



Progress report

進度報告

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To do list

Person Re-Identification system

1. Person detection	Object detection (MobileNetSSD) Openpose	done cont.
2. 特徵擷取網路架構	Torchreid (A Library for Deep Learning Person Re-Identification in Pytorch) backbone: ResNet50 backbone: OSNet Tensorflow -> Pytorch 世超-DSPF (backbone: SE-ResNeXt)	done cont. cont.
3. 動態每天分群給編號 4. 使用3的分群編號， 去做Query識別	Feature Extractor Compute distance matrix Real-time system	done done cont.
5. 確認編號的確實身份	結合Line bod應用	cont.

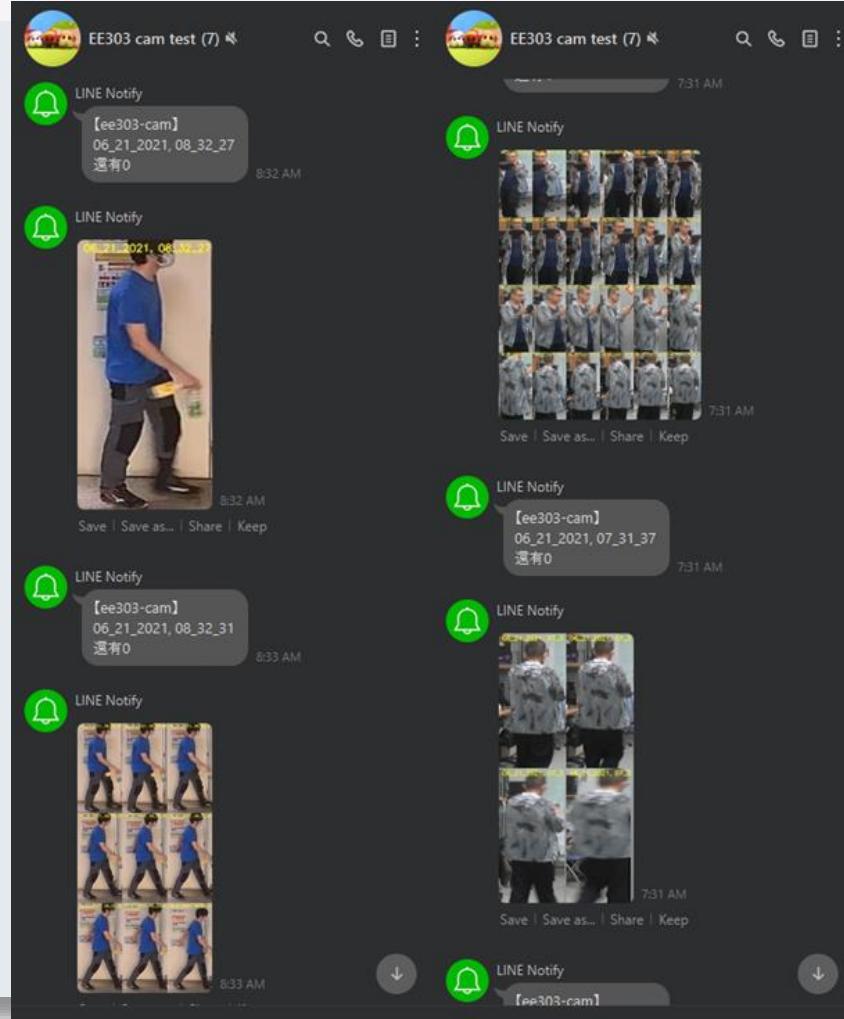


What I have done

Person detection

Object detection (MobileNetSSD)	done
Openpose	cont.

1	aeroplane
2	bicycle
3	bird
4	boat
5	bottle
6	bus
7	car
8	cat
9	chair
10	cow
11	dining table
12	dog
13	horse
14	motorbike
15	person
16	pottedplant
17	sheep
18	sofa
19	train
20	TV monitor



Use person bounding box

```
main_loop:
...
blob = cv2.dnn.blobFromImage(
    cv2.resize(frame, (300, 300)), 1.0 / 127.5, (300, 300), 127.5)
net.setInput(blob)
detections = net.forward()

for i in range(0, detections.shape[2]):
    confidence = detections[0, 0, i, 2]

    if confidence > threshold:

        idx = int(detections[0, 0, i, 1])

        bounding_box = detections[0, 0, i, 3:7] * \
            np.array([origin_w, origin_h, origin_w, origin_h])
        x_start, y_start, x_end, y_end = bounding_box.astype(
            'int')

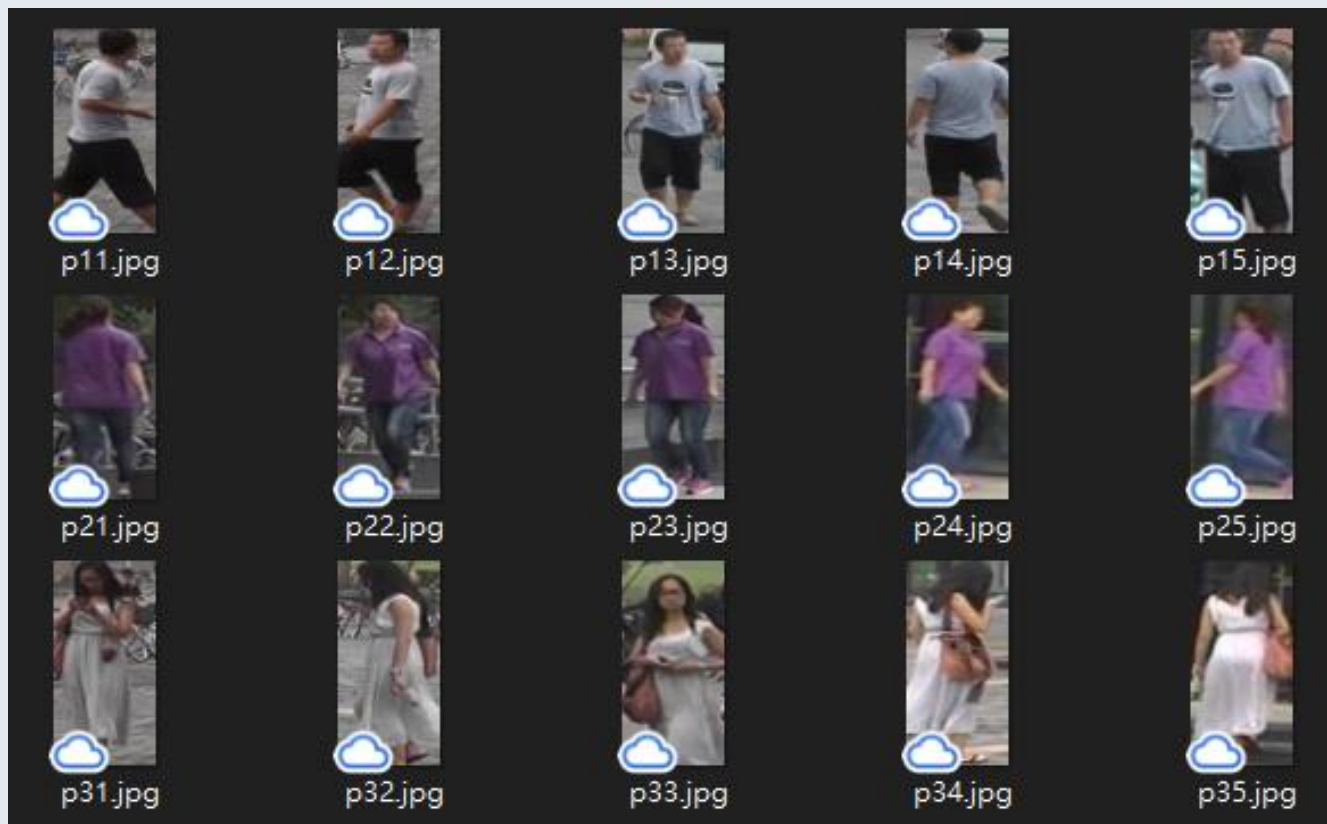
        if idx == 15:
            #偵測到人
            person_bounding_box = frame[y_start:y_end, x_start:x_end].astype(
                'uint8')
            run_re_id(person_bounding_box)
            ...
```

Torchreid training ResNet50

Hyperparamters	
Data set	Market1501
batch_size_train	32
model name	resnet50
optimizer	adam
lr	0.0003
lr_scheduler	single_step
stepsize	20
max_epoch	60

Results	
mAP	67.40%
Rank-1	85.00%
Elapsed	01:15:48

Test image



Torchreid ResNet50 Feature Extractor

layer name	output size	18-layer	34-layer	50-layer	101-layer	152-layer
conv1	112×112	7×7, 64, stride 2				
conv2_x	56×56	3×3 max pool, stride 2				
		$\begin{bmatrix} 3 \times 3, 64 \\ 3 \times 3, 64 \end{bmatrix} \times 2$	$\begin{bmatrix} 3 \times 3, 64 \\ 3 \times 3, 64 \end{bmatrix} \times 3$	$\begin{bmatrix} 1 \times 1, 64 \\ 3 \times 3, 64 \\ 1 \times 1, 256 \end{bmatrix} \times 3$	$\begin{bmatrix} 1 \times 1, 64 \\ 3 \times 3, 64 \\ 1 \times 1, 256 \end{bmatrix} \times 3$	$\begin{bmatrix} 1 \times 1, 64 \\ 3 \times 3, 64 \\ 1 \times 1, 256 \end{bmatrix} \times 3$
conv3_x	28×28	$\begin{bmatrix} 3 \times 3, 128 \\ 3 \times 3, 128 \end{bmatrix} \times 2$	$\begin{bmatrix} 3 \times 3, 128 \\ 3 \times 3, 128 \end{bmatrix} \times 4$	$\begin{bmatrix} 1 \times 1, 128 \\ 3 \times 3, 128 \\ 1 \times 1, 512 \end{bmatrix} \times 4$	$\begin{bmatrix} 1 \times 1, 128 \\ 3 \times 3, 128 \\ 1 \times 1, 512 \end{bmatrix} \times 4$	$\begin{bmatrix} 1 \times 1, 128 \\ 3 \times 3, 128 \\ 1 \times 1, 512 \end{bmatrix} \times 8$
conv4_x	14×14	$\begin{bmatrix} 3 \times 3, 256 \\ 3 \times 3, 256 \end{bmatrix} \times 2$	$\begin{bmatrix} 3 \times 3, 256 \\ 3 \times 3, 256 \end{bmatrix} \times 6$	$\begin{bmatrix} 1 \times 1, 256 \\ 3 \times 3, 256 \\ 1 \times 1, 1024 \end{bmatrix} \times 6$	$\begin{bmatrix} 1 \times 1, 256 \\ 3 \times 3, 256 \\ 1 \times 1, 1024 \end{bmatrix} \times 23$	$\begin{bmatrix} 1 \times 1, 256 \\ 3 \times 3, 256 \\ 1 \times 1, 1024 \end{bmatrix} \times 36$
conv5_x	7×7	$\begin{bmatrix} 3 \times 3, 512 \\ 3 \times 3, 512 \end{bmatrix} \times 2$	$\begin{bmatrix} 3 \times 3, 512 \\ 3 \times 3, 512 \end{bmatrix} \times 3$	$\begin{bmatrix} 1 \times 1, 512 \\ 3 \times 3, 512 \\ 1 \times 1, 2048 \end{bmatrix} \times 3$	$\begin{bmatrix} 1 \times 1, 512 \\ 3 \times 3, 512 \\ 1 \times 1, 2048 \end{bmatrix} \times 3$	$\begin{bmatrix} 1 \times 1, 512 \\ 3 \times 3, 512 \\ 1 \times 1, 2048 \end{bmatrix} \times 3$
	1×1	average pool, 1000-d fc, softmax				
FLOPs		1.8×10^9	3.6×10^9	3.8×10^9	7.6×10^9	11.3×10^9

Input image size (256, 128)

ResNet50

Output feature (1, 2048)

Feature Extractor

(ResNet50)

image_list_p1

image_list_p2

p21

p14

p23

p11

p12

p22

p31

p23

```
features_1 = extractor(image_list_p1)
features_2 = extractor(image_list_p2)
print("features_1 ",features_1)
print("features_2 ",features_2)
```

=>>>

features_1

```
torch.Size([5, 2048])
[0.0785, 0.3583, 0.3029, ..., 0.0520, 0.0000, 1.0073],
[0.0664, 0.0527, 0.3985, ..., 0.0327, 0.0454, 1.1085],
[0.5219, 1.2649, 0.0334, ..., 0.0906, 0.1035, 0.2110],
[0.0283, 0.2683, 0.5947, ..., 0.3594, 0.0563, 0.2664],
[0.4860, 1.0593, 0.0162, ..., 0.0346, 0.0486, 0.1544]
```

features_2

```
torch.Size([3, 2048])
[0.3463, 1.4591, 0.0021, ..., 0.1234, 0.0718, 0.1940],
[0.0032, 0.0800, 0.2054, ..., 0.1168, 0.0088, 0.6943],
[0.4860, 1.0593, 0.0162, ..., 0.0346, 0.0486, 0.1544]
```

```
image_list_p1 = [
    'test_image/p11.jpg',
    'test_image/p12.jpg',
    'test_image/p22.jpg',
    'test_image/p31.jpg',
    'test_image/p23.jpg',
]
```

```
image_list_p2 = [
    'test_image/p21.jpg',
    'test_image/p14.jpg',
    'test_image/p23.jpg'
]
```

Compute distance matrix

use **euclidean**

$$d(x, y) := \sqrt{(x_1 - y_1)^2 + (x_2 - y_2)^2 + \dots + (x_n - y_n)^2} = \sqrt{\sum_{i=1}^n (x_i - y_i)^2}$$

image_list_p1

image_list_p2

p21

p14

p23

p11

p12

p22

p31

p23

689

71

666

679

82

663

54

652

45

662

602

615

69

625

0

```
image_list_p1 = [
    'test_image/p11.jpg',
    'test_image/p12.jpg',
    'test_image/p22.jpg',
    'test_image/p31.jpg',
    'test_image/p23.jpg',
]
```

```
image_list_p2 = [
    'test_image/p21.jpg',
    'test_image/p14.jpg',
    'test_image/p23.jpg',
]
```

```
features_1 = extractor(image_list_p1)
```

```
features_2 = extractor(image_list_p2)
```

```
distmat = metrics.compute_distance_matrix(features_1, features_2)
```

```
print(distmat)
```

```
=>>>
```

```
[689.1064, 71.9318, 666.6927],
[679.5190, 82.5314, 663.0724],
[ 54.5557, 652.5553, 45.2095],
[662.2687, 602.6262, 615.7032],
[ 69.6683, 625.9954, 0.0000]
```

Compute distance matrix

use cosine

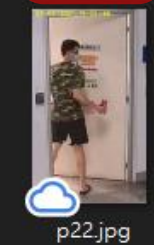
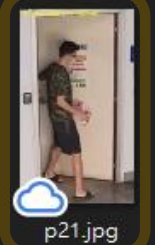
```
#cosine
distmat2 = metrics.compute_distance_matrix(features_1, features_2, metric='cosine')
print(distmat2)
```

```
=>>>
[0.5796531438827515, 0.0591344833374023, 0.5568877458572388],
[0.5758468508720398, 0.0683749318122863, 0.5579483509063721],
[0.04654204845428467, 0.545458018779754, 0.038355231285095215],
[0.5825185775756836, 0.517381191253662, 0.5374826192855835],
[0.06009882688522339, 0.527764081954956, 1.1920928955078125e-07]
```

```
features_1 = extractor(image_list_p1)
features_2 = extractor(image_list_p2)
```

```
distmat = metrics.compute_distance_matrix(features_1, features_2)
print(distmat)
```

```
=>>>
[689.1064, 71.9318, 666.6927],
[679.5190, 82.5314, 663.0724],
[ 54.5557, 652.5553, 45.2095],
[662.2687, 602.6262, 615.7032],
[ 69.6683, 625.9954, 0.0000]
```



False Positive
(一樣)

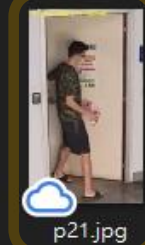
False Negative
(不一樣)

EE3F test

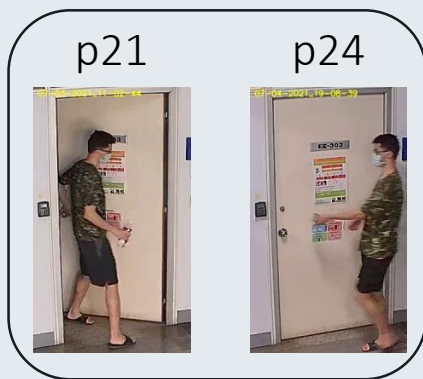
image total = 6+7=13

if V>110	Prediction					
	p21	p12	p23	p33	p14	p35
p11	50.9017	20.3506	247.2152	381.1587	77.9178	370.6215
p22	54.5635	127.8018	102.7195	340.7387	140.1973	341.6727
p13	75.845	35.5033	277.5178	385.7975	48.2986	380.8727
p34	452.6635	440.9745	373.1768	89.053	358.6364	84.5405
p24	188.0116	220.797	78.1254	227.1243	132.516	252.0209
p15	142.2623	76.3605	228.548	268.5231	36.2512	279.7692
p41	199.574	179.9592	343.4623	363.6335	186.4541	368.7164

	Ground truth					
	p21	p12	p23	p33	p14	p35
p11		1			1	
p22	1		1			
p13		1			1	
p34				1		1
p24	1		1			
p15		1			1	
p41						



False Positive
(一樣)



False Negative
(不一樣)



Next to do

Next to do

Person Re-Identification system		
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5. 確認編號的確實身份	結合Line bod應用	cont.