03-02-2026

MLA0202

V. Rakshana
192425137

Set 7-7

MODEL PRACTICAL

1) **Aim:-** To study whether Breast Cancer classification problem can solve using Naive Bayes classifier, analyze dataset, process it & evaluates its performance using a Confusion matrix.

Algorithm:-

→ Load Breast Cancer dataset

→ The dataset is split into training & testing sets & a Naive Bayes Classifier using training data.

→ The model evaluate Confusion matrix and accuracy score.

Program:-

```
impost pandas as pd
from sklearn.datasets impost load_breast_cancer.
data = load_breast_cancer()
print ("First five rows: \n", df.head())
print ("\n Null values: \n", df.isnull().sum())
df.fillna(df.mode().iloc[0], inplace = True)
x = df.drop ('target', axis=1)
y = df['target']
print ("\n Confusion Matrix: \n", Confusion_matrix (y_test,
                                                    y_pred))
print ("Accuracy", accuracy_score (ytest, y_pred)
```

Output:-

Accuracy ≈ 93-96 %

Result:-

The program executed successfully.

2)

Aim:- To find most specific hypothesis using Find s-algorithm using given dataset.

Algorithm:-

→ Initialize hypothesis with positive example.

⇒ Generalize only when attribute value differ.

⇒ Ignore negative examples

⇒ Output final hypothesis.

Program:-

```
data = [ ['Big', 'Red', 'Circle', 'No'], ['Small', 'Red', 'Triangle',
                                          'Yes']]
hypothesis = None
for row in data:
    if row[-1] == 'Yes':
        if hypothesis is None:
            hypothesis = row[:-1]
        else:
            for i in range(len(hypothesis)):
                if hypothesis[i] != row[i]:
                    hypothesis[i] = '?'
print("Most specific hypothesis:", hypothesis)
```

Output:-

Most specific hypothesis : ['Small', '?', 'Circle']

Result:-

The program executed Successfully.

3) **Aim :-** To implement polynomial Regression & analyze its performance.

Program :-

```
import numpy as np
from sklearn.preprocessing import polynomialFeatures
X = np.array([[1,2,3,4,5]]).reshape(-1,1)
y = np.array([1,4,9,16,25])
poly = PolynomialFeatures(degree=2)
X_poly = poly.fit_transform(X)
model = LinearRegression()
model.fit(X_poly, y)
y_pred = model.predict(X_poly)
print("R2 Score:", r2_score(y, y_pred))
```

Output :-

```
R2 Score : 1.0
```

Result :-

The program has executed successfully.

Algorithm :-

⇒ Data preparation & Transformation.

⇒ A Linear Regression model is trained using transformed polynomial features & target values.

⇒ The trained model predicts outputs for given data & its performance is evaluated using R2 score, indicating how well model fits data.

4) Aim:- To implement the kNN classification algorithm using a sample dataset.

Algorithm:-

⇒ The Iris dataset is loaded & divided into training & testing sets to evaluate performance.
⇒ A kNN classifier is created with K=5 neighbors & trained using training data.
⇒ The trained model predicts class labels for test data, performance is measured using accuracy score.

Program:-

```
from sklearn.datasets import load_iris
from sklearn.metrics import accuracy_scores.
iris = load_iris()
x = iris.data
y = iris.target
x_train, x_test, y_train, y_test = train_test_split(x, y,
test_size=0.3, random_state=42)
knn = kNeighborsClassifier(n_neighbors=5)
knn.fit(x_train, y_train)
y_pred = knn.predict(x_test)
print("kNN Accuracy", accuracy_score(y_test,
                                      y_pred))
```

Output:-

kNN Accuracy: ~97%

Result:-

The program has been executed successfully.