1. ADABOOST CLASSIFIER

```
CODE:-
```

2019103555

```
import pandas as pd
from sklearn.ensemble import AdaBoostClassifier
from sklearn.model_selection import train_test_split
df = pd.read_csv("Iris.csv")
array = df.values
X = df.iloc[:, :-1]
y = df.iloc[:, -1]
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.35)
seed = 10
num\_trees = 15
print("Using Ada Boost Classifiers, with no. of trees = ", num_trees)
model = AdaBoostClassifier(n_estimators=num_trees, random_state=seed)
model.fit(X_train, y_train)
y_pred = model.predict(X_test)
from sklearn.metrics import confusion_matrix
from sklearn.metrics import classification report
from sklearn.metrics import accuracy_score
print("Accuracy = ", accuracy_score(y_pred, y_test))
y_true = y_test
print("\nConfusion Matrix: \n", confusion_matrix(y_true, y_pred))
matrix = classification_report(y_true, y_pred)
print("\nClassification report : \n", matrix)
```

## **OUTPUT:-**

```
TERMINAL
         JUPYTER PROBLEMS OUTPUT DEBUG CONSOLE
(base) C:\Users\bmspr\OneDrive - Anna University\Documents\SEM5\ML\lab\anaconda\lab13>C:/ProgramData/Anaconda3/python.exe "c:/Users/b
Using Ada Boost Classifiers, with no. of trees = 15
Accuracy = 0.9811320754716981
Confusion Matrix:
 [[16 0 0]
  0 17 0]
 [ 0 1 19]]
Classification report :
                  precision
                               recall f1-score
                                                  support
                      1.00
   Iris-setosa
                                1.00
                                          1.00
Iris-versicolor
                      0.94
                                          0.97
                      1.00
                                          0.97
                                                      20
Iris-virginica
                                0.95
      accuracy
                                          0.98
                                                      53
      macro avg
                      0.98
                                0.98
                                          0.98
                                                      53
  weighted avg
                      0.98
                                0.98
                                          0.98
                                                      53
```

## 2. BAGGING CLASSIFIER

## CODE:-

```
import numpy as np
        import pandas as pd
        from sklearn import metrics
        # classifier
        from sklearn.tree import DecisionTreeClassifier
[166] 🗸 0.3s
                                                                                                                                          Python
        from sklearn.datasets import load_iris
        from sklearn.model_selection import train_test_split
[167] 🗸 0.3s
                                                                                                                                          Python
                                                                                                                         def bagging(X, y):
    n_samples = X.shape[0]
            # doing random sampling with replacement
            indices = np.random.choice(n_samples, size=n_samples, replace=True)
return X.iloc[indices], y.iloc[indices]
[168] 🗸 0.3s
                                                               + Code + Markdown
       class BaggedClassifier:
            def __init__(self, n_estimators, n_neighbours=5):
                self.n_estimators = n_estimators
self.n_neighbours = n_neighbours
                self.classifiers = []
            def fit(self, X, y):
                for _ in range(self.n_estimators):
                    clf = DecisionTreeClassifier(max_depth=4)
                    # getting random sample for the given input
                    X_sample, y_sample = bagging(X, y)
                    # fitting the data on the given input
                    clf.fit(X_sample, y_sample)
                    self.classifiers.append(clf)
            def predict(self, X):
                preds = np.array([clf.predict(X) for clf in self.classifiers])
                preds = np.swapaxes(preds, 0, 1)
                # majority vote
                y_pred = [np.argmax(np.bincount(pred)) for pred in preds]
                return y_pred
[169] 🗸 0.3s
```

```
dataset = load iris()
      df = pd.DataFrame({
       'species': dataset.target
      })
[170] 🗸 0.3s
      print('______')
      print(df.sample(5))
      X = df.iloc[:,:4]
      y = df.iloc[:,-1]
      X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3)
[171] 		 0.3s
         -----DATASET-
      sepal length sepal width petal length petal width species
                     3.0 1.3
                                             0.2
         4.4
                                                        0
   38
   4
                           3.6
                                       1.4
                                                   0.2
                                                             0
              3.6
5.2
3.5
7.2
3.6
6.0
2.9
                                      1.5
                                                  0.2
2.5
1.5
                                                           0
   27
   109
                                       6.1
                                                             2
                                      4.5
   78
      print('\nBuilding random forest classifier')
      clf = BaggedClassifier(n_estimators=50)
      clf.fit(X_train, y_train)
      print('number of classifiers:', clf.n_estimators)
[172] 🗸 0.2s
   Building random forest classifier
    number of classifiers: 50
      y_pred = clf.predict(X_test)
       print()
       print('accuracy:', metrics.accuracy_score(y_test, y_pred))
       print('confusion matrix:\n', metrics.confusion_matrix(y_test, y_pred))
       print("Classification Report: \n", metrics.classification_report(y_test, y_pred))
[173] 🗸 0.1s
    accuracy: 0.9111111111111111
    confusion matrix:
     [[12 0 0]
     [ 0 14 3]
     [ 0 1 15]]
    Classification Report:
                 precision
                           recall f1-score support
                            1.00
             0
                    1.00
                                      1.00
                                                 12
                     0.93
                             0.82
                                      0.87
                                                 17
             1
             2
                    0.83
                             0.94
                                      0.88
                                                 16
                                      0.91
                                                 45
       accuracy
                    0.92
                             0.92
                                       0.92
                                                 45
      macro avg
                    0.92
                             0.91
                                      0.91
    weighted avg
                                                 45
```