

Exp-3

Aim

To analyze the performance of a classifier using statistical evaluation metrics.

OBJECTIVE

- To compute statistical performance measurement such as precision, recall, F1-score, and confusion matrix.
- To interpret the classifier's behavior on test data.

PSEUDOCODE.

```
#USE THE TRAINED MODEL FROM EXP-2
```

```
y_pred = model.predict(x_test)
```

```
from sklearn.metrics import confusion_matrix,  
classification_report.
```

```
cm = confusion_matrix(y_test, y_pred)
```

```
print("Confusion Matrix : \n", cm)
```

```
report = classification_report(y_test, y_pred)
```

```
print("Classification Report \n", report).
```

OBSERVATION

Accuracy

Precision

Recall

F1-Score

Classification Report -- Logistic Regression --

	precision	recall	f1-score	support
setosa	1.00	1.00	1.00	19
versicolor	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

Classification	Report --	Linear Regression	(Round
	precision	recall	supp
setosa	0.92	1.00	0.84 13
versicolor	1.00	1.00	0.76 13
virginica	1.00	0.62	1.00 19
			n = 83 45

accuracy		0.89	0.83	45
macro avg	0.92	0.87	0.87	45
weighted avg	0.91	0.87	0.87	45

RESULT

The classifier was evaluated using statistical parameters. It showed excellent precision, recall, F1-score, confirming its reliability for multiclass classification tasks.

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LAB 3

```
from sklearn.datasets import load_breast_cancer
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.neighbors import KNeighborsClassifier
from sklearn.svm import SVC
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import classification_report, accuracy_score

# -----
# Load dataset
# -----
data = load_breast_cancer()
X, y = data.data, data.target

# -----
# Split into train and test
# -----
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42)

# -----
# Scale features (important for KNN and SVM)
# -----
scaler = StandardScaler()
X_train_scaled = scaler.fit_transform(X_train)
X_test_scaled = scaler.transform(X_test)

# -----
# KNN Classifier
# -----
knn = KNeighborsClassifier(n_neighbors=5)
knn.fit(X_train_scaled, y_train)
knn_preds = knn.predict(X_test_scaled)
```

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x_train, x_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42)

```
# -----
# Scale features (important for KNN and SVM)
# -----
scaler = StandardScaler()
X_train_scaled = scaler.fit_transform(X_train)
X_test_scaled = scaler.transform(X_test)

# -----
# KNN Classifier
# -----
knn = KNeighborsClassifier(n_neighbors=5)
knn.fit(X_train_scaled, y_train)
knn_preds = knn.predict(X_test_scaled)

# -----
# SVM Classifier
# -----
svm = SVC(kernel='rbf') # or 'linear'
svm.fit(X_train_scaled, y_train)
svm_preds = svm.predict(X_test_scaled)

# -----
# Decision Tree Classifier
# -----

dt = DecisionTreeClassifier(random_state=42)
dt.fit(X_train, y_train) # no scaling needed for trees
dt_preds = dt.predict(X_test)

# -----
# Results
# -----

print("\n--- KNN Classifier ---")
```

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--- KNN Classifier ---
Accuracy: 0.9590643274853801
Classification Report:

	precision	recall	f1-score	support
0	0.95	0.94	0.94	63
1	0.96	0.97	0.97	108
accuracy			0.96	171
macro avg	0.96	0.95	0.96	171
weighted avg	0.96	0.96	0.96	171

--- SVM Classifier ---
Accuracy: 0.9766881871345029
Classification Report:

	precision	recall	f1-score	support
0	0.97	0.97	0.97	63
1	0.98	0.98	0.98	108
accuracy			0.98	171
macro avg	0.97	0.97	0.97	171
weighted avg	0.98	0.98	0.98	171

...

accuracy			0.94	171
macro avg	0.93	0.94	0.94	171
weighted avg	0.94	0.94	0.94	171

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Upcoming Earnings

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