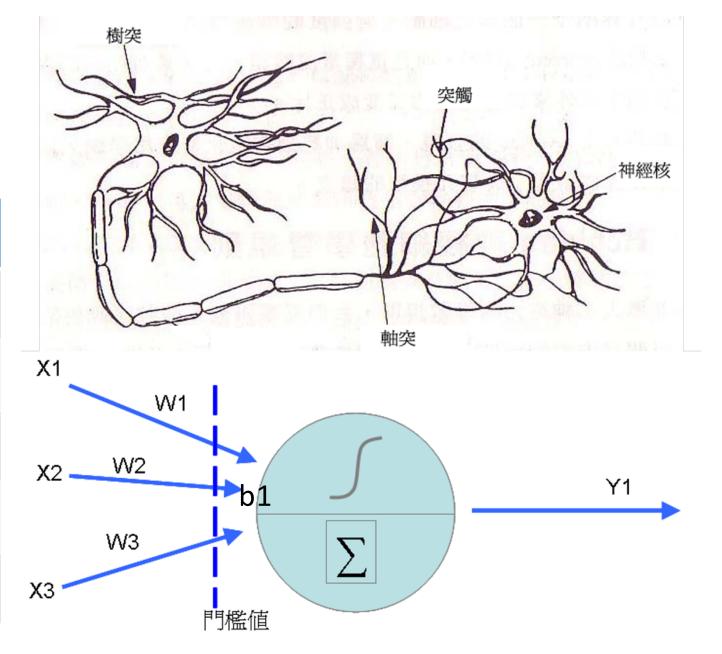


深度學習簡介

項目	說明
Input X	模仿輸入神經元
Output Y	模仿接收神經元
Weight W	模仿軸突
Bias b	模仿突觸
Activation Function	模仿神經傳導的運作 方式

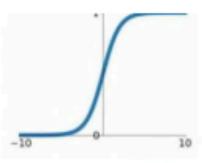


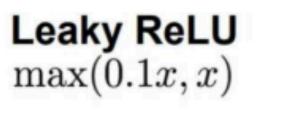
Y1=sigmoid(X1*W1+X2*W2+X3*W3+b1)

激活函數(Activation Function)

Sigmoid

$$\sigma(x) = \frac{1}{1 + e^{-x}}$$

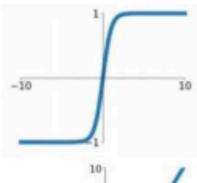






tanh

tanh(x)

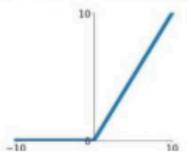


Maxout

$$\max(w_1^T x + b_1, w_2^T x + b_2)$$

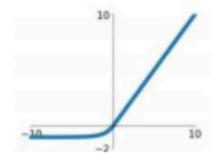
ReLU

 $\max(0,x)$

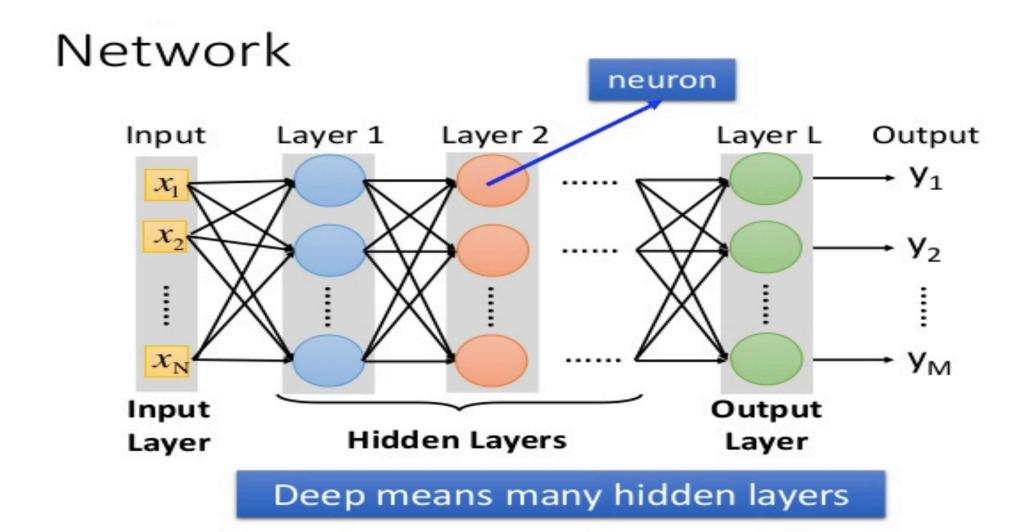


ELU

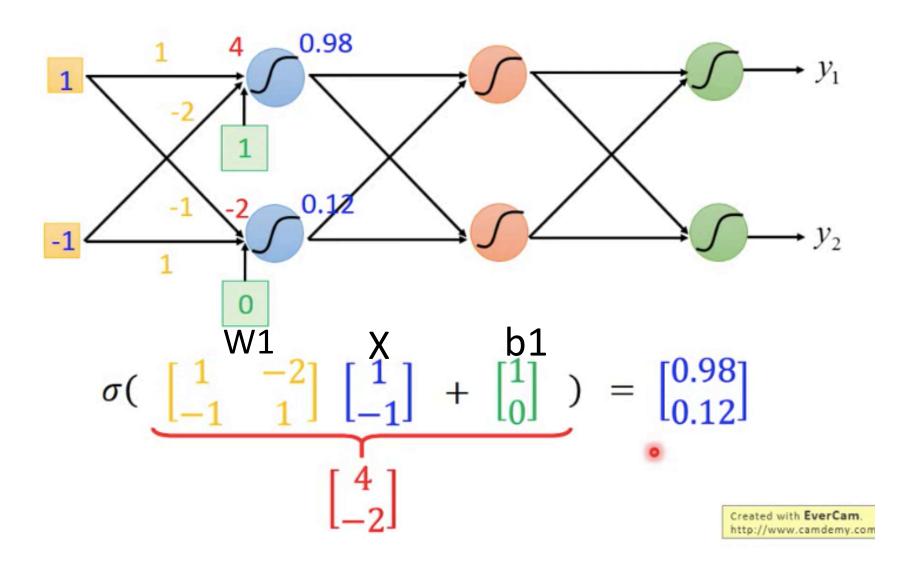
$$\begin{cases} x & x \ge 0 \\ \alpha(e^x - 1) & x < 0 \end{cases}$$



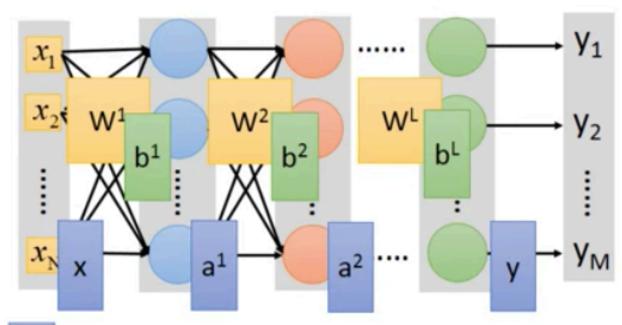
多層感知器模型



Matrix Operation

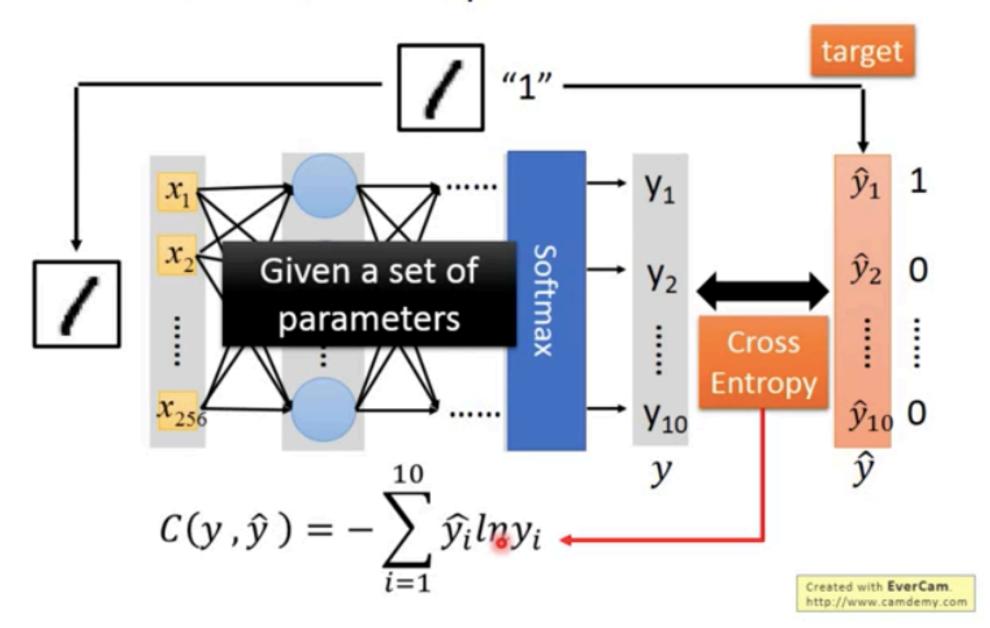


Neural Network



$$y = f(x)$$
Using parallel computing techniques to speed up matrix operation

Loss for an Example



Softmax函數

$$\sigma(x_j) = \frac{e^{x_j}}{\sum_i e^{x_i}}$$

```
exp(3.0) = 20.0855

exp(5.0) = 148.4132

exp(2.0) = 7.3891

sum = 175.8878
```

And the softmax values are:

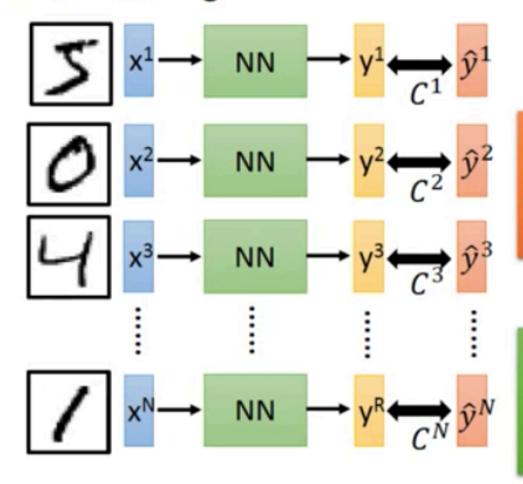
```
s(3.0) = 20.0855 / 175.8878 = 0.12

s(5.0) = 148.4132 / 175.8878 = 0.84

s(2.0) = 7.3891 / 175.8878 = 0.04
```

Total Loss

For all training data ...



Total Loss:

$$\overset{\bullet}{L} = \sum_{n=1}^{N} C^n$$



Find *a function in function set* that
minimizes total loss L



Find the network

parameters θ* that

minimize total loss L

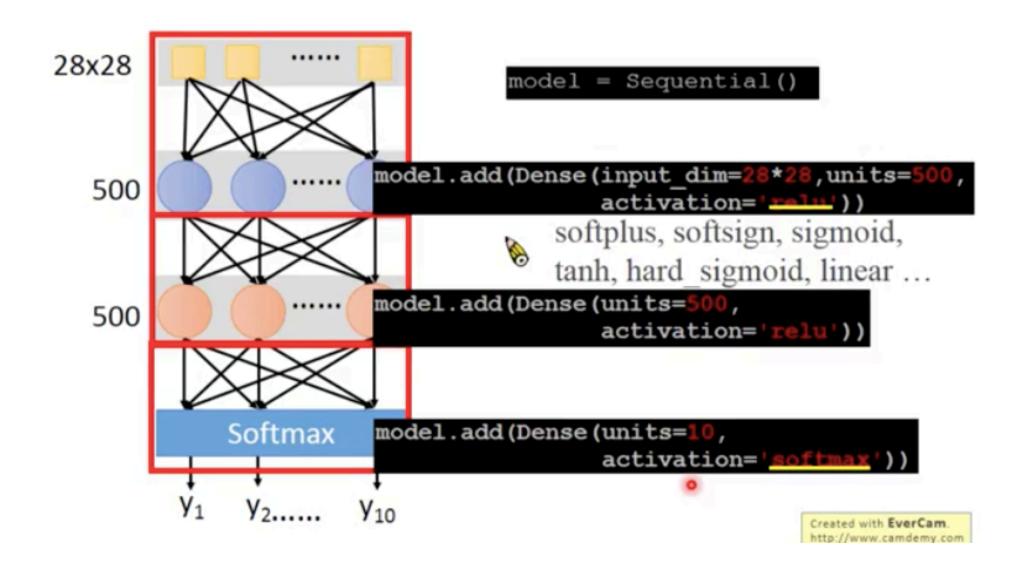
http://www.camdemy.co

Gradient 如何進行 Descent Optimizer Loss 誤差 更新W和b **Function** Back Propagation Network neuron Output Layer 1 Layer 2 Layer L Input 取出 模型 真實 Mnist 預處理 輸出 資料集 數字 feature y_M Output Input **Hidden Layers** Layer Layer Deep means many hidden layers 更新參數Reference https://www.cnblogs.com/charlotte77/p/5629865.html

Let's Demo



Keras: Building a Network

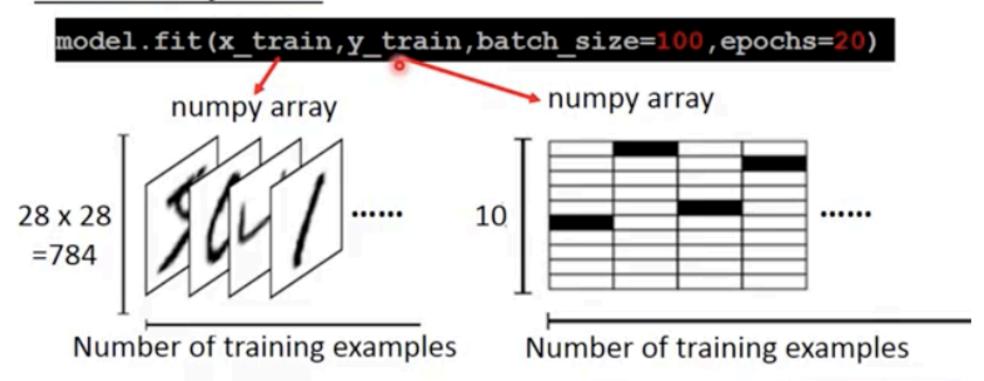


Configuration

Several alternatives: https://keras.io/objectives/

SGD, RMSprop, Adagrad, Adadelta, Adam, Adamax, Nadam

Pick the best function



Keras

Save and load models

http://keras.io/getting-started/faq/#how-can-i-save-a-keras-model

How to use the neural network (testing): •

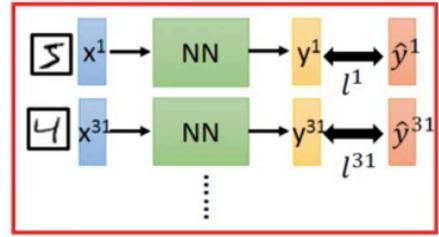
```
case 1: print('Total loss on Testing Set:', score[0])
print('Accuracy of Testing Set:', score[1])
```

```
case 2: result = model.predict(x_test)
```

We do not really minimize total loss!

Mini-batch

Mini-batch



 $\frac{\text{Mini-batch}}{\text{Min}} = \frac{\text{Mini-batch}}{\text{Mini-batch}} = \frac{\text{Mini-ba$

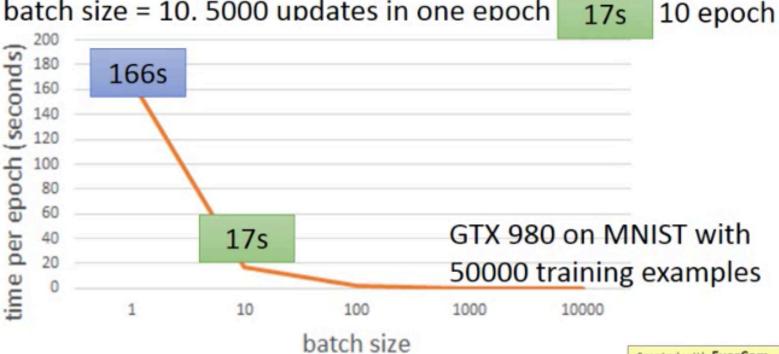
- Randomly initialize network parameters
- Pick the 1st batch $L' = l^1 + l^{31} + \cdots$ Update parameters once
- Pick the 2^{nd} batch $L'' = l^2 + l^{16} + \cdots$ Update parameters once :
- Until all mini-batches have been picked

one epoch

Repeat the above process

Speed

- Smaller batch size means more updates in one epoch
 - E.g. 50000 examples
 - 166s 1 epoch batch size = 1, 50000 updates in one epoch
 - batch size = 10. 5000 updates in one epoch



http://www.camdemy.com