

Neural Networks Basics

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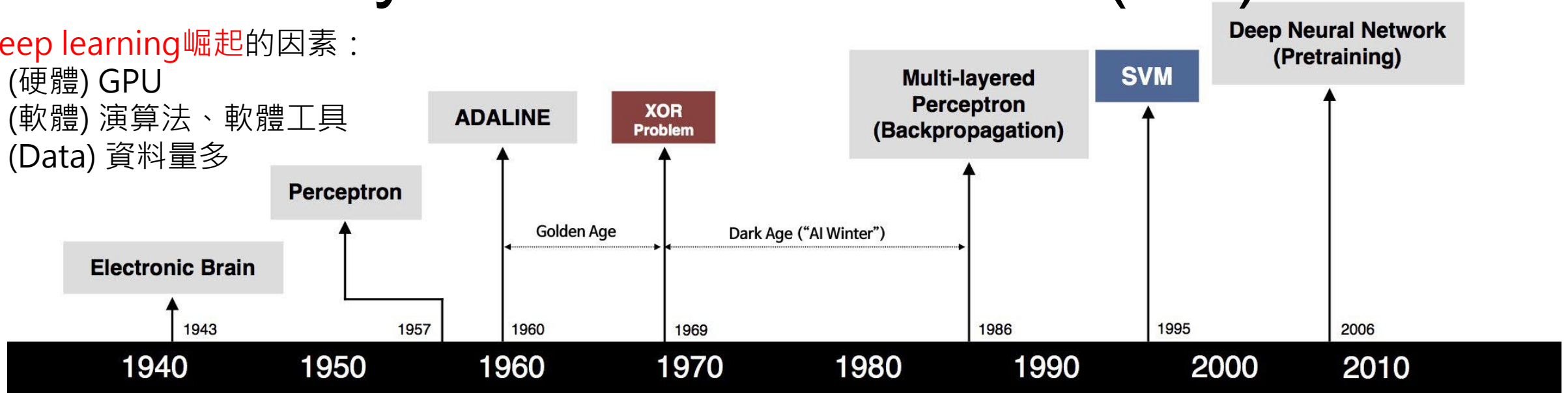
Outline

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2. Biological Inspiration
3. Artificial Neuron
4. Illustrative Example: Neural Representation
5. Activation Functions
6. Network Architectures
 - A Layer of Neurons
 - Multiple Layers of Neurons

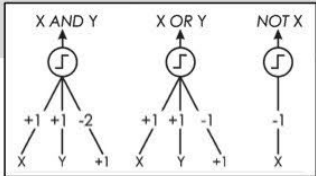
Brief History of Neural Networks (1/4)

Deep learning 崛起的因素：

1. (硬體) GPU
2. (軟體) 演算法、軟體工具
3. (Data) 資料量多



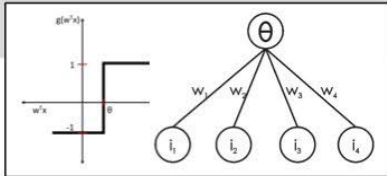
S. McCulloch – W. Pitts



- Adjustable Weights
- Weights are not Learned



F. Rosenblatt



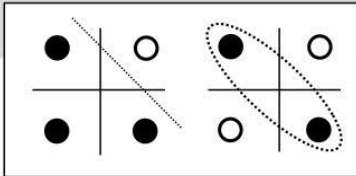
- Learnable Weights and Threshold



B. Widrow – M. Hoff



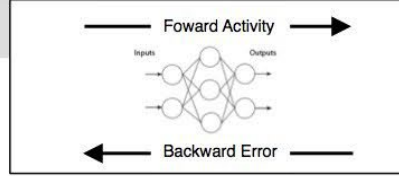
M. Minsky – S. Papert



- XOR Problem



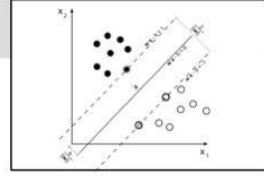
D. Rumelhart – G. Hinton – R. Williams



- Solution to nonlinearly separable problems
- Big computation, local optima and overfitting



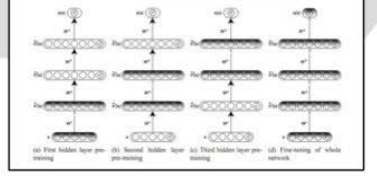
V. Vapnik – C. Cortes



- Limitations of learning prior knowledge
- Kernel function: Human Intervention



G. Hinton – S. Ruslan



- Hierarchical feature Learning

用人工自己去調
權重
不是學習而來的

Ex: 考兩科，其中一科一百分就有獎勵，考兩科更好，白色的代表沒有一百分

expression1	expression2	result
1	0	1
0	1	1
0	0	0
1	1	1

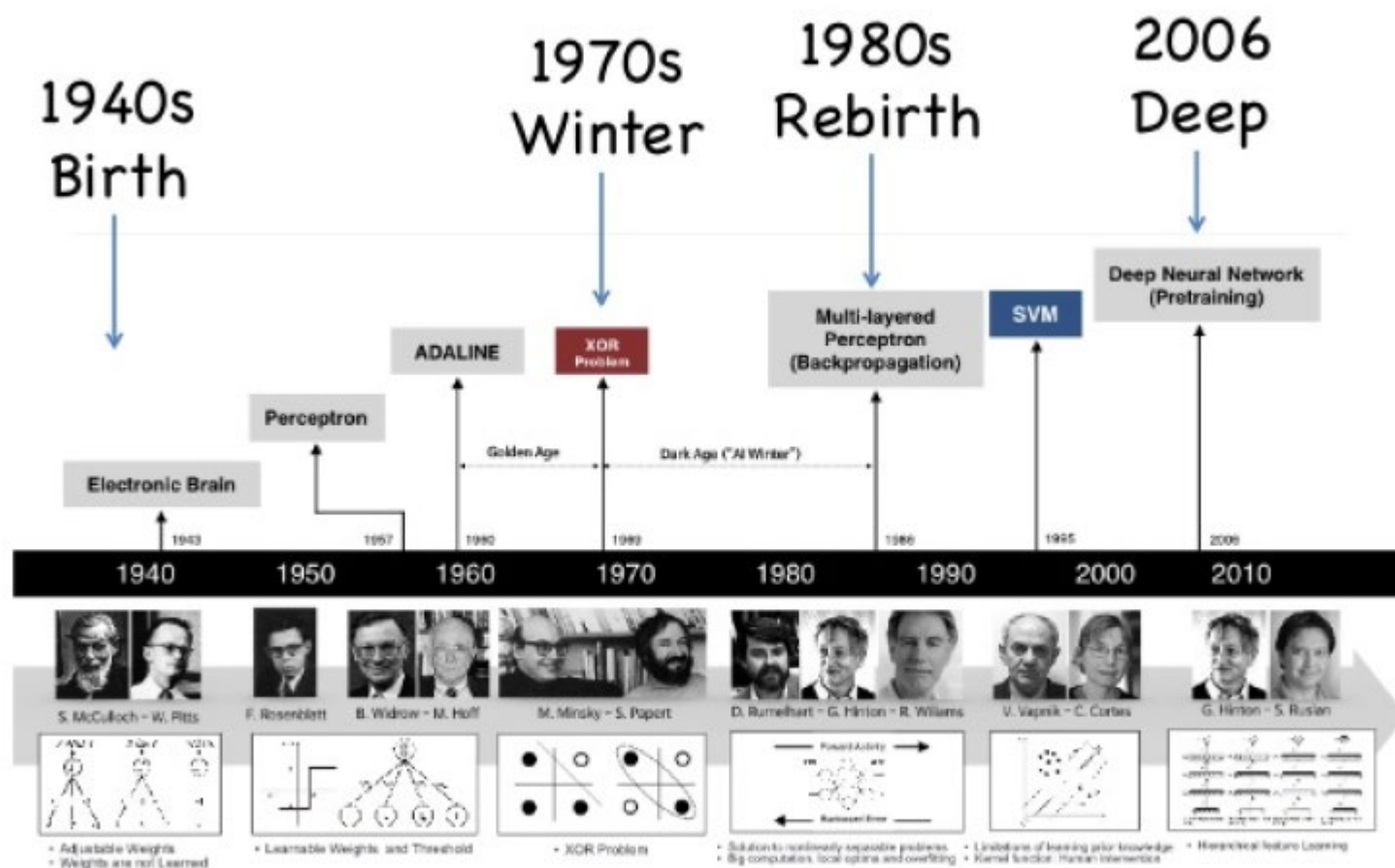
Brief History of Neural Networks (2/4)

- 1943: McCulloch & Pitts
 - Show that neurons can be combined to construct a **Turing machine**
- 1958: Rosenblatt
 - Shows that **perceptron** will converge if what they are trying to learn can be represented
- 1950s: Rosenblatt, Widrow & Hoff
 - First practical networks and learning rules
 - The **Widrow-Hoff learning rule** is still in use today
- 1969: Minsky & Papert
 - Show **the limitations of perceptron**, killing research in neural networks for a decade

Brief History of Neural Networks (3/4)

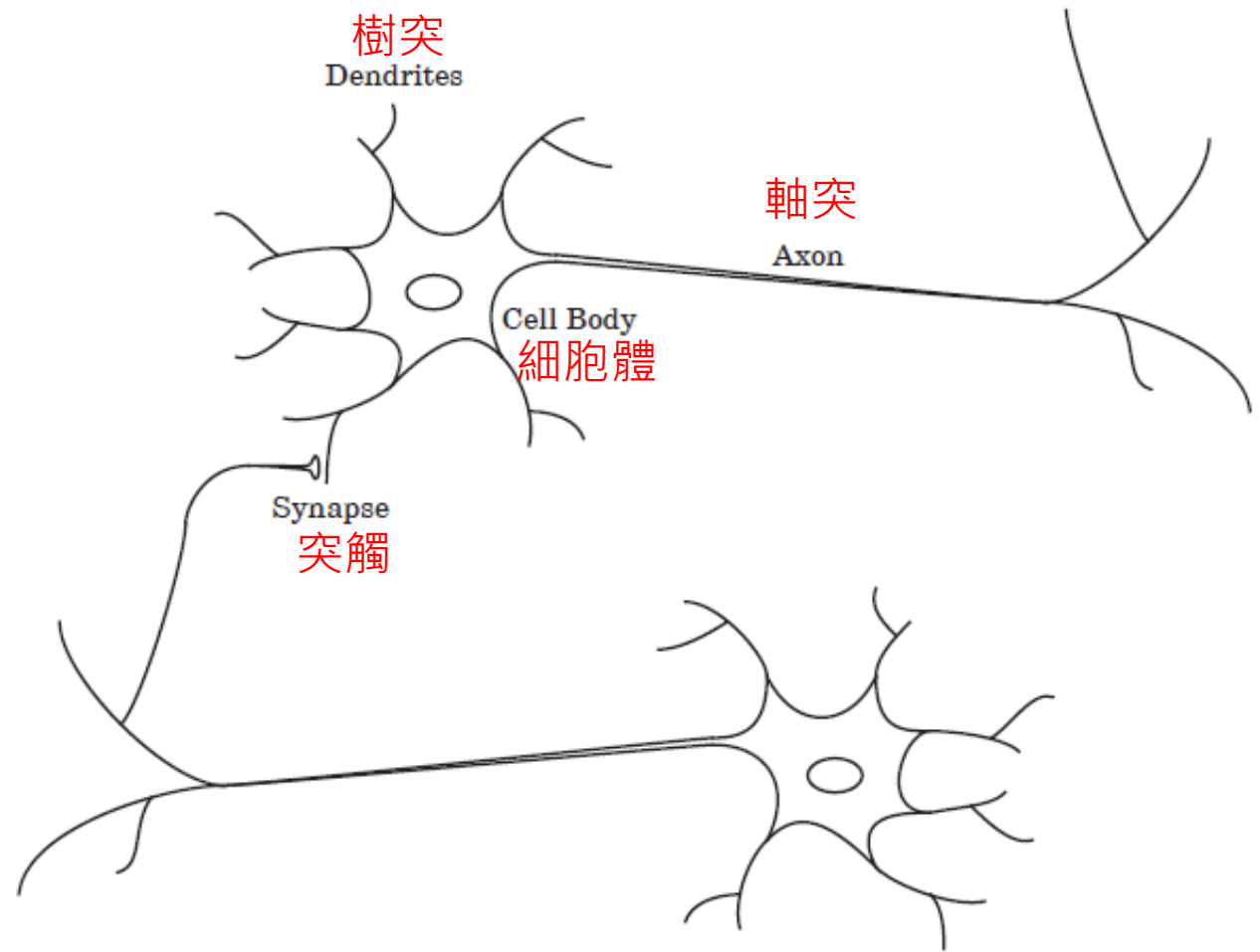
- 1980s: Rumelhart, Hinton & Williams
 - The **backpropagation algorithm** revitalizes the field
- 1998: Yan LeCun
 - **Convolutional Neural Networks with Backpropagation** for document analysis
- 2006: The Hinton lab solves the training problem for **Deep Neural Networks**
- 2012 - present:
 - A variety of **deep learning algorithms** are increasingly emerging

Brief History of Neural Networks (4/4)



Biological Inspiration (1/3)

- Biological motivation
 - Biological learning system (brain)
 - Complex network of neurons

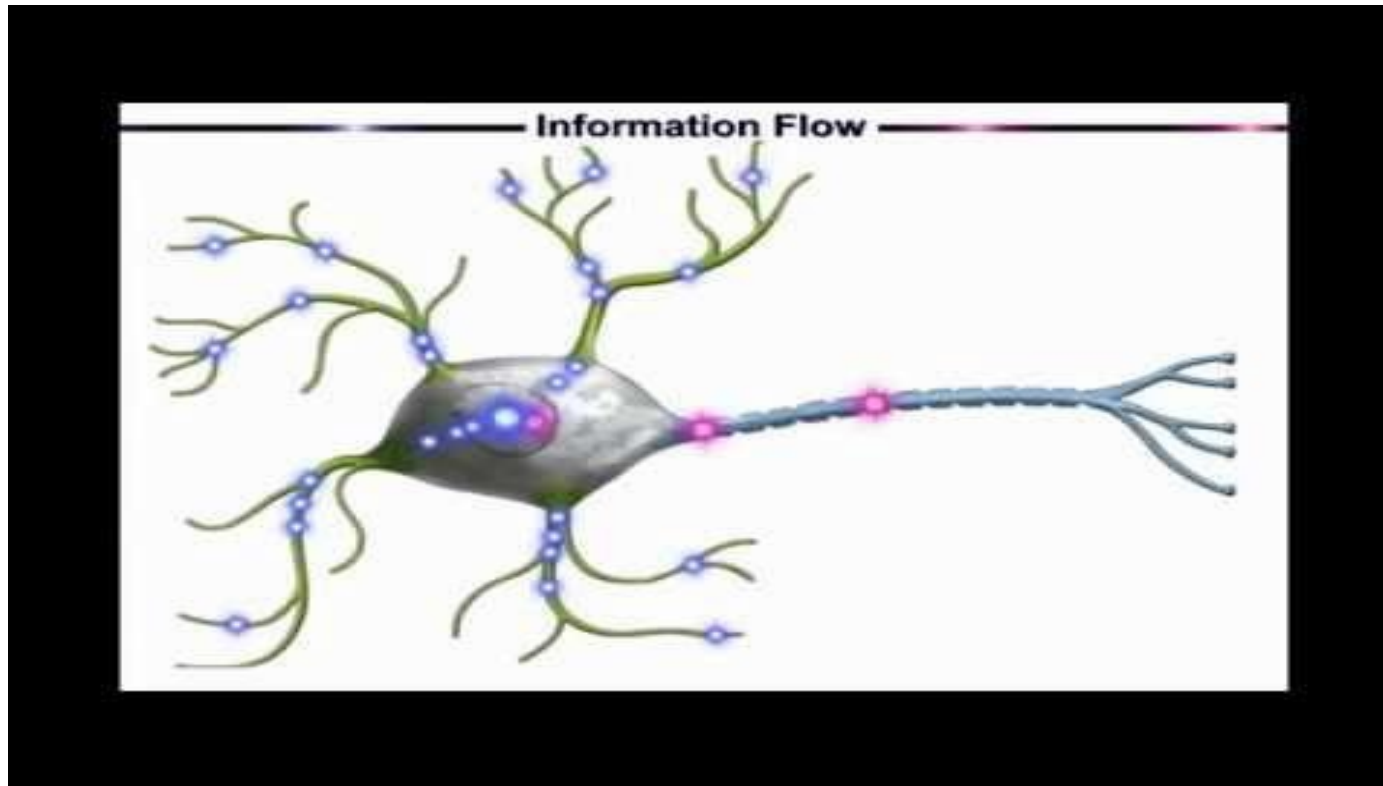


Biological Inspiration (2/3)

- A neuron consists of a **cell body** (細胞體)(中心體), **dendrites** (樹突)(輸入) and an **axon** (軸突)(輸出).
- Neurons are massively interconnected, where an interconnection is between the **axon** of one neuron and a **dendrite** of another neuron.
- This connection is referred to as a **synapse** (突觸)-> 用來決定每個輸入值的重要程度

Biological Inspiration (3/3)

Neural Networks: A Simple Explanation

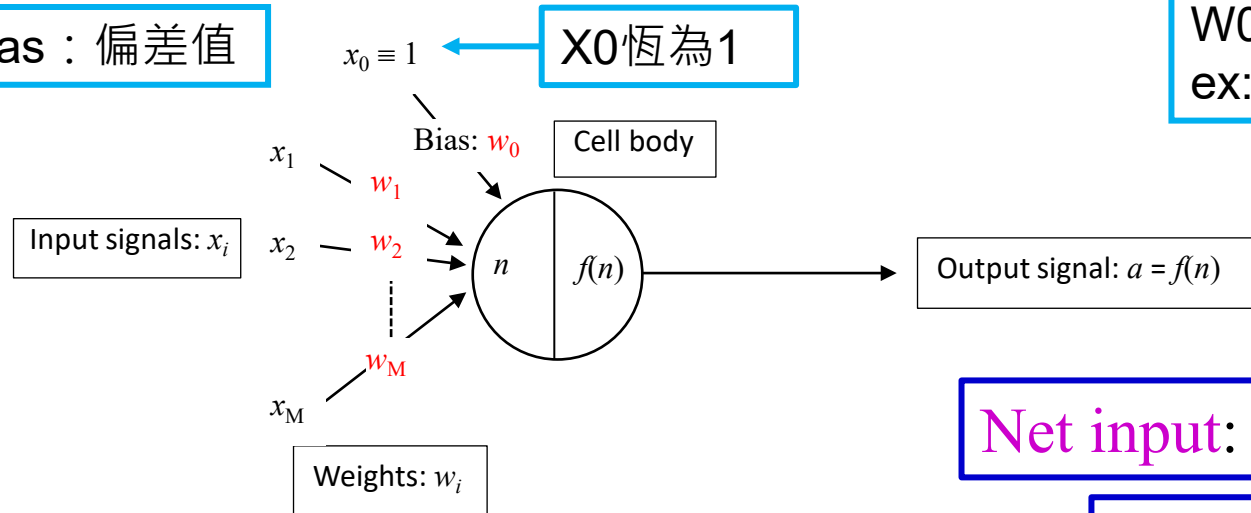


https://youtu.be/gcK_5x2KsLA

Artificial Neuron (1/2)

- An **artificial neuron** is a model of a **biological neuron**.
 - Based on model of brain
 - Network of simple units
 - **Real-valued inputs & outputs**

Bias : 偏差值



w_0 : 用來移動直線
ex: 讓線不通過原點

$$\text{Net input: } n = w_0 x_0 + w_1 x_1 + \dots + w_M x_M$$

Activation Function: $f(n)$

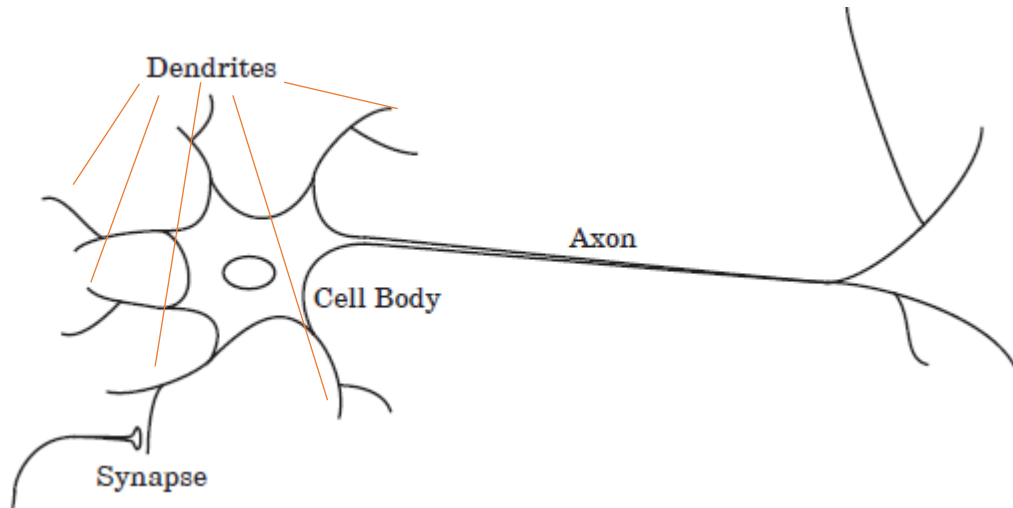
Artificial Neuron (2/2)

Bias and Weight

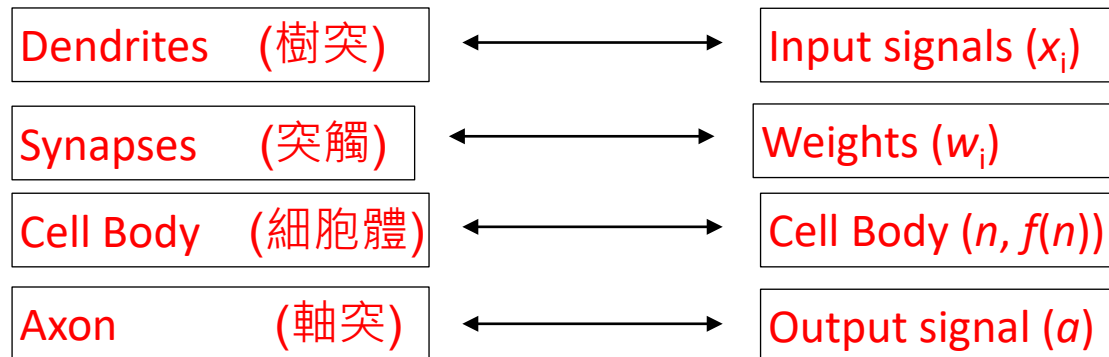
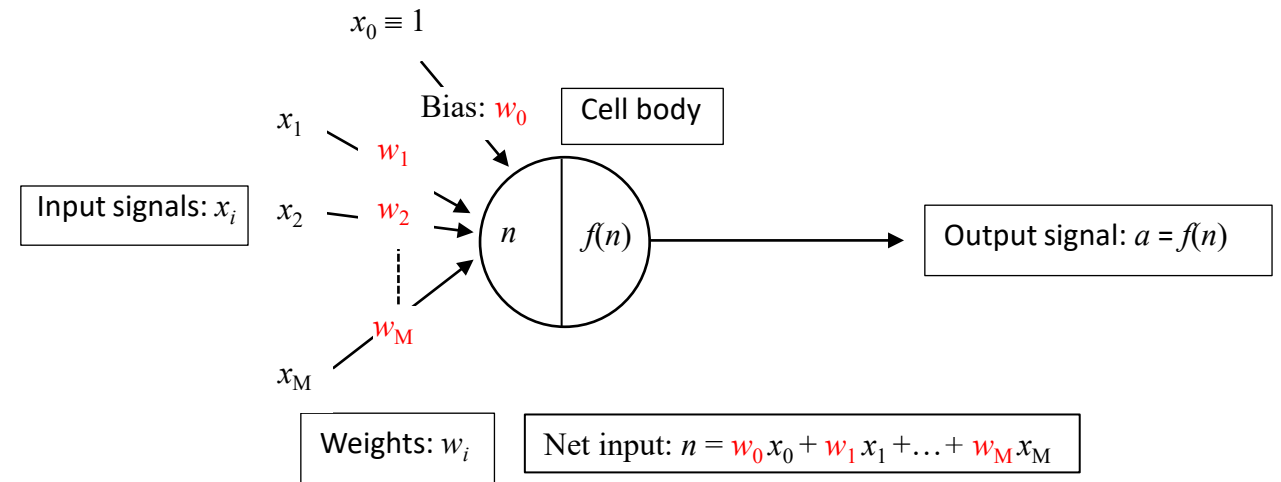
- The **bias** b is much like a **weight** w , except that it has a **constant input of 1**. It can be **omitted** for some situation.
- **The bias** $b (= w_0)$ and **weights** w_i , $i = 1, 2, \dots, M$, are **adjustable** 可調整的 scalar parameters of the neuron. They can be adjusted by some **learning algorithm** so that the neuron input/output relationship meets some particular goal.

Biological Neuron Vs. Artificial Neuron

Biological Neuron



Artificial Neuron



Illustrative Example for Linear Classification

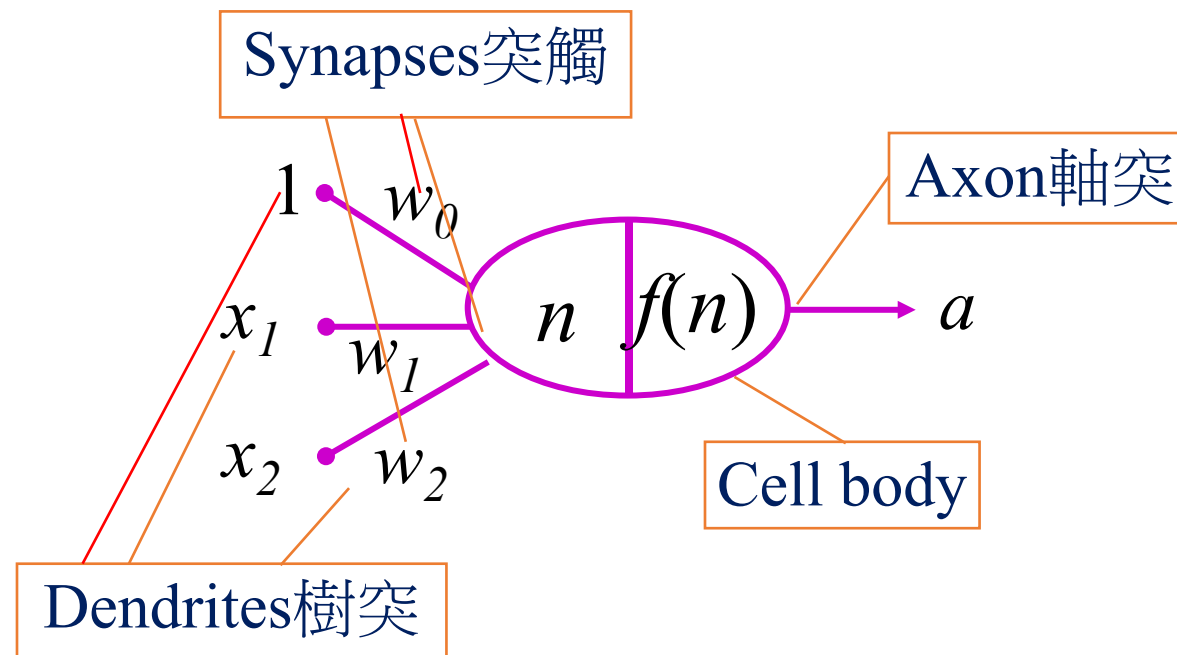
- Dataset D
 - Students take a class on a **Pass/Fail** basis

Student s	x_1 (Midterm)	x_2 (Final)	y (Pass/Fail)
A	80	60	+1 (Pass)
B	50	50	-1 (Fail)
C	90	80	+1 (Pass)
D	30	60	-1 (Fail)
E	40	90	+1 (Pass)
F	90	50	+1 (Pass)

Illustrative Example: Neural Representation

Students' Performance Dataset

Students	x_1 (Midterm)	x_2 (Final)	y (Pass/Fail)
A	80	60	+1
B	50	50	-1
C	90	80	+1
D	30	60	-1
E	40	90	+1
F	90	50	+1



$$\begin{aligned}
 f(x_1, x_2) &= \text{sign}((w_1 x_1 + w_2 x_2) - \text{threshold}) \\
 &= \text{sign}(\underbrace{w_0}_{- \text{threshold}(\text{門檻值})} \underbrace{x_0}_1 + w_1 x_1 + w_2 x_2)
 \end{aligned}$$

Ex:
 $0.4 x_1 + 0.6 x_2 > 50$ 及格
 則 $0.4 = w_1$, $0.6 = w_2$

Activation Functions (1/3) (= transfer function)

- Activation function $f(n)$:
 - Define the output of a node when given the inputs of the node
 - The output of a neuron can either **inhibit** (抑制、不觸發) or **excite** (激發、觸發) the net input n

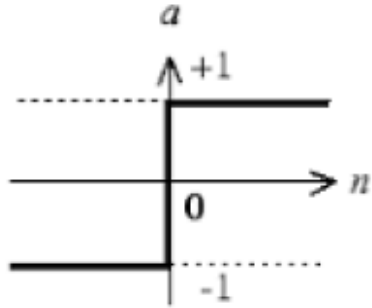
Ex: 及格 -> 觸發
不及格 -> 不觸發

函式考試必考!!

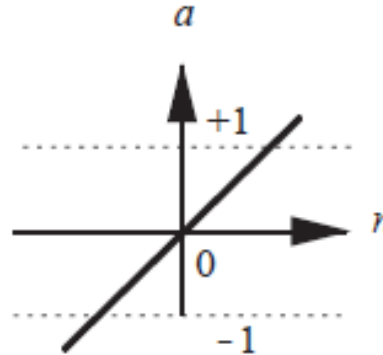
一個東西有很多名稱，表示它很重要，表示有個領域的人來做研究，並用不同的名稱來稱呼它

Activation Functions (2/3)

- The activation function f may be a **linear** or **nonlinear** function of n

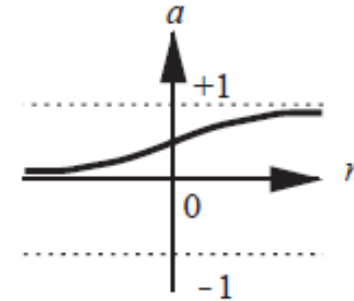


Sign function:
 $f(n) = \text{sign}(n)$



Linear function:
 $f(n) = n$

$0.4 X1 + 0.6 X2 = 0$
Ex: 老師完全不調分，
也沒有門檻值



Logistic sigmoid function:

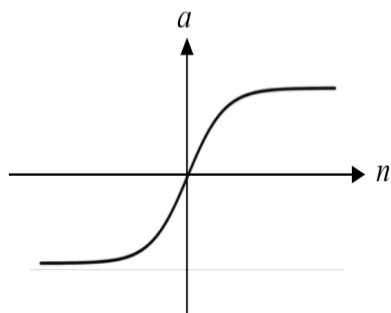
$$f(n) = \sigma(n) = \frac{1}{1+e^{-n}}$$

$$n = 0 \rightarrow \sigma(n) = \frac{1}{2}$$

$$n = \infty \rightarrow \sigma(n) = 1$$

Activation Functions (3/3)

- The activation function f may be a **linear** or **nonlinear** function of n

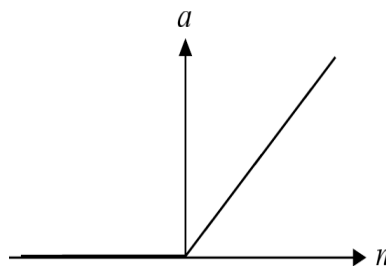


Hyperbolic Tangent Tanh
function (雙曲函數) :

$$f(n) = \frac{e^n - e^{-n}}{e^n + e^{-n}}$$

是**Logistic sigmoid function** 的變形

Ex: 把同學的分數壓縮在1和-1之間



Rectified Linear Unit (ReLU)
function (線性整流函數) :

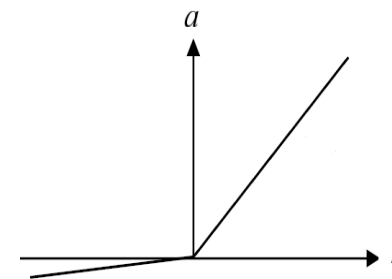
$$f(n) = \max(0, n)$$

-> 變形的線性函數

訊號很強 就希望它的輸出也很強

如果訊號很弱 甚至是負的 就當作它們

不重要



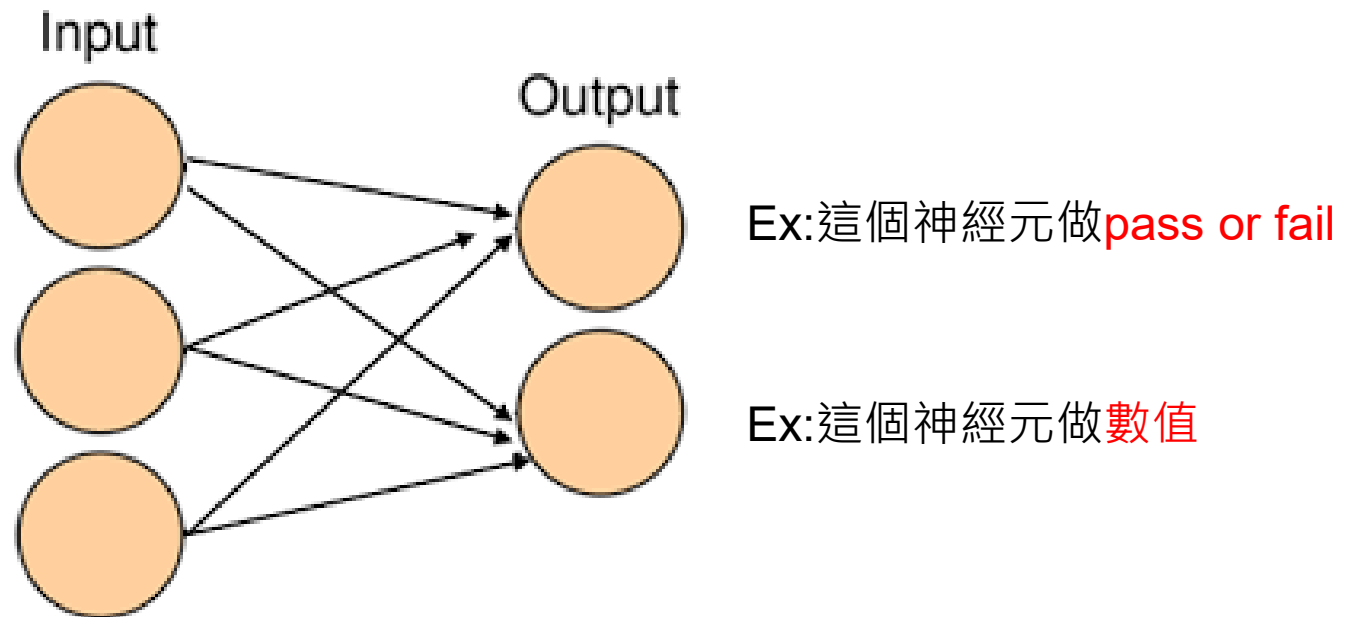
Leaky ReLU function:
 $f(n) = \max(0.01n, n)$

實務上，負的有跟沒有差不多，所以比較

prefer **Rectified Linear Unit ReLU**

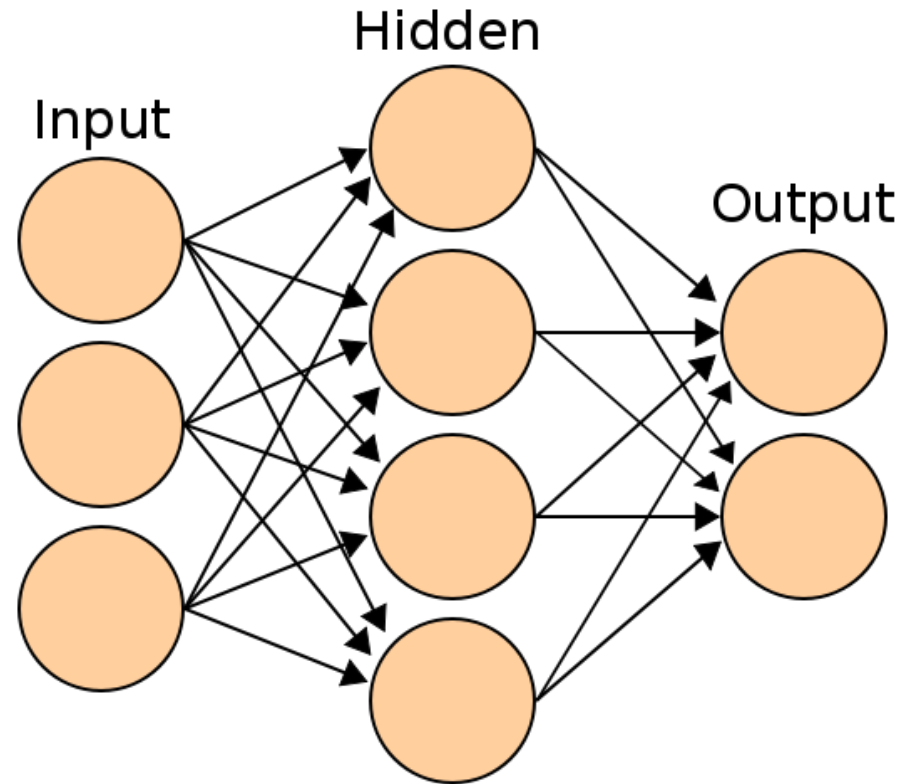
Network Architectures (1/3)

- A Layer of Neurons



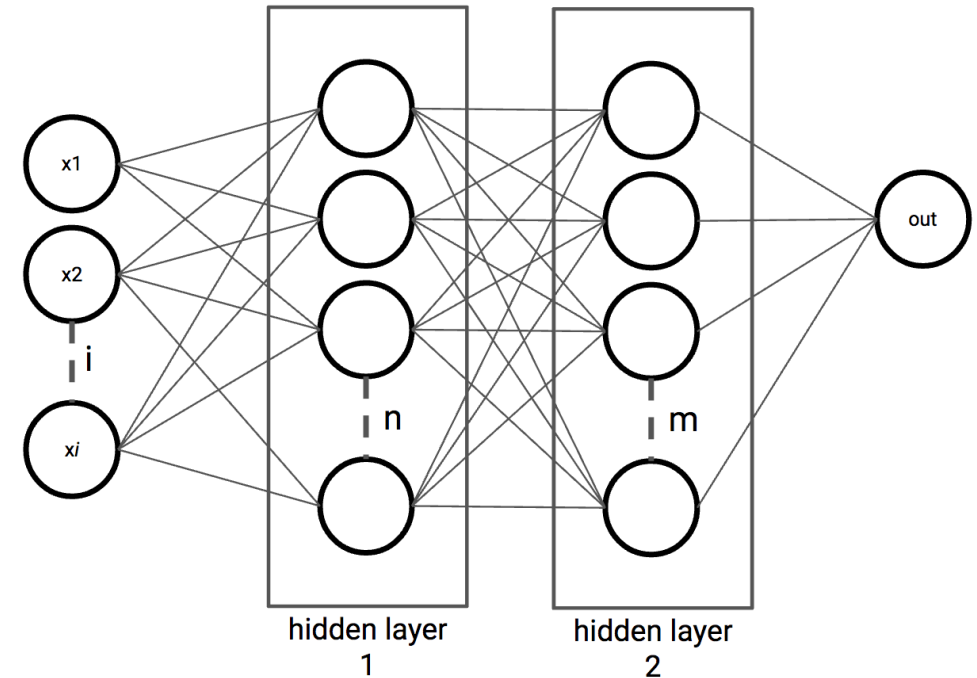
Network Architectures (2/3)

- Multiple Layers of Neurons



Network Architectures (3/3)

- Multiple Layers of Neurons
 - The layer that receives inputs is called the **input layer**.
 - The outputs of the network are generated from the **output layer**.
 - Any layer between the input and the output layers is called a **hidden layer**.



-> 如果問題很簡單 不需要隱藏層

輸入層和輸出層永遠只有一個 但是**隱藏層可以有很多個**

Ex: 如果有兩個隱藏層 一個輸出層 我們稱它為三層的神經元

因為真正有神經元的是隱藏層一 隱藏層二和輸出層 (但也有人說它是四層 ex:有人說 tree 的 root 是在 level 1 也有人說在 level 0)

Representational Power of Feedforward (前饋) Networks

- Boolean functions
 - Every boolean function can be represented exactly by some network with **two** layers (*i.e.*, one single hidden layer) of units
 - But might require exponential (in number of inputs) hidden units
- Continuous functions
 - Every bounded continuous function can be approximated with arbitrarily small error by a network with **two** layers
- Arbitrary functions
 - Any function can be approximated to arbitrary accuracy by a network with **three** layers (*i.e.*, two hidden layers)

Note: Although **a limited depth** of feedforward networks can provide a very expressive hypothesis space for Backpropagation, the number of neurons in **a hidden layer can be exponential** in the input dimension.

Summary

- An **Artificial Neural Network** (ANN) is a computational model inspired by **biological network systems**.
- A **biological** neuron:
 - **Dendrites**, a **cell body**, and an **axon**.
 - **Synapse** is between the axon of one neuron with the dendrite of another neuron.
- An artificial neuron:
 - A set of input nodes, a set of weights, an activation function that relates the total synaptic input to the output (activation) of the neuron.
- Activation function
 - Define the output of a neuron
- The essential elements of an artificial network are an input layer, an output layer, and may contain hidden layers.

References

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- [2] Neural Network Design (2nd Edition), Martin T Hagan , Howard B Demuth , Mark H Beale , and Orlando De Jesús, Martin Hagan, 2014.
- [3] Pattern Recognition and Machine Learning, Christopher Bishop, Springer-Verlag New York, 2006.
- [4] Machine Learning, Tom M. Mitchell, McGraw-Hill, 1997.