

# Exercise 9

Take the iris flower dataset and use different Classifier model such as Random Forest, Decision Tree, SVM and logistic Regression and use cross\_val\_score method to measure the performance of the best classifier

```
In [102... import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sn
from sklearn.datasets import load_iris
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.svm import SVC
from sklearn.ensemble import RandomForestClassifier
from sklearn.tree import DecisionTreeClassifier
from sklearn.model_selection import StratifiedKFold
from sklearn.model_selection import cross_val_score
from sklearn.metrics import confusion_matrix
```

```
In [103... data = load_iris()
dir(data)
```

```
Out[103]: ['DESCR',
'data',
'data_module',
'feature_names',
'filename',
'frame',
'target',
'target_names']
```

```
In [104... df = pd.DataFrame(data.data, columns=data.feature_names)
df.head()
```

```
Out[104]:
```

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)
0	5.1	3.5	1.4	0.2
1	4.9	3.0	1.4	0.2
2	4.7	3.2	1.3	0.2
3	4.6	3.1	1.5	0.2
4	5.0	3.6	1.4	0.2

```
In [105... df['target'] = data.target
df['flower name'] = df.target.apply(lambda x: data.target_names[x])
df.head()
```

```
Out[105]:
```

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)	target	flower name
0	5.1	3.5	1.4	0.2	0	setosa
1	4.9	3.0	1.4	0.2	0	setosa
2	4.7	3.2	1.3	0.2	0	setosa
3	4.6	3.1	1.5	0.2	0	setosa
4	5.0	3.6	1.4	0.2	0	setosa

```
In [106... X = df.drop(['target', 'flower name'], axis=1)
y = df['target']
```

```
In [107... X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2)
```

```
In [108... kf = StratifiedKFold()
kf
```

```
Out[108]: StratifiedKFold(n_splits=5, random_state=None, shuffle=False)
```

```
In [109... lr = cross_val_score(LogisticRegression(max_iter=1000), X, y, cv=kf)
lr
```

```
Out[109]: array([0.96666667, 1.          , 0.93333333, 0.96666667, 1.          ])
```

```
In [110... svm = cross_val_score(SVC(kernel="linear"), X, y, cv=kf)
svm
```

```
Out[110]: array([0.96666667, 1.          , 0.96666667, 0.96666667, 1.          ])
```

```
In [111]: dt = cross_val_score(DecisionTreeClassifier(), X, y, cv=kf)
dt
```

```
Out[111]: array([0.96666667, 0.96666667, 0.9          , 0.93333333, 1.          ])
```

```
In [112]: rf = cross_val_score(RandomForestClassifier(n_estimators=40), X, y, cv=kf)
rf
```

```
Out[112]: array([0.96666667, 0.96666667, 0.93333333, 0.93333333, 1.          ])
```

```
In [113]: print(f"Logistic Regression avg score : {np.mean(lr)}")
print(f"SVM avg score : {np.mean(svm)}")
print(f"Decision tree avg score : {np.mean(dt)}")
print(f"Random Forest Classifier avg score : {np.mean(rf)}")
```

Logistic Regression avg score : 0.9733333333333334  
SVM avg score : 0.9800000000000001  
Decision tree avg score : 0.9533333333333334  
Random Forest Classifier avg score : 0.96

## SVM performed the best

```
In [114]: X_test.head()
```

```
Out[114]:
```

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)
54	6.5	2.8	4.6	1.5
27	5.2	3.5	1.5	0.2
119	6.0	2.2	5.0	1.5
110	6.5	3.2	5.1	2.0
99	5.7	2.8	4.1	1.3

```
In [115]: df.iloc[13]
```

```
Out[115]: sepal length (cm)    4.3
sepal width (cm)             3.0
petal length (cm)            1.1
petal width (cm)             0.1
target                       0
flower name                   setosa
Name: 13, dtype: object
```

```
In [116]: model = SVC(kernel="linear")
model.fit(X_train, y_train)
```

```
Out[116]:
```

▼ SVC

SVC(kernel='linear')

```
In [117]: predict = model.predict(X_test)
predict
```

```
Out[117]: array([1, 0, 2, 2, 1, 2, 2, 2, 1, 0, 1, 0, 1, 2, 2, 2, 2, 2, 2, 2, 1,
        2, 0, 1, 0, 2, 2, 1, 0])
```

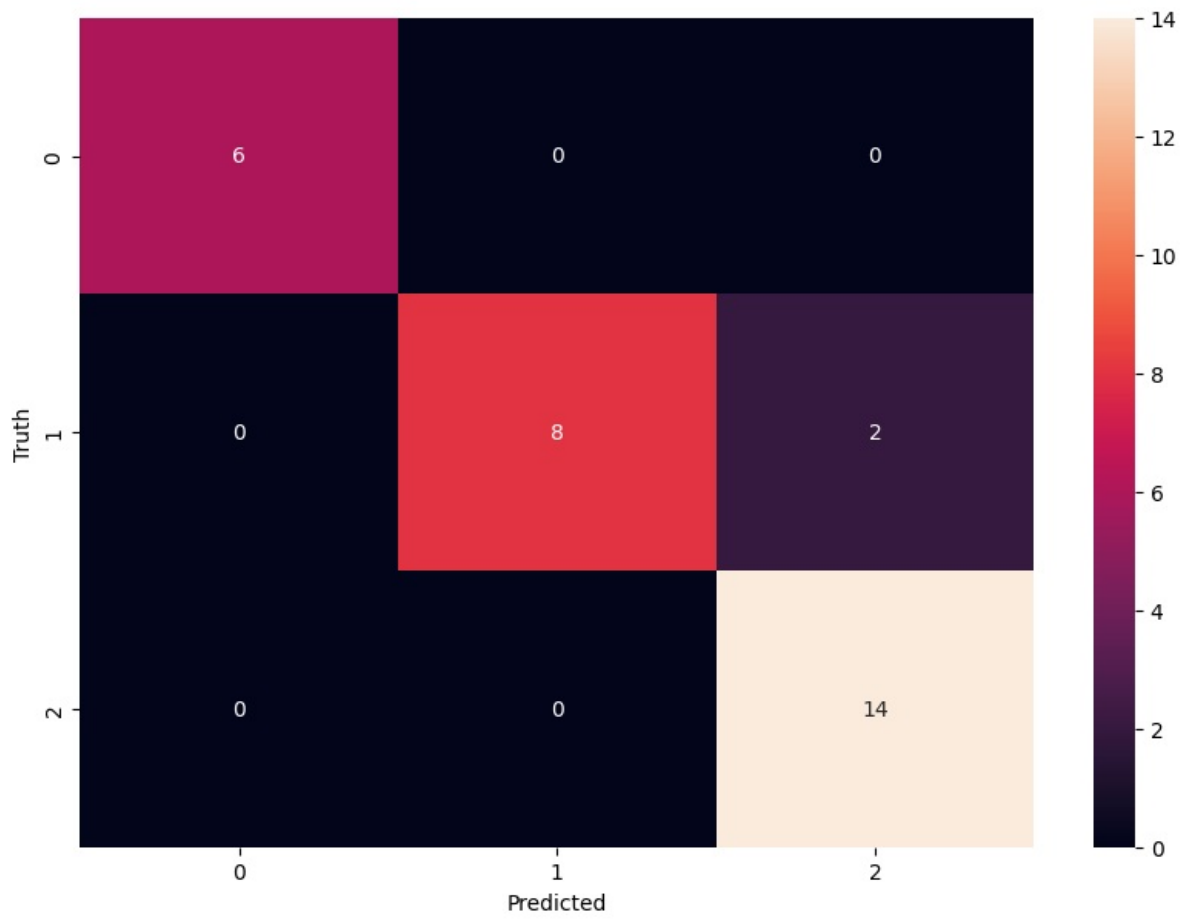
```
In [118]: predict1 = predict.astype(np.int32) # Adjust the data type if necessary
y_test1 = y_test.astype(np.int32)
right = np.sum(predict1 == y_test1)
wrong = np.sum(predict1 != y_test1)

print(f"Num of Correct predictions {right}")
print(f"Num of Wrong predictions {wrong}")
```

Num of Correct predictions 28  
Num of Wrong predictions 2

```
In [120]: cm = confusion_matrix(y_test, predict)
plt.figure(figsize=(10, 7))
sn.heatmap(cm, annot=True)
plt.xlabel("Predicted")
plt.ylabel("Truth")
```

```
Out[120]: Text(95.7222222222221, 0.5, 'Truth')
```



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