

水果辨識 CNN(HW4)

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方法

第一題：套用的 pretrained model 是 VGG16

資料的取得和範例程式相同，在 generator 的地方加入 VGG16 的 preprocessing function


```
generator(horizontal_flip=True,preprocessing_function=tf.keras.applications.vgg16.preprocess_input,validation_split=0.1)
generator(preprocessing_function=tf.keras.applications.vgg16.preprocess_input)
```


在載入 pretrained model 時，將 pooling 的參數寫入 'avg'。代表在 include Top = False 的情況下會在最後一層的 output 出來後先套一個 GlobalAveragePooling 再接到我們的 layer。

VGG16 後再接一個 filter 為 256 的 Dense Layer 和一個決定 class 的 Dense Layer 作結束

```
def fruit_model_pretrained(class_num,height,width,channel):
    model = tf.keras.models.Sequential(name="fruit_model_pretrained")
    pretrained = tf.keras.applications.VGG16(include_top=False,input_shape=(height,width,channel),pooling='avg',weights='imagenet')
    model.add(pretrained)
    model.add(tf.keras.layers.Dense(256,activation='relu'))
    model.add(tf.keras.layers.Dense(class_num,activation='softmax'))
    pretrained.trainable = False
    model.compile(loss=tf.keras.losses.SparseCategoricalCrossentropy(),optimizer='adam', metrics=['accuracy'])
    return model
```

Summary 如下圖

 `fruit_model.summary()`

 `Model: "fruit_model_pretrained"`

Layer (type)	Output Shape	Param #
vgg16 (Functional)	(None, 512)	14714688
dense (Dense)	(None, 256)	131328
dense_1 (Dense)	(None, 131)	33667

Total params: 14,879,683
Trainable params: 164,995
Non-trainable params: 14,714,688

Epoch 數設定為 20 下去訓練

```
history = fruit_model.fit(train_generator, epochs=20, validation_data=valid_generator)

Epoch 1/20
1905/1905 [=====] - 105s 39ms/step - loss: 0.2998 - accuracy: 0.9471 - val_loss: 0.3095 - val_accuracy: 0.9356
Epoch 2/20
1905/1905 [=====] - 72s 38ms/step - loss: 0.0411 - accuracy: 0.9902 - val_loss: 0.0853 - val_accuracy: 0.9746
Epoch 3/20
1905/1905 [=====] - 73s 38ms/step - loss: 0.0366 - accuracy: 0.9915 - val_loss: 0.0917 - val_accuracy: 0.9803
Epoch 4/20
1905/1905 [=====] - 73s 38ms/step - loss: 0.0275 - accuracy: 0.9944 - val_loss: 0.0929 - val_accuracy: 0.9801
```

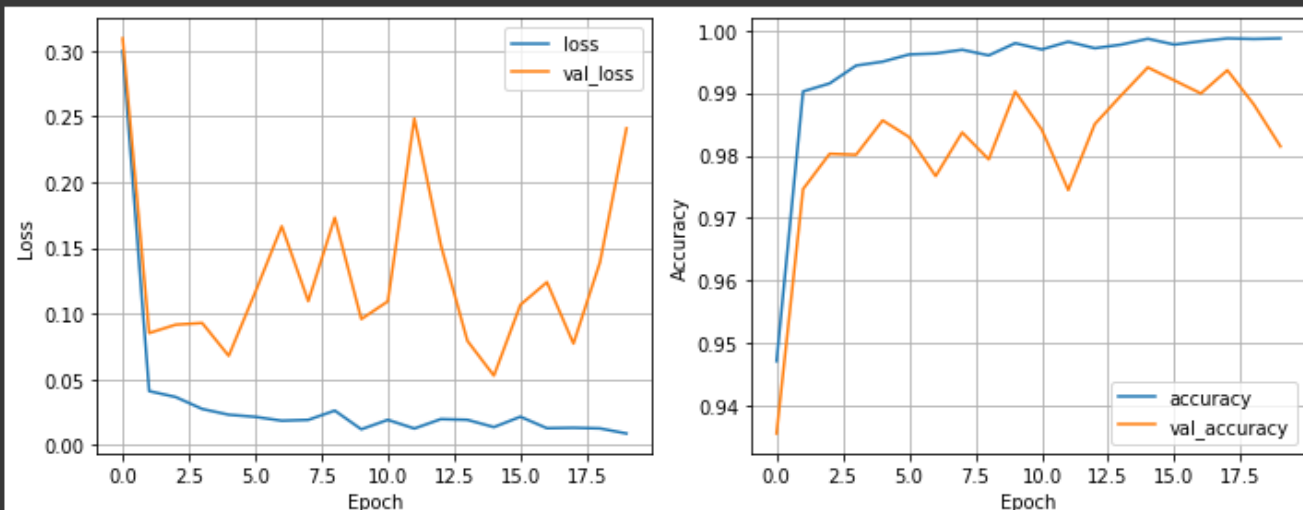
每次大概都跑 70 幾秒 精準度最後都在 0.9 以上

```
Epoch 17/20
1905/1905 [=====] - 73s 38ms/step - loss: 0.0129 - accuracy: 0.9982 - val_loss: 0.1240 - val_accuracy: 0.9899
Epoch 18/20
1905/1905 [=====] - 73s 38ms/step - loss: 0.0132 - accuracy: 0.9987 - val_loss: 0.0773 - val_accuracy: 0.9936
Epoch 19/20
1905/1905 [=====] - 73s 38ms/step - loss: 0.0127 - accuracy: 0.9986 - val_loss: 0.1392 - val_accuracy: 0.9881
Epoch 20/20
1905/1905 [=====] - 73s 38ms/step - loss: 0.0089 - accuracy: 0.9987 - val_loss: 0.2411 - val_accuracy: 0.9814
```

執行結果:

Training and Validation loss / accuracy

```
plt.tight_layout()
plt.show()
```



Evaluation

```
test_generator.reset()
loss, acc = fruit_model.evaluate(test_generator, verbose=0)
print('loss : {:.4f} , acuracy : {:.4f}'.format(loss, acc))

loss : 1.3954 , acuracy : 0.9499
```

第二題:

資料獲取如同範例程式，前處理也相同

```
test_path = os.path.sep.join([file.split(os.path.sep)[-1]+'/'+'fruit-images-dataset-master', 'Test'])

# 定義訓練影像資料產生器
train_datagen = tf.keras.preprocessing.image.ImageDataGenerator(rescale=1./255, validation_split=0.1)
# 定義測試影像資料產生器
test_datagen = tf.keras.preprocessing.image.ImageDataGenerator(rescale=1./255)

# 訓練資料產生器
train_generator = train_datagen.flow_from_directory(training_path, target_size=(100, 100), batch_size=32, class_mode='sparse', subset='training')
# 驗證資料產生器
valid_generator = train_datagen.flow_from_directory(training_path, target_size=(100, 100), batch_size=32, class_mode='sparse', subset='validation')
```

資料進來後先進入一個 conv2D 的 layer

```
def fruit_model(height, width, channel, classes, name):
    input = tf.keras.layers.Input((height, width, channel))

    conv1 = tf.keras.layers.Conv2D(16, (3, 3), padding='valid', activation='relu')(input)

    x = tf.keras.layers.BatchNormalization()(conv1)
    x = tf.keras.layers.ReLU()(x)
    x = tf.keras.layers.Conv2D(16, (3, 3), padding = 'same', activation='relu')(x)
```

後面接一個 2-layer 的 Dense block (BN->Relu->conv->BN->relu->conv)

```
x = tf.keras.layers.BatchNormalization()(conv1)
x = tf.keras.layers.ReLU()(x)
x = tf.keras.layers.Conv2D(16, (3, 3), padding = 'same', activation='relu')(x)
x = tf.keras.layers.BatchNormalization()(x)
x = tf.keras.layers.ReLU()(x)
x = tf.keras.layers.Conv2D(16, (3, 3), padding = 'same', activation='relu')(x)
out = tf.keras.layers.Concatenate()([x, conv1])

z = tf.keras.layers.BatchNormalization()(out)
z = tf.keras.layers.ReLU()(z)
z = tf.keras.layers.Conv2D(16, (3, 3), padding = 'same', activation='relu')(z)
z = tf.keras.layers.BatchNormalization()(z)
z = tf.keras.layers.ReLU()(z)
z = tf.keras.layers.Conv2D(16, (3, 3), padding = 'same', activation='relu')(z)
out1 = tf.keras.layers.Concatenate()([z, out, conv1])

pool = tf.keras.layers.MaxPooling2D((3, 3))(out1)
conv2 = tf.keras.layers.Conv2D(32, (3, 3), padding='valid', activation='relu')(pool)
```

離開 Dense block 後做一個 maxpooling 接兩層 conv2D 然後 flatten
之後經過 Relu 的 Dense layer 然後 分類出 class

```
pool = tf.keras.layers.MaxPooling2D((3, 3))(out1)
conv2 = tf.keras.layers.Conv2D(32, (3, 3), padding='valid', activation='relu')(pool)
conv3 = tf.keras.layers.Conv2D(32, (3, 3), padding='valid', activation='relu')(conv2)

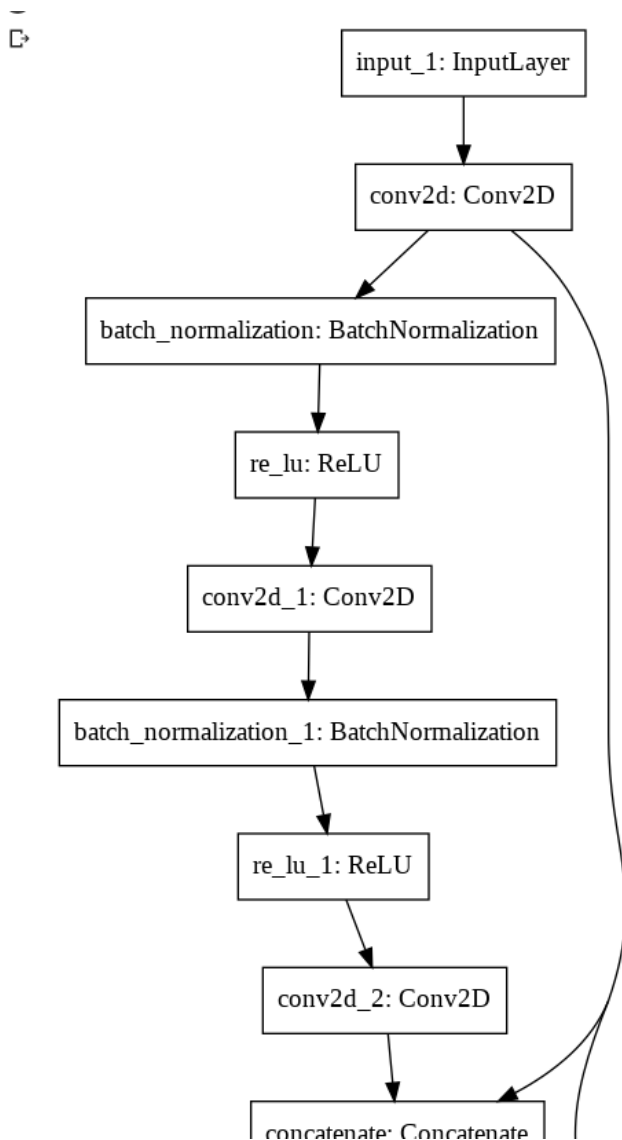
flat = tf.keras.layers.Flatten()(conv3)
dense = tf.keras.layers.Dense(classes, activation='relu')(flat)
dense2 = tf.keras.layers.Dense(classes, activation='softmax')(dense)

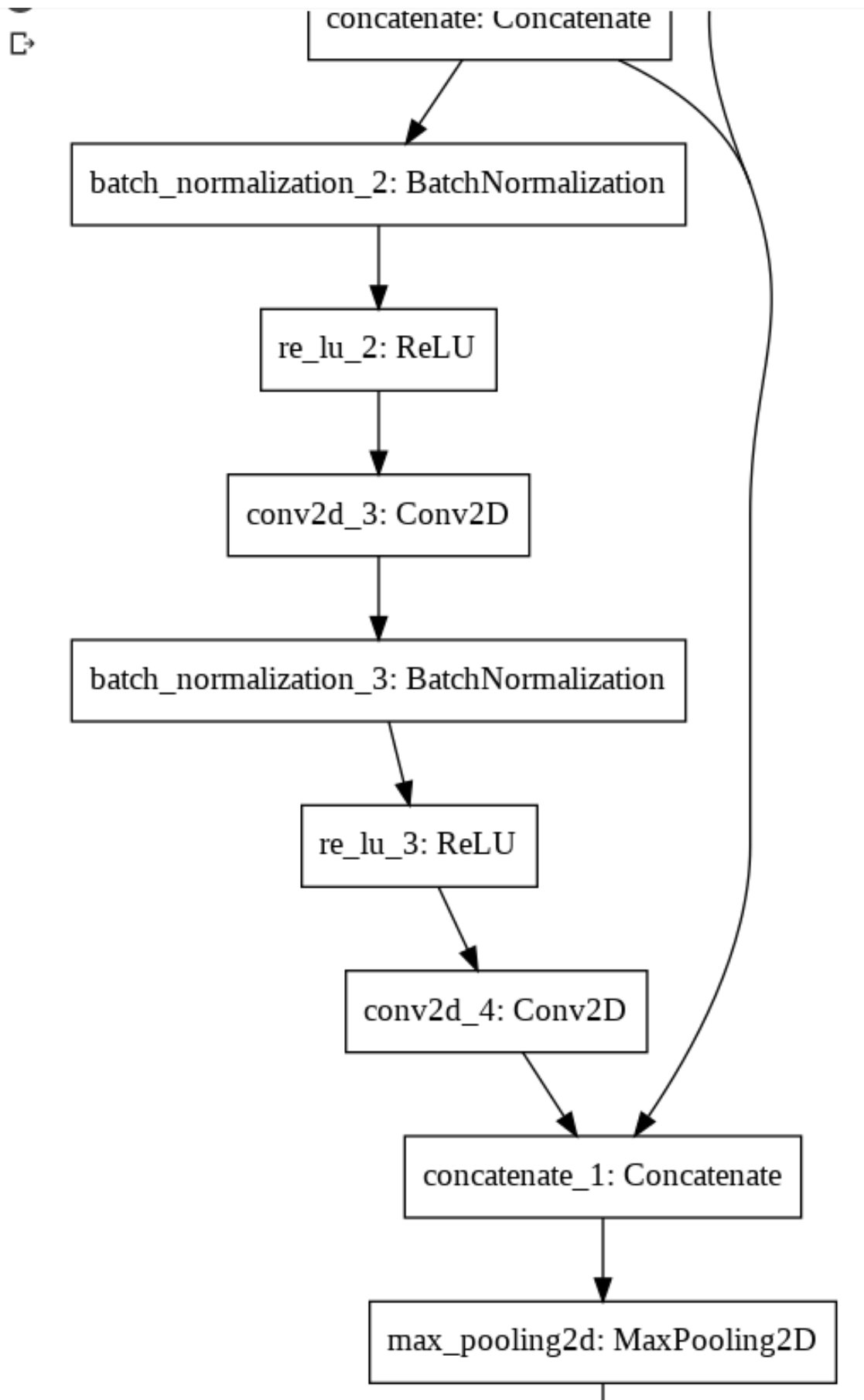
model = tf.keras.Model(inputs=input, outputs=dense2, name=name)
model.summary()
model.compile(loss='sparse_categorical_crossentropy', optimizer='adam', metrics=['accuracy'])

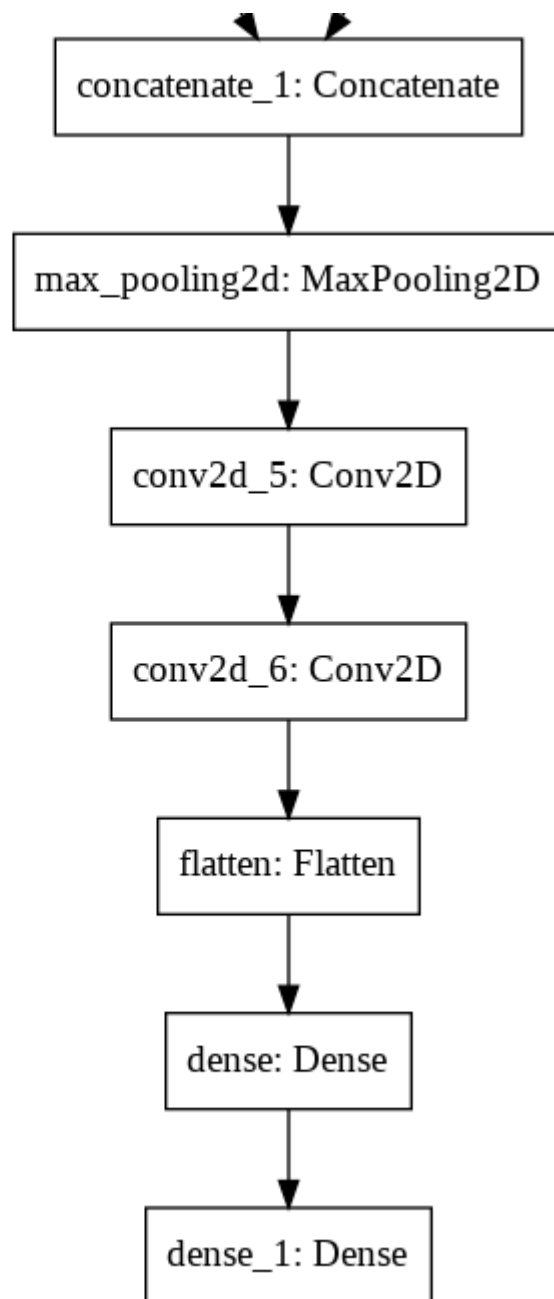
return model

mod = fruit_model(100, 100, 3, train_generator.num_classes, "fruit_model")
```

Plot 出來如下圖 (Summary 有點亂就不放了)







訓練如下圖 10 個 epoch

```
history = mod.fit(train_generator, epochs=10, validation_data=valid_generator)
```

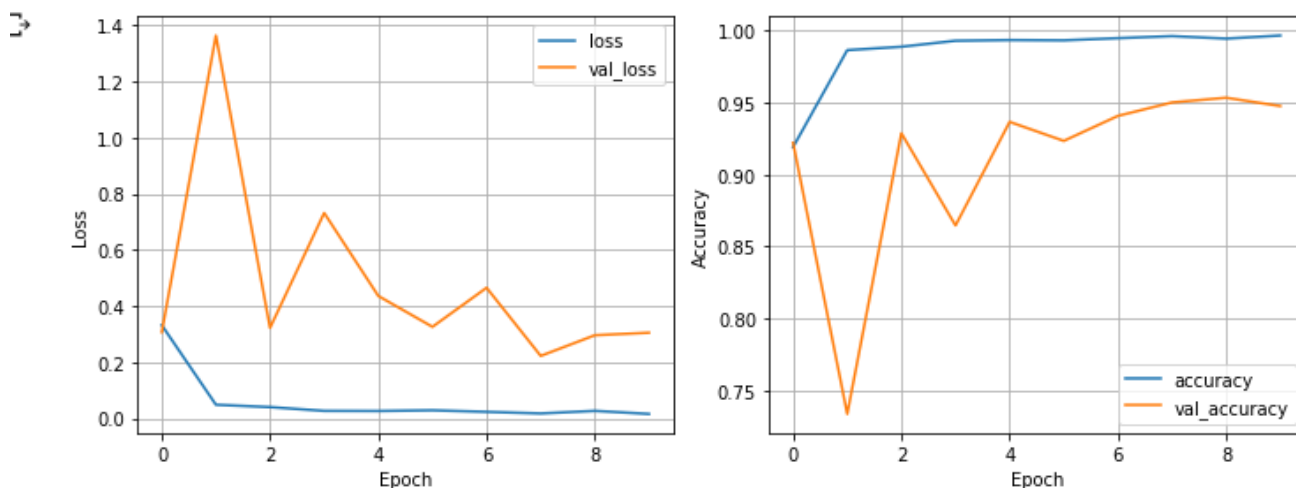
```

Epoch 1/10
1905/1905 [=====] - 113s 42ms/step - loss: 0.3320 - accuracy: 0.9191 - val_loss: 0.3072 - val_accuracy: 0.9221
Epoch 2/10
1905/1905 [=====] - 81s 42ms/step - loss: 0.0488 - accuracy: 0.9865 - val_loss: 1.3627 - val_accuracy: 0.7337
Epoch 3/10
1905/1905 [=====] - 81s 42ms/step - loss: 0.0404 - accuracy: 0.9888 - val_loss: 0.3224 - val_accuracy: 0.9288
Epoch 4/10
1905/1905 [=====] - 80s 42ms/step - loss: 0.0270 - accuracy: 0.9930 - val_loss: 0.7314 - val_accuracy: 0.8646
Epoch 5/10
1905/1905 [=====] - 80s 42ms/step - loss: 0.0265 - accuracy: 0.9935 - val_loss: 0.4359 - val_accuracy: 0.9366
Epoch 6/10
1905/1905 [=====] - 80s 42ms/step - loss: 0.0290 - accuracy: 0.9933 - val_loss: 0.3260 - val_accuracy: 0.9234
Epoch 7/10
1905/1905 [=====] - 80s 42ms/step - loss: 0.0234 - accuracy: 0.9948 - val_loss: 0.4651 - val_accuracy: 0.9408
Epoch 8/10
1905/1905 [=====] - 80s 42ms/step - loss: 0.0178 - accuracy: 0.9962 - val_loss: 0.2226 - val_accuracy: 0.9501
Epoch 9/10
1905/1905 [=====] - 80s 42ms/step - loss: 0.0269 - accuracy: 0.9945 - val_loss: 0.2962 - val_accuracy: 0.9534
Epoch 10/10
1905/1905 [=====] - 80s 42ms/step - loss: 0.0161 - accuracy: 0.9966 - val_loss: 0.3051 - val_accuracy: 0.9476
  
```

執行結果:

Training and Validation loss / accuracy

```
plt.grid(True)
plt.legend()
plt.tight_layout()
plt.show()
```



Evaluation

```
test_generator.reset()
loss, acc = mod.evaluate(test_generator, verbose=0)
print('loss : {:.4f} , accuracy : {:.4f}'.format(loss, acc))
```

```
loss : 0.5867 , accuracy : 0.9258
```

結論

寫到 GPU 被強制關起來只好換一個帳號跑，不然一個 epoch 要跑 1 個小時多。做 CNN 時還算蠻幸運的，沒碰到什麼問題。一開始抓 test generator 的時候一直 find 0 photos 最後發現少下一個參數，折騰了好久。

參考文獻

作業範例程式

上課簡報