

ML Lab 6

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Procedure:

Load and read the data: The dataset is loaded using pandas for easy data manipulation.

Split the data: The data is split into training and testing sets, ensuring that the model is trained on one portion of the data and evaluated on another.

Normalize the data: StandardScaler is used to normalize the feature values, bringing them to a common scale with mean 0 and standard deviation 1.

Train models: Three machine learning models—SVM (Support Vector Machine), KNN (K-Nearest Neighbors), and Naive Bayes—are trained on the normalized training data.

Evaluate models: The models are evaluated on the test data using several performance metrics:

Accuracy

F1 Score

Precision

Recall

Compare models: A plot is created to visually compare the performance of the models across the evaluation metrics (Accuracy, F1 Score, Precision, and Recall).

```
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.preprocessing import StandardScaler as ss
from sklearn.metrics import accuracy_score, classification_report, precision_score, f1_score, recall_score
from sklearn.model_selection import train_test_split as tt
from sklearn.svm import SVC

from sklearn.neighbors import KNeighborsClassifier as knn
from sklearn.naive_bayes import GaussianNB as gnb

data = pd.read_csv('all_data_3_.csv')
x = data.drop('class', axis=1)
y = data['class']
sc = ss()
x_scaled = sc.fit_transform(x)
x = pd.DataFrame(x_scaled, columns=x.columns)
x_train, x_test, y_train, y_test = tt(x, y, test_size=0.2, random_state=42)
model = SVC(kernel='linear')
model.fit(x_train, y_train)
y_pred = model.predict(x_test)

print("SVM :")
svm_accuracy = accuracy_score(y_test, y_pred)
svm_precision = precision_score(y_test, y_pred, average='weighted')
svm_recall = recall_score(y_test, y_pred, average='weighted')
svm_f1 = f1_score(y_test, y_pred, average='weighted')
print(classification_report(y_test, y_pred))

print("K nearest neighbors:")
knn=knn(n_neighbors=5)
knn.fit(x_train,y_train)
y_pred=knn.predict(x_test)
knn_accuracy = accuracy_score(y_test, y_pred)
knn_precision = precision_score(y_test, y_pred, average='weighted')
knn_recall = recall_score(y_test, y_pred, average='weighted')
knn_f1 = f1_score(y_test, y_pred, average='weighted')
print(classification_report(y_test,y_pred))
```

```

print("Navie Bayes:")
nb=gnb()
nb.fit(x_train,y_train)
y_pred=nb.predict(x_test)
nb_accuracy = accuracy_score(y_test, y_pred)
nb_precision = precision_score(y_test, y_pred,average='weighted')
nb_recall = recall_score(y_test, y_pred,average='weighted')
nb_f1 = f1_score(y_test, y_pred,average='weighted')
print(classification_report(y_test,y_pred))

# Plotting the accuracy comparison
models = ['SVM', 'KNN', 'Naive Bayes']
accuracies = [svm_accuracy, knn_accuracy, nb_accuracy]
plt.figure(figsize=(8, 5))
plt.bar(models, accuracies, color=['magenta', 'turquoise', 'lime'],edgecolor='black', linewidth=1)
plt.xlabel("Classifiers")
plt.ylabel("Accuracy Score")
plt.ylim(0.6, 1)
plt.title("Performance Comparison of Classifiers")
plt.show()

#precision comparison
models = ['SVM', 'KNN', 'Naive Bayes']
accuracies = [svm_precision, knn_precision, nb_precision]
plt.figure(figsize=(8, 5))
plt.bar(models, accuracies, color=['magenta', 'turquoise', 'lime'],edgecolor='black', linewidth=1)
plt.xlabel("Classifiers")
plt.ylabel("Precision Score")
plt.ylim(0.6, 1)
plt.title("Performance Comparison of Classifiers")
plt.show()

```

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#f1 comparison
models = ['SVM', 'KNN', 'Naive Bayes']
accuracies = [svm_f1, knn_f1, nb_f1]
plt.figure(figsize=(8, 5))
plt.bar(models, accuracies, color=['magenta', 'turquoise', 'lime'],edgecolor='black', linewidth=1)
plt.xlabel("Classifiers")
plt.ylabel("F1 Score")
plt.ylim(0.6, 1)
plt.title("Performance Comparison of Classifiers")
plt.show()

#recall comparison
models = ['SVM', 'KNN', 'Naive Bayes']
accuracies = [svm_recall, knn_recall, nb_recall]
plt.figure(figsize=(8, 5))
plt.bar(models, accuracies, color=['magenta', 'turquoise', 'lime'],edgecolor='black', linewidth=1)
plt.xlabel("Classifiers")
plt.ylabel("Recall Score")
plt.ylim(0.6, 1)
plt.title("Performance Comparison of Classifiers")
plt.show()

```

Output:

SVM :

	precision	recall	f1-score	support
bruteForce	1.00	1.00	1.00	37
httpFlood	1.00	0.98	0.99	98
icmp-echo	0.99	1.00	1.00	130
normal	1.00	1.00	1.00	134
slowloris	1.00	1.00	1.00	160
slowpost	1.00	1.00	1.00	99
tcp-syn	0.99	0.99	0.99	190
udp-flood	1.00	1.00	1.00	152
accuracy			1.00	1000
macro avg	1.00	1.00	1.00	1000
weighted avg	1.00	1.00	1.00	1000

K nearest neighbors:

	precision	recall	f1-score	support
bruteForce	1.00	1.00	1.00	37
httpFlood	1.00	0.97	0.98	98
icmp-echo	0.98	1.00	0.99	130
normal	1.00	0.99	0.99	134
slowloris	0.98	1.00	0.99	160
slowpost	1.00	1.00	1.00	99
tcp-syn	1.00	1.00	1.00	190
udp-flood	1.00	0.99	0.99	152
accuracy			0.99	1000
macro avg	0.99	0.99	0.99	1000
weighted avg	0.99	0.99	0.99	1000

Navie Bayes:

	precision	recall	f1-score	support
bruteForce	0.35	1.00	0.52	37
httpFlood	1.00	0.87	0.93	98
icmp-echo	0.61	0.74	0.67	130
normal	0.88	0.79	0.83	134
slowloris	0.62	1.00	0.77	160
slowpost	0.65	1.00	0.79	99
tcp-syn	0.94	0.43	0.59	190
udp-flood	0.94	0.20	0.34	152
accuracy			0.70	1000
macro avg	0.75	0.75	0.68	1000
weighted avg	0.79	0.70	0.67	1000



