

# Random Forest For Visual Input

In [1]:

```
#download the model as a pickle file
import pickle as cPickle
def RFR_export_model(rfr,file_path):
    with open(file_path, 'wb') as f:
        cPickle.dump(rfr, f)
```

In [2]:

```
# 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
import matplotlib.colors as mcolors
import numpy as np

# Modelling
from sklearn.ensemble import RandomForestClassifier
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_Model(data_mode,
              csv_path,
              data_dir,
              approach_type,
              batch_size=32,
              threshold=0.5,
              train_verbose=True):

    X_y_data = CautDataloaderRegular.get_X_y_TrainTest_Visual(csv_path="C:\\Work\\606Capst",
                                                              data_dir=data_dir,
                                                              data_mode=data_mode,
                                                              approach_type=approach_type, # sec
                                                              verbose=True)

    if data_mode == "OpenFace" and approach_type == "average":
        X_train, y_train = X_y_data[0], X_y_data[1]
        X_test, y_test = X_y_data[2], X_y_data[3]
    else:
        X_train, y_train = X_y_data[0].reshape(-1, X_y_data[0].shape[1] * X_y_data[0].shape[2])
        X_test, y_test = X_y_data[2].reshape(-1, X_y_data[2].shape[1] * X_y_data[2].shape[2])
        X_train[np.isnan(X_train)] = 0
        X_test[np.isnan(X_test)] = 0

    # Setup model:
    #fitting and evaluating
    rf = RandomForestClassifier(n_estimators=120)

    # 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)
```

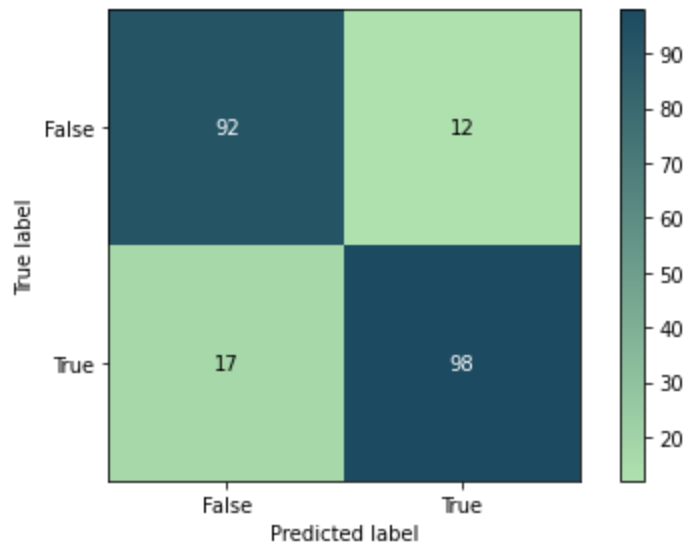
## OpenFace Average:

In [3]:

```
rfr = RFR_Model(data_mode="OpenFace",
                 csv_path="C:\\Work\\606Capstone\\Video_chunks\\CSV\\",
                 data_dir="C:\\Work\\606Capstone\\Video_chunks\\Excel\\",
                 approach_type="average",
                 threshold=0.5,
                 train_verbose=False)
RFR_export_model(rfr,"C:\\Work\\606Capstone\\Video_chunks\\Models\\OpenFaceAverage_RFR.pic")
```

[illegible]

Accuracy: 0.867579908675799  
Precision 0.8909090909090909  
Recall 0.8521739130434782



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Metrics Rates:

- True Positive : 98
- False Positive : 12
- True Negative : 92
- False Negative : 17
- True Positive Rate : 0.8521739130434782
- True Negative Rate : 0.8846153846153846
- Positive Predictive Value: 0.8909090909090909
- Negative predictive value: 0.8440366972477065
- False Positive Rate : 0.11538461538461539
- False Negative Rate : 0.14782608695652175
- False Discovery Rate : 0.10909090909090909

## OpenFace Sequential:

```
In [4]: rfr = RFR_Model(data_mode="OpenFace",  
                      csv_path="C:\\Work\\606Capstone\\Video_chunks\\CSV\\",  
                      data_dir="C:\\Work\\606Capstone\\Video_chunks\\Excel\\",  
                      approach_type="sequential",  
                      threshold=0.5,  
                      train_verbose=False)  
RFR_export_model(rfr,"C:\\Work\\606Capstone\\Video_chunks\\Models\\OpenFaceSequential_RFR.
```

Selected csv\_path: C:\Work\606Capstone\Video\_chunks\CSV\

Processed 100 / 520

- Sample shape & label:
  - X\_data: (90, 27)
  - y\_data: 0

Processed 200 / 520

- Sample shape & label:
  - X\_data: (90, 27)
  - y\_data: 0

Processed 300 / 520

- Sample shape & label:
  - X\_data: (90, 27)
  - y\_data: 1

Processed 400 / 520

- Sample shape & label:

```

- X_data: (90, 27)
- y_data: 0
Processed 500 / 520
- Sample shape & label:
- X_data: (90, 27)
- y_data: 1
Casting collected data to .numpy array type...
Data is collected. Returning X and y data.
@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@

```

```

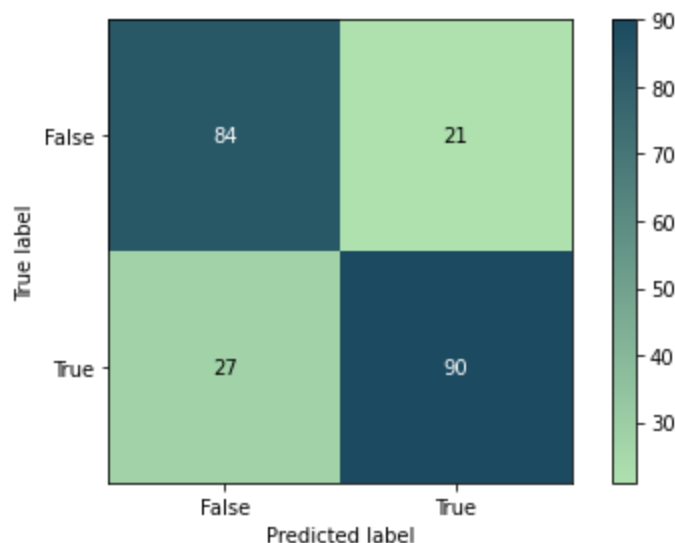
Processed 100 / 223
- Sample shape & label:
- X_data: (90, 27)
- y_data: 1
Processed 200 / 223
- Sample shape & label:
- X_data: (90, 27)
- y_data: 0
Casting collected data to .numpy array type...
Data is collected. Returning X and y data.
@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@@

```

```

Accuracy: 0.7837837837837838
Precision 0.8108108108108109
Recall 0.7692307692307693

```



```

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Metrics Rates:
- True Positive           : 90
- False Positive          : 21
- True Negative           : 84
- False Negative          : 27
- True Positive Rate      : 0.7692307692307693
- True Negative Rate      : 0.8
- Positive Predictive Value: 0.8108108108108109
- Negative predictive value: 0.7567567567567568
- False Positive Rate     : 0.2
- False Negative Rate     : 0.23076923076923078
- False Discovery Rate    : 0.1891891891891892

```

## MediaPipe Sequential:

```

In [5]: rfr = RFR_Model(data_mode="MediaPipe",
                        csv_path="C:\\Work\\606Capstone\\Video_chunks\\CSV\\",

```

```

data_dir="C:\\Work\\606Capstone\\Video_chunks\\MediaPipe\\",
approach_type=None,
threshold=0.5,
train_verbose=False)
RFR_export_model(rfr,"C:\\Work\\606Capstone\\Video_chunks\\Models\\MediaPipeSequential_RFR")

```

Selected csv\_path: C:\Work\606Capstone\Video\_chunks\CSV\

Processed 100 / 520

```

- Sample shape & label:
  - X_data: (90, 36)
  - y_data: 0

```

Processed 200 / 520

```

- Sample shape & label:
  - X_data: (90, 36)
  - y_data: 0

```

Processed 300 / 520

```

- Sample shape & label:
  - X_data: (90, 36)
  - y_data: 1

```

Processed 400 / 520

```

- Sample shape & label:
  - X_data: (90, 36)
  - y_data: 0

```

Processed 500 / 520

```

- Sample shape & label:
  - X_data: (90, 36)
  - y_data: 1

```

Processed 100 / 223

```

- Sample shape & label:
  - X_data: (90, 36)
  - y_data: 1

```

Processed 200 / 223

```

- Sample shape & label:
  - X_data: (90, 36)
  - y_data: 1

```

Gathered data shapes:

X\_train.shape: (520, 90, 36)

y\_train.shape: (520,)

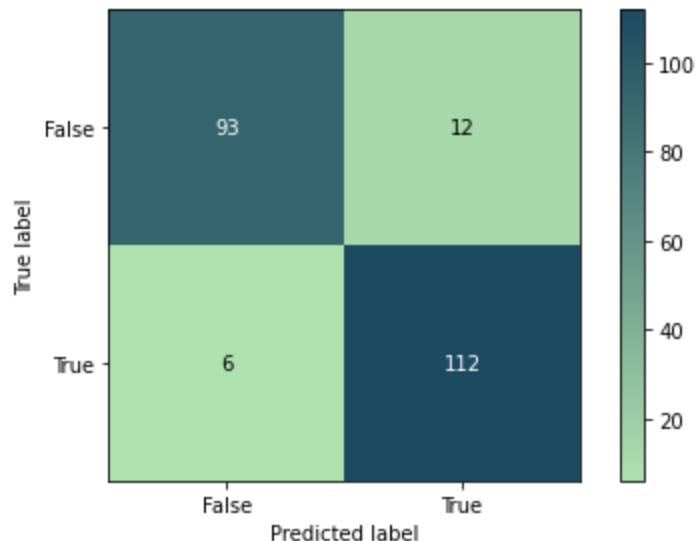
X\_test.shape: (223, 90, 36)

y\_test.shape: (223,)

Accuracy: 0.9192825112107623

Precision 0.9032258064516129

Recall 0.9491525423728814



Metrics Rates:

- True Positive : 112

- False Positive	:	12
- True Negative	:	93
- False Negative	:	6
- True Positive Rate	:	0.9491525423728814
- True Negative Rate	:	0.8857142857142857
- Positive Predictive Value:	:	0.9032258064516129
- Negative predictive value:	:	0.9393939393939394
- False Positive Rate	:	0.11428571428571428
- False Negative Rate	:	0.05084745762711865
- False Discovery Rate	:	0.0967741935483871