```
In [1]:
        import pandas as pd
        import numpy as np
        from sklearn.datasets import load digits
In [2]: digits=load_digits()
In [3]: digits
Out[3]: {'data': array([[ 0., 0., 5., ..., 0., 0., 0.],
                [0., 0., 0., \dots, 10., 0., 0.],
                [0., 0., 0., \ldots, 16., 9., 0.],
                [0., 0., 1., \ldots, 6., 0., 0.],
                [ 0., 0., 2., ..., 12., 0., 0.],
                [0., 0., 10., \ldots, 12., 1., 0.]
         'target': array([0, 1, 2, ..., 8, 9, 8]),
         'frame': None,
         'feature names': ['pixel 0 0',
           'pixel_0_1',
          'pixel 0 2',
           'pixel 0 3',
           'pixel 0 4',
           'pixel_0_5',
           'pixel 0 6',
           'pixel 0 7',
           'pixel_1_0',
          'pixel_1_1',
          'n + v ~ 1 1 2
In [4]: digits.target
Out[4]: array([0, 1, 2, ..., 8, 9, 8])
In [5]: digits.target names
Out[5]: array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
In [6]: from sklearn.datasets import load digits
```

In [7]: | for image, label in zip(digits.data[0:5], digits.target[0:5]):

```
print(image)
               print(label)
                              9.
                                  1.
                                       0.
                                           0.
                                                0.
                                                    0. 13. 15. 10. 15.
                                                                           5.
                                                                               0.
                                                                                   0.
                                                                                        3.
          [ 0.
                     5. 13.
           15.
                     0.11.
                              8.
                                  0.
                                       0.
                                           4. 12.
                                                    0.
                                                         0.
                                                             8.
                                                                  8.
                                                                      0.
                                                                           0.
                                                                               5.
                                                                                   8.
                                                                                        0.
                                                         7.
                          0.
                              0.
                                  4. 11.
                                           0.
                                                1. 12.
                                                             0.
                                                                  0.
                                                                      2. 14.
                                                                               5. 10. 12.
            0.
                 9.
                     8.
                              6. 13. 10.
            0.
                 0.
                     0.
                          0.
                                           0.
                                                0.
                                                    0.1
                                  5.
                                       0.
          [ 0.
                 0.
                     0. 12. 13.
                                           0.
                                                0.
                                                    0. 0. 11. 16.
                                                                      9.
                                                                               0.
                                                                                        0.
                                                                           0.
                                                                                   0.
            3. 15. 16.
                          6.
                              0.
                                  0.
                                       0.
                                           7. 15. 16. 16.
                                                             2.
                                                                  0.
                                                                      0.
                                                                           0.
                                                                               0.
                                                                                   1. 16.
                     0.
                          0.
                              0.
                                  0.
                                       1. 16. 16.
                                                    6. 0.
                                                             0.
                                                                  0.
                                                                      0.
                                                                           1. 16. 16.
                              0. 11. 16. 10.
            0.
                 0.
                     0.
                          0.
          [ 0.
                 0.
                         4. 15. 12.
                                       0.
                                           0.
                                                0.
                                                    0.
                                                        3. 16. 15. 14.
                                                                                        0.
                     0.
                                                                           0.
            8. 13.
                     8. 16.
                              0.
                                  0.
                                       0.
                                           0.
                                                1.
                                                    6. 15. 11.
                                                                  0.
                                                                      0.
                                                                           0.
                                                                               1.
                                                                                   8. 13.
                                  9. 16. 16.
                                                5.
                                                        0.
                                                             0.
                 1.
                     0.
                          0.
                              0.
                                                    0.
                                                                  0.
                                                                      3. 13. 16. 16. 11.
            5.
                     0.
                          0.
                              0.
                                   3. 11. 16.
                                                9.
                                                    0.1
                    7. 15. 13.
          [ 0.
                 0.
                                  1.
                                       0.
                                           0.
                                                0.
                                                    8. 13.
                                                             6. 15.
                                                                                        2.
                                                                      4.
                                                                           0.
            1. 13. 13.
                          0.
                              0.
                                  0.
                                       0.
                                           0.
                                                2. 15. 11.
                                                             1.
                                                                  0.
                                                                      0.
                                                                               0.
                                                                                   0.
                                                                                        1.
                                                                           0.
                     1.
                          0.
                              0.
                                  0.
                                       0.
                                           0.
                                                1. 10. 8.
                                                             0.
                                                                  0.
                                                                      0.
                                                                           8.
                                                                                    5. 14.
           12. 12.
                              7. 13. 13.
                                           9.
            9.
                 0.
                     0.
                          0.
                                                0.
                                                    0.1
 In [8]: | for i,j in enumerate(['ele1','ele2','ele3']):
              print(i)
               print(j)
          0
          ele1
          ele2
          2
          ele3
 In [9]:
          #Display some of the images and those labels
          import matplotlib.pyplot as plt
In [10]:
          plt.figure(figsize=(20,4))
          for index,(image,label) in enumerate(zip(digits.data[100:105],digits.target[100:1
              plt.subplot(1,5,index+1)
               plt.imshow(np.reshape(image,(8,8)),cmap=plt.cm.gray)
              plt.title(f"Training: {label}",fontsize=20)
               Training: 4
                                 Training: 0
                                                   Training: 5
                                                                     Training: 3
                                                                                       Training: 6
```

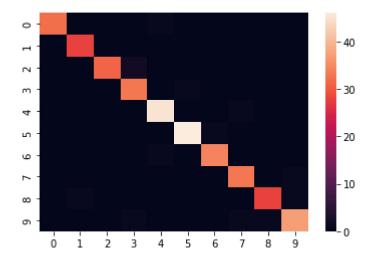
```
In [11]: X=digits.data
         y=digits.target
In [12]: X
Out[12]: array([0, 1, 2, ..., 8, 9, 8])
In [13]: X.shape
Out[13]: (1797, 64)
In [14]: | y.shape
Out[14]: (1797,)
In [15]: from sklearn.model_selection import train_test_split
In [16]: X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.2,random_state=42)
In [17]: X train
Out[17]: array([[ 0., 0., 3., ..., 13., 4.,
                                               0.],
                [ 0., 0., 9., ..., 3., 0.,
                                               0.],
                       0., 0., ...,
                [ 0.,
                                      6.,
                [ 0., 0., 9., ..., 16., 2.,
                                               0.],
                [ 0., 0., 1., ..., 0., 0.,
                                               0.],
                [0., 0., 1., ..., 1., 0.,
In [18]: y_train
Out[18]: array([6, 0, 0, ..., 2, 7, 1])
In [19]: | from sklearn.preprocessing import StandardScaler
In [20]: | sc = StandardScaler()
         X_train = sc.fit_transform(X_train)
         X test = sc.transform(X test)
In [21]: X_train.shape
Out[21]: (1437, 64)
In [22]: from sklearn.discriminant_analysis import LinearDiscriminantAnalysis
In [28]:
         lda=LinearDiscriminantAnalysis(n_components=9)
         X_train=lda.fit_transform(X_train,y_train)
         X_test=lda.transform(X_test)
```

```
In [29]: lda.explained variance ratio
Out[29]: array([0.27851663, 0.19023768, 0.17379022, 0.11077975, 0.08773235,
                0.0654018 , 0.04249174, 0.02968284, 0.02136698])
In [30]: |X_train.shape
Out[30]: (1437, 9)
In [31]: X train[0]
Out[31]: array([ 3.58237017, -1.82036205, -3.00354787, 2.13332268, 0.26151916,
                -1.5496613 , -0.19386593, -0.50007375, 0.40930703])
In [32]: #model building
In [33]: | from sklearn.ensemble import RandomForestClassifier
In [35]: rf1=RandomForestClassifier(n estimators=100,random state=42)
In [36]: rf1.fit(X train,y train)
Out[36]: RandomForestClassifier(random_state=42)
In [39]: y pred1=rf1.predict(X test)
In [40]: y_pred1
Out[40]: array([6, 9, 3, 7, 2, 1, 5, 2, 5, 3, 1, 9, 4, 0, 4, 2, 3, 7, 8, 8, 4, 3,
                9, 7, 5, 6, 3, 5, 6, 3, 4, 9, 1, 4, 4, 6, 9, 4, 7, 6, 6, 9, 1, 3,
                6, 1, 3, 0, 6, 5, 5, 1, 9, 5, 6, 0, 9, 0, 0, 1, 0, 4, 5, 2, 4, 5,
                7, 0, 7, 5, 9, 5, 5, 4, 7, 0, 7, 5, 5, 9, 9, 0, 2, 3, 8, 0, 6, 4,
                4, 9, 1, 2, 8, 3, 5, 2, 9, 4, 4, 4, 4, 3, 5, 3, 1, 3, 5, 9, 4, 2,
                7, 7, 4, 4, 1, 9, 2, 7, 8, 7, 2, 6, 9, 4, 0, 7, 2, 7, 5, 8, 7, 5,
                7, 9, 0, 6, 6, 4, 2, 8, 0, 9, 4, 6, 9, 9, 6, 9, 0, 5, 5, 6, 6, 0,
                6, 4, 3, 9, 3, 7, 7, 2, 9, 0, 4, 5, 3, 6, 5, 9, 9, 8, 4, 2, 1, 3,
                7, 7, 2, 2, 3, 9, 8, 0, 3, 2, 3, 5, 6, 9, 9, 4, 1, 5, 4, 2, 3, 6,
                4, 8, 5, 9, 5, 7, 8, 9, 4, 8, 1, 5, 4, 4, 9, 6, 1, 8, 6, 0, 4, 5,
                2, 7, 4, 6, 4, 5, 6, 0, 3, 2, 3, 6, 7, 1, 5, 1, 4, 7, 6, 9, 1, 5,
                5, 1, 4, 2, 8, 8, 9, 8, 7, 6, 2, 2, 2, 3, 4, 8, 8, 3, 6, 0, 3, 7,
                7, 0, 1, 0, 4, 5, 1, 5, 3, 6, 0, 4, 1, 0, 0, 3, 6, 5, 9, 7, 3, 5,
                5, 9, 9, 8, 5, 3, 3, 2, 0, 5, 8, 3, 4, 0, 2, 4, 6, 4, 3, 4, 5, 0,
                5, 2, 1, 3, 1, 4, 1, 1, 7, 0, 1, 5, 2, 1, 2, 8, 7, 0, 6, 4, 8, 8,
                5, 1, 8, 4, 5, 8, 7, 9, 8, 6, 0, 6, 2, 0, 7, 9, 8, 9, 5, 2, 7, 7,
                1, 8, 7, 4, 3, 8, 3, 5])
```

```
In [41]: y_test
Out[41]: array([6, 9, 3, 7, 2, 1, 5, 2, 5, 2, 1, 9, 4, 0, 4, 2, 3, 7, 8, 8, 4, 3,
                 9, 7, 5, 6, 3, 5, 6, 3, 4, 9, 1, 4, 4, 6, 9, 4, 7, 6, 6, 9, 1, 3,
                 6, 1, 3, 0, 6, 5, 5, 1, 9, 5, 6, 0, 9, 0, 0, 1, 0, 4, 5, 2, 4, 5,
                 7, 0, 7, 5, 9, 5, 5, 4, 7, 0, 4, 5, 5, 9, 9, 0, 2, 3, 8, 0, 6, 4,
                 4, 9, 1, 2, 8, 3, 5, 2, 9, 0, 4, 4, 4, 3, 5, 3, 1, 3, 5, 9, 4, 2,
                 7, 7, 4, 4, 1, 9, 2, 7, 8, 7, 2, 6, 9, 4, 0, 7, 2, 7, 5, 8, 7, 5,
                 7, 7, 0, 6, 6, 4, 2, 8, 0, 9, 4, 6, 9, 9, 6, 9, 0, 3, 5, 6, 6, 0,
                 6, 4, 3, 9, 3, 9, 7, 2, 9, 0, 4, 5, 3, 6, 5, 9, 9, 8, 4, 2, 1, 3,
                 7, 7, 2, 2, 3, 9, 8, 0, 3, 2, 2, 5, 6, 9, 9, 4, 1, 5, 4, 2, 3, 6,
                 4, 8, 5, 9, 5, 7, 8, 9, 4, 8, 1, 5, 4, 4, 9, 6, 1, 8, 6, 0, 4, 5,
                 2, 7, 4, 6, 4, 5, 6, 0, 3, 2, 3, 6, 7, 1, 5, 1, 4, 7, 6, 8, 8, 5,
                 5, 1, 6, 2, 8, 8, 9, 9, 7, 6, 2, 2, 2, 3, 4, 8, 8, 3, 6, 0, 9, 7,
                 7, 0, 1, 0, 4, 5, 1, 5, 3, 6, 0, 4, 1, 0, 0, 3, 6, 5, 9, 7, 3, 5,
                 5, 9, 9, 8, 5, 3, 3, 2, 0, 5, 8, 3, 4, 0, 2, 4, 6, 4, 3, 4, 5, 0,
                 5, 2, 1, 3, 1, 4, 1, 1, 7, 0, 1, 5, 2, 1, 2, 8, 7, 0, 6, 4, 8, 8,
                 5, 1, 8, 4, 5, 8, 7, 9, 8, 5, 0, 6, 2, 0, 7, 9, 8, 9, 5, 2, 7, 7,
                1, 8, 7, 4, 3, 8, 3, 5])
In [42]: #accuracy score
In [48]: from sklearn.metrics import accuracy score, confusion matrix, classification report
In [46]:
         accuracy_score(y_test,y_pred1)
Out[46]: 0.963888888888888
In [52]: y=confusion matrix(y test,y pred1)
In [50]: c=classification report(y test,y pred1)
In [51]: print(c)
                        precision
                                     recall f1-score
                                                         support
                    0
                             1.00
                                       0.97
                                                 0.98
                                                              33
                     1
                             0.97
                                       1.00
                                                 0.98
                                                              28
                     2
                             1.00
                                       0.94
                                                 0.97
                                                              33
                     3
                             0.92
                                       0.97
                                                 0.94
                                                              34
                     4
                             0.96
                                       0.98
                                                 0.97
                                                              46
                     5
                             0.98
                                                 0.98
                                                              47
                                       0.98
                     6
                             0.97
                                       0.97
                                                 0.97
                                                              35
                     7
                             0.94
                                       0.97
                                                 0.96
                                                              34
                     8
                             0.97
                                       0.93
                                                 0.95
                                                              30
                     9
                             0.95
                                       0.93
                                                 0.94
                                                              40
                                                 0.96
                                                             360
             accuracy
                             0.96
                                       0.96
                                                 0.96
                                                             360
            macro avg
         weighted avg
                             0.96
                                       0.96
                                                 0.96
                                                             360
```

```
In [53]: import seaborn as sns
sns.heatmap(y)
```

Out[53]: <AxesSubplot:>



```
In [54]: def get_misclassified_index(y_pred,y_test):
    misclassification=[]#help us out to get the misclassified index value
    for index,(predicted,actual) in enumerate(zip(y_pred,y_test)):
        if predicted!=actual:
            misclassification.append(index)

    return misclassification
```

```
In [57]: misclassification = get_misclassified_index(y_pred1,y_test)
```

In [58]: misclassification

Out[58]: [9, 76, 97, 133, 149, 159, 186, 239, 240, 244, 249, 262, 339]

In []: