```
In [ ]:
          # Features (X)
          X = digits.data
          digits.data.shape # 1797 pictures of size 64 == 8x8
         (1797, 64)
Out[]:
In [ ]:
          # Image at first index
          X[0]
         array([ 0.,
                       0., 5., 13., 9., 1.,
                                                   0.,
                                                        0., 0., 0., 13., 15., 1
Out[]:
         0.,
                 15.,
                       5.,
                             0.,
                                  0., 3., 15.,
                                                   2.,
                                                        0., 11.,
         4.,
                 12.,
                       0.,
                             0., 8.,
                                        8.,
                                             0., 0.,
                                                        5.,
                                                              8.,
                                                                   0.,
                                                                         0.,
                                                                             9.,
         8.,
                 0., 0.,
                                                                         2., 14.,
                            4., 11., 0.,
                                             1., 12., 7., 0.,
                                                                   0.,
         5.,
                 10., 12., 0., 0., 0., 6., 13., 10., 0., 0., 0.])
In [ ]:
          # Labels (y)
          y = digits.target
          digits.target.shape
         (1797,)
Out[]:
In [ ]:
          # First index image label
          y[0]
Out[]:
In [ ]:
          # Plot first 10 digits images
          plt.figure(figsize=(20,4))
          for index, (image, label) in enumerate(zip(digits.data[0:10], digit
              plt.subplot(1,10,index+1)
              plt.imshow(np.reshape(image, (8,8)), cmap=plt.cm.gray)
              plt.title(f'Training: {label}', fontsize=20)
         Training: 0 Training: 1 Training: 2 Training: 3 Training: 4 Training: 5 Training: 6 Training: 7 Training: 8 Training: 9
In [ ]:
          # split the data
          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,
                                                                   random_state=0)
In [ ]:
          print(f'Train input: {X_train.shape}')
          print(f
                   Train labels:
                                  {y_train.shape}
          print(f'Test input: {X_test.shape}')
          print(f'Test labels: {y_test.shape}')
         Train input: (1437, 64)
         Train labels: (1437,)
         Test input: (360, 64)
         Test labels: (360,)
In [ ]:
          # Train model
          from sklearn.linear_model import LogisticRegression
          log_reg = LogisticRegression().fit(X_train, y_train)
          log_reg
         c:\users\awon\miniconda3\lib\site-packages\sklearn\linear_model\_log
         istic.py:814: ConvergenceWarning: lbfgs failed to converge (status=
         1):
         STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
         Increase the number of iterations (max_iter) or scale the data as sh
             https://scikit-learn.org/stable/modules/preprocessing.html
         Please also refer to the documentation for alternative solver option
             https://scikit-learn.org/stable/modules/linear_model.html#logist
         ic-regression
           n_iter_i = _check_optimize_result(
         LogisticRegression()
Out[]:
In [ ]:
          # Make predictions on first 10 test images
          log_reg.predict(X_test[:10])
         array([2, 8, 2, 6, 6, 7, 1, 9, 8, 5])
Out[]:
In [ ]:
          # Make predictions on all test images
          y_pred = log_reg.predict(X_test)
          y_pred[:10] # same labels predicted as above
         array([2, 8, 2, 6, 6, 7, 1, 9, 8, 5])
Out[]:
In [ ]:
          y_test.ndim, y_pred.ndim
         (1, 1)
Out[]:
In [ ]:
          # Accuracy score
          from sklearn.metrics import accuracy_score
          acc_score = accuracy_score(y_test, y_pred)
          print(f'Accuracy score: {acc_score}')
         Accuracy score: 0.966666666666667
In [ ]:
          # Confusion matrix
          from sklearn.metrics import confusion_matrix
          cm = confusion_matrix(y_test, y_pred)
         array([[27,
                       Θ,
                            Θ,
                                Θ,
                                     Θ,
                                         Θ,
                                             Θ,
                                                  Θ,
                                                      Θ,
                                                          0],
Out[ ]:
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                                    Θ,
                                         Θ,
                                             Θ,
                 [ 0,
                      34,
                            Θ,
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                                                          0],
                                    Θ,
                 [ 0,
                       0, 35,
                                1,
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                                             Θ,
                                                  Θ,
                                                      Θ,
                   Θ,
                                                      Θ,
                            0, 29,
                 [
                       Θ,
                                    Θ,
                                         Θ,
                                             Θ,
                                                  Θ,
                                                          0],
                                             Θ,
                            Θ,
                                0, 29,
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                       Θ,
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                                                           3],
                                         0, 43,
                            Θ,
                                    Θ,
                 L
                   Θ,
                       1,
                                Θ,
                                                  Θ,
                                                      Θ,
                                                          0],
                 [
                   Θ,
                       Θ,
                            Θ,
                                Θ,
                                    1,
                                        Θ,
                                             0, 38,
                                                      Θ,
                                                          0],
                                                     36,
                                                      0, 40]], dtype=int64)
                 [ 0,
                            Θ,
                                Θ,
                                    Θ,
                                         1,
                                             Θ,
                                                  Θ,
In [ ]:
          # Plot confusion matrix
          import seaborn as sns
          plt.figure(figsize=(9,9))
          sns.heatmap(cm, annot=True, fmt='.3f', linewidths=.5,
                       square=True, cmap='Spectral')
          plt.xlabel('Predicted Output')
          plt.ylabel('Acutal Output')
          plt.title(f'Accuracy Score: {acc_score:.2f}', size=15);
                               Accuracy Score: 0.97
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                    34.000
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                                                                              - 25
         Acutal Output
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                    0.000
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                      i
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                                                             8
                                                                   9
                                    Predicted Output
In [ ]:
          # Get misclassified index labels
          index = 0
          misclassifiedIndexes = []
          for label, predict in zip(y_test, y_pred):
              if label != predict:
                   misclassifiedIndexes.append(index)
              index += 1
In [ ]:
          # View misclassified index labels
          misclassifiedIndexes
         [56, 84, 94, 118, 124, 130, 181, 196, 235, 315, 331, 335]
Out[]:
In [ ]:
          # Plot misclassified index labels with known labels (y_test)
          plt.figure(figsize=(20,4))
          for plotIndex, badIndex in enumerate(misclassifiedIndexes[0:5]): #
              plt.subplot(1, 5, plotIndex+1)
              plt.imshow(np.reshape(X_test[badIndex], (8,8)), cmap=plt.cm.gray
              plt.title(f'Actual: {y_test[badIndex]}, Predicted: {y_pred[badI]
          Actual: 5, Predicted: 9
                          Actual: 5, Predicted: 9
                                         Actual: 7, Predicted: 4
                                                        Actual: 6, Predicted: 1
                                                                        Actual: 8, Predicted: 1
```

In []:

In []:

Import libraries
import numpy as np
import pandas as pd

Import online data

digits = load_digits()

Load data

import matplotlib.pyplot as plt

from sklearn.datasets import load_digits