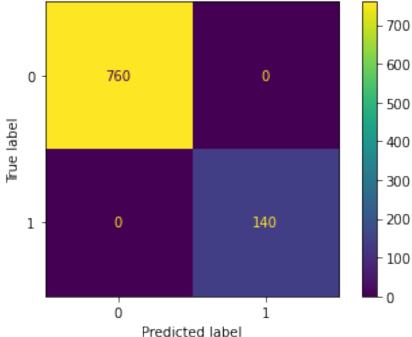
IMPORT NEEDED LIBRARIES

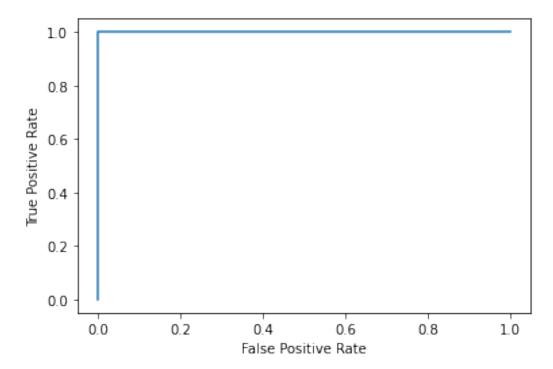
```
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn.neighbors import KNeighborsClassifier
from sklearn.linear model import LogisticRegression
from sklearn.ensemble import StackingClassifier
import xgboost
import pandas as pd
from sklearn.model selection import KFold
from sklearn.model selection import cross val score
import time
from sklearn.model selection import train test split
from sklearn import metrics
from sklearn.metrics import precision recall fscore support
import matplotlib.pyplot as plt
import tensorflow as tf
from tensorflow import keras
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
BINARY DATASET
IMPORT AND READ BINARY DATASET
names =
['pH','TDS','Turbidity','Phospate','Nitrate','Iron','COD(mg/L)','Chlor
ine','Sodium','Class']
bdataframe = pd.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)
SPECIFYING BASE CLASSIFIERS
dtc = DecisionTreeClassifier()
rfc = RandomForestClassifier()
knn = KNeighborsClassifier()
xgb = xgboost.XGBClassifier()
```

```
clf = [('dtc',dtc),('rfc',rfc),('knn',knn),('xgb',xgb)]
lr = LogisticRegression()
CREATING AND FITTING STACKING CLASSIFIER TO MODEL
bstart = time.time()
stacking = StackingClassifier(estimators = clf,final estimator = lr)
b_stacking_model = stacking.fit(Xb_train, Yb_train)
bend = time.time()
# total time taken
print(f"Runtime of the Stacking is {bend - bstart}")
Runtime of the Stacking is 3.124774217605591
ACCURACY
Yb pred = b stacking model.predict(Xb test)
print("Accuracy on test data:", metrics.accuracy score(Yb test,
Yb pred))
Yb val pred = b stacking 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()
                                              700
                                              600
             760
                               0
```



AUC-ROC CURVE

```
fpr, tpr, _ = roc_curve(Yb_test, Yb_pred)
roc_display = RocCurveDisplay(fpr=fpr, tpr=tpr).plot()
```



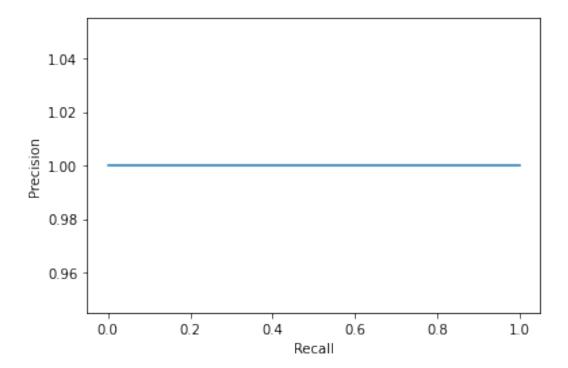
PRECISION-RECALL-F1 SCORE

```
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

```
results = cross_val_score(stacking,Xb,Yb,cv = 5,scoring = 'accuracy')
print("Cross validation score:", results.mean())
```

MULTICLASS DATASET

IMPORT AND READ MULTICLASS DATASET

```
mdataframe = pd.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 BAGGING CLASSIFIER TO MODEL

```
mstart = time.time()
m_stacking_model = stacking.fit(Xm_train, Ym_train)
mend = time.time()
# total time taken
print(f"Runtime of the Stacking is {mend - mstart}")
```

Runtime of the Stacking is 7.993763208389282

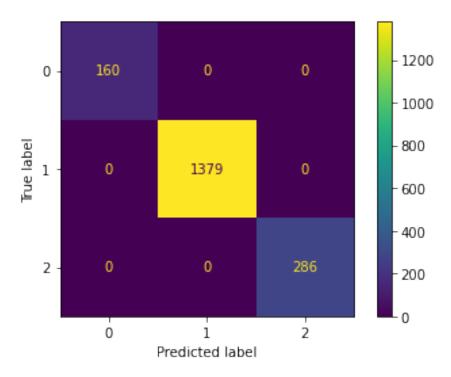
ACCURACY

```
Ym_pred = m_stacking_model.predict(Xm_test)
print("Accuracy on test data:", metrics.accuracy_score(Ym_test,
Ym_pred))
Ym_val_pred = m_stacking_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)
cm_display = ConfusionMatrixDisplay(cm).plot()



PRECISION-RECALL-F1 SCORE

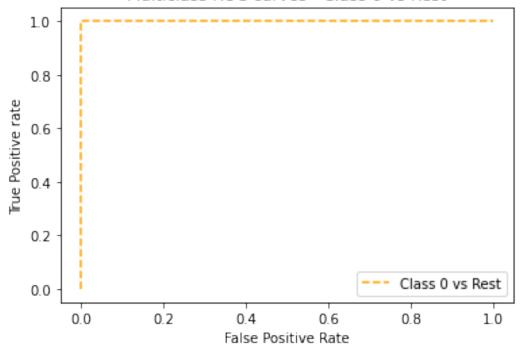
```
print("Precision-Recall-Fscore[weighted] on test data:",
precision_recall_fscore_support(Ym_test, Ym_pred, average='weighted'))
print("Precision-Recall-Fscore[weighted] on validation data:",
precision_recall_fscore_support(Ym_val, Ym_val_pred,
average='weighted'))
```

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)

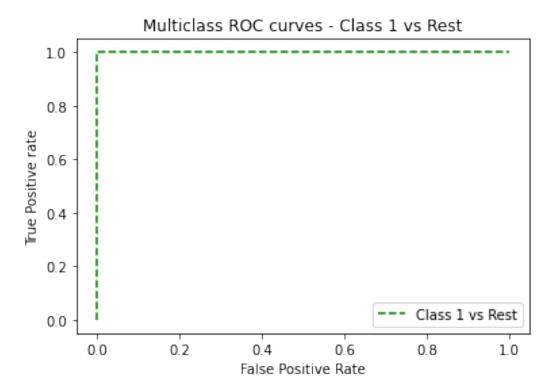
AUC-ROC CURVE

```
n_class = 3
Ym_pred_prob = m_stacking_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)
```

Multiclass ROC curves - Class 0 vs Rest

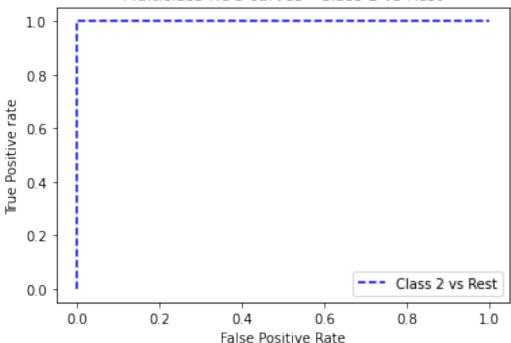


```
plt.plot(fpr[1], tpr[1], linestyle='--',color='green', label='Class 1
vs Rest')
plt.title('Multiclass ROC curves - Class 1 vs Rest')
plt.xlabel('False Positive Rate')
plt.ylabel('True Positive rate')
plt.legend(loc='best')
plt.savefig('Multiclass ROC - class 1',dpi=300)
```



```
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

```
score = cross_val_score(stacking,Xm,Ym,cv = 5,scoring = 'accuracy')
print("The accuracy score of is:",score.mean())
The accuracy score of is: 1.0
BASE MODELS ACCURACY
clf1 = [dtc,rfc,knn,xgb,lr]
clf2 = [dtc,rfc,knn,xqb]
print("BINARY DATASET")
for algo in clf1:
   bscore = cross val score( algo, Xb, Yb, cv = 5, scoring = 'accuracy')
   print("The accuracy score of {} is:".format(algo),bscore.mean())
print("\n\n\n")
print("MULTICLASS DATASET")
for algo in clf2:
   mscore = cross val score( algo,Xm,Ym,cv = 5,scoring = 'accuracy')
   print("The accuracy score of {} is:".format(algo), mscore.mean())
BINARY DATASET
The accuracy score of RandomForestClassifier() is: 0.9994
The accuracy score of KNeighborsClassifier() is: 0.9260000000000002
/usr/local/lib/python3.7/dist-packages/sklearn/linear model/
logistic.py:818: ConvergenceWarning: lbfgs failed to converge
```

```
(status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
Increase the number of iterations (max iter) or scale the data as
shown in:
    https://scikit-learn.org/stable/modules/preprocessing.html
Please also refer to the documentation for alternative solver options:
https://scikit-learn.org/stable/modules/linear model.html#logistic-
regression
  extra_warning_msg=_LOGISTIC_SOLVER_CONVERGENCE_MSG,
/usr/local/lib/python3.7/dist-packages/sklearn/linear_model/_logistic.
py:818: ConvergenceWarning: lbfgs failed to converge (status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
Increase the number of iterations (max iter) or scale the data as
shown in:
    https://scikit-learn.org/stable/modules/preprocessing.html
Please also refer to the documentation for alternative solver options:
https://scikit-learn.org/stable/modules/linear model.html#logistic-
regression
  extra warning msg= LOGISTIC SOLVER CONVERGENCE MSG,
/usr/local/lib/python3.7/dist-packages/sklearn/linear model/ logistic.
py:818: ConvergenceWarning: lbfqs failed to converge (status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
Increase the number of iterations (max iter) or scale the data as
shown in:
    https://scikit-learn.org/stable/modules/preprocessing.html
Please also refer to the documentation for alternative solver options:
https://scikit-learn.org/stable/modules/linear model.html#logistic-
regression
  extra warning msg= LOGISTIC SOLVER CONVERGENCE MSG,
/usr/local/lib/python3.7/dist-packages/sklearn/linear model/ logistic.
py:818: ConvergenceWarning: lbfqs failed to converge (status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
Increase the number of iterations (max iter) or scale the data as
shown in:
    https://scikit-learn.org/stable/modules/preprocessing.html
Please also refer to the documentation for alternative solver options:
https://scikit-learn.org/stable/modules/linear model.html#logistic-
regression
  extra warning msg= LOGISTIC SOLVER CONVERGENCE MSG,
/usr/local/lib/python3.7/dist-packages/sklearn/linear model/ logistic.
py:818: ConvergenceWarning: lbfgs failed to converge (status=1):
```

STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

Increase the number of iterations (max_iter) or scale the data as shown in:

https://scikit-learn.org/stable/modules/preprocessing.html
Please also refer to the documentation for alternative solver options:

https://scikit-learn.org/stable/modules/linear_model.html#logisticregression

extra_warning_msg=_LOGISTIC_SOLVER_CONVERGENCE_MSG,

The accuracy score of LogisticRegression() is: 0.8934

MULTICLASS DATASET

The accuracy score of DecisionTreeClassifier() is: 0.9999013806706115 The accuracy score of RandomForestClassifier() is: 0.999802761341223 The accuracy score of KNeighborsClassifier() is: 0.9271130176541735 The accuracy score of XGBClassifier() is: 0.999802761341223