## Machine Learning Lab – Simplified Experiments

# **Ex 1:** Naive Bayes Classifier – Gender Classification

#### Aim:

Predict whether a person is **Male** or **Female** using simple features like long hair and forehead width.

```
import pandas as pd
from sklearn.model selection import train test split
from sklearn.naive bayes import GaussianNB
from sklearn.metrics import accuracy score
# Sample data
data = {
  'long hair': [1,0,1,0,1,0,0,1],
  'forehead width cm': [6.5,6.0,5.8,5.5,6.2,5.7,5.9,6.1],
  'gender': ['Female', 'Male', 'Female', 'Male', 'Female', 'Male', 'Female']
}
df = pd.DataFrame(data)
X = df[['long_hair','forehead_width_cm']]
y = df['gender']
# Split train-test
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=1)

# Train model
model = GaussianNB()
model.fit(X_train, y_train)

# Predict
y_pred = model.predict(X_test)
print("Predictions:", y_pred)
print("Accuracy:", accuracy_score(y_test, y_pred))

Output
Predictions: ['Female' 'Male' 'Female']
Accuracy: 1.0
```

## **Ex 2:** Linear Regression – Predict Marks from Study Hours

#### Aim:

Predict student marks based on hours studied.

import numpy as np

from sklearn.linear model import LinearRegression

$$X = \text{np.array}([[2],[4],[6],[8]])$$
# Hours studied  
 $y = \text{np.array}([20,40,60,80])$ # Marks  
 $model = \text{LinearRegression}()$ 

```
model.fit(X, y)
# Predict marks for 5 hours
pred = model.predict([[5]])
print("Predicted marks for 5 hours:", pred[0])
```

# Output

Predicted marks for 5 hours: 50.0

## **Q** Ex 3: K-Nearest Neighbors (KNN) – Iris Flower Classification

#### Aim:

Classify flowers using petal and sepal features.

from sklearn.datasets import load iris

from sklearn.model\_selection import train\_test\_split

from sklearn.neighbors import KNeighborsClassifier

from sklearn.metrics import accuracy score

X, y = iris.data, iris.target

X\_train, X\_test, y\_train, y\_test = train\_test\_split(X, y, test\_size=0.2, random state=0)

knn = KNeighborsClassifier(n neighbors=3)

knn.fit(X\_train, y\_train)

```
y_pred = knn.predict(X_test)
print("Accuracy:", accuracy score(y test, y pred))
Output
Accuracy: 1.0
```



### **Ex 4: Decision Tree Classifier**

#### Aim:

Use a decision tree to classify fruits as Apple or Orange based on weight and texture.

from sklearn.tree import DecisionTreeClassifier

```
# Features: [Weight, Texture(1=Smooth, 0=Rough)]
X = [[140,1],[130,1],[150,0],[170,0]]
y = ['Apple','Apple','Orange','Orange']
model = DecisionTreeClassifier()
model.fit(X, y)
pred = model.predict([[145,1]])
print("Prediction:", pred[0])
Output
```

Prediction: Apple

## **Ex 5: K-Means Clustering**

#### Aim:

```
Group students into clusters based on marks in two subjects.
```

import numpy as np

from sklearn.cluster import KMeans

import matplotlib.pyplot as plt

```
X = \text{np.array}([[70,80],[65,60],[90,95],[40,30],[35,40],[80,85]])
```

```
kmeans = KMeans(n clusters=2, random state=0)
```

kmeans.fit(X)

```
print("Cluster Centers:\n", kmeans.cluster_centers_)
```

print("Labels:", kmeans.labels )

plt.scatter(X[:,0], X[:,1], c=kmeans.labels\_, cmap='rainbow')

plt.scatter(kmeans.cluster\_centers\_[:,0], kmeans.cluster\_centers\_[:,1], color='black', marker='X')

plt.title("Student Clusters")

plt.show()

# Output

**Cluster Centers:** 

[[37.5 35.]

[76.25 80.]]

Labels: [1 1 1 0 0 1]

# Ex 6: Logistic Regression – Pass/Fail Prediction

### Aim:

Predict if a student passes based on study hours.

import numpy as np

from sklearn.linear model import LogisticRegression

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

y = np.array([0,0,0,0,1,1,1,1]) # 1=Pass, 0=Fail

model = LogisticRegression()

model.fit(X, y)

print("Predicted (5 hours):", model.predict([[5]])[0])

print("Predicted (2 hours):", model.predict([[2]])[0])

# Output

Predicted (5 hours): 1

Predicted (2 hours): 0



# **Ex 7: Support Vector Machine (SVM) – Binary Classification**

#### Aim:

Classify whether points belong to Class 0 or 1.

from sklearn import svm

## **II** Ex 8: Evaluation Metrics – Confusion Matrix

## Aim:

Evaluate model performance using accuracy and confusion matrix.

from sklearn.metrics import confusion\_matrix, accuracy\_score

Confusion Matrix:

```
[[1 \ 1]]
```

[1 3]]

Accuracy: 0.67

## Advanced Data Structures (ADS) – Simplified Experiments

# **Ex 1: Huffman Coding**

#### Aim:

Compress text using shorter binary codes for frequent characters.

import heapq

```
def huffman(text):
    freq = {c: text.count(c) for c in set(text)}
    heap = [[f, [c, ""]] for c, f in freq.items()]
    heapq.heapify(heap)

while len(heap) > 1:
    lo = heapq.heappop(heap)
    hi = heapq.heappop(heap)
    for p in lo[1:]:
        p[1] = '0' + p[1]
    for p in hi[1:]:
        p[1] = '1' + p[1]
    heapq.heappush(heap, [lo[0]+hi[0]] + lo[1:] + hi[1:])
```

```
return sorted(heapq.heappop(heap)[1:], key=lambda p: (len(p[-1]), p))

text = "HELLO"

codes = huffman(text)

print("Huffman Codes:", codes)

Output
```

## **Ex 2: Longest Common Subsequence (LCS)**

Find the longest sequence common to both strings.

Huffman Codes: [['L', '0'], ['O', '10'], ['H', '110'], ['E', '111']]

#### Aim:

def lcs(a, b):
 m, n = len(a), len(b)
 dp = [[0]\*(n+1) for \_ in range(m+1)]
 for i in range(m):
 for j in range(n):
 if a[i] == b[j]:
 dp[i+1][j+1] = dp[i][j] + 1
 else:
 dp[i+1][j+1] = max(dp[i][j+1], dp[i+1][j])
 return dp[m][n]

print("LCS length:", lcs("ACGTGCA", "GTCGACG"))

LCS length: 4

## **Ex 3: Traveling Salesman Problem (Nearest Neighbor)**

#### Aim:

Find a short path visiting all cities once.

import math

```
cities = [(0,0),(2,3),(5,4),(1,1)]
visited = [0]
total = 0
while len(visited) < len(cities):
  last = visited[-1]
  next city = min(
     [i for i in range(len(cities)) if i not in visited],
     key=lambda i: math.dist(cities[last], cities[i])
  )
  total += math.dist(cities[last], cities[next_city])
  visited.append(next city)
total += math.dist(cities[visited[-1]], cities[0])
print("Path:", visited)
print("Distance:", round(total, 2))
```

## Output

Path: [0, 3, 1, 2]

Distance: 12.06

# Ex 4: Randomized Quick Sort

#### Aim:

Sort numbers by choosing a random pivot each time.

import random

```
def quicksort(arr):
  if len(arr) <= 1:
     return arr
  pivot = random.choice(arr)
  left = [x for x in arr if x < pivot]
  mid = [x \text{ for } x \text{ in arr if } x == pivot]
  right = [x \text{ for } x \text{ in arr if } x > pivot]
  return quicksort(left) + mid + quicksort(right)
data = [10, 7, 8, 9, 1, 5]
print("Original:", data)
print("Sorted:", quicksort(data))
Output
```

Original: [10, 7, 8, 9, 1, 5]

Sorted: [1, 5, 7, 8, 9, 10]

## **Ex 5: Hash Table (Chaining)**

```
Aim:
Store and search values using hashing.
class HashTable:
  def init (self, size):
     self.size = size
    self.table = [[] for _ in range(size)]
  def insert(self, key):
     self.table[key % self.size].append(key)
  def display(self):
     for i, b in enumerate(self.table):
       print(i, ":", b)
h = HashTable(5)
for k in [10, 15, 20, 25, 30]:
  h.insert(k)
h.display()
Output
0:[10, 15, 20, 25, 30]
1:[]
```

2:[]

```
3:[]
```

4:[]

# **⊗** Ex 6: Graph – Maximum Flow (Edmonds-Karp)

## Aim:

Find max flow from source to sink in a flow network.

from collections import deque

```
def bfs(r, s, t, p):
  visited = [False]*len(r)
  queue = deque([s])
  visited[s] = True
  while queue:
     u = queue.popleft()
     for v in range(len(r)):
       if not visited[v] and r[u][v] > 0:
          queue.append(v)
          visited[v] = True
          p[v] = u
  return visited[t]
def max flow(graph, s, t):
  r = [row[:] for row in graph]
  p = [-1]*len(r)
```

```
flow = 0
  while bfs(r, s, t, p):
    path_flow = float('inf')
    v = t
    while v != s:
       u = p[v]
       path_flow = min(path_flow, r[u][v])
       v = u
    flow += path_flow
     v = t
    while v != s:
       u = p[v]
       r[u][v] = path flow
       r[v][u] += path_flow
       v = u
  return flow
graph = [
[0,16,13,0,0,0],
[0,0,10,12,0,0],
[0,4,0,0,14,0],
[0,0,9,0,0,20],
[0,0,0,7,0,4],
[0,0,0,0,0,0]
```

```
]
print("Max Flow:", max flow(graph, 0, 5))
Output
Max Flow: 23
Ex 7: Graph Coloring
Aim:
Color a graph so that no adjacent vertices have the same color.
def is safe(v, g, c, color):
  for i in range(len(g)):
    if g[v][i] == 1 and color[i] == c:
       return False
  return True
def color graph(g, m, v, color):
  if v == len(g):
    return True
  for c in range(1, m+1):
    if is safe(v, g, c, color):
       color[v] = c
       if color_graph(g, m, v+1, color):
         return True
       color[v] = 0
  return False
```

```
graph = [
[0,1,1,1],
[1,0,1,0],
[1,1,0,1],
[1,0,1,0]
]
m = 3
color = [0]*len(graph)
color_graph(graph, m, 0, color)
print("Colors:", color)

Output
Colors: [1, 2, 3, 2]
```

# **Ex 8: Parallel Merge Sort (Simplified Single-Core Version)**

## Aim:

Sort numbers by repeatedly dividing and merging.

```
def merge_sort(arr):
  if len(arr) > 1:
    mid = len(arr)//2
    L = arr[:mid]
    R = arr[mid:]
    merge_sort(L)
```

merge sort(R)

$$i = j = k = 0$$

while  $i < len(L)$  and  $j < len(R)$ :

 $if L[i] < R[j]$ :

 $arr[k] = L[i]$ ;  $i += 1$ 
 $else$ :

 $arr[k] = R[j]$ ;  $j += 1$ 
 $k += 1$ 
 $arr[k:] = L[i:] + R[j:]$ 

data = [9, 5, 1, 4, 3]

merge\_sort(data)

print("Sorted:", data)

# Output

Sorted: [1, 3, 4, 5, 9]