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Department Of Information Technology

A.Y. 2025-2026
Class: BE-ITA/B, Semester: VII
Subject: Data Science Lab

Experiment – 9

1. **Aim:** To Evaluate classification algorithms.
2. **Objectives:** Students should be define and apply metrics to measure the performance of various learning algorithms
3. **Prerequisite:** Python basics
4. **Pre-Experiment Exercise:**
Theory:

Compare

- i. Popular Classification Algorithms:
- ii. Logistic Regression
- iii. Naive Bayes
- iv. K-Nearest Neighbors
- v. Decision Tree
- vi. Support Vector Machines

5. **Laboratory Exercise**

Procedure:

- i. Compare above listed classifier algorithms and apply metrics to measure the performances of each.

6. **Post-Experiments Exercise:**

A. Extended Theory:

- i. Explain various performance metrices.

7. **Conclusion:**

- i. Write what was performed in the program (s) .
- ii. What is the significance of program and what Objective is achieved?

8. **References:**

- a. https://scikit-learn.org/stable/modules/cross_validation.html

Lab Exercise:

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# Import libraries
import pandas as pd
import numpy as np
from sklearn.datasets import load_breast_cancer
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import accuracy_score, precision_score, recall_score, f1_score, classification_report
from sklearn.linear_model import LogisticRegression
from sklearn.naive_bayes import GaussianNB
from sklearn.neighbors import KNeighborsClassifier
from sklearn.tree import DecisionTreeClassifier
from sklearn.svm import SVC
from sklearn.ensemble import RandomForestClassifier

# Load dataset
data = load_breast_cancer()
X = pd.DataFrame(data.data, columns=data.feature_names)
y = pd.Series(data.target) # 0 = malignant, 1 = benign

print("Dataset loaded successfully!")
print("Shape:", X.shape)
print(X.head())

# Train-test split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.25, random_state=42)

# Feature scaling
scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)

# Initialize classifiers
models = [
    "Logistic Regression": LogisticRegression(max_iter=1000),
    "Naive Bayes": GaussianNB(),
    "K-Nearest Neighbors": KNeighborsClassifier(n_neighbors=5),
    "Decision Tree": DecisionTreeClassifier(max_depth=5, random_state=42),
    "Support Vector Machine": SVC(kernel='rbf', C=2, gamma='auto', random_state=42),
    "Random Forest": RandomForestClassifier(n_estimators=200, max_depth=8, random_state=42)
]

# Train, predict and evaluate
results = []

for name, model in models.items():
    model.fit(X_train, y_train)
    y_pred = model.predict(X_test)

    results.append({
        "Model": name,
        "Accuracy": round(accuracy_score(y_test, y_pred), 4),
        "Precision": round(precision_score(y_test, y_pred), 4),
        "Recall": round(recall_score(y_test, y_pred), 4),
        "F1-Score": round(f1_score(y_test, y_pred), 4)
    })

# Display results
results_df = pd.DataFrame(results).sort_values(by='Accuracy', ascending=False).reset_index(drop=True)
print("\nPerformance Comparison of Classifiers:\n")
print(results_df)

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Classification Report:
      precision    recall   f1-score   support
0       0.96     0.98     0.97      54
1       0.99     0.98     0.98      89

   accuracy      macro avg     weighted avg
accuracy           0.98     0.98     0.98
macro avg           0.98     0.98     0.98
weighted avg         0.98     0.98     0.98

# Detailed report for the best model
best_model_name = results_df.iloc[0]['Model']
print(f"\nBest Model: {best_model_name}")

best_model = models[best_model_name]
y_best_pred = best_model.predict(X_test)
print("\nClassification Report:\n")
print(classification_report(y_test, y_best_pred))

Dataset loaded successfully!
Shape: (569, 30)
   mean radius  mean texture  mean perimeter  mean area  mean smoothness \
0       17.99      10.38       122.80     1001.0      0.11840
1       20.57      17.77       132.90     1326.0      0.08474
2       19.69      21.25       130.00     1203.0      0.10960
3       11.42      20.38       77.58      386.1      0.14250
4       20.29      14.34       135.10     1297.0      0.10030

   mean compactness  mean concavity  mean concave points  mean symmetry \
0        0.27760      0.3001       0.14710      0.2419
1        0.07864      0.0869       0.07017      0.1812
2        0.15990      0.1974       0.12790      0.2069
3        0.28390      0.2414       0.10520      0.2597
4        0.13280      0.1980       0.10430      0.1809

   mean fractal dimension ...  worst radius  worst texture  worst perimeter \
0        0.07871    ...       25.38       17.33      184.60
1        0.05667    ...       24.99       23.41      158.80
2        0.05999    ...       23.57       25.53      152.50
3        0.09744    ...       14.91       26.50      98.87
4        0.05883    ...       22.54       16.67      152.20

   worst area  worst smoothness  worst compactness  worst concavity \
0      2019.0       0.1622       0.6656       0.7119
1      1956.0       0.1238       0.1866       0.2416
2      1709.0       0.1444       0.4245       0.4504
3       567.7       0.2098       0.8663       0.6869
4      1575.0       0.1374       0.2050       0.4000

   worst concave points  worst symmetry  worst fractal dimension
0            0.2654       0.4601       0.11890
1            0.1860       0.2750       0.08902
2            0.2430       0.3613       0.08758
3            0.2575       0.6638       0.17300
4            0.1625       0.2364       0.07678

[5 rows x 30 columns]

Performance Comparison of Classifiers:

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	Model	Accuracy	Precision	Recall	F1-Score
0	Logistic Regression	0.979	0.9886	0.9775	0.9831
1	Support Vector Machine	0.979	0.9778	0.9888	0.9832
2	Random Forest	0.965	0.9667	0.9775	0.9721
3	Decision Tree	0.958	0.9770	0.9551	0.9659
4	K-Nearest Neighbors	0.958	0.9663	0.9663	0.9663
5	Naive Bayes	0.951	0.9659	0.9551	0.9605

Best Model: Logistic Regression