Start coding or generate with AI.

Title of Project

Handwritten Digit Prediction - Classification Analysis

Objective

The objective of this project is to build a machine learning model that can accurately predict handwritten digits.

Data Source

The dataset used for this project is the "digits" dataset from scikit-learn's built-in datasets. It contains images of handwritten digits, where each image is an 8x8 pixel matrix, and the target variable represents the actual digit value (0-9).

Import Library

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
```

Import Data

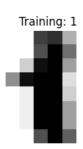
from sklearn.datasets import load_digits

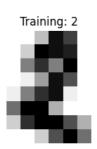
Data Preprocessing

Flatten Image

```
df = load_digits()
_, axes = plt.subplots(nrows = 1, ncols = 4, figsize = (10, 3))
for ax, image, label in zip(axes, df.images, df.target):
    ax.set_axis_off()
    ax.imshow(image, cmap=plt.cm.gray_r, interpolation = 'nearest')
    ax.set_title('Training: %i' % label)
```









df.images.shape

```
→ (1797, 8, 8)
```

df.images[0]

```
df.images[0].shape
```

```
→ (8, 8)
```

```
n_samples = len(df.images)
 data = df.images.reshape((n_samples, -1))
  data[0]
⇒ array([0., 0., 5., 13., 9., 1., 0., 0., 0., 0., 13., 15., 10., 15., 5., 0., 0., 3., 15., 2., 0., 11., 8., 0., 0., 4.,
                              12., 0., 0., 8., 8., 0., 0., 5., 8., 0., 0., 9., 8., 0., 0., 4., 11., 0., 1., 12., 7., 0., 0., 2., 14., 5.,
                              10., 12., 0., 0., 0., 6., 13., 10., 0., 0., 0.])
  data[0].shape
<del>→</del> (64,)
 data.shape
→ (1797, 64)
Scaling Data
 data.min()
 <del>→</del> 0.0
  data.max()
 → 1.0
data = data/16
data.min()
→ 0.0
data.max()
→ 0.0625
 data[0]
                              [0. , 0. , 0.01953125, 0.05078125, 0.03515625, 0.00390625, 0. , 0. , 0. , 0. , 0.
\rightarrow array([0.
                              0. , 0.
                                                                                    , 0.01171875, 0.05859375, 0.0078125 ,
                                                       , 0.04296875, 0.03125 , 0. , 0.
                              0.015625 , 0.046875 , 0. , 0.
                                                                                                                                                   , 0.03125
                                                                        , 0.
                               0.03125 , 0.
                                                                                                                    , 0.01953125, 0.03125
                              0. , 0. , 0.03515625, 0.03125 , 0.

0. , 0.015625 , 0.04296875, 0. , 0.00390625 , 0.046875 , 0.02734375, 0. , 0. , 0.046875 , 0.01953125 , 0.0390625 , 0.046875 , 0.01953125 , 0.0390625 , 0.046875 , 0.01953125 , 0.0390625 , 0.046875 , 0.01953125 , 0.0390625 , 0.046875 , 0.01953125 , 0.0390625 , 0.046875 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 , 0.01953125 
                              0. , 0. , 0. , 0.0234375 , 0.05078125, 0.0390625 , 0. , 0. , 0. ])
Train test split Data
  from \ sklearn.model\_selection \ import \ train\_test\_split
  xtrain, xtest, ytrain, ytest = train_test_split(data, df.target, test_size = 0.3)
  xtrain.shape, xtest.shape, ytrain.shape, ytest.shape
T ((1257, 64), (540, 64), (1257,), (540,))
Random Forest model
 from sklearn.ensemble import RandomForestClassifier
 rf = RandomForestClassifier()
rf.fit(xtrain, ytrain)
               ▼ RandomForestClassifier ① ?
             RandomForestClassifier()
```

Predict Test Data

```
y_pred = rf.predict(xtest)
y_pred
\Rightarrow array([0, 4, 4, 1, 6, 3, 0, 3, 9, 7, 0, 1, 4, 8, 3, 9, 3, 2, 0, 9, 5, 8,
            9, 6, 7, 8, 7, 0, 0, 7, 4, 9, 0, 1, 2, 3, 5, 4, 8, 7, 2, 9, 0, 8, 6, 3, 7, 7, 6, 7, 6, 6, 3, 5, 7, 7, 1, 3,
               4, 5, 3, 6, 3, 1, 8, 4, 2, 2, 8, 9, 9, 5, 3,
                                                                2, 2, 2, 5, 9,
            8, 3, 6, 5, 9, 8, 0, 4, 4, 1, 0, 0, 7, 5, 4, 0, 9, 6, 0, 4, 3,
            2, 4, 7, 3, 9, 0, 3, 0, 3, 4, 0, 6, 3, 6, 5, 5, 8, 1, 5,
            7, 8, 9, 2, 4, 3, 6, 9, 1, 7, 3, 1, 5, 8, 1, 0, 1, 8, 2, 5, 6, 9,
            3, 9, 6, 1, 1, 7, 4, 1, 2, 1, 5, 9, 0, 6, 1, 2, 3, 4, 7, 3, 1,
            6, 0, 6, 0, 2, 3, 2, 7, 8, 0, 6, 8, 2, 0, 6, 8, 0, 6, 4, 3, 6, 4,
            1, 1, 9, 9, 2, 2, 1, 7, 8, 5, 2, 4, 4, 7, 1, 3, 2, 9, 1, 8, 9, 3,
               1, 4, 9, 1, 0, 8, 2, 3, 0, 7, 1, 7, 0, 0, 9, 8, 6, 0,
            9, 0, 3, 6, 2, 4, 4, 9, 7, 4, 8, 3, 6, 5, 8, 2, 3, 2, 2, 2, 5, 7,
               3, 8, 6, 0, 0, 8, 4, 0, 0, 0, 0, 9, 5, 6, 9, 4,
               4, 6, 5, 4, 5, 1, 0, 4, 3, 0, 1, 4, 8, 6, 6, 3, 0, 9, 4, 7,
               0, 7,
                      5, 7, 5, 9, 8, 9, 4, 7, 0, 5, 1, 3, 7,
                                                                4, 4, 6, 8, 1,
               3, 5, 0, 0, 4, 2, 8, 5, 7, 7, 1,
                                                   2, 7, 2, 6, 9, 9, 9, 8, 5,
            4, 1, 2, 3, 5, 3, 9, 5, 1, 0, 3, 8, 3, 0, 0, 5, 1, 5, 3, 8, 3, 6,
            6, 7, 7, 5, 0, 6, 4, 1, 3, 7, 4, 2, 5, 9, 6, 7, 9, 5, 8, 2, 2, 8,
            4, 2, 6, 6, 0, 3, 4, 4, 1, 0, 7, 5, 9, 8, 5, 8, 4, 8, 2,
               6, 9, 1, 0, 6, 6, 3, 2, 0, 2, 1, 0, 5, 8, 2, 4, 2, 9,
            1, 8, 7, 0, 4, 4, 0, 7, 2, 2, 5, 1, 4, 3, 5, 6, 7, 1, 3, 6, 5, 4,
               3, 1, 9, 0, 4, 4, 7, 6, 5, 9, 8, 5, 9, 5, 9, 6, 3, 6, 3, 1, 0,
               5, 9, 0, 0, 9, 8, 5, 3, 3, 7, 7, 1, 7, 5, 5, 3, 4, 9, 2, 8, 1, 3, 2, 5, 1, 8, 7, 7, 1, 8, 7, 8, 0, 5, 9, 7, 7, 0, 0, 8, 9, 9,
            3, 0, 6, 1, 2, 9, 5, 6, 7, 5, 8, 1])
```

Model Accuracy

```
from sklearn.metrics import confusion_matrix, classification_report
confusion_matrix(ytest, y_pred)
```

```
\rightarrow array([[63, 0, 0, 0,
                           0,
                              0, 0, 0,
            0, 48, 0, 0, 0, 1, 0,
                                      0,
                                             0],
            0,
               1, 47,
                       0,
                           0,
                                  0,
            0,
                0, 0, 61, 0,
                              0,
                                  0,
                                      0,
                                             0],
            0,
                0,
                   0,
                       0, 52,
                              0,
                                  0,
                                      1,
                                         0,
                                             0],
                                      0,
                0, 0, 0, 0, 52,
                                  0,
                                             1],
            0,
                0,
                   0,
                       0,
                           0,
                              0, 53,
                                     0,
                                         0,
                                             0],
          [ 0,
                0, 0,
                       0,
                          0,
                              0, 0,51,
                                         0,
                                             0],
                              0,
                          0,
                                  0,
                                     0, 51,
            0,
                   0,
                       0,
                                             01.
                3,
                0, 0,
                       1,
                                 0,
                           0,
                              1,
                                     1,
                                         0, 5211)
```

Model Evaluation

print(classification_report(ytest, y_pred))

_	precision	recall	f1-score	support
0	1.00	1.00	1.00	63
1	0.92	0.98	0.95	49
2	1.00	0.98	0.99	48
3	0.98	1.00	0.99	61
4	1.00	0.98	0.99	53
5	0.96	0.98	0.97	53
6	1.00	1.00	1.00	53
7	0.96	1.00	0.98	51
8	1.00	0.94	0.97	54
9	0.98	0.95	0.96	55
accuracy			0.98	540
macro avg	0.98	0.98	0.98	540
weighted avg	0.98	0.98	0.98	540

Prediction

```
from PIL import Image
import numpy as np

# Load the image and convert it to grayscale
image = Image.open('download.png').convert('L')

# Resize the image to 8x8 pixels (the same size as the dataset images)
image = image.resize((8, 8))
```

```
# Convert the image to a numpy array
new_image = np.array(image)

# Flatten the 8x8 image into a 1D array for prediction
new_image_flattened = new_image.flatten().reshape(1, -1)

# Scale the feature variables using the previously defined scaler
new_image_scaled = scaler.transform(new_image_flattened)

# Make a prediction using the model
predicted_digit = model.predict(new_image_scaled)

# Print the predicted digit
print("Predicted Digit:", predicted_digit[0])
```

Predicted Digit: 4