

深度學習入門

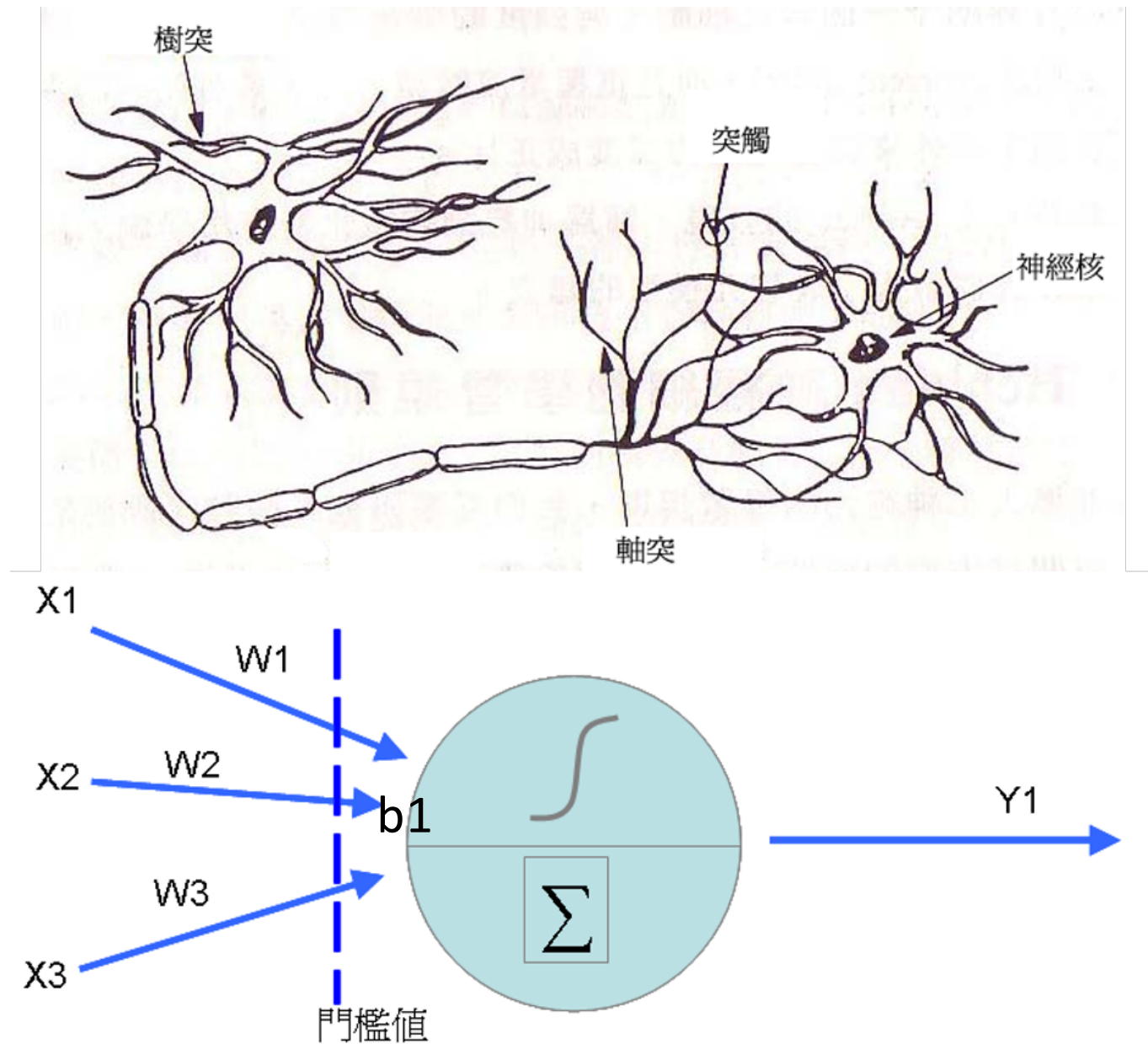
Keras

DE Sharing

何信賢

深度學習簡介

項目	說明
Input X	模仿輸入神經元
Output Y	模仿輸出神經元
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

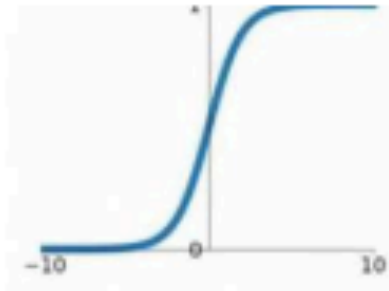


$$Y1 = \text{sigmoid}(X1 * W1 + X2 * W2 + X3 * W3 + b1)$$

激活函數(Activation Function)

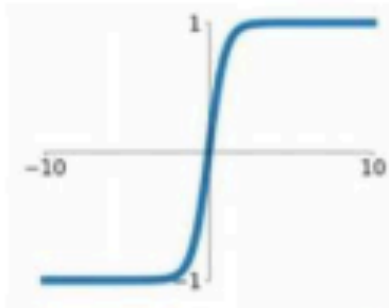
Sigmoid

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



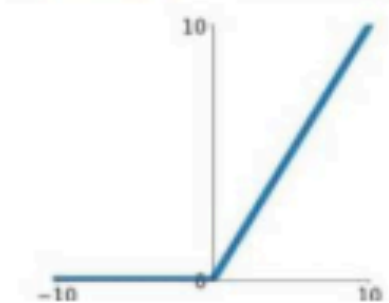
tanh

$$\tanh(x)$$



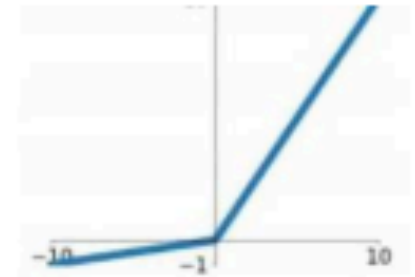
ReLU

$$\max(0, x)$$



Leaky ReLU

$$\max(0.1x, x)$$

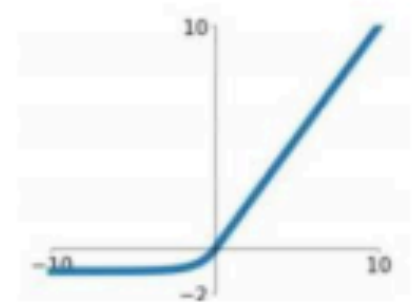


Maxout

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

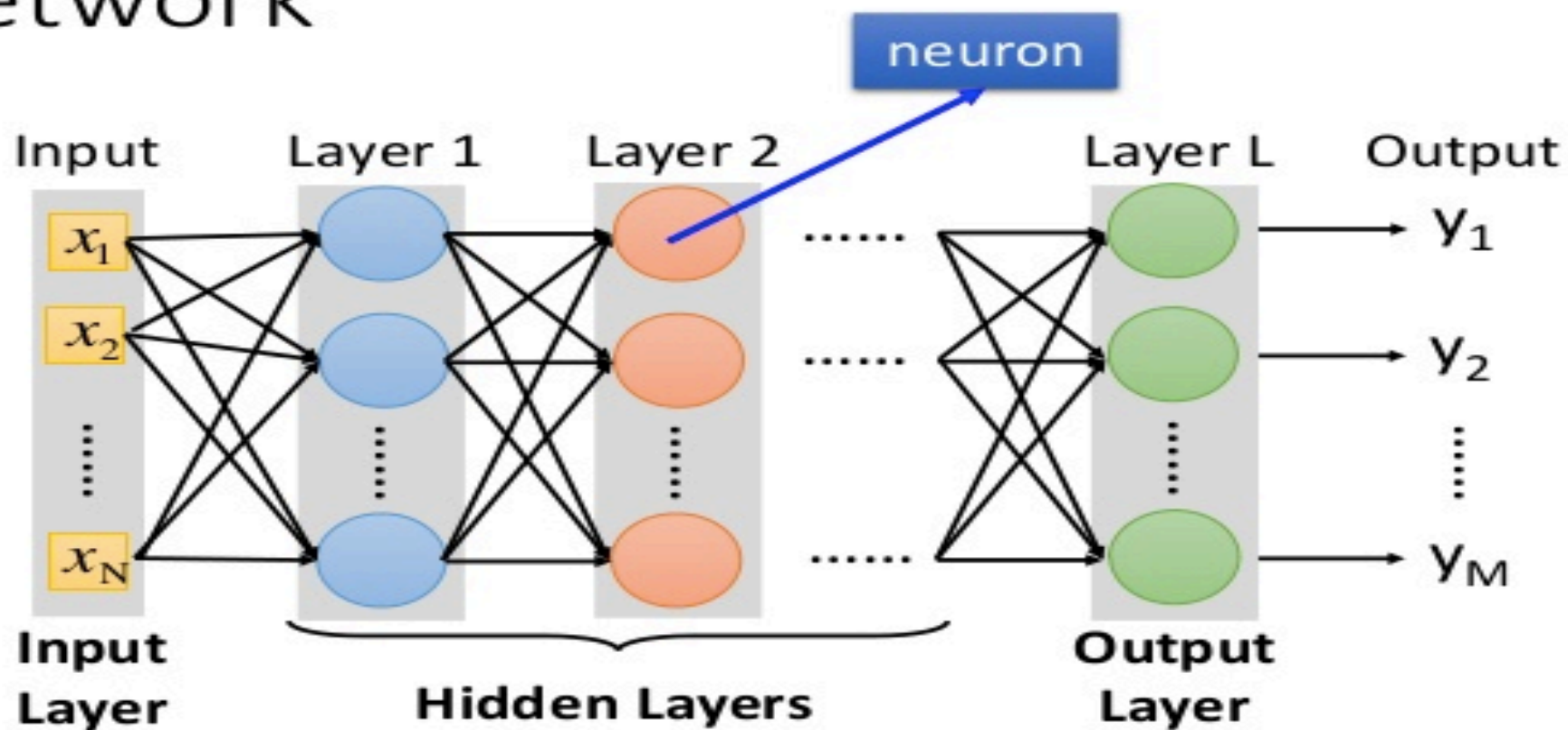
ELU

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



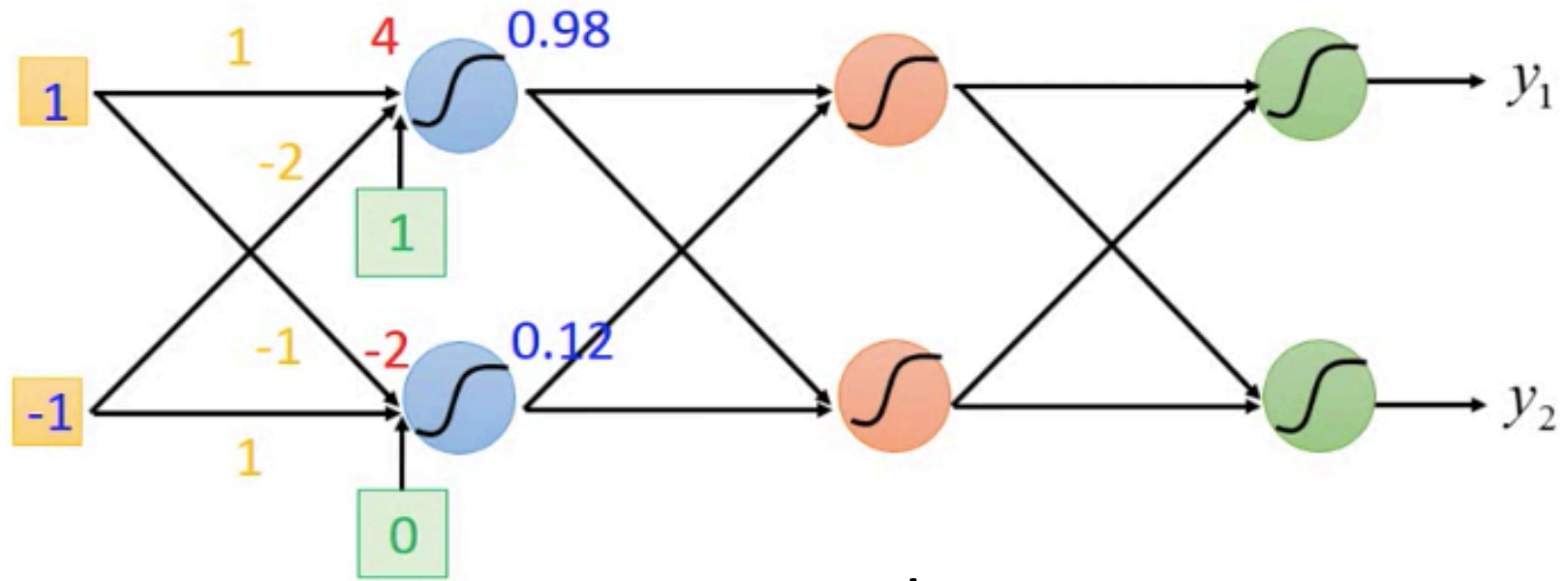
多層感知器模型

Network



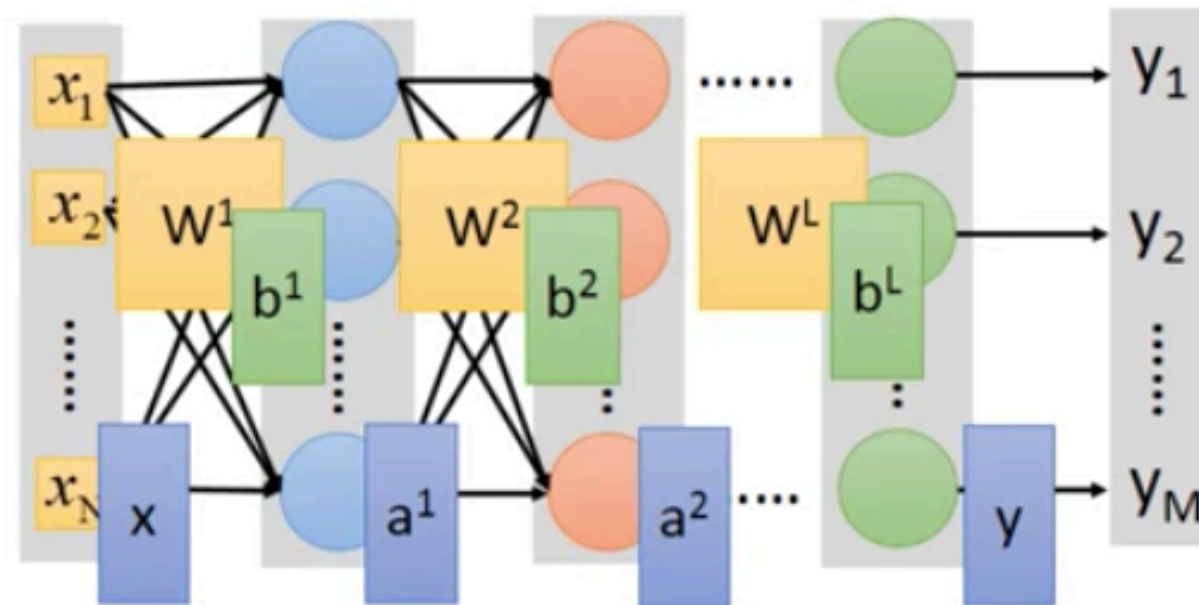
Deep means many hidden layers

Matrix Operation



$$\sigma\left(\underbrace{\begin{bmatrix} 1 & -2 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \end{bmatrix}}_{\begin{bmatrix} 4 \\ -2 \end{bmatrix}} \right) = \begin{bmatrix} 0.98 \\ 0.12 \end{bmatrix}$$

Neural Network

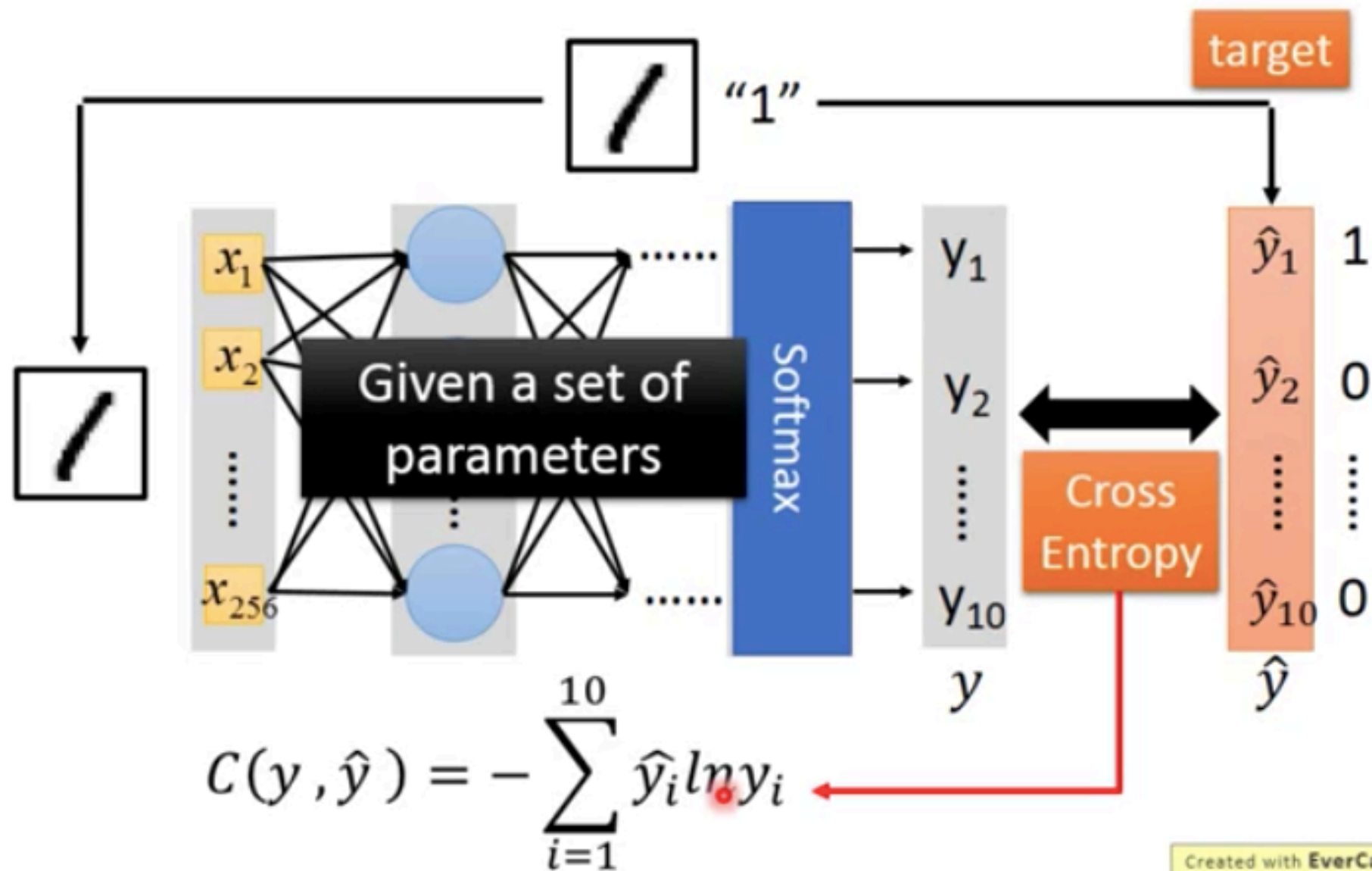


$$y = f(x)$$

Using parallel computing techniques
to speed up matrix operation

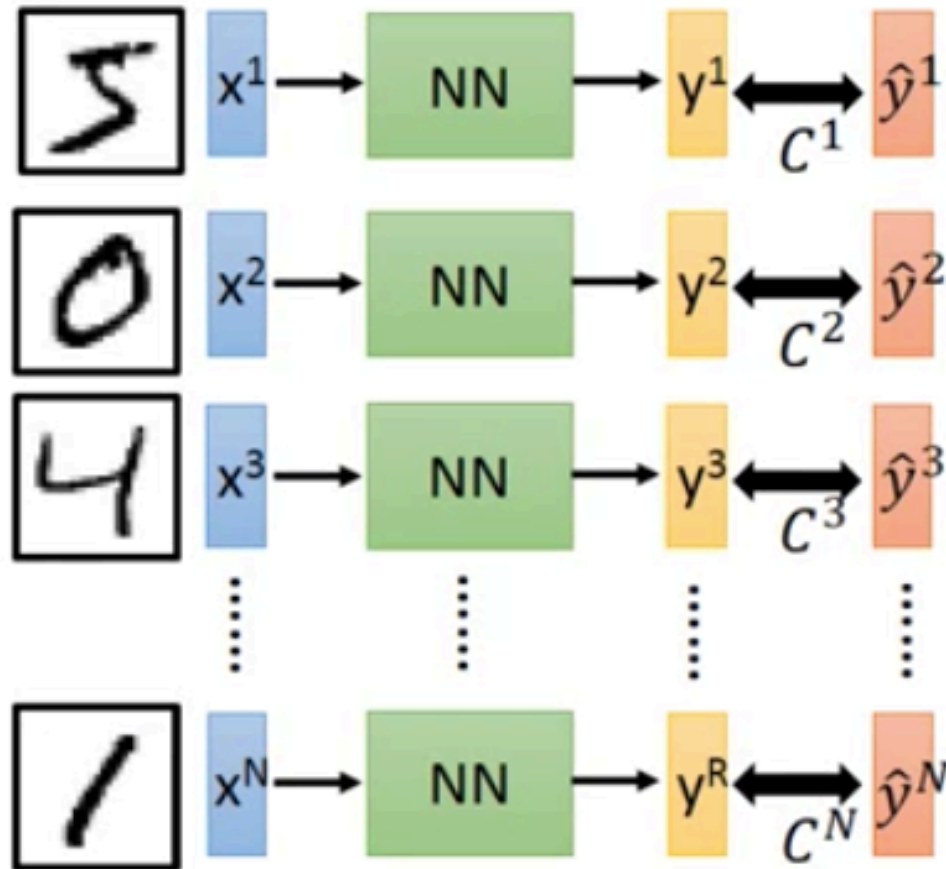
$$= \sigma(W^L \dots \sigma(W^2 \sigma(W^1 x + b^1) + b^2) \dots + b^L)$$

Loss for an Example



Total Loss

For all training data ...



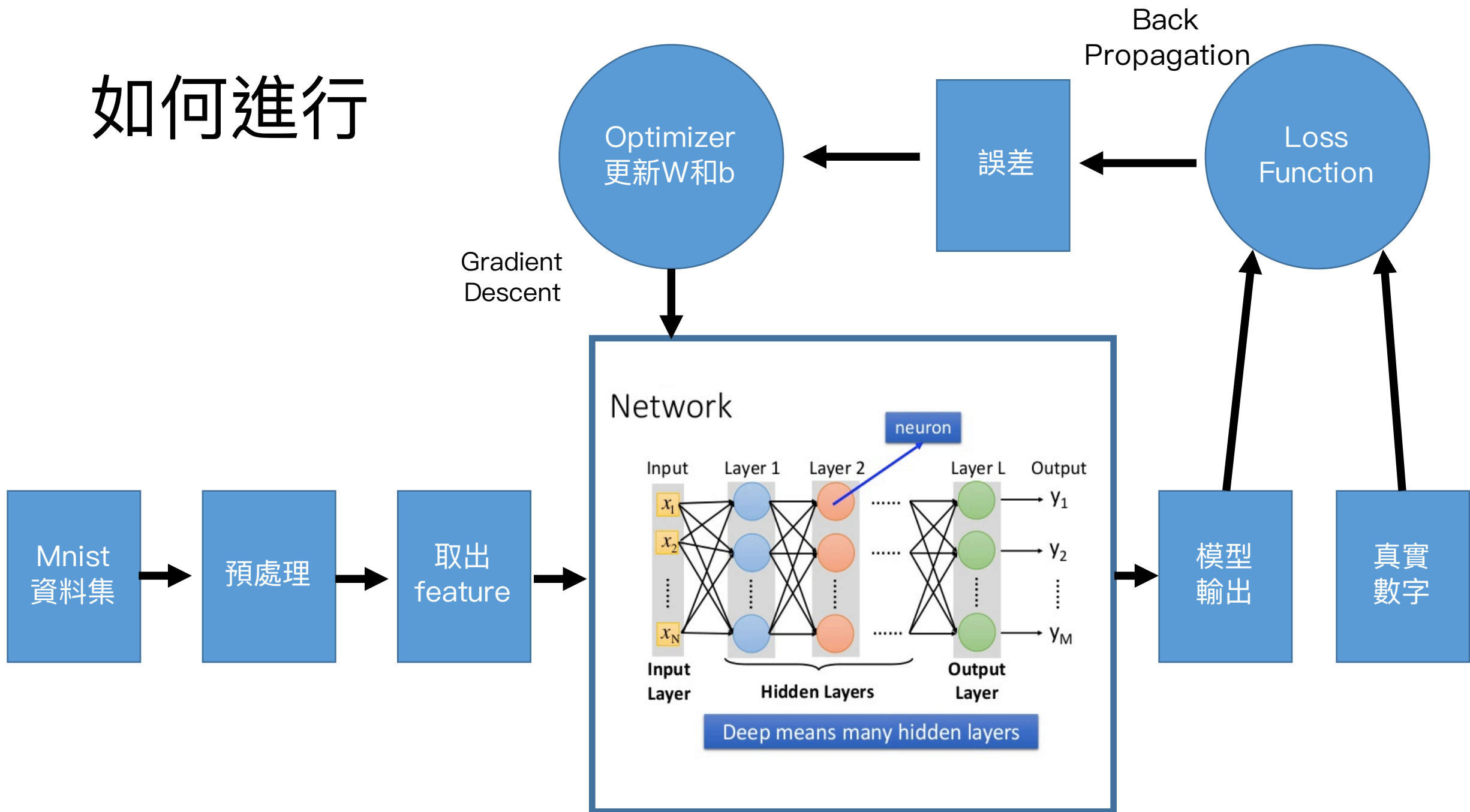
Total Loss:

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

Find a function in function set that minimizes total loss L

Find the network parameters θ^* that minimize total loss L

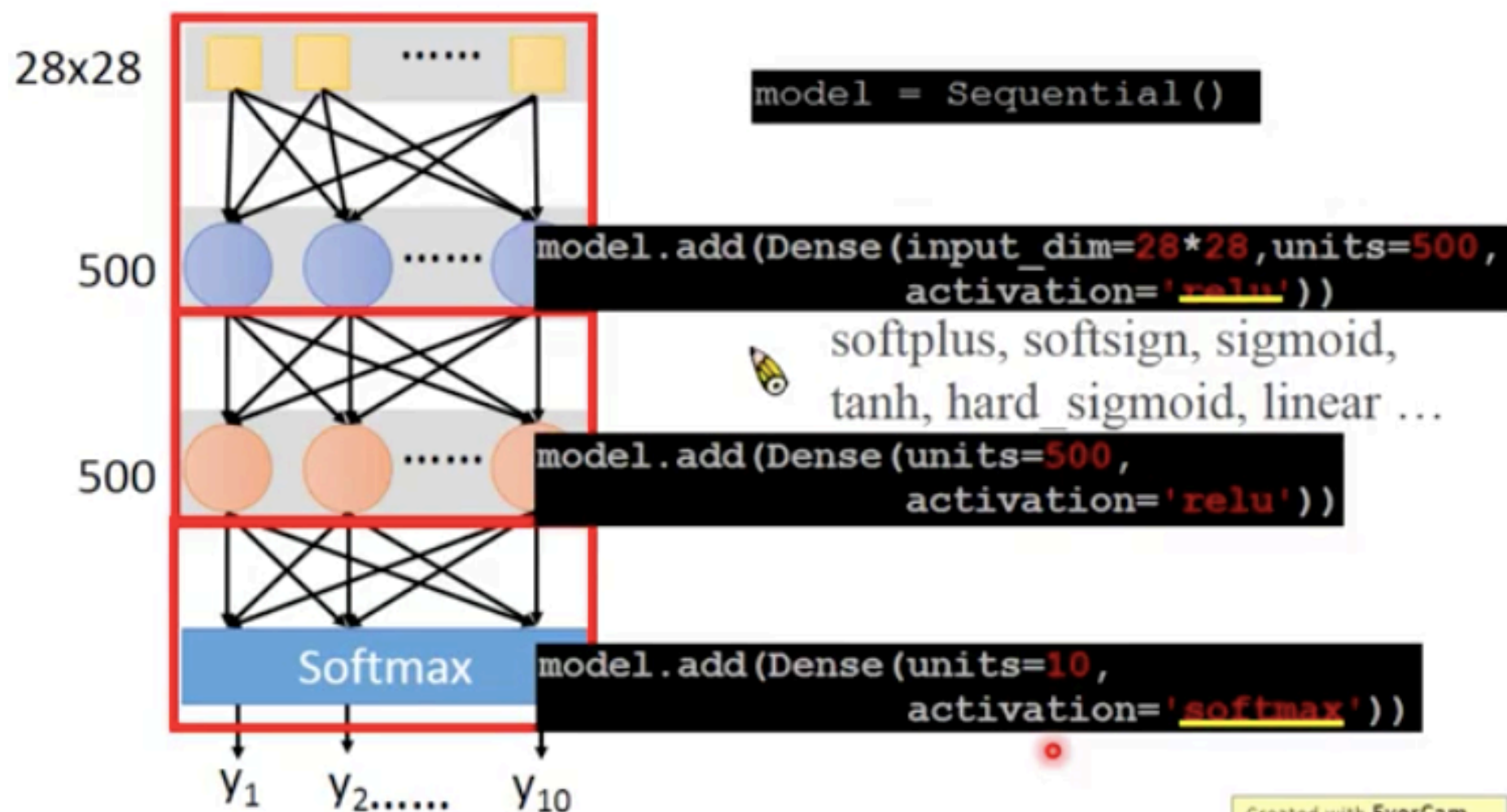
如何進行



Let's Demo



Keras: Building a Network



Configuration

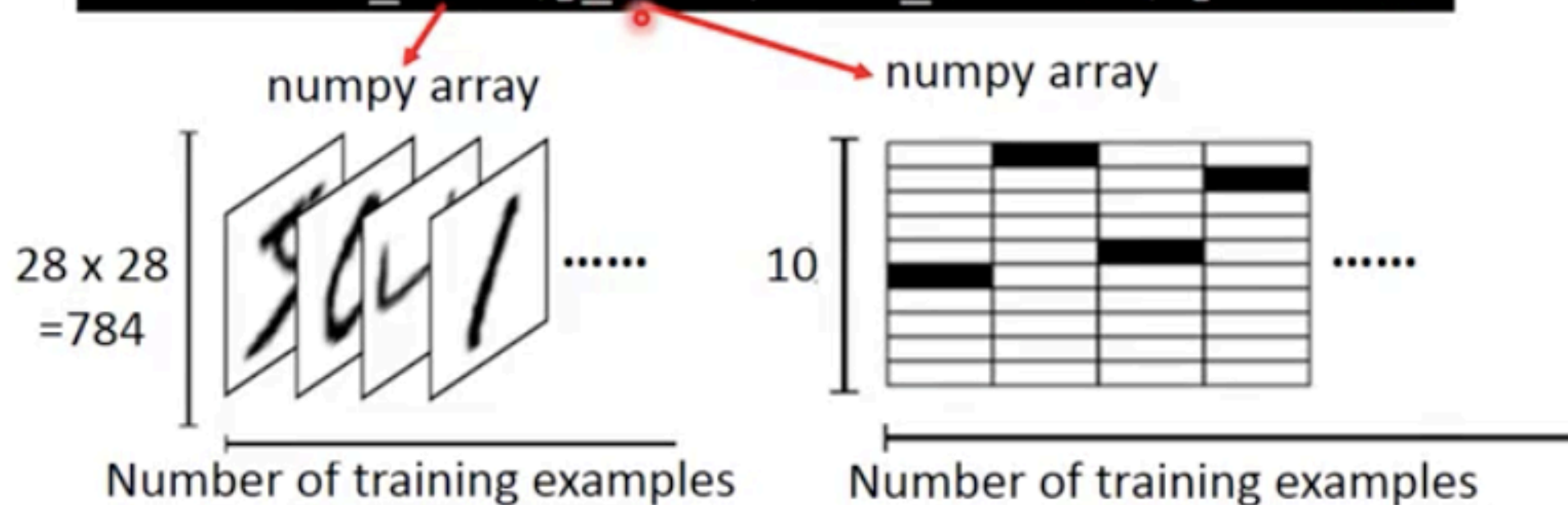
Several alternatives: <https://keras.io/objectives/>

```
model.compile(loss='categorical_crossentropy',  
              optimizer='adam',  
              metrics=['accuracy'])
```

SGD, RMSprop, Adagrad, Adadelata, Adam, Adamax, Nadam

Pick the best function

```
model.fit(x_train, y_train, batch_size=100, epochs=20)
```



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:

```
score = model.evaluate(x_test,y_test)
print('Total loss on Testing Set:', score[0])
print('Accuracy of Testing Set:', score[1])
```

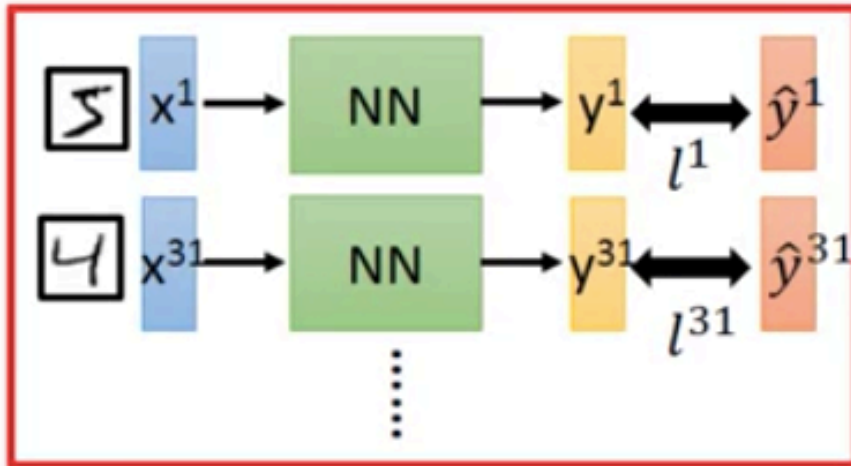
case 2:

```
result = model.predict(x_test)
```

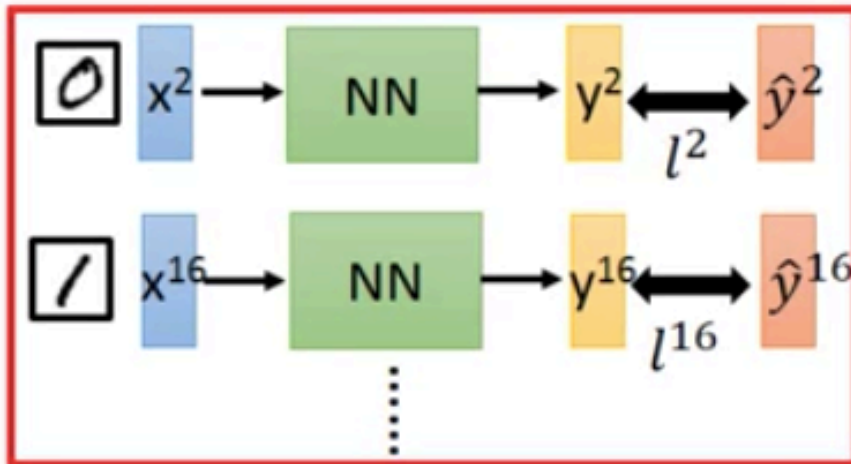
We do not really minimize total loss!

Mini-batch

Mini-batch



Mini-batch



- Randomly initialize network parameters

- Pick the 1st batch
 $L' = l^1 + l^{31} + \dots$
Update parameters once
- Pick the 2nd batch
 $L'' = l^2 + l^{16} + \dots$
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
 - batch size = 1, 50000 updates in one epoch 166s 1 epoch
 - batch size = 10. 5000 updates in one epoch 17s 10 epoch

