

GitHub link:

Introduction:

我使用的建置 cnn model 的函式庫是使用 kera，接著為了使用 GPU 去運算，使用了 tensorflow_gpu，雖然我自己使用的 GPU 很不夠力，不過還是稍微比 CPU 快，而主要是使用 anaconda 的 jupyter notebook 去寫的。

以下為函式庫以及 GPU 的部分，classes_labels 與 category 主要用來分類使用

```
import glob
import os
import numpy as np
import tensorflow as tf
import keras
import keras.backend.tensorflow_backend as KTF
import tensorflow.keras.layers as Layers
import tensorflow.keras.optimizers as Optimizer
from keras.utils import np_utils
from keras.regularizers import l2
from keras.models import Sequential
from keras.layers import MaxPooling2D
from keras.layers import Dense, Dropout, Input, BatchNormalization, Activation, Conv2D
from keras.layers import AveragePooling2D, Input, Flatten
from keras.optimizers import Adam
from keras.regularizers import l2
from keras import backend as K
from keras.models import Model
from sklearn.utils import shuffle
from PIL import Image

classes_labels = {'bedroom':0, 'coast':1, 'forest':2, 'highway':3, 'insidecity':4, 'kitchen':5, 'livingroom':6, 'mountain':7, 'office':8,
category={0: 'bedroom', 1: 'coast', 2: 'forest', 3: 'highway', 4: 'insidecity', 5: 'kitchen', 6: 'livingroom', 7: 'mountain', 8: 'office', 9: 'open'

print("Num GPUs Available: ", len(tf.config.experimental.list_physical_devices('GPU'))

gpus = tf.config.experimental.list_physical_devices('GPU')
if gpus:
    # Restrict TensorFlow to only allocate 1GB of memory on the first GPU
    try:
        tf.config.experimental.set_virtual_device_configuration(
            gpus[0],
            [tf.config.experimental.VirtualDeviceConfiguration(memory_limit=4096)])
        logical_gpus = tf.config.experimental.list_logical_devices('GPU')
        print(len(gpus), "Physical GPUs,", len(logical_gpus), "Logical GPUs")
    except RuntimeError as e:
        # Virtual devices must be set before GPUs have been initialized
        print(e)
```

以下為前置讀檔的部分，由路徑讀取測資，接著 size 取為 150*150，這是我訓練出來感覺準度比較好的大小，接著在做一些單純的 reshape 處理。

```
X_size=150
Y_size=150
num=0
X_train=np.zeros((X_size,Y_size))
Y_train=np.zeros(1)

for folders in glob.glob(r'C:\Users\Wei\Desktop\DL&CV\cs-ioc5008-hw1\dataset\dataset\train\*'):
    print(folders)
    label=os.path.basename(folders)
    print(label)
    for filename in os.listdir(folders):
        img_dir=os.path.join(folders, filename)
        Img=Image.open(img_dir)
        test=Img.resize((X_size,Y_size),Image.BILINEAR)
        test=np.array(test,dtype=float)/255
        X_train=np.append(X_train,test)
        Y_train = np.vstack((Y_train,classes_labels[label]))
    print(num)
    num+=1

Y_train=np.delete(Y_train,0,axis=0)
X_train=X_train[X_size*Y_size:]
X_train=X_train.reshape(num,X_size,Y_size,1)
X,y=shuffle(X_train,Y_train,random_state=817328462)
X_4D=X.reshape(X.shape[0],X_size,Y_size,1).astype('float32')
y_OneHot = np_utils.to_categorical(y)
```






以下為 model 的建置，我自己是從最簡單的 conv 層,pooling 層,dense 層慢慢往上加上去建置的，因為測試資料有點少，我在訓練的時候有參考過網路上其他比較強的 model，但是很容易 OOM，所以就以比較單純的建置方法下手，接著訓練後也發現說很容易 overfitting，所以會加入 regularizer 以及 dropout. 訓練了很久之後，交上去 kaggle 的準確度到達 69 分

```
model=tf.keras.Sequential()
model.add(Layers.Conv2D(220,kernel_size=(3,3),activation='relu',input_shape=(150,150,1)))
model.add(Layers.Conv2D(180,kernel_size=(3,3),kernel_regularizer=l2(0.01), bias_regularizer=l2(0.01),activation='relu'))
model.add(Layers.MaxPool2D(5,5))
model.add(Layers.Dropout(rate=0.1))
model.add(Layers.Conv2D(180,kernel_size=(3,3),kernel_regularizer=l2(0.01), bias_regularizer=l2(0.01),activation='relu'))
model.add(Layers.Dropout(rate=0.2))
model.add(Layers.Conv2D(140,kernel_size=(3,3),activation='relu'))
model.add(Layers.Dropout(rate=0.2))
model.add(Layers.Conv2D(100,kernel_size=(3,3),activation='relu'))
model.add(Layers.Dropout(rate=0.2))
model.add(Layers.Conv2D(50,kernel_size=(3,3),activation='relu'))
model.add(Layers.Dropout(rate=0.2))
model.add(Layers.MaxPool2D(5,5))
model.add(Layers.Dropout(rate=0.5))
model.add(Layers.Flatten())
model.add(Layers.Dense(180,activation='relu'))
model.add(Layers.Dense(100,activation='relu'))
model.add(Layers.Dense(50,activation='relu'))
model.add(Layers.Dropout(rate=0.5))
model.add(Layers.Dense(13,activation='softmax'))

model.compile(optimizer=Optimizer.Adam(lr=0.0001),loss='sparse_categorical_crossentropy',metrics=['accuracy'])

model.summary()
```

```
model.fit(X_4D,y,batch_size=32,epochs=1000,validation_split=0.05,verbose=1,shuffle=True)
```

40	0856703		0.69423	7	6h
Your Best Entry 					
Your submission scored 0.69423, which is an improvement of your previous score of 0.66250. Great job!  Tweet this!					
41	0856066		0.60865	8	1h
42	Baseline		0.56153	1	1mo

程式最後的部分就是匯入 **test data** 以及預測跟寫檔了~

```
X_size=150
Y_size=150
num=0

X_test = np.zeros((X_size,Y_size))

for folders in glob.glob(r'C:\Users\Wei\Desktop\DL&CV\cs-ioc5008-hw1\dataset\dataset\test\*'):
    Img=Image.open(folders)
    test=Img.resize((X_size,Y_size),Image.BILINEAR)
    test=np.array(test, dtype=float)/255
    X_test=np.append(X_test,test)
    print(num)
    num+=1

X_test=X_test[X_size*Y_size:]
X_test=X_test.reshape(num,X_size,Y_size,1)

list=[]
for filename in glob.glob(r'C:\Users\Wei\Desktop\DL&CV\cs-ioc5008-hw1\dataset\dataset\test\*'):
    temp=os.path.basename(filename)
    list.append(os.path.splitext(temp)[0])

ANS = model.predict_classes(X_test)
ans=[]
for i in range (0,1041):
    print(ANS[i])
    ans.append(category[ANS[i]])

f=open("submission.csv", "w") ##寫檔
f.write("{}{}\n".format("Id", "label"))
for x in zip(list, ans):
    f.write("{}{}\n".format(x[0], x[1]))
f.close()
```

Summay

我覺得我自己在使用 **model** 所下的功夫不夠，而準確度只達到快 **70%**，跟其他人相比還差太多，而測資數量有點少，很容易 **overfitting**，我覺得我所使用的正規化跟 **dropout** 的方式有待加強，然後設計的時候，有時候參數太大也很容易 **OOM**，所以設計起來要很有 **sense**，能夠好好的去做 **data preprocess** 以及 **layer** 的加入，該如何去處理 **overfitting** 的問題，所以要訓練到準度高達 **95%** 以上看起來真的很有難度，我也很想見識看看前幾名的做法，想看看其他人的做法。