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In [3]: ##HAND-WRITTEN DIGIT PREDICTION
        #import libraries
        import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
In [4]: #import data
         from sklearn.datasets import load_digits
        df=load_digits()
In [6]: _, 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)
                                                             Training: 2
                                                                                     Training: 3
            Training: 0
                                     Training: 1
In [8]: #Data Preprocessing
        df.images.shape
Out[8]: (1797, 8, 8)
In [9]: df.images[0]
Out[9]: 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., 0., 6., 13., 10., 0., 0., 0.]])
In [10]: df.images[0].shape
Out[10]: (8, 8)
In [11]: len(df.images)
Out[11]: 1797
In [12]: n_samples=len(df.images)
         data=df.images.reshape((n_samples, -1))
In [13]: data[0]
Out[13]: 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.]
In [14]: data[0].shape
Out[14]: (64,)
In [15]: data.shape
Out[15]: (1797, 64)
In [16]: #Scaling Image Data
         data.min()
Out[16]: 0.0
In [17]: data.max()
Out[17]: 16.0
In [18]: data=data/16
In [19]: data.min()
Out[19]: 0.0
In [20]: data.max()
Out[20]: 1.0
In [21]: data[0]
Out[21]: array([0. , 0. , 0.3125, 0.8125, 0.5625, 0.0625, 0. , 0. ,
               0. , 0. , 0.8125, 0.9375, 0.625 , 0.9375, 0.3125, 0. ,
               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. ])
In [22]: #Train Test Split Data
         from sklearn.model_selection import train_test_split
In [24]: X_train, X_test, y_train, y_test=train_test_split(data, df.target, test_size=0.3)
In [25]: X_train.shape, X_test.shape, y_train.shape, y_test.shape
Out[25]: ((1257, 64), (540, 64), (1257,), (540,))
In [26]: #Random Forest Model
        from sklearn.ensemble import RandomForestClassifier
In [27]: rf=RandomForestClassifier()
In [28]: rf.fit(X_train,y_train)
Out[28]:
         ▼ RandomForestClassifier
         RandomForestClassifier()
In [29]: #Predict Test Data
        y_pred=rf.predict(X_test)
In [30]: y_pred
Out[30]: array([8, 8, 3, 0, 5, 8, 9, 6, 9, 1, 5, 2, 5, 5, 3, 2, 5, 3, 8, 2, 0, 8,
               1, 3, 9, 6, 0, 1, 5, 6, 0, 3, 0, 7, 3, 9, 6, 1, 1, 3, 5, 7, 0, 0,
               6, 5, 3, 5, 6, 9, 3, 4, 6, 6, 2, 2, 1, 3, 2, 5, 3, 2, 0, 2, 6, 7,
               5, 8, 9, 7, 7, 4, 4, 2, 4, 1, 4, 8, 4, 3, 0, 9, 0, 7, 3, 0, 7, 8,
               2, 5, 5, 0, 9, 1, 5, 4, 6, 0, 8, 2, 8, 3, 3, 5, 3, 1, 5, 0, 8, 6,
               3, 6, 7, 5, 0, 4, 6, 9, 9, 7, 0, 0, 9, 3, 9, 9, 5, 3, 5, 9, 5, 7,
               7, 9, 1, 4, 0, 5, 6, 1, 1, 5, 7, 2, 5, 7, 3, 1, 6, 8, 7, 5, 1, 7,
               9, 1, 7, 2, 4, 0, 3, 5, 7, 1, 0, 9, 3, 8, 4, 7, 4, 6, 4, 3, 5, 8,
               6, 7, 5, 6, 6, 0, 7, 4, 3, 4, 4, 8, 5, 6, 1, 1, 3, 0, 6, 4, 0, 1,
               9, 2, 9, 5, 4, 9, 7, 2, 8, 9, 5, 7, 4, 9, 2, 0, 7, 8, 0, 2, 7, 7,
               3, 7, 8, 7, 2, 1, 0, 0, 9, 8, 1, 9, 9, 6, 0, 9, 8, 5, 2, 1, 2, 5,
               3, 1, 3, 7, 1, 4, 3, 9, 8, 3, 3, 0, 1, 7, 3, 9, 6, 1, 9, 6, 4, 3,
               9, 8, 0, 4, 6, 5, 4, 3, 6, 6, 8, 3, 6, 5, 3, 9, 0, 8, 1, 5, 6, 4,
               9, 1, 4, 1, 7, 4, 5, 5, 5, 9, 1, 7, 9, 0, 9, 2, 4, 8, 0, 0, 1, 7,
               8, 3, 1, 2, 3, 3, 9, 0, 3, 3, 1, 7, 1, 9, 1, 7, 1, 1, 6, 9, 7,
               4, 5, 3, 7, 2, 8, 1, 8, 3, 1, 6, 0, 1, 6, 3, 4, 5, 5, 8, 4, 4, 3,
               7, 3, 0, 4, 9, 9, 8, 6, 2, 7, 7, 1, 7, 2, 2, 0, 7, 5, 2, 8, 2, 0,
               3, 7, 7, 9, 7, 6, 5, 6, 3, 4, 6, 9, 4, 8, 1, 7, 3, 0, 3, 9, 9, 0,
               7, 3, 6, 9, 7, 4, 5, 2, 1, 1, 0, 2, 6, 3, 8, 5, 0, 9, 3, 2, 8, 9,
               1, 4, 9, 0, 2, 9, 7, 4, 9, 8, 7, 7, 9, 2, 9, 6, 7, 8, 6, 9, 2, 3,
               5, 4, 1, 0, 7, 7, 0, 3, 3, 2, 8, 5, 8, 6, 4, 5, 3, 6, 6, 5, 7, 1,
               3, 1, 2, 2, 5, 3, 7, 9, 1, 7, 8, 9, 6, 2, 8, 3, 8, 0, 3, 7, 1, 9,
               5, 0, 2, 8, 9, 7, 6, 0, 0, 2, 6, 8, 5, 1, 8, 3, 9, 8, 1, 6, 1, 5,
               8, 5, 3, 3, 3, 0, 5, 2, 7, 1, 5, 8, 5, 9, 3, 2, 1, 9, 9, 9, 6, 2,
               3, 8, 3, 7, 8, 9, 0, 2, 7, 5, 7, 8])
In [31]: #Model Accuracy
         from sklearn.metrics import confusion_matrix,classification_report
In [32]: confusion_matrix(y_test,y_pred)
Out[32]: array([[52, 0, 0, 0, 0, 0, 0, 0, 0],
               [ 0, 54, 0, 0, 0, 0, 0, 0, 0],
               [ 0, 0, 43, 0, 0, 0, 0, 0, 0, 0],
               [ 0, 0, 1, 69, 0, 1, 0, 1, 1, 0],
               [ 0, 1, 0, 0, 39, 0, 0, 1, 0, 0],
               [ 0, 0, 0, 0, 56, 0, 0, 0, 0],
               [ 0, 0, 0, 0, 0, 1, 48, 0, 0, 0],
               [ 0, 0, 0, 0, 0, 0, 58, 0, 1],
               [0, 0, 0, 0, 0, 0, 0, 1, 48, 0],
               [ 0, 0, 0, 0, 0, 0, 1, 1, 62]], dtype=int64)
In [33]: print(classification_report(y_test,y_pred))
                    precision
                                recall f1-score support
                                                       52
                         1.00
                                  1.00
                                           1.00
                         0.98
                                  1.00
                                           0.99
                                                       54
                                                       43
                         0.98
                                  1.00
                                           0.99
                                                       73
                         1.00
                                  0.95
                                           0.97
                                                       41
                         1.00
                                  0.95
                                           0.97
```

0.97

1.00

0.94

0.96

0.98

0.98

accuracy

macro avg

1.00

0.98

0.98

0.98

0.97

0.98

0.98

0.99

0.96

0.97

0.98

0.98

0.98

56

49

59

49

64

540

540

weighted avg 0.98 0.98 0.98 540

Τn Γ 1·