EXPERIMENT – 1

AIM : Write a program to implement Linear Regression using Scikit-Learn

CODE :

```
from sklearn.linear_model import LinearRegression
import numpy as np
X = np.array([[1], [2], [3], [4], [5]])
y = np.array([2, 4, 6, 8, 10])
model = LinearRegression()
model.fit(X, y)
pred = model.predict([[6]])
print("Predicted value for 6:", pred[0])
```

OUTPUT:

```
Predicted value for 6: 12.0
```

EXPERIMENT - 2

AIM: Logistic Regression Classification

CODE :

```
from sklearn.linear_model import LogisticRegression

X = [[1], [2], [3], [4]]

y = [0, 0, 1, 1]

model = LogisticRegression()

model.fit(X, y)

print('Prediction for 2.5:', model.predict([[2.5]])[0])
```

OUTPUT:

```
Prediction for 2.5: 0
```

EXPERIMENT - 3

AIM: Write a program to find the K-Nearest Neighbors (KNN)

CODE:

```
from sklearn.neighbors import KNeighborsClassifier

X = [[1,2], [2,3], [3,4], [6,7]]

y = [0, 0, 1, 1]

model = KNeighborsClassifier(n_neighbors=3)

model.fit(X, y)

print('Class for [4,5]:', model.predict([[4,5]])[0])
```

OUTPUT:

```
Class for [4,5]: 1
```

EXPERIMENT – 4

AIM: Write a program to implement Decision Tree Classifier

CODE:

```
from sklearn.tree import DecisionTreeClassifier
X = [[0,0], [1,1], [2,2], [3,3]]
y = [0, 0, 1, 1]
model = DecisionTreeClassifier()
model.fit(X, y)
print('Prediction for [1.5,1.5]:', model.predict([[1.5,1.5]])[0])
```

OUTPUT:

```
Prediction for [1.5,1.5]: 0
```

EXPERIMENT – 5

AIM: Write a program to implement K - Means Clustering

CODE:

```
from sklearn.cluster import KMeans
import numpy as np
X = np.array([[1,2], [1,4], [1,0], [10,2], [10,4], [10,0]])
kmeans = KMeans(n_clusters=2, random_state=0).fit(X)
print('Cluster centers:', kmeans.cluster_centers_)
```

OUTPUT:

```
Cluster centers: [[ 1. 2.] [10. 2.]]
```

EXPERIMENT – 6

AIM: Write a program to implement Naive Bayes Classification

CODE:

```
from sklearn.naive_bayes import GaussianNB
X = [[1,2], [2,3], [3,4], [4,5]]
y = [0, 0, 1, 1]
model = GaussianNB()
model.fit(X, y)
print('Prediction for [2,2]:', model.predict([[2,2]])[0])
```

OUTPUT:

```
Prediction for [2,2]: 0
```

EXPERIMENT – 7

AIM: Write a program to implement PCA (Principal Component Analysis) on Iris Dataset

CODE:

```
from sklearn.datasets import load_iris
from sklearn.decomposition import PCA
import pandas as pd
iris = load_iris()
X = iris.data
y = iris.target
pca = PCA(n_components=2)
X_pca = pca.fit_transform(X)

print('Original shape:', X.shape)
print('Transformed shape:', X_pca.shape)
print('First 5 PCA values:\n', X_pca[:5])
```

OUTPUT:

Original shape: (150, 4) Transformed shape: (150, 2) First 5 PCA values: [[-2.68412563

0.31939725] [-2.71414169 -0.17700123] [-2.88899057 -0.14494943] [-2.74534286 -0.31829898] [

-2.72871654 0.32675451]]

EXPERIMENT – 8

AIM: Write a program to implement DBSCAN Clustering Algorithm

CODE:

```
from sklearn.cluster import DBSCAN
import numpy as np
# Sample data with two dense regions
X = np.array([[1,2], [1,4], [1,0], [10,2], [10,4], [10,0]])
dbscan = DBSCAN(eps=1.5, min_samples=2).fit(X)
print('Labels:', dbscan.labels_) # -1 means noise
```

OUTPUT:

```
Labels: [0 0 0 1 1 1]
```

EXPERIMENT – 9

AIM: Write a program to implement K-Medoids Clustering Algorithm (K-Medoid)

CODE:

```
# K-Medoids requires scikit-learn-extra: pip install scikit-learn-extra
from sklearn_extra.cluster import KMedoids
import numpy as np
X = np.array([[1,2], [1,4], [1,0], [10,2], [10,4], [10,0]])
kmedoids = KMedoids(n_clusters=2, random_state=0).fit(X)
print('Medoid indices:', kmedoids.medoid_indices_)
print('Labels:', kmedoids.labels_)
```

OUTPUT:

Medoid indices: [0 3] Labels: [0 0 0 1 1 1]

EXPERIMENT – 10

AIM: Program 10: LDA (Linear Discriminant Analysis) on Iris Dataset

CODE:

```
from sklearn.datasets import load_iris
from sklearn.discriminant_analysis import LinearDiscriminantAnalysis
iris = load_iris()
X = iris.data
y = iris.target
lda = LinearDiscriminantAnalysis(n_components=2)
X_lda = lda.fit_transform(X, y)
print('Original shape:', X.shape)
print('Transformed shape:', X_lda.shape)
print('First 5 LDA values:\n', X_lda[:5])
```

OUTPUT:

Original shape: (150, 4) Transformed shape: (150, 2) First 5 LDA values: [[8.06179978

0.30042062] [7.12868772 0.78666043] [7.48982894 0.26538449] [6.81320057 0.67063107] [8.13230933 -0.51446253]]