

课程作业三：基于支持向量机的手写数字识别 github: <https://github.com/dypw/homework3>

```
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)

# # Input data files are available in the read-only "../input/" directory
# # For example, running this (by clicking run or pressing Shift+Enter) will list
# the files in the input directory

# import os
# for dirname, _, filenames in os.walk('/content/minist'):
#     for filename in filenames:
#         print(os.path.join(dirname, filename))

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn
import sklearn
%matplotlib inline

from tensorflow.keras.datasets import mnist
(X_train, Y_train), (X_test, Y_test) = mnist.load_data()

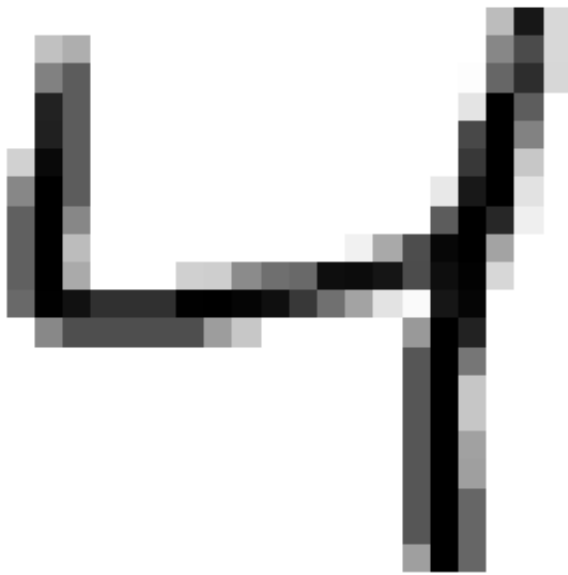
print('X_train Shape :' + str(X_train.shape))    #shape of Data
print('Y_train Shape :' + str(Y_train.shape))
print('X_test Shape :' + str(X_test.shape))
print('Y_test Shape :' + str(Y_test.shape))

X_train Shape :(60000, 28, 28)
Y_train Shape :(60000,)
X_test Shape :(10000, 28, 28)
Y_test Shape :(10000,)

X_pixels =X_train.flatten()    #flattening the input pixel from 28*28 to 784
X=X_pixels.reshape(60000,784)
Xtest = X_test.flatten()
Xtest = Xtest.reshape(10000,784)
print(X.shape)
print(Xtest.shape)

(60000, 784)
(10000, 784)
```

```
import matplotlib
some_digit = X[2]
some_digit_image = some_digit.reshape(28, 28)
plt.imshow(some_digit_image, cmap = matplotlib.cm.binary,
            interpolation="nearest")
plt.axis("off")
plt.show()
```



```

from sklearn.svm import SVC
svc = SVC(kernel='rbf', C=1).fit(X, Y_train)    #training the data
y_pred = svc.predict(Xtest)
#importing confusion matrix
from sklearn.metrics import confusion_matrix
confusion = confusion_matrix(Y_test, y_pred)
print('Confusion Matrix\n')
print(confusion)

```

Confusion Matrix

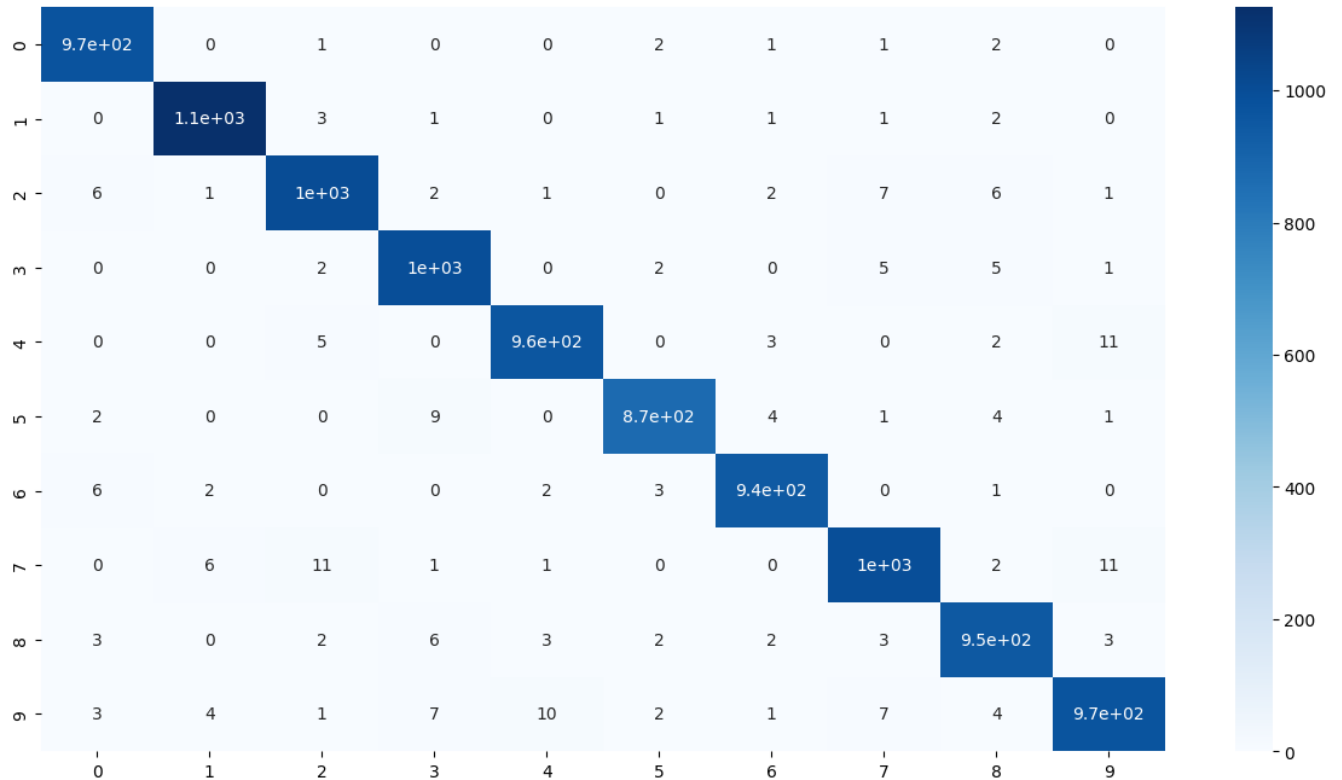
```

[[ 973     0     1     0     0     2     1     1     2     0]
 [    0 1126     3     1     0     1     1     1     2     0]
 [    6     1 1006     2     1     0     2     7     6     1]
 [    0     0     2  995     0     2     0     5     5     1]
 [    0     0     5     0  961     0     3     0     2    11]
 [    2     0     0     9     0  871     4     1     4     1]
 [    6     2     0     0     2     3  944     0     1     0]
 [    0     6    11     1     1     0     0  996     2    11]
 [    3     0     2     6     3     2     2     3  950     3]
 [    3     4     1     7    10     2     1     7     4  970]]

```

```
import seaborn as sns
plt.figure(figsize=(15,8))
sns.heatmap(confusion, annot=True,
            cmap='Blues')
```

<Axes: >



```
#importing accuracy_score, precision_score, recall_score, f1_score
from sklearn.metrics import accuracy_score, precision_score, recall_score, f1_score
print('\nAccuracy: {:.2f}\n'.format(accuracy_score(Y_test, y_pred)))

print('Micro Precision: {:.2f}'.format(precision_score(Y_test, y_pred, average='
```

```

print('Micro Recall: {:.2f}'.format(recall_score(Y_test, y_pred, average='micro'))
print('Micro F1-score: {:.2f}\n'.format(f1_score(Y_test, y_pred, average='micro'))

print('Macro Precision: {:.2f}'.format(precision_score(Y_test, y_pred, average='macro'))
print('Macro Recall: {:.2f}'.format(recall_score(Y_test, y_pred, average='macro'))
print('Macro F1-score: {:.2f}\n'.format(f1_score(Y_test, y_pred, average='macro'))

print('Weighted Precision: {:.2f}'.format(precision_score(Y_test, y_pred, average='weighted'))
print('Weighted Recall: {:.2f}'.format(recall_score(Y_test, y_pred, average='weighted'))
print('Weighted F1-score: {:.2f}'.format(f1_score(Y_test, y_pred, average='weighted'))

from sklearn.metrics import classification_report
print('\nClassification Report\n')
print(classification_report(Y_test, y_pred, target_names=['Class 1', 'Class 2',

```

Accuracy: 0.98

Micro Precision: 0.98
 Micro Recall: 0.98
 Micro F1-score: 0.98

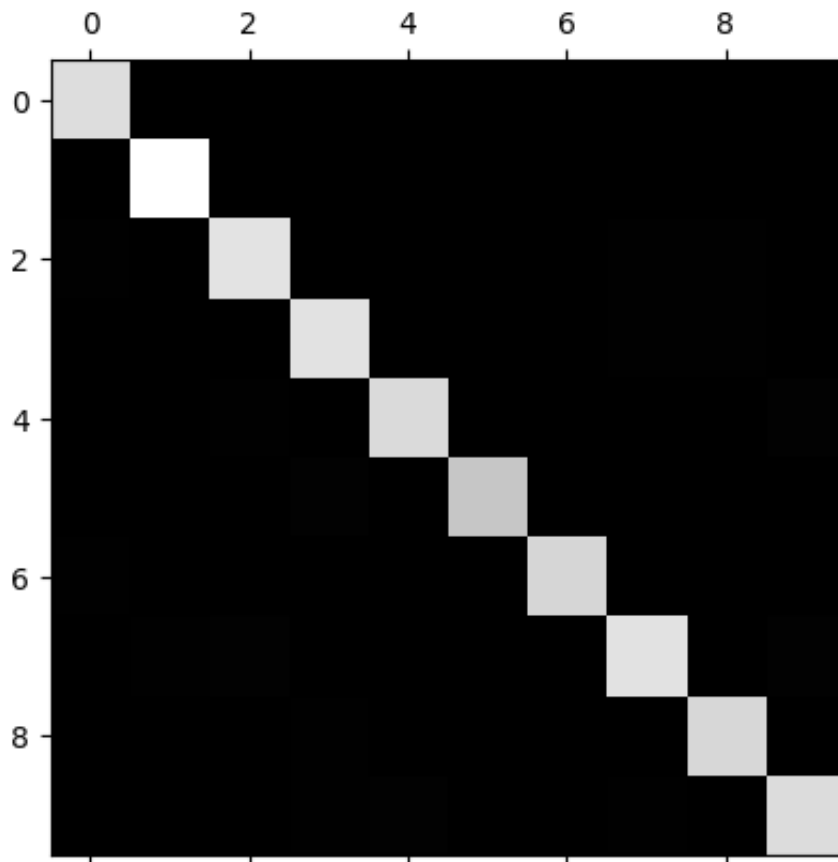
Macro Precision: 0.98
 Macro Recall: 0.98
 Macro F1-score: 0.98

Weighted Precision: 0.98
 Weighted Recall: 0.98
 Weighted F1-score: 0.98

Classification Report

	precision	recall	f1-score	support
Class 1	0.98	0.99	0.99	980
Class 2	0.99	0.99	0.99	1135
Class 3	0.98	0.97	0.98	1032
Class 4	0.97	0.99	0.98	1010
Class 5	0.98	0.98	0.98	982
Class 6	0.99	0.98	0.98	892
Class 7	0.99	0.99	0.99	958
Class 8	0.98	0.97	0.97	1028
Class 9	0.97	0.98	0.97	974
Class 10	0.97	0.96	0.97	1009
accuracy			0.98	10000
macro avg	0.98	0.98	0.98	10000
weighted avg	0.98	0.98	0.98	10000

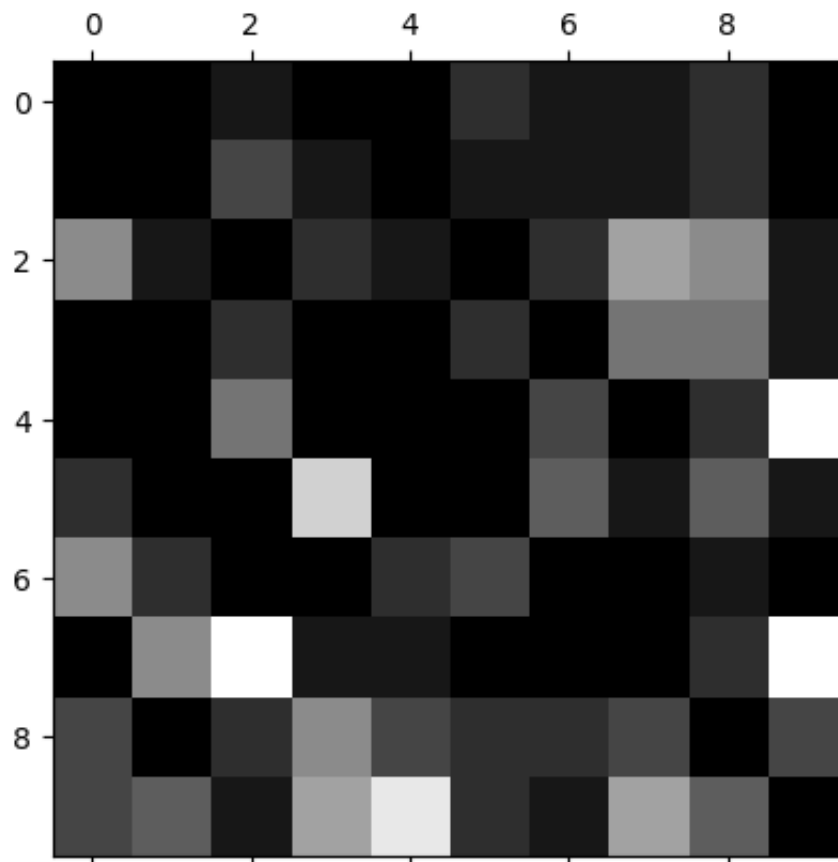
```
plt.matshow(confusion, cmap=plt.cm.gray)
plt.show()
```



Error Analysis

```
row_sums = confusion.sum(axis=1, keepdims=True)
norm_conf_mx = confusion / row_sums
```

```
np.fill_diagonal(confusion, 0)
plt.matshow(confusion, cmap=plt.cm.gray)
plt.show()
```



```
from sklearn.model_selection import cross_val_predict
sv = SVC(kernel='rbf', C=1)
y_train_pred = cross_val_predict(sv, X, Y_train, cv=3)
```

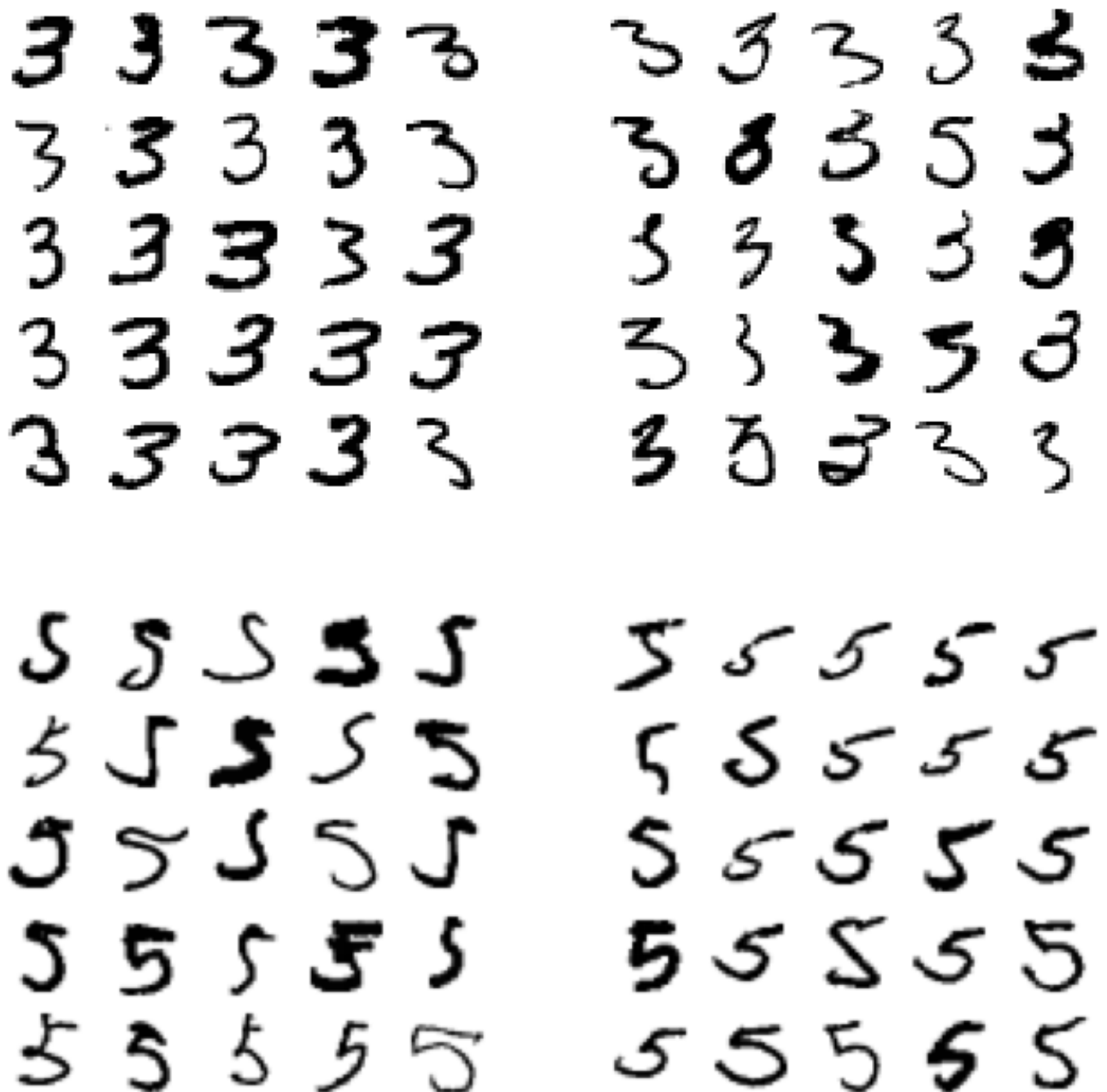
```
# EXTRA
def plot_digits(instances, images_per_row=10, **options):
    size = 28
    images_per_row = min(len(instances), images_per_row)
    images = [instance.reshape(size,size) for instance in instances]
    n_rows = (len(instances) - 1) // images_per_row + 1
    row_images = []
    n_empty = n_rows * images_per_row - len(instances)
    images.append(np.zeros((size, size * n_empty)))
    for row in range(n_rows):
        rimages = images[row * images_per_row : (row + 1) * images_per_row]
        row_images.append(np.concatenate(rimages, axis=1))
    image = np.concatenate(row_images, axis=0)
    plt.imshow(image, cmap = matplotlib.cm.binary, **options)
    plt.axis("off")
```



```

cl_a, cl_b = 3, 5
X_aa = X[(Y_train == cl_a) & (y_train_pred == cl_a)]
X_ab = X[(Y_train == cl_a) & (y_train_pred == cl_b)]
X_ba = X[(Y_train == cl_b) & (y_train_pred == cl_a)]
X_bb = X[(Y_train == cl_b) & (y_train_pred == cl_b)]
plt.figure(figsize=(8,8))
plt.subplot(221); plot_digits(X_aa[:25], images_per_row=5)
plt.subplot(222); plot_digits(X_ab[:25], images_per_row=5)
plt.subplot(223); plot_digits(X_ba[:25], images_per_row=5)
plt.subplot(224); plot_digits(X_bb[:25], images_per_row=5)
plt.show()

```



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