

EfficientNet 環境安裝及測試

```
!pip install pandas numpy matplotlib sklearn
```

```
!pip install keras==2.3.1
```

```
!pip install tensorflow
```

```
!pip install protobuf==3.20.*
```

```
!pip list | find /i "tensorflow"
```

```
In [3]: from tensorflow.keras.applications.efficientnet import EfficientNetB0, preprocess_
```

```
In [4]: # 載入 EfficientNetB0 模型
efficientnetB0_model = EfficientNetB0(weights='imagenet',
                                       include_top=False,
                                       input_shape=(224, 224, 3))
```

Downloading data from https://storage.googleapis.com/keras-applications/efficientnetB0_notop.h5

16705208/16705208 [=====] - 5s 0us/step

Fingerprint

```
In [1]: import glob, cv2
from sklearn.model_selection import train_test_split
import numpy as np
from keras.utils import to_categorical
from tensorflow.keras.applications.efficientnet import EfficientNetB0, preprocess_
from PIL import Image
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Dropout, GlobalAveragePooling2D
import matplotlib.pyplot as plt
%matplotlib inline
```

Load Dataset

```
In [2]: x_real = np.load('dataset_c/x_real.npz')['data']
y_real = np.load('dataset_c/y_real.npy')
x_easy = np.load('dataset_c/x_easy.npz')['data']
y_easy = np.load('dataset_c/y_easy.npy')
x_medium = np.load('dataset_c/x_medium.npz')['data']
y_medium = np.load('dataset_c/y_medium.npy')
x_hard = np.load('dataset_c/x_hard.npz')['data']
y_hard = np.load('dataset_c/y_hard.npy')

print(x_real.shape, y_real.shape)

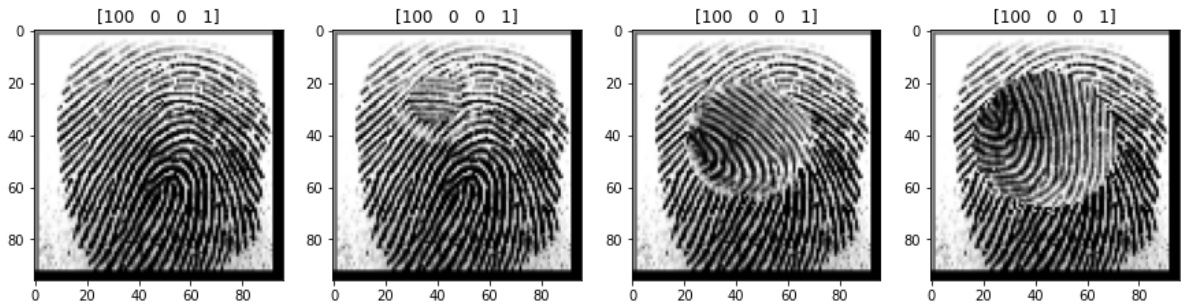
plt.figure(figsize=(15, 10))
plt.subplot(1, 4, 1)
plt.title(y_real[0])
plt.imshow(x_real[0].squeeze()) # cmap='gray'
plt.subplot(1, 4, 2)
plt.title(y_easy[0])
plt.imshow(x_easy[0].squeeze())
plt.subplot(1, 4, 3)
```

```
plt.title(y_medium[0])
plt.imshow(x_medium[0].squeeze())
plt.subplot(1, 4, 4)
plt.title(y_hard[0])
plt.imshow(x_hard[0].squeeze())
```

(6000, 96, 96, 3) (6000, 4)

C:\Users\ArcherSeven\anaconda3\envs\efficientnet\lib\site-packages\matplotlib\tex
t.py:1223: FutureWarning: elementwise comparison failed; returning scalar instead,
but in the future will perform elementwise comparison
if s != self._text:

Out[2]: <matplotlib.image.AxesImage at 0x2d035a78648>



```
In [3]: print(x_real.shape)
print(y_real.shape)
print(x_easy.shape)
print(y_easy.shape)
print(x_medium.shape)
print(y_medium.shape)
print(x_hard.shape)
print(y_hard.shape)
```

(6000, 96, 96, 3)
(6000, 4)
(17931, 96, 96, 3)
(17931, 4)
(17067, 96, 96, 3)
(17067, 4)
(14272, 96, 96, 3)
(14272, 4)

Train Test Split

```
In [4]: x_all_data = np.concatenate([x_easy, x_medium, x_hard], axis=0)
label_all_data = np.concatenate([y_easy, y_medium, y_hard], axis=0)
print(x_all_data.shape, label_all_data.shape)

permutation = list(np.random.permutation(x_all_data.shape[0])) # permutation 隨機排
x_data = x_all_data[permutation][:6000]
label_data = label_all_data[permutation][:6000]

x_train, x_val, label_train, label_val = train_test_split(x_data, label_data, test

print(x_data.shape, label_data.shape) #
print(x_train.shape, label_train.shape) # 訓練
print(x_val.shape, label_val.shape) # 驗證
```

(49270, 96, 96, 3) (49270, 4)
(6000, 96, 96, 3) (6000, 4)
(4800, 96, 96, 3) (4800, 4)
(1200, 96, 96, 3) (1200, 4)

```
In [5]: permutation = list(np.random.permutation(x_real.shape[0])) # permutation 隨機排列
x_test = x_real[permutation][:1200]
label_test = y_real[permutation][:1200]

print(x_test.shape, label_test.shape) # 測試

(1200, 96, 96, 3) (1200, 4)
```

(暫無使用) Make Label Dictionary Lookup Table

```
In [6]: # ID(3) 性別(1) 左右(1) 指頭(1): index
# {'100001': 0, '100004': 1, '100002': 2, ....}
label_real_dict = {}

for i, y in enumerate(y_real):
    key = y.astype(str)
    key = ''.join(key).zfill(6)

    label_real_dict[key] = i
len(label_real_dict)
```

Out[6]: 6000

EfficientNetB0 model

```
In [7]: # 載入 EfficientNetB0 模型
efficientnetB0_model = EfficientNetB0(weights='imagenet',
                                       include_top=False,
                                       input_shape=(224, 224, 3))
```

train data 處理

```
In [8]: # 調整 X_train 的圖片尺寸
print('調整X_train的圖片尺寸...')
X_train_new = np.array(
    [np.asarray(Image.fromarray(x_train[i]).resize(
        (224, 224))) for i in range(0, len(x_train))])
X_train_new = X_train_new.astype('float32')

調整X_train的圖片尺寸...
```

```
In [9]: X_train_new.shape
```

Out[9]: (4800, 224, 224, 3)

```
In [11]: # 訓練資料的資料前處理
train_input = preprocess_input(X_train_new)
# 使用 EfficientNetB0 模型預測訓練資料的特徵資料
print('使用 EfficientNetB0 模型預測訓練資料的特徵資料...')
train_features = efficientnetB0_model.predict(train_input, verbose=1)
```

使用 EfficientNetB0 模型預測訓練資料的特徵資料...
150/150 [=====] - 121s 804ms/step

```
In [12]: train_features.shape
```

Out[12]: (4800, 7, 7, 1280)

val data 處理

```
In [13]: # 調整 X_val 的圖片尺寸
print('調整X_val的圖片尺寸...')
X_val_new = np.array(
    [np.asarray(Image.fromarray(x_val[i]).resize(
        (224, 224))) for i in range(0, len(x_val))])
X_val_new = X_val_new.astype('float32')
```

調整X_val的圖片尺寸...

```
In [14]: X_val_new.shape
```

Out[14]: (1200, 224, 224, 3)

```
In [15]: # 驗證資料的資料前處理
val_input = preprocess_input(X_val_new)
# 使用 EfficientNetB0 模型預測驗證資料的特徵資料
print('使用 EfficientNetB0 模型預測驗證資料的特徵資料...')
val_features = efficientnetB0_model.predict(val_input, verbose=1)
```

使用 EfficientNetB0 模型預測驗證資料的特徵資料...
38/38 [=====] - 31s 795ms/step

```
In [16]: val_features.shape
```

Out[16]: (1200, 7, 7, 1280)

test data 處理

```
In [17]: # 調整 X_test 的圖片尺寸
print('調整X_test的圖片尺寸...')
X_test_new = np.array(
    [np.asarray(Image.fromarray(x_test[i]).resize(
        (224, 224))) for i in range(0, len(x_test))])
X_test_new = X_test_new.astype('float32')
```

調整X_test的圖片尺寸...

```
In [18]: X_test_new.shape
```

Out[18]: (1200, 224, 224, 3)

```
In [19]: # 測試資料的資料前處理
test_input = preprocess_input(X_test_new)
# 使用 EfficientNetB0 模型預測測試資料的特徵資料
print('使用 EfficientNetB0 模型預測測試資料的特徵資料...')
test_features = efficientnetB0_model.predict(test_input, verbose=1)
```

使用 EfficientNetB0 模型預測測試資料的特徵資料...
38/38 [=====] - 30s 789ms/step

```
In [20]: test_features.shape
```

Out[20]: (1200, 7, 7, 1280)

標籤正規化

```
In [21]: print('訓練 :')
id_label_train = to_categorical(label_train[:,0]-1) # 減1是因為id為 1~600
print('身分',id_label_train.shape)
gender_label_train = to_categorical(label_train[:,1])
print('性別',gender_label_train.shape)
lr_label_train = to_categorical(label_train[:,2])
print('左右',lr_label_train.shape)
finger_label_train = to_categorical(label_train[:,3])
print('指頭',finger_label_train.shape)

print('驗證 :')
id_label_val = to_categorical(label_val[:,0]-1)
print('身分',id_label_val.shape)
gender_label_val = to_categorical(label_val[:,1])
print('性別',gender_label_val.shape)
lr_label_val = to_categorical(label_val[:,2])
print('左右',lr_label_val.shape)
finger_label_val = to_categorical(label_val[:,3])
print('指頭',finger_label_val.shape)

print('測試 :')
id_label_test = to_categorical(label_test[:,0]-1)
print('身分',id_label_test.shape)
gender_label_test = to_categorical(label_test[:,1])
print('性別',gender_label_test.shape)
lr_label_test = to_categorical(label_test[:,2])
print('左右',lr_label_test.shape)
finger_label_test = to_categorical(label_test[:,3])
print('指頭',finger_label_test.shape)
```

```
訓練 :
身分 (4800, 600)
性別 (4800, 2)
左右 (4800, 2)
指頭 (4800, 5)
驗證 :
身分 (1200, 600)
性別 (1200, 2)
左右 (1200, 2)
指頭 (1200, 5)
測試 :
身分 (1200, 600)
性別 (1200, 2)
左右 (1200, 2)
指頭 (1200, 5)
```

id model

```
In [23]: from keras.callbacks import EarlyStopping
# 建立 EarlyStopping 物件
es = EarlyStopping(monitor='val_loss', mode='min',
                   verbose=1, patience=5)

from keras.callbacks import ModelCheckpoint
# 建立 ModelCheckpoint 物件
filename = './data/id_weights.h5'
mc = ModelCheckpoint(filename, monitor='val_accuracy',
                     mode='max', verbose=0,
```

```

save_best_only=True)

# 定義模型
model = Sequential()
model.add(GlobalAveragePooling2D(
    input_shape=train_features.shape[1:]))
model.add(Dropout(0.5))
model.add(Dense(600, activation='softmax'))

# 編譯模型
model.compile(loss='categorical_crossentropy', optimizer='adam',
              metrics=['accuracy'])

# 訓練模型
# validation_split: 0~1之間的浮點數，用來指定訓練集的一定比例數據作為驗證集。
# validation_data: 形式為 (X, y) 或 (X, y, sample_weights) 的tuple，是指定的驗證集。
history = model.fit(train_features, id_label_train,
                    validation_data=(val_features, id_label_val),
                    epochs=100, batch_size=32, verbose=0,
                    callbacks=[es, mc])

# 評估模型
print('\nTesting ...')
loss, accuracy = model.evaluate(test_features, id_label_test, verbose=1)
print('測試資料集的準確度 = {:.2f}'.format(accuracy))

```

WARNING:tensorflow:AutoGraph could not transform <function Model.make_train_function.<locals>.train_function at 0x000002D197D45D38> and will run it as-is.
Please report this to the TensorFlow team. When filing the bug, set the verbosity to 10 (on Linux, `export AUTOGRAPH_VERBOSITY=10`) and attach the full output.
Cause: 'arguments' object has no attribute 'posonlyargs'
To silence this warning, decorate the function with @tf.autograph.experimental.do_not_convert

WARNING: AutoGraph could not transform <function Model.make_train_function.<locals>.train_function at 0x000002D197D45D38> and will run it as-is.
Please report this to the TensorFlow team. When filing the bug, set the verbosity to 10 (on Linux, `export AUTOGRAPH_VERBOSITY=10`) and attach the full output.
Cause: 'arguments' object has no attribute 'posonlyargs'
To silence this warning, decorate the function with @tf.autograph.experimental.do_not_convert

WARNING:tensorflow:AutoGraph could not transform <function Model.make_test_function.<locals>.test_function at 0x000002D1A1980438> and will run it as-is.
Please report this to the TensorFlow team. When filing the bug, set the verbosity to 10 (on Linux, `export AUTOGRAPH_VERBOSITY=10`) and attach the full output.
Cause: 'arguments' object has no attribute 'posonlyargs'
To silence this warning, decorate the function with @tf.autograph.experimental.do_not_convert

WARNING: AutoGraph could not transform <function Model.make_test_function.<locals>.test_function at 0x000002D1A1980438> and will run it as-is.
Please report this to the TensorFlow team. When filing the bug, set the verbosity to 10 (on Linux, `export AUTOGRAPH_VERBOSITY=10`) and attach the full output.
Cause: 'arguments' object has no attribute 'posonlyargs'
To silence this warning, decorate the function with @tf.autograph.experimental.do_not_convert

Epoch 38: early stopping

Testing ...

38/38 [=====] - 0s 4ms/step - loss: 3.7823 - accuracy: 0.3592

測試資料集的準確度 = 0.36

In [24]: `import matplotlib.pyplot as plt`
`%matplotlib inline`

```

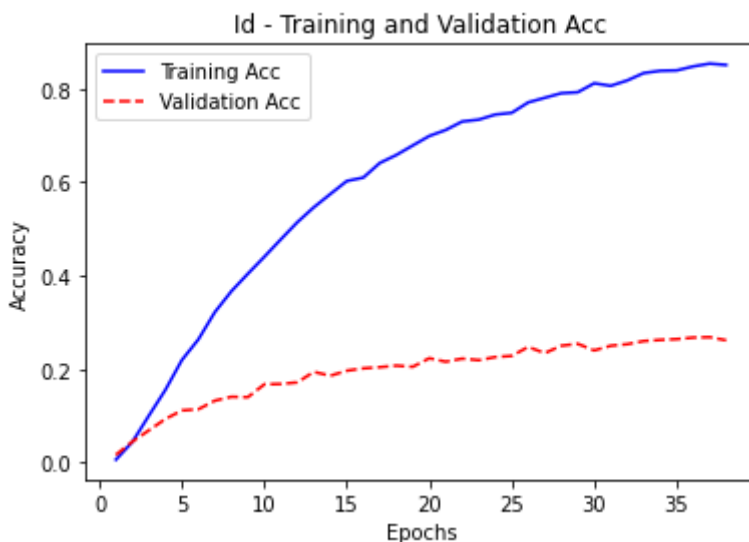
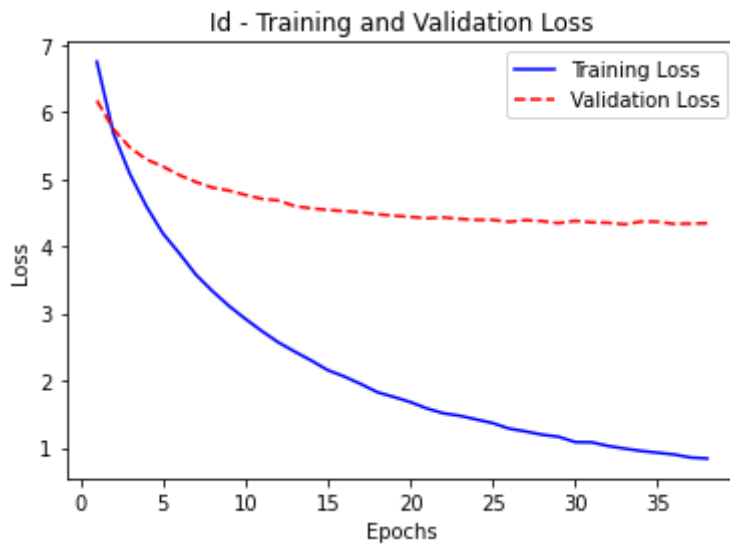
# 顯示訓練和驗證損失
loss = history.history['loss']

```

```

epochs = range(1, len(loss) + 1)
val_loss = history.history['val_loss']
plt.plot(epochs, loss, 'b-', label='Training Loss')
plt.plot(epochs, val_loss, 'r--', label='Validation Loss')
plt.title('Id - Training and Validation Loss')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.legend()
plt.show()
# 顯示訓練和驗證準確度 注意 accyracy 要改成 acc · val_accuracy => val_acc · 因為keras版本
acc = history.history['accuracy']
epochs = range(1, len(acc) + 1)
val_acc = history.history['val_accuracy']
plt.plot(epochs, acc, 'b-', label='Training Acc')
plt.plot(epochs, val_acc, 'r--', label='Validation Acc')
plt.title('Id - Training and Validation Acc')
plt.xlabel('Epochs')
plt.ylabel('Accuracy')
plt.legend()
plt.show()

```



gender model

```

In [26]: from keras.callbacks import EarlyStopping
# 建立 EarlyStopping 物件
es = EarlyStopping(monitor='val_loss', mode='min',

```

```
        verbose=1, patience=5)

from keras.callbacks import ModelCheckpoint
# 建立 ModelCheckpoint 物件
filename = './data/gender_weights.h5'
mc = ModelCheckpoint(filename, monitor='val_accuracy',
                     mode='max', verbose=0,
                     save_best_only=True)

# 定義模型
model = Sequential()
model.add(GlobalAveragePooling2D(
    input_shape=train_features.shape[1:]))
model.add(Dropout(0.5))
model.add(Dense(2, activation='softmax'))

# 編譯模型
model.compile(loss='binary_crossentropy', optimizer='adam',
              metrics=['accuracy'])

# 訓練模型
# validation_split : 0~1 之間的浮點數，用來指定訓練集的一定比例數據作為驗證集。
# validation_data : 形式為 (X, y) 或 (X, y, sample_weights) 的 tuple，是指定的驗證集。
history = model.fit(train_features, gender_label_train,
                    validation_data=(val_features, gender_label_val),
                    epochs=100, batch_size=32, verbose=0,
                    callbacks=[es, mc])

# 評估模型
print('\nTesting ...')
loss, accuracy = model.evaluate(test_features, gender_label_test, verbose=1)
print('測試資料集的準確度 = {:.2f}'.format(accuracy))
```


WARNING:tensorflow:AutoGraph could not transform <function Model.make_train_function.<locals>.train_function at 0x000002D1A50CED38> and will run it as-is.
 Please report this to the TensorFlow team. When filing the bug, set the verbosity to 10 (on Linux, `export AUTOGRAPH_VERBOSITY=10`) and attach the full output.
 Cause: 'arguments' object has no attribute 'posonlyargs'
 To silence this warning, decorate the function with @tf.autograph.experimental.do_not_convert

WARNING: AutoGraph could not transform <function Model.make_train_function.<locals>.train_function at 0x000002D1A50CED38> and will run it as-is.
 Please report this to the TensorFlow team. When filing the bug, set the verbosity to 10 (on Linux, `export AUTOGRAPH_VERBOSITY=10`) and attach the full output.
 Cause: 'arguments' object has no attribute 'posonlyargs'
 To silence this warning, decorate the function with @tf.autograph.experimental.do_not_convert

WARNING:tensorflow:AutoGraph could not transform <function Model.make_test_function.<locals>.test_function at 0x000002D1A51749D8> and will run it as-is.
 Please report this to the TensorFlow team. When filing the bug, set the verbosity to 10 (on Linux, `export AUTOGRAPH_VERBOSITY=10`) and attach the full output.
 Cause: 'arguments' object has no attribute 'posonlyargs'
 To silence this warning, decorate the function with @tf.autograph.experimental.do_not_convert

WARNING: AutoGraph could not transform <function Model.make_test_function.<locals>.test_function at 0x000002D1A51749D8> and will run it as-is.
 Please report this to the TensorFlow team. When filing the bug, set the verbosity to 10 (on Linux, `export AUTOGRAPH_VERBOSITY=10`) and attach the full output.
 Cause: 'arguments' object has no attribute 'posonlyargs'
 To silence this warning, decorate the function with @tf.autograph.experimental.do_not_convert

Epoch 17: early stopping

Testing ...

38/38 [=====] - 0s 3ms/step - loss: 0.4357 - accuracy: 0.8167

測試資料集的準確度 = 0.82

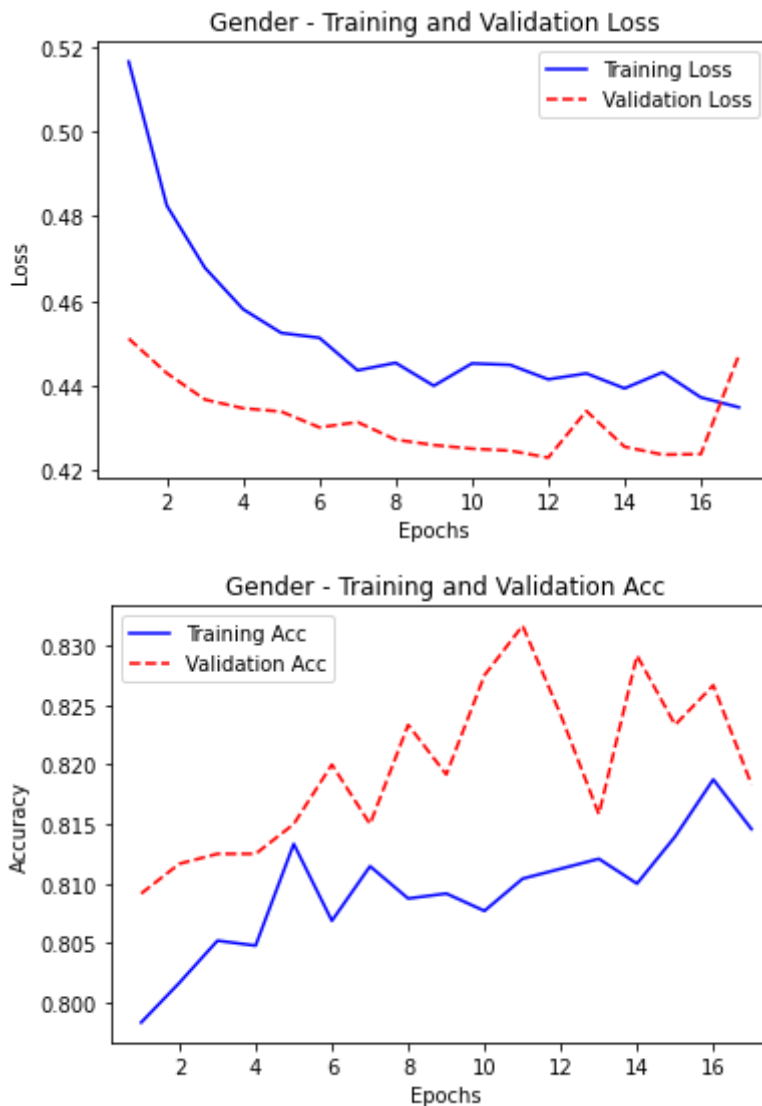
In [27]: `import matplotlib.pyplot as plt`
`%matplotlib inline`

顯示訓練和驗證損失

```
loss = history.history['loss']
epochs = range(1, len(loss) + 1)
val_loss = history.history['val_loss']
plt.plot(epochs, loss, 'b-', label='Training Loss')
plt.plot(epochs, val_loss, 'r--', label='Validation Loss')
plt.title('Gender - Training and Validation Loss')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.legend()
plt.show()
```

顯示訓練和驗證準確度 注意 accyracy 要改成 acc · val_accuracy => val_acc · 因為keras版本

```
acc = history.history['accuracy']
epochs = range(1, len(acc) + 1)
val_acc = history.history['val_accuracy']
plt.plot(epochs, acc, 'b-', label='Training Acc')
plt.plot(epochs, val_acc, 'r--', label='Validation Acc')
plt.title('Gender - Training and Validation Acc')
plt.xlabel('Epochs')
plt.ylabel('Accuracy')
plt.legend()
plt.show()
```



lr model

```
In [28]: from keras.callbacks import EarlyStopping
# 建立 EarlyStopping 物件
es = EarlyStopping(monitor='val_loss', mode='min',
                   verbose=1, patience=5)

from keras.callbacks import ModelCheckpoint
# 建立 ModelCheckpoint 物件
filename = './data/lr_weights.h5'
mc = ModelCheckpoint(filename, monitor='val_accuracy',
                     mode='max', verbose=0,
                     save_best_only=True)

# 定義模型
model = Sequential()
model.add(GlobalAveragePooling2D(
    input_shape=train_features.shape[1:]))
model.add(Dropout(0.5))
model.add(Dense(2, activation='softmax'))

# 編譯模型
model.compile(loss='binary_crossentropy', optimizer='adam',
              metrics=['accuracy'])

# 訓練模型
# validation_split: 0~1 之間的浮點數，用來指定訓練集的一定比例數據作為驗證集。
```

```
# validation_data : 形式為 (X·y) 或 (X·y·sample_weights) 的tuple · 是指定的驗證集。
history = model.fit(train_features, lr_label_train,
                    validation_data=(val_features, lr_label_val),
                    epochs=100, batch_size=32, verbose=0,
                    callbacks=[es, mc])

# 評估模型
print('\nTesting ...')
loss, accuracy = model.evaluate(test_features, lr_label_test, verbose=1)
print('測試資料集的準確度 = {:.2f}'.format(accuracy))
```

WARNING:tensorflow:AutoGraph could not transform <function Model.make_train_function.<locals>.train_function at 0x000002D1B1E521F8> and will run it as-is.
Please report this to the TensorFlow team. When filing the bug, set the verbosity to 10 (on Linux, `export AUTOGRAPH_VERBOSITY=10`) and attach the full output.
Cause: 'arguments' object has no attribute 'posonlyargs'
To silence this warning, decorate the function with @tf.autograph.experimental.do_not_convert

WARNING: AutoGraph could not transform <function Model.make_train_function.<locals>.train_function at 0x000002D1B1E521F8> and will run it as-is.
Please report this to the TensorFlow team. When filing the bug, set the verbosity to 10 (on Linux, `export AUTOGRAPH_VERBOSITY=10`) and attach the full output.
Cause: 'arguments' object has no attribute 'posonlyargs'
To silence this warning, decorate the function with @tf.autograph.experimental.do_not_convert

WARNING:tensorflow:AutoGraph could not transform <function Model.make_test_function.<locals>.test_function at 0x000002D1B1E8CE58> and will run it as-is.
Please report this to the TensorFlow team. When filing the bug, set the verbosity to 10 (on Linux, `export AUTOGRAPH_VERBOSITY=10`) and attach the full output.
Cause: 'arguments' object has no attribute 'posonlyargs'
To silence this warning, decorate the function with @tf.autograph.experimental.do_not_convert

WARNING: AutoGraph could not transform <function Model.make_test_function.<locals>.test_function at 0x000002D1B1E8CE58> and will run it as-is.
Please report this to the TensorFlow team. When filing the bug, set the verbosity to 10 (on Linux, `export AUTOGRAPH_VERBOSITY=10`) and attach the full output.
Cause: 'arguments' object has no attribute 'posonlyargs'
To silence this warning, decorate the function with @tf.autograph.experimental.do_not_convert

Epoch 40: early stopping

Testing ...

38/38 [=====] - 0s 3ms/step - loss: 0.5751 - accuracy: 0.7033

測試資料集的準確度 = 0.70

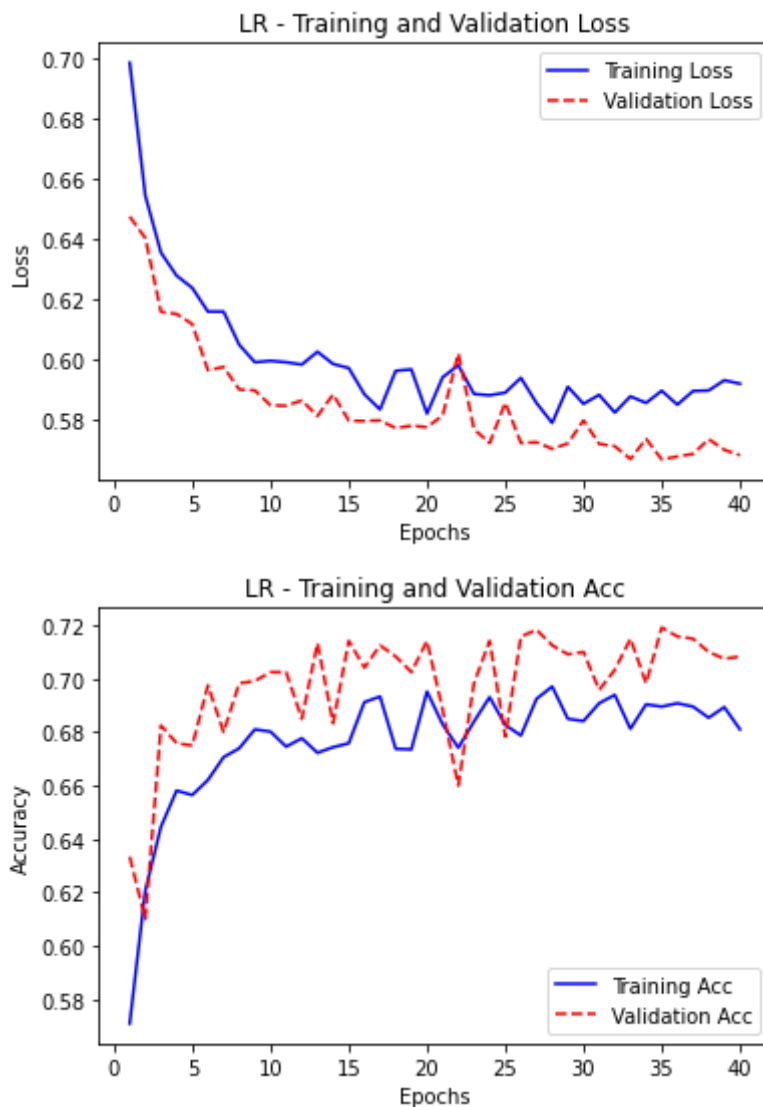
In [29]: `import matplotlib.pyplot as plt`
`%matplotlib inline`

顯示訓練和驗證損失

```
loss = history.history['loss']
epochs = range(1, len(loss) + 1)
val_loss = history.history['val_loss']
plt.plot(epochs, loss, 'b-', label='Training Loss')
plt.plot(epochs, val_loss, 'r--', label='Validation Loss')
plt.title('LR - Training and Validation Loss')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.legend()
plt.show()
```

顯示訓練和驗證準確度 注意 accyracy 要改成 acc · val_accuracy => val_acc · 因為keras版本較新
acc = history.history['accuracy']
epochs = range(1, len(acc) + 1)
val_acc = history.history['val_accuracy']
plt.plot(epochs, acc, 'b-', label='Training Acc')

```
plt.plot(epochs, val_acc, 'r--', label='Validation Acc')
plt.title('LR - Training and Validation Acc')
plt.xlabel('Epochs')
plt.ylabel('Accuracy')
plt.legend()
plt.show()
```



finger model

```
In [30]: from keras.callbacks import EarlyStopping
# 建立 EarlyStopping 物件
es = EarlyStopping(monitor='val_loss', mode='min',
                   verbose=1, patience=5)

from keras.callbacks import ModelCheckpoint
# 建立 ModelCheckpoint 物件
filename = './data/finger_weights.h5'
mc = ModelCheckpoint(filename, monitor='val_accuracy',
                     mode='max', verbose=0,
                     save_best_only=True)

# 定義模型
model = Sequential()
model.add(GlobalAveragePooling2D(
    input_shape=train_features.shape[1:]))
model.add(Dropout(0.5))
```

```

model.add(Dense(5, activation='softmax'))

# 編譯模型
model.compile(loss='categorical_crossentropy', optimizer='adam',
              metrics=['accuracy'])

# 訓練模型
# validation_split: 0~1之間的浮點數，用來指定訓練集的一定比例數據作為驗證集。
# validation_data: 形式為 (X·y) 或 (X·y·sample_weights) 的tuple，是指定的驗證集。
history = model.fit(train_features, finger_label_train,
                    validation_data=(val_features, finger_label_val),
                    epochs=100, batch_size=32, verbose=0,
                    callbacks=[es, mc])

# 評估模型
print('\nTesting ...')
loss, accuracy = model.evaluate(test_features, finger_label_test, verbose=1)
print('測試資料集的準確度 = {:.2f}'.format(accuracy))

```

WARNING:tensorflow:AutoGraph could not transform <function Model.make_train_function.<locals>.train_function at 0x000002D1B2050A68> and will run it as-is.

Please report this to the TensorFlow team. When filing the bug, set the verbosity to 10 (on Linux, `export AUTOGRAPH_VERBOSITY=10`) and attach the full output.

Cause: 'arguments' object has no attribute 'posonlyargs'

To silence this warning, decorate the function with @tf.autograph.experimental.do_not_convert

WARNING: AutoGraph could not transform <function Model.make_train_function.<locals>.train_function at 0x000002D1B2050A68> and will run it as-is.

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WARNING:tensorflow:AutoGraph could not transform <function Model.make_test_function.<locals>.test_function at 0x000002D1B20C8EE8> and will run it as-is.

Please report this to the TensorFlow team. When filing the bug, set the verbosity to 10 (on Linux, `export AUTOGRAPH_VERBOSITY=10`) and attach the full output.

Cause: 'arguments' object has no attribute 'posonlyargs'

To silence this warning, decorate the function with @tf.autograph.experimental.do_not_convert

WARNING: AutoGraph could not transform <function Model.make_test_function.<locals>.test_function at 0x000002D1B20C8EE8> and will run it as-is.

Please report this to the TensorFlow team. When filing the bug, set the verbosity to 10 (on Linux, `export AUTOGRAPH_VERBOSITY=10`) and attach the full output.

Cause: 'arguments' object has no attribute 'posonlyargs'

To silence this warning, decorate the function with @tf.autograph.experimental.do_not_convert

Epoch 24: early stopping

Testing ...

38/38 [=====] - 0s 3ms/step - loss: 0.8844 - accuracy: 0.6508

測試資料集的準確度 = 0.65

In [31]: `import matplotlib.pyplot as plt`
`%matplotlib inline`

顯示訓練和驗證損失

```

loss = history.history['loss']
epochs = range(1, len(loss) + 1)
val_loss = history.history['val_loss']
plt.plot(epochs, loss, 'b-', label='Training Loss')
plt.plot(epochs, val_loss, 'r--', label='Validation Loss')
plt.title('Finger - Training and Validation Loss')
plt.xlabel('Epochs')
plt.ylabel('Loss')

```

```

plt.legend()
plt.show()
# 顯示訓練和驗證準確度 注意 accyracy 要改成 acc · val_accuracy => val_acc · 因為keras版本
acc = history.history['accuracy']
epochs = range(1, len(acc) + 1)
val_acc = history.history['val_accuracy']
plt.plot(epochs, acc, 'b-', label='Training Acc')
plt.plot(epochs, val_acc, 'r--', label='Validation Acc')
plt.title('Finger - Training and Validation Acc')
plt.xlabel('Epochs')
plt.ylabel('Accuracy')
plt.legend()
plt.show()

```

