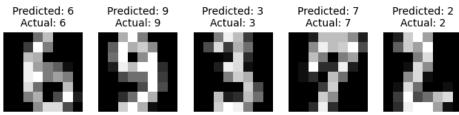
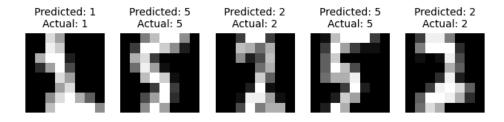
```
In [ ]: '''
       NAME: MANE SHIVRAJ PANDURANG
       ROLL NO.37
       COURSE: AI&DS, SUB:ML(Machine Learning)
In [ ]: '''
       PRACTICAL NO:03
       A. Implementation of Support Vector Machines (SVM) for classifying images of
       handwritten digits into their respective numerical classes (0 to 9).''
In [6]: import pandas as pd
       import numpy as np
       import matplotlib.pyplot as plt
       from sklearn import datasets
       from sklearn.model_selection import train_test_split
       from sklearn.svm import SVC
       from sklearn.metrics import accuracy_score, classification_report, confusion_matrix
In [2]: digits = datasets.load_digits()
       X = digits.data # Features (pixel values)
       y = digits.target # Target labels (0 to 9)
       X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
       svm_model = SVC(kernel='linear', C=1.0)
        svm_model.fit(X_train, y_train)
Out[2]: ▼
               SVC
       SVC(kernel='linear')
In [3]: y_pred = svm_model.predict(X_test)
        accuracy = accuracy_score(y_test, y_pred)
       confusion_mat = confusion_matrix(y_test, y_pred)
       classification_rep = classification_report(y_test, y_pred)
       print("Accuracy:", accuracy)
       print("Confusion Matrix:\n", confusion_mat)
       print("Classification Report:\n", classification_rep)
      Accuracy: 0.977777777777777
      Confusion Matrix:
       [[33 0 0 0 0 0 0 0 0]
       [02800000000]
       [003300000000]
       [00032010001]
       [01004500000]
       [00000470000]
       [00000035000]
       [00000003301]
       [00000100290]
       [0001100137]]
      Classification Report:
                              recall f1-score support
                    precision
                a
                       1.00
                                1.00
                                         1.00
                                                    33
                1
                       0.97
                                1.00
                                         0.98
                                                    28
                2
                       1.00
                                1.00
                                         1.00
                                                    33
                       0.97
                                0.94
                                         0.96
                3
                                                    34
                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
                       1.00
                                0.97
                                         0.98
                                                    30
                8
                9
                       0.95
                                0.93
                                         0.94
                                                    40
          accuracy
                                         0.98
                                                   360
                       0.98
                                0.98
         macro avg
                                         0.98
                                                   360
      weighted avg
                       0.98
                                0.98
                                         0.98
                                                   360
```

```
In [4]: plt.figure(figsize=(8, 6))
for i in range(10):
    plt.subplot(2, 5, i + 1)
    plt.imshow(X_test[i].reshape(8, 8), cmap='gray')
    plt.title(f'Predicted: {y_pred[i]}\nActual: {y_test[i]}', fontsize=10)
    plt.axis('off')
plt.show()

Predicted: 6  Predicted: 9  Predicted: 3  Predicted: 7  Predicted: 2
```





In []: