

## 深度學習TensorFlow實務

驗證碼辨識

Lab7

# 1. 驗證碼介紹

### 什麼是驗證碼?

■ 驗證碼(Captcha)是根據隨機字元生成一幅圖片,然後在圖片中加入干擾像素,使用者必須手動填入,防止有人利用機器人自動批量註冊、灌水、發垃圾廣告等等

驗證碼的作用是驗證用戶是真人還是機器人; 設計理念是對人友好, 對機器難。





#### CAPTCHA 測試定義

- 全自動區分電腦和人類的公開圖靈測試(Completely Automated Public Turing test to tell Computers and Humans Apart, CAPTCHA)
- 根據 CAPTCHA 測試的定義,產生驗證碼圖片的演算法必須公開,即使該演算法可能有專利保護。這樣做是證明想破解就需要解決一個不同的人工智慧難題,而非僅靠發現原來的秘密演算法,而後者可以用還原工程等途徑得到

# 驗證碼種類

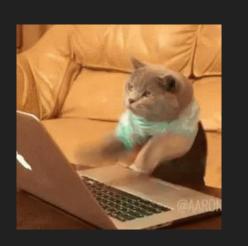
Scheme	T. 1 ( )	E1-	Security	Features	- Excluded Characters						
	Website(s)	Example	Anti-segmentation	Anti-recognition							
Wikipedia	wikipedia.org	weanhicks	Overlapping characters, Enligh letters	Kotation, distortion, waving							
Microsoft	{live, bing, miscosoft}.com {office, linkedin}.com	G/SYA	Overlapping characters, solid background	Different font styles, varied font sizes, rotation, waving	0, 1, 5, D, G, I, Q, U						
eBay	ebay.com	434525	Overlapping characters, Only arabic numerals	Character rotating, distortion and waving	-						
Baidu	{baidu, qq}.com	THE P	Occluding lines, overlapping, only Enligh letters	Varied font size, color, rotation, disortion and waving	Z						
Google	<pre>google.{com,co.in,co.jp, co.uk,ru,com.br,fr com.hk,it,ca,es,com.mx} youtube.com</pre>	hedodeg	Overlapping characters, Enligh letters	Varied font sizes & color, rotation, disortion, waving	-						
Alipay	<pre>{alipay, tmall}.com {taobao, login.tmall}.com alipayexpress.com</pre>	YDKT	English letters and arabic numerals, overlapping characters	Rotation and distortion	0, 1, I, L, O						
JD	jd.com	WV5U	English letters and arabic numerals, overlapping characters	Rotation and distortion	0, 1, 2, 7, 9, D, G, I, J, L, O, P, Q, Z						
Qihu360	360. cn	PUXIG	English letters and arabic numerals, overlapping characters	Varied font sizes, rotation and distortion	0, I, L, O, T, i, l, o, t, q						
Sina	sina.cn	DOURE	English letters and arabic numerals, overlapping characters	Rotation, distortion, waving	1, 9, 0, D, I, J, L, O, T, i, j, l, o, t, g, r						
Weibo	weibo.cn	46811	English letters and arabic numerals, overlapping characters, occluding lines	Rotation and distortion	0, 1, 5, D, G, I, Q, U						
Sohu	sohu.com	DE	Complex background, occluding lines, and overlapping	Varied font size, color and rotation	0, 1, i, l, o, z						
Table 1: Text-based captcha schemes tested in our experiments.											

# 2. 驗證碼生成

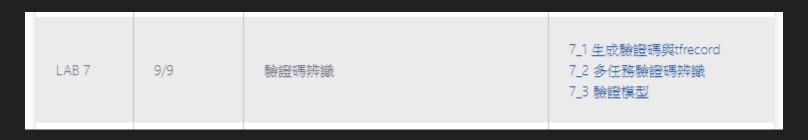
#### 驗證碼來源?

- 或許你可以寫一支爬蟲程序,截取個5萬張驗證碼,再自己手動標記答案上去,但這樣有點太費時了,或許我們可以試着模仿產生一些驗證碼看看。
- 不過當然,我們產生的訓練集必須非常接近真實的驗證碼,否則最後訓練完可能用在真實的驗證碼上效果會非常的差。





■ Code: 7\_1 生成驗證碼與tfrecord.ipynb







■ 定義驗證碼包含 0-9 字元

```
In [1]: from captcha.image import ImageCaptcha
import numpy as np
from PIL import Image
import random
import sys
import os
import tensorflow as tf
import matplotlib.pyplot as plt

In [2]: number = [str(i) for i in range(10)]
print(number)
CHAR_SET = number
['0', '1', '2', '3', '4', '5', '6', '7', '8', '9']
```

■ 隨機生成4個字元,並輸出驗證碼圖片

```
In [3]: # <u>隨機生成4個字元</u>
        def random_captcha_text(char_set=number,captcha_size=4):
            return ''.join(random.choices(CHAR SET,k=4))
        # 生成對應字元的驗證碼
        def gen captcha text and image(path):
            image = ImageCaptcha()
            captcha text = random captcha text()
            captcha = image .generate(captcha text)
            image .write(captcha text, path + captcha text + '.jpg')
        path = 'captcha/images/'
        num = 10000
        if not os.path.exists(path):
            os.makedirs(path)
        for i in range(num):
            gen captcha text and image(path)
            sys.stdout.write('\r>> Creating image %d/%d' % (i+1,num))
            sys.stdout.flush()
        sys.stdout.write('\n')
        sys.stdout.flush()
        >> Creating image 10000/10000
```

0000 0000.jpg	00 01 0001.jpg	0 0 2 0002.jpg	0004.jpg	0005.jpg	0006.jpg	<b>0</b> 007	00 08 0008.jpg	0 009 0009.jpg	0: 010.jpg
007 1 0011.jpg	0012.jpg	00/A 0013.jpg	001A ·	00 1.5 0015.jpg	0; 0.16 0016.jpg	00: 17. 0017.jpg	00.18 pg(.8100	00 <b>2</b> 0	0024 0024.jpg
00.23 0027.jpg	002 <b>8</b> 0028.jpg	0030.jpg	0032.jpg	003-5 0035.jpg	0037.jpg	0038.jpg	0040 0040.jpg	0042 0042.jpg	0043.jpg
00 A 4	00 4 5 0045.jpg	004 6 0046.jpg	0047.jpg	00 5 0 0050.jpg	0051.jpg	0053.jpg	00 55 0055.jpg	0 0 56 0056.jpg	005 7 0057.jpg
0 0 58 0058.jpg	0066.jpg	0068.jpg	0071.jpg	0 0 12 0072.jpg	00/5 0075.jpg	0081.jpg	0.00 3 0083.jpg	00 <b>84</b> 0084.jpg	0 08 5 0085.jpg
0 0 % 0086.jpg	0087.jpg	0089.jpg	0 090 0090.jpg	0091.jpg	0094.jpg	0. <b>0</b> 5 0095.jpg	00.96 0096.jpg	0.0.98 0098.jpg	0099.jpg
0 102 0102.jpg	0106 0106.jpg	07.0 ) 0107.jpg	0108.jpg	01 11 0111.jpg	0115.jpg	07:16 0116.jpg	07 77 0117.jpg	0 1 19 0119.jpg	0:1:0 0120.jpg
0121 0121.jpg	0.10.3 0123.jpg	0125.jpg	01 <b>6</b> 0126.jpg	01.29 0129.jpg	0 1 3 0 0130.jpg	0131.jpg	013A 0134.jpg	0.73.5 0135.jpg	0137.jpg
0142.jpg	0143.jpg	0146.jpg	0 <b>14</b> / 0147.jpg	0148.jpg	0:149 0149.jpg	0 15 1 0151.jpg	0 7 53 0153.jpg	0.7 · 5 · 4 · 0154.jpg	0156.jpg
0158.jpg	0159.jpg	07 60 0160.jpg	0 (A : 6) 0162.jpg	0 . <b>7</b> 63	0164.jpg	0 1 65 0165.jpg	0166.jpg	016 7. 0167.jpg	0168.jpg

## 3. 驗證碼辨識方法

### 驗證碼辨識-方法—

- 把標籤轉為向量,向量長度為40
- 有一個驗證碼:0782
- 它的標籤可以轉為長度為40的向量



0

7

8

2

■ 訓練方法跟 0-9 手寫數字辨識類似

### 驗證碼辨識-方法二

- 將一張驗證碼圖片拆成 4 個標籤
- 有一個驗證碼:0782



Label 1:0000000100

Label\_2:0000000010

Label\_3:0010000000

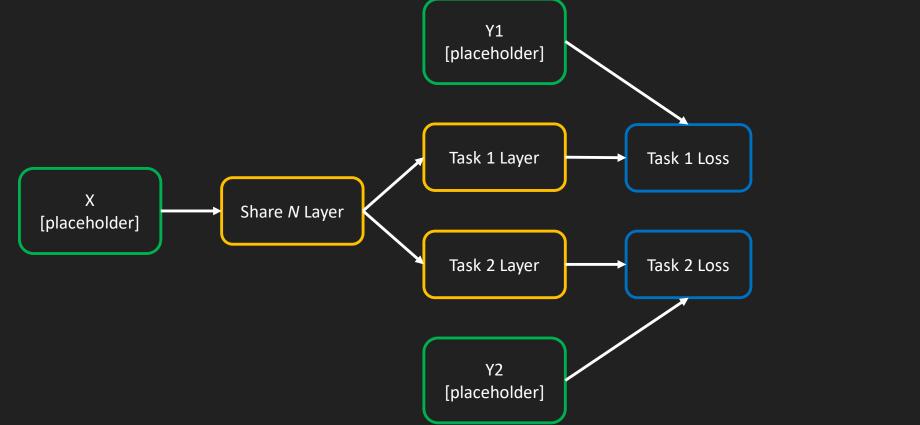
■ 可以使用多任務學習(Multi-task Learning)



# 4. 多任務學習

### 建立多任務圖

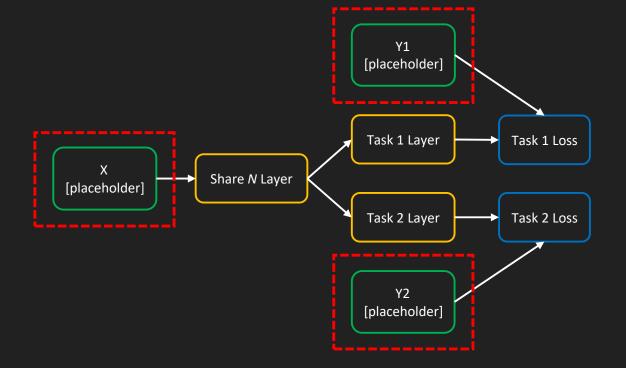
■ 多任務的一個特點是,單一個張量輸入(X)以及多個輸出(Y\_1, Y\_2)。因此在定義 placeholder 時要定義多個輸出。同樣也需要有多個損失函數用於分別計算每個任務的損失



import Tensorflow as tf

#### # 定義佔位符

```
X = tf.placeholder("float", [10, 10], name="X")
Y1 = tf.placeholder("float", [10, 20], name="Y1")
Y2 = tf.placeholder("float", [10, 20], name="Y2")
```



import Tensorflow as tf

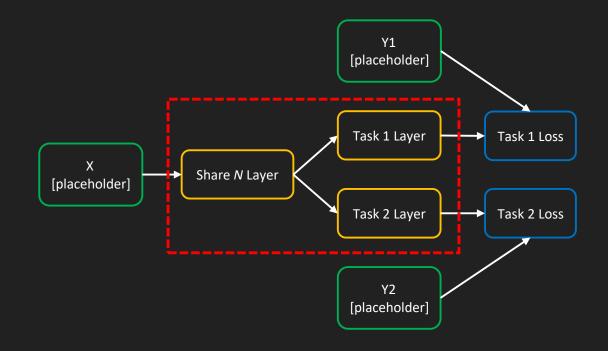
#### # 定義佔位符

```
X = tf.placeholder("float", [10, 10], name="X")
Y1 = tf.placeholder("float", [10, 20], name="Y1")
Y2 = tf.placeholder("float", [10, 20], name="Y2")
```

#### # 定義權重

```
initial_shared_layer_weights = np.random.rand(10,20)
initial_Y1_layer_weights = np.random.rand(20,20)
initial_Y2_layer_weights = np.random.rand(20,20)
```

```
shared_layer_weights = tf.Variable(initial_shared_layer_weights, name="share_W",
dtype="float32")
Y1_layer_weights = tf.Variable(initial_Y1_layer_weights, name="share_Y1", dtype="float32")
Y2 layer weights = tf.Variable(initial Y2 layer weights, name="share Y2", dtype="float32")
```



import Tensorflow as tf

#### # 定義佔位符

```
X = tf.placeholder("float", [10, 10], name="X")
Y1 = tf.placeholder("float", [10, 20], name="Y1")
Y2 = tf.placeholder("float", [10, 20], name="Y2")
```

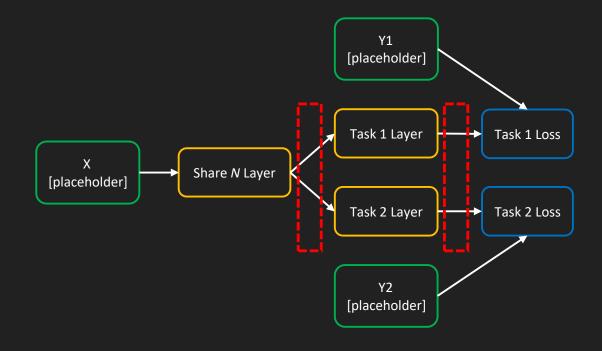
#### # 定義權重

```
initial_shared_layer_weights = np.random.rand(10,20)
initial_Y1_layer_weights = np.random.rand(20,20)
initial_Y2_layer_weights = np.random.rand(20,20)
```

```
shared_layer_weights = tf.Variable(initial_shared_layer_weights, name="share_W", dtype="float32")
Y1_layer_weights = tf.Variable(initial_Y1_layer_weights, name="share_Y1", dtype="float32")
Y2 layer weights = tf.Variable(initial Y2 layer weights, name="share Y2", dtype="float32")
```

#### # 使用relu激活函數

```
shared_layer = tf.nn.relu(tf.matmul(X,shared_layer_weights))
Y1_layer = tf.nn.relu(tf.matmul(shared_layer,Y1_layer_weights))
Y2_layer = tf.nn.relu(tf.matmul(shared_layer,Y2_layer_weights))
```



import Tensorflow as tf

#### # 定義佔位符

```
X = tf.placeholder("float", [10, 10], name="X")
Y1 = tf.placeholder("float", [10, 20], name="Y1")
Y2 = tf.placeholder("float", [10, 20], name="Y2")
```

#### # 定義權重

```
initial_shared_layer_weights = np.random.rand(10,20)
initial_Y1_layer_weights = np.random.rand(20,20)
initial_Y2_layer_weights = np.random.rand(20,20)
```

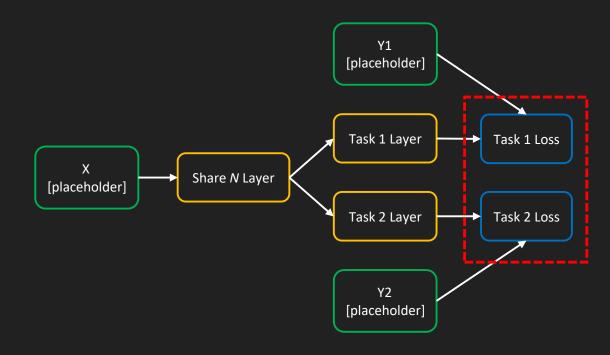
```
shared_layer_weights = tf.Variable(initial_shared_layer_weights, name="share_W", dtype="float32")
Y1_layer_weights = tf.Variable(initial_Y1_layer_weights, name="share_Y1", dtype="float32")
Y2 layer weights = tf.Variable(initial Y2 layer weights, name="share Y2", dtype="float32")
```

#### # 使用relu激活函數

```
shared_layer = tf.nn.relu(tf.matmul(X,shared_layer_weights))
Y1_layer = tf.nn.relu(tf.matmul(shared_layer,Y1_layer_weights))
Y2 layer = tf.nn.relu(tf.matmul(shared layer,Y2 layer weights))
```

#### # 計算loss

```
Y1_Loss = tf.nn.l2_loss(Y1-Y1_layer)
Y2_Loss = tf.nn.l2_loss(Y2-Y2_layer)
```

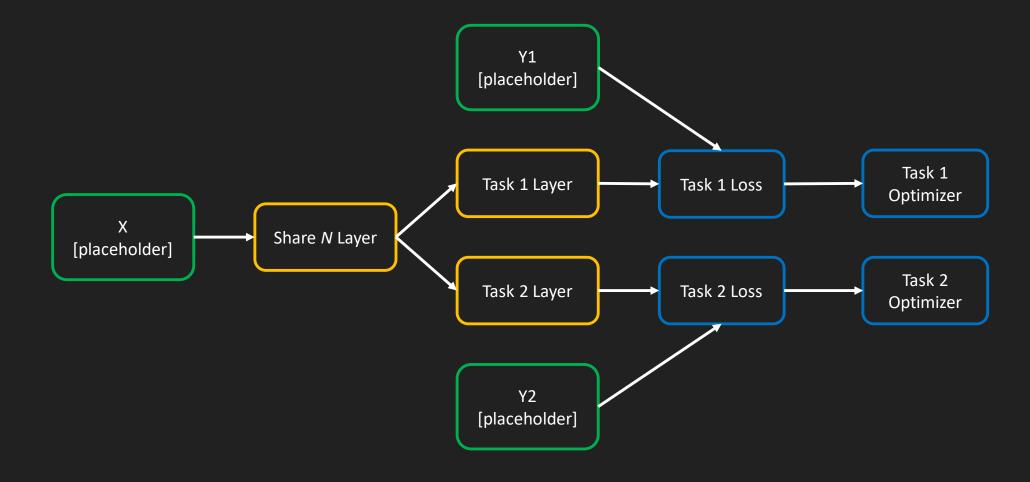


## 多任務學習

- 有了網路的建構後,接下來是訓練,有兩種方式:
  - 交替訓練
  - 聯合訓練

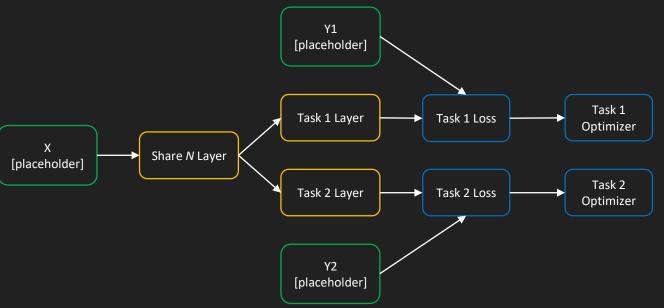
## 多任務學習

■ 交替訓練



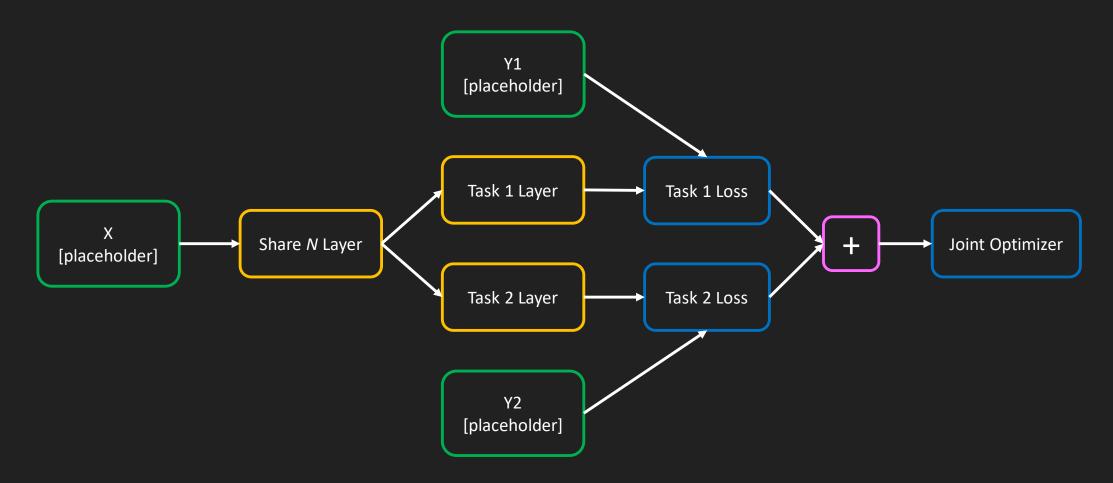
### 交替訓練

```
# 優化器
Y1 op = tf.train.AdamOptimizer().minimize(Y1 Loss)
Y2_op = tf.train.AdamOptimizer().minimize(Y2_Loss)
# Calculation (Session) Code
with tf.Session() as session:
    session.run(tf.initialize all variables())
    for iters in range(10):
        if np.random.rand() < 0.5:</pre>
            _, Y1_loss = session.run([Y1_op, Y1_Loss],
                              X: np.random.rand(10,10)*10,
                              Y1: np.random.rand(10,20)*10,
                              Y2: np.random.rand(10,20)*10
                               })
            print(Y1_loss)
        else:
            _, Y2_loss = session.run([Y2_op, Y2_Loss],
                              X: np.random.rand(10,10)*10,
                              Y1: np.random.rand(10,20)*10,
                              Y2: np.random.rand(10,20)*10
                               })
            print(Y2_loss)
```

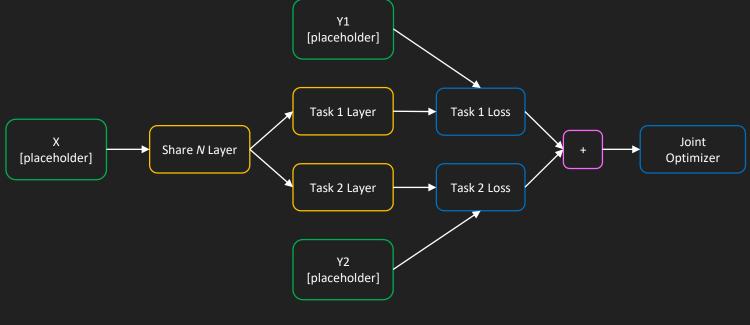


## 多任務學習

■ 聯合訓練



### 交替訓練



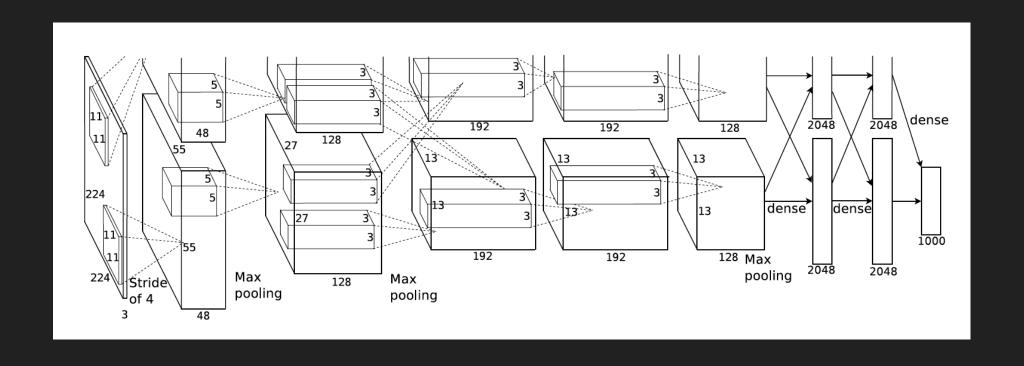
```
# 計算loss
Joint Loss = Y1 Loss + Y2 Loss
# 優化器
Optimiser = tf.train.AdamOptimizer().minimize(Joint Loss)
# Calculation (Session) Code
with tf.Session() as session:
    session.run(tf.initialize all variables())
    _, Joint_Loss = session.run([Optimiser, Joint_Loss],
                      X: np.random.rand(10,10)*10,
                      Y1: np.random.rand(10,20)*10,
                      Y2: np.random.rand(10,20)*10
                      })
    print(Joint_Loss)
```

# 4. 程式碼

- 開啟「7\_1 生成驗證碼與tfrecord.ipynb」
- 將驗證碼圖片與數字標籤轉成 tfrecord
  - train.tfrecord
  - test.tfrecord

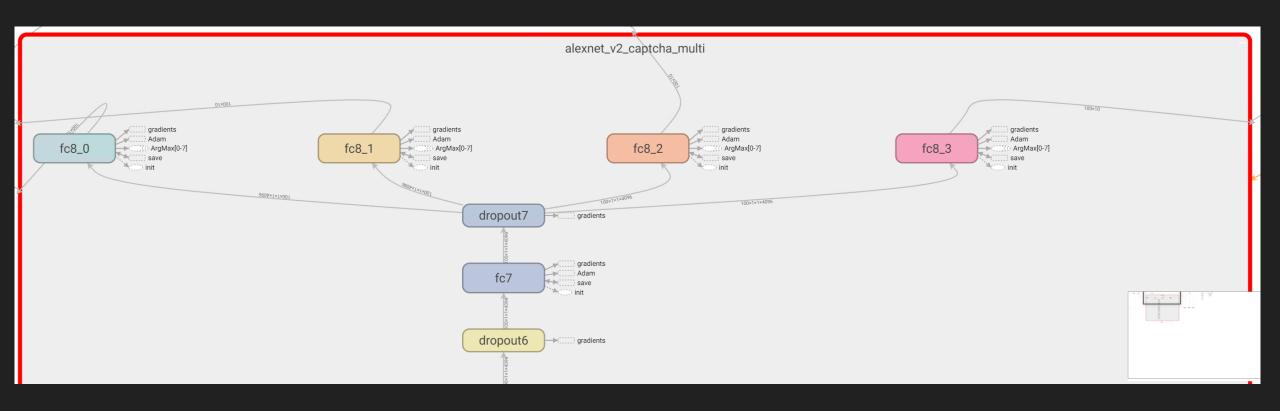
## 多任務學習模型訓練

■ 以 AlexNetv2 為基礎網路



### 多任務學習模型訓練

- 將原本 AlexNetv2 輸出,由單一任務改為多任務
- 4個網路輸出:fc8\_0,fc8\_1,fc8\_2,fc8\_3

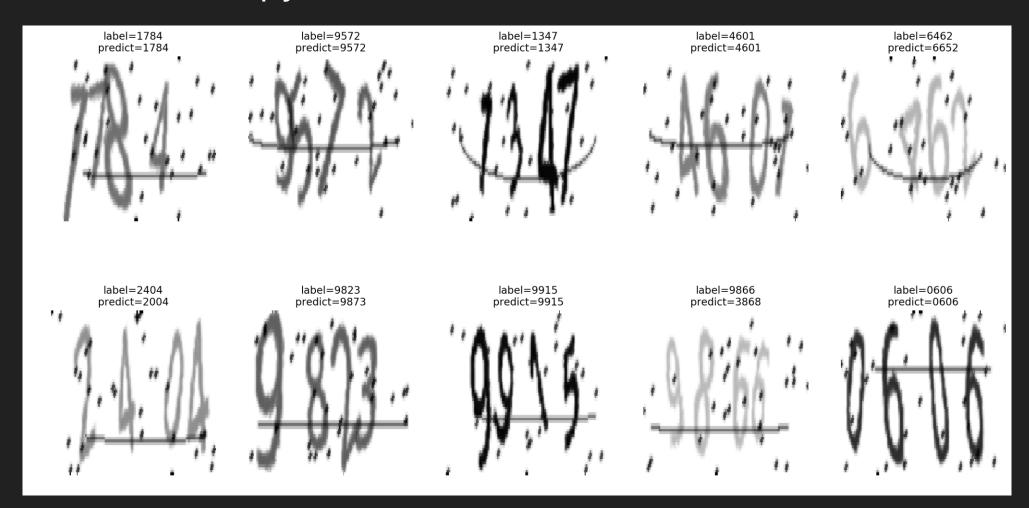


### 多任務學習模型訓練

```
Iter:0/1359 epoch:1,
                                                                    Learning_rate:0.00100
                       Loss:10.469
                                    Accuracy: 0.15, 0.15, 0.09, 0.12
                                                                    Learning rate:0.00100
Iter:20/1359 epoch:1,
                                    Accuracy: 0.10, 0.06, 0.07, 0.11
                        Loss:2.311
Iter:40/1359 epoch:1,
                        Loss:2.302
                                    Accuracy: 0.12, 0.09, 0.12, 0.14
                                                                    Learning rate:0.00100
                                    Accuracy: 0.14, 0.07, 0.06, 0.04
                                                                    Learning rate:0.00100
Iter:60/1359 epoch:2,
                        Loss:2.303
Iter:80/1359 epoch:2,
                        Loss:2.299
                                    Accuracy: 0.14, 0.10, 0.17, 0.09
                                                                    Learning rate:0.00100
                                     Accuracy: 0.13, 0.11, 0.11, 0.12
                                                                    Learning rate: 0.00100
Iter:100/1359 epoch:3,
                        Loss:2.299
Iter:120/1359 epoch:3,
                        Loss:2.303
                                     Accuracy: 0.16, 0.11, 0.08, 0.09
                                                                    Learning rate:0.00100
Iter:140/1359 epoch:4,
                        Loss:2.303
                                     Accuracy: 0.11, 0.09, 0.11, 0.09
                                                                    Learning rate:0.00100
Iter:160/1359 epoch:4,
                        Loss:2.301
                                     Accuracy:0.16,0.15,0.12,0.12
                                                                    Learning rate:0.00100
Iter:180/1359 epoch:4,
                        Loss:2.301
                                     Accuracy: 0.08, 0.14, 0.09, 0.16
                                                                     Learning rate: 0.00100
                                     Accuracy: 0.09, 0.12, 0.16, 0.12
Iter:200/1359 epoch:5,
                        Loss:2.302
                                                                     Learning rate: 0.00100
Iter:220/1359 epoch:5,
                        Loss:2.304
                                     Accuracy: 0.12, 0.14, 0.11, 0.09
                                                                     Learning rate: 0.00100
                                     Accuracy: 0.11, 0.18, 0.07, 0.12
Iter:240/1359 epoch:6,
                        Loss:2.303
                                                                     Learning rate: 0.00100
Iter:260/1359 epoch:6,
                                     Accuracy: 0.09, 0.09, 0.09, 0.10
                                                                     Learning rate:0.00100
                        Loss:2.306
                                     Accuracy: 0.08, 0.11, 0.09, 0.11
                                                                    Learning rate:0.00100
Iter:280/1359 epoch:7,
                        Loss:2.307
                        Loss:2.303
                                     Accuracy: 0.06, 0.12, 0.12, 0.12
                                                                    Learning rate:0.00100
Iter:300/1359 epoch:7,
Iter:320/1359 epoch:8,
                        Loss:2.305
                                     Accuracy: 0.11, 0.05, 0.16, 0.09
                                                                     Learning rate:0.00050
                         Loss:2.296
                                     Accuracy: 0.06, 0.12, 0.16, 0.09
                                                                     Learning rate: 0.00050
Iter:340/1359 epoch:8,
                                     Accuracy: 0.14, 0.10, 0.16, 0.16
                                                                     Learning rate:0.00050
Iter:360/1359 epoch:8,
                         Loss:2.248
                                                                     Learning rate:0.00050
Iter:380/1359 epoch:9,
                         Loss:2.181
                                     Accuracy:0.18,0.17,0.16,0.20
Iter:400/1359 epoch:9,
                         Loss:2.080
                                      Accuracy:0.14,0.28,0.21,0.13
                                                                     Learning rate:0.00050
Iter:420/1359 epoch:10,
                         Loss:2.095
                                      Accuracy: 0.21, 0.23, 0.23, 0.19
                                                                      Learning rate: 0.00050
Iter:440/1359 epoch:10,
                          Loss:1.957
                                      Accuracy: 0.23, 0.23, 0.22, 0.23
                                                                      Learning rate: 0.00050
Iter:460/1359 epoch:11,
                          Loss:1.834
                                                                      Learning rate:0.00050
                                      Accuracy: 0.32, 0.37, 0.29, 0.25
Iter:480/1359 epoch:11,
                                      Accuracy: 0.38, 0.25, 0.17, 0.28
                                                                      Learning rate:0.00050
                          Loss:1.814
Iter:500/1359 epoch:12,
                          Loss:1.588
                                      Accuracy: 0.43, 0.41, 0.29, 0.46
                                                                      Learning rate: 0.00050
```

#### 驗證模型

#### ■ 7\_3 驗證模型.ipynb

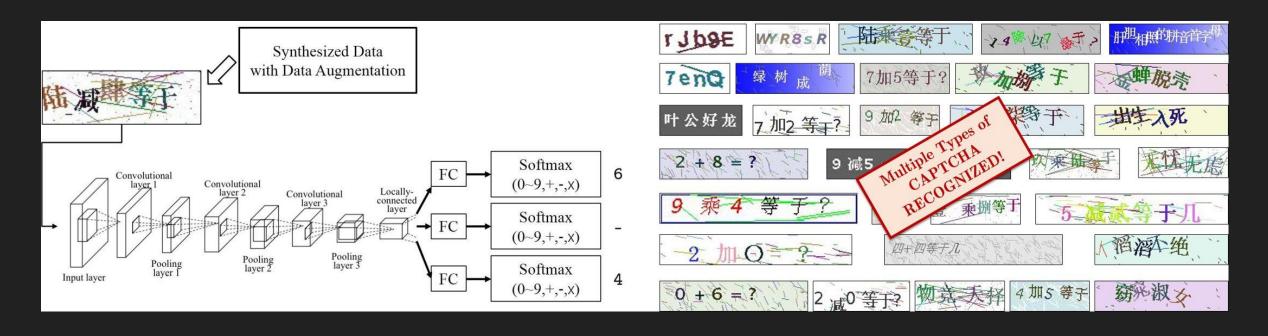


#### Reference

- JasonLiTW, simple-railway-captcha-solver, 基於 CNN 的台鐵訂票驗證碼辨識以及驗證性高的訓練集產生器, GitHub [Link]
- Parker-Lyu,TensorFLow-Learning,B站上煉數成金的公開課筆記,GitHub [Link]
- Jonathan Godwin Multi-Task Learning in Tensorflow [Link]

# Good Project about CAPTCHA

- CAPTCHA Cracking: A CNN Based OCR Module for Web Crawler
- http://wangchuan.github.io/archive/projects/captcha-crack/



# -END-

■ 安裝...

> pip install captcha