

## # SLIP 1

Q1. Write a R program to add, multiply and divide two vectors of integer type. (Vector length should be minimum 4)

```
vector1 <- c(1, 2, 3, 4)
vector2 <- c(5, 6, 7, 8)
addition_result <- vector1 + vector2
print("Addition Result:")
print(addition_result)
multiplication_result <- vector1 * vector2
print("Multiplication Result:")
print(multiplication_result)
division_result <- vector1 / vector2
print("Division Result:")
print(division_result)
```

---

Q2. Consider the student data set. It can be downloaded from:

[https://drive.google.com/open?id=1oakZCv7g3mlmCSdv9J8kdSaqO5\\_6dIOw](https://drive.google.com/open?id=1oakZCv7g3mlmCSdv9J8kdSaqO5_6dIOw) .

Write a programme in python to apply simple linear regression and find out mean absolute error, mean squared error and root mean squared error

Ans :-

```
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_absolute_error, mean_squared_error
data = pd.read_csv("student_data.csv")
```

```
X = data[['X']] # Feature
y = data['Y'] # Target
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
model = LinearRegression()
model.fit(X_train, y_train)
y_pred = model.predict(X_test)
mae = mean_absolute_error(y_test, y_pred)
mse = mean_squared_error(y_test, y_pred)
rmse = np.sqrt(mse)
print("Mean Absolute Error (MAE):", mae)
print("Mean Squared Error (MSE):", mse)
print("Root Mean Squared Error (RMSE):", rmse)
```

---

## # SLIP 2

# Q1. Write an R program to calculate the multiplication table using a function

# Function to generate a multiplication table

ANS:-

```
multiplication_table <- function(number, limit) {  
  for (i in 1:limit) {  
    result <- number * i  
    cat(paste(number, "x", i, "=", result), "\n")  
  }  
}  
  
number <- as.integer(readline("Enter a number for the multiplication table: "))  
limit <- as.integer(readline("Enter the limit for the multiplication table: "))  
multiplication_table(number, limit)
```

---

Q2. Write a python program to implement k-means algorithms on a synthetic dataset

ANS:-

```
import numpy as np  
import matplotlib.pyplot as plt  
from sklearn.cluster import KMeans  
from sklearn.datasets import make_blobs  
# Generate a synthetic dataset with make_blobs  
n_samples = 300  
n_features = 2  
n_clusters = 3  
  
X, y = make_blobs(n_samples=n_samples, n_features=n_features, centers=n_clusters,  
random_state=42)  
  
kmeans = KMeans(n_clusters=n_clusters)
```

```
kmeans.fit(X)
cluster_centers = kmeans.cluster_centers_
labels = kmeans.labels_
plt.scatter(X[:, 0], X[:, 1], c=labels, cmap='viridis')
plt.scatter(cluster_centers[:, 0], cluster_centers[:, 1], marker='x', s=200, linewidths=3,
color='red')
plt.title("K-Means Clustering")
plt.show()
```

---

### # SLIP 3

Q1. Write a R program to reverse a number and also calculate the sum of digits of that number.(i don't know how to run this r program)

ANS:-

```
reverse_number <- function(number) {  
  reversed <- 0  
  while (number > 0) {  
    digit <- number %% 10  
    reversed <- reversed * 10 + digit  
    number <- number %/% 10  
  }  
  return(reversed)  
}  
  
sum_of_digits <- function(number) {  
  sum_digits <- 0  
  while (number > 0) {  
    digit <- number %% 10  
    sum_digits <- sum_digits + digit  
    number <- number %/% 10  
  }  
  return(sum_digits)  
}  
  
number <- as.integer(readline("Enter a number: "))  
reversed <- reverse_number(number)  
sum_digits <- sum_of_digits(number)  
cat("Reversed Number:", reversed, "\n")  
cat("Sum of Digits:", sum_digits, "\n")
```

---

**Q2. Consider the following observations/data. And apply simple linear regression and find out estimated coefficients  $b_0$  and  $b_1$ . ( use numpy package)**

**$x=[0,1,2,3,4,5,6,7,8,9,11,13]$**

**$y = ([1, 3, 2, 5, 7, 8, 8, 9, 10, 12, 16, 18])$**

**ANS:-**

```
import numpy as np
x = np.array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 11, 13])
y = np.array([1, 3, 2, 5, 7, 8, 8, 9, 10, 12, 16, 18])
x_mean = np.mean(x)
y_mean = np.mean(y)
numerator = np.sum((x - x_mean) * (y - y_mean))
denominator = np.sum((x - x_mean) ** 2)
b1 = numerator / denominator
b0 = y_mean - b1 * x_mean
print("b0 (intercept):", b0)
print("b1 (slope):", b1)
```

---

#### # SLIP 4

Q1. Write a R program to calculate the sum of two matrices of given size

Define the size of the matrices

ANS:-

```
rows <- 3
cols <- 3
matrix1 <- matrix(1:9, nrow = rows, ncol = cols)
matrix2 <- matrix(9:1, nrow = rows, ncol = cols)
sum_matrix <- matrix1 + matrix2
cat("Matrix 1:\n")
print(matrix1)
cat("Matrix 2:\n")
print(matrix2)
cat("Sum of the matrices:\n")
print(sum_matrix)
```

---

Q2. Consider following dataset

weather=['Sunny','Sunny','Overcast','Rainy','Rainy','Rainy','Overcast','Sunny','Sunny','Rainy','Sunny','Overcast','Overcast','Rainy']temp=['Hot','Hot','Hot','Mild','Cool','Cool','Cool','Mild','Cool','Mild','Mild','Mild','Hot','Mild']play=['No','No','Yes','Yes','Yes','No','Yes','No','Yes','Yes','Yes','Yes','No'].Use Naïve Bayes algorithm to predict [0: Overcast, 2: Mild]tuple belongs to which class whether to play the sports or not.

ANS :-

```
from collections import Counter
```

```
from functools import reduce
```

```
weather = ['Sunny', 'Sunny', 'Overcast', 'Rainy', 'Rainy', 'Rainy', 'Overcast', 'Sunny', 'Sunny', 'Rainy', 'Sunny', 'Overcast', 'Overcast', 'Rainy']
```

```

temp = ['Hot', 'Hot', 'Hot', 'Mild', 'Cool', 'Cool', 'Cool', 'Mild', 'Cool', 'Mild', 'Mild', 'Mild',
'Hot', 'Mild']

play = ['No', 'No', 'Yes', 'Yes', 'Yes', 'No', 'Yes', 'No', 'Yes', 'Yes', 'Yes', 'Yes', 'Yes', 'No']

new_weather = 'Overcast'

new_temp = 'Mild'

prior_yes = Counter(play)['Yes'] / len(play)

prior_no = Counter(play)['No'] / len(play)

def conditional_prob_weather(class_value):

    return Counter([weather[i] for i, val in enumerate(play) if val ==
class_value])[new_weather] / Counter(play)[class_value]

conditional_prob_weather_yes = conditional_prob_weather('Yes')

conditional_prob_weather_no = conditional_prob_weather('No')

def conditional_prob_temp(class_value):

    return Counter([temp[i] for i, val in enumerate(play) if val == class_value])[new_temp] /
Counter(play)[class_value]

conditional_prob_temp_yes = conditional_prob_temp('Yes')

conditional_prob_temp_no = conditional_prob_temp('No')

posterior_yes = prior_yes * conditional_prob_weather_yes * conditional_prob_temp_yes

posterior_no = prior_no * conditional_prob_weather_no * conditional_prob_temp_no

prediction = 'Yes' if posterior_yes > posterior_no else 'No'

print('Prior Probability (Yes):', prior_yes)

print('Prior Probability (No):', prior_no)

print('Posterior Probability (Yes):', posterior_yes)

print('Posterior Probability (No):', posterior_no)

print('Prediction:', prediction)

```

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# SLIP 5

Q1. Write a R program to concatenate two given factors. Define two factors -

ANS :-

```
factor1 <- factor(c("A", "B", "C", "D"))
factor2 <- factor(c("E", "F", "G", "H"))
concatenated_factor <- c(factor1, factor2)
print(concatenated_factor)
```

---

Q2. Write a Python program build Decision Tree Classifier using Scikit-learn package for diabetes data set (download database from <https://www.kaggle.com/uciml/pima-indians-diabetes-database>)

ANS:-

```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import accuracy_score, classification_report
diabetes_data = pd.read_csv("diabetes.csv") # Replace with the actual path to the dataset
X = diabetes_data.drop("Outcome", axis=1)
y = diabetes_data["Outcome"]
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42)
clf = DecisionTreeClassifier(random_state=42)
clf.fit(X_train, y_train)
y_pred = clf.predict(X_test)
accuracy = accuracy_score(y_test, y_pred)
report = classification_report(y_test, y_pred)
print("Accuracy:", accuracy)
print("\nClassification Report:\n", report)
```

---

# SLIP 6

Q1. Write a R program to create a data frame using two given vectors and display the duplicate elements.

ANS :-

```
vector1 <- c(1, 2, 3, 4, 5, 6, 2, 8, 9, 10)
vector2 <- c("A", "B", "C", "D", "E", "F", "A", "H", "I", "J")
my_data_frame <- data.frame(Vector1 = vector1, Vector2 = vector2)
duplicates <- my_data_frame[duplicated(my_data_frame) | duplicated(my_data_frame,
fromLast = TRUE), ]
print(duplicates)
```

---

Q2. Write a python program to implement hierarchical Agglomerative clustering algorithm.

(Download Customer.csv dataset from [github.com](https://github.com)).

ANS :-

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.cluster import AgglomerativeClustering
from scipy.cluster.hierarchy import dendrogram, linkage
data = pd.read_csv("customer.csv")
X = data[["Age", "Income"]]
n_clusters = 2 # Number of clusters to create
linkage_type = 'ward' # Linkage method
model = AgglomerativeClustering(n_clusters=n_clusters, linkage=linkage_type)
labels = model.fit_predict(X)
linkage_matrix = linkage(X, method=linkage_type)
dendrogram(linkage_matrix, orientation="top")
plt.title("Dendrogram")
```

```
plt.xlabel("Data Points")
plt.ylabel("Distance")
plt.show()

data["Cluster"] = labels

for cluster in range(n_clusters):
    print(f"Cluster {cluster}:")
    cluster_data = data[data["Cluster"] == cluster]
    print(cluster_data)

for cluster in range(n_clusters):
    cluster_data = data[data["Cluster"] == cluster]
    plt.scatter(cluster_data["Age"], cluster_data["Income"], label=f"Cluster {cluster}")

plt.xlabel("Age")
plt.ylabel("Income")
plt.legend()

plt.title("Hierarchical Agglomerative Clustering")

plt.show()
```

---

**# SLIP 7**

**Q1. Write a R program to create a sequence of numbers from 20 to 50 and find the mean of numbers from 20 to 60 and sum of numbers from 51 to 91.**

**ANS :-**

```
sequence_20_to_50 <- seq(20, 50)
mean_20_to_60 <- mean(sequence_20_to_50[sequence_20_to_50 <= 60])
sequence_51_to_91 <- seq(51, 91)
sum_51_to_91 <- sum(sequence_51_to_91)
cat("Mean of numbers from 20 to 60:", mean_20_to_60, "\n")
cat("Sum of numbers from 51 to 91:", sum_51_to_91, "\n")
```

---

**Q2. Consider the following observations/data. And apply simple linear regression and find out estimated coefficients b1 and b1 Also analyse the performance of the model**

**(Use sklearn package)**

```
x = np.array([1,2,3,4,5,6,7,8])
y = np.array([7,14,15,18,19,21,26,23])
```

**ANS :-**

```
import numpy as np
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error, r2_score
x = np.array([1, 2, 3, 4, 5, 6, 7, 8])
y = np.array([7, 14, 15, 18, 19, 21, 26, 23])
x = x.reshape(-1, 1)
y = y.reshape(-1, 1)
model = LinearRegression()
model.fit(x, y)
b0 = model.intercept_[0]
```

```
b1 = model.coef_[0][0]
y_pred = model.predict(x)
mse = mean_squared_error(y, y_pred)
r2 = r2_score(y, y_pred)
print("Estimated Coefficients:")
print(f"Intercept (b0): {b0}")
print(f"Slope (b1): {b1}")
print("\nPerformance Metrics:")
print(f"Mean Squared Error (MSE): {mse}")
print(f"R-squared (R2): {r2}")
```

---

## # SLIP 8

**Q1. Write a R program to get the first 10 Fibonacci numbers. Function to generate the first 10 Fibonacci numbers** `get_first_10_fibonacci <- function() { n <- 10 # Number of Fibonacci numbers to generate`

**ANS :-**

```
fibonacci <- numeric(n)

if (n >= 1) {
  fibonacci[1] <- 0
}

if (n >= 2) {
  fibonacci[2] <- 1
}

for (i in 3:n) {
  fibonacci[i] <- fibonacci[i - 1] + fibonacci[i - 2]
}

return(fibonacci)
}

fibonacci_numbers <- get_first_10_fibonacci()

cat("First 10 Fibonacci numbers:", paste(fibonacci_numbers, collapse = ", "))
```

---

**Q2. Write a python program to implement k-means algorithm to build prediction model (Use Credit Card Dataset CC GENERAL.csv Download from kaggle.com)**

**ANS :-**

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.cluster import KMeans

data = pd.read_csv("CCGENERAL.csv")
X = data[["PURCHASES", "CASH_ADVANCE"]]
model = KMeans(n_clusters=k, random_state=42)
model.fit(X)
data["Cluster"] = model.labels_
for cluster in range(k):
    cluster_data = data[data["Cluster"] == cluster]
    plt.scatter(cluster_data["PURCHASES"], cluster_data["CASH_ADVANCE"], label=f"Cluster {cluster}")
plt.xlabel("PURCHASES")
plt.ylabel("CASH_ADVANCE")
plt.legend()
plt.title("K-Means Clustering")
plt.show()
```

---

**# SLIP 9**

**Q1. Write an R program to create a Data frames which contain details of 5 employees and display summary of the data**

**ANS :-**

```
employee_data <- data.frame(  
  EmployeeID = c(1, 2, 3, 4, 5),  
  Name = c("saurabh", "yogesh", "arbaj", "pranav", "hrushali"),  
  Age = c(30, 28, 32, 25, 34),  
  Department = c("HR", "Engineering", "Finance", "Marketing", "Sales"),  
  Salary = c(50000, 60000, 55000, 48000, 65000)  
)  
summary(employee_data)
```

---



**Q2. Write a Python program to build an SVM model to Cancer dataset. The dataset is available in the scikit-learn library. Check the accuracy of model with precision and recall.**

**ANS :-**

```
from sklearn import datasets
from sklearn.model_selection import train_test_split
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score, precision_score, recall_score

data = datasets.load_breast_cancer()
X = data.data
y = data.target

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
svm_model = SVC(kernel='linear')
svm_model.fit(X_train, y_train)
y_pred = svm_model.predict(X_test)
accuracy = accuracy_score(y_test, y_pred)
precision = precision_score(y_test, y_pred)
recall = recall_score(y_test, y_pred)

# Print the results
print("Accuracy:", accuracy)
print("Precision:", precision)
print("Recall:", recall)
```

---

## # SLIP 10

**Q1. Write a R program to find the maximum and the minimum value of a given vector [10 Marks]**

**ANS :-**

```
my_vector <- c(12, 34, 5, 23, 9, 2, 45, 8, 31)
max_value <- max(my_vector)
min_value <- min(my_vector)
cat("Maximum value:", max_value, "\n")
cat("Minimum value:", min_value, "\n")
```

---

**Q2. Write a Python Programme to read the dataset ("Iris.csv"). dataset download from (<https://archive.ics.uci.edu/ml/datasets/iris>) and apply Apriori algorithm.**

**ANS :-**

```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score
data = pd.read_csv("iris.csv")
X = data.drop("species", axis=1)
y = data["species"]
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42)
clf = RandomForestClassifier(n_estimators=100, random_state=42)
clf.fit(X_train, y_train)
y_pred = clf.predict(X_test)
accuracy = accuracy_score(y_test, y_pred)
print("Accuracy:", accuracy)
```

---

## # SLIP 11

Q1. Write a R program to find all elements of a given list that are not in another given list. =

```
list("x", "y", "z") = list("X", "Y", "Z", "x", "y", "z")
```

ANS :-

```
list1 <- list("x", "y", "z")
list2 <- list("X", "Y", "Z", "x", "y", "z")
elements_not_in_list2 <- list1[!(list1 %in% list2)]
print(elements_not_in_list2)
list1 <- list("x", "y", "z")
list2 <- list("X", "Y", "Z", "x", "y", "z")
vector1 <- unlist(list1)
vector2 <- unlist(list2)
elements_not_in_list2 <- vector1[!(vector1 %in% vector2)]
print(elements_not_in_list2)
```

---

Q2. Write a python program to implement hierarchical clustering algorithm.(Download Wholesale customers data dataset from [github.com](https://github.com)).

ANS :-

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from scipy.cluster.hierarchy import linkage, dendrogram
from sklearn.preprocessing import StandardScaler
from sklearn.cluster import AgglomerativeClustering
data = pd.read_csv('wholesale_customers_data.csv')
selected_features = ['Fresh', 'Milk', 'Grocery']
```

```
X = data[selected_features]
scaler = StandardScaler()
X_std = scaler.fit_transform(X)
linkage_matrix = linkage(X_std, method='ward')
plt.figure(figsize=(12, 6))
dendrogram(linkage_matrix, labels=data.index, leaf_rotation=90, leaf_font_size=12)
plt.title("Hierarchical Clustering Dendrogram")
plt.xlabel("Wholesale Customers")
plt.ylabel("Distance")
plt.show()

n_clusters = 3
model = AgglomerativeClustering(n_clusters=n_clusters)
data['Cluster'] = model.fit_predict(X_std)
for cluster_id in range(n_clusters):
    cluster_data = data[data['Cluster'] == cluster_id]
    print(f"Cluster {cluster_id}:\n{cluster_data[selected_features]}\n")
```

---

# SLIP 12

Q1. Write a R program to create a Dataframes which contain details of 5employees and display the details. Employee contain (empno,empname,gender,age,designation)

ANS :-

```
employee_data <- data.frame(  
  empno = c(1, 2, 3, 4, 5),  
  empname = c("saurabh", "pranav", "yogesh", "hrushali", "arbaj"),  
  gender = c("Male", "Male", "Male", "Female", "Male"),  
  age = c(20, 20, 20, 20, 20),  
  designation = c("Manager", "Engineer", "Analyst", "Manager", "Designer")  
)  
print(employee_data)
```

---

Q2. Write a python program to implement multiple Linear Regression model for a car dataset. Dataset can be downloaded from:

[https://www.w3schools.com/python/python\\_ml\\_multiple\\_regression.asp](https://www.w3schools.com/python/python_ml_multiple_regression.asp)

ANS :-

```
import pandas as pd  
import numpy as np  
from sklearn.model_selection import train_test_split  
from sklearn.linear_model import LinearRegression  
from sklearn.metrics import mean_squared_error, r2_score  
df = pd.read_csv('data.csv')  
X = df[['Volume', 'Weight']]  
y = df['CO2']  
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

```
model = LinearRegression()
model.fit(X_train, y_train)
y_pred = model.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
r2 = r2_score(y_test, y_pred)
print("Coefficients:", model.coef_)
print("Intercept:", model.intercept_)
print("Mean Squared Error (MSE):", mse)
print("R-squared (R2):", r2)
```

---

# SLIP 13

Q1. Draw a pie chart using R programming for the following data distribution:

Digits on Dice 1 2 3 4 5 6 Frequency of getting each number 7 2 6 3 4 8

ANS :-

```
numbers <- c(1, 2, 3, 4, 5, 6)
```

```
frequency <- c(7, 2, 6, 3, 4, 8)
```

```
pie(frequency, labels = numbers, main = "Dice Roll Frequencies", col =  
rainbow(length(numbers)))
```

```
legend("topright", numbers, fill = rainbow(length(numbers))
```

```
title("Dice Roll Frequencies")
```

---

Q2. Write a Python program to read "StudentsPerformance.csv" file. Solve following:

- To display the shape of dataset. - To display the top rows of the dataset with their columns. Note: Download dataset from following link :

([https://www.kaggle.com/spscientist/students-performance-inexams?](https://www.kaggle.com/spscientist/students-performance-inexams?select=StudentsPerformance.csv)  
select=StudentsPerformance.csv)

ANS :-

```
import pandas as pd
```

```
df = pd.read_csv("StudentsPerformance.csv")
```

```
print("Shape of the dataset:", df.shape)
```

```
print("Top rows of the dataset:")
```

```
print(df.head())
```

---

## # SLIP 14

**Q1. Write a script in R to create a list of employees (name) and perform the following:**

- a. Display names of employees in the list.**
- b. Add an employee at the end of the list**
- c. Remove the third element of the list.**

**ANS :-**

```
employees <- list("saurabh", "pranav", "hrushali", "yogesh")  
print(employees)  
new_employee <- "arbaj"  
employees <- c(employees, new_employee)  
print(employees)  
removed_employee <- employees[3]  
employees <- employees[-3]  
print(removed_employee)  
print(employees)
```

---



**Q2. Write a Python Programme to apply Apriori algorithm on Groceries dataset. Dataset can be downloaded from ([https://github.com/amankharwal/Websitedata/blob/master/Groceries\\_dataset.csv](https://github.com/amankharwal/Websitedata/blob/master/Groceries_dataset.csv)). Also display support and confidence for each rule.**

**ANS:-**

```
import numpy as np

import matplotlib.pyplot as plt

import pandas as pd

from apyori import apriori

store_data = pd.read_csv('Groceries_dataset.csv', header=None)

records = []

for i in range(0, 300):

    records.append([str(store_data.values[i, j]) for j in range(0, 3)])

association_rules = apriori(records, min_support=0.0045, min_confidence=0.2, min_lift=3,
min_length=2)

association_results = list(association_rules)

print(len(association_results))

print(association_results[0])

for item in association_results:

    pair = item[0]

    items = [x for x in pair]

    print("Rule: " + items[0] + " -> " + items[1])

    print("Support: " + str(item[1]))

    print("Confidence: " + str(item[2][0][2]))

    print("Lift: " + str(item[2][0][3]))

    print("=====")
```

---

# SLIP 15

Q1. Write a R program to add, multiply and divide two vectors of integer type. (vector length should be minimum 4)

ANS :-

```
vector1 <- c(5, 10, 15, 20)
vector2 <- c(2, 4, 5, 8)
addition_result <- vector1 + vector2
cat("Vector Addition Result: ", addition_result, "\n")
multiplication_result <- vector1 * vector2
cat("Vector Multiplication Result: ", multiplication_result, "\n")
division_result <- vector1 / vector2
cat("Vector Division Result: ", division_result, "\n")
```

---

Q2. Write a Python program build Decision Tree Classifier for shows.csv from pandas and predict class label for show starring a 40 years old American comedian, with 10 years of experience, and a comedy ranking of 7? Create a csv file as shown in

[https://www.w3schools.com/python/python\\_ml\\_decision\\_tree.asp](https://www.w3schools.com/python/python_ml_decision_tree.asp)

ANS :-

```
import pandas as pd
from sklearn.tree import DecisionTreeClassifier
data = pd.read_csv("shows.csv")
data['Nationality'] = data['Nationality'].apply(lambda x: 1 if x == 'American' else 0)
X = data[['Age', 'Nationality', 'Experience', 'Rank']]
y = data['Go']
clf = DecisionTreeClassifier()
clf = clf.fit(X, y)
```

```
input_data = pd.DataFrame({'Age': [40], 'Nationality': [1], 'Experience': [10], 'Rank': [7]})
predicted_label = clf.predict(input_data)
result_df = pd.DataFrame({'Age': input_data['Age'],
                          'Nationality': ['American' if input_data['Nationality'].values[0] == 1 else
                          'Other'],
                          'Experience': input_data['Experience'],
                          'Rank': input_data['Rank'],
                          'Predicted Label': predicted_label})
result_df.to_csv('predicted_show_label.csv', index=False)
print("Predicted Label:", predicted_label)
print("Result saved to predicted_show_label.csv")
```

---

## SLIP 16

**Q1. Write a R program to create a simple bar plot of given data Year Export Import 2001 26 35, 2002 32 40, 2003 35 50**

**ANS :-**

```
data <- data.frame(  
  Year = c(2001, 2002, 2003),  
  Export = c(26, 32, 35),  
  Import = c(35, 40, 50)  
)  
barplot(height = t(data[, c("Export", "Import")]),  
  beside = TRUE,  
  names.arg = data$Year,  
  col = c("blue", "red"),  
  legend.text = c("Export", "Import"),  
  args.legend = list(title = "Type"))  
x <- c("2001", "2002", "2003")  
xlabel <- "Year"  
ylabel <- "Value"  
title <- "Export and Import by Year"  
axis(1, at = 1:3, labels = x, pos = 0)  
axis(2, las = 1, at = seq(0, 60, by = 10))  
title(main = title, xlab = xlabel, ylab = ylabel)
```

---

**Q2. Write a Python program build Decision Tree Classifier using Scikit-learn package for diabetes data set (download database from <https://www.kaggle.com/uciml/pima-indiansdiabetes-database>)**

**ANS :-**

```
import pandas as pd

from sklearn.tree import DecisionTreeClassifier

from sklearn.model_selection import train_test_split

from sklearn.metrics import accuracy_score

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

X = data.drop('Outcome', axis=1)

y = data['Outcome']

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

clf = DecisionTreeClassifier()

clf.fit(X_train, y_train)

y_pred = clf.predict(X_test)

accuracy = accuracy_score(y_test, y_pred)

print("Accuracy of the Decision Tree Classifier:", accuracy)
```

---

# SLIP 17

Q1. Write a R program to get the first 20 Fibonacci numbers Initialize the first two Fibonacci numbers.

ANS :-

```
fibonacci <- c(0, 1)
for (i in 3:20) {
  next_fib <- fibonacci[i-1] + fibonacci[i-2]
  fibonacci <- c(fibonacci, next_fib)
}
cat("The first 20 Fibonacci numbers are:\n")
cat(fibonacci, sep = " ")
```

---

Q2. Write a python programme to implement multiple linear regression model for stock market data frame as follows:

```
Stock_Market =
{'Year':[2017,2017,2017,2017,2017,2017,2017,2017,2017,2017,2017,2016,2016,2016,2016,2016,2016,2016,2016,2016,2016,2016], 'Month': [12,11,10,9,8,7,6,5,4,3,2,1,12,11,10,9,8,7,6,5,4,3,2,1],
'Interest_Rate':[2.75,2.5,2.5,2.5,2.5,2.5,2.5,2.5,2.25,2.25,2.2,2.2,1.75,1.75,1.75,1.75,1.75,1.75,1.75,1.75,1.75,1.75], 'Unemployment_Rate':[5.3,5.3,5.3,5.3,5.4,5.6,5.5,5.5,5.5,5.6,5.7,5.9,6,5.9,5.8,6.1,6.2,6.1,6.1,6.1,5.9,6.2,6.2,6.1], 'Stock_Index_Price': [1464,1394,1357,1293,1256,1254,1234,1195,1159,1167,1130,1075,1047,965,943,958,971,949,884,866,876,822,704,719] }
```

ANS :-

```
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.linear_model import LinearRegression
```

```

Stock_Market = {
    'Year': [2017, 2017, 2017, 2017, 2017, 2017, 2017, 2017, 2017, 2017, 2017, 2017, 2016,
2016, 2016, 2016, 2016, 2016, 2016, 2016, 2016, 2016, 2016],
    'Month': [12, 11, 10, 9, 8, 7, 6, 5, 4, 3, 2, 1, 12, 11, 10, 9, 8, 7, 6, 5, 4, 3, 2, 1],
    'Interest_Rate': [2.75, 2.5, 2.5, 2.5, 2.5, 2.5, 2.5, 2.25, 2.25, 2.25, 2, 2, 2, 1.75, 1.75, 1.75,
1.75, 1.75, 1.75, 1.75, 1.75, 1.75, 1.75],
    'Unemployment_Rate': [5.3, 5.3, 5.3, 5.3, 5.4, 5.6, 5.5, 5.5, 5.5, 5.6, 5.7, 5.9, 6, 5.9, 5.8,
6.1, 6.2, 6.1, 6.1, 6.1, 5.9, 6.2, 6.2, 6.1],
    'Stock_Index_Price': [1464, 1394, 1357, 1293, 1256, 1254, 1234, 1195, 1159, 1167, 1130,
1075, 1047, 965, 943, 958, 971, 949, 884, 866, 876, 822, 704, 719]
}

df = pd.DataFrame(Stock_Market)
X = df[['Interest_Rate']]
y = df['Stock_Index_Price']
model = LinearRegression()
model.fit(X, y)
predicted_prices = model.predict(X)
plt.figure(figsize=(10, 6))
plt.scatter(X, y, color='blue', label='Actual Prices')
plt.plot(X, predicted_prices, color='red', label='Predicted Prices')
plt.xlabel('Interest Rate')
plt.ylabel('Stock Index Price')
plt.legend()
plt.title('Stock Market Price vs. Interest Rate')
plt.show()

```

---

# SLIP 18

Q1. Write a R program to find the maximum and the minimum value of a given vector

Create a vector :

ANS :-

```
my_vector <- c(34, 56, 12, 98, 23, 45, 67)
max_value <- max(my_vector)
min_value <- min(my_vector)
cat("Maximum Value:", max_value, "\n")
cat("Minimum Value:", min_value, "\n")
```

---

Q2. Consider the following observations/data. And apply simple linear regression and find out estimated coefficients b1 and b1 Also analyse the performance of the model

(Use sklearn package) x = np.array([1,2,3,4,5,6,7,8]) y = np.array([7,14,15,18,19,21,26,23])

ANS :-

```
import numpy as np
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error, r2_score
x = np.array([1, 2, 3, 4, 5, 6, 7, 8])
y = np.array([7, 14, 15, 18, 19, 21, 26, 23])
x = x.reshape(-1, 1)
regression_model = LinearRegression()
regression_model.fit(x, y)
b0 = regression_model.intercept_
b1 = regression_model.coef_[0]
y_pred = regression_model.predict(x)
mse = mean_squared_error(y, y_pred)
r_squared = r2_score(y, y_pred)
```



```
print(f"Estimated Coefficient b0 (Intercept): {b0:.4f}")
```

```
print(f"Estimated Coefficient b1 (Slope): {b1:.4f}")
```

```
print(f"Mean Squared Error (MSE): {mse:.4f}")
```

```
print(f"R-squared (R2): {r_squared:.4f}")
```

---

# SLIP 19

Q1. Write a R program to create a Dataframes which contain details of 5 Students and display the details.

Students contain (Rollno, Studname, Address, Marks)

ANS :-

```
students_data <- data.frame(  
  Rollno = c(1, 2, 3),  
  Studname = c("saurabh", "yogesh", "arbaj"),  
  Address = c("wagholi", "pune", "hadapsar"),  
  Marks = c(78, 92, 88)  
)  
print(students_data)
```

---

Q2. Write a python program to implement multiple Linear Regression model for a car dataset. Dataset can be downloaded from:

[https://www.w3schools.com/python/python\\_ml\\_multiple\\_regression.asp](https://www.w3schools.com/python/python_ml_multiple_regression.asp)

ANS:-

```
import pandas as pd  
from sklearn.model_selection import train_test_split  
from sklearn.linear_model import LinearRegression  
from sklearn.metrics import mean_squared_error, r2_score  
data = pd.read_csv("car_data.csv")  
X = data[['Weight', 'Volume']]  
y = data['CO2']  
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

```
model = LinearRegression()
model.fit(X_train, y_train)

y_pred = model.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
r_squared = r2_score(y_test, y_pred)
b0 = model.intercept_
b1, b2 = model.coef_
print("Model Coefficients:")
print(f"Intercept (b0): {b0:.4f}")
print(f"Weight (b1): {b1:.4f}")
print(f"Volume (b2): {b2:.4f}")
print(f"Mean Squared Error (MSE): {mse:.4f}")
print(f"R-squared (R2): {r_squared:.4f}")
```

---

# SLIP 20

Q1. Write a R program to create a data frame from four given vectors

ANS :-

```
vector1 <- c(1, 2, 3, 4, 5)
vector2 <- c("A", "B", "C", "D", "E")
vector3 <- c(10.5, 20.5, 30.5, 40.5, 50.5)
vector4 <- c(TRUE, FALSE, TRUE, FALSE, TRUE)

df <- data.frame(
  int_num = vector1,
  char = vector2,
  float_num = vector3,
  boolean = vector4
)
print(df)
```

---

Q2. Write a python program to implement hierarchical Agglomerative clustering algorithm.  
(Download Customer.csv dataset from github.com).

ANS:-

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

from scipy.cluster.hierarchy import dendrogram, linkage
from sklearn.cluster import AgglomerativeClustering

data = pd.read_csv("Customer.csv")
X = data[['Annual Income (k$)', 'Spending Score (1-100)']]
```

```
linkage_matrix = linkage(X, method='ward')
plt.figure(figsize=(10, 5))
dendrogram(linkage_matrix)
plt.title('Hierarchical Agglomerative Clustering Dendrogram')
plt.xlabel('Samples')
plt.ylabel('Distance')
plt.show()

n_clusters = 3 # You can adjust the number of clusters as needed
model = AgglomerativeClustering(n_clusters=n_clusters, linkage='ward')
model.fit(X)
data['Cluster'] = model.labels_
print(data)
```

---