

**St. Francis Institute of Technology, Mumbai-400 103**  
**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 metrics.

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](https://scikit-learn.org/stable/modules/cross_validation.html)

## Lab Exercise:

```
# 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)
```

## Classification Report:

	precision	recall	f1-score	support
0	0.96	0.98	0.97	54
1	0.99	0.98	0.98	89
accuracy			0.98	143
macro avg	0.98	0.98	0.98	143
weighted avg	0.98	0.98	0.98	143

```
# 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:

	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