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1. (1%) 請說明你實作的 CNN model，其模型架構、訓練參數和準確率為何？

(Collaborators: b05902127 )

(1) Layer (type)	Output Shape	Param #
batch_normalization_1 (Batch Normalization)	(None, 48, 48, 1)	4
conv2d_1 (Conv2D)	(None, 48, 48, 32)	160
batch_normalization_2 (Batch Normalization)	(None, 48, 48, 32)	128
conv2d_2 (Conv2D)	(None, 48, 48, 32)	4128
batch_normalization_3 (Batch Normalization)	(None, 48, 48, 32)	128
max_pooling2d_1 (MaxPooling2D)	(None, 24, 24, 32)	0
batch_normalization_4 (Batch Normalization)	(None, 24, 24, 32)	128
conv2d_3 (Conv2D)	(None, 24, 24, 32)	4128
batch_normalization_5 (Batch Normalization)	(None, 24, 24, 32)	128
conv2d_4 (Conv2D)	(None, 24, 24, 32)	4128
batch_normalization_6 (Batch Normalization)	(None, 24, 24, 32)	128
max_pooling2d_2 (MaxPooling2D)	(None, 12, 12, 32)	0
batch_normalization_7 (Batch Normalization)	(None, 12, 12, 32)	128
flatten_1 (Flatten)	(None, 4608)	0
batch_normalization_8 (Batch Normalization)	(None, 4608)	18432
dropout_1 (Dropout)	(None, 4608)	0
dense_1 (Dense)	(None, 512)	2359808
dense_2 (Dense)	(None, 512)	262656

batch_normalization_9 (Batch Normalization)	(None, 512)	2048
dense_3 (Dense)	(None, 7)	3591
accuracy: 0.60936		
(2) batch_normalization_1 (Batch Normalization)	(None, 48, 48, 1)	4
conv2d_1 (Conv2D)	(None, 48, 48, 63)	630
batch_normalization_2 (Batch Normalization)	(None, 48, 48, 63)	252
conv2d_2 (Conv2D)	(None, 48, 48, 127)	72136
batch_normalization_3 (Batch Normalization)	(None, 48, 48, 127)	508
max_pooling2d_1 (MaxPooling2D)	(None, 24, 24, 127)	0
batch_normalization_4 (Batch Normalization)	(None, 24, 24, 127)	508
conv2d_3 (Conv2D)	(None, 24, 24, 255)	291720
batch_normalization_5 (Batch Normalization)	(None, 24, 24, 255)	1020
conv2d_4 (Conv2D)	(None, 24, 24, 511)	1173256
batch_normalization_6 (Batch Normalization)	(None, 24, 24, 511)	2044
max_pooling2d_2 (MaxPooling2D)	(None, 12, 12, 511)	0
batch_normalization_7 (Batch Normalization)	(None, 12, 12, 511)	2044
flatten_1 (Flatten)	(None, 73584)	0
batch_normalization_8 (Batch Normalization)	(None, 73584)	294336
dropout_1 (Dropout)	(None, 73584)	0
dense_1 (Dense)	(None, 512)	37675520
dense_2 (Dense)	(None, 512)	262656

batch_normalization_9 (Batch Normalization)	(None, 512)	2048
dropout_2 (Dropout)	(None, 512)	0
dense_3 (Dense)	(None, 7)	3591
accuracy: <b>0.64140</b>		
(3)input_1 (InputLayer)	(None, 48, 48, 1)	0
batch_normalization_1 (Batch Normalization)	(None, 48, 48, 1)	4
conv2d_1 (Conv2D)	(None, 48, 48, 64)	640
batch_normalization_2 (Batch Normalization)	(None, 48, 48, 64)	256
conv2d_2 (Conv2D)	(None, 48, 48, 64)	36928
batch_normalization_3 (Batch Normalization)	(None, 48, 48, 64)	256
max_pooling2d_1 (MaxPooling2D)	(None, 24, 24, 64)	0
batch_normalization_4 (Batch Normalization)	(None, 24, 24, 64)	256
conv2d_3 (Conv2D)	(None, 24, 24, 128)	73856
batch_normalization_5 (Batch Normalization)	(None, 24, 24, 128)	512
conv2d_4 (Conv2D)	(None, 24, 24, 128)	147584
batch_normalization_6 (Batch Normalization)	(None, 24, 24, 128)	512
max_pooling2d_2 (MaxPooling2D)	(None, 12, 12, 128)	0
batch_normalization_7 (Batch Normalization)	(None, 12, 12, 128)	512
conv2d_5 (Conv2D)	(None, 12, 12, 256)	295168
batch_normalization_8 (Batch Normalization)	(None, 12, 12, 256)	1024
conv2d_6 (Conv2D)	(None, 12, 12, 256)	590080

batch_normalization_9 (Batch Normalization)	(None, 12, 12, 256)	1024
conv2d_7 (Conv2D)	(None, 12, 12, 256)	590080
batch_normalization_10 (Batch Normalization)	(None, 12, 12, 256)	1024
max_pooling2d_3 (MaxPooling2D)	(None, 6, 6, 256)	0
batch_normalization_11 (Batch Normalization)	(None, 6, 6, 256)	1024
conv2d_8 (Conv2D)	(None, 6, 6, 512)	1180160
batch_normalization_12 (Batch Normalization)	(None, 6, 6, 512)	2048
conv2d_9 (Conv2D)	(None, 6, 6, 512)	2359808
batch_normalization_13 (Batch Normalization)	(None, 6, 6, 512)	2048
conv2d_10 (Conv2D)	(None, 6, 6, 512)	2359808
batch_normalization_14 (Batch Normalization)	(None, 6, 6, 512)	2048
max_pooling2d_4 (MaxPooling2D)	(None, 3, 3, 512)	0
batch_normalization_15 (Batch Normalization)	(None, 3, 3, 512)	2048
conv2d_11 (Conv2D)	(None, 3, 3, 512)	2359808
batch_normalization_16 (Batch Normalization)	(None, 3, 3, 512)	2048
conv2d_12 (Conv2D)	(None, 3, 3, 512)	2359808
batch_normalization_17 (Batch Normalization)	(None, 3, 3, 512)	2048
conv2d_13 (Conv2D)	(None, 3, 3, 512)	2359808
batch_normalization_18 (Batch Normalization)	(None, 3, 3, 512)	2048
max_pooling2d_5 (MaxPooling2D)	(None, 2, 2, 512)	0
batch_normalization_19 (Batch Normalization)	(None, 2, 2, 512)	2048

flatten_1 (Flatten)	(None, 2048)	0
batch_normalization_20 (Batch Normalization)	(None, 2048)	8192
dropout_1 (Dropout)	(None, 2048)	0
dense_1 (Dense)	(None, 512)	1049088
dense_2 (Dense)	(None, 512)	262656
batch_normalization_21 (Batch Normalization)	(None, 512)	2048
dense_3 (Dense)	(None, 512)	262656
batch_normalization_22 (Batch Normalization)	(None, 512)	2048
dense_4 (Dense)	(None, 7)	3591

accuracy: 0.65254

- (1%) 請嘗試 data normalization, data augmentation, 說明實行方法並且說明對準確率有什麼樣的影響？

藉由添加 batch\_normalization 層完成 data normalization，準確率由 56% 上升至 60%; from keras.preprocessing.image import ImageDataGenerator 來處理 data augmentation，準確率由 60% 上升至 65%。

- (1%) 觀察答錯的圖片中，哪些 class 彼此間容易用混？[繪出 confusion matrix 分析]
- (1%) 從(1)(2)可以發現，使用 CNN 的確有些好處，試繪出其 saliency maps，觀察模型在做 classification 時，是 focus 在圖片的哪些部份？
- (1%) 承(4) 利用上課所提到的 gradient ascent 方法，觀察特定層的 filter 最容易被哪種圖片 activate 與觀察 filter 的 output。