

彩色影像處理

作業 3：簡化的 CNN 數值辨識

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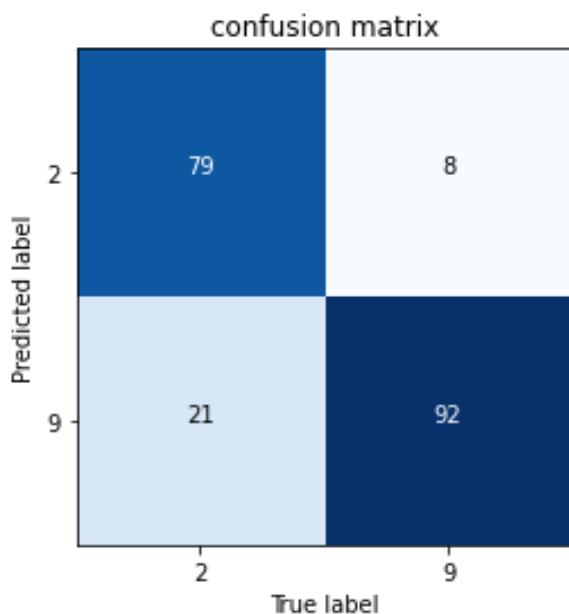
完成了可以隨意指定 0~9 之間兩類別的比較，若要更改數值，請在下圖地方更改，路徑改成存放位置的地方。其餘程式碼部分已有註解，有疑問再請詳看。

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

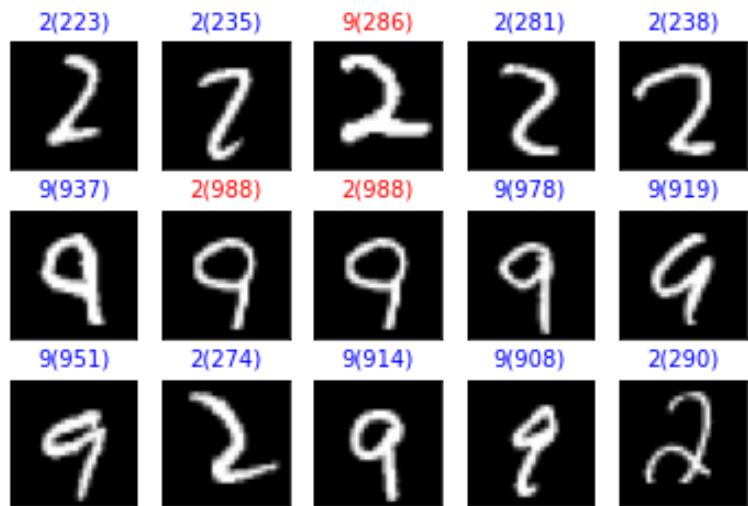
kernel1 = np.array([[-1, -1, 1, -1, 0, 1, -1, 1, 1]]).reshape(3, 3)
#預設的filter1
kernel12 = np.random.random(size=(3, 3))
kernel12 = (np.array((kernel12 / np.sum(kernel12))) - (1/9)) * 20
#自訂的filter2

# num1 , num2 為可指定 0 ~9
num1 = 2
num2 = 9
```

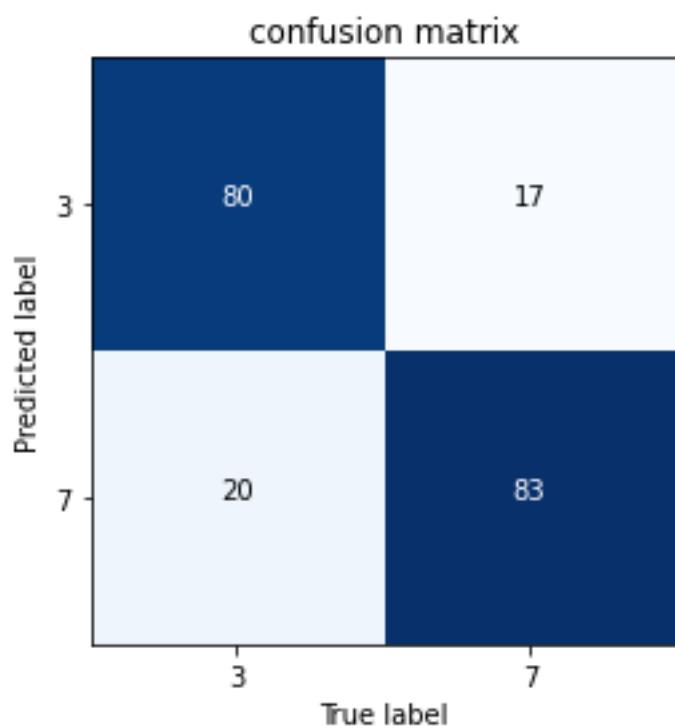
類別 : 2、9

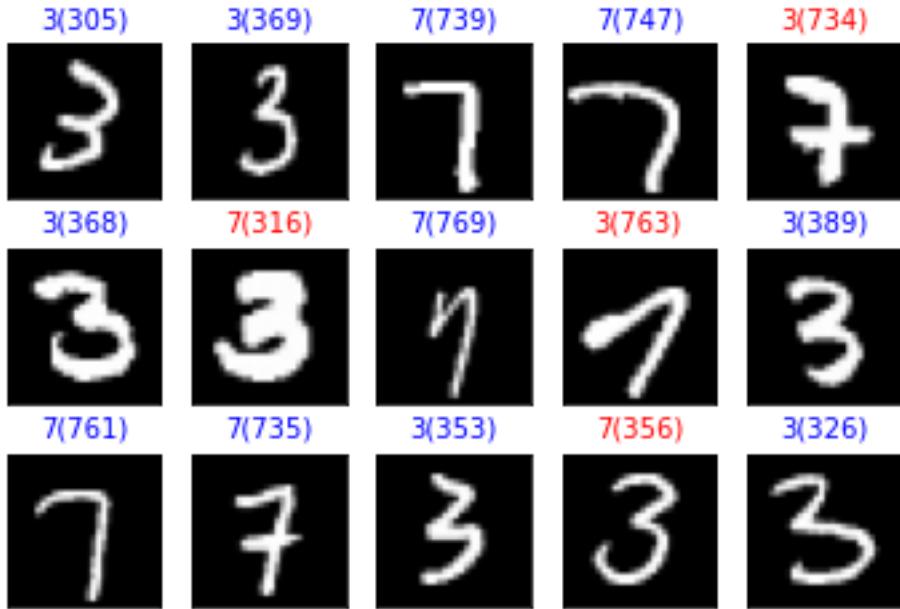


隨機 15 個影像分類結果



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 /

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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')

```

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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)

```