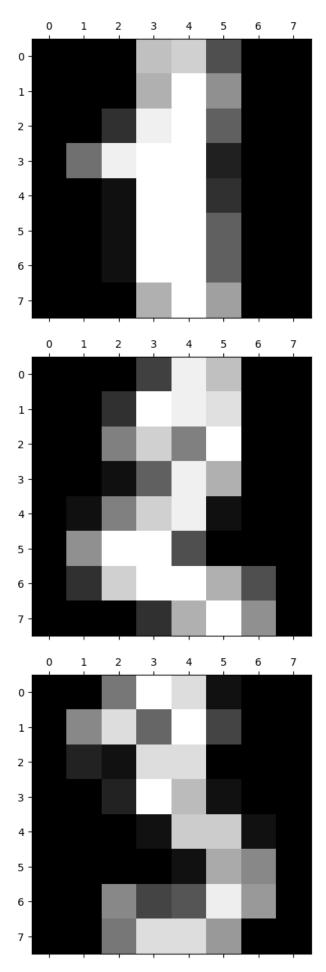
7/21/24, 3:38 PM ML practical 5

```
In [1]: '''
        NAME: MANE SHIVRAJ PANDURANG
        COURSE: AI&DS, SUB:ML(Machine Learning)
        CLASS: BE
Out[1]: '\nNAME: MANE SHIVRAJ PANDURANG\nROLL NO.37\nCOURSE: AI&DS, SUB:ML(Machine Learning)\nCLASS: BE \n'
In [2]: '''
        PRACTICAL.NO:05 (B)
            Use different voting mechanism and Apply AdaBoost (Adaptive Boosting), Gradient Tree
        Boosting (GBM), XGBoost classification on Iris dataset and compare the performance of three
        models using different evaluation measures.
Out[2]: '\nPRACTICAL.NO:05 (B)\n Use different voting mechanism and Apply AdaBoost (Adaptive Boosting), Gradient Tree\nBo
         osting (GBM), XGBoost classification on Iris dataset and compare the performance of three\nmodels using different ev
         aluation measures.\n\n'
In [3]: import pandas as pd
        from sklearn.datasets import load_digits
In [4]: digits = load_digits()
        dir(digits)
        ['DESCR', 'data', 'feature_names', 'frame', 'images', 'target', 'target_names']
        %matplotlib inline
        import matplotlib.pyplot as plt
In [5]: plt.gray()
        for i in range(4):
         plt.matshow(digits.images[i])
       <Figure size 640x480 with 0 Axes>
             0
                                             5
                                                    6
                                                          7
       0 -
       1 -
       2 -
       3 -
       4 -
       5
```

6



7/21/24, 3:38 PM ML practical 5

```
In [6]: df = pd.DataFrame(digits.data)
         df.head()
Out[6]:
                                            7
                                                8
                                                   9 ... 54 55
                                                                 56
                                                                     57
                                                                          58
                                                                               59
                                                                                    60
                                                                                         61 62
                                                                                                 63
         0 0.0 0.0 5.0 13.0
                             9.0
                                  1.0
                                      0.0 0.0 0.0 0.0 ... 0.0 0.0 0.0 0.0 6.0 13.0
                                                                                  10.0
                                                                                        0.0 0.0 0.0
         1 0.0 0.0 0.0 12.0 13.0
                                  2 0.0 0.0 0.0
                                                                              3.0
                                                                                  11.0
                                                                                       16.0
                                                                                            9.0 0.0
         3 0.0 0.0 7.0 15.0 13.0
                                  1.0
                                      0.0 0.0 0.0 8.0 ... 9.0 0.0 0.0 0.0
                                                                         7.0
                                                                             13.0
                                                                                  13.0
                        1.0 11.0
                                  2.0 16.0
         4 0.0 0.0 0.0
                                                                                        4.0 0.0 0.0
        5 rows × 64 columns
In [7]: df['target'] = digits.target
         df[0:12]
                       2
                                      5
                                              7
                                                             55 56 57
                                                                          58
                                                                               59
                                                                                    60
                                                                                         61
                                                                                            62
                                                                                                63 target
          0.0 0.0
                      5.0 13.0
                               9.0
                                    1.0
                                         0.0 0.0 0.0 0.0 ... 0.0 0.0 0.0
                                                                         6.0
                                                                             13.0 10.0
                                                                                        0.0
                                                                                           0.0 0.0
                                                                                                        0
          1 0.0 0.0
                     0.0 12.0
                              13.0
                                    5.0
                                         0.0 0.0
                                                0.0
                                                    0.0
                                                        ... 0.0 0.0 0.0
                                                                         0.0
                                                                             11.0
                                                                                  16.0
                                                                                       10.0
                                                                                           0.0
                                                                                               0.0
          2 0.0 0.0
                                         0.0 \quad 0.0 \quad 0.0 \quad 0.0
                                                                         0.0
                                                                                                         2
                     0.0
                          4.0
                              15.0 12.0
                                                        ... 0.0 0.0 0.0
                                                                              3.0 11.0
                                                                                       16.0
                                                                                            9.0 0.0
                     7.0 15.0
                              13.0
                                         0.0 0.0
                                                 0.0
                                                    8.0
                                                        ... 0.0 0.0 0.0
                                                                             13.0
          3 0.0 0.0
                                    1.0
                                                                         7.0
                                                                                  13.0
                                                                                        9.0
                                                                                           0.0
          4 0.0 0.0
                     0.0
                          1.0
                              11.0
                                    0.0
                                         0.0 0.0
                                                0.0 0.0 ... 0.0 0.0 0.0
                                                                         0.0
                                                                              2.0
                                                                                  16.0
                                                                                        4.0
                                                                                           0.0 0.0
                                                                                                        4
          5 0.0 0.0
                    12.0 10.0
                               0.0
                                    0.0
                                         0.0 0.0
                                                0.0
                                                    0.0 ... 0.0 0.0 0.0
                                                                         9.0
                                                                             16.0
                                                                                  16.0
                                                                                       10.0
                                                                                           0.0
                                                                                               0.0
          6 0.0 0.0
                     0.0 12.0
                              13.0
                                    0.0
                                         0.0 0.0 0.0
                                                     0.0
                                                         ... 0.0 0.0 0.0
                                                                         1.0
                                                                              9.0
                                                                                  15.0
                                                                                       11.0
                                                                                           3.0 0.0
                                                                                                        6
          7 0.0 0.0
                                        15.0 1.0 0.0 0.0 ... 0.0 0.0 0.0
                     7.0
                          8.0 13.0 16.0
                                                                        13.0
                                                                              5.0
                                                                                   0.0
                                                                                        0.0 0.0 0.0
          8 0.0 0.0
                      9.0
                         14.0
                               8.0
                                    1.0
                                         0.0
                                             0.0
                                                 0.0
                                                     0.0
                                                         ... 0.0 0.0 0.0
                                                                        11.0
                                                                             16.0
                                                                                  15.0
                                                                                       11.0
                                                                                                         8
                                                                                            1.0 0.0
          9 0.0 0.0 11.0 12.0
                               0.0
                                    0.0
                                         0.0 0.0 0.0 2.0
                                                        ... 0.0 0.0 0.0
                                                                         9.0
                                                                             12.0
                                                                                  13.0
                                                                                        3.0
                                                                                           0.0 0.0
                                                                                                        9
         10 0.0 0.0
                     1.0
                          9.0
                              15.0 11.0
                                         0.0 0.0 0.0 0.0 ... 0.0 0.0 0.0
                                                                         1.0
                                                                             10.0 13.0
                                                                                        3.0
                                                                                           0.0 0.0
                                                                                                        0
         11 0.0 0.0
                     0.0
                          0.0 14.0 13.0
                                         1.0 0.0 0.0 0.0 ... 0.0 0.0 0.0
                                                                         0.0
                                                                              1.0 13.0 16.0 1.0 0.0
        12 rows × 65 columns
In [8]: # Train and the model and prediction
         X = df.drop('target',axis='columns')
         y = df.target
         from sklearn.model_selection import train_test_split
         X_train, X_test, y_train, y_test = train_test_split(X,y,test_size=0.2)
         from sklearn.ensemble import RandomForestClassifier
         model = RandomForestClassifier(n_estimators=20)
         model.fit(X_train, y_train)
         RandomForestClassifier(n_estimators=20)
         model.score(X_test, y_test)
Out[8]: 0.94722222222222
In [9]: y_predicted = model.predict(X_test)
In [10]: # Confusion Matrix
         from sklearn.metrics import confusion_matrix
         cm = confusion_matrix(y_test, y_predicted)
         cm
```

7/21/24, 3:38 PM ML practical 5

```
Out[10]: array([[35, 0, 0, 0, 0,
                                  0, 0,
                                  0,
                                      0,
                                         0,
                                             0,
                                                0],
               [ 0, 37, 0, 0, 0,
                0,
                   0, 36, 1, 0,
                                  0,
                                      0,
                                         1,
                                             0,
                                                 1],
                       0, 42, 0,
                0,
                    0,
                                  0,
                                      0,
                                         0,
                                             0,
                                                 1],
                0,
                                     0,
                    0,
                       0, 0, 35, 0,
                                         0, 1,
                                                 0],
                0,
                    0,
                        0, 0, 0, 34, 1,
                                         0, 1,
                                                 0],
                    0, 0, 0, 0,
                                 0, 28, 0, 0,
                                                0],
                2,
                              0,
                                                0],
                0,
                   0, 0, 0,
                                  0,
                                      0, 35, 0,
               [ 0,
                    2,
                       1, 1,
                               0, 1,
                                      0, 1, 21, 1],
               [ 0,
                    0, 0, 0, 0, 2, 0, 0, 1, 38]], dtype=int64)
In [11]: %matplotlib inline
        import matplotlib.pyplot as plt
        import seaborn as sn
        plt.figure(figsize=(10,7))
        sn.heatmap(cm, annot=True)
        plt.xlabel('Predicted')
        plt.ylabel('Truth')
```

Out[11]: Text(95.722222222221, 0.5, 'Truth')

