

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？

(Collaborators:)

答：

CNN:

- 架構

Total params: 4,746,471, Trainable params: 4,741,415, Non-trainable params: 5,056

```
def build_model():
    filt_size = (3, 3)
    model = Sequential()
    model.add(Convolution2D(32, filt_size, input_shape=(48,48,1), activation='relu', padding='same'))
    model.add(Convolution2D(32, filt_size, activation='relu', padding='same'))
    model.add(BatchNormalization())
    model.add(MaxPooling2D((2,2)))
    model.add(Dropout(0.5))

    model.add(Convolution2D(64, filt_size, activation='relu', padding='same'))
    model.add(Convolution2D(64, filt_size, activation='relu', padding='same'))
    model.add(BatchNormalization())
    model.add(MaxPooling2D((2,2)))
    model.add(Dropout(0.5))

    model.add(Convolution2D(128, filt_size, activation='relu', padding='same'))
    model.add(Convolution2D(128, filt_size, activation='relu', padding='same'))
    model.add(Convolution2D(128, filt_size, activation='relu', padding='same'))
    model.add(BatchNormalization())
    model.add(MaxPooling2D((2,2)))
    model.add(Dropout(0.5))

    model.add(Convolution2D(256, filt_size, activation='relu', padding='same'))
    model.add(Convolution2D(256, filt_size, activation='relu', padding='same'))
    #model.add(Convolution2D(256, filt_size, activation='relu', padding='same'))
    model.add(BatchNormalization())
    model.add(MaxPooling2D((2,2)))
    model.add(Dropout(0.5))

    model.add(Flatten())
    model.add(Dense(1024, activation='relu'))
    model.add(BatchNormalization())
    model.add(Dropout(0.5))
    model.add(Dense(1024, activation='relu'))
    model.add(BatchNormalization())
    model.add(Dropout(0.5))
    model.add(Dense(7))
    model.add(Activation('softmax'))
```

- 參數： batch size = 128, epoch = 100
- public score: 0.66815, private score: 0.65310

DNN:

- 架構

Total params: 4,995,719, Trainable params: 4,985,095, Non-trainable params: 10,624

```
def build_model():
    model = Sequential()
    model.add(Flatten(input_shape=(48, 48, 1)))

    model.add(Dense(64, activation='relu'))
    model.add(Dense(64, activation='relu'))
    model.add(BatchNormalization())
    model.add(Dropout(0.5))

    model.add(Dense(128, activation='relu'))
    model.add(Dense(128, activation='relu'))
    model.add(BatchNormalization())
    model.add(Dropout(0.5))

    model.add(Dense(512, activation='relu'))
    model.add(Dense(512, activation='relu'))
    model.add(Dense(512, activation='relu'))
    model.add(BatchNormalization())
    model.add(Dropout(0.5))

    model.add(Dense(512, activation='relu'))
    model.add(Dense(512, activation='relu'))
    #model.add(Convolution2D(256, filt_size, activation='relu', padding='same'))
    model.add(BatchNormalization())
    model.add(Dropout(0.5))

    model.add(Dense(1024, activation='relu'))
    model.add(BatchNormalization())
    model.add(Dropout(0.5))

    model.add(Dense(1024, activation='relu'))
    model.add(BatchNormalization())
    model.add(Dropout(0.5))

    model.add(Dense(1024, activation='relu'))
    model.add(BatchNormalization())
    model.add(Dropout(0.5))
    model.add(Dense(1024, activation='relu'))
    model.add(BatchNormalization())
    model.add(Dropout(0.5))
    model.add(Dense(7))
    model.add(Activation('softmax'))
```

- 參數：batch size = 128, epoch = 100
- public score: 0.34104, private score: 0.33797

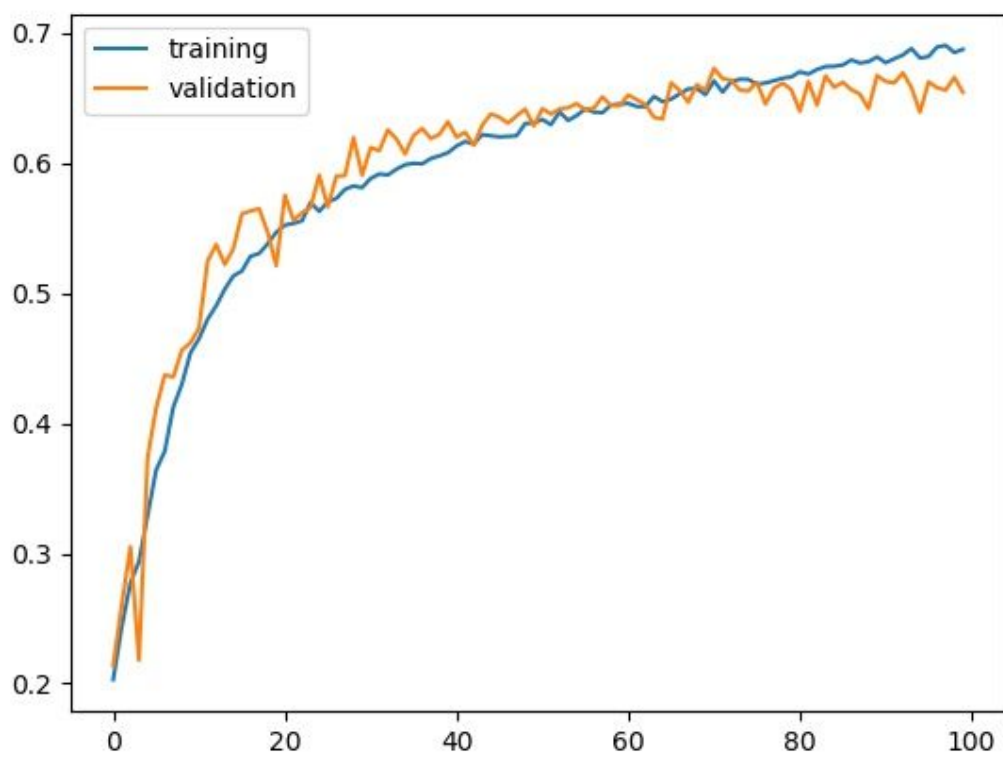
雖然DNN架構未做太多的調整，但是其預測結果非常差，而且由第二題的圖可以發現結果

波動很大且並未達到收斂。

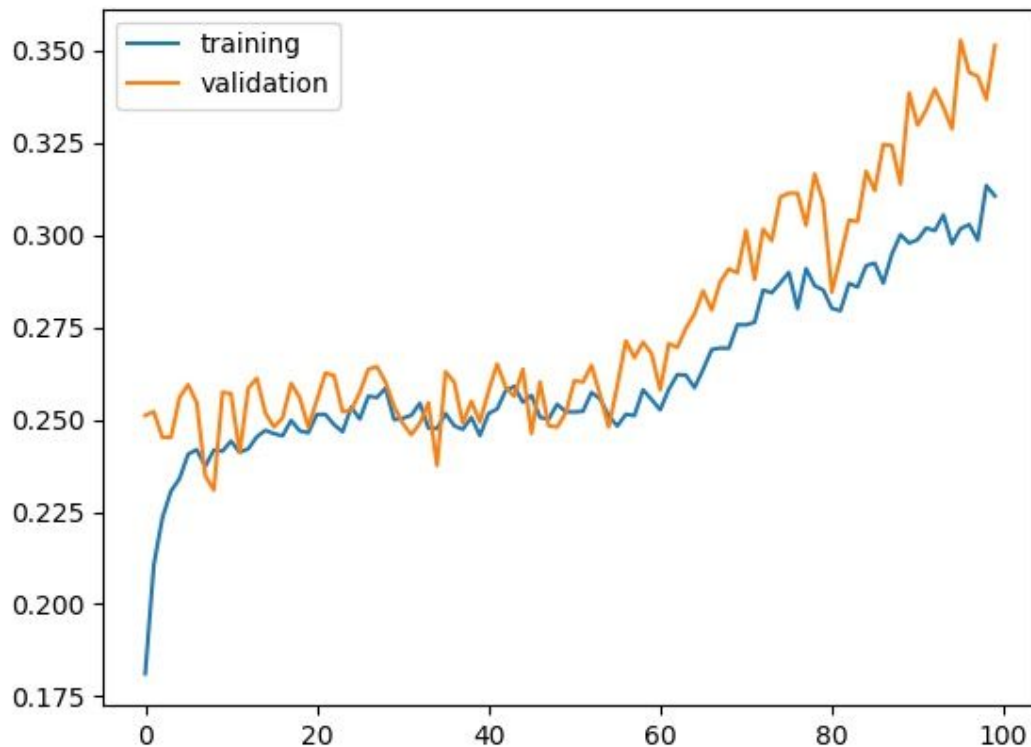
2. (1%) 承上題，請分別畫出這兩個model的訓練過程 (i.e., loss/accuracy v.s. epoch)
(Collaborators:)

答：

CNN



DNN



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？

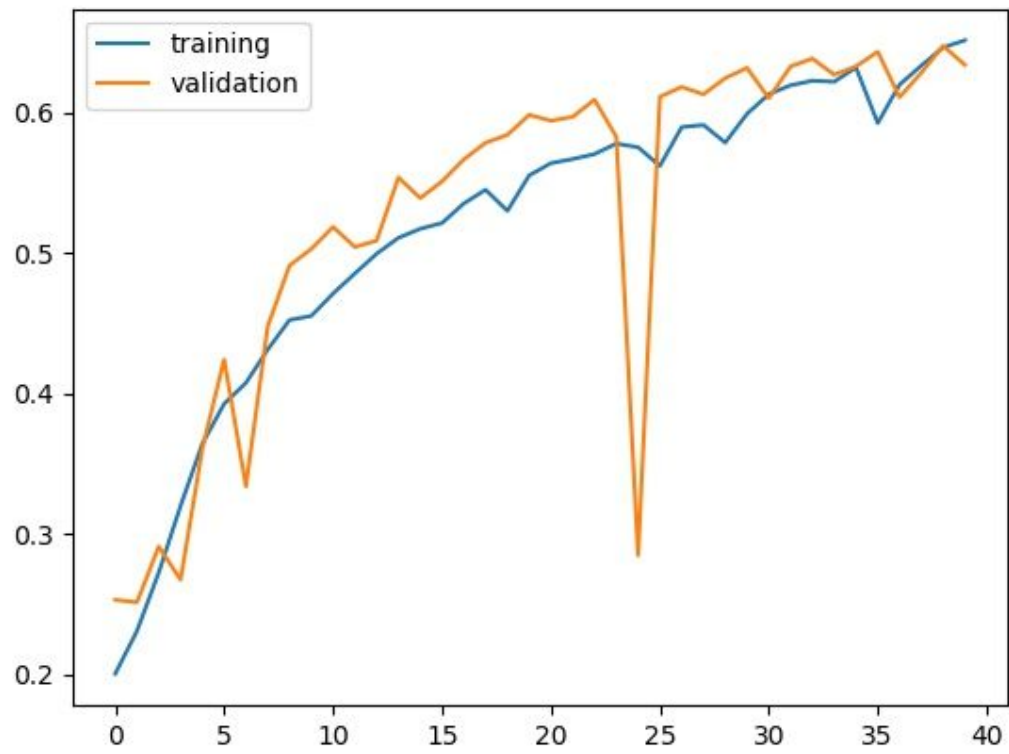
(Collaborators:)

答：

在本題實驗中，我使用CNN的架構執行40個epoch來做比較，normalization的實驗中是使用 $(x - \text{mean}) / (\text{std} * 225)$ ，進行normalization後準確率有略為上升一些，且波動也比較小，收斂的速度似乎較快。而 data augmentation 我則是使用 Keras 的 image generator，並給圖片進行部分的 shift 及 rotate，但準確率反而下降不少，可能是image generator的參數沒有調整好。

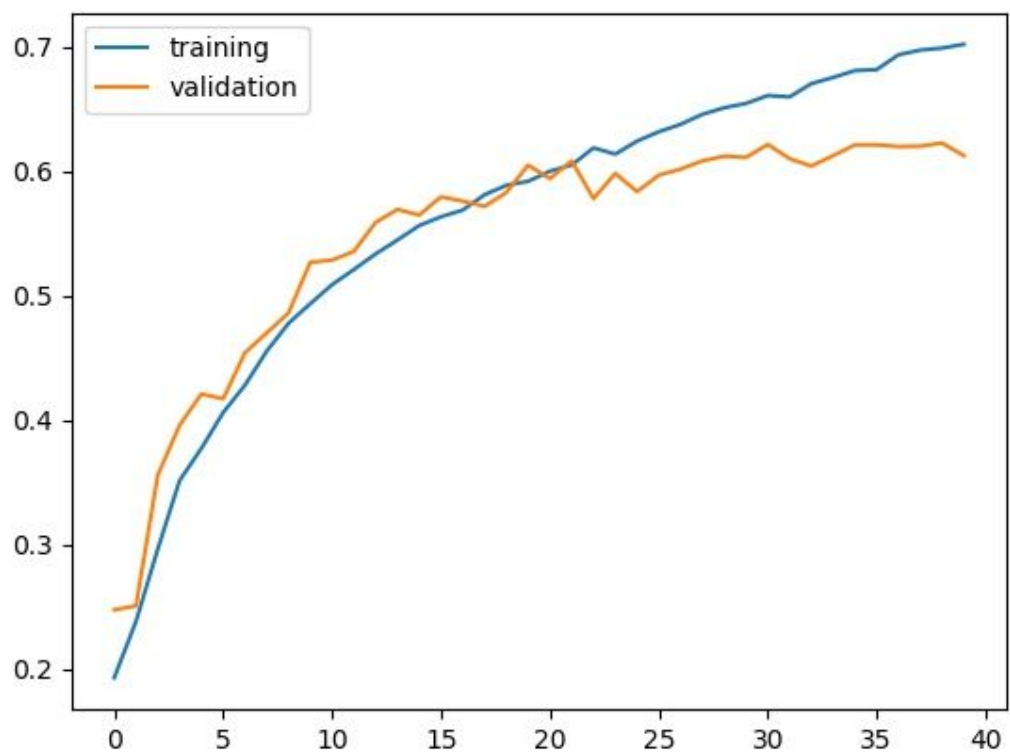
(a) data normalization (X) , data augmentation (X)

public score:0.63360 private score:0.61827



(b) data normalization (O) , data augmentation (X)

public score:0.63555 private score:0.62886



(c) data normalization (X) , data augmentation (O)

public score:0.47868 private score:0.47673



(d) data normalization (O) , data augmentation (O)

同第一題結果

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]

(Collaborators:)

答：

從 confusion matrix 中可以觀察到，最準確的是高興這個分類。而容易混淆的則是難過 vs 中立。

