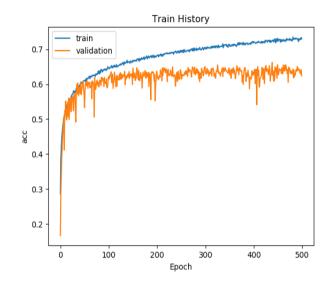
學號:R05546014系級:工工碩二 姓名:洪紹綺

## 註:四到五題都取#777 圖片

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

答:

Layer (type)	Output	Shape	Param #
conv2d_1 (Conv2D)	(None,	48, 48, 64)	1664
batch_normalization_1 (Batch	(None,	48, 48, 64)	256
leaky_re_lu_1 (LeakyReLU)	(None,	48, 48, 64)	0
zero_padding2d_1 (ZeroPaddin	(None,	52, 52, 64)	0
max_pooling2d_1 (MaxPooling2	(None,	24, 24, 64)	0
zero_padding2d_2 (ZeroPaddin	(None,	26, 26, 64)	0
conv2d_2 (Conv2D)	(None,	24, 24, 128)	73856
batch_normalization_2 (Batch	(None,	24, 24, 128)	512
leaky_re_lu_2 (LeakyReLU)	(None,	24, 24, 128)	0
zero_padding2d_3 (ZeroPaddin	(None,	26, 26, 128)	0
dropout_1 (Dropout)	(None,	26, 26, 128)	0
conv2d_3 (Conv2D)	(None,	24, 24, 64)	73792
batch_normalization_3 (Batch	(None,	24, 24, 64)	256
leaky_re_lu_3 (LeakyReLU)	(None,	24, 24, 64)	0
average_pooling2d_1 (Average	(None,	11, 11, 64)	0
zero_padding2d_4 (ZeroPaddin	(None,	13, 13, 64)	0
conv2d_4 (Conv2D)	(None,	11, 11, 128)	73856
batch_normalization_4 (Batch	(None,	11, 11, 128)	512
leaky_re_lu_4 (LeakyReLU)	(None,	11, 11, 128)	0
max_pooling2d_2 (MaxPooling2	(None,	5, 5, 128)	0
zero_padding2d_5 (ZeroPaddin	(None,	7, 7, 128)	0
dropout_2 (Dropout)	(None,	7, 7, 128)	0
conv2d_5 (Conv2D)	(None,	5, 5, 64)	73792
batch_normalization_5 (Batch	(None,	5, 5, 64)	256
leaky_re_lu_5 (LeakyReLU)	(None,	5, 5, 64)	0
zero_padding2d_6 (ZeroPaddin	(None,	7, 7, 64)	0
average_pooling2d_2 (Average	(None,	3, 3, 64)	0
flatten_1 (Flatten)	(None,	576)	0
dense_1 (Dense)	(None,		590848
dense_2 (Dense)	(None,	512)	524800
dropout_3 (Dropout)	(None,	512)	0
dense_3 (Dense)	(None,	1024)	525312
dropout_4 (Dropout)	(None,	1024)	0
dense_4 (Dense)	(None,	7)	7175
activation_1 (Activation)	(None,	7)	0
Total params: 1,946,887 Trainable params: 1,945,991 Non-trainable params: 896			



Train 的準確率大概在 0.7 左右,持續上升,但 validation 在 0.6 左右擺盪了 400 個跌代左右。

Kaggle 準確率:( 0.65784+0.65032)/2=0.65408

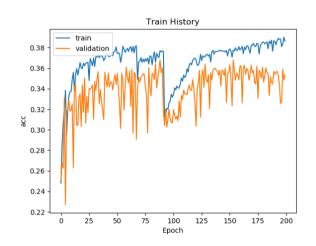
2. (1%) 承上題,請用與上述 CNN 接近的參數量,實做簡單的 DNN model。其模型架構、訓練過程和準確率為何?試與上題結果做比較,並說明你觀察到了什麼?

(Collaborators: )

答:DNN 參數雖較 CNN 多,但準確率極差。

DNN 模型比 CNN 略多 1%參數, Train 跟 Validation 準確率在 100 個跌代間在 0.35 上下擺盪停滯。

Layer (type)	Output Shape	Param #
dense_1 (Dense)	(None, 256)	590080
dense_2 (Dense)	(None, 256)	65792
dense_3 (Dense)	(None, 1024)	263168
dense_4 (Dense)	(None, 512)	524800
dropout_1 (Dropout)	(None, 512)	0
dense_5 (Dense)	(None, 1024)	525312
dropout_2 (Dropout)	(None, 1024)	0
dense_6 (Dense)	(None, 7)	7175
activation 1 (Activation)	(None, 7)	0



3. (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]

(Collaborators: )

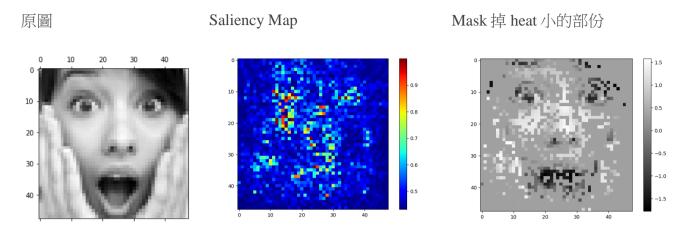
答:從矩陣可以看出生氣、難過與恐懼有些會混淆,符合現實人類表情狀況; 驚訝與恐懼之間有時也會混淆,難過與無表情也會混淆。

	Angry	Hate	Fear	Joy	Sad	Surprise	Neutra	al
 predic label 0 1 2	[Info					Matri 5 16 1 32		All 315 24 318
3 4 5 6 All	16 41 8 23 287	1 1 0 0 22	13 30 17 20 251	424 12 14 27 500	20 190 3 49 384		26 51 7 220 377	511 327 236 347 2078

4. (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?

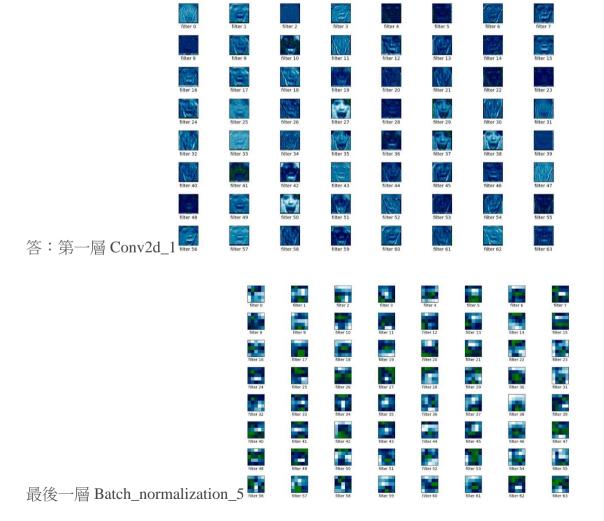
(Collaborators:林冠廷、洪唯凱、張少豪)

答:從圖中可以看出,模型分類的時候主要是抓眼睛跟嘴巴,反應較明顯。



5. (1%) 承(1)(2),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。

(Collaborators: 林冠廷、洪唯凱、張少豪)



眼睛嘴巴較明顯的圖片較易 activate。