

HW3 Report

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1. (2%) 請說明你實作的 **CNN model**，其模型架構、訓練參數和準確率為何？並請用與上述 **CNN** 接近的參數量，實做簡單的 **DNN model**，同時也說明其模型架構、訓練參數和準確率為何？並說明你觀察到了什麼？

CNN:

Layer (type)	Output Shape	Param #
conv2d_1 (Conv2D)	(None, 48, 48, 64)	640
leaky_re_lu_1 (LeakyReLU)	(None, 48, 48, 64)	0
batch_normalization_1 (Batch Normalization)	(None, 48, 48, 64)	256
max_pooling2d_1 (MaxPooling2D)	(None, 24, 24, 64)	0
dropout_1 (Dropout)	(None, 24, 24, 64)	0
flatten_1 (Flatten)	(None, 4608)	0
dense_1 (Dense)	(None, 256)	1179904
batch_normalization_5 (Batch Normalization)	(None, 256)	1024
dropout_5 (Dropout)	(None, 256)	0
Total params: 2,871,815		
Trainable params: 2,868,359		
Non-trainable params: 3,456		

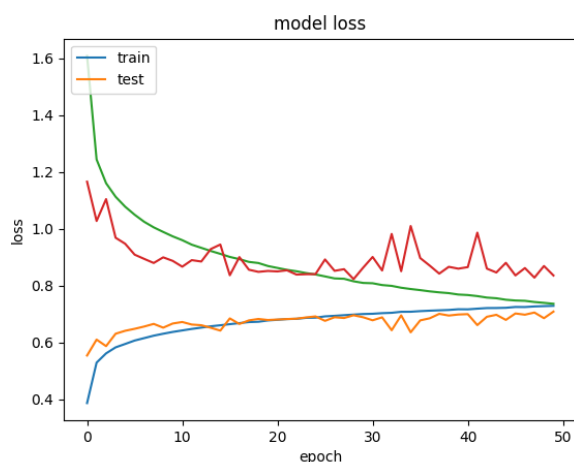
DNN:

Layer (type)	Output Shape	Param #
flatten_1 (Flatten)	(None, 2304)	0
dense_1 (Dense)	(None, 1024)	2360320
leaky_re_lu_1 (LeakyReLU)	(None, 1024)	0
batch_normalization_1 (Batch Normalization)	(None, 1024)	4096
dropout_1 (Dropout)	(None, 1024)	0
dense_2 (Dense)	(None, 512)	524800
leaky_re_lu_2 (LeakyReLU)	(None, 512)	0
batch_normalization_2 (Batch Normalization)	(None, 512)	2048
dropout_2 (Dropout)	(None, 512)	0
dense_3 (Dense)	(None, 7)	3591
activation_1 (Activation)	(None, 7)	0
Total params: 2,894,855		
Trainable params: 2,891,783		
Non-trainable params: 3,072		

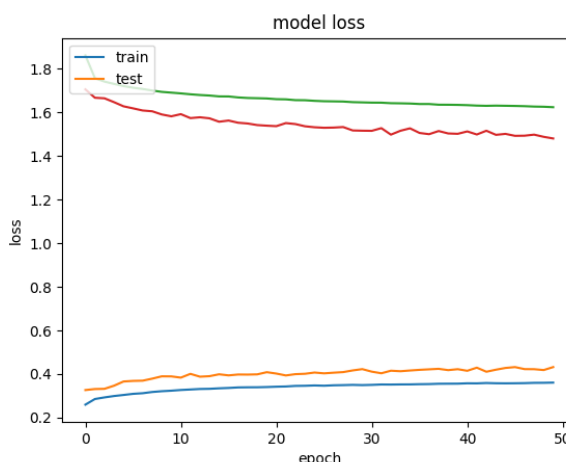
兩者都用了差不多兩百八十幾萬的參數，用batch_size = 128, epoch = 50的參數訓練，結果看起來，CNN的方法效果好很多。

2. (1%) 承上題，請分別畫出這兩個model的訓練過程 (i.e., loss/accuracy v. s. epoch)

CNN:



DNN:



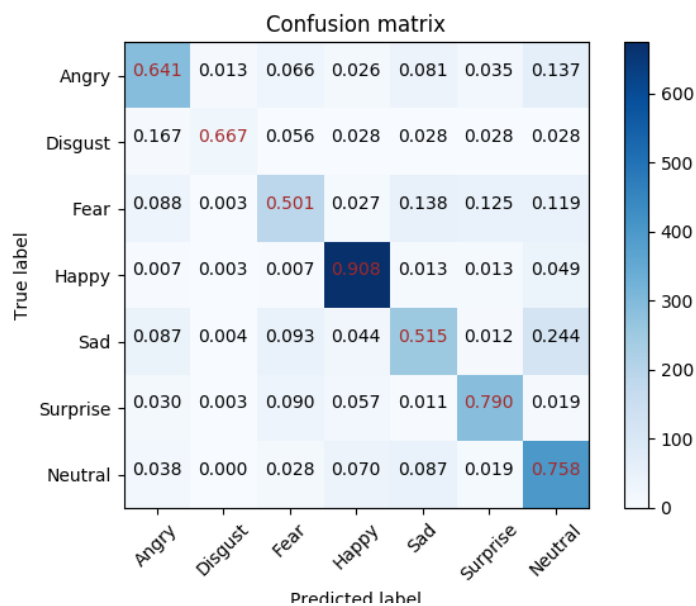
兩張圖中上面兩條線（綠色、紅色）分別代表training和validation的loss，而下面兩條線（藍色、橘色）分別代表training和validation的accuracy。

3. (1%) 請嘗試 **data normalization**, **data augmentation**,說明實作方法並且說明實行前後對準確率有什麼樣的影響？

	public acc	private acc
No data normalization & no data augmentation	0.65394	0.62635
Only data normalization	0.63722	0.63137
Both data normalization & data augmentation	0.68347	0.67957

做了data augmentation會增加training data的量，所以會訓練得比較好，結果也是如此，再來，做data normalization也會進步一些。但data augmentation才是結果變好的主要原因。

4. (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]



從confusion matrix中分析，發現錯誤的predict中，sad最容易被辨認成neutral。