

TNT Sessional Exam

Q1: Implement the following algorithms on your dataset 'Iris Form Docker.csv'
With the various performance metrics: (10 marks). Divide dataset into train data and test data in 80% and 20% ratio.

TABLE:

	Sensitivity/ Recall	Specificity	F-Score	Precision	Accuracy
Logistic Regression	100.0	1.0	100.0	100.0	100.0
Decision Tree	100.0	1.0	100.0	100.0	100.0
Gradient Boost	100.0	1.0	100.0	100.0	100.0
Random Forest	100.0	1.0	100.0	100.0	100.0

CODE:

importing necessary libraries pandas for data frame matplotlib for barplot sklearn for regression models

```
In [1]: ▶ import pandas as pd
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn import preprocessing
from sklearn.linear_model import LogisticRegression
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import GradientBoostingClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import confusion_matrix
from sklearn.metrics import accuracy_score
from sklearn.metrics import recall_score
from sklearn.metrics import precision_score
from sklearn.metrics import f1_score

import warnings
warnings.filterwarnings("ignore")
```

importing data frame from csv file using read_csv function

```
In [2]: ▶ df = pd.read_csv('iris-write-from-docker.csv', encoding = "ISO-8859-1")
df
```

Out[2]:

	sepal_length	sepal_width	petal_length	petal_width	class
0	5.1	3.5	1.4	0.2	Iris-setosa
1	4.9	3.0	1.4	0.2	Iris-setosa
2	4.7	3.2	1.3	0.2	Iris-setosa
3	4.6	3.1	1.5	0.2	Iris-setosa
4	5.0	3.6	1.4	0.2	Iris-setosa
...
145	6.7	3.0	5.2	2.3	Iris-virginica
146	6.3	2.5	5.0	1.9	Iris-virginica
147	6.5	3.0	5.2	2.0	Iris-virginica
148	6.2	3.4	5.4	2.3	Iris-virginica
149	5.9	3.0	5.1	1.8	Iris-virginica

150 rows × 5 columns

```
In [3]: le = preprocessing.LabelEncoder()  
df['class'] = le.fit_transform(df['class'])
```

```
In [4]: df['class'].value_counts()
```

```
Out[4]: 0    50  
        1    50  
        2    50  
        Name: class, dtype: int64
```

```
In [5]: corr = df.corr()  
corr
```

Out[5]:

	sepal_length	sepal_width	petal_length	petal_width	class
sepal_length	1.000000	-0.109369	0.871754	0.817954	0.782561
sepal_width	-0.109369	1.000000	-0.420516	-0.356544	-0.419446
petal_length	0.871754	-0.420516	1.000000	0.962757	0.949043
petal_width	0.817954	-0.356544	0.962757	1.000000	0.956464
class	0.782561	-0.419446	0.949043	0.956464	1.000000

Executing Test Train split with class as the target column

```
In [6]: x = df.drop(columns=['class'])
y = df['class']
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size = 0.2)
x_train.shape, y_train.shape, x_test.shape, y_test.shape
```

Out[6]: ((120, 4), (120,), (30, 4), (30,))

Resulting lists

```
In [7]: algos = []
accuracy = []
recall = []
precision = []
f1Score = []
specificity = []
```

Logistic Regression

```
In [8]: ► algo = "Logistic Regression"
model = LogisticRegression()
model.fit(x_train, y_train)
y_pred = model.predict(x_test)
print(algo)
print(confusion_matrix(y_test, y_pred), '\n\n')
acc = accuracy_score(y_test, y_pred) * 100
print('Accuracy:', acc)
rec = recall_score(y_test, y_pred, average='weighted') * 100
print('Recall:', rec)
pre = precision_score(y_test, y_pred, average='weighted') * 100
print('Precision:', pre)
f1s = f1_score(y_test, y_pred, average='weighted') * 100
print('F score:', f1s)

algos.append(algo)
accuracy.append(acc)
precision.append(pre)
f1Score.append(f1s)
```

```
Logistic Regression
[[ 7  0  0]
 [ 0 12  0]
 [ 0  0 11]]
```

```
Accuracy: 100.0
Recall: 100.0
Precision: 100.0
F score: 100.0
```

Decision Tree

```
In [9]: ▶ algo = "Decision Tree"
model = DecisionTreeClassifier()
model.fit(x_train, y_train)
y_pred = model.predict(x_test)
print(algo)
print(confusion_matrix(y_test, y_pred), '\n\n')
acc = accuracy_score(y_test, y_pred) * 100
print('Accuracy:', acc)
rec = recall_score(y_test, y_pred, average='weighted') * 100
print('Recall:', rec)
pre = precision_score(y_test, y_pred, average='weighted') * 100
print('Precision:', pre)
f1s = f1_score(y_test, y_pred, average='weighted') * 100
print('F score:', f1s)

algos.append(algo)
accuracy.append(acc)
recall.append(rec)
precision.append(pre)
f1Score.append(f1s)
```

Decision Tree

```
[[ 7  0  0]
 [ 0 12  0]
 [ 0  0 11]]
```

Accuracy: 100.0
Recall: 100.0
Precision: 100.0
F score: 100.0

Gradient Boost

```
In [10]: ▶ algo = "GradientBoost"
model = GradientBoostingClassifier()
model.fit(x_train, y_train)
y_pred = model.predict(x_test)
print(algo)
print(confusion_matrix(y_test, y_pred), '\n\n')
acc = accuracy_score(y_test, y_pred) * 100
print('Accuracy:', acc)
rec = recall_score(y_test, y_pred, average='weighted') * 100
print('Recall:', rec)
pre = precision_score(y_test, y_pred, average='weighted') * 100
print('Precision:', pre)
f1s = f1_score(y_test, y_pred, average='weighted') * 100
print('F score:', f1s)

algorithms.append(algo)
accuracy.append(acc)
recall.append(rec)
precision.append(pre)
f1Score.append(f1s)
```

```
GradientBoost
[[ 7  0  0]
 [ 0 12  0]
 [ 0  0 11]]
```

```
Accuracy: 100.0
Recall: 100.0
Precision: 100.0
F score: 100.0
```


Random Forest

```
In [11]: ► algo = "Random forest"
model = RandomForestClassifier()
model.fit(x_train, y_train)
y_pred = model.predict(x_test)
print(algo)
print(confusion_matrix(y_test, y_pred), '\n\n')
acc = accuracy_score(y_test, y_pred) * 100
print('Accuracy:', acc)
rec = recall_score(y_test, y_pred, average='weighted') * 100
print('Recall:', rec)
pre = precision_score(y_test, y_pred, average='weighted') * 100
print('Precision:', pre)
f1s = f1_score(y_test, y_pred, average='weighted') * 100
print('F score:', f1s)

algorithms.append(algo)
accuracy.append(acc)
recall.append(rec)
precision.append(pre)
f1Score.append(f1s)
```

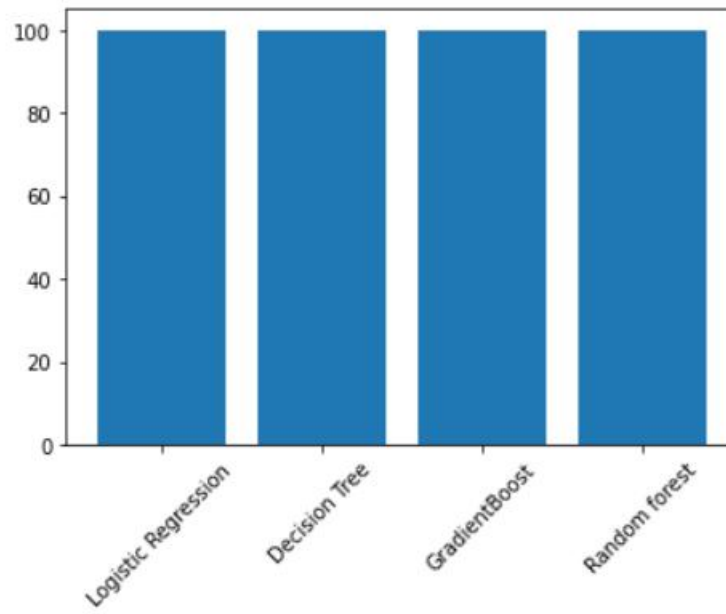
```
Random forest
[[ 7  0  0]
 [ 0 12  0]
 [ 0  0 11]]
```

```
Accuracy: 100.0
Recall: 100.0
Precision: 100.0
F score: 100.0
```

Q2: Draw the graph(barplot) for each algorithm,keep on y-axis the accuracy and thus compare the accuracy in each case.(2 marks)

CODE:

```
In [12]: ▶ plt.bar(algos, accuracy)
plt.xticks(rotation = 45)
plt.show()
```



Q4: Write the python program using loop to print sum of first 10 natural numbers (2 marks).

CODE:

```
In [1]: ▶ sum = 0
        for x in range(1, 11):
            sum = sum + x

        print("Sum of first 10 Natural Numbers is:", sum)
```

OUTPUT:

```
Sum of first 10 Natural Numbers is: 55
```