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3.6.10.13. Simple visualization and classification of the digits dataset

Plot the first few samples of the digits dataset and a 2D representation built using PCA, then do a simple classification

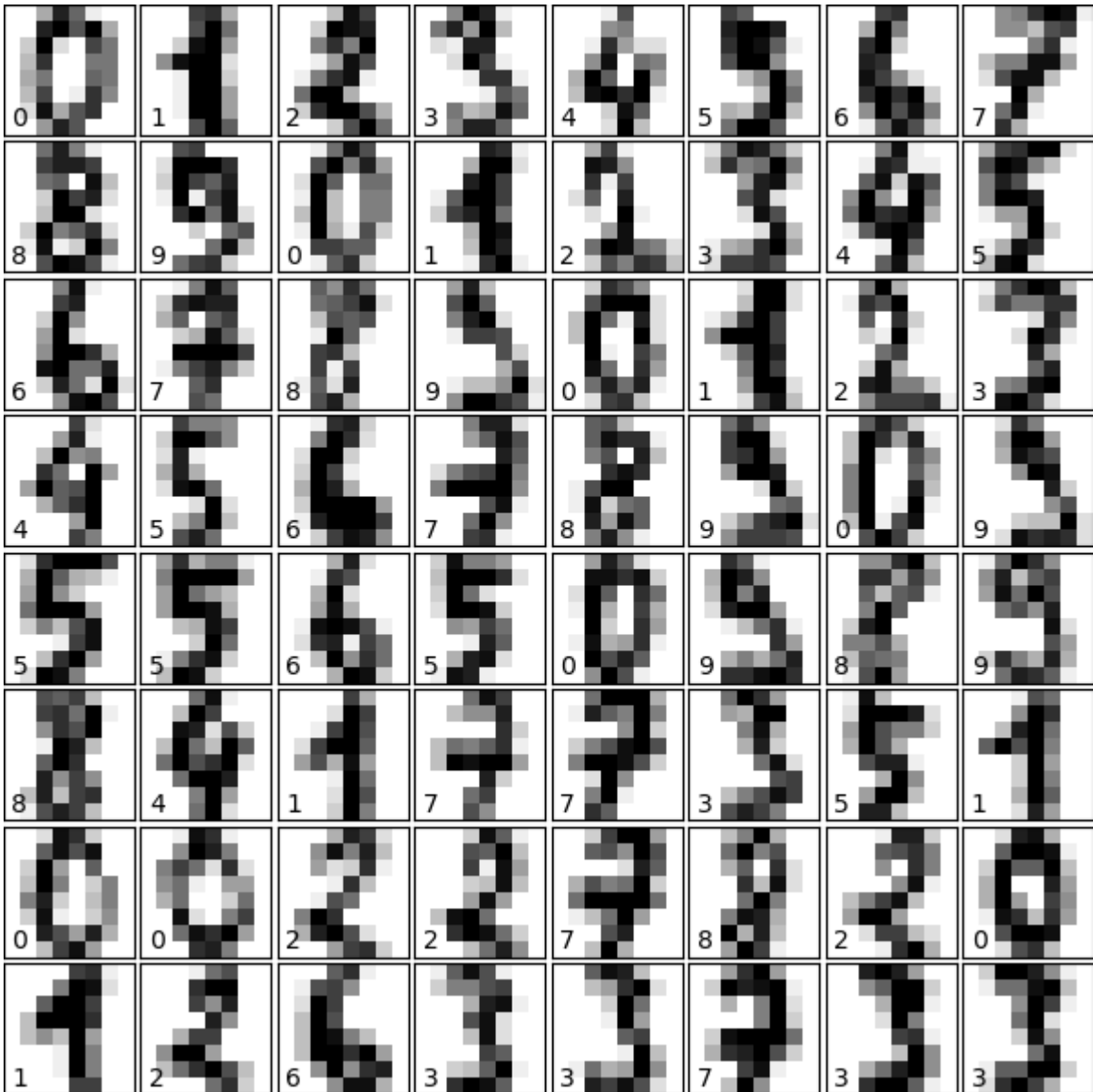
```
from sklearn.datasets import load_digits
digits = load_digits()
```

Plot the data: images of digits

Each data in a 8x8 image

```
from matplotlib import pyplot as plt
fig = plt.figure(figsize=(6, 6)) # figure size in inches
fig.subplots_adjust(left=0, right=1, bottom=0, top=1, hspace=0.05, wspace=0.05)

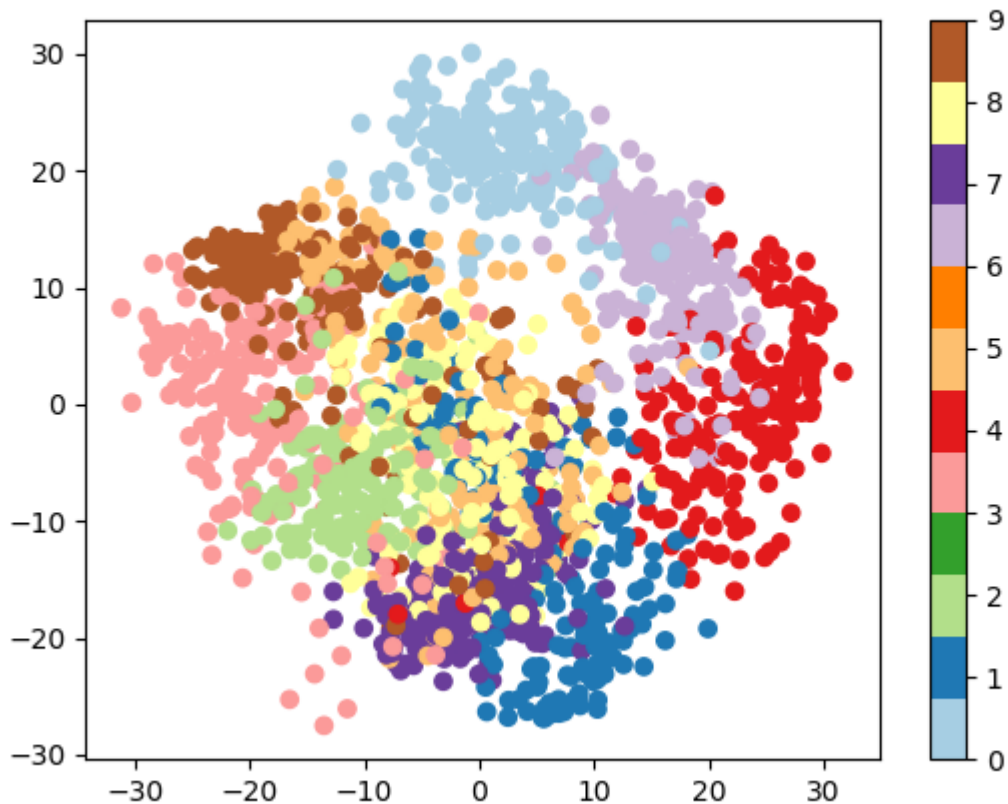
for i in range(64):
    ax = fig.add_subplot(8, 8, i + 1, xticks=[], yticks=[])
    ax.imshow(digits.images[i], cmap=plt.cm.binary, interpolation='nearest')
    # label the image with the target value
    ax.text(0, 7, str(digits.target[i]))
```



Plot a projection on the 2 first principal axis

```
plt.figure()
```

```
from sklearn.decomposition import PCA
pca = PCA(n_components=2)
proj = pca.fit_transform(digits.data)
plt.scatter(proj[:, 0], proj[:, 1], c=digits.target, cmap="Paired")
plt.colorbar()
```



Classify with Gaussian naive Bayes

```
from sklearn.naive_bayes import GaussianNB
from sklearn.model_selection import train_test_split

# split the data into training and validation sets
X_train, X_test, y_train, y_test = train_test_split(digits.data, digits.target)

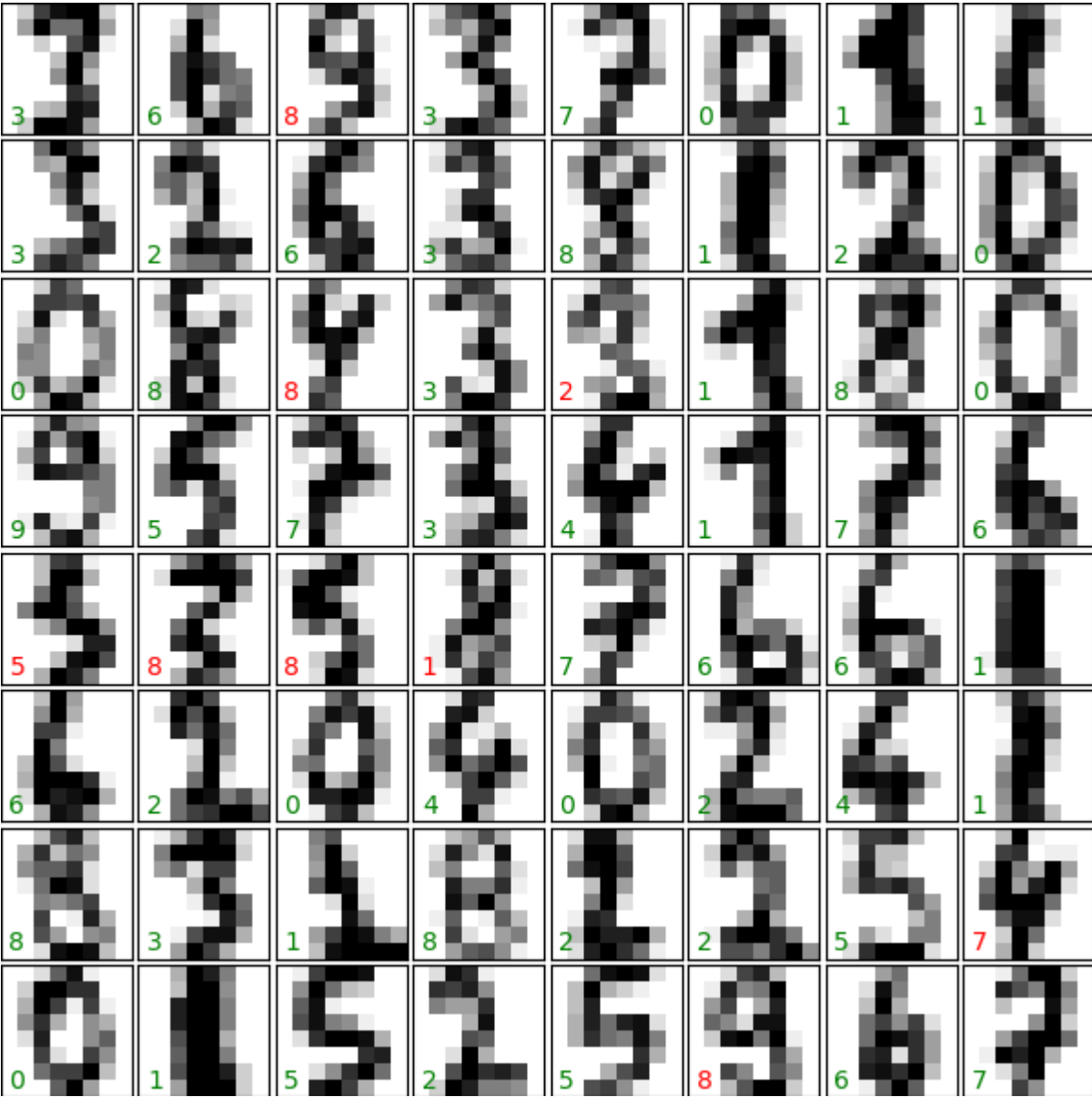
# train the model
clf = GaussianNB()
clf.fit(X_train, y_train)

# use the model to predict the labels of the test data
predicted = clf.predict(X_test)
expected = y_test

# Plot the prediction
fig = plt.figure(figsize=(6, 6)) # figure size in inches
fig.subplots_adjust(left=0, right=1, bottom=0, top=1, hspace=0.05, wspace=0.05)

# plot the digits: each image is 8x8 pixels
for i in range(64):
    ax = fig.add_subplot(8, 8, i + 1, xticks=[], yticks=[])
    ax.imshow(X_test.reshape(-1, 8, 8)[i], cmap=plt.cm.binary,
               interpolation='nearest')
```

```
# label the image with the target value
if predicted[i] == expected[i]:
    ax.text(0, 7, str(predicted[i]), color='green')
else:
    ax.text(0, 7, str(predicted[i]), color='red')
```



Quantify the performance

First print the number of correct matches

```
matches = (predicted == expected)
print(matches.sum())
```

Out: 402

The total number of data points

```
print(len(matches))
```

Out: 450

And now, the ration of correct predictions

```
matches.sum() / float(len(matches))
```

Print the classification report

```
from sklearn import metrics
print(metrics.classification_report(expected, predicted))
```

Out:

	precision	recall	f1-score	support
0	0.98	1.00	0.99	53
1	0.82	0.88	0.85	42

	2	0.90	0.91	0.90	57
	3	0.94	0.76	0.84	38
	4	0.97	0.85	0.91	40
	5	0.93	0.96	0.95	45
	6	0.95	0.97	0.96	40
	7	0.78	0.98	0.87	51
	8	0.78	0.87	0.82	46
	9	1.00	0.66	0.79	38
avg / total		0.90	0.89	0.89	450

Print the confusion matrix

```
print(metrics.confusion_matrix(expected, predicted))

plt.show()
```

Out:

```
[[53  0  0  0  0  0  0  0  0  0]
 [ 0 37  1  0  0  0  2  1  1  0]
 [ 0  2 52  0  1  0  0  0  2  0]
 [ 0  0  3 29  0  1  0  1  4  0]
 [ 0  1  0  0 34  0  0  4  1  0]
 [ 0  0  0  1  0 43  0  0  1  0]
 [ 0  0  0  0  0  1 39  0  0  0]
 [ 0  0  0  0  0  1  0 50  0  0]
 [ 0  4  0  0  0  0  0  2 40  0]
 [ 1  1  2  1  0  0  0  6  2 25]]
```

Total running time of the script: (0 minutes 1.780 seconds)

Download Python source code: plot_digits_simple_classif.py

Download Jupyter notebook: plot_digits_simple_classif.ipynb