

# Computational Intelligence Lecture 2: What Are Neural Networks

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Neuron Modeling of NN Activation Function

Neural Architecture



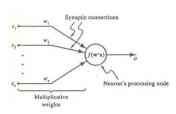


## Neuron Modeling of NN

- McCullouch-Pitts model is introduced in 1943 and the first network is designed
- ► They found out that more precise computations is achieved by combining several neurons in a NN system.
- ► The model considers several drastic simplifications:
  - ▶ It allows only binary states (0-1)
  - Operates under a discrete time assumption
  - Wights and neuron's threshold are fixed
- ▶ Nowadays, Computing algorithms employ a varieties of neuron models with more diversified features.







- ▶ The main artificial neuron models that is used later in this course is:
- ► Each neuron consists of a processing element with synaptic input connections and a single output
- ▶ The neuron output is defined as

$$o = f(W^T X) = f(\sum_{i=1}^n w_i x_i)$$

where  $W = [w_1 \ w_2 \ ... \ w_n]^T$  is the weight vector and  $X = [x_1 \ x_2 \ ... \ x_n]^T$  is input vector.

•  $f(W^TX)$  is activation function.





## **Activation Function**

▶ The simplest definition of activation function is binary with threshold.



where  $net = W^T X$ , and  $\theta$  is threshold level to fire

neuron

► Therefore, output y is defined as

$$y = \begin{cases} 1 & net \ge \theta \\ 0 & net < \theta \end{cases}$$

 $\blacktriangleright$  Any function f(net) that is monotonically nondecreasing and continuous s.t.  $net \in R$  and  $f(net) \in (-1,1)$  can be considered as a NN activation function



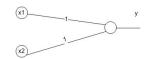
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## ► Example:And

| <i>x</i> <sub>1</sub> | <i>x</i> <sub>2</sub> | $\rightarrow$ | у |
|-----------------------|-----------------------|---------------|---|
| 1                     | 1                     |               | 1 |
| 1                     | 0                     |               | 0 |
| 0                     | 1                     |               | 0 |
| 0                     | 0                     |               | 0 |

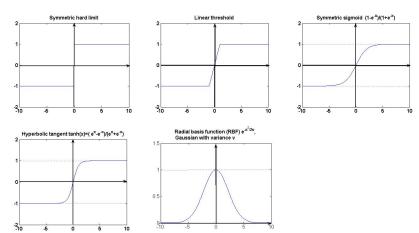
$$y = \left\{ egin{array}{ll} 1 & \textit{net} \geq \theta \\ 0 & \textit{net} < \theta \end{array} 
ight., \theta = 2$$







- ► Most popular activation functions:
  - ▶ Linear It is usually used in output layer when continuous functions are required (such as in control): f(net) = net





- ▶ In discrete NN, the output of each neuron can be
  - ▶ unipolar binary: 0 and 1

Neuron Modeling of NN

- ▶ bipolar: -1 and 1
- ▶ Sometimes, unipolar functions cannot represent the output properly.
- ▶ Unipolar functions are not proper functions for generalization as well



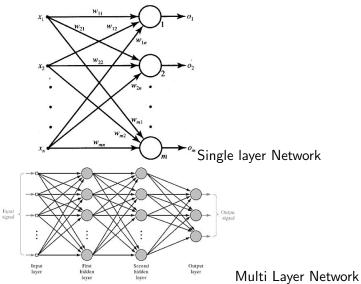
## Neural Architecture

- Neurons at NN are arranged in layers
- ▶ Neurons in the same layer behave in the same manner.
- ► Key factors in determining the behavior of a neuron are its activation function and the pattern of its weight connections
- ▶ Within each layer, neurons usually have the same activation function and the same pattern of connections to other neurons.
- Neural nets are often classified to:
  - 1. Single Layer
    - includes one layer of connection weights.
    - ▶ input units: the units which receive signals from the outside world
    - output units which the response of the net can be read.

#### 2. Multi Layer

- It has layers of nodes between the input units and the output units. (hidden units)
- ► Multilayer nets can solve more complicated problems than can single-layer nets, but training may be more\_difficult.





► The NN based on type of the connection can also be categorized to:

### 1. Feed forward Networks

- ► The signals flow from the input units to the output units, in a forward direction.
- Like Multilayer perceptrons, RBF, etc

### 2. Feedback Networks

- ► It can be obtained from the feed forward network and a feedback connection form the neurons' outputs to their inputs.
- Like Hopfield networks

