CNN Final Project

第一組

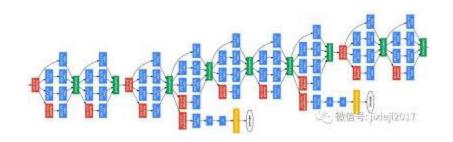
組員:張聚洋、莊宗縉、許智堯

大綱

- 動機
- 4 Steps
 - EDA
 - Data Preprocessing
 - Model Establishment
 - Model Evaluation
- 遇到的問題
- 結論

動機

- 首先是,為什麼我們選擇這個主題?
- 我們試著理解手機上的手寫辨識背後的原理。想 先以一個經典的題目來測試一般卷積神經網路的 精確度跟強度
- 我們預計先使用一般卷積層作訓練,若無法精確 辨識再使用GoogleNet下去做,最後的備案是 ResNet。





4 Steps



Step1 - EDA

- 1.Data Visualization
- 2.看資料分布



Step2 - Preprocessing

- 1.OneHot-Encoding
- 2.增加維度
- 3.Normalization



Step3 - Model

- 1.Convolutional Layer
- 2.Max Pooling
- 3. Fully Connected Layer



Step4 - Evaluation

- 1.Confusion Matrix
- 2. Visualization

EDA

• Data visualization:看資料分布的平不平均,好不好。若資料不好,可能需要做刪除空值或augmentation。我們是利用直方圖去看,發現我們的資料分布很好,很平均,漂亮!

```
plt.hist(y_Train)

(array([5923., 6742., 5958., 6131., 5842., 5421., 5918., 6265., 5851., 5949.]),
  array([0., 0.9, 1.8, 2.7, 3.6, 4.5, 5.4, 6.3, 7.2, 8.1, 9.]),
  <a list of 10 Patch objects>)

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Data Preprocessing

- x_train:60000筆、x_test:10000筆;再者 我們Validation的split比例為0.2
 - 我們先將 y label 使用 One hot Encoding 用成維 度型使之不要有類別關係而導致很難訓練
 - 我們將資料增加一個維度,使我們可以做convnet。
 - 爾後為了讓資料訓練速度更快,我們決定讓數據 收斂,使用正規化(將其標準化),像我們下面這 張圖

Data Preprocessing

```
# 多加一個顏色的維度
x_Train4D=x_Train.reshape(x_Train.shape[0],28,28,1).astype('float32')
x_Test4D=x_Test.reshape(x_Test.shape[0],28,28,1).astype('float32')
```

```
print('x_train_image:',x_Train.shape)
print('y_train_label:',y_Train.shape)

x_train_image: (60000, 28, 28)
y_train_label: (60000,)

print('x_test_image:',x_Test.shape)
print('y_test_label:',y_Test.shape)

x_test_image: (10000, 28, 28)
y_test_label: (10000,)
```

```
# 將數值縮小到0~1 #標準化
x_Train4D_normalize = x_Train4D / <mark>255</mark>
x_Test4D_normalize = x_Test4D / <mark>255</mark>
```

Model Establishment

Non-trainable params: 0

Model: "sequential"			
Layer (type)	Output	Shape	Param #
conv2d (Conv2D)	(None,	28, 28, 16)	416
max_pooling2d (MaxPooling2D)	(None,	14, 14, 16)	0
conv2d_1 (Conv2D)	(None,	14, 14, 36)	14436
max_pooling2d_1 (MaxPooling2	(None,	7, 7, 36)	0
dropout (Dropout)	(None,	7, 7, 36)	0
flatten (Flatten)	(None,	1764)	0
dense (Dense)	(None,	128)	225920
dropout_1 (Dropout)	(None,	128)	0
dense_1 (Dense)	(None,	10)	1290

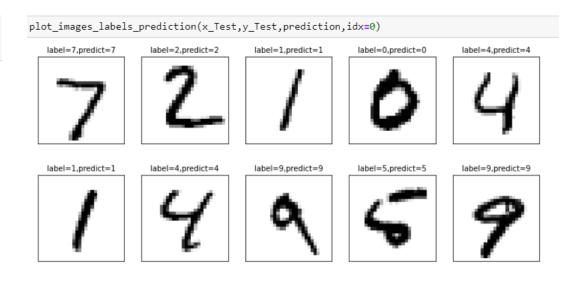
Model Establishment-Loss Function

- 我們使用的是Categorical crossentropy,並將 optimizer 設成 Adam
- Batch Size = 300, Epoch = 20, Iterations = 160

Model Evaluation

- 觀察Accuracy
- Confusion Matrix 查看各類別資料的辨識情況
 - X Precision
 - X Recall
 - X F₁ score

.mport od.cros		(y_Te				,coln	ames	=['pr	edict	t'])
predict label	0	1	2	3	4	5	6	7	8	9
0	978	1	0	0	0	0	0	1	0	0
1	0	1135	0	0	0	0	0	0	0	0
2	1	1	1028	0	0	0	0	2	0	0
3	0	0	0	1006	0	2	0	0	2	0
4	0	0	0	0	979	0	0	0	1	2
5	1	0	0	6	0	884	1	0	0	0
6	2	2	1	0	1	2	950	0	0	0
7	0	3	1	0	0	0	0	1022	1	1
8	1	1	1	1	0	1	0	1	967	1
9	0	1	0	0	5	3	0	0	1	999

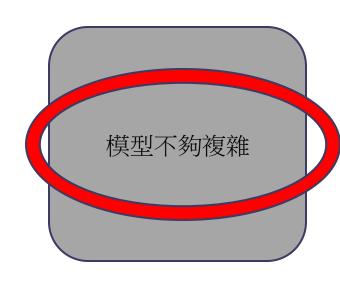


訓練過程所遭遇的問題

- Training Model Accuracy低:
 - 一開始訓練結果與網路上他人結果相比還略顯低, 因此開始推斷以下原因 =>

Feature Collision

資料分布有問題

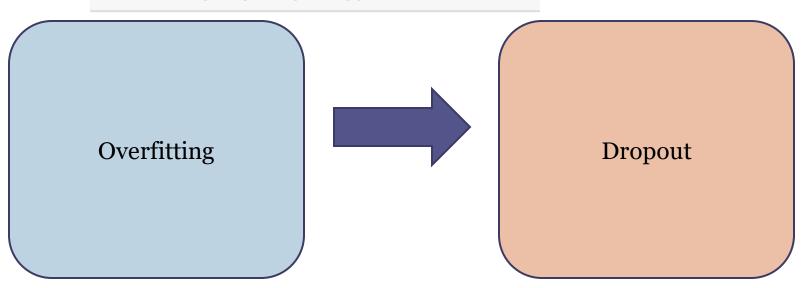


訓練過程所遭遇的問題

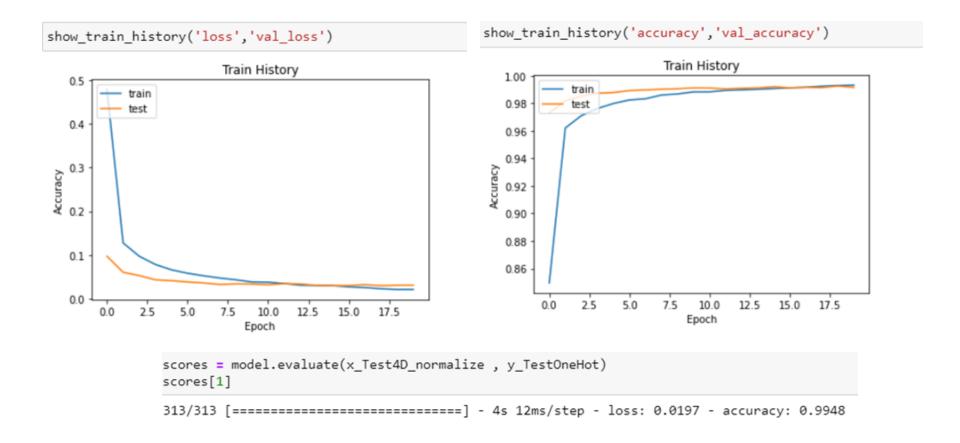
• Test Model Accuracy小低於Training Model Accuracy:

將訓練模型的準確度有效提升後,又發現 Test 出來的準確度居然相較訓練的還來的低 =>

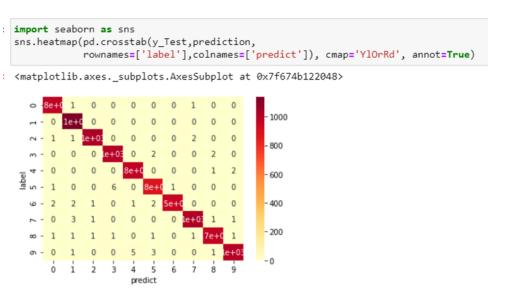
Drop掉部分神經元避免overfitting model.add(Dropout(0.25))

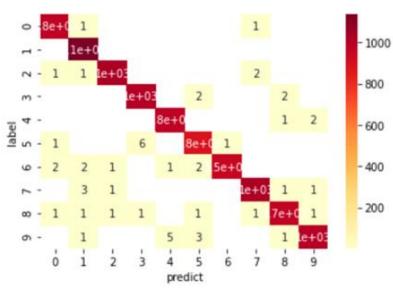


訓練成果



Model Evaluation結果



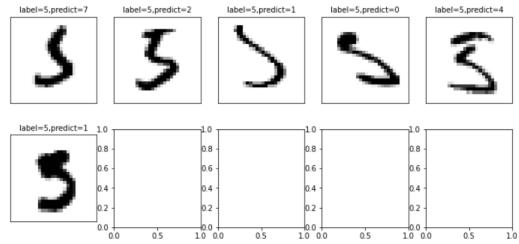


我們發現對角線(辨識正確數)非常高,因此判定訓練結果佳,而且只有零星錯誤

辨識錯誤癥結點

df[(df.label==5)&(df.predict==3)]

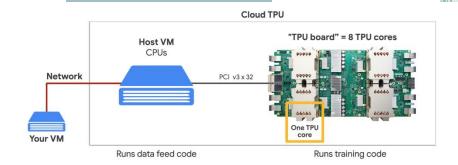
	label	predict
340	5	3
674	5	3
1393	5	3
1737	5	3
2597	5	3
5937	5	3



In []: plot_images_labels_prediction([x_Test[i] for i in df[(df.label==5)&(df.predict==3)].index],[y_Test[i] for i in df[(df.label==5)&(df.label==5)&(df.label==5)&(df.label==5)&(df.label==5)&(df.label==5)&(df.label==5)&(df.label==5)&(df.label==5)&(df.label==5)&(df.label==5)&(df.label==5)&(df.label==5)&(df.label==5)&(df.label==5)&(df.label==5)&(df.label==5)&(df.label==5)&(df.label==5)&(df.label==5

結論

- Datasets Source : Keras datasets
- 執行環境:Google Colab TPU
 1.Cloud TPU v3 Pod
 2.每秒超過 100 千兆次浮點運算
 - 3.32 TB HBM
 - 4.2-D 環面網狀網路
- 使用Module:
- 1. Matplotlib
- 2. Keras
- 3. Numpy
- 4. Pandas
- 5. tensorflow



結論

- 我們在各步驟裡所使用的技術
 - Input X:Image;Output Y:Label(數字)
 - Datasets: MNIST Handwritten Digit Classification Dataset
 - Data Visualization
 - Data Preprocessing(OneHot-Encoding \
 Normalization)

結論

- Convolutional Layer(Activation Function: Relu)
- Max Pooling
- Fully Connected (Activation Function : Softmax)
- Loss Function(Categorical cross entropy)
- Confusion Matrix