

彩色影像處理

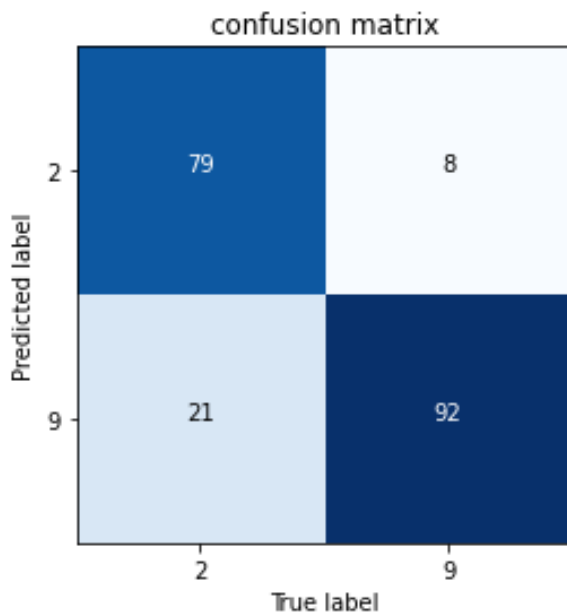
作業 3：簡化的 CNN 數值辨識

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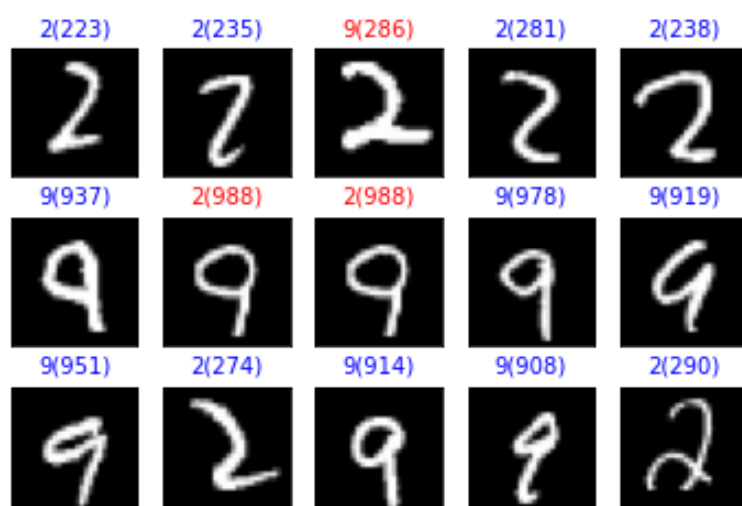
完成了可以隨意指定 0~9 之間兩類別的比較，若要更改數值，請在下圖地方更改，路徑改成存放位置的地方。其餘程式碼部分已有註解，有疑問再請詳看。

```
trainPath = 'C:\\Users\\User\\Desktop\\NTUST\\image_processing\\HW3_readme\\train1000\\'  
testPath = 'C:\\Users\\User\\Desktop\\NTUST\\image_processing\\HW3_readme\\test1000\\'  
  
kernel1 = np.array([(-1, -1, 1, -1, 0, 1, -1, 1, 1)]).reshape(3, 3)  
#預設的filter1  
kernel2 = np.random.random(size=(3, 3))  
kernel2 = (np.array((kernel2 / np.sum(kernel2))) - (1/9)) * 20  
#自訂的filter2  
  
# num1 , num2 為可指定 0 ~9  
num1 = 2  
num2 = 9
```

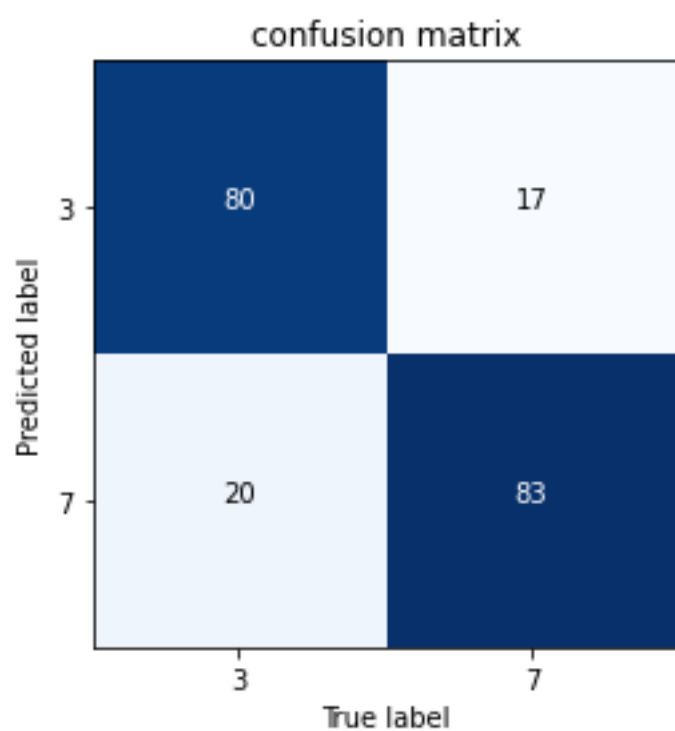
類別：2、9

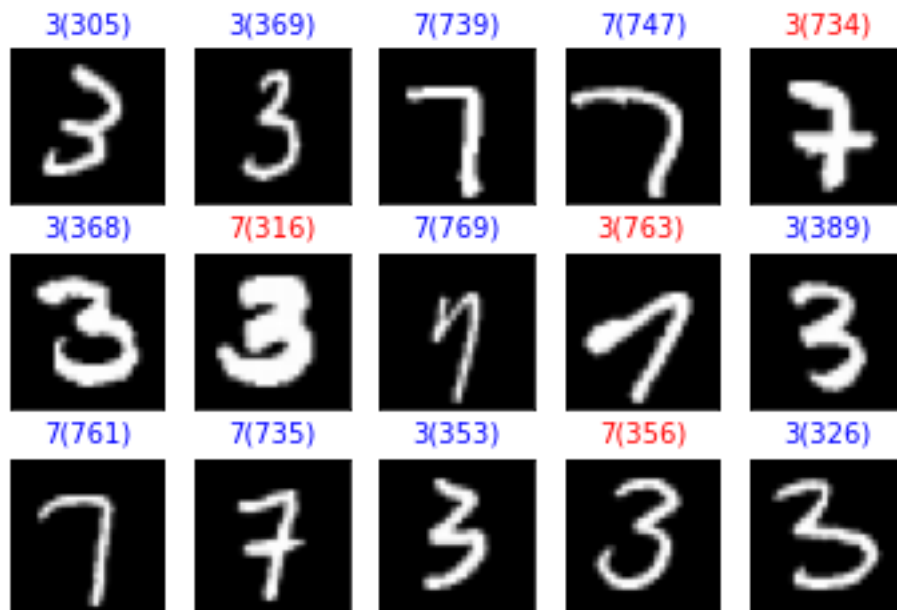


隨機 15 個影像分類結果



類別 3、7





Code :

```
import numpy as np
import cv2
from sklearn.metrics import classification_report
import itertools
import matplotlib.pyplot as plt
from sklearn.metrics import confusion_matrix
import random

trainPath = 'C:\\Users\\User\\Desktop\\NTUST\\image_processing\\HW3
readme\\train1000\\'
testPath = 'C:\\Users\\User\\Desktop\\NTUST\\image_processing\\HW3
readme\\test1000\\'

kernel1 = np.array([[-1, -1, 1, -1, 0, 1, -1, 1, 1]]).reshape(3, 3)
#預設的 filter1
kernel2 = np.random.random(size=(3, 3))
kernel2 = (np.array((kernel2 / np.sum(kernel2))) - (1/9)) * 20
#自訂的 filter2

# num1 , num2 為可指定 0 ~9
num1 = 2
```

```
num2 = 9
```

```
#讀取檔案，只取 train 的部分即可完成作業需求
```

```
def load_image(num1 , num2):
```

```
    train = []
```

```
    test = []
```

```
    num1 = num1 * 100
```

```
    num2 = num2 * 100
```

```
    for i in range(num1 , num1 + 100):
```

```
        img = cv2.imread(trainPath + str(i + 1) + ".png", 0)
```

```
        train.append(cv2.resize(img, (8, 8)))
```

```
    for i in range(num2,num2 + 100):
```

```
        img = cv2.imread(trainPath + str(i + 1) + ".png", 0)
```

```
        train.append(cv2.resize(img, (8, 8)))
```

```
    return train
```

```
#max pooling
```

```
def pooling(inputMap, poolSize=3, poolStride=2, mode='max'):
```

```
    """INPUTS:
```

```
        inputMap - input array of the pooling layer
```

```
        poolSize - X-size(equivalent to Y-size) of receptive field
```

```
        poolStride - the stride size between successive pooling squares
```

```
    OUTPUTS:
```

```
        outputMap - output array of the pooling layer
```

```
        Padding mode - 'edge'
```

```
    """
```

```
    # inputMap sizes
```

```
    in_row, in_col = np.shape(inputMap)
```

```
    # outputMap sizes
```

```
    out_row, out_col = int(np.floor(in_row / poolStride)), int(np.floor(in_col /
```

```

poolStride))
    row_remainder, col_remainder = np.mod(in_row, poolStride),
np.mod(in_col, poolStride)
    if row_remainder != 0:
        out_row += 1
    if col_remainder != 0:
        out_col += 1
    outputMap = np.zeros((out_row, out_col))

    # padding
    temp_map = np.lib.pad(inputMap, ((0, poolSize - row_remainder), (0,
poolSize - col_remainder)), 'edge')

    # max pooling
    for r_idx in range(0, out_row):
        for c_idx in range(0, out_col):
            startX = c_idx * poolStride
            startY = r_idx * poolStride
            poolField = temp_map[startY:startY + poolSize, startX:startX +
poolSize]

            poolOut = np.max(poolField)
            outputMap[r_idx, c_idx] = poolOut

    # retrun outputMap
    return outputMap
#第一層捲積及完成 max-pooling
def conv_pool_layer1(img, kernel):
    # Conv_and_Pooling
    conv_img = cv2.filter2D(img, -1, kernel)
    max_img = pooling(conv_img, 2)
    return max_img
#第二層捲積及完成 max-pooling
def conv_pool_layer2(img, kernel):
    # Conv_and_Pooling
    conv_img = cv2.filter2D(img[:, :, 0], -1, kernel) + cv2.filter2D(img[:, :, 1], -1,
kernel)
    max_img = pooling(conv_img, 2)
    return max_img

```


This function prints and plots the confusion matrix.
Normalization can be applied by setting `normalize=True`.

```
"""
```

```
if normalize:
```

```
    cm = cm.astype('float') / cm.sum(axis=1)[:, np.newaxis]
    print("Normalized confusion matrix")
```

```
else:
```

```
    print('Confusion matrix, without normalization')
```

```
cm = cm.T
```

```
print(cm)
```

```
plt.imshow(cm, interpolation='nearest', cmap=cmap)
```

```
plt.title(title)
```

```
tick_marks = np.arange(len(classes))
```

```
plt.xticks(tick_marks, classes)
```

```
plt.yticks(tick_marks, classes)
```

```
fmt = '.2f' if normalize else 'd'
```

```
thresh = cm.max() / 2.
```

```
for i, j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):
```

```
    plt.text(j, i, format(cm[i, j], fmt),
             horizontalalignment="center",
             color="white" if cm[i, j] > thresh else "black")
```

```
plt.tight_layout()
```

#隨機找 15 張圖，看其預測結果

```
def plot_images_labels_prediction(labels, prediction, num1, num2):
```

```
    # fig = plt.gcf()
```

```
    # fig.set_size_inches(12, 14)
```

```
    plt.figure()
```

```
    for i in range(0, 15):
```

```
        r = random.randint(0, 199)
```

```
        png = num1 * 100 + r if r < 100 else (num2 - 1) * 100 + r
```

```
        img = cv2.imread(testPath + str(png + 1) + ".png", 0)
```

```
        ax = plt.subplot(3, 5, 1 + i)
```

```
        ax.imshow(img, cmap='gray')
```

```
title = str(prediction[r]) + "(" + str(png) + ")"
```

```
if labels[r] == prediction[r]:
```

```
    color = 'b'
```

```
else:
```

```
    color = 'r'
```

```
ax.set_title(title, fontsize=10, color=color)
```

```
ax.set_xticks([])
```

```
ax.set_yticks([])
```

```
def generate_confusion_mat(y_true, y_pred):
```

```
    target_names = [str(num1), str(num2)]
```

```
    cnf_matrix = confusion_matrix(y_true, y_pred)
```

```
    plot_confusion_matrix(cnf_matrix, classes=target_names, normalize=False,  
                           title='confusion matrix')
```

```
    plot_images_labels_prediction(y_true, y_pred, num1, num2)
```

```
    plt.show()
```

```
if __name__ == '__main__':
```

```
    train = load_image(num1, num2)
```

```
    y_true, y_pred = model_train(num1, num2, train, kernel1, kernel2)
```

```
    generate_confusion_mat(y_true, y_pred)
```