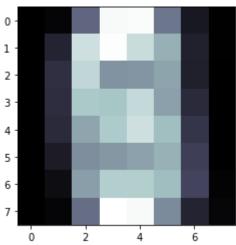
Performing PCA with SVM on digits dataset and checking prformance with and without PCA

```
#load dataset
from sklearn import datasets
digits = datasets.load_digits()
digits.data.shape
[→ (1797, 64)
from matplotlib import pyplot as plt
fig = plt.figure(figsize=(8, 6))
# plot several images
for i in range(15):
   ax = fig.add_subplot(3, 5, i + 1, xticks=[], yticks=[])
   ax.imshow(digits.images[i], cmap=plt.cm.bone)
C→
#splitting
from sklearn.model selection import train test split
X_train, X_test, y_train, y_test = train_test_split(digits.data,
       digits.target, random_state=142)
print(X_train.shape, X_test.shape)
   (1347, 64) (450, 64)
from sklearn import decomposition
pca = decomposition.PCA(n_components=50, whiten=True)
pca.fit(X_train)
```

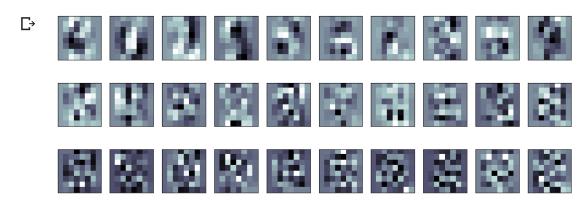
PCA(copy=True, iterated_power='auto', n_components=50, random_state=None, svd_solver='auto', tol=0.0, whiten=True)

cmatplotlib.image.AxesImage at 0x7f2814d35f28>

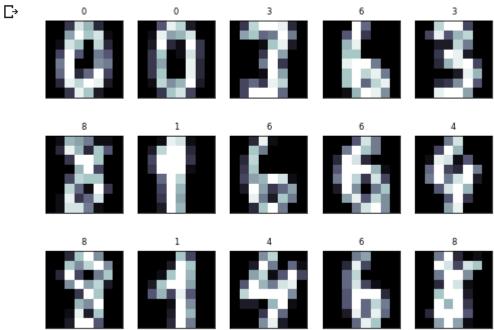


print(pca.components_.shape)

[→ (50, 64)



```
142_01.ipynb - Colaboratory
X_test_pca = pca.transform(X_test)
print(X_train_pca.shape)
    (1347, 50)
Гэ
print(X_test_pca.shape)
    (450, 50)
#with pcm
from sklearn import svm
clf = svm.SVC(C=5., gamma=0.001)
clf.fit(X_train_pca, y_train)
     SVC(C=5.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
         decision_function_shape='ovr', degree=3, gamma=0.001, kernel='rbf',
         max_iter=-1, probability=False, random_state=None, shrinking=True,
         tol=0.001, verbose=False)
import numpy as np
fig = plt.figure(figsize=(8, 6))
for i in range(15):
    ax = fig.add_subplot(3, 5, i + 1, xticks=[], yticks=[])
    ax.imshow(X_test[i].reshape(digits.images[0].shape),
              cmap=plt.cm.bone)
    y_pred = clf.predict(X_test_pca[i, np.newaxis])[0]
    color = ('black' if y_pred == y_test[i] else 'red')
    ax.set_title(y_pred, fontsize='small', color=color)
С→
```



from sklearn import metrics y_pred = clf.predict(X_test_pca) print(metrics.classification_report(y_test, y_pred))

```
precision
                                  recall f1-score
С→
                                                       support
                0
                         1.00
                                    1.00
                                               1.00
                                                             52
                1
                         0.94
                                    0.96
                                               0.95
                                                             48
                2
                                                             50
                         0.98
                                    1.00
                                               0.99
                3
                                    0.94
                                               0.97
                                                             51
                         1.00
                4
                         1.00
                                    0.96
                                               0.98
                                                             46
                5
                         0.97
                                    0.97
                                               0.97
                                                             31
                6
                         1.00
                                    1.00
                                               1.00
                                                             50
                7
                         0.97
                                    1.00
                                               0.99
                                                             38
                8
                         0.88
                                    0.95
                                               0.91
                                                             39
                9
                         0.93
                                    0.91
                                               0.92
                                                             45
                                               0.97
                                                            450
        accuracy
                         0.97
                                    0.97
                                               0.97
                                                            450
       macro avg
                         0.97
                                    0.97
                                               0.97
                                                            450
    weighted avg
```

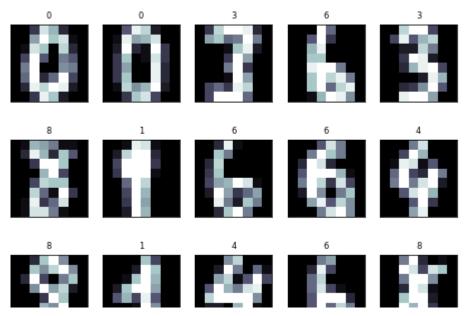
print(metrics.confusion_matrix(y_test, y_pred))

```
Г⇒
    [[52
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                      0
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                            0
                               0
                                  0]
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```

```
#without pcm
clf = svm.SVC(C=5., gamma=0.001)
clf.fit(X_train, y_train)
```

SVC(C=5.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0, decision_function_shape='ovr', degree=3, gamma=0.001, kernel='rbf', max_iter=-1, probability=False, random_state=None, shrinking=True, tol=0.001, verbose=False)

 \Box



y_pred = clf.predict(X_test)
print(metrics.classification_report(y_test, y_pred))

₽	precision	recall	f1-score	support
0	1.00	1.00	1.00	52
1	1.00	1.00	1.00	48
2	1.00	1.00	1.00	50
3	1.00	0.94	0.97	51
4	1.00	1.00	1.00	46
5	0.94	1.00	0.97	31
6	1.00	1.00	1.00	50
7	0.97	1.00	0.99	38
8	0.97	1.00	0.99	39
9	0.98	0.96	0.97	45
accuracy			0.99	450
macro avg	0.99	0.99	0.99	450
weighted avg	0.99	0.99	0.99	450

print(metrics.confusion_matrix(y_test, y_pred))

```
0]
  48
      0
         0
                             0]
   0 50
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            0
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      0 48
            0
               1
                         1 1]
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                   0
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               0 50
                            0]
            0
      0
         0
                   0 38
                             0]
         0
0
      0
            0
               0
                   0
                      0 39
                             0]
                1
                         0 43]]
```