrix = confusion_matrix(y_test, y_pred)  
plt.figure()  
sns.heatmap(conf_matrix, annot=True, fmt="d", cmap="Blues",  
            xticklabels=["UnSustainable", "Sustainable"],  
            yticklabels=["UnSustainable", "Sustainable"])  
plt.xlabel("Predicted")  
plt.ylabel("Actual")  
plt.title("Confusion Matrix")  
plt.show()
```



In [ ]: