Start coding or generate with AI.

Hand Written Digit Prediction - classification Analysis

Import Library

```
import pandas as pd

import numpy as np

import matplotlib.pyplot as plt

Import Data

from sklearn.datasets import load_digits

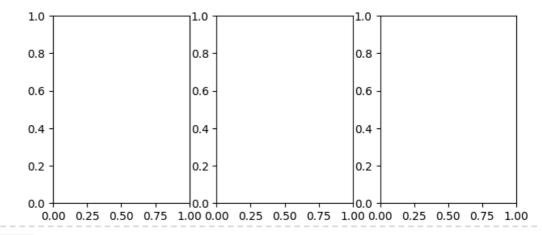
df = load_digits()

_, 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, camp=plt.cm.gray_r, interpolation="nearest")
    ax.set_tittle("Training: %i" %label)
```



```
AttributeError
                                           Traceback (most recent call last)
<ipython-input-14-65822857cd19> in <cell line: 2>()
      2 for ax, image, label in zip(axes,df.images,df.target):
            ax.set_axis_off()
            ax.imshow(image, camp=plt.cm.gray_r, interpolation="nearest")
---> 4
            ax.set_tittle("Training: %i" %label)
                                   🗘 6 frames
/usr/local/lib/python3.10/dist-packages/matplotlib/artist.py in _update_props(self,
props, errfmt)
   1195
                            func = getattr(self, f"set_{k}", None)
   1196
                            if not callable(func):
-> 1197
                                raise AttributeError(
                                    errfmt.format(cls=type(self), prop_name=k))
   1198
                            ret.append(func(v))
   1199
```

AttributeError: AxesImage.set() got an unexpected keyword argument 'camp'



Next steps:

Explain error

Data processing

Flatten Image

```
df.images.shape
→ (1797, 8, 8)
df.images[0]
    array([[ 0., 0., 5., 13., 9., 1., 0.,
                 0., 13., 15., 10., 15., 5.,
           [ 0.,
                3., 15., 2., 0., 11.,
            0.,
                4., 12., 0., 0., 8.,
                                        8.,
                5., 8., 0., 0., 9.,
           [ 0.,
                 4., 11., 0., 1., 12.,
                                        7.,
                 2., 14., 5., 10., 12.,
                                        0.,
                0., 6., 13., 10., 0., 0.,
```

df.images[0].shape

Scaling Image Data

```
0. , 0.1875, 0.9375, 0.125 , 0. , 0.6875, 0.5 , 0. ,
0. , 0.25 , 0.75 , 0. , 0. , 0.5 , 0.5 , 0. ,
0. , 0.3125, 0.5 , 0. , 0. , 0.5625, 0.5 , 0. ,
0. , 0.25 , 0.6875, 0. , 0.0625, 0.75 , 0.4375, 0. ,
0. , 0.125 , 0.875 , 0.3125, 0.625 , 0.75 , 0. , 0. ,
0. , 0. , 0.375 , 0.8125, 0.625 , 0. , 0. , 0. ])
```

Train Test Split Data

```
from sklearn.model_selection import train_test_split

x_train, x_test, y_train, y_test = train_test_split(data, df.target, test_size=0.3)

x_train.shape, x_test.shape, y_train.shape, y_test.shape

$\infty$ ((1257, 64), (540, 64), (1257,), (540,))
```

Random Forest Model

Predict Test Data

y pred = rf.predict(x test)

```
y_pred

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

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

Model Accuracy

from sklearn.metrics import confusion_matrix, classification_report

confusion_matrix(y_test,y_pred)

```
0,
                                         0,
                                                 0],
\rightarrow \overline{\phantom{a}} array([[51, 0,
                              0,
                                  0,
                                     0,
           [ 0, 57, 2,
                              0,
                                  0,
                                     0,
                                                  0],
           Γ0,
                         0,
                             0,
                 0,52,
                                  0,
                                      0,
                                          0,
                                              0,
                                                  0],
           [ 0,
                 0, 0, 43,
                             0,
                                  0,
                                      0,
                                          0,
                                              0,
                                                  0],
                     0,
                                 0,
           [ 0,
                         0, 49,
                                     0,
                                         1,
                 0,
                                                  0],
                         0,
           Γ0,
                 0, 0,
                             0, 51,
                                     1,
                                         0,
                                             0,
           [0, 0, 0, 0, 0,
                                 0, 63,
                                          0, 0,
           [0, 0, 0, 0, 0,
                                  0, 0, 60,
                                              0,
                                                  1],
                     0,
           [0, 3,
                                          0, 50,
                         0,
                              0,
                                  0,
                                     0,
                                                  0],
           [0, 0,
                     0,
                         1,
                             0,
                                  0,
                                     0, 0, 1, 53]])
```

print(classification_report(y_test,y_pred))

→	precision	recall	f1-score	support
0	1.00	1.00	1.00	51
1	0.95	0.97	0.96	59
2	0.96	1.00	0.98	52
3	0.98	1.00	0.99	43
4	1.00	0.98	0.99	50
5	1.00	0.96	0.98	53
6	0.98	1.00	0.99	63
7	0.98	0.98	0.98	61
8	0.98	0.94	0.96	53
9	0.96	0.96	0.96	55
accuracy			0.98	540
macro avg	0.98	0.98	0.98	540