Random Forest For Audio Input

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In [3]:
         # Importing our Utilities functions:
        from utils import CautDataloaderRegular
        import os
        import pandas as pd
        import numpy as np
        import math
        import matplotlib.pyplot as plt
         # Modelling
        import pandas as pd
        import tensorflow as tf
        from tensorflow.keras.models import Sequential
        from tensorflow.keras.layers import Dense, Dropout
        from keras.layers import LSTM
        from sklearn.metrics import accuracy score, confusion matrix, precision score, recall score
        from sklearn.model selection import RandomizedSearchCV, train test split
        from scipy.stats import randint
        def RFR grid search (csv path,
                              data dir,
                              feature type="MFCC",
                              threshold=0.5):
            X y data = CautDataloaderRegular.get X y TrainTest Audio(csv path=csv path,
                                                                        data dir=data dir,
                                                                        feature type=feature type,
                                                                        input length in seconds=3,
                                                                        class to num dict={"truth": (
                                                                        verbose=False)
            X train, y train = X y data[0].reshape(-1, X y data[0].shape[1] * X y data[0].shape[2]
            X \text{ test}, y \text{ test} = X y \text{ data}[2].reshape(-1, X y data}[2].shape[1] * X y data}[2].shape[2]),
                 'n estimators': randint(50,500),
                 'max depth': randint(1,20)
            rfc = RandomForestClassifier()
            rand search = RandomizedSearchCV(rfc,
                                          param distributions = params,
                                           n iter=5,
                                           cv=5)
            rand search.fit(X train, y_train)
            best rf = rand search.best estimator
            return best rf.get params()['max depth'], best rf.get params()['n estimators']
        def RFR Model Audio (csv path,
                              data dir,
                              feature type="MFCC",
                              threshold=0.5,
                              train verbose=True):
            X y data = CautDataloaderRegular.get X y TrainTest Audio(csv path=csv path,
                                                                        data dir=data dir,
                                                                        feature type=feature type,
                                                                        input length in seconds=3,
                                                                        class to num dict={"truth": (
                                                                        verbose=train verbose)
```

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X_{\text{test}}, y_{\text{test}} = X_y_{\text{data}[2].\text{reshape}(-1, X_y_{\text{data}[2].\text{shape}[1]} * X_y_{\text{data}[2].\text{shape}[2]),
             print(X train.shape, y train.shape, X test.shape, y test.shape)
             depth, estimators = RFR grid search(csv path,
                                                     data dir,
                                                     feature type="MFCC",
                                                     threshold=0.5)
             # Setup model:
             #fitting and evaluating
             rf = RandomForestClassifier(n estimators=estimators, max depth=depth)
             # fit the model:
             rf.fit(X train, y train)
              # predict on test data:
             y pred = rf.predict(X test)
             # get metrics (accuracy, precision, recall)
             accuracy = accuracy score(y test, y pred)
             precision = precision score(y test, y pred)
             recall = recall score(y test, y pred)
             print("Accuracy:", accuracy)
             print("Precision", precision)
             print("Recall", recall)
             # show confusion matrix
             CautDataloaderRegular.plot confusion matrix(y test, y pred)
In [4]:
         RFR Model Audio(csv path="C:\\Work\\606Capstone\\Video chunks\\CSV\\",
                           data dir="C:\\Work\\606Capstone\\Video chunks\\audio features\\",
                           feature type="MFCC",
                           threshold=0.5,
                           train verbose=True)
        data dir updated to: C:\Work\606Capstone\Video chunks\audio features\MFCC audio features
        Selected csv path: C:\Work\606Capstone\Video chunks\CSV\
        Processed 100 / 520
          - Audio sample shape & label:
            - X data: (130, 20)
            - y data: 0
        Processed 200 / 520
          - Audio sample shape & label:
             - X data: (130, 20)
            - y data: 0
        Processed 300 / 520
          - Audio sample shape & label:
            - X data: (130, 20)
            - y data: 1
        Processed 400 / 520
          - Audio sample shape & label:
             - X data: (130, 20)
            - y data: 0
        Processed 500 / 520
          - Audio sample shape & label:
            - X data: (130, 20)
            - y data: 1
        Processed 100 / 223
          - Audio sample shape & label:
            - X data: (130, 20)
            - y data: 1
```

X train, y train = X y data[0].reshape(-1, X y data[0].shape[1] * X y data[0].shape[2]

Processed 200 / 223

- Audio sample shape & label:

- X data: (130, 20)

- y data: 1

Gathered data shapes:

X train.shape: (520, 130, 20)

y train.shape: (520,)

X_test.shape: (223, 130, 20)

y test.shape: (223,)

(520, 2600) (520,) (223, 2600) (223,)

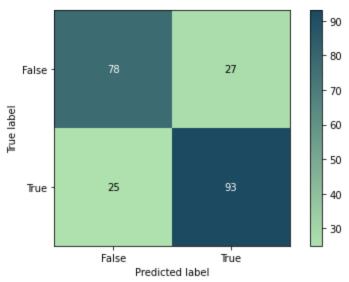
data dir updated to: C:\Work\606Capstone\Video chunks\audio features\MFCC audio features

Selected csv path: C:\Work\606Capstone\Video chunks\CSV\

Accuracy: 0.7668161434977578

Precision 0.775

Recall 0.788135593220339



Metrics Rates:

- True Positive : 93 - False Positive : 27 - True Negative : 78 - False Negative : 25

- True Positive Rate : 0.788135593220339 - True Negative Rate : 0.7428571428571429

- Positive Predictive Value: 0.775

- Negative predictive value: 0.7572815533980582 - False Positive Rate : 0.2571428571428571 - False Negative Rate : 0.211864406779661

- False Discovery Rate : 0.225