



國立雲林科技大學
電子工程系
Department of Electronic Engineering

教育部補助AI應用領域系列課程-
人工智慧計算晶片設計和應用人才培育

LAB6-EfficientDet

國立雲林科技大學

夏世昌 特聘教授 / 電子工程系

王斯弘 助理教授 / 前瞻學位學士學程

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YunTech 國立雲林科技大學
National Yunlin University of Science & Technology





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目錄

1. 環境
2. 網路、程式
3. label



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今天LAB請在 colab 跑
在es508 的 cuda(11.1) 不相容

要在自己電腦跑推薦 cuda10.2



在自己電腦處理資料環節會遇到的問題 (callback)

找出你的資料夾路徑: Python37/site-

packages/tensorflow_examples/lite/model_maker/core/task/model_spec/object_detector_spec.py

```
.....TB=TensorBoard(  
.....log_dir='./logs',  
.....histogram_freq=0,  
.....write_graph=True,  
.....write_images=False,  
.....update_freq="epoch",  
.....embeddings_freq=0,  
.....embeddings_metadata=None,  
.....)  
.....callbacks=train_lib.get_callbacks(config.as_dict(), val_dataset)  
.....callbacks.append(TB)  
.....print("callbacks", callbacks)  
.....model.fit(  
.....train_dataset,  
.....epochs=epochs,  
.....steps_per_epoch=steps_per_epoch,  
.....callbacks=callbacks,  
.....validation_data=val_dataset,  
.....validation_steps=validation_steps)  
.....return model
```

手動添加程式，為了繪製訓練曲線

找不到直接到conda搜尋這個檔



所需套件

安裝指令: `pip install tf-lite-model-maker`

TFLite 模型製作器的公共 API，一個用於訓練自定義 TFLite 模型的遷移學習庫。

Reference: https://www.tensorflow.org/lite/api_docs/python/tf-lite_model_maker





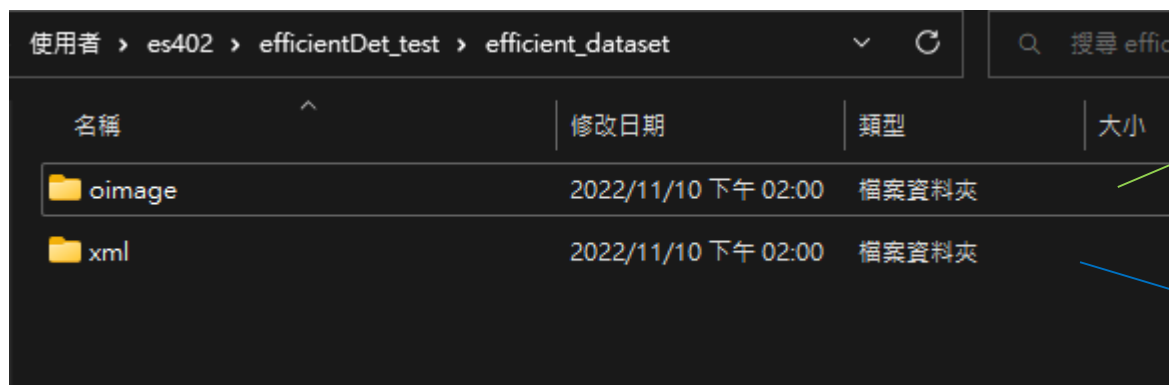
目錄

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dataset

雲端上的 efficient_dataset 壓縮檔



| 使用者 > es402 > efficientDet_test > efficient_dataset | | | |
|---|---------------------|-------|----|
| 名稱 | 修改日期 | 類型 | 大小 |
| oimage | 2022/11/10 下午 02:00 | 檔案資料夾 | |
| xml | 2022/11/10 下午 02:00 | 檔案資料夾 | |

圖片

標籤



匯入函式庫

```
import numpy as np
import os
import pycocotools
import tqdm as notebook_tqdm
import tensorflow as tf
assert tf.__version__.startswith('2')

import tflite_model_maker

from tflite_model_maker.config import ExportFormat
from tflite_model_maker import model_spec
from tflite_model_maker import object_detector

tf.get_logger().setLevel('ERROR')
from absl import logging
logging.set_verbosity(logging.ERROR)

print(tf.__version__)
print(tflite_model_maker.__version__)
```

2.7.0
0.3.2



設定資料夾路徑

```
images_in = './oimage/'  
annotations_in = './xml/'
```

舉例：這是在colab連結到我的雲端 我把dataset 放在雲端內的efficient_dataset中

```
images_in = '/content/drive/MyDrive/efficient_dataset/oimage/'  
annotations_in = '/content/drive/MyDrive/efficient_dataset/xml/'
```



資料處理

```
import os
import random
import shutil

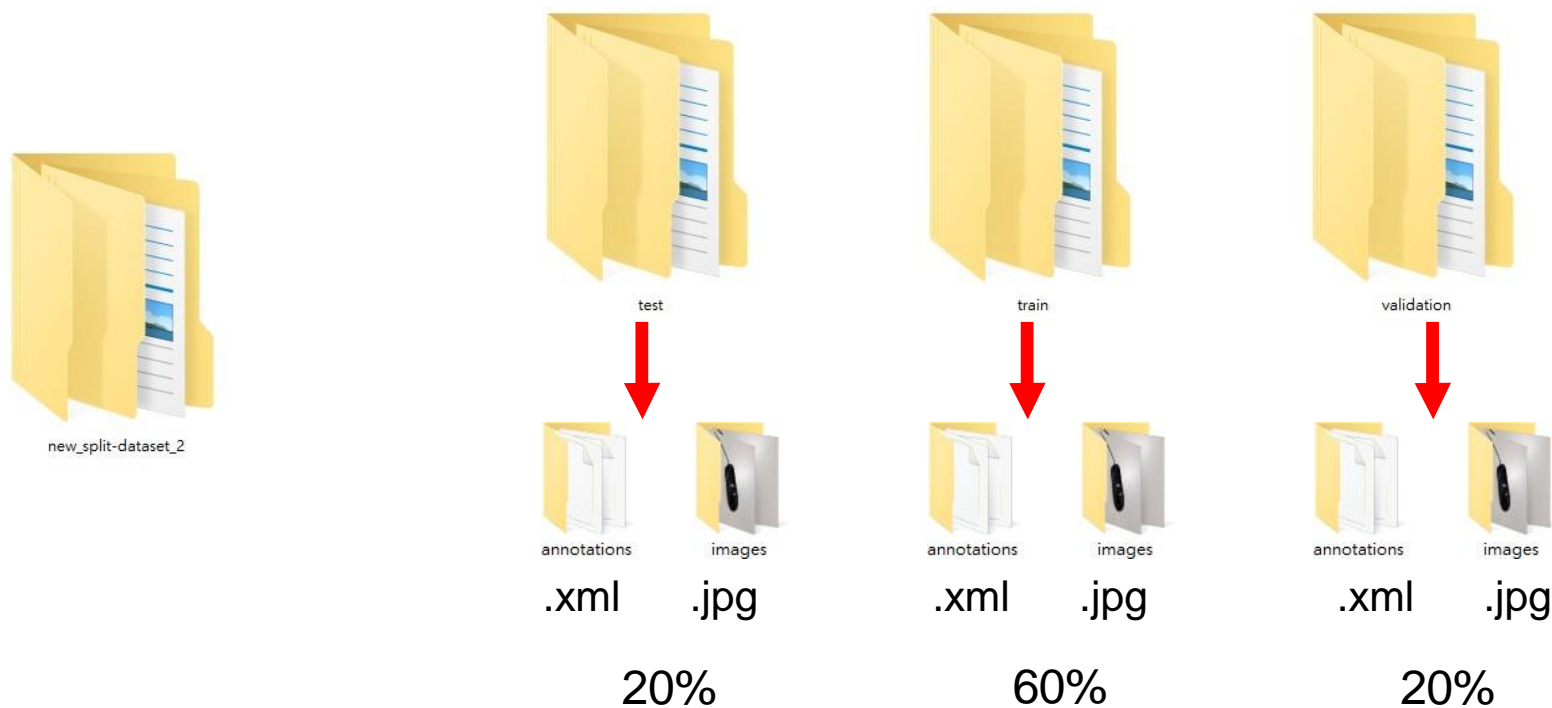
def split_dataset(images_path, annotations_path, val_split, test_split, out_path):

    """將已排序的圖像/註釋目錄拆分為訓練、驗證和測試集。
```



資料處理

```
train_dir, val_dir, test_dir = split_dataset(images_in, annotations_in,  
                                              val_split=0.2, test_split=0.2,  
                                              out_path='new_split-dataset_2')
```



注意圖片副檔名需要 .jpg

資料處理

label_map需要新增你建立的標籤名稱

```
label_map= ["M", "K", "S"]  
train_images_dir = './new_split-dataset_2/train/images/'  
train_annotations_dir = './new_split-dataset_2/train/annotations/'  
val_images_dir = './new_split-dataset_2/validation/images/'  
val_annotations_dir = './new_split-dataset_2/validation/annotations/'  
test_images_dir = './new_split-dataset_2/test/images/'  
test_annotations_dir = './new_split-dataset_2/test/annotations/'
```

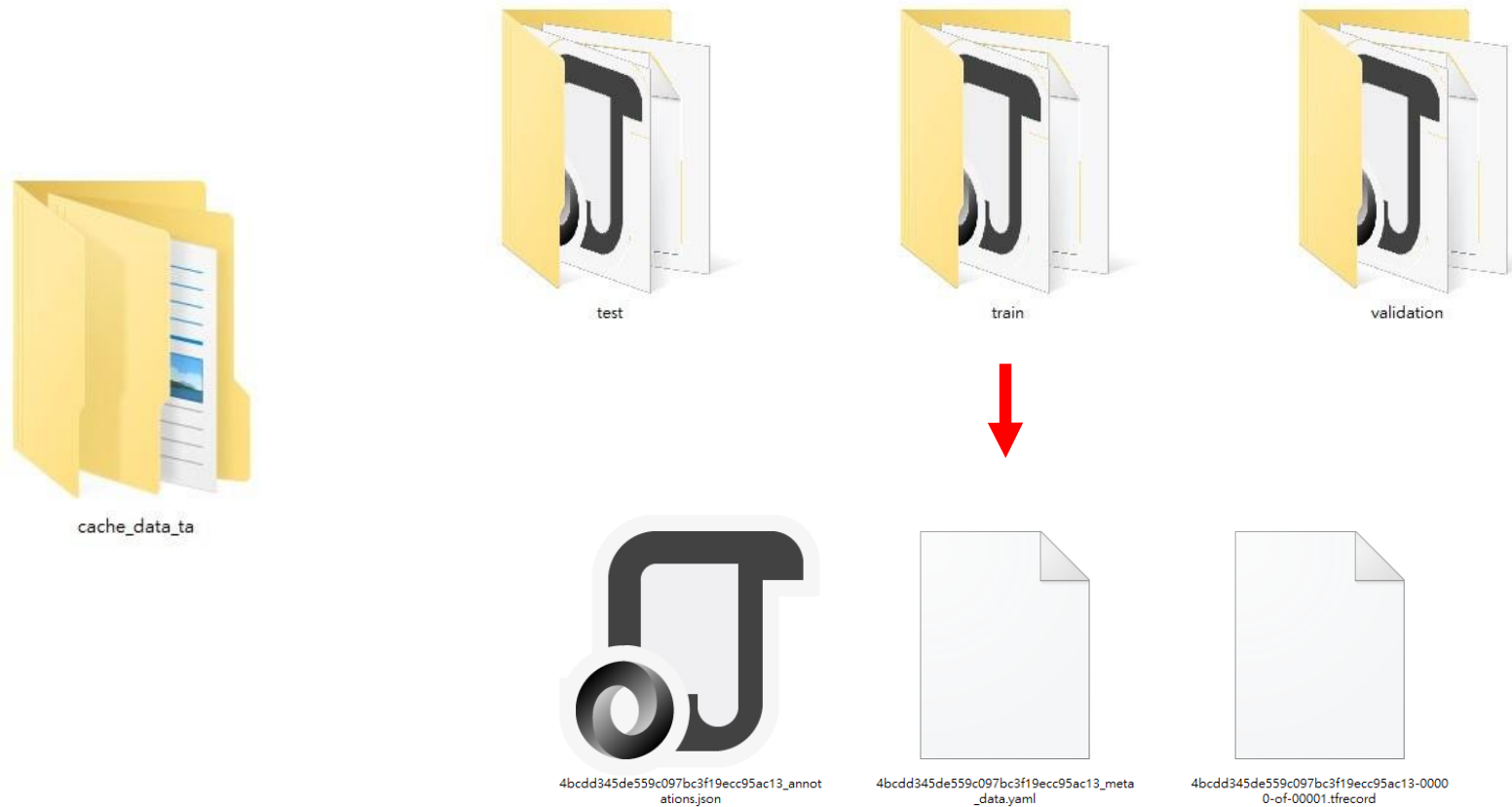


資料處理

```
train_data = object_detector.DataLoader.from_pascal_voc(  
    train_images_dir, train_annotations_dir, label_map=label_map, cache_dir="./cache_data_ta/train",  
  
validation_data = object_detector.DataLoader.from_pascal_voc(  
    val_images_dir, val_annotations_dir, label_map=label_map, cache_dir="./cache_data_ta/validation"  
  
test_data = object_detector.DataLoader.from_pascal_voc(  
    test_images_dir, test_annotations_dir, label_map=label_map, cache_dir="./cache_data_ta/test", num_
```



資料處理



資料處理

```
train_data = object_detector.DataLoader.from_cache('./cache_data_ta/train/4bcdd345de559c097bc3f19ecc95ac13')  
validation_data = object_detector.DataLoader.from_cache('./cache_data_ta/validation/4bcdd345de559c097bc3f19ecc95ac13')  
test_data = object_detector.DataLoader.from_cache('./cache_data_ta/test/4bcdd345de559c097bc3f19ecc95ac13')
```

```
print(f'train count: {len(train_data)}')  
print(f'validation count: {len(validation_data)}')  
print(f'test count: {len(test_data)}')
```

```
train count: 291  
validation count: 97  
test count: 97
```



訓練

```
spec = object_detector.EfficientDetLite0Spec()
```

```
model = object_detector.create(train_data=train_data,  
                               model_spec=spec,  
                               validation_data=validation_data,  
                               epochs=200,  
                               batch_size=16,  
                               train_whole_model=True,  
                               do_train=True)
```

- Epochs、batch_size可以適當調整





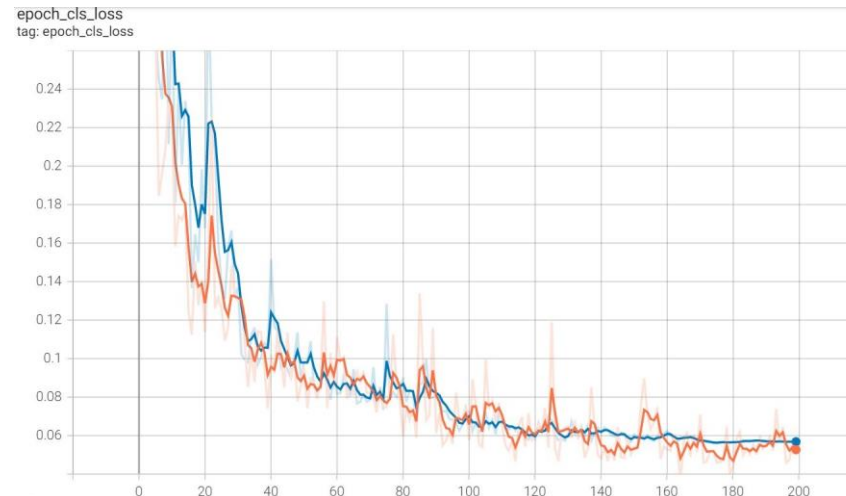
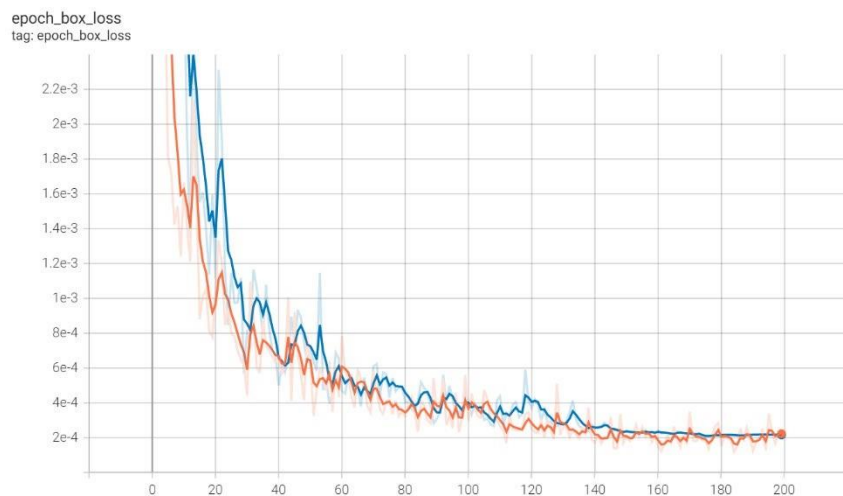
```
callbacks [<keras.callbacks.ModelCheckpoint object at 0x0000025987893348>, <keras.callbacks.TensorBoard object at  
0x0000025989A41588>, <tensorflow_examples.lite.model_maker.third_party.efficientdet.keras.train_lib.COCOCallback o  
bject at 0x0000025989A39AC8>, <keras.callbacks.TensorBoard object at 0x00000257C26591C8>]  
Epoch 1/200  
18/18 [=====] - 43s 657ms/step - det_loss: 1.7116 - cls_loss: 1.1293 - box_loss: 0.0116 -  
reg_l2_loss: 0.0634 - loss: 1.7750 - learning_rate: 0.0140 - gradient_norm: 1.1063 - val_det_loss: 1.7066 - val_cl  
s_loss: 1.1288 - val_box_loss: 0.0116 - val_reg_l2_loss: 0.0634 - val_loss: 1.7700  
Epoch 2/200  
18/18 [=====] - 10s 536ms/step - det_loss: 1.2997 - cls_loss: 0.8761 - box_loss: 0.0085 -  
reg_l2_loss: 0.0634 - loss: 1.3631 - learning_rate: 0.0200 - gradient_norm: 1.7163 - val_det_loss: 1.1881 - val_cl  
s_loss: 0.8238 - val_box_loss: 0.0073 - val_reg_l2_loss: 0.0634 - val_loss: 1.2515  
Epoch 3/200  
18/18 [=====] - 9s 521ms/step - det_loss: 0.8016 - cls_loss: 0.5283 - box_loss: 0.0055 -  
reg_l2_loss: 0.0635 - loss: 0.8650 - learning_rate: 0.0200 - gradient_norm: 1.8455 - val_det_loss: 1.1127 - val_cl  
s_loss: 0.7659 - val_box_loss: 0.0069 - val_reg_l2_loss: 0.0635 - val_loss: 1.1762  
Epoch 4/200  
18/18 [=====] - 9s 518ms/step - det_loss: 0.6037 - cls_loss: 0.4209 - box_loss: 0.0037 -  
reg_l2_loss: 0.0635 - loss: 0.6672 - learning_rate: 0.0200 - gradient_norm: 1.9602 - val_det_loss: 0.9970 - val_cl  
s_loss: 0.6466 - val_box_loss: 0.0070 - val_reg_l2_loss: 0.0636 - val_loss: 1.0605
```



驗證

終端機上 `&> tensorboard --logdir=.\logs\`

沒有跳出網頁請把終端機輸出的網頁複製到瀏覽器





驗證

Evaluate the model

```
model.evaluate(test_data)
```

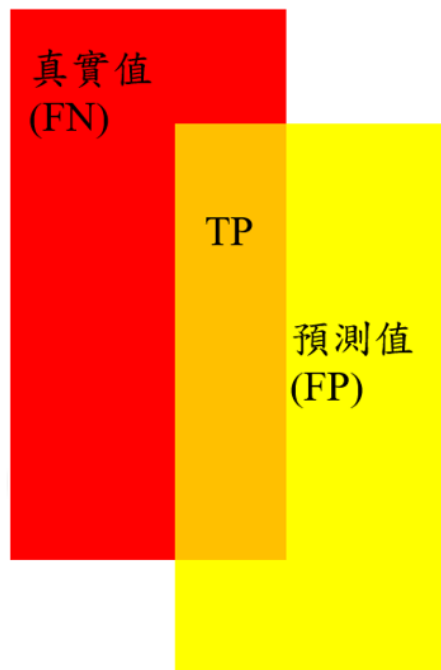
```
2/2 [=====] - 12s 5s/step
```

```
{'AP': 0.94376695,  
 'AP50': 1.0,  
 'AP75': 1.0,  
 'APs': -1.0,  
 'APm': -1.0,  
 'AP1': 0.943777,  
 'ARmax1': 0.9462973,  
 'ARmax10': 0.9634138,  
 'ARmax100': 0.9634138,  
 'ARs': -1.0,  
 'ARm': -1.0,  
 'AR1': 0.9634138,  
 'AP_/M': 0.938481,  
 'AP_/K': 0.9886693,  
 'AP_/S': 0.9790732,  
 'AP_/P': 0.8688444}
```



驗證

$$MIoU = \frac{TP}{FP + FN + TP}$$



- AP: IoU[0.5:0.95], Stride:0.05 , (mean Average Precision)
- AP50: IoU > 50% (mean Average Precision)
- AP75: IoU > 75% (mean Average Precision)



匯出權重檔

```
TFLITE_FILENAME = 'efficientdet.tflite'  
LABELS_FILENAME = 'labels.txt'  
SAVED_FILENAME = 'model'
```

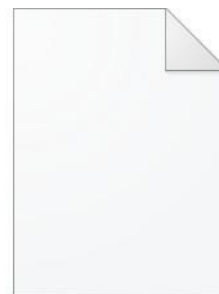
```
model.export(export_dir='./export', tflite_filename=TFLITE_FILENAME, label_filename=LABELS_FILENAME, saved_model_file_name=SAVED_FILENAME,  
              export_format=[ExportFormat.TFLITE, ExportFormat.LABEL, ExportFormat.SAVED_MODEL])
```



export



model



efficientdet.tflite



labels.txt

測試

TFLite_test.ipynb

```
import tensorflow as tf
import cv2 as cv
import numpy as np
import glob

tf.compat.v1.reset_default_graph()
interpreter = tf.lite.Interpreter('./export/'+ 'efficientdet.tflite') # 讀取 .tflite 模型
interpreter.allocate_tensors() # 初始化模型
```



測試

TFLite_test.ipynb

```
import tensorflow as tf
import cv2 as cv
import numpy as np
import glob

tf.compat.v1.reset_default_graph()
interpreter = tf.lite.Interpreter('./export/'+ 'efficientdet.tflite') # 讀取 .tflite 模型
interpreter.allocate_tensors() # 初始化模型
```



測試

```
w,h,c=input_details[0]['shape'][1:4] # 獲取輸入圖片大小
img_file='./new_split-dataset_2/test/images/20220425_230252_006.jpg' # 一個圖片路徑
imgc = 0
Pred = 0

img=cv.imread(img_file) # 讀取圖片
img=img.astype(np.uint8) # 轉整數
img=cv.resize(img,(w,h)) # 調整圖片大小
img=cv.cvtColor(img,cv.COLOR_BGR2RGB) # BGR to RGB
img=img.reshape((w,h,c)) # 調整陣列大小
img=np.expand_dims(img,0) # 轉 4D 陣列

interpreter.set_tensor(input_details[0]['index'], img) # 輸入資料到模型
interpreter.invoke() # 模型計算輸出

detection_scores = interpreter.get_tensor(output_details[0]['index']) # 獲取輸出各框的得分
detection_boxes = interpreter.get_tensor(output_details[1]['index']) # 獲取輸出框座標 (百分比)
num_boxes = interpreter.get_tensor(output_details[2]['index']) # 獲取輸出總框數量
detection_classes = interpreter.get_tensor(output_details[3]['index']) # 輸出框類別
```

測試

```
outimg=cv.imread(img_file)
for i in range(0,num_boxes):
    # 把百分比座標轉為實際圖片座標
    detection_boxes[i,0]=int(detection_boxes[i,0]*outimg.shape[0])
    detection_boxes[i,1]=int(detection_boxes[i,1]*outimg.shape[1])
    detection_boxes[i,2]=int(detection_boxes[i,2]*outimg.shape[0])
    detection_boxes[i,3]=int(detection_boxes[i,3]*outimg.shape[1])
    # 得分大於 0.5 (可自行調整) 才畫出框及文字敘述

    if detection_scores[i]>0.5:
        outimg=cv.rectangle(outimg, (int(detection_boxes[i,1]), int(detection_boxes[i,0])), (int(detection_boxes[i,3]
        if detection_classes[i] ==1:
            ID = 'Mouse'
        elif detection_classes[i] ==2:
            ID = 'Keyboard'
        elif detection_classes[i] ==3:
            ID = 'Screen'
        outimg=cv.putText(outimg,'ID: %s'%(ID), (int(detection_boxes[i,1]),int(detection_boxes[i,0]-3)),cv.FONT_HERSHEY
cv.imshow('out', outimg)
key = cv.waitKey(0)
if key>0:
    cv.destroyAllWindows()
```



測試

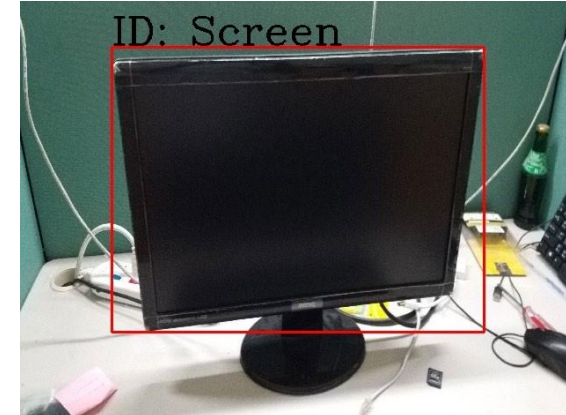
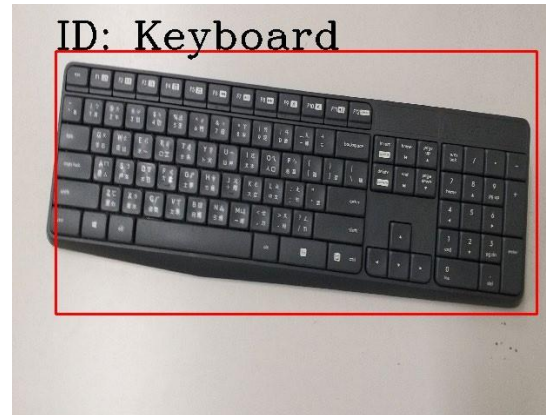
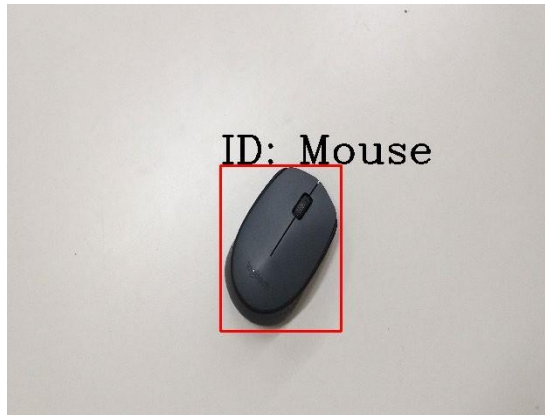
Colab無法使用imshow 改成存圖片

```
cv.imwrite('out.jpg',outimg)
'''

cv.imshow('out', outimg)
key = cv.waitKey(0)
if key>0:
    cv.destroyAllWindows()
'''
```



測試





目錄

1. 環境
2. 資料集、網路、程式
3. **label**

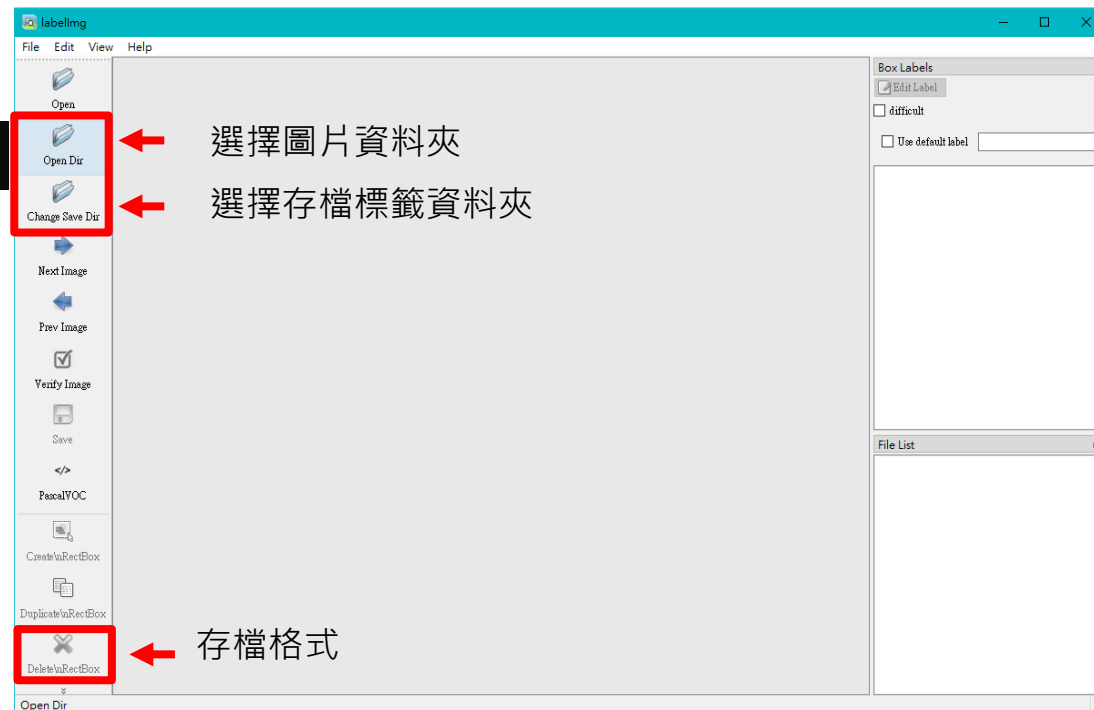


Label

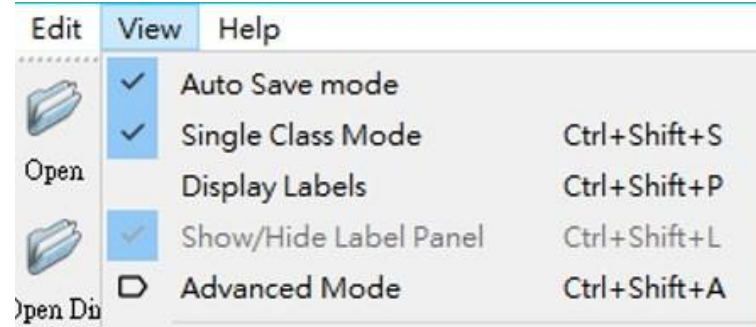
pip install labeling

開啟labelimg

```
(keras) C:\Users\es402\Desktop>labelimg
```



Label





Label

手機請盡量調低相機像素、並打橫著拍，我的圖片是3264x1836

W 框選

A 上一張

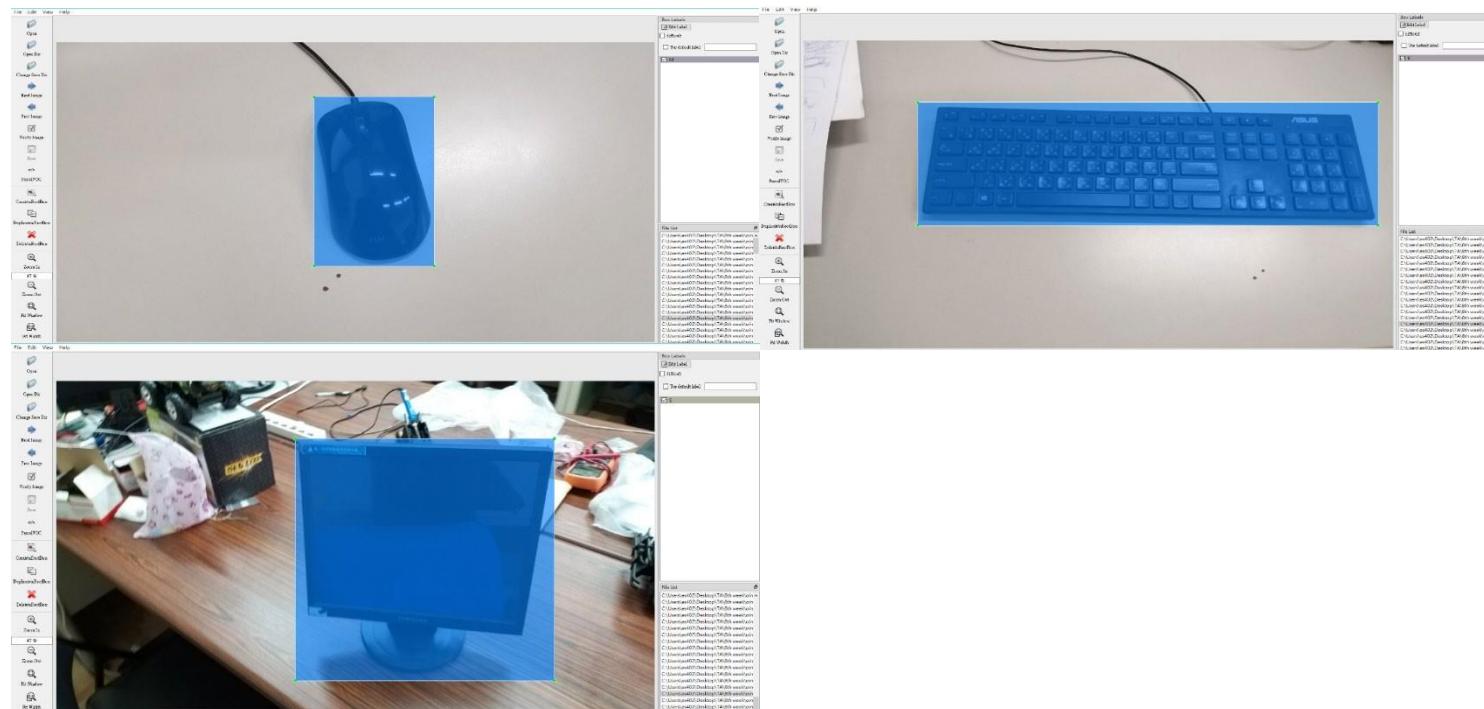
D 下一張

Delete 刪除框選

Ctrl+S 存檔



Label



Label:
M: 滑鼠
K: 鍵盤
S: 螢幕

```
label_map= ["M", "K", "S"]
train_images_dir = './content/drive/MyDrive/Colab Notebooks/efficientdet/train/images/'
train_annotations_dir = './content/drive/MyDrive/Colab Notebooks/efficientdet/train/annotations/'
val_images_dir = './content/drive/MyDrive/Colab Notebooks/efficientdet/validation/images/'
val_annotations_dir = './content/drive/MyDrive/Colab Notebooks/efficientdet/validation/annotations/'
test_images_dir = './content/drive/MyDrive/Colab Notebooks/efficientdet/test/images/'
test_annotations_dir = './content/drive/MyDrive/Colab Notebooks/efficientdet/test/annotations/'
```





END

