# 2024 鐵人賽 – 我數學就爛要怎麼來學 DNN 模型安全 Day 05 – DNN 模型基本概念

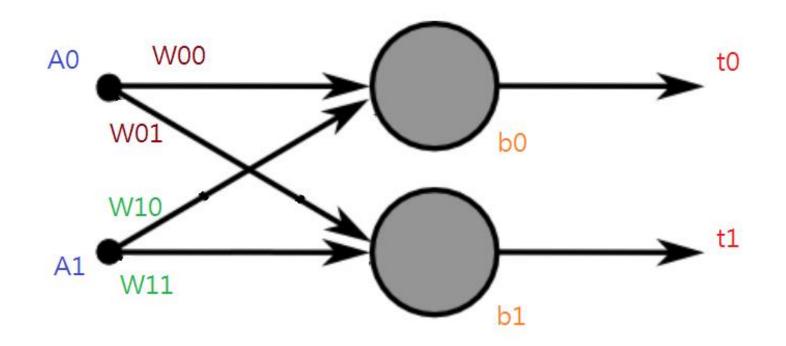
## 大綱

- ■類神經網路
  - 激勵函數
  - 損失函數
  - 最佳化計算
- ■結論



# 回憶一下之前的鸚鵡模型

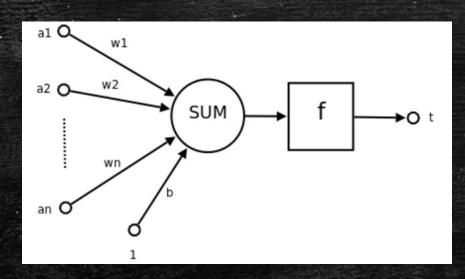
$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \times \begin{bmatrix} A_0 \\ A_1 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \end{bmatrix} = \begin{bmatrix} A_0 \\ A_1 \end{bmatrix}$$



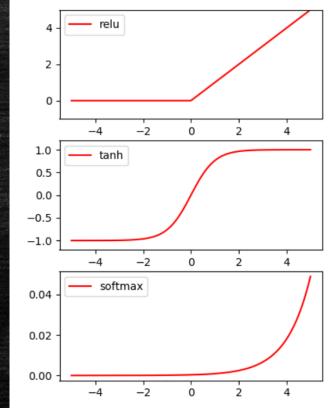
接著要讓機器用數據去學習 w,b 這兩種參數

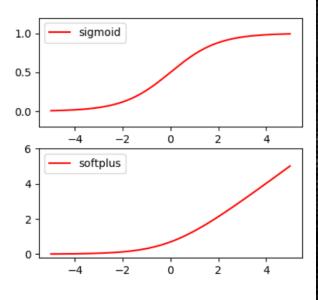
### 激勵函數

線性的東西組合出來還是線性,所以需要一些函數產生 非線性的結果



https://joe1in.github.io/dev\_no tes/ml/tensorflow/syntax.html #activation





#### 損失函數

https://chih-sheng-huang821.medium.com/%E6%A9%9F%E5%99%A8-%E6%B7%B1%E5%BA%A6%E5%AD%B8%E7%BF%92-%E5%9F%BA%E7%A4%8E%E4%BB%8B%E7%B4%B9-%E6%90%8D%E5%A4%B1%E5%87%BD%E6%95%B8-loss-function-2dcac5ebb6cb

■ 定義預測出來的結果 和實際數值差距的函 式

Loss: 
$$L = \frac{1}{N} \sum_{n} e_n$$

$$e = |y - \hat{y}|$$
 L is mean absolute error (MAE)  
 $e = (y - \hat{y})^2$  L is mean square error (MSE)

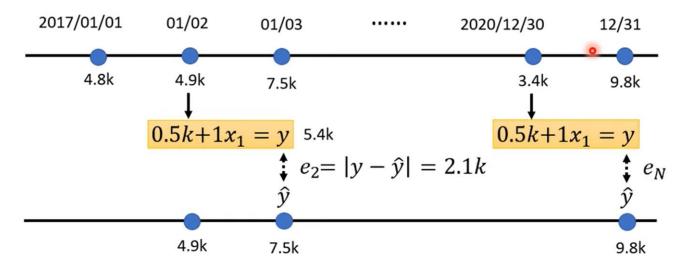
If y and  $\hat{y}$  are both probability distributions  $\longrightarrow$  Cross-entropy

# 2. Define Loss from Training Data > Loss: how good a set of

- Loss is a function of parameters L(b, w)
- values is.

$$L(0.5k, 1)$$
  $y = b + wx_1 \longrightarrow y = 0.5k + 1x_1$  How good it is?

Data from 2017/01/01 - 2020/12/31

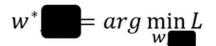


#### 最佳化計算

■ 對損失函數針對模型參數做偏微分求出梯度,在依照梯度的反方向做調整

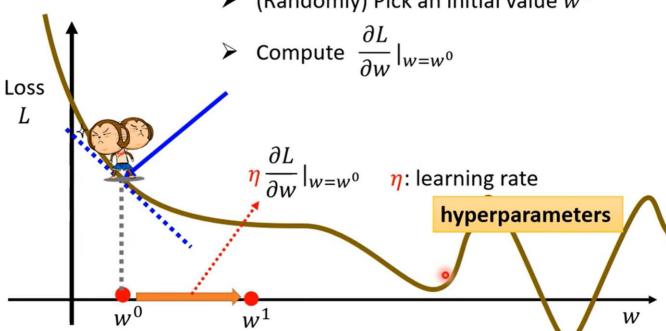
Source of image: http://chico386.pixnet.net/album/photo/171572850

#### 3. Optimization



#### **Gradient Descent**

 $\triangleright$  (Randomly) Pick an initial value  $w^0$ 



#### 來用類神經網路算數學

■ 來寫個程式預設 y = 2x + 1 看看

```
In []: # 建立屬於自己的 model
       model = Sequential()
       model.add(Dense(1, input_dim=1, activation='linear'))
In []: # 設定一些 function 後進行 compile
       model.compile(optimizer = tf.optimizers.Adam(),
                   loss='mean squared error')
       model.summary()
In []: #顯示模型的參數來看看
       old parameter = model.get weights()
       print(model.get_weights())
In []: # 開始傳入資料做訓練, epochs 代表要訓練幾回, batch size 代表一次要讀取的資料作訓練的數量
       model.fit(x, y, epochs=15, batch size=50)
In []: # 顯示訓練後模型的參數來看看
       print(old_parameter)
       print(model.get_weights())
       print(model.predict([10, 5, 200, 13]))
```

https://github.com/christianversloot/machinelearning-articles/blob/main/can-neuralnetworks-approximate-mathematical-functions.md

#### 結論

類神經網路 Deep Neural Network (深度神經網路) 算是最基本的應用類型,針對不同應用情境還有 CNN、RNN

