## **Unit 3.2 Graded Assignment = Implementation of classification algorithm**

## **Instructions:**

Implement a single classification model of your choice and try to achieve at least an 80% F1 score on the wine dataset.

## Submitted by:

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## **Solution:**

1. First we will import all the required libraries.

```
Import all the required Libaries

property pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

from sklearn.metrics import confusion_matrix
from sklearn.metrics import classification_report

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```

2. Import the builtin data from the sklearn library.

```
# Load the wine dataset
from sklearn.datasets import load_wine
data = load_wine()

# Just for Visualization
df = pd. DataFrame(data.data, columns=data.feature_names)
df.head()

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# Just for Visualization
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3. We have split our dataset into test size 0.2 which means 80 % training data and 20% testing data and a random state of 42 so that any bias ness could be removed.

4. Here we have implemented the decision tree with max depth 3.

```
Decision Tree Classifier Implementation

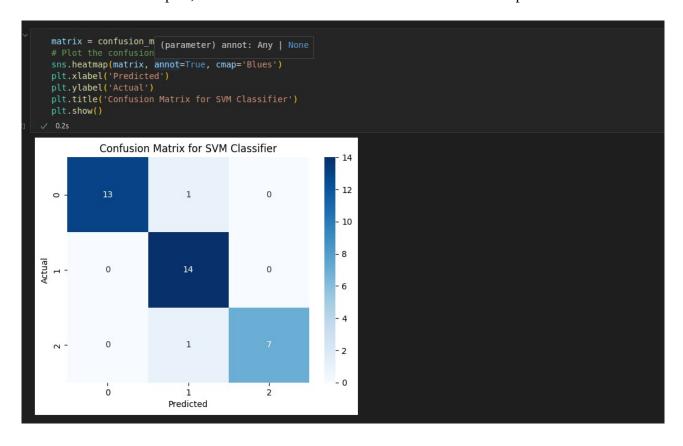
from sklearn.tree import DecisionTreeClassifier, plot_tree

# Create a decision tree classifier with max depth of 3
dt = DecisionTreeClassifier(max_depth=3)

# Train the model using the training data
dt.fit(X_train, y_train)

# Use the trained model to predict the labels of the test data
y_pred = dt.predict(X_test)
```

In our classification report, it can be seen that our f1 score is 0.96 with a complete decision tree.



```
# Print the classification report
report = classification_report(y_test, y_pred, target_names=data.target_names)
   print(report)
                 precision
                                                         support
     class_0
class_1
                                   0.93
                                                0.96
                                                               14
                       0.88
                                                0.93
0.93
                                   1.00
                                                               14
      class 2
                                   0.88
                                                0.94
    accuracy
   macro avg
                       0.96
weighted avg
                       0.95
                                                0.94
```