

LAB Assignment No 4

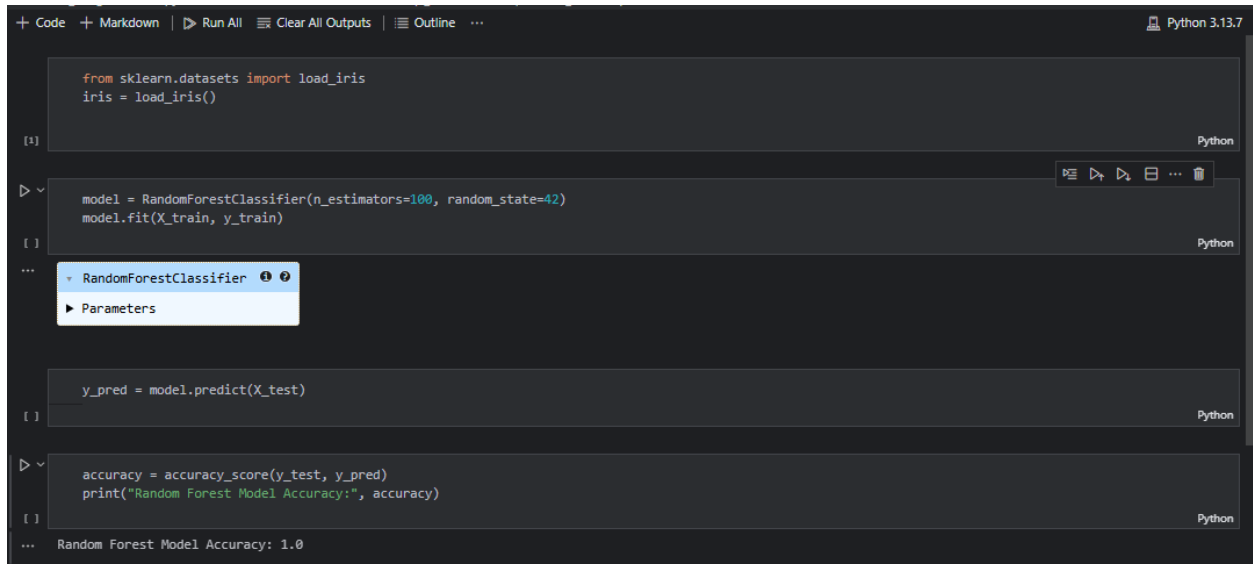
Topic: Random Forest and Support Vector Machine Classifier

Question 1

Classify flower species using Random Forest.

Task:

1. Load the *Iris dataset* from `sklearn.datasets`.
2. Split into training (70%) and testing (30%) sets.
3. Train a **Random Forest Classifier**.
4. Predict flower species on the test set.
5. Calculate and print **model accuracy**.



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from sklearn.datasets import load_iris
iris = load_iris()

(1) Python

▶ ▼
model = RandomForestClassifier(n_estimators=100, random_state=42)
model.fit(X_train, y_train)

[] Python

...
▼ RandomForestClassifier ⓘ ⓘ
  ▶ Parameters

y_pred = model.predict(X_test)

[] Python

▶ ▼
accuracy = accuracy_score(y_test, y_pred)
print("Random Forest Model Accuracy:", accuracy)

[] Python

... Random Forest Model Accuracy: 1.0
```

Question 2

Use SVM on Breast Cancer Dataset and Classify tumors as malignant or benign.

Task:

1. Load the *Breast Cancer* dataset using `sklearn.datasets.load_breast_cancer`.
2. Train an **SVM classifier** (use `SVC(kernel='linear')`).
3. Evaluate the model using **accuracy** and **confusion matrix**.

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D v
print(data.feature_names[:5])
print(data.data.shape)
from sklearn.datasets import load_breast_cancer
from sklearn.model_selection import train_test_split
from sklearn.svm import SVC

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

[1] Python

... ['mean radius' 'mean texture' 'mean perimeter' 'mean area'
      'mean smoothness']
(569, 30)

from sklearn.svm import SVC
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42)
svm_model = SVC(kernel='linear')
svm_model.fit(X_train, y_train)

[2] Python

... SVC 0.0
Parameters

D v
from sklearn.metrics import accuracy_score, confusion_matrix
y_pred = svm_model.predict(X_test)
accuracy = accuracy_score(y_test, y_pred)
print("Model Accuracy:", accuracy)
cm = confusion_matrix(y_test, y_pred)
print("Confusion Matrix:")
print(cm)

[3] Python

... Model Accuracy: 0.9649122807017544
Confusion Matrix:
[[ 59   4]
 [  2 106]]
```

Question 3

Use Random Forest on CSV Dataset (Custom) : Predict student pass/fail based on study hours and scores.

Task:

1. Load a CSV file (e.g., students.csv) with columns: study_hours, attendance, marks, result.
2. Train a **Random Forest Classifier** to predict result (Pass/Fail).
3. Display **accuracy score** and **feature importance**.

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import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score

[1] Python

df = pd.read_csv("student_lab4 q3.csv")
X = df[['study_hours', 'attendance', 'marks']]
y = df['result']
y = y.map({'Pass': 1, 'Fail': 0})

[2] Python

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42)

[3] Python

rf_model = RandomForestClassifier(random_state=42)
rf_model.fit(X_train, y_train)

...
RandomForestClassifier
Parameters

[4] Python

y_pred = rf_model.predict(X_test)
print("Predicted Results (first 10):", y_pred[:10])
print("Actual Results (first 10):", y_test[:10].values)

[5] Python

... Predicted Results (first 10): [0 1 1 1 1 0 1 1 1 1]
Actual Results (first 10): [0 1 1 1 1 0 1 1 1 1]

[5] Python

y_pred = rf_model.predict(X_test)
print("Predicted Results (first 10):", y_pred[:10])
print("Actual Results (first 10):", y_test[:10].values)

... Predicted Results (first 10): [0 1 1 1 1 0 1 1 1 1]
Actual Results (first 10): [0 1 1 1 1 0 1 1 1 1]

[6] Python

y_pred = rf_model.predict(X_test)

[6] Python

accuracy = accuracy_score(y_test, y_pred)
print("Accuracy:", accuracy)

[7] Python

... Accuracy: 1.0

[7] Python

importances = rf_model.feature_importances_
feature_names = X.columns
for name, importance in zip(feature_names, importances):
    print(f"{name}: {importance:.4f}")

[8] Python

... study_hours: 0.3571
attendance: 0.3564
marks: 0.2865
```

Question 4

Use SVM on Digits Dataset and to identify the: Handwritten digit recognition.

Task:

1. Load the *Digits dataset* from `sklearn.datasets.load_digits`.
2. Train an **SVM classifier** with an RBF kernel.
3. Test on unseen data.
4. Print **accuracy** and visualize some **misclassified samples**.

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from sklearn.datasets import load_digits

digits = load_digits()
X = digits.data # features (image pixels)
y = digits.target # labels (0-9 digits)

[1] Python

from sklearn.model_selection import train_test_split
from sklearn.svm import SVC

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42)

svm_model = SVC(kernel='rbf', gamma=0.001) # RBF kernel use kiya
svm_model.fit(X_train, y_train) # Model train hua

[2] Python

SVC
Parameters

y_pred = svm_model.predict(X_test)

[3] Python

from sklearn.metrics import accuracy_score, confusion_matrix
import matplotlib.pyplot as plt
import numpy as np
print("Model Accuracy:", accuracy_score(y_test, y_pred))
misclassified = np.where(y_test != y_pred)[0]
plt.figure(figsize=(10, 4))
for i, index in enumerate(misclassified[:5]):
    plt.subplot(1, 5, i + 1)
    plt.imshow(X_test[index].reshape(8, 8), cmap='gray')
    plt.title(f"True: {y_test[index]}\nPred: {y_pred[index]}")
    plt.axis('off')
plt.show()

[4] Python

Model Accuracy: 0.9987487487487487

True: 7 Pred: 9
True: 3 Pred: 5
True: 9 Pred: 7
True: 3 Pred: 8
True: 9 Pred: 3
```

Question 5:

Compare Random Forest vs SVM on Same Dataset (you can choose any dataset): Compare two models on the same data.

Task:

- Use the *Wine dataset* from `sklearn.datasets.load_wine`.
- Train both:

`RandomForestClassifier(n_estimators=100)`

- SVC(kernel='rbf')
- Print accuracy of both models.
- Conclude which performs better.

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import pandas as pd
from sklearn.datasets import load_wine
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.svm import SVC
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import accuracy_score, confusion_matrix

[1] Python

wine = load_wine()
X = wine.data
y = wine.target
target_names = wine.target_names

[2] Python

X_train, X_test, y_train, y_test = train_test_split(
    X, y, test_size=0.3, random_state=42, stratify=y
)

[3] Python

rf_model = RandomForestClassifier(n_estimators=100, random_state=42)
rf_model.fit(X_train, y_train)
y_pred_rf = rf_model.predict(X_test)
acc_rf = accuracy_score(y_test, y_pred_rf)

[4] Python

scaler = StandardScaler()
X_train_s = scaler.fit_transform(X_train)
X_test_s = scaler.transform(X_test)

svm_model = SVC(kernel='rbf', random_state=42)
svm_model.fit(X_train_s, y_train)
y_pred_svm = svm_model.predict(X_test_s)
acc_svm = accuracy_score(y_test, y_pred_svm)

[5] Python
```

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[4] Python

acc_rf = accuracy_score(y_test, y_pred_rf)

[5] Python

scaler = StandardScaler()
X_train_s = scaler.fit_transform(X_train)
X_test_s = scaler.transform(X_test)

svm_model = SVC(kernel='rbf', random_state=42)
svm_model.fit(X_train_s, y_train)
y_pred_svm = svm_model.predict(X_test_s)
acc_svm = accuracy_score(y_test, y_pred_svm)

[6] Python

print("Random Forest Accuracy: {:.4f} ({:.2f}%)".format(acc_rf, acc_rf*100))
print("SVM (RBF) Accuracy : {:.4f} ({:.2f}%)".format(acc_svm, acc_svm*100))

[6] Python

... Random Forest Accuracy: 1.0000 (100.00%)
SVM (RBF) Accuracy : 0.9815 (98.15%)

if acc_rf > acc_svm:
    conclusion = "Random Forest performs better on this split."
elif acc_svm > acc_rf:
    conclusion = "SVM performs better on this split."
else:
    conclusion = "Both models perform equally well."

print("Conclusion:", conclusion)

[7] Python

... Conclusion: Random Forest performs better on this split.
```