

Exp-2

AIM

To implement a classifier using an open-source dataset.

OBJECTIVE.

- To understand and implement a supervised machine learning classifier.
- To train and test the classifier on an open source data (Iris.).
- To evaluate the classifier's performance using accuracy metrics.

PSEUDOCODE

```
IMPORT PANDAS AS PD  
FROM sklearn.model-selection IMPORT train-test-split.  
FROM sklearn.linear-model IMPORT LogisticRegression.  
FROM sklearn.metrics IMPORT accuracy-score.
```

```
DATA = load-iris-dataset()
```

```
X = data.features
```

```
Y = data.labels
```

```
X_train, X_test, Y_train, Y_test = train-test-split(X, Y,  
                                                    test_size=0.2)
```

```
model = LogisticRegression()
```

```
model.fit(X_train, Y_train)
```

```
Y_pred = model.predict(X_test)
```

```
accuracy = accuracy-score(Y_test, Y_pred)
```

```
PRINT("ACCURACY", accuracy)
```

OBSERVATION

DATASET: IRIS Dataset (150 Samples, 4 features, 3 classes)

CLASSIFIER: Logistic Regression

Test/Train Split: 80% - Training, 20% - Testing

ACCURACY :

Classification Report:

	Precision	recall	f1-score	support
Setosa	1.00	1.00	1.00	19
Versicolour	1.00	0.62	0.76	13
virginica	0.72	1.00	0.84	13
accuracy			0.89	45
macro avg	0.91	0.87	0.87	45
weighted avg	0.92	0.89	0.88	45

RESULT

A Classifier was successfully implemented using the open-source Iris dataset. The model achieved high accuracy indicating effective classification of flower species.

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LAB_2

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```
# -----  
# LAB 2  
# -----  
from sklearn.datasets import load_iris  
from sklearn.model_selection import train_test_split  
from sklearn.preprocessing import StandardScaler  
from sklearn.neighbors import KNeighborsClassifier  
from sklearn.metrics import classification_report, accuracy_score  
  
# -----  
# Load data  
# -----  
iris = load_iris()  
X = iris.data  
y = iris.target  
  
# -----  
# Split  
# -----  
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42)  
  
# -----  
# Scale  
# -----  
scaler = StandardScaler()  
X_train_scaled = scaler.fit_transform(X_train)  
X_test_scaled = scaler.transform(X_test)  
  
# -----  
# KNN Model  
# -----
```

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```
# Split
# -----
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42)

# -----
# Scale
# -----
scaler = StandardScaler()
X_train_scaled = scaler.fit_transform(X_train)
X_test_scaled = scaler.transform(X_test)

# -----
# KNN Model
# -----
knn = KNeighborsClassifier(n_neighbors=1)
knn.fit(X_train_scaled, y_train)

# -----
# Predict
# -----
y_pred = knn.predict(X_test_scaled)

# -----
# Results
# -----
print("Classification Report:\n", classification_report(y_test, y_pred))
print("Accuracy Score:", accuracy_score(y_test, y_pred))
```

[1]

Python

... Classification Report:

	precision	recall	f1-score	support
0	1.00	1.00	1.00	19
1	1.00	0.92	0.96	13

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```
# RESULTS
# -----
print("Classification Report:\n", classification_report(y_test, y_pred))
print("Accuracy Score:", accuracy_score(y_test, y_pred))
```

[1]

Python

Classification Report:

	precision	recall	f1-score	support
0	1.00	1.00	1.00	19
1	1.00	0.92	0.96	13
2	0.93	1.00	0.96	13
accuracy			0.98	45
macro avg	0.98	0.97	0.97	45
weighted avg	0.98	0.98	0.98	45

Accuracy Score: 0.9777777777777777

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