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Roll No.: 22522913

Lab: 6

```
1
                                             Predictive Analytics for Hospitals
       Step:1
In [1]:
            1 import pandas as pd
In [2]:
            1 data = pd.read_csv('diabetes.csv')
In [3]:
            1 data.head(10)
   Out[3]:
              Pregnancies Glucose
                               BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction Age Outcome
           0
                     6
                           148
                                        72
                                                   35
                                                          0 33.6
                                                                                     50
                                                                                              1
                                                                               0.627
                            85
                                        66
                                                   29
                                                          0 26.6
                                                                               0.351
                                                                                     31
                                                                                              0
                     8
                           183
                                        64
                                                    0
                                                          0 23.3
                                                                               0.672
                                                                                     32
                            89
                                        66
                                                         94 28.1
                                                                               0.167
                     0
                           137
                                        40
                                                   35
                                                        168 43.1
                                                                               2.288
                                                                                     33
                           116
                                        74
                                                    0
                                                          0 25.6
                                                                               0.201
                                                                                     30
                     3
                            78
                                        50
                                                   32
                                                         88 31.0
                                                                               0.248
                                                                                     26
                     10
                           115
                                        0
                                                    0
                                                          0 35.3
                                                                               0.134
                                                                                     29
                                                                                              0
                     2
                           197
                                        70
                                                   45
                                                        543 30.5
                                                                               0.158
                                                                                     53
                           125
                                        96
                                                    0
                                                          0.0
                                                                               0.232
In [4]:  ▶ 1 data.shape
   Out[4]: (768, 9)
In [5]: ▶
           1 data.columns
   dtype='object')
In [6]: ▶ 1 | data.dtypes
   Out[6]: Pregnancies
                                       int64
           Glucose
                                      int64
           BloodPressure
                                      int64
           SkinThickness
                                      int64
           Insulin
                                      int64
           BMI
                                     float64
          {\tt DiabetesPedigreeFunction}
                                     float64
           Age
                                      int64
           Outcome
                                      int64
           dtype: object
Out[7]: <bound method IndexOpsMixin.value_counts of 0</pre>
                                                          0.627
                 0.351
                 0.672
           2
                 0.167
           3
           4
                 2.288
           763
                 0.171
           764
                 0.340
           765
                 0.245
           766
                 0.349
           Name: DiabetesPedigreeFunction, Length: 768, dtype: float64>
```

In [8]: 1 data.describe() Out[8]: Pregnancies Glucose BloodPressure SkinThickness BMI DiabetesPedigreeFunction Age Insulin Outcome 768.000000 768.000000 768.000000 768.000000 768.000000 768.000000 768.000000 768.000000 768.000000 count 120.894531 69.105469 33.240885 3.845052 20.536458 79.799479 31.992578 0.471876 0.348958 mean 3.369578 31.972618 19.355807 15.952218 115.244002 7.884160 0.331329 11.760232 0.476951 std min 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.078000 21.000000 0.000000 25% 1.000000 99.000000 62.000000 0.000000 0.000000 27.300000 0.243750 24.000000 0.000000 50% 3.000000 117.000000 72.000000 23.000000 30.500000 32.000000 0.372500 29.000000 0.000000 75% 6.000000 140.250000 80.000000 32.000000 127.250000 36.600000 0.626250 41.000000 1.000000 max 17.000000 199.000000 122.000000 99.000000 846.000000 67.100000 2.420000 81.000000 1.000000 step:2 In [9]: ▶ 1 import numpy as np import seaborn as sns 3 import matplotlib.pyplot as plt In [10]: M 1 data.drop_duplicates() Out[10]: Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction Age Outcome 33.6 50 1 0 6 148 72 35 0 0.627 1 1 29 0.351 31 0 85 66 0 26.6 2 8 183 64 Ω 23.3 0.672 32 1 0 3 1 89 66 23 94 28.1 0.167 21 0 0 4 137 40 35 2.288 33 168 43.1 1 763 10 101 76 48 63 0 180 32.9 0.171 764 2 122 70 27 0 36.8 0.340 27 0 5 72 23 30 0 765 121 112 26.2 0.245 766 126 60 0 30.1 0.349 47

768 rows × 9 columns

93

70

31

30.4

0.315

23

0

767

```
In [54]: ▶
                    1 import seaborn as sns
                       import matplotlib.pyplot as plt
                    3 plt.figure(figsize=(10,10))
                    4 sns.heatmap(data.head(10), cmap='BuPu',annot=True, linewidth=.10)
     Out[54]: <AxesSubplot:>
                          6
                               1.5e+02
                                           72
                                                    35
                                                             0
                                                                     34
                                                                             0.63
                                                                                       50
                                                                                                               - 500
                          1
                                  85
                                           66
                                                   29
                                                             0
                                                                     27
                                                                             0.35
                                                                                       31
                                                                                               0
                          8
                               1.8e+02
                                           64
                                                    0
                                                             0
                                                                     23
                                                                             0.67
                                                                                       32
                                                                                                1
                                                                                                               400
                                  89
                                           66
                                                   23
                                                            94
                                                                     28
                                                                             0.17
                                                                                       21
                                                                                                0
                                                                                                               - 300
                         0
                               1.4e+02
                                          40
                                                    35
                                                         1.7e+02
                                                                     43
                                                                              23
                                                                                       33
                                                                                               1
                               1.2e+02
                                           74
                                                    0
                                                             0
                                                                     26
                                                                              0.2
                                                                                       30
                                                                                                0
                                                                                                               200
                                  78
                                           50
                                                   32
                                                            88
                                                                     31
                                                                                               1
                          3
                                                                             0.25
                                                                                       26
                         10
                               1.2e+02
                                           0
                                                    0
                                                             0
                                                                     35
                                                                             0.13
                                                                                      29
                                                                                               0
                                                                                                               - 100
                          2
                                2e+02
                                           70
                                                   45
                                                                     30
                                                                             0.16
                                                                                      53
                                                                                               1
                         8
                               1.2e+02
                                                             0
                                                                     0
                                          96
                                                    0
                                                                             0.23
                                                                                      54
                                                                                               1
                                                                                                              - 0
                                                                                      Age
                                                                     Μ̈́
                                                                              DiabetesPedigreeFunction
                                                                                               Outcome
                                  Glucose
                                           BloodPressure
```

```
step:3
```

```
In [12]: ▶
          1 X=data["Age"]
            y=data['Outcome']
In [13]:
          1 X = np.array(X).reshape(-1, 1)
In [14]:
          1 from sklearn.model_selection import train_test_split
          2 from sklearn.linear_model import LogisticRegression
In [15]:
          1 X_train,x_test,Y_train,y_test = train_test_split(X,y,test_size=.25,random_state=11)
In [16]: ▶
          1 logr = LogisticRegression()
          2 logr.fit(X_train,Y_train)
  Out[16]: LogisticRegression()
In [17]: ▶
         1 logr.predict(x_test)
  0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 1, 0,
              0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0,
              1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0,
              0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0,
              0, 1, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0], dtype=int64)
```

```
In [18]: ▶
             1 print("Coef :",logr.coef_)
              2 print("Intercept :",logr.intercept_)
            Coef : [[0.04278121]]
            Intercept : [-2.06807344]
In [19]: ► 1 logr.predict([[60]])
   Out[19]: array([1], dtype=int64)
In [20]: ▶
             1 lrf = logr.coef_*60 + logr.intercept_
In [21]: ▶
             1 from scipy.special import expit
In [22]: ▶
             1 output=expit(lrf)
In [23]: ▶
             1 output
   Out[23]: array([[0.62217709]])
In [24]:
             1 if output > 0.5 :
                   print(" YES he will become diabetic")
             3 else :
                     print(" NO he will not be diabetic")
             4
             YES he will become diabetic
        step:4
In [25]: ▶
             1 X=data[['Glucose','BMI','Age']]
              2 y=data['Outcome']
             1 X_train,x_test,Y_train,y_test = train_test_split(X,y,test_size=.25,random_state=11)
In [26]:
In [27]: ▶
             1 logr = LogisticRegression()
              2 logr.fit(X_train,Y_train)
   Out[27]: LogisticRegression()
Out[28]: array([0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0,
                   1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
                   0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 1, 0, 1, 0, 0, 0,
                   0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0,
                   1, 1, 1, 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1,
                   0, 1, 0, 1, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0,
                   1, 0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0,
                   0, 0, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0,
                   0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0], dtype=int64)
In [29]: ▶
             1 print("Coef :",logr.coef_)
              2 print("Intercept :",logr.intercept_)
            Coef : [[0.03358049 0.07889299 0.02722911]]
            Intercept : [-8.37441801]
In [30]: | 1 | logr.predict([[150,30,40]])
            C:\Users\arulk\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X does not have valid feature names, but Logist
            icRegression was fitted with feature names
              warnings.warn(
   Out[30]: array([1], dtype=int64)
In [31]: ▶
             1 lrf = logr.coef_*150*30*40 + logr.intercept_
In [32]:
             1 output=expit(lrf)
```

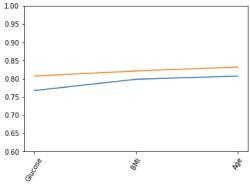
```
In [33]: ▶ 1 output
   Out[33]: array([[1., 1., 1.]])
C:\Users\arulk\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X does not have valid feature names, but Logist
           icRegression was fitted with feature names
             warnings.warn(
Out[35]: array([False, True])
In [36]: ▶ 1 output1
   Out[36]: array([[0.47038229, 0.52961771]])
        Step:5
In [37]: ▶
            1 | X1=data.drop(['Outcome'],axis=1)
In [38]:
            1 X train,x test,Y train,y test = train test split(X,y,test size=.25,random state=11)
In [39]: ▶
            1 LOR=LogisticRegression()
             2 LOR.fit(X_train,Y_train)
   Out[39]: LogisticRegression()
In [40]:  ▶ 1 LOR.predict(x_test)
   Out[40]: array([0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0,
                  1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
                 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 1, 0, 1, 0, 0, 0,
                 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0,
                 1, 1, 1, 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1,
                 0, 1, 0, 1, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0,
                 1, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0,
                 0, 0, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0,
                 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0], dtype=int64)
In [41]: ▶
            1 y_pred=logr.predict(x_test)
             2 y_pred
   1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
                 0,\ 0,\ 0,\ 0,\ 0,\ 1,\ 0,\ 0,\ 0,\ 0,\ 0,\ 1,\ 1,\ 0,\ 1,\ 0,\ 1,\ 0,\ 0,\ 0,
                 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0,
                 1, 1, 1, 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1,
                 0,\ 1,\ 0,\ 1,\ 0,\ 0,\ 1,\ 0,\ 1,\ 0,\ 0,\ 0,\ 0,\ 0,\ 0,\ 1,\ 0,\ 1,\ 0,\ 0,
                 1, 0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0,
                 0, 0, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0,
                 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0], dtype=int64)
In [42]: ▶
            1 from sklearn.metrics import roc_auc_score
               lor_auc=roc_auc_score(y_test,y_pred)
            3
Auc: 0.6756854256854258
```

step:6

```
In [44]: ▶
              1 def get_auc(var,tar,data):
              2
                     fx=data[var]
                     fy=data[tar]
              3
              4
                     LOR4=LogisticRegression()
              5
                     LOR4.fit(fx,fy)
                     pred=LOR4.predict_proba(fx)[:,1]
              6
                     auc_val = roc_auc_score(y,pred)
              7
                     return auc_val
C:\Users\arulk\anaconda3\lib\site-packages\sklearn\utils\validation.py:993: DataConversionWarning: A column-vector y was pas
             sed when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
              y = column_or_1d(y, warn=True)
   Out[45]: 0.8109328358208956
In [46]: N 1 get_auc(['Pregnancies', 'BloodPressure', 'SkinThickness'], ['Outcome'], data)
            C:\Users\arulk\anaconda3\lib\site-packages\sklearn\utils\validation.py:993: DataConversionWarning: A column-vector y was pas
             sed when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
              y = column_or_1d(y, warn=True)
   Out[46]: 0.6444962686567164
In [47]: ▶
              1 def best_next(current,cand,tar,data):
                     best_auc=-1
                     best_var=None
              4
                     for i in cand:
                         auc_v = get_auc(current+[i],tar,data)
              5
              6
                         if auc_v>=best_auc:
              7
                            best_auc=auc_v
              8
                            best_var=i
                     return best_var
              1 current=['Insulin','BMI','DiabetesPedigreeFunction','Age']
In [48]: ▶
              2 cand=['Pregnancies','Glucose','BloodPressure','SkinThickness']
              3 tar=['Outcome']
              4 next_var = best_next(current,cand,tar,data)
            C:\Users\arulk\anaconda3\lib\site-packages\sklearn\utils\validation.py:993: DataConversionWarning: A column-vector y was pas
             sed when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
              y = column_or_1d(y, warn=True)
            C:\Users\arulk\anaconda3\lib\site-packages\sklearn\utils\validation.py:993: DataConversionWarning: A column-vector y was pas
             sed when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
              y = column_or_1d(y, warn=True)
             C:\Users\arulk\anaconda3\lib\site-packages\sklearn\utils\validation.py:993: DataConversionWarning: A column-vector y was pas
             sed when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
              y = column_or_1d(y, warn=True)
             C:\Users\arulk\anaconda3\lib\site-packages\sklearn\utils\validation.py:993: DataConversionWarning: A column-vector y was pas
             sed when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
               y = column_or_1d(y, warn=True)
In [49]: ► 1 next_var
   Out[49]: 'Glucose'
```

```
In [50]: ▶
              1 tar =['Outcome']
               2 current=[]
               3 cand=['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin', 'BMI', 'DiabetesPedigreeFunction', 'Age']
              4 \text{ max\_num} = 7
                 num_it = min(max_num,len(cand))
                 for i in range(0,num_it):
                     next var = best next(current.cand.tar.data)
              8
                      current += [next_var]
              9
                      cand.remove(next var)
              10
                     print("variable add in step "+str(i+1)+' is '+ next_var +" .")
             11 print(current)
             C:\Users\aruɪk\anaconda3\lib\site-packages\sklearn\utils\validation.py:993: DataConversionwarning: A column-vector y was
             passed when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
               y = column_or_1d(y, warn=True)
             variable add in step 3 is Pregnancies .
             C:\Users\arulk\anaconda3\lib\site-packages\sklearn\utils\validation.py:993: DataConversionWarning: A column-vector y was
             passed when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
               y = column_or_1d(y, warn=True)
             C:\Users\arulk\anaconda3\lib\site-packages\sklearn\utils\validation.py:993: DataConversionWarning: A column-vector y was
             passed when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
               y = column_or_1d(y, warn=True)
             C:\Users\arulk\anaconda3\lib\site-packages\sklearn\utils\validation.py:993: DataConversionWarning: A column-vector y was
             passed when a 1d array was expected. Please change the shape of y to (n samples, ), for example using ravel().
               y = column_or_1d(y, warn=True)
             C:\Users\arulk\anaconda3\lib\site-packages\sklearn\utils\validation.py:993: DataConversionWarning: A column-vector y was
             passed when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
               y = column_or_1d(y, warn=True)
             C:\Users\arulk\anaconda3\lib\site-packages\sklearn\utils\validation.py:993: DataConversionWarning: A column-vector y was
             passed when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
               V - column on 1d/v warn-True)
         step:7
In [58]:
              1 X_train,X_test,y_train,y_test = train_test_split(X,y,test_size = 0.5,stratify =y)
In [60]:
          М
              1 pred2 = logr.predict_proba(X_test)
In [61]: ▶
              1 train = pd.concat([X_train,y_train], axis=1)
               2 test = pd.concat([X_test,y_test], axis=1)
              1 def auc_train_test(variables, target, train, test):
In [62]: ▶
                      X_train = train[variables]
              3
                      X_test = test[variables]
               4
                      Y_train = train[target]
               5
                      Y test = test[target]
                      logreg = LogisticRegression()
              6
              7
                      logreg.fit(X_train, Y_train)
              8
                      predictions_train = logreg.predict_proba(X_train)[:,1]
                      predictions_test = logreg.predict_proba(X_test)[:,1]
              10
                      auc_train = roc_auc_score(Y_train, predictions_train)
              11
                      auc_test = roc_auc_score(Y_test,predictions_test)
              12
                      return(auc_train, auc_test)
In [64]: ▶
              1 auc_values_train = []
                 auc_values_test = []
                 variables_evaluate = []
              4
                 for v in X.columns:
                      variables_evaluate.append(v)
                      auc_train, auc_test = auc_train_test(variables_evaluate,["Outcome"],train,test)
               6
                      auc_values_train.append(auc_train)
                      auc_values_test.append(auc_test)
             C:\Users\arulk\anaconda3\lib\site-packages\sklearn\utils\validation.py:993: DataConversionWarning: A column-vector y was pas
             sed when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
               y = column_or_1d(y, warn=True)
             C:\Users\arulk\anaconda3\lib\site-packages\sklearn\utils\validation.py:993: DataConversionWarning: A column-vector y was pas
             sed when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
               y = column_or_1d(y, warn=True)
             C:\Users\arulk\anaconda3\lib\site-packages\sklearn\utils\validation.py:993: DataConversionWarning: A column-vector y was pas
             sed when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
               y = column_or_1d(y, warn=True)
```

```
In [69]: ▶
              1 import matplotlib.pyplot as plt
              2 import numpy as np
              3 x = np.array(range(0,len(auc_values_train)))
              4 my_train = np.array(auc_values_train)
              5 my_test = np.array(auc_values_test)
              6 plt.xticks(x,X.columns,rotation=60)
                plt.plot(x,my_train)
              8 plt.plot(x,my_test)
              9
                plt.ylim((0.6,1.0))
             10
                plt.show()
```



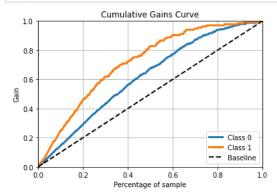
Step:8

In [70]: ▶

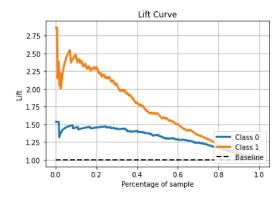
```
1 !pip install scikit-plot
  from scikitplot.estimators import plot_feature_importances
3 from scikitplot.metrics import plot_confusion_matrix, plot_roc
```

Successfully installed scikit-plot-0.3.7

```
Collecting scikit-plot
  Downloading scikit_plot-0.3.7-py3-none-any.whl (33 kB)
Requirement already satisfied: matplotlib>=1.4.0 in c:\users\arulk\anaconda3\lib\site-packages (from scikit-plot) (3.5.1)
Requirement already satisfied: scikit-learn>=0.18 in c:\users\arulk\anaconda3\lib\site-packages (from scikit-plot) (1.0.2)
Requirement already satisfied: scipy>=0.9 in c:\users\arulk\anaconda3\lib\site-packages (from scikit-plot) (1.7.3)
Requirement already satisfied: joblib>=0.10 in c:\users\arulk\anaconda3\lib\site-packages (from scikit-plot) (1.1.0)
Requirement already satisfied: python-dateutil>=2.7 in c:\users\arulk\anaconda3\lib\site-packages (from matplotlib>=1.4.0->s
cikit-plot) (2.8.2)
Requirement already satisfied: pillow>=6.2.0 in c:\users\arulk\anaconda3\lib\site-packages (from matplotlib>=1.4.0->scikit-p
lot) (9.0.1)
Requirement already satisfied: numpy>=1.17 in c:\users\arulk\anaconda3\lib\site-packages (from matplotlib>=1.4.0->scikit-plo
t) (1.21.5)
Requirement already satisfied: packaging>=20.0 in c:\users\arulk\anaconda3\lib\site-packages (from matplotlib>=1.4.0->scikit
-plot) (21.3)
Requirement already satisfied: kiwisolver>=1.0.1 in c:\users\arulk\anaconda3\lib\site-packages (from matplotlib>=1.4.0->scik
it-plot) (1.3.2)
Requirement already satisfied: fonttools>=4.22.0 in c:\users\arulk\anaconda3\lib\site-packages (from matplotlib>=1.4.0->scik
it-plot) (4.25.0)
Requirement already satisfied: pyparsing>=2.2.1 in c:\users\arulk\anaconda3\lib\site-packages (from matplotlib>=1.4.0->sciki
t-plot) (3.0.4)
Requirement already satisfied: cycler>=0.10 in c:\users\arulk\anaconda3\lib\site-packages (from matplotlib>=1.4.0->scikit-pl
ot) (0.11.0)
Requirement already satisfied: six>=1.5 in c: \arconda3 \lib \site-packages (from python-dateutil>=2.7-\mbox{matplotlib} \label{lib}.
>=1.4.0->scikit-plot) (1.16.0)
Requirement already satisfied: threadpoolctl>=2.0.0 in c:\users\arulk\anaconda3\lib\site-packages (from scikit-learn>=0.18->
scikit-plot) (2.2.0)
Installing collected packages: scikit-plot
```



<Figure size 504x504 with 0 Axes>



In []: 🔰 1