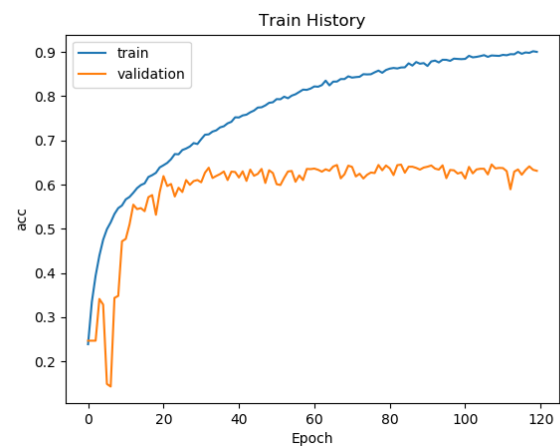
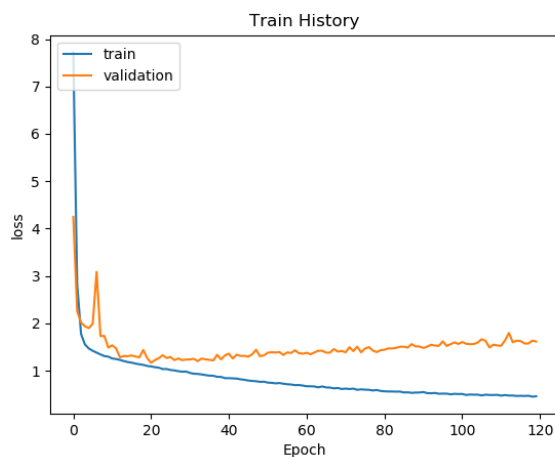


1. (1%) 請說明你實作的 CNN model，其模型架構、訓練過程和準確率為何？

Layer (type)	Output Shape	Param #		
conv2d_1 (Conv2D)	(None, 44, 44, 64)	1664	conv2d_2 (Conv2D)	(None, 20, 20, 128)
leaky_re_lu_1 (LeakyReLU)	(None, 44, 44, 64)	0	leaky_re_lu_2 (LeakyReLU)	(None, 20, 20, 128)
batch_normalization_1 (Batch Normalization)	(None, 44, 44, 64)	256	batch_normalization_2 (Batch Normalization)	(None, 20, 20, 128)
max_pooling2d_1 (MaxPooling2D)	(None, 22, 22, 64)	0	max_pooling2d_2 (MaxPooling2D)	(None, 10, 10, 128)
dropout_1 (Dropout)	(None, 22, 22, 64)	0	dropout_2 (Dropout)	(None, 10, 10, 128)
conv2d_3 (Conv2D)	(None, 8, 8, 512)	590336	conv2d_4 (Conv2D)	(None, 2, 2, 512)
leaky_re_lu_3 (LeakyReLU)	(None, 8, 8, 512)	0	leaky_re_lu_4 (LeakyReLU)	(None, 2, 2, 512)
batch_normalization_3 (Batch Normalization)	(None, 8, 8, 512)	2048	batch_normalization_4 (Batch Normalization)	(None, 2, 2, 512)
max_pooling2d_3 (MaxPooling2D)	(None, 4, 4, 512)	0	average_pooling2d_1 (Average Pooling2D)	(None, 1, 1, 512)
			dropout_3 (Dropout)	(None, 4, 4, 512)
			dropout_4 (Dropout)	(None, 1, 1, 512)
			flatten_1 (Flatten)	(None, 512)
			dense_1 (Dense)	(None, 512)
			dropout_5 (Dropout)	(None, 512)
			dense_2 (Dense)	(None, 512)
			activation_1 (Activation)	(None, 512)
			dropout_6 (Dropout)	(None, 512)
			dense_3 (Dense)	(None, 7)
			activation_2 (Activation)	(None, 7)

- Total params: 3,559,431
 - Trainable params: 3,556,999
 - Non-trainable params: 2,432



Final Valid ACC are around 65%

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:)

Epoch 180/180

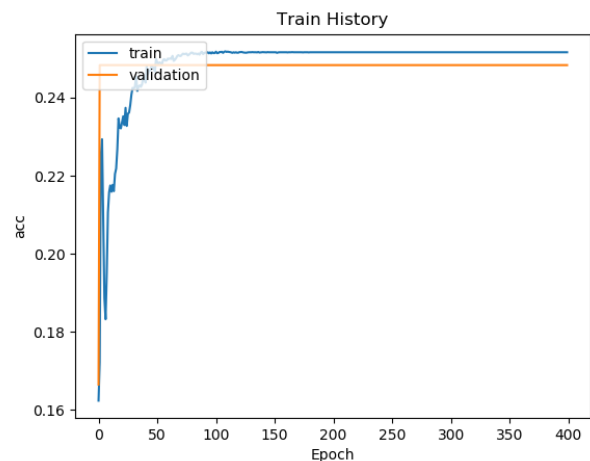
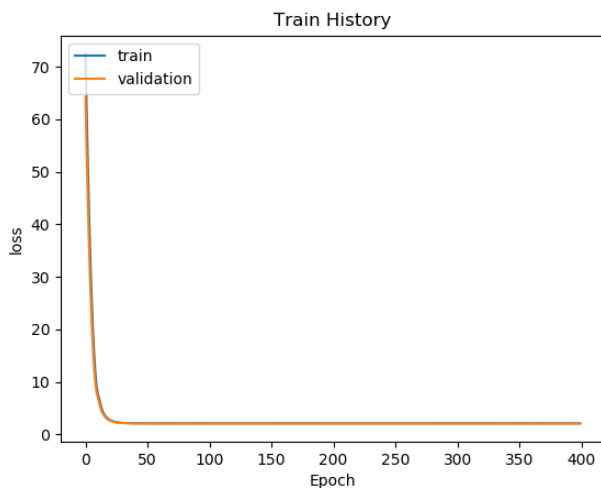
25838/25838 [=====] - 0s 16us/step

- loss: 2.1008 - acc: 0.2516 - val_loss: 2.0967 - val_acc: 0.2483

DNN sucks, it the model will be in local optimal 0.2483 even more often than CNN for image processing.

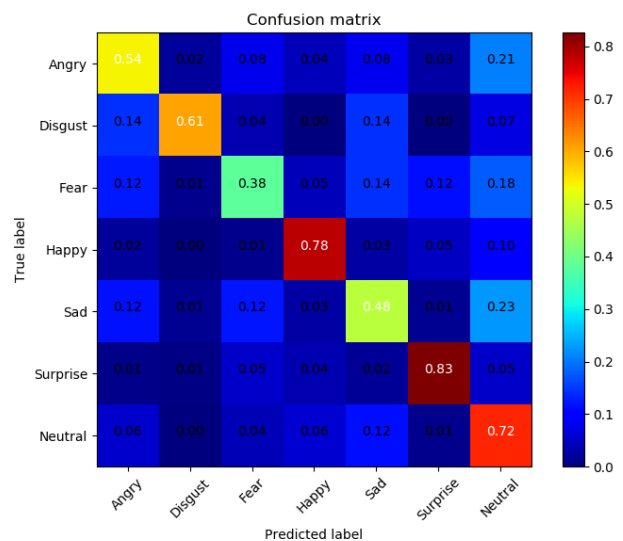
Layer (type)	Output Shape	Param #
dense_118 (Dense)	(None, 1024)	2360320
dropout_78 (Dropout)	(None, 1024)	0
dense_119 (Dense)	(None, 512)	524800
activation_78 (Activation)	(None, 512)	0
dropout_79 (Dropout)	(None, 512)	0
dense_120 (Dense)	(None, 512)	262656
activation_79 (Activation)	(None, 512)	0
dropout_80 (Dropout)	(None, 512)	0
dense_121 (Dense)	(None, 128)	65664
activation_80 (Activation)	(None, 128)	0
dropout_81 (Dropout)	(None, 128)	0
dense_122 (Dense)	(None, 256)	33024

activation_81 (Activation)	(None, 256)	0
dropout_82 (Dropout)	(None, 256)	0
dense_123 (Dense)	(None, 256)	65792
activation_82 (Activation)	(None, 256)	0
dropout_83 (Dropout)	(None, 256)	0
dense_124 (Dense)	(None, 128)	32896
activation_83 (Activation)	(None, 128)	0
dropout_84 (Dropout)	(None, 128)	0
dense_125 (Dense)	(None, 7)	903
activation_84 (Activation)	(None, 7)	0

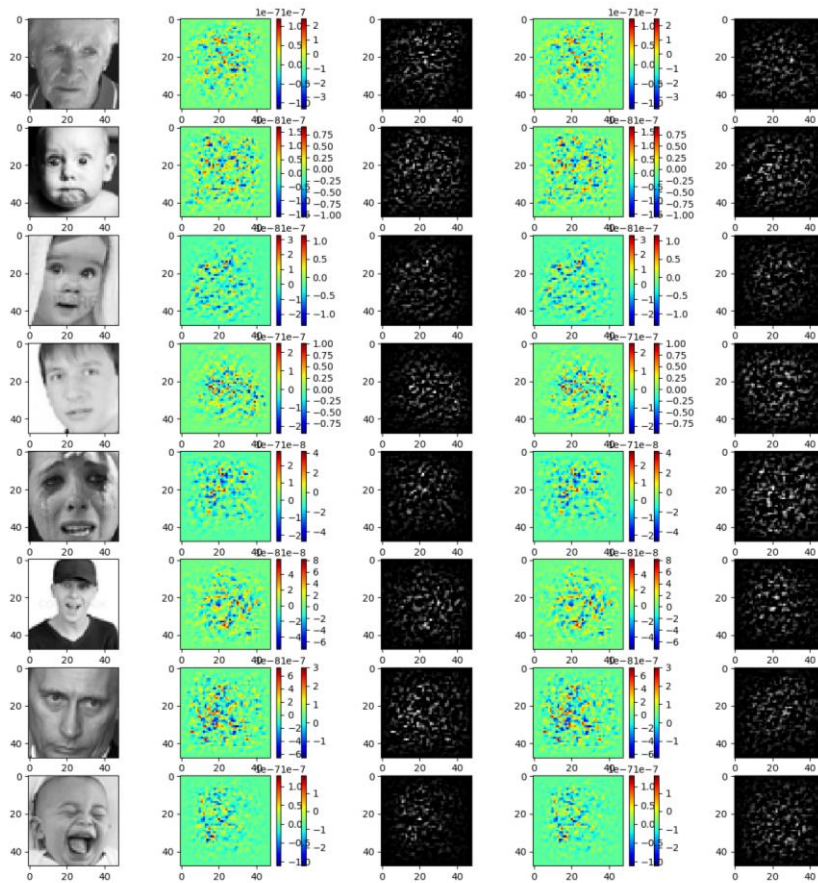


3. (1%) 觀察答錯的圖片中，哪些 class 彼此間容易用混？[繪出 confusion matrix 分析]
(Collaborators:)

It appears that class “Neutral” are a bit easy to be classified as incorrect ; but are good Overall.



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: r05546003 r05546024 r05546026)
If mostly focus on the central facial area, or the nose area.



答：

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: r05546003 r05546024 r05546026)

答：It appears filter 9, 21, 29 are activated.

Input image and conv2d_2 filters

