```
In [2]: import numpy as np
         import matplotlib.pyplot as plt
         from sklearn import datasets
         from sklearn.model selection import train test split
         from sklearn import svm
         from sklearn import metrics
 In [4]: # Load the digits dataset
         digits = datasets.load_digits()
In [10]: # Split the data into features (X) and labels (y)
         X = digits.data
         y = digits.target
         print(X)
         print(y)
       [[ 0. 0. 5. ... 0. 0. 0.]
         [ 0. 0. 0. ... 10. 0. 0.]
         [ 0. 0. 0. ... 16. 9. 0.]
         . . .
         [ 0. 0. 1. ... 6. 0. 0.]
         [0. 0. 2. ... 12. 0. 0.]
         [ 0. 0. 10. ... 12. 1. 0.]]
       [0 1 2 ... 8 9 8]
In [12]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_sta
In [14]: # Create an SVM classifier (linear kernel)
         clf = svm.SVC(kernel='linear')
In [16]: # Fit the classifier on the training data
         clf.fit(X_train, y_train)
Out[16]: -
                SVC
         SVC(kernel='linear')
In [18]: # Predict on the test data
         y_pred = clf.predict(X_test)
In [20]: # Calculate accuracy
         accuracy = metrics.accuracy_score(y_test, y_pred)
         print("Accuracy : ", accuracy)
       In [22]: # Confusion matrix
         confusion_matrix = metrics.confusion_matrix(y_test, y_pred)
         print("Confusion Matrix : ")
         print(confusion matrix)
```

```
Confusion Matrix :
                                 0
                                    0
                                           0]
                             0
          [ 0 28
                   0
                      0
                          0
                                 0
                                    0
                                       0
                                           01
               0 33
                      0
                          0
                             0
                                 0
                                    0
                                       0
                                           0]
               0
                   0 32
                          0
                             1
                                0
                                    0
                                       0
                                           1]
               1
                   0
                      0 45
                             0
                                0
                                    0
                                       0
                                           0]
               0
                   0
                      0
                          0 47
                                0
                                    0
                                       0
                                           01
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               0
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                      0
                          0
                             0 35
                                    0
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                          0
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                                0 33
               0
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                          0
                             1
                                0
                                    0 29
                                0
                                   1 0 37]]
               0
                   0
                      1
                          1
                             0
In [24]: # Classification report
          classification_report = metrics.classification_report(y_test, y_pred)
          print("Classification Report : ")
          print(classification_report)
         Classification Report :
                         precision
                                       recall f1-score
                                                             support
                                          1.00
                     0
                              1.00
                                                     1.00
                                                                   33
                                                     0.98
                     1
                              0.97
                                          1.00
                                                                   28
                     2
                              1.00
                                          1.00
                                                     1.00
                                                                   33
                                                                   34
                     3
                              0.97
                                          0.94
                                                     0.96
                     4
                              0.98
                                          0.98
                                                     0.98
                                                                   46
                     5
                              0.96
                                          1.00
                                                     0.98
                                                                   47
                     6
                              1.00
                                          1.00
                                                     1.00
                                                                   35
                     7
                              0.97
                                          0.97
                                                     0.97
                                                                   34
                     8
                              1.00
                                          0.97
                                                     0.98
                                                                   30
                     9
                              0.95
                                          0.93
                                                     0.94
                                                                   40
             accuracy
                                                     0.98
                                                                  360
            macro avg
                              0.98
                                          0.98
                                                     0.98
                                                                  360
         weighted avg
                              0.98
                                          0.98
                                                     0.98
                                                                  360
In [26]: # Visualize some of the test images and their predicted labels
          plt.figure(figsize=(15, 8))
          for i in range(10):
               plt.subplot(5, 5, i + 1)
               plt.imshow(X_test[i].reshape(8, 8), cmap=plt.cm.gray_r)
               plt.title(f"Predicted : {y_pred[i]}, Actual : {y_test[i]}")
               plt.axis('on')
        Predicted: 6, Actual: 6
                            Predicted: 9, Actual: 9
                                                Predicted: 3, Actual: 3
                                                                    Predicted: 7, Actual: 7
                                                                                        Predicted: 2, Actual: 2
                            Predicted: 5, Actual: 5
        Predicted: 1, Actual: 1
                                                                                                   ctual : 2
                                                Predicted: 2,
                                                          Actual : 2
                                                                    Predicted: 5, Actual: 5
                                                                                        Predicte
 In [ ]:
```