In [1]: **import** pandas **as** pd import numpy as np import matplotlib.pyplot as plt from sklearn.datasets import load_digits df=load_digits() In [5]: __,axes=plt.subplots(nrows=1,ncols=4,figsize=(10,3)) for ax,image,lable 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'%lable) Training: 0 Training: 1 Training: 2 Training: 3 df.images.shape (1797, 8, 8)Out[7]: In [8]: df.images[0] 0., 5., 13., 9., 1., array([[0., 0., 0., 13., 15., 10., 15., 0., 3., 15., 2., 0., 11., 8., 8., 4., 12., 0., 0., 8., 5., 8., 0., 0., 9., 8., 4., 11., 0., 1., 12., 2., 14., 5., 10., 12., 0., 0.], [0., 0., 6., 13., 10., 0., 0., 0.]In [9]: df.images[0].shape (8, 8)Out[9]: len(df.images) Out[10]: In [12]: n_samples=len(df.images) data=df.images.reshape((n_samples, -1)) In [13]: data[0] array([0., 0., 5., 13., 9., 1., 0., 0., 0., Out[13]: 8., 15., 5., 0., 0., 3., 15., 2., 0., 11., 12., 0., 0., 8., 8., 0., 0., 5., 8., 0., 0., 0., 4., 11., 0., 1., 12., 7., 0., 0., 2., 14., 5., 10., 12., 0., 0., 0., 6., 13., 10., 0., In [14]: data[0].shape Out[14]: (64,) In [15]: data.shape Out[15]: (1797, 64) In [16]: data.min() Out[16]: In [17]: data.max() 16.0 Out[17]: data=data/16 In [18]: In [20]: data.min() 0.0 Out[20]: In [21]: data.max() Out[21]: In [22]: data[0] , 0. , 0.3125, 0.8125, 0.5625, 0.0625, 0. array([0. Out[22]: , 0. , 0.8125, 0.9375, 0.625 , 0.9375, 0.3125, 0. , 0.1875, 0.9375, 0.125 , 0. , 0.6875, 0.5 Θ. , 0. , 0.25 , 0.75 , 0. , 0. , 0.5 , 0.5 , 0. , 0.5625, 0.5 , 0.3125, 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. , 0.375 , 0.8125, 0.625 , 0.]) from sklearn.model_selection import train_test_split In [23]: In [28]: x_train,x_test,y_train,y_test=train_test_split(data, df.target, test_size=0.3) In [29]: x_train.shape,x_test.shape,y_train.shape,y_test.shape ((1257, 64), (540, 64), (1257,), (540,))Out[29]: from sklearn.ensemble import RandomForestClassifier In [30]: rf=RandomForestClassifier() In [31]: In [32]: rf.fit(x_train,y_train) Out[32]: ▼ RandomForestClassifier RandomForestClassifier() In [33]: y_pred=rf.predict(x_test) In [34]: y_pred array([5, 8, 8, 3, 7, 7, 7, 6, 6, 5, 9, 6, 7, 7, 6, 9, 3, 5, 7, 9, 8, 3, Out[34]: 2, 7, 4, 1, 7, 2, 7, 6, 5, 4, 2, 5, 5, 2, 7, 5, 7, 0, 0, 4, 7, 1, 8, 3, 0, 4, 7, 5, 5, 2, 8, 5, 8, 2, 3, 2, 9, 6, 5, 2, 0, 8, 1, 1, 8, 9, 2, 1, 9, 2, 9, 0, 4, 3, 1, 5, 9, 7, 0, 8, 1, 0, 6, 4, 8, 1, 1, 3, 3, 5, 1, 1, 4, 7, 4, 2, 3, 8, 9, 5, 8, 6, 0, 7, 7, 6, 3, 3, 6, 7, 5, 8, 3, 1, 6, 9, 1, 6, 6, 2, 2, 4, 3, 2, 3, 0, 1, 9, 5, 7, 7, 3, 0, 4, 3, 5, 0, 3, 8, 9, 7, 1, 5, 4, 9, 6, 4, 4, 0, 8, 2, 1, 3, 2, 6, 3, 4, 4, 6, 1, 8, 1, 7, 1, 7, 7, 6, 5, 9, 6, 2, 4, 8, 0, 7, 9, 7, 4, 7, 7, 3, 3, 2, 6, 3, 1, 4, 9, 9, 8, 8, 1, 0, 1, 1, 6, 3, 7, 9, 4, 7, 7, 2, 8, 4, 7, 2, 4, 1, 1, 4, 5, 7, 0, 2, 5, 0, 9, 2, 3, 9, 9, 4, 4, 1, 1, 8, 6, 8, 3, 1, 8, 6, 8, 7, 5, 9, 8, 5, 3, 5, 4, 7, 4, 5, 3, 5, 5, 4, 4, 4, 1, 2, 9, 5, 4, 9, 6, 5, 5, 2, 7, 2, 7, 3, 7, 3, 5, 7, 0, 6, 2, 0, 1, 3, 5, 1, 2, 2, 6, 9, 8, 5, 3, 9, 9, 0, 2, 0, 1, 5, 2, 7, 1, 8, 0, 4, 7, 6, 9, 5, 3, 5, 4, 1, 1, 8, 0, 5, 4, 5, 4, 0, 7, 1, 4, 9, 7, 9, 8, 6, 1, 5, 5, 7, 8, 4, 6, 2, 8, 1, 5, 8, 9, 6, 6, 1, 4, 8, 7, 5, 2, 7, 4, 3, 9, 6, 8, 5, 3, 1, 6, 1, 4, 7, 8, 5, 2, 8, 2, 6, 4, 3, 3, 7, 3, 2, 1, 8, 7, 8, 7, 4, Θ, 0, 2, 6, 9, 9, 6, 9, 4, 2, 8, 7, 2, 3, 7, 7, 7, Ο, 9, 6, 6, 6, 6, 4, 9, 4, 2, 3, 1, 1, 3, 0, 0, 2, 1, 6, 6, 3, 3, 4, 3, 6, 2, 7, 0, 2, 0, 5, 3, 6, 7, 8, 1, 8, 2, 1, 2, 5, 3, 6, 0, 7, 5, 2, 0, 7, 4, 0, 7, 4, 0, 7, 8, 0, 3, 4, 9, 1, 1, 9, 9, 7, 6, 6, 1, 4, 5, 3, 5, 9, 4, 8, 0, 3, 0, 6, 6, 8, 9, 7, 3, 5, 3, 0, 8, 1, 7, 9, 5, 3, 9, 0, 3, 5, 5, 5, 4, 7]) In [35]: **from** sklearn.metrics **import** confusion_matrix,classification_report confusion_matrix(y_test,y_pred) In [36]: array([[43, 0, 0], Out[36]: 0, 51, Θ, Θ, Θ, 0], Θ, Θ, Θ, Θ, 52, 0], Θ, Θ, Θ, 1, Θ, Θ, Θ, Θ, Θ, 1, 0, 56, Θ, 2, Θ, Θ, Θ, 0], Θ, Θ, 0, 55, Θ, 3, Θ, Ο, Θ, 0], Θ, Θ, Θ, Θ, 55, Θ, Θ, Θ, Θ, 1], Θ, Θ, Θ, 2, Θ, 52, Θ, 0], Θ, Θ, Θ, Ο, Θ, Θ, Θ, Θ, Θ, 65, Θ, 1], Θ, [0, 3, Θ, Θ, Θ, 1, 46, 0], 1, 2, [0, Ο, Ο, Θ, Θ, Θ, 1, 46]], dtype=int64) Ο, In [37]: y_pred array([5, 8, 8, 3, 7, 7, 7, 6, 6, 5, 9, 6, 7, 7, 6, 9, 3, 5, 7, 9, 8, 3, 2, 7, 4, 1, 7, 2, 7, 6, 5, 4, 2, 5, 5, 2, 7, 5, 7, 0, 0, 4, 7, 1, 8, 3, 0, 4, 7, 5, 5, 2, 8, 5, 8, 2, 3, 2, 9, 6, 5, 2, 0, 8, 1, 1, 9, 2, 1, 9, 2, 9, 0, 4, 3, 1, 5, 9, 7, 0, 8, 1, 0, 6, 4, 8, 1, 3, 3, 5, 1, 1, 4, 7, 4, 2, 3, 8, 9, 5, 8, 6, 0, 7, 7, 7, 5, 8, 3, 1, 6, 9, 1, 6, 6, 2, 2, 4, 3, 2, 3, 0, 1, 5, 7, 7, 3, 0, 4, 3, 5, 0, 3, 8, 9, 7, 1, 5, 4, 9, 6, 3, 2, 6, 4, 6, 1, 8, 1, 7, 1, 7, 4, 0, 8, 2, 3, 4, 7, 6, 5, 1, 9, 7, 4, 7, 7, 2, 4, 8, Θ, 7, 3, 3, 2, 6, 3, 1, 4, 9, 9, 3, 7, 9, 0, 1, 1, 6, 4, 7, 7, 2, 8, 4, 7, 2, 4, 1, 1, 4, 3, 9, 7, Θ, 2, 5, 0, 9, 2, 9, 4, 4, 1, 1, 8, 6, 8, 3, 1, 8, 6, 5, 9, 5, 4, 7, 4, 5, 3, 2, 7, 8, 5, 3, 5, 5, 4, 4, 4, 1, 9, 5, 7, 2, 7, 6, 7, 3, 7, 2, 5, 5, 2, 3, 5, 0, 6, Θ, 1, 3, 2, 0, 6, 9, 8, 5, 3, 9, 9, 0, 5, 2, 7, Ο, 1, 1, 8, 9, 5, 3, 5, 4, 1, 1, 8, 0, 5, 6, 4, 5, 4, 0, 7, 1, 9, 7, 9, 6, 1, 7, 1, 6, 8, 5, 5, 8, 4, 6, 2, 8, 5, 8, 7, 8, 5, 3, 1, 6, 4, 1, 5, 2, 7, 4, 3, 4, 9, 6, 5, 2, 8, 2, 7, 8, 6, 4, 3, 3, 7, 3, 2, 1, 8, 7, Θ, 2, 6, 9, 9, 6, 9, 4, 8, 7, 2, 8, 7, 2, 3, 7, 7, 1, 5, 9, 6, 6, 6, 6, 4, 7, 7, 9, 4, 2, 3, 1, 1, 3, 0, 2, 0, 2, 1, 6, 6, 3, 3, 4, 3, 6, 2, 7, 0, 2, 0, 5, 3, 6, 3, 7, 8, 1, 8, 2, 1, 2, 5, 3, 6, 0, 7, 5, 2, 0, 7, 1, 6, 1, 4, 0, 7, 4, 0, 7, 8, 0, 3, 4, 9, 1, 1, 9, 9, 7, 6, 4, 5, 3, 5, 9, 4, 8, 0, 3, 0, 6, 6, 8, 9, 7, 3, 5, 3, 0, 8, 1, 7, 9, 5, 3, 9, 0, 3, 5, 5, 5, 4, 7]) In [39]: **from** sklearn.metrics **import** confusion_matrix,classification_report In [40]: confusion_matrix(y_test,y_pred) Θ, Θ, 0], array([[43, 0, Out[40]: 0], 0, 51, Θ, Θ, Θ, Θ, Θ, Θ, Θ, Θ, 1, 52, Θ, Θ, Θ, Θ, Θ, 0], Θ, 1, Θ, 56, Θ, 2, Θ, Θ, 0], Θ, Θ, Θ, Θ, 55, Θ, 3, Θ, Θ, 0], 55, Θ, Θ, Θ, Θ, Θ, Θ, Θ, Θ, 1], Θ, 2, Θ, 52, Θ, Ο, Θ, Θ, Θ, 0], Θ, Θ, Ο, Θ, Θ, Θ, Θ, 65, Θ, 1], Θ, Θ, 3, Θ, Θ, 1, Θ, 1, 46, 0], [0, Θ, Θ, Θ, Θ, Θ, Θ, 2, 1, 46]], dtype=int64) In [41]: print(classification_report(y_test,y_pred)) precision recall f1-score support 0 43 1.00 1.00 1.00 1 0.91 0.95 51 1.00 53 1.00 0.98 0.99 1.00 0.95 0.97 59 58 0.96 0.95 0.96 5 0.95 0.98 0.96 56 6 1.00 0.96 0.98 54 7 0.92 0.98 0.95 66 8 0.98 0.90 0.94 51 9 0.96 0.94 0.95 49 accuracy 0.96 540 0.97 0.96 540 macro avg 0.97 weighted avg 540 0.97 0.96 0.96 In []: