

# Homework 2 Report

資工所碩二 R06922132 何羿辰

EE5184 - Machine Learning

1. (18) 請說明你實作的 CNN model，其模型架構、訓練過程和準確率為何？

| Layer (type)                                 | Output Shape        | Param # |
|--|---------------------|---------|
| conv2d_1 (Conv2D)                            | (None, 32, 48, 48)  | 320     |
| batch_normalization_1 (Batch Normalization)  | (None, 32, 48, 48)  | 128     |
| activation_1 (Activation)                    | (None, 32, 48, 48)  | 0       |
| conv2d_2 (Conv2D)                            | (None, 32, 48, 48)  | 9248    |
| batch_normalization_2 (Batch Normalization)  | (None, 32, 48, 48)  | 128     |
| activation_2 (Activation)                    | (None, 32, 48, 48)  | 0       |
| conv2d_3 (Conv2D)                            | (None, 32, 48, 48)  | 9248    |
| batch_normalization_3 (Batch Normalization)  | (None, 32, 48, 48)  | 128     |
| activation_3 (Activation)                    | (None, 32, 48, 48)  | 0       |
| max_pooling2d_1 (MaxPooling2D)               | (None, 32, 24, 24)  | 0       |
| conv2d_4 (Conv2D)                            | (None, 64, 24, 24)  | 18496   |
| batch_normalization_4 (Batch Normalization)  | (None, 64, 24, 24)  | 256     |
| activation_4 (Activation)                    | (None, 64, 24, 24)  | 0       |
| conv2d_5 (Conv2D)                            | (None, 64, 24, 24)  | 36928   |
| batch_normalization_5 (Batch Normalization)  | (None, 64, 24, 24)  | 256     |
| activation_5 (Activation)                    | (None, 64, 24, 24)  | 0       |
| conv2d_6 (Conv2D)                            | (None, 64, 24, 24)  | 36928   |
| batch_normalization_6 (Batch Normalization)  | (None, 64, 24, 24)  | 256     |
| activation_6 (Activation)                    | (None, 64, 24, 24)  | 0       |
| max_pooling2d_2 (MaxPooling2D)               | (None, 64, 12, 12)  | 0       |
| conv2d_7 (Conv2D)                            | (None, 128, 12, 12) | 73856   |
| batch_normalization_7 (Batch Normalization)  | (None, 128, 12, 12) | 512     |
| activation_7 (Activation)                    | (None, 128, 12, 12) | 0       |
| conv2d_8 (Conv2D)                            | (None, 128, 12, 12) | 147584  |
| batch_normalization_8 (Batch Normalization)  | (None, 128, 12, 12) | 512     |
| activation_8 (Activation)                    | (None, 128, 12, 12) | 0       |
| conv2d_9 (Conv2D)                            | (None, 128, 12, 12) | 147584  |
| batch_normalization_9 (Batch Normalization)  | (None, 128, 12, 12) | 512     |
| activation_9 (Activation)                    | (None, 128, 12, 12) | 0       |
| max_pooling2d_3 (MaxPooling2D)               | (None, 128, 6, 6)   | 0       |
| flatten_1 (Flatten)                          | (None, 4608)        | 0       |
| dense_1 (Dense)                              | (None, 1024)        | 4719616 |
| batch_normalization_10 (Batch Normalization) | (None, 1024)        | 4096    |
| activation_10 (Activation)                   | (None, 1024)        | 0       |
| dropout_1 (Dropout)                          | (None, 1024)        | 0       |
| dense_2 (Dense)                              | (None, 7)           | 7175    |
| Total params: 5,213,767                      |                     |         |
| Trainable params: 5,210,375                  |                     |         |
| Non-trainable params: 3,392                  |                     |         |

訓練過程:

固定參數: Batch size 固定為 128 · optimizer 固定使用 ADAM

Preprocess:

```
data_generator_train = ImageDataGenerator(
    rotation_range=15, width_shift_range=0.1, height_shift_range=0.1, horizontal_flip=True, data_format='channels_first')
```

變動參數: 前 75 個 epoch 固定使用 lr 為 0.01 · 之後的 75~300epoch 從 lr=0.001 開始 reduce lr 尋找

最佳解

訓練前有先切出 10%資料作為 validation set · 剩下的 90%則為 training set

準確率:最後準確率為 0.68152

2. (1%) 承上題，請用與上述 CNN 接近的參數量，實做簡單的 DNN model，其模型架構、訓練過程和準確率為何？試與上題結果做比較，並說明你觀察到了什麼？

| Layer (type)                                | Output Shape | Param # |   |             |        |
|---|--------------|---------|---|-------------|--------|
| flatten_1 (Flatten)                         | (None, 2304) | 0       | dense_6 (Dense)                             | (None, 512) | 262656 |
| dense_1 (Dense)                             | (None, 1024) | 2360320 | batch_normalization_6 (Batch Normalization) | (None, 512) | 2048   |
| batch_normalization_1 (Batch Normalization) | (None, 1024) | 4096    | activation_6 (Activation)                   | (None, 512) | 0      |
| activation_1 (Activation)                   | (None, 1024) | 0       | dropout_2 (Dropout)                         | (None, 512) | 0      |
| dense_2 (Dense)                             | (None, 1024) | 1049600 | dense_7 (Dense)                             | (None, 256) | 131328 |
| batch_normalization_2 (Batch Normalization) | (None, 1024) | 4096    | batch_normalization_7 (Batch Normalization) | (None, 256) | 1024   |
| activation_2 (Activation)                   | (None, 1024) | 0       | activation_7 (Activation)                   | (None, 256) | 0      |
| dense_3 (Dense)                             | (None, 1024) | 1049600 | dense_8 (Dense)                             | (None, 256) | 65792  |
| batch_normalization_3 (Batch Normalization) | (None, 1024) | 4096    | batch_normalization_8 (Batch Normalization) | (None, 256) | 1024   |
| activation_3 (Activation)                   | (None, 1024) | 0       | activation_8 (Activation)                   | (None, 256) | 0      |
| dropout_1 (Dropout)                         | (None, 1024) | 0       | dense_9 (Dense)                             | (None, 256) | 65792  |
| dense_4 (Dense)                             | (None, 512)  | 524800  | batch_normalization_9 (Batch Normalization) | (None, 256) | 1024   |
| batch_normalization_4 (Batch Normalization) | (None, 512)  | 2048    | activation_9 (Activation)                   | (None, 256) | 0      |
| activation_4 (Activation)                   | (None, 512)  | 0       | dropout_3 (Dropout)                         | (None, 256) | 0      |
| dense_5 (Dense)                             | (None, 512)  | 262656  | dense_10 (Dense)                            | (None, 7)   | 1799   |
| batch_normalization_5 (Batch Normalization) | (None, 512)  | 2048    |   |             |        |
| activation_5 (Activation)                   | (None, 512)  | 0       |   |             |        |
|   |              |         | Total params: 5,795,847                     |             |        |
|   |              |         | Trainable params: 5,785,095                 |             |        |
|   |              |         | Non-trainable params: 10,752                |             |        |

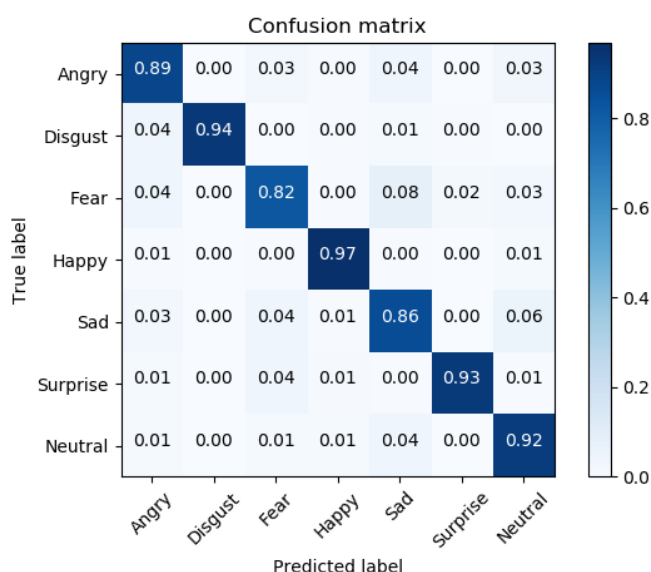
訓練過程:

為了比較結果，所以訓練過程完全比照 CNN 的作法

準確率: 0.49902，雖然參數量比 CNN 稍微多了一些，但準確率卻差很多，可以判斷 DNN 對於影

像辨識的準確度比 CNN 效率要差

3. (1%) 觀察答錯的圖片中，哪些 class 彼此間容易用混？並說明你觀察到了什麼？[繪出 confusion matrix 分析]



從 confusion matrix 中可以看出，Happy 的判斷最準確，Fear 判斷的準確度最差，而 Fear 最容易被誤判為 Sad，也許是 Fear 與 Sad 兩者的情緒表現本來就很相似，另外 Sad 也很容易被誤判為 Neutral

4. (1.5%, each 0.5%) CNN time/space complexity:

For a. b. Given a CNN model as

```
model = Sequential()
model.add(Conv2D(filters=6,
                  strides=(3, 3),
                  padding = "valid",
                  kernel_size=(2,2),
                  input_shape=(8,8,5),
                  activation='relu'))
model.add(Conv2D(filters=4,
                  strides=(2, 2),
                  padding = "valid",
                  kernel_size=(2,2),
                  activation='relu'))
```

And for the c. given the parameter as:

kernel size = (k,k);

channel size = c;

input shape of each layer = (n,n);

padding = p;

strides = (s,s);

- a. How many parameters are there in each layer(Hint: you may consider whether the number of parameter is related with)

Layer A:

Layer B:

- b. How many multiplications/additions are needed for a forward pass(each layer).

Layer A:

Layer B:

- c. What is the time complexity of convolutional neural networks?(note: you must use big-O upper bound, and there are l(lower case of L) layer, you can use  $C_l$ ,  $C_{l-1}$  as lth and l-1th layer)

(a) 計算公式為:  $((\text{kernel\_size}) * \text{stride} + 1) * \text{filters}$

Layer A:  $((2 * 2) * 5 + 1) * 6 = 126\#$

Layer B:  $((2*2)*6+1)*4=100\#$

(b)

Layer A:

Additions:  $(2*2*5-1)*3*3*6=1026\#$

Multiplications:  $(2*2*5)*3*3*6=1080\#$

Layer B:

Additions:  $(2*2*6-1)*1*1*4=92\#$

Multiplications:  $(2*2*6)*1*1*4=96\#$

(c) 所求為所有卷積層的時間複雜度相加，而每個卷積層的複雜度為：

輸出特徵圖面積乘上卷積核面積乘上輸入/輸出通道數(filter size)

假設  $c_1$  為 input data channel size

$$O\left(\sum_{i=2}^l \left(\frac{n_i - k_i + 2 * p_i}{s_i} + 1\right)^2 * k_i^2 * c_{i-1} * c_i\right) \#$$

5. (1.5%, each 0.5%) PCA practice: Problem statement: Given 10 samples in 3D space.  $(1, 2, 3), (4, 8, 5), (3, 12, 9), (1, 8, 5), (5, 14, 2), (7, 4, 1), (9, 8, 9), (3, 8, 1), (11, 5, 6), (10, 11, 7)$
- (1) What are the principal axes?
  - (2) Compute the principal components for each sample.
  - (3) Reconstruction error if reduced to 2D. (Calculate the L2-norm)

(a) 以下三題皆有先做 centering

Step1: 歸一化減去均值

$$\text{mean}_1 = \frac{1 + 4 + 3 + 1 + 5 + 7 + 9 + 3 + 11 + 10}{10} = 5.4$$

$$\text{mean}_2 = \frac{2 + 8 + 12 + 8 + 14 + 4 + 8 + 8 + 5 + 11}{10} = 8.0$$

$$\text{mean}_3 = \frac{3 + 5 + 9 + 5 + 2 + 1 + 9 + 1 + 6 + 7}{10} = 4.8$$

10 samples 減去均值後:  $(-4.4, -6, -1.8), (-1.4, 0, 0.2), (-2.4, 4, 4.2), (-4.4, 0, 0.2), (-0.4, 6, -2.8), (1.6, -4, -$

3.8),(3.6,0,4.2),(-2.4,0,-3.8),(5.6,-3,1.2),(4.6,3,2.2)

**Step2:斜方差求 eigenvalue 與 eigenvector**

斜方差矩陣求得為: 
$$\begin{bmatrix} 13.37 & 0.56 & 3.64 \\ 0.55 & 13.55 & 3.22 \\ 3.64 & 3.22 & 9.06 \end{bmatrix}$$

矩陣的 eigenvalue 與 eigenvector 依 eigenvalue 大到小排序為:

Eigenvalue: 16.99715933,12.9228041,6.08

Eigenvector: (-0.61,-0.58,-0.52),( 0.67,-0.73,0.02),( 0.40,0.33,-0.85) #

Principle axes 即為這些 eigenvector(老師投影片裡的 $u_i$ )

(b) 將各個 principle axis 與歸一化後的  $x$  相乘做投影可得 principle component(老師投影片裡的 $u_i^T x$ )

**1'st principle component:**

[7.18 0.75 -3.07 2.6 -1.82 3.35 -4.41 3.46 -2.31 -5.75] #

**2'nd principle component:**

[1.37 -0.94 -4.45 -2.97 -4.75 3.91 2.55 -1.73 6.03 0.97] #

**3'rd principle component:**

[-2.25 -0.73 -3.18 -1.92 4.25 2.52 -2.13 2.27 0.20 0.97] #

(c) Reduce 到 2D → 取前兩個 vector 重建  $\tilde{x}$  ( $\tilde{x} = u_i u_i^T x$ )

重建後的  $\tilde{x}=(1.9,2.75,1.08),(4.29,8.24,4.37),(4.27,13.07,6.28)$

取  $x$  與  $\tilde{x}$  的:

L2 norm error(兩者相減後平方相加) 可得 L2-norm error 為 54.72 #

若採用 L2 norm(兩者相減後平方相加取開根號) 可得 L2-norm 為 20.47 #