# Neuron Model

- Neural network is a general name including both
  - Biological neural networks (e.g. human nervous system)
  - Artificial neural networks
- Our main topic is artificial neural networks (ANNs)
- We will sometimes say "neural network" to refer to an ANN

- Biological neural networks (such as human brain) compute in a different way from today's computers
- The brain is a highly complex, nonlinear, and parallel computer
- It can organize its own structure (connected neurons) to perform certain computations much faster than current computers

- (Artificial) neural network is a machine that is designed to model the way in which the brain performs a particular task or function of interest; usually
  - implemented by using electronic components
  - or simulated in software on a computer
- Our interest will mostly be on a group of ANNs which do useful computations after a *learning* process
- As the name implies, it is a network of smaller computing units called *neurons*

- (Definition by Alexander & Morton 1990)
  - A neural network is a massively parallel distributed processor made up of simple processing units, which has a natural propensity for storing experimental knowledge and making it available for use. It resembles the brain in two respects:
    - Knowledge is acquired by the network from its environment through a learning process
    - Interneuron connection strengths, known as synaptic weights, are used to store the acquired knowledge.

- The procedure used to perform the learning process is called a *learning algorithm* 
  - The main idea here is to modify the synaptic weights of the network in some way so as to achieve a desired objective

# Benefits of ANNs

- Nonlinearity: Neurons can be linear or nonlinear.
   Nonlinearity also comes from the networking. This is an important property particularly when we are working on nonlinear problems.
- <u>Input-Output mapping</u>: An ANN learns how to map inputs to outputs from examples. This is similar to nonparametric statistical inference (a branch of statistics) and tabula rasa learning (biology)
- Adaptivity: An ANN trained to work for a specific case can easily be retrained to deal with minor changes in conditions. In fact, it can be designed to do this in a changing environment. But, there is often a critical line between an adaptive system and a robust one.

# Benefits of ANNs

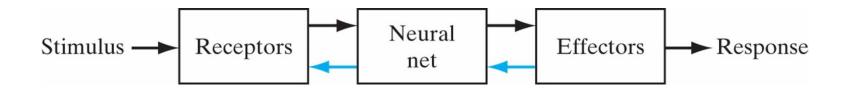
- Evidential Response: An ANN can be designed not only to give us a decision but also to give us how confident it is in that decision.
- <u>Contextual Information:</u> Knowledge is represented by the structure. Every neuron is potentially affected by all others in the network. Therefore, contextual information is dealt with naturally.
- <u>Fault Tolerance</u>: In hardware form, ANNs are fault tolerant in the sense that, if a neuron fails the general performance is only slightly degraded.

# Benefits of ANNs

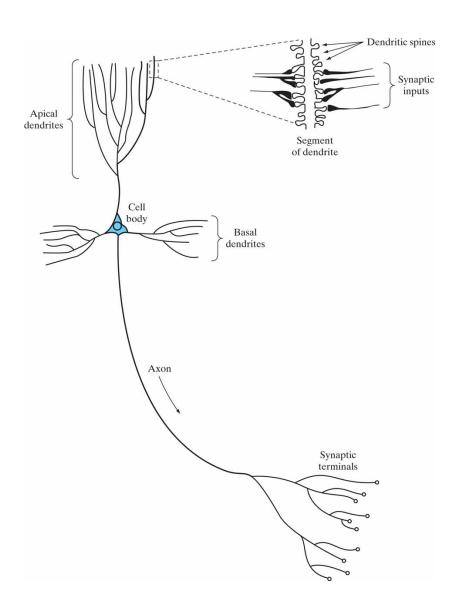
- VLSI Implementability: An ANN is well suited to be implemented using very-large-scale-integrated (VLSI) technology.
- Uniformity of Analysis and Design: Same notation (neurons being the main unit, etc.) is used in all domains involving the application of neural networks.
- Neurobiological Analogy: ANNs are motivated by analogy with the brain, which is a living proof that fault tolerant parallel processing is not only physically possible but also fast and powerful.

#### **Human Brain**

- May be viewed as a three-stage system as below
  - Brain (neural net); Receptors convert stimuli into electrical impulses; Effectors convert electrical impulses into responses (system outputs)
  - Left to right arrows: forward transmission: Right to left: feedback
- Neurons are five to six orders of magnitude slower than silicon logic gates
  - Neural events happen in 10<sup>-3</sup> s range, whereas silicon gate events happen in 10<sup>-9</sup> s
- Yet, brain makes up for this by having extremely many neurons and complex interconnections between them
  - There are approximately 10 billion neurons in the human cortex and 60 trillion connections (synapses)
- Also, brain is energy efficient (10<sup>-16</sup> joules per operation per second)
  - Computers today have about 10<sup>-6</sup> joules per operation per second)

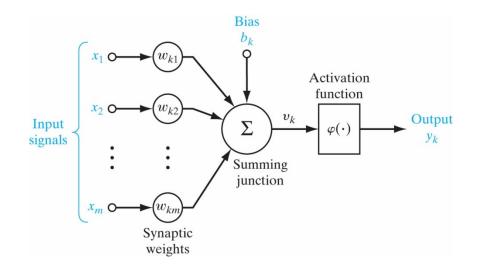


# The Pyramidal Cell



# **Artificial Neuron Models**

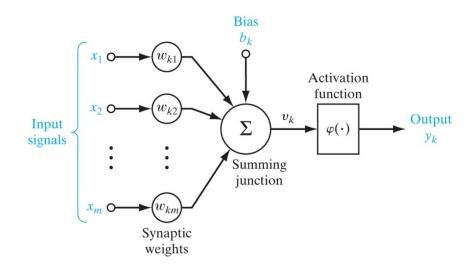
- A neuron is the fundamental information processing unit of a neural network
- The diagram on the right shows a neuron model including
  - A set of synapses (Each with a weight,  $w_{ki}$ )
  - An adder (a linear combiner)
  - An activation function
  - A bias value  $(b_k)$  to modify the net input of activation function



# Artificial Neuron Models

 Mathematically, the following pair of equations describe neuron k

$$u_k = \sum_{i=1}^{m} w_{ki} x_i$$
$$y_k = \varphi(u_k + b_k)$$



# **Artificial Neuron Model**

• Use of bias  $(b_k)$  applies an affine transformation to  $u_k$ 

$$v_k = u_k + b_k$$

 $v_k$  is called activation potential (or induced local field)

 Using activation potential, instead of the previous equations we can write

$$v_k = \sum_{i=0}^{\infty} w_{ki} x_i$$
$$y_k = \varphi(v_k)$$

where  $x_0 = 1$ , and  $w_{k0} = b_k$ 

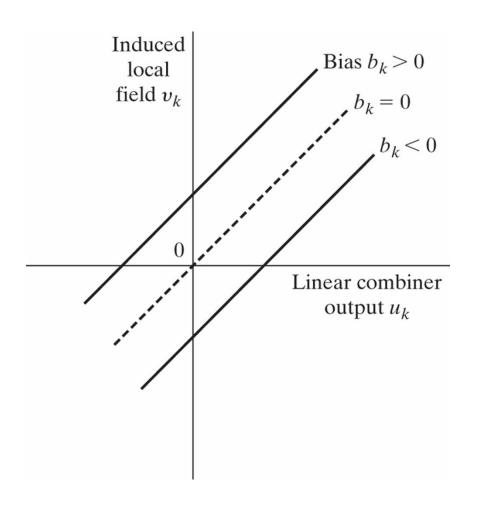


Figure 7 Another nonlinear model of a neuron;  $w_{k0}$  accounts for the bias  $b_k$ .

