

Keras_yolov3實作教學

2019/12/19

TA-侑學

Outline

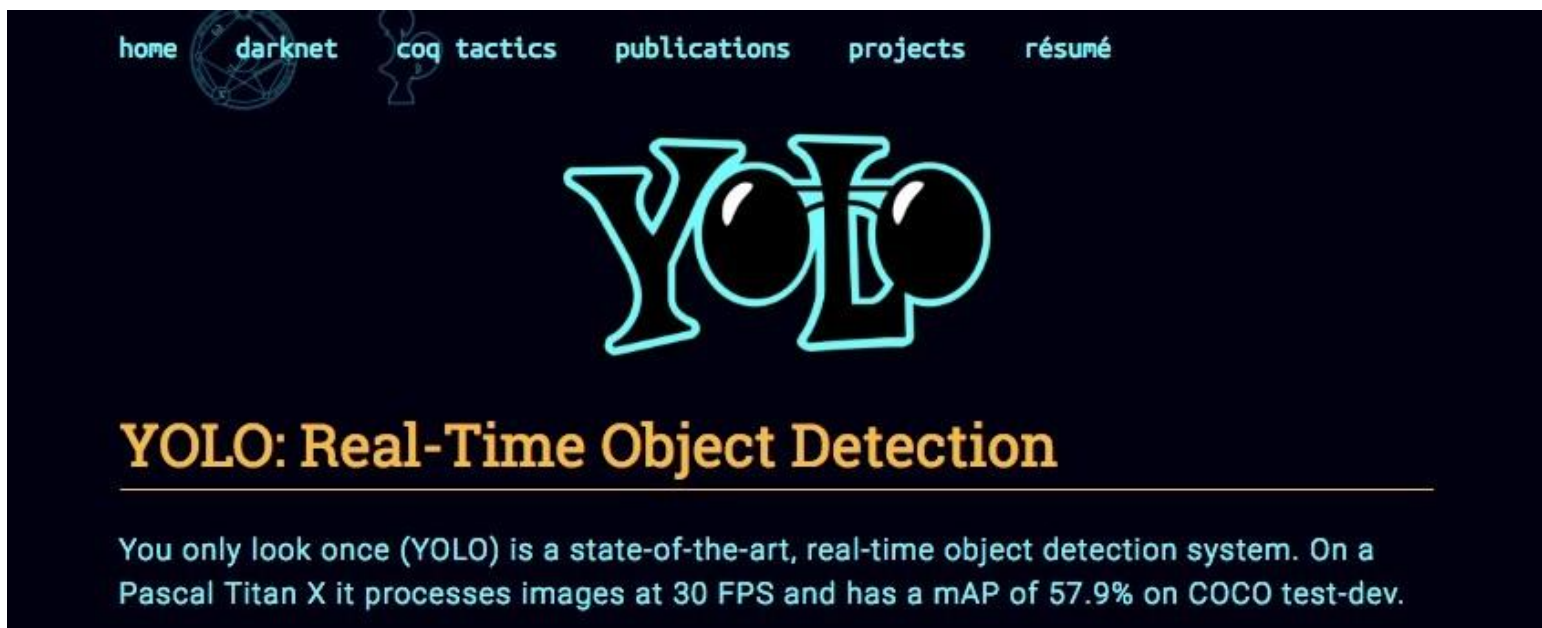
- Introduction
- 實作
- 訓練自己的model

檔案連結-請先下載

- <https://drive.google.com/file/d/1gViUGFXEVYmyQm6s5aiWp7VVhNGzl4tQ/view?usp=sharing>

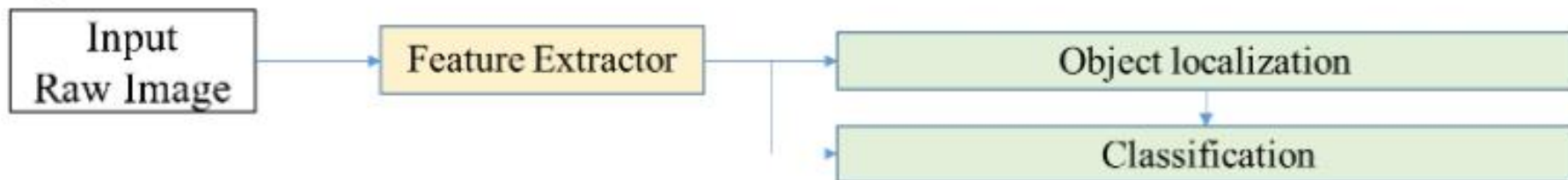
Introduction

- YOLO全名 - You only look once，顧名思義就是只要學習過一一次即可，代表學習的速度很快。
- YOLO是one stage的物件偵測方法，也就是只需要對圖片作一次CNN架構便能夠判斷圖形內的物體位置與類別。



現今兩種Object detection方法

Two-stage Object detection

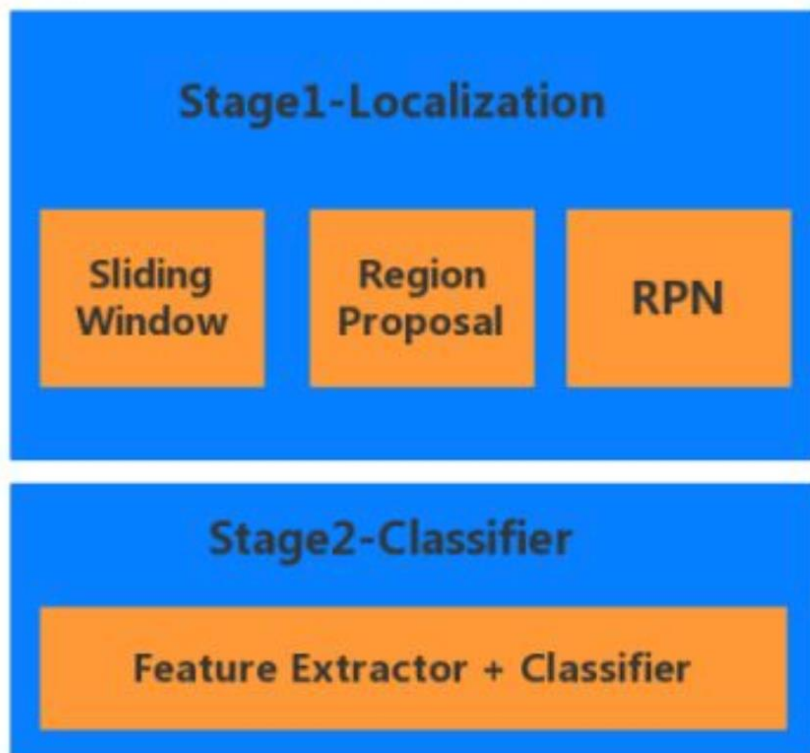


One-stage Object detection



現今兩種Object detection方法

Two-stage methods
(RCNN, FRCNN)



One-stage methods
(YOLO, SSD, FPN)



Yolo發展- YoloV1

- 1.從 RCNN、fast RCNN、faster RCNN、Yolo 的思路一路發展上來，Yolo 最大的特色是直接 end-to-end 做物件偵測，利用整張圖片作為神經網路的輸入，直接預測 bounding box 坐標位置、bounding box 含物體的 confidence 和物體所屬的類別。
- 2. YoloV1 計算快速，能夠達到 real-time 速度需求，缺點是對位置的預測不夠精確，且小物體預測效果較差。

Yolo發展 - YoloV2

- 1. YoloV2 針對 YoloV1 的缺點做了一些改進：
- 2. 引入 Faster RCNN 中的 anchor box，不再直接 mapping bounding box 的座標，而是預測相對於 anchor box 的參數，並使用 K-Means 求 anchor box 比例。
- 3. 去掉 fully connected layer，改成全部皆為 conv layer。
- 4. 每層加上 batch normalization，去掉 dropout。
- 5. 增加解析度：增加 ImageNet pre-train 的解析度，從 224×224 提升至 448×448 。

YoloV3-邊界框預測

- 邊界框預測 - YOLOv3在邊界框的預測上，改成以logistic regression來預測邊界框包含物體的分數。YOLOv3只把與ground-truth重疊率最高的那個邊界框認作滿分1，不是被認作最好的那些邊界框，就不會對分類或是邊界框座標的損失有做貢獻。
- 分類預測 - YOLOv3改用independent logistic classifiers，並搭配binary cross-entropy來衡量分類上的損失。

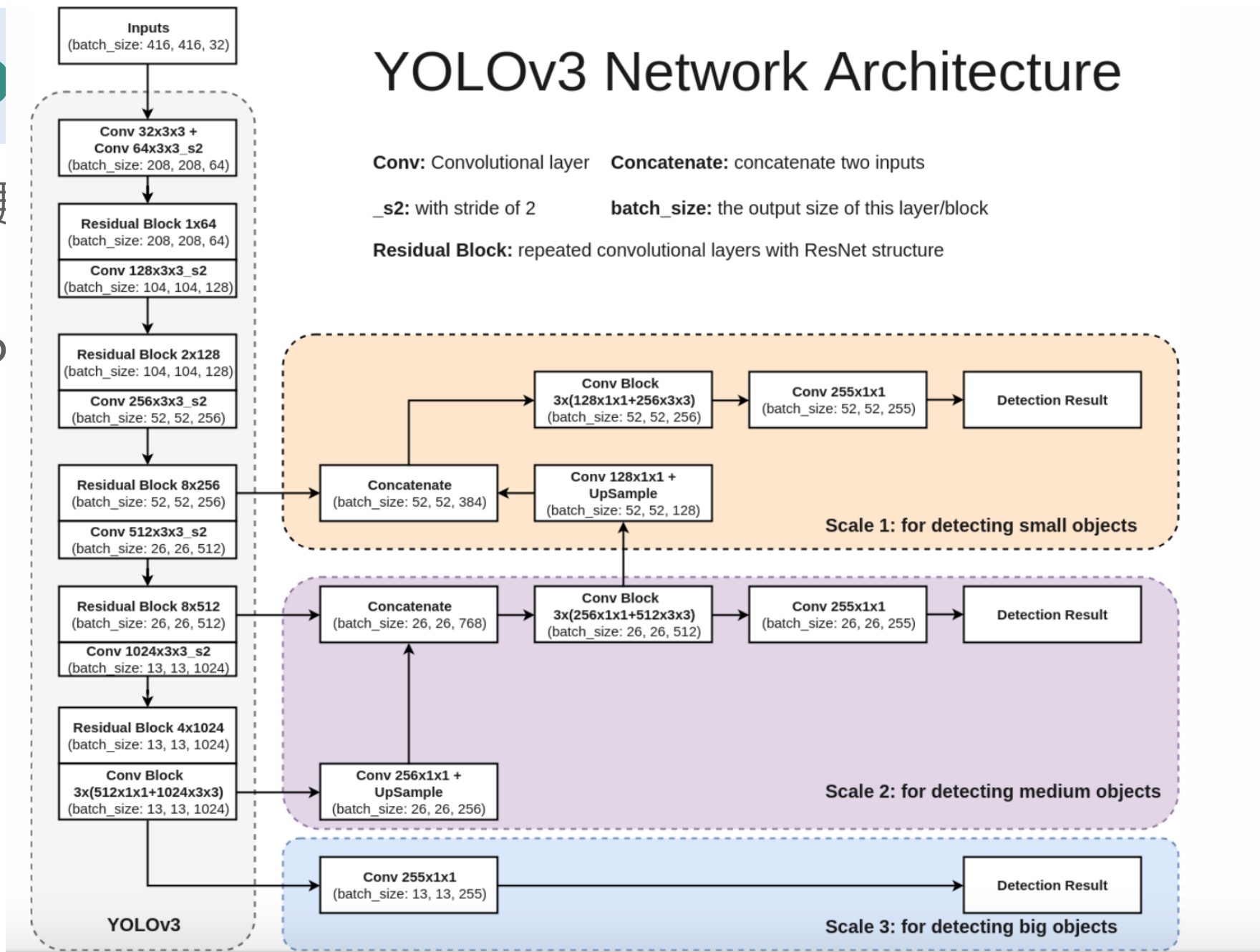
Yo

YOLOv3 Network Architecture

Conv: Convolutional layer **Concatenate:** concatenate two inputs
_s2: with stride of 2 **batch_size:** the output size of this layer/block
Residual Block: repeated convolutional layers with ResNet structure

• 1. 夏

• 2. P



s)

Why use YOLO?

- 簡單的安裝步驟
- 不需要調動太多的參數，即可有良好的訓練效果

Keras-yolov3

- Google drive 連結：<https://drive.google.com/file/d/1QYu-pUQ8Dwlx7AKdW1k4-rifz8xouhOd/view?usp=sharing>
- 此次實作的keras-yolov3版本為：<https://github.com/experiencor/keras-yolo3>
- 詳細運作可看此篇的說明

YOLO3 (Detection, Training, and Evaluation)

Dataset and Model

Dataset	mAP	Demo	Config	Model
Kangaroo Detection (1 class) (https://github.com/experiencor/kangaroo)	95%	https://youtu.be/UR03UDHvoLY	check zoo	http://bit.ly/2Xqz8qL
Raccoon Detection (1 class) (https://github.com/experiencor/raccoon_dataset)	98%	https://youtu.be/lxLyLIL7OsU	check zoo	http://bit.ly/2Xqz8qL
Red Blood Cell Detection (3 classes) (https://github.com/experiencor/BCCD_Dataset)	84%	https://imgur.com/a/uJI2IRI	check zoo	http://bit.ly/2Xqz8qL
VOC (20 classes) (http://host.robots.ox.ac.uk/pascal/VOC/voc2012/)	72%	https://youtu.be/0RmOI6hcfBI	check zoo	http://bit.ly/2Xqz8qL

Todo list:

- ☒ Yolo3 detection

資料夾內部

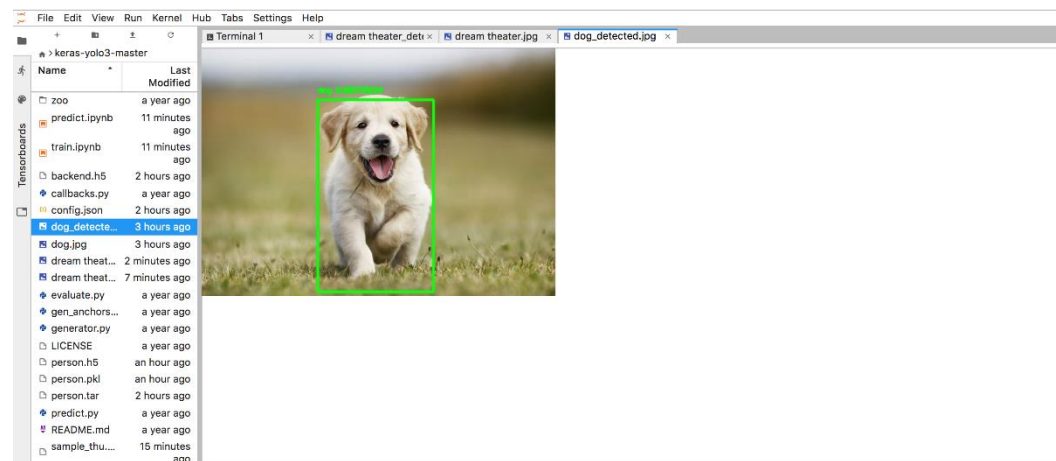
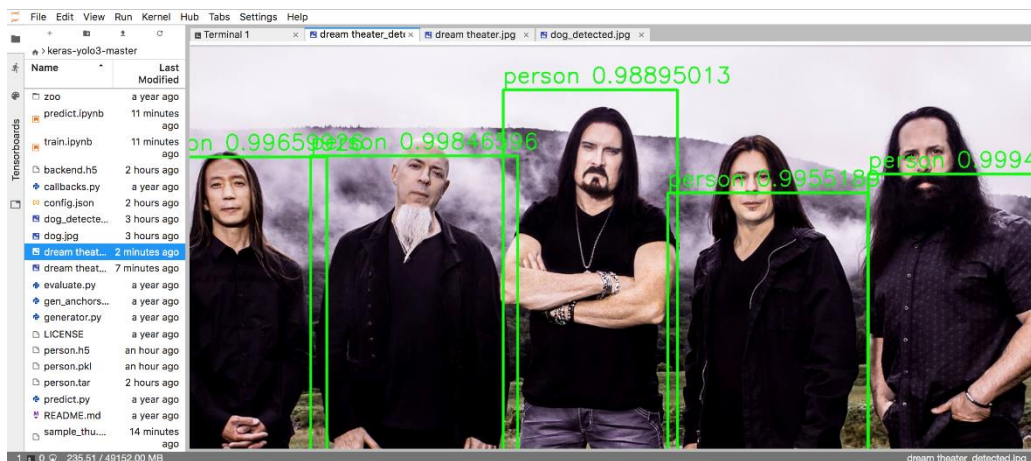
- 主要使用train.py, predict.py , train.ipynb , predict.ipynb
- 當訓練的時候如果怕會斷線可使用.py的方式執行，在hub上可使用.ipynb執行即可
- 畫框的部分則是在utils中的bbox.py檔案裡面

資料夾內部

- 可使用train.ipynb檔案進行操作，如果使用ipynb檔就不需要再另外執行cmd指令
- Train.ipynb裡包含了predict的程式(最後一段)，訓練完model後可直接執行，或者另外拉出來，就不用每次都重新執行程式。
- 主要會用到的train.py 以及predict.py，畫框的部分則是在utils中的bbox.py檔案裡面

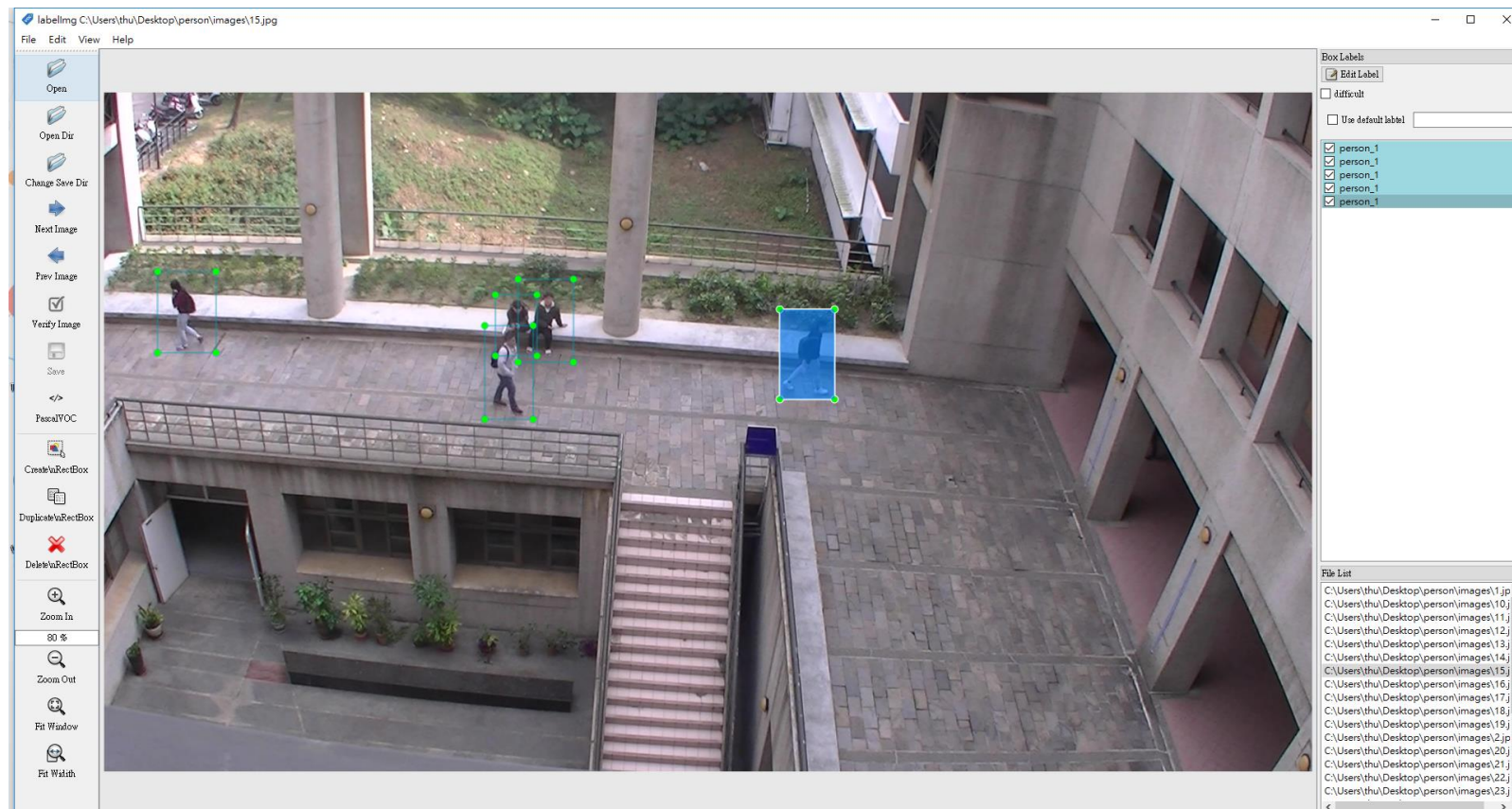
測試

- 可執行這行指令產出結果
- `python3 yolo3_one_file_to_detect_them_all.py -w yolo3.weights -i dog.jpg`
- `-w` 代表載入權重的位置，`-i` 代表載入檔案的位置
- 執行完後會呈現detect 結果，可自行測試



Keras-yolov3

- 要進行訓練之前需要準備 image 跟 label ，也就是上上禮拜學的標記 。
- 將label儲存成Pascal VOC格式之後再存儲的時候比較不會有問題
- 如果label還不熟悉得請參考之前的資料標記講義



Keras-yolov3

- 在做資料標記時，要特別掌握每個框的大小，不能差距過大，否則很容易誤判。
- 要注意再丟進去訓練之前一定要檢查每一張相片都要有label到，否則會噴錯。
- 至於要標記多少張則是需要case by case
- 在做訓練的時候，執行train的時候就不能執行其他程式了，會造成OOM(Out Of Memory)。

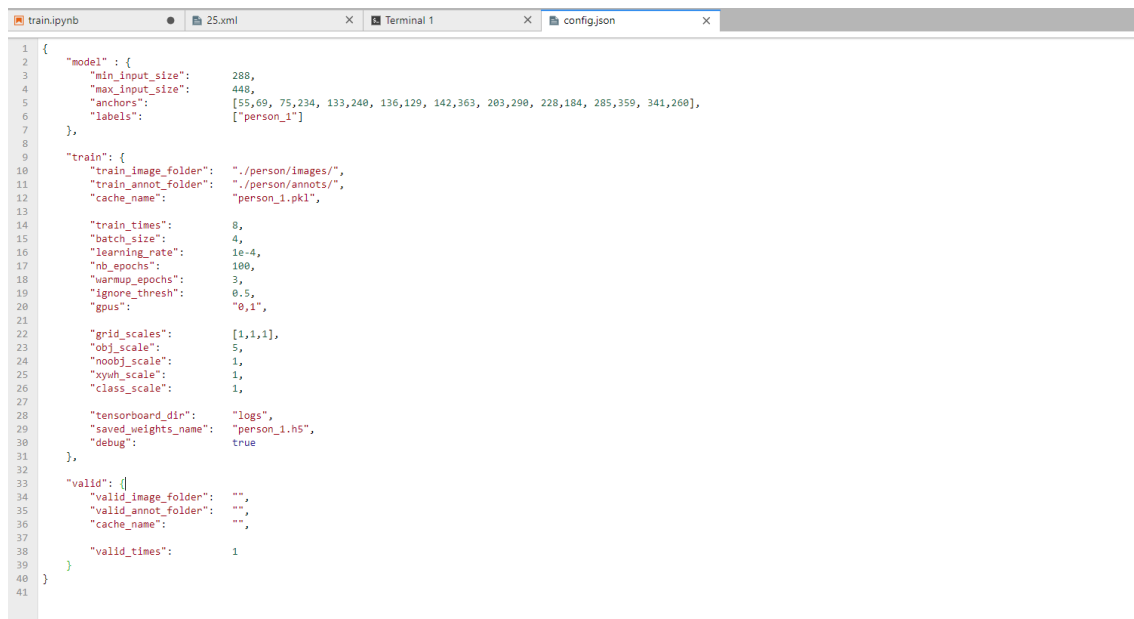
Keras-yolov3

- 常用需要更動的參數

Names	參數說明
labels	照片裡有label到的名子
anchors	錨點，可用預設或gen_anchor.py
train_image_folder	存放訓練資料照片的路徑
train_annot_folder	存放訓練資料labels的路徑
cache_name	pickle 檔，會儲存訓練的紀錄(資料更動後需要刪除，否則則會讀取前一次的訓練資料)
saved_weights_name	你所要存放的權重名稱
nb_epochs	訓練的圈數

Keras-yolov3

- 再準備好資料之後可新增一資料夾(ex:person)，然後打開config.json去修改你要讀取的資料。
- 容易OOM的話可將batch size調小
- 如果有複數Labels可用,隔開(ex:["person_1" , " apple"])



```
1 {
2   "model": {
3     "min_input_size": 288,
4     "max_input_size": 448,
5     "anchors": [55,69, 75,234, 133,240, 136,129, 142,363, 203,290, 228,184, 285,359, 341,260],
6     "labels": ["person_1"]
7   },
8   "train": {
9     "train_image_folder": "./person/images/",
10    "train_annot_folder": "./person/annots/",
11    "cache_name": "person_1.pkl",
12    "train_times": 8,
13    "batch_size": 4,
14    "learning_rate": 1e-4,
15    "nb_epochs": 100,
16    "warmup_epochs": 3,
17    "ignore_thresh": 0.5,
18    "gpus": "0,1",
19    "grid_scales": [1,1,1],
20    "obj_scale": 5,
21    "noobj_scale": 1,
22    "xywh_scale": 1,
23    "class_scale": 1,
24    "tensorboard_dir": "logs",
25    "saved_weights_name": "person_1.h5",
26    "debug": true
27  },
28  "valid": {
29    "valid_image_folder": "",
30    "valid_annot_folder": "",
31    "cache_name": "",
32    "valid_times": 1
33  }
34 }
```

Python檔執行

Keras-yolov3

- 如果不知自己的anchors要設多少可使用(也可用預設即可)
- `python3 gen_anchors.py -c config.json` 獲取推薦值

```
jovyan@jupyter-kevincho-40aiacademy-2etw:~/keras-yolo3$ python3 gen_anchors.py -c config.json
[Errno 21] Is a directory: './person/anns/.ipynb_checkpoints'
Ignore this bad annotation: ./person/anns/.ipynb_checkpoints
./person/images/1.jpg
./person/images/10.jpg
./person/images/11.jpg
./person/images/12.jpg
./person/images/13.jpg
./person/images/14.jpg
./person/images/15.jpg
./person/images/16.jpg
./person/images/17.jpg
./person/images/18.jpg
./person/images/19.jpg
./person/images/2.jpg
./person/images/20.jpg
./person/images/21.jpg
./person/images/22.jpg
./person/images/23.jpg
./person/images/24.jpg
./person/images/25.jpg
./person/images/26.jpg
./person/images/3.jpg
./person/images/4.jpg
./person/images/5.jpg
./person/images/6.jpg
./person/images/7.jpg
./person/images/8.jpg
./person/images/9.jpg
iteration 1: dists = 249.85984916983142
iteration 2: dists = 35.16874714947593
iteration 3: dists = 20.06209145177064
iteration 4: dists = 9.900296449156652
iteration 5: dists = 12.248173094546878
iteration 6: dists = 7.459747519496492
iteration 7: dists = 4.4981561237806975
iteration 8: dists = 1.3073398704477113

average IOU for 9 anchors: 0.93
13,48, 15,40, 15,57, 16,47, 17,58, 18,53, 19,48, 20,63, 20,53
```

Keras-yolov3

- 將以上參數設定完畢後便可開始訓練
- `python3 train.py -c config.json`

```
jovyan@jupyter-kevincho-40aiacademy-2etw:~/keras-yolo3$ python3 train.py -c config.json
/opt/conda/lib/python3.6/site-packages/h5py/_init_.py:36: FutureWarning: Conversion of the second argument of issubdtype from `float` to `np.floating` is deprecated. In future, it will
be treated as `np.float64 == np.dtype(float).type`.
  from ._conv import register_converters as _register_converters
Using TensorFlow backend.
valid_annot_folder not exists. Splitting the training set.
Seen labels:    {'person_1': 135}

Given labels:   ['person_1']

Training on:    ['person_1']

2019-03-17 15:20:51.028416: I tensorflow/core/platform/cpu_feature_guard.cc:141] Your CPU supports instructions that this TensorFlow binary was not compiled to use: SSE4.1 SSE4.2 AVX AVX2
FMA
2019-03-17 15:20:51.291583: I tensorflow/core/common_runtime/gpu/gpu_device.cc:1405] Found device 0 with properties:
name: GeForce GTX 1080 Ti major: 6 minor: 1 memoryClockRate(GHz): 1.582
pciBusID: 0000:0e:00.0
totalMemory: 10.92GiB freeMemory: 274.50MiB
2019-03-17 15:20:51.291649: I tensorflow/core/common_runtime/gpu/gpu_device.cc:1484] Adding visible gpu devices: 0
2019-03-17 15:20:51.580121: I tensorflow/core/common_runtime/gpu/gpu_device.cc:965] Device interconnect StreamExecutor with strength 1 edge matrix:
2019-03-17 15:20:51.580182: I tensorflow/core/common_runtime/gpu/gpu_device.cc:971]      0
2019-03-17 15:20:51.580194: I tensorflow/core/common_runtime/gpu/gpu_device.cc:984] 0:    N
2019-03-17 15:20:51.580348: I tensorflow/core/common_runtime/gpu/gpu_device.cc:1097] Created TensorFlow device (/job:localhost/replica:0/task:0/device:GPU:0 with 210 MB memory) -> physical
1 GPU (device: 0, name: GeForce GTX 1080 Ti, pci bus id: 0000:0e:00.0, compute capability: 6.1)
2019-03-17 15:20:51.581102: E tensorflow/stream_executor/cuda/cuda_driver.cc:903] failed to allocate 210.50M (220725248 bytes) from device: CUDA_ERROR_OUT_OF_MEMORY
```


Keras-yolov3

- Train 完之後會跑出.pkl以及.h5檔
- 這時候可用python3 predict.py -c config.json -i (檔案)來查看預測
- 檔案預設存放在output資料夾內

```
(myenv) jovyan@jupyter-kevincho-40aiacademy-2etw:~/keras-yolo3$ python3 predict.py -c config.json -i sample_thu.mp4
/opt/conda/lib/python3.6/site-packages/h5py/_init_.py:36: FutureWarning: Conversion of the second argument of issubdtype from `float` to `np.floating` is deprecated. In future, it will
be treated as `np.float64 == np.dtype(float).type`.
  from ._conv import register_converters as _register_converters
Using TensorFlow backend.
2019-03-17 16:03:34.053971: I tensorflow/core/platform/cpu_feature_guard.cc:141] Your CPU supports instructions that this TensorFlow binary was not compiled to use: SSE4.1 SSE4.2 AVX AVX2
FMA
2019-03-17 16:03:34.350549: I tensorflow/core/common_runtime/gpu/gpu_device.cc:1405] Found device 0 with properties:
name: GeForce GTX 1080 Ti major: 6 minor: 1 memoryClockRate(GHz): 1.582
pciBusID: 0000:0e:00.0
totalMemory: 10.92GiB freeMemory: 274.50MiB
2019-03-17 16:03:34.350614: I tensorflow/core/common_runtime/gpu/gpu_device.cc:1484] Adding visible gpu devices: 0
2019-03-17 16:03:34.742895: I tensorflow/core/common_runtime/gpu/gpu_device.cc:965] Device interconnect StreamExecutor with strength 1 edge matrix:
2019-03-17 16:03:34.742971: I tensorflow/core/common_runtime/gpu/gpu_device.cc:971]      0
2019-03-17 16:03:34.742991: I tensorflow/core/common_runtime/gpu/gpu_device.cc:984] 0:    N
2019-03-17 16:03:34.743166: I tensorflow/core/common_runtime/gpu/gpu_device.cc:1097] Created TensorFlow device (/job:localhost/replica:0/task:0/device:GPU:0 with 210 MB memory) -> physical
1 GPU (device: 0, name: GeForce GTX 1080 Ti, pci bus id: 0000:0e:00.0, compute capability: 6.1)
2019-03-17 16:03:34.744300: E tensorflow/stream_executor/cuda/cuda_driver.cc:903] failed to allocate 210.50M (220725248 bytes) from device: CUDA_ERROR_OUT_OF_MEMORY
```

Jupyter notebook

Train

- 改動需要用到的參數, train_annot_folder, train_image_folder, labels, saved_weights_name, nb_epochs 等
- 如果有複數labels的時候, Seen labels 與Given labels需要對齊, 否則output的labels會出錯
- 當有出現Epochs的訓練時代表已開始訓練

```
[Errno 21] Is a directory: './person/anns/.ipynb_checkpoints'  
Ignore this bad annotation: ./person/anns/.ipynb_checkpoints  
valid_annot_folder not exists. Splitting the training set.  
Seen labels:   {'person_1': 135}
```

```
Given labels:  ['person_1']
```

```
Training on:   ['person_1']
```

```
/opt/conda/lib/python3.6/site-packages/keras/callbacks.py:999: UserWarning: `epsilon` argument is deprecated and will be removed, use `min_delta` instead.  
  warnings.warn("`epsilon` argument is deprecated and "
```

```
Epoch 1/103
```

```
resizing: 416 416
```

```
resizing: 288 288
```

```
resizing: 288 288
```

```
resizing: 288 288
```

```
resizing: 320 320
```

```
resizing: 352 352
```

```
resizing: 288 288
```

```
resizing: 416 416
```

```
resizing: 320 320
```

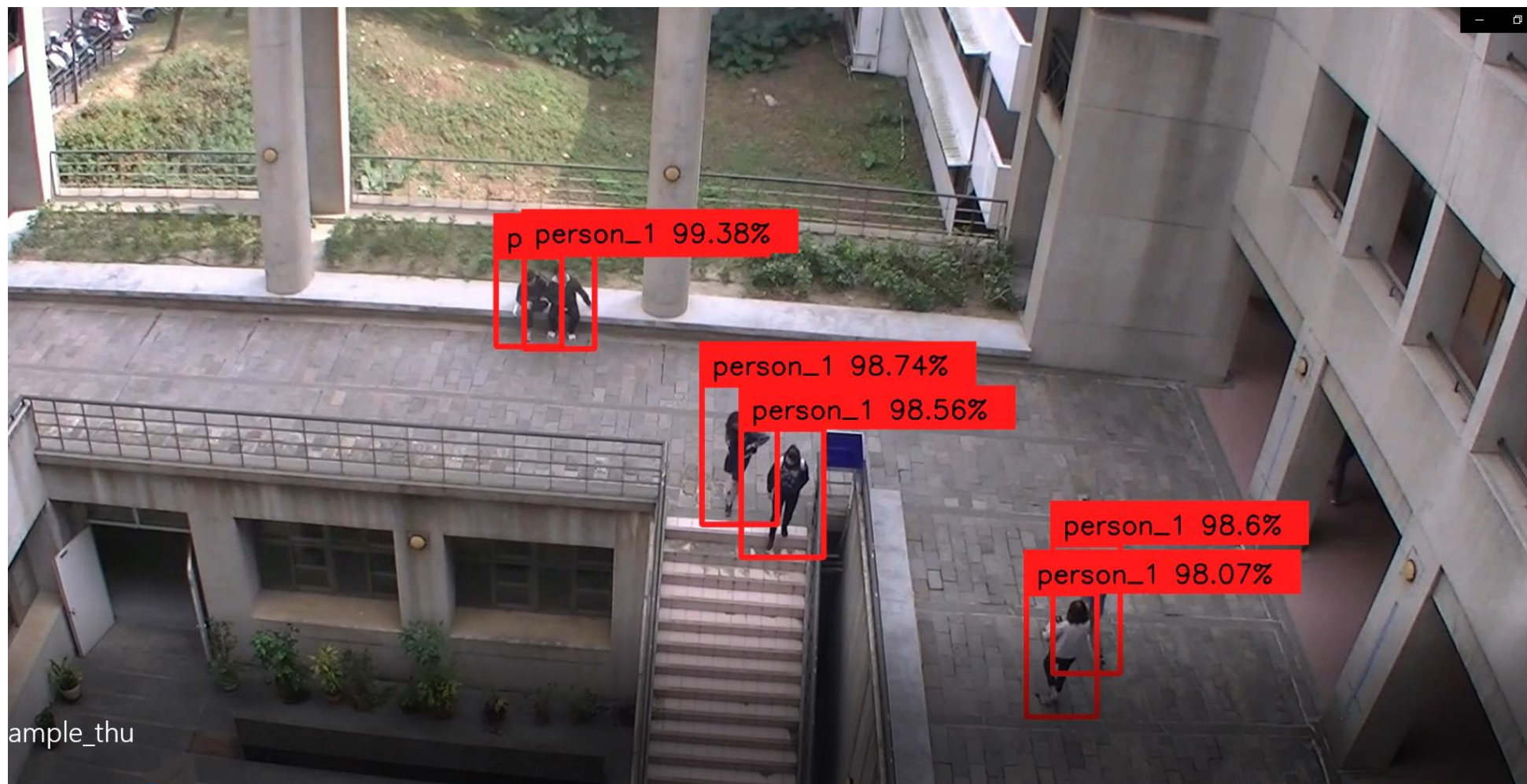
```
- 40s - loss: 563.0123 - yolo_layer_1_loss: 55.7123 - yolo_layer_2_loss: 156.2393 - yolo_layer_3_loss: 3  
51.0607
```

Predict

- 訓練結束後可執行Predict.ipynb，並更改input_path,labels,與load_model裡的名字
- 在input_path中可放入.mp4或圖片檔皆可predict
- Predict完的結果可在output資料夾裡查看

Keras-yolov3

- 察看結果



實作參考

- Yolo實作影片
- <https://www.youtube.com/watch?v=DhkuUVzK2Vw&t=4s>
- 看Yolo3代碼的筆記（寫得也就自己能到懂）
- <https://www.twblogs.net/a/5bf34200bd9eee040518bf66>

AND

After Detection

What else you can do?

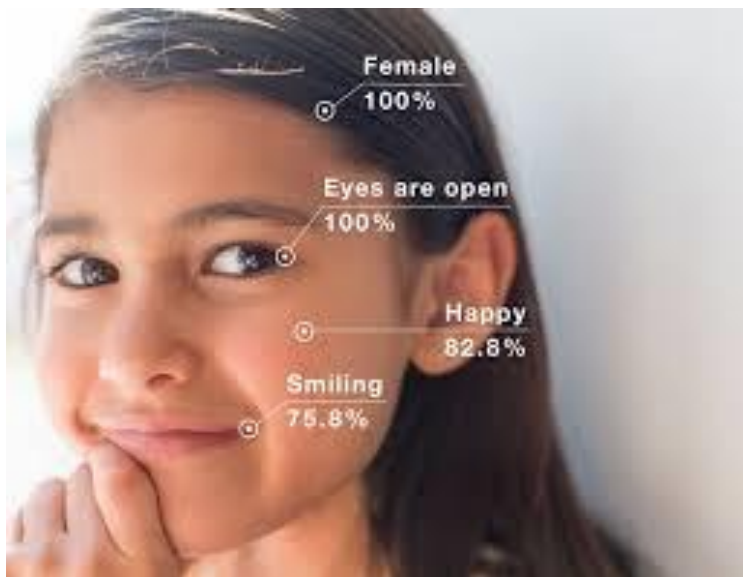
Face recognition

Face recognition



Mobi

▶ 原理



- 人臉識別是由一系列的幾個相關問題組成的：
- 1.找到一張圖片中的所有人脸。
- 2.對於每一張臉來說，無論光線明暗或面朝別處，它依舊能夠識別出是同一個人的臉。
- 3.能夠在每一張臉上找出可用於與他人區分的獨特之處，比如說眼睛有多大，臉有多長等等。
- 4.將這張臉的特點與已知的所有人脸進行比較，以確定這個人的姓名。

Tracking

Tracking



Pose Estimation

Real-time Multi-Person 2D Pose Estimation Using Part Affinity Fields

Zhe Cao, Tomas Simon, Shih-En Wei, Yaser Sheikh
Carnegie Mellon University

Semantic Segmentation

Semantic Segmentation

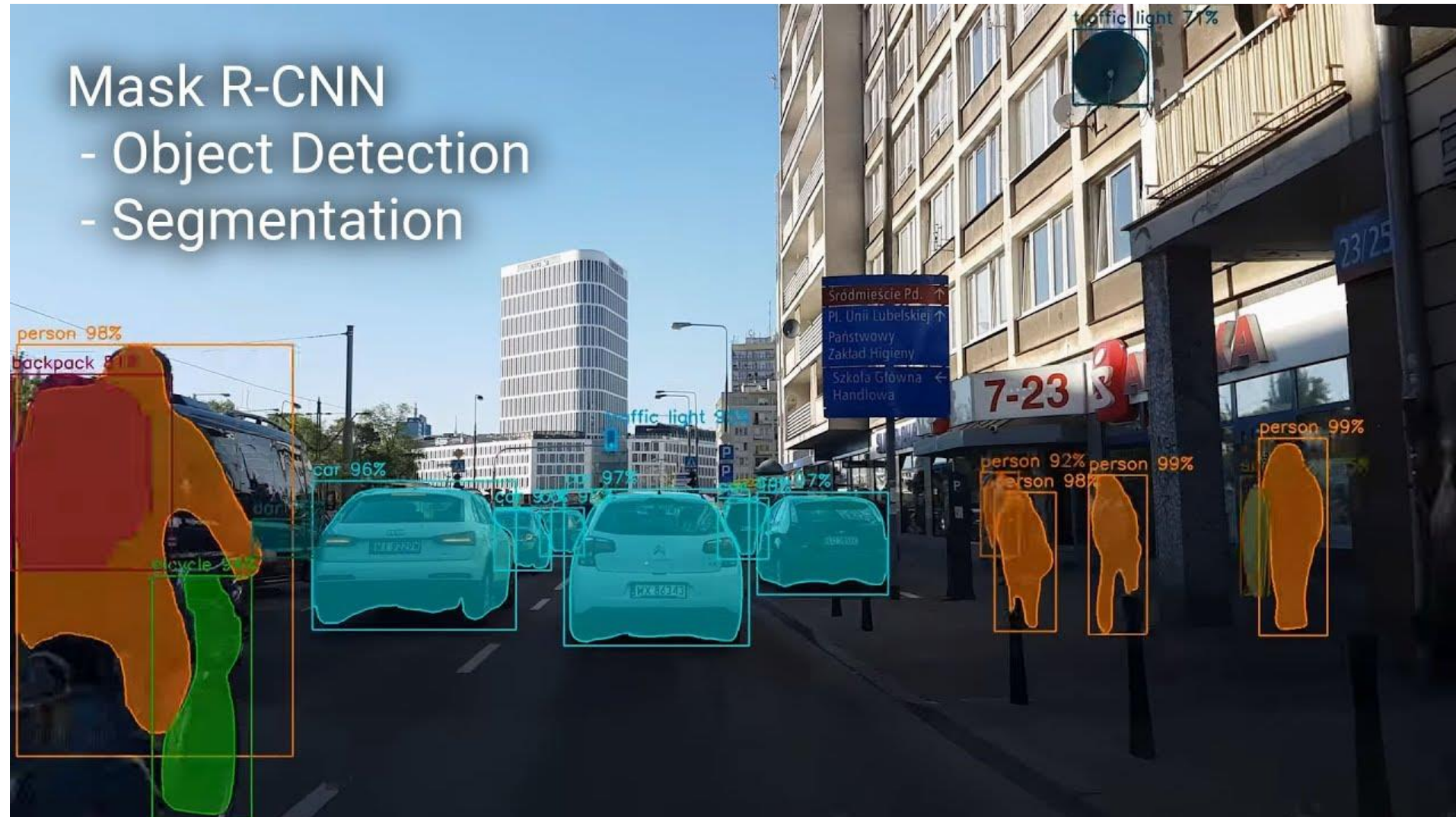
ICNet for Real-Time Semantic Segmentation on High-Resolution Images

Hengshuang Zhao¹ Xiaojuan Qi¹ Xiaoyong Shen¹ Jianping Shi² Jiaya Jia¹

¹The Chinese University of Hong Kong ²SenseTime Group Limited

*Each frame in the video is processed independently at the rate of 30 fps on a 1024*2048 resolution image.*

Semantic Segmentation



Person re-identification

Person re-identification

Cam1



Cam2



Start_Time:0secs
End_time:1secs
Video_Total time:1secs

Thanks