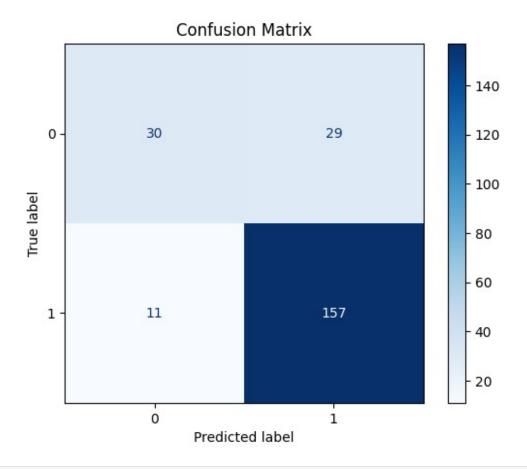
```
import pandas as pd
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import accuracy score,
classification report, ConfusionMatrixDisplay
from sklearn.metrics import confusion matrix
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.model selection import train test split
from sklearn import preprocessing
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestClassifier
from xgboost import XGBClassifier
df = pd.read_csv('pd_speech_features.csv')
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 756 entries, 0 to 755
Columns: 755 entries, id to class
dtypes: float64(749), int64(6)
memory usage: 4.4 MB
df.head()
                   PPE
                            DFA
                                    RPDE
                                          numPulses
                                                     numPeriodsPulses
   id
     gender
/
0
   0
            1 0.85247 0.71826 0.57227
                                                240
                                                                   239
                       0.69481 0.53966
                                                234
                                                                   233
1
   0
            1
              0.76686
2
    0
            1 0.85083 0.67604 0.58982
                                                232
                                                                   231
                                                178
                                                                   177
    1
              0.41121 0.79672 0.59257
            0 0.32790
                        0.79782 0.53028
                                                236
                                                                   235
   1
   meanPeriodPulses
                     stdDevPeriodPulses locPctJitter
                                                        ... \
0
           0.008064
                               0.000087
                                              0.00218
                                                        . . .
1
           0.008258
                               0.000073
                                              0.00195
2
           0.008340
                               0.000060
                                              0.00176
3
           0.010858
                               0.000183
                                              0.00419
4
           0.008162
                               0.002669
                                              0.00535
   tgwt kurtosisValue dec 28
                              tgwt kurtosisValue dec 29 \
0
                      1.5620
                                                 2.6445
1
                      1.5589
                                                 3.6107
2
                      1.5643
                                                 2.3308
3
                      3.7805
                                                 3.5664
4
                      6.1727
                                                 5.8416
```

```
tqwt kurtosisValue dec 30
                               tgwt kurtosisValue dec 31 \
0
                                                   4.2105
                       3.8686
1
                      23.5155
                                                  14.1962
2
                       9.4959
                                                  10.7458
3
                       5.2558
                                                  14.0403
4
                       6.0805
                                                   5.7621
   tgwt kurtosisValue dec 32
                               tgwt kurtosisValue dec 33 \
0
                       5.1221
                                                   4.4625
1
                      11.0261
                                                   9.5082
2
                      11.0177
                                                   4.8066
3
                       4.2235
                                                   4.6857
4
                       7.7817
                                                  11.6891
   tqwt kurtosisValue dec 34
                               tgwt kurtosisValue dec 35
0
                       2.6202
                                                   3.0004
1
                       6.5245
                                                   6.3431
2
                       2.9199
                                                   3.1495
3
                       4.8460
                                                   6.2650
4
                       8.2103
                                                   5.0559
   tgwt kurtosisValue dec 36
                               class
0
                      18.9405
                                   1
1
                      45.1780
                                   1
2
                                   1
                       4.7666
3
                       4.0603
                                   1
4
                       6.1164
                                   1
[5 rows x 755 columns]
null values=df.isnull().sum()
sum(null values!=0)
0
y = df.loc[:,'class']
X = df.drop(['class', 'id'], axis=1)
X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.3, random_state=80)
sc = StandardScaler()
X train = sc.fit transform(X train)
X test = sc.transform(X test)
clf = DecisionTreeClassifier(max_depth=5, random_state=42)
clf.fit(X train, y train)
DecisionTreeClassifier(max depth=5, random state=42)
```

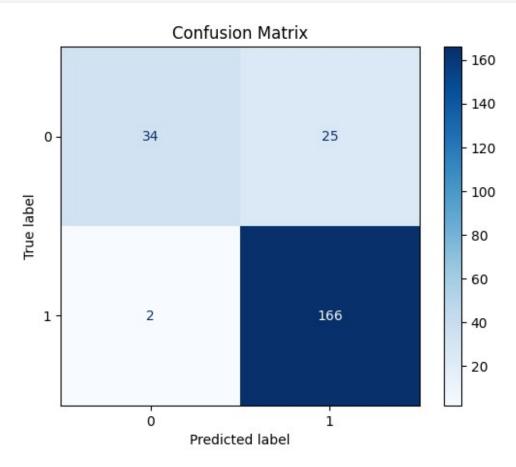
```
y pred = clf.predict(X test)
accuracy = accuracy_score(y_test, y_pred)
print("Accuracy:", accuracy)
print("Classification Report:\n", classification report(y test,
y pred))
Accuracy: 0.8237885462555066
Classification Report:
                            recall f1-score
               precision
                                               support
                             0.51
                                                   59
           0
                   0.73
                                       0.60
           1
                   0.84
                             0.93
                                       0.89
                                                  168
                                       0.82
                                                  227
    accuracy
                   0.79
                                       0.74
   macro avg
                             0.72
                                                  227
weighted avg
                   0.81
                             0.82
                                       0.81
                                                  227
cm = confusion_matrix(y_test, y_pred, labels=clf.classes_)
disp = ConfusionMatrixDisplay(confusion matrix=cm,
display labels=clf.classes )
disp.plot(cmap='Blues')
plt.title("Confusion Matrix")
```

plt.show()



```
rf model = RandomForestClassifier (n estimators = 100, max depth = 5,
random state =42)
rf_model.fit(X_train, y_train)
RandomForestClassifier(max_depth=5, random_state=42)
y_pred = rf_model.predict(X_test)
accuracy = accuracy_score(y_test, y_pred)
print("Accuracy:", accuracy)
print("Classification Report:\n", classification_report(y_test,
y pred))
Accuracy: 0.8810572687224669
Classification Report:
               precision
                             recall f1-score
                                                support
                   0.94
                              0.58
                                        0.72
                                                    59
           0
           1
                   0.87
                              0.99
                                        0.92
                                                    168
                                        0.88
                                                   227
    accuracy
   macro avg
                   0.91
                              0.78
                                        0.82
                                                    227
weighted avg
                   0.89
                              0.88
                                        0.87
                                                   227
```

```
cm = confusion_matrix(y_test, y_pred, labels=rf_model.classes_)
disp = ConfusionMatrixDisplay(confusion_matrix=cm,
display_labels=clf.classes_)
disp.plot(cmap='Blues')
plt.title("Confusion Matrix")
plt.show()
```



```
xgb = XGBClassifier(n_estimators =100 , max_depth =5 , learning_rate
=0.1)

xgb.fit(X_train, y_train )

XGBClassifier(base_score=None, booster=None, callbacks=None, colsample_bylevel=None, colsample_bynode=None, colsample_bytree=None, device=None, early_stopping_rounds=None, enable_categorical=False, eval_metric=None, feature_types=None, gamma=None, grow_policy=None, importance_type=None, interaction_constraints=None, learning_rate=0.1, max_bin=None,
```

```
max_cat_threshold=None, max_cat_to_onehot=None,
              max delta step=None, max depth=5, max leaves=None,
              min child weight=None, missing=nan,
monotone constraints=None,
              multi strategy=None, n estimators=100, n jobs=None,
              num parallel tree=None, random state=None, ...)
y pred = xgb.predict(X test)
accuracy = accuracy_score(y_test, y_pred)
print("Accuracy:", accuracy)
print("Classification Report:\n", classification_report(y_test,
y pred))
Accuracy: 0.8942731277533039
Classification Report:
               precision
                            recall f1-score
                                               support
           0
                   0.91
                             0.66
                                       0.76
                                                    59
           1
                   0.89
                             0.98
                                       0.93
                                                   168
                                       0.89
                                                   227
    accuracy
                   0.90
   macro avg
                             0.82
                                       0.85
                                                   227
weighted avg
                   0.90
                             0.89
                                       0.89
                                                   227
cm = confusion matrix(y test, y pred, labels=xgb.classes )
disp = ConfusionMatrixDisplay(confusion_matrix=cm,
display_labels=xgb.classes_)
disp.plot(cmap='Blues')
plt.title("Confusion Matrix")
```

plt.show()

