IMPORT NEEDED LIBRARIES

```
from sklearn.ensemble import AdaBoostClassifier
from sklearn import datasets
from sklearn.model selection import train test split
from sklearn import model selection
from sklearn import metrics
from sklearn.metrics import precision recall fscore support
import pandas
import time
from sklearn.metrics import confusion matrix
from sklearn.metrics import ConfusionMatrixDisplay
from sklearn.metrics import roc curve
from sklearn.metrics import RocCurveDisplay
from sklearn.metrics import precision recall curve
from sklearn.metrics import PrecisionRecallDisplay
import matplotlib.pyplot as plt
BINARY DATASET
```

IMPORT AND READ BINARY DATASET

```
names =
['pH','TDS','Turbidity','Phospate','Nitrate','Iron','COD(mg/L)','Chlor
ine','Sodium','Class']
bdataframe = pandas.read csv("binary.csv", names=names)
array = bdataframe.values
Xb = array[:,0:9]
Yb = array[:,9]
SPLITTING TRAIN, TEST AND VALIDATION DATA
Xb train, Xb test, Yb train, Yb test = train test split(Xb, Yb,
test size=0.3)
Xb test, Xb val, Yb test, Yb val = train test split(Xb test, Yb test,
test size=0.4)
CREATING AND FITTING ADABOOST CLASSIFIER TO MODEL
bstart = time.time()
abc = AdaBoostClassifier(n estimators=10,
                         learning rate=0.5)
b abc model = abc.fit(Xb train, Yb train)
bend = time.time()
# total time taken
print(f"Runtime of the Adaboost is {bend - bstart}")
Runtime of the Adaboost is 0.054878950119018555
ACCURACY
```

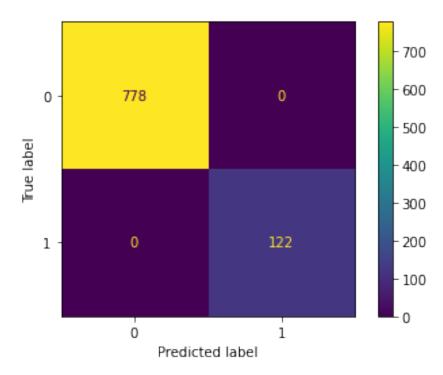
```
Yb_pred = b_abc_model.predict(Xb_test)
print("Accuracy on test data:", metrics.accuracy_score(Yb_test,
Yb_pred))
Yb_val_pred = b_abc_model.predict(Xb_val)
print("Accuracy on validation data:", metrics.accuracy_score(Yb_val,
Yb_val_pred))
```

Accuracy on test data: 1.0

Accuracy on validation data: 1.0

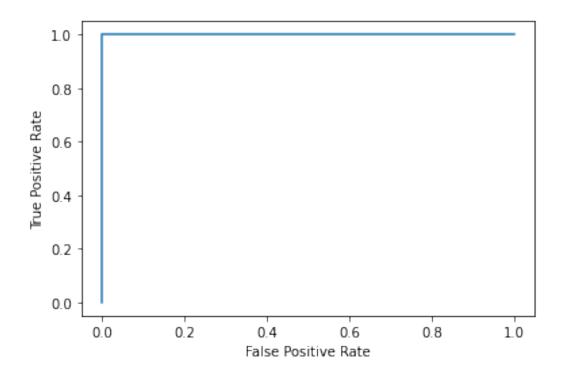
CONFUSION MATRIX

cm = confusion_matrix(Yb_test, Yb_pred)
cm_display = ConfusionMatrixDisplay(cm).plot()



AUC-ROC CURVE

fpr, tpr, _ = roc_curve(Yb_test, Yb_pred)
roc display = RocCurveDisplay(fpr=fpr, tpr=tpr).plot()



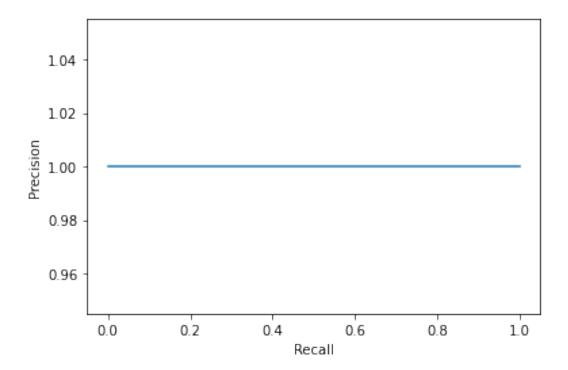
PRECISION-RECALL-F1 SCORE

```
print("Precision-Recall-Fscore[binary] on test data:",
precision_recall_fscore_support(Yb_test, Yb_pred, average='binary'))
print("Precision-Recall-Fscore[binary] on validation data:",
precision_recall_fscore_support(Yb_val, Yb_val_pred,
average='binary'))
```

Precision-Recall-Fscore[binary] on test data: (1.0, 1.0, 1.0, None) Precision-Recall-Fscore[binary] on validation data: (1.0, 1.0, 1.0, None)

PRECISION RECALL DISPLAY

```
prec, recall, _ = precision_recall_curve(Yb_test, Yb_pred)
pr_display = PrecisionRecallDisplay(precision=prec,
recall=recall).plot()
```



CROSS VALIDATION SCORE

```
seed = 7
kfold = model_selection.KFold(n_splits=10, random_state=seed,
shuffle=True)
results = model_selection.cross_val_score(abc, Xb, Yb, cv=kfold)
print("Cross validation score:",results.mean())
```

MULTICLASS DATASET

IMPORT AND READ MULTICLASS DATASET

```
mdataframe = pandas.read_csv("multi.csv", names=names)
array = mdataframe.values
Xm = array[:,0:9]
Ym = array[:,9]
```

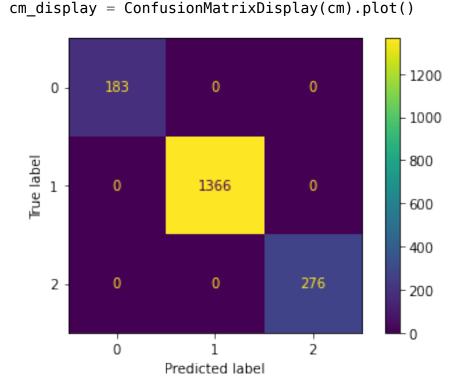
SPLITTING TRAIN, TEST AND VALIDATION DATA

```
Xm_train, Xm_test, Ym_train, Ym_test = train_test_split(Xm, Ym,
test_size=0.3)
Xm_test, Xm_val, Ym_test, Ym_val = train_test_split(Xm_test, Ym_test,
test_size=0.4)
```

CREATING AND FITTING ADABOOST CLASSIFIER TO MODEL

```
mstart = time.time()
m_abc_model = abc.fit(Xm_train, Ym_train)
mend = time.time()
```

```
# total time taken
print(f"Runtime of the Adaboost is {mend - mstart}")
Runtime of the Adaboost is 0.09352445602416992
ACCURACY
Ym_pred = m_abc_model.predict(Xm_test)
print("Accuracy on test data:", metrics.accuracy_score(Ym_test, Ym_pred))
Ym_val_pred = m_abc_model.predict(Xm_val)
print("Accuracy on validation data:", metrics.accuracy_score(Ym_val, Ym_val_pred))
Accuracy on test data: 1.0
Accuracy on validation data: 1.0
CONFUSION MATRIX
cm = confusion matrix(Ym test, Ym pred)
```



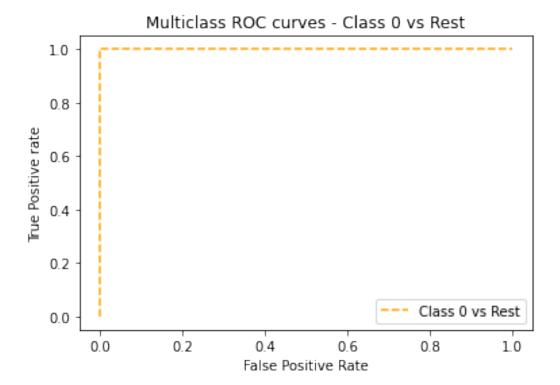
PRECISION-RECALL-F1 SCORE

```
print("Precision-Recall-Fscore[weighted] on test data:",
precision_recall_fscore_support(Ym_test, Ym_pred, average='weighted'))
print("Precision-Recall-Fscore[weighted] on validation data:",
precision_recall_fscore_support(Ym_val, Ym_val_pred,
average='weighted'))
```

```
Precision-Recall-Fscore[weighted] on test data: (1.0, 1.0, 1.0, None)
Precision-Recall-Fscore[weighted] on validation data: (1.0, 1.0, 1.0, None)

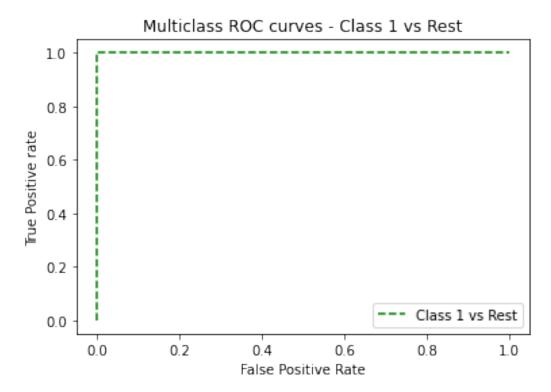
AUC-ROC CURVE
```

```
n_class = 3
Ym_pred_prob = m_abc_model.predict_proba(Xm_test)
fpr = {}
tpr = {}
thresh = {}
for i in range(n_class):
    fpr[i], tpr[i], thresh[i] = roc_curve(Ym_test, Ym_pred_prob[:,i],
    pos_label=i)
plt.plot(fpr[0], tpr[0], linestyle='--',color='orange', label='Class 0
vs Rest')
plt.title('Multiclass ROC curves - Class 0 vs Rest')
plt.xlabel('False Positive Rate')
plt.ylabel('True Positive rate')
plt.legend(loc='best')
plt.savefig('Multiclass ROC - class 0',dpi=300)
```

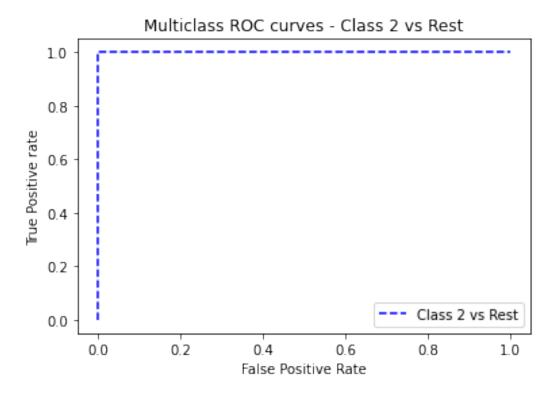


```
plt.plot(fpr[1], tpr[1], linestyle='--',color='green', label='Class 1
vs Rest')
plt.title('Multiclass ROC curves - Class 1 vs Rest')
plt.xlabel('False Positive Rate')
plt.ylabel('True Positive rate')
```

```
plt.legend(loc='best')
plt.savefig('Multiclass ROC - class 1',dpi=300)
```



```
plt.plot(fpr[1], tpr[1], linestyle='--',color='blue', label='Class 2
vs Rest')
plt.title('Multiclass ROC curves - Class 2 vs Rest')
plt.xlabel('False Positive Rate')
plt.ylabel('True Positive rate')
plt.legend(loc='best')
plt.savefig('Multiclass ROC - class 2',dpi=300)
```



CROSS VALIDATION SCORE

results = model_selection.cross_val_score(abc, Xm, Ym, cv=kfold)
print("Cross validation score:",results.mean())

Cross validation score: 0.999802761341223