Hand Written Digit Prediction - Classifiction Analysis

The digits dataset consists of 88 pixel images of digits. The images attribute of the dataset stores 88 arrays of grayscale values for each image. We will use these arrays to visualize the first 4 images. The target attribute of the dataset stores the digit each image represents

▼ Import Library

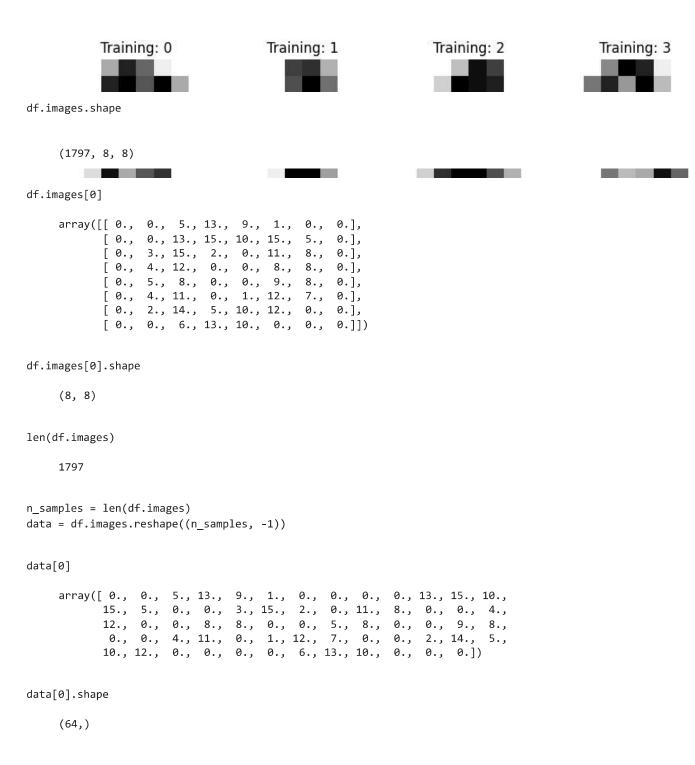
```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plot
```

▼ Import Data

```
from sklearn.datasets import load_digits

df = load_digits()

_, axes = plot.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=plodft.cm.gray_r, interpolation="nearest")
    ax.set_title("Training: %i" % label)
```



```
data.shape
(1797, 64)
```

- Scaling Imaging Data

```
data.min()
    0.0
data.max()
    16.0
data = data/16
data.min()
    0.0
data.max()
    1.0
data[0]
    array([0.
               , 0. , 0.3125, 0.8125, 0.5625, 0.0625, 0. , 0.
             , 0. , 0.8125, 0.9375, 0.625 , 0.9375, 0.3125, 0.
              , 0.1875, 0.9375, 0.125 , 0. , 0.6875, 0.5 , 0.
            , 0.25 , 0.75 , 0. , 0. , 0.5 , 0.5 , 0.
              , 0.3125, 0.5 , 0. , 0. , 0.5625, 0.5 , 0.
              , 0.25 , 0.6875, 0. , 0.0625, 0.75 , 0.4375, 0.
              , 0.125 , 0.875 , 0.3125, 0.625 , 0.75 , 0. , 0.
               , 0. , 0.375 , 0.8125, 0.625 , 0. , 0.
```

→ Train Test Split Data

from sklearn.model_selection import train_test_split

Random Forest Model

Predict Test Data

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

```
9, 1, 8, 7, 0, 9, 2, 4, 8, 4, 5, 2, 2, 2, 1, 9, 8, 7, 0, 3, 9, 5, 5, 7, 9, 6, 1, 3, 7, 8, 1, 4, 4, 4, 7, 1, 7, 6, 0, 3, 5, 5, 7, 7, 9, 6, 6, 7, 3, 1, 7, 7, 6, 8, 5, 5, 6, 6, 1, 0, 1, 8, 1, 7, 7, 6, 4, 8, 1, 5, 4, 6, 9, 8, 5, 2, 7, 8, 9, 6, 6, 3, 3, 1, 3, 9, 1, 9, 2, 9, 0, 5, 2, 1, 3, 3, 8, 6, 1, 8, 4, 0, 6, 5, 6, 9, 5, 6, 9, 3, 2, 3, 7, 7, 6, 9, 9, 3, 2, 1, 1, 2, 5, 2, 2, 1, 4, 0, 0, 5, 7, 1, 6, 3, 6, 5, 0, 0, 9, 4, 9, 2, 4, 5, 0, 5, 4, 9, 2, 6, 3, 9, 1, 4, 3, 0, 1, 1, 3, 8, 0, 9, 6, 4, 5, 6])
```

Model Accuracy

confusion_matrix(y_test, y_pred)

```
from sklearn.metrics import confusion_matrix, classification_report
```

```
array([[42, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0],
        [0, 52, 0, 0, 0, 0, 0, 0, 0, 0, 0],
        [1, 1, 47, 1, 0, 0, 0, 0, 0, 0, 1],
        [0, 0, 0, 63, 0, 0, 0, 2, 1, 0],
        [0, 0, 0, 0, 54, 0, 0, 2, 0, 0],
        [0, 0, 0, 0, 0, 52, 1, 0, 0, 1],
        [0, 0, 0, 0, 0, 0, 52, 1, 0, 0, 0],
        [0, 0, 0, 0, 0, 0, 50, 0, 0],
        [0, 0, 0, 0, 0, 0, 0, 50, 0, 0],
        [0, 0, 0, 0, 1, 0, 1, 46, 0],
        [0, 0, 0, 0, 1, 0, 1, 46, 0],
```

print(classification_report(y_test, y_pred))

	precision	recall	f1-score	support
0	0.98	1.00	0.99	42
1	0.95	1.00	0.97	52
2	1.00	0.92	0.96	51
3	0.97	0.95	0.96	66
4	0.98	0.96	0.97	56
5	0.96	0.96	0.96	54
6	0.98	0.98	0.98	61
7	0.91	1.00	0.95	50
8	0.98	0.92	0.95	50
9	0.97	0.97	0.97	58
accuracy			0.97	540
macro avg	0.97	0.97	0.97	540
weighted avg	0.97	0.97	0.97	540

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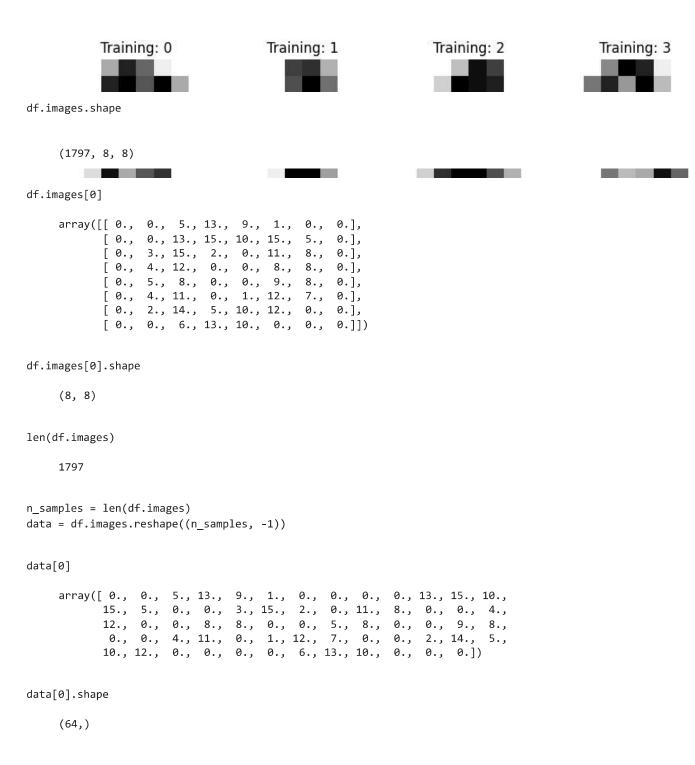
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            0, 1, 3, 9, 5, 3, 4, 0, 2, 7, 2, 5, 1, 2, 4, 9, 0, 7, 7, 8, 9, 0,
            2, 1, 5, 4, 5, 3, 1, 4, 8, 1, 5, 0, 3, 8, 7, 7, 3, 6, 4, 2, 5, 6,
            8, 3, 4, 7, 2, 5, 7, 8, 8, 2, 6, 1, 4, 3, 4, 0, 6, 9, 8, 9, 9, 7,
            8, 9, 4, 0, 3, 6, 4, 0, 6, 6, 4, 9, 3, 4, 8, 2, 9, 4, 9, 3, 3, 3,
            8, 9, 5, 8, 4, 1, 3, 9, 1, 9, 7, 2, 1, 8, 7, 8, 7, 2, 8, 6, 3, 1,
            9, 4, 2, 8, 2, 4, 8, 6, 5, 9, 5, 8, 5, 2, 0, 6, 6, 0, 8, 3, 1, 5,
            3, 4, 2, 9, 7, 1, 4, 1, 2, 5, 9, 0, 5, 6, 4, 6, 1, 0, 7, 3, 3, 8,
            3, 7, 7, 1, 3, 1, 3, 6, 5, 1, 9, 2, 1, 8, 5, 2, 3, 2, 7, 7, 5, 2,
            5, 9, 6, 3, 7, 4, 3, 1, 0, 6, 0, 9, 3, 5, 7, 2, 9, 4, 4, 1, 7, 5,
            1, 3, 7, 0, 4, 1, 5, 0, 3, 0, 3, 4, 2, 9, 0, 9, 3, 8, 8, 7, 8, 6,
            0, 6, 6, 4, 1, 4, 8, 2, 3, 6, 4, 2, 5, 2, 8, 1, 1, 3, 9, 4, 3, 6,
            5, 7, 8, 3, 3, 2, 3, 0, 7, 9, 8, 6, 2, 6, 4, 8, 9, 4, 8, 4, 1, 8,
            4, 7, 2, 6, 3, 5, 7, 3, 8, 9, 4, 4, 6, 5, 6, 5, 0, 6, 7, 3, 0, 9,
            1, 3, 1, 4, 6, 8, 6, 3, 5, 6, 9, 6, 2, 4, 7, 9, 1, 6, 9, 7, 6, 6,
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```

```
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        [0, 0, 0, 0, 0, 52, 1, 0, 0, 1],
        [0, 0, 0, 0, 0, 0, 52, 1, 0, 0, 0],
        [0, 0, 0, 0, 0, 0, 50, 0, 0],
        [0, 0, 0, 0, 0, 0, 0, 50, 0, 0],
        [0, 0, 0, 0, 1, 0, 1, 46, 0],
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