# HAP\_ModelSelection\_Evaluation

March 5, 2022

```
[1]: from ipynb.fs.full.HAP_DataProcessing import *
```

<class 'pandas.core.frame.DataFrame'> Int64Index: 595 entries, 1 to 927 Data columns (total 14 columns):

| #                             | Column   | Non-Null Count | Dtype   |  |
|-------------------------------|----------|----------------|---------|--|
|                               |          |                |         |  |
| 0                             | age      | 595 non-null   | float64 |  |
| 1                             | sex      | 595 non-null   | float64 |  |
| 2                             | ср       | 595 non-null   | float64 |  |
| 3                             | trestbps | 595 non-null   | float64 |  |
| 4                             | chol     | 595 non-null   | float64 |  |
| 5                             | fbs      | 595 non-null   | float64 |  |
| 6                             | restecg  | 595 non-null   | float64 |  |
| 7                             | thalach  | 595 non-null   | float64 |  |
| 8                             | exang    | 595 non-null   | float64 |  |
| 9                             | oldpeak  | 595 non-null   | float64 |  |
| 10                            | slope    | 595 non-null   | float64 |  |
| 11                            | ca       | 595 non-null   | float64 |  |
| 12                            | thal     | 595 non-null   | float64 |  |
| 13                            | num      | 595 non-null   | int64   |  |
| dtypes: float64(13), int64(1) |          |                |         |  |

memory usage: 69.7 KB

None

## Logistic Regression

```
[2]: import scipy.optimize as opt
    import statsmodels.api as sm
    from sklearn import preprocessing
   X = np.asarray(hap_df[['age', 'sex', 'cp', 'trestbps', 'chol', _
    y = np.asarray(hap_df['num'])
    # normalization of the dataset
    X = preprocessing.StandardScaler().fit(X).transform(X)
```

```
# Train-and-Test -Split
     from sklearn.model_selection import train_test_split
     X_train, X_test, y_train, y_test = train_test_split(
            X, y, test_size = 0.3, random_state = 4)
     print ('Train set:', X_train.shape, y_train.shape)
     print ('Test set:', X_test.shape, y_test.shape)
    Train set: (416, 13) (416,)
    Test set: (179, 13) (179,)
[3]: from sklearn.linear_model import LogisticRegression
     logreg = LogisticRegression()
     logreg.fit(X_train, y_train)
     logreg_y_pred = logreg.predict(X_test)
     # Evaluation and accuracy
     from sklearn.metrics import accuracy_score
     print('')
     print('Accuracy score is = ',
           accuracy_score(y_test, logreg_y_pred))
```

Accuracy score is = 0.8938547486033519



| The details for confusion matrix is = |           |        |          |         |
|---------------------------------------|-----------|--------|----------|---------|
|                                       | precision | recall | f1-score | support |
|                                       |           |        |          |         |
| 0                                     | 0.92      | 0.87   | 0.90     | 94      |
| 1                                     | 0.87      | 0.92   | 0.89     | 85      |
|                                       |           |        |          |         |
| accuracy                              |           |        | 0.89     | 179     |
| macro avg                             | 0.89      | 0.89   | 0.89     | 179     |
| weighted avg                          | 0.90      | 0.89   | 0.89     | 179     |

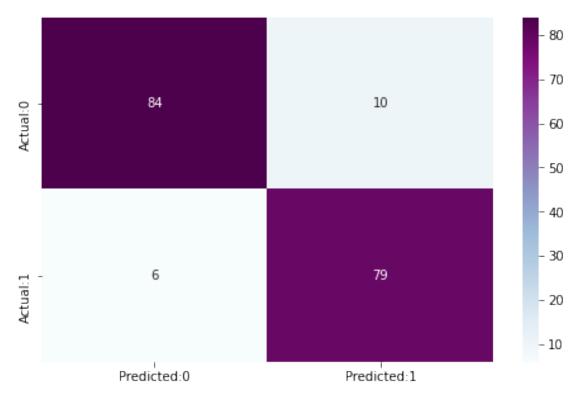
## 2 Decission Tree

```
[5]: from sklearn import tree
  dtree = tree.DecisionTreeClassifier()
  dtree.fit(X_train, y_train)
  dtree_y_pred = dtree.predict(X_test)

# Evaluation and accuracy
  from sklearn.metrics import accuracy_score
  print('')
  print('Accuracy score is = ',
```

```
accuracy_score(y_test, dtree_y_pred))
```

Accuracy score is = 0.9106145251396648

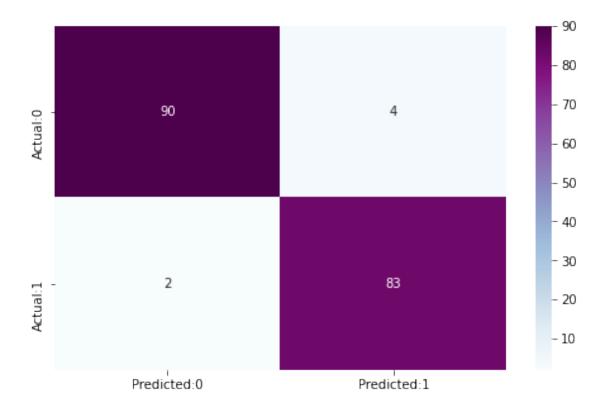


The details for confusion matrix is = precision recall f1-score support

| 0            | 0.93 | 0.89 | 0.91 | 94  |
|--------------|------|------|------|-----|
| 1            | 0.89 | 0.93 | 0.91 | 85  |
| accuracy     |      |      | 0.91 | 179 |
| macro avg    | 0.91 | 0.91 | 0.91 | 179 |
| weighted avg | 0.91 | 0.91 | 0.91 | 179 |

#### 3 Random Forest

Accuracy score is = 0.9664804469273743



The details for confusion matrix is = precision recall f1-score support 0 0.98 0.96 0.97 94 1 0.95 0.98 0.97 85 accuracy 0.97 179 0.97 0.97 179 macro avg 0.97 weighted avg 0.97 0.97 0.97 179

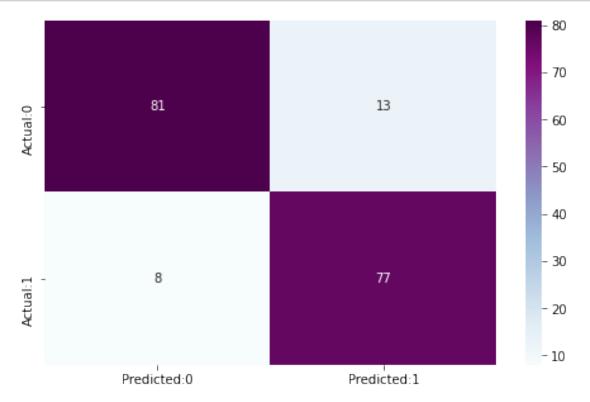
# 4 Support Vector Machine

```
[9]: from sklearn.svm import SVC
svc = SVC(kernel="linear",probability=True)
svc.fit(X_train, y_train)
svc_y_pred = svc.predict(X_test)

# Evaluation and accuracy
from sklearn.metrics import accuracy_score
print('')
print('Accuracy score is = ',
```

```
accuracy_score(y_test, svc_y_pred))
```

Accuracy score is = 0.88268156424581



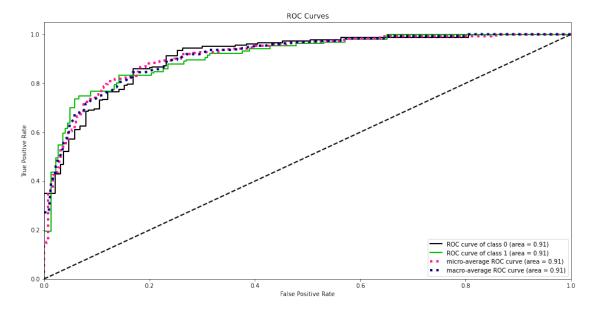
The details for confusion matrix is = precision recall f1-score support

| 0            | 0.91 | 0.86 | 0.89 | 94  |
|--------------|------|------|------|-----|
| 1            | 0.86 | 0.91 | 0.88 | 85  |
|              |      |      |      |     |
| accuracy     |      |      | 0.88 | 179 |
| macro avg    | 0.88 | 0.88 | 0.88 | 179 |
| weighted avg | 0.88 | 0.88 | 0.88 | 179 |

### 5 Model evaluation

Populating the interactive namespace from numpy and matplotlib

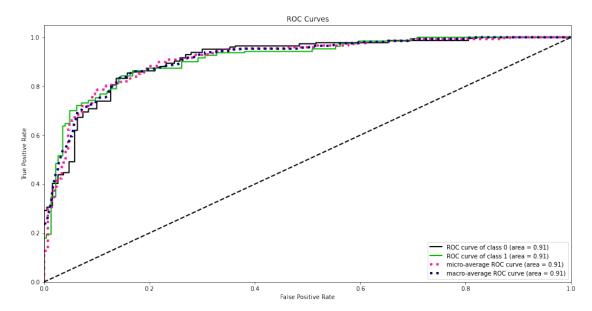
```
C:\Users\aditya.sumbaraju\Anaconda3\lib\site-
packages\IPython\core\magics\pylab.py:159: UserWarning: pylab import has
clobbered these variables: ['cm']
`%matplotlib` prevents importing * from pylab and numpy
  warn("pylab import has clobbered these variables: %s" % clobbered +
```



```
[12]: # SVC %pylab inline
```

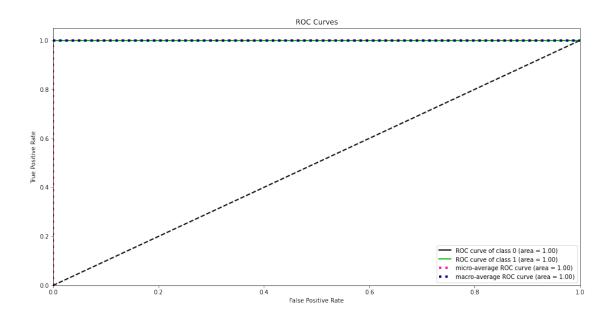
```
pylab.rcParams['figure.figsize'] = (16,8)
svc_y_probas = svc.predict_proba(X_train)
skplt.metrics.plot_roc(y_train, svc_y_probas)
plt.show()
```

Populating the interactive namespace from numpy and matplotlib

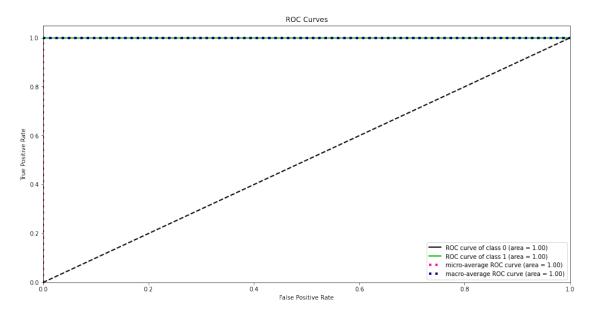


```
[13]: # Random Forest
%pylab inline
   pylab.rcParams['figure.figsize'] = (16,8)
   rfc_y_probas = rf.predict_proba(X_train)
   skplt.metrics.plot_roc(y_train, rfc_y_probas)
   plt.show()
```

Populating the interactive namespace from numpy and matplotlib



Populating the interactive namespace from numpy and matplotlib



### 6 Model Results

|   | МО                    | DEL | ACCURACY_SCORE |
|---|-----------------------|-----|----------------|
| 1 | Random Forest Classif | ier | 0.97           |
| 2 | Decision T            | ree | 0.91           |
| 3 | Logistic Regress      | ion | 0.89           |
| 4 | SVM Classif           | ier | 0.88           |

[]: