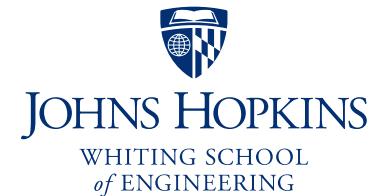


# **Johns Hopkins Engineering**

## **Applied Machine Learning for Mechanical Engineers**

Machine Learning Fundamentals, Part 1, C



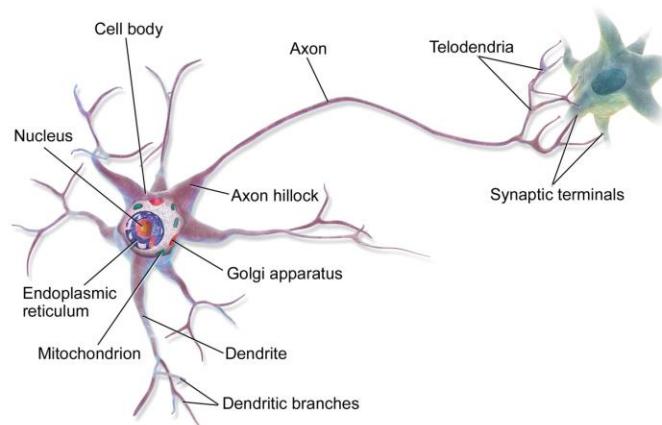
# Perceptron and Neural Networks

- By the end of this lecture you will be able to:
  - Describe perceptron
  - Describe activation/transfer functions
  - Describe multi-layer perceptron
  - Describe neural network optimization

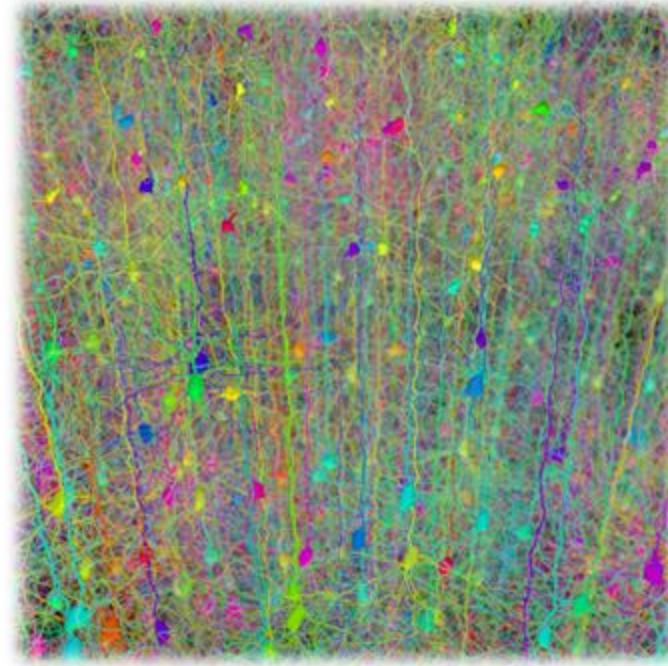
# Perceptron and Neural Networks

## ■ Perceptron

- Brain is a giant network of neurons connected to one another



Simplified biological model of a neuron in brain

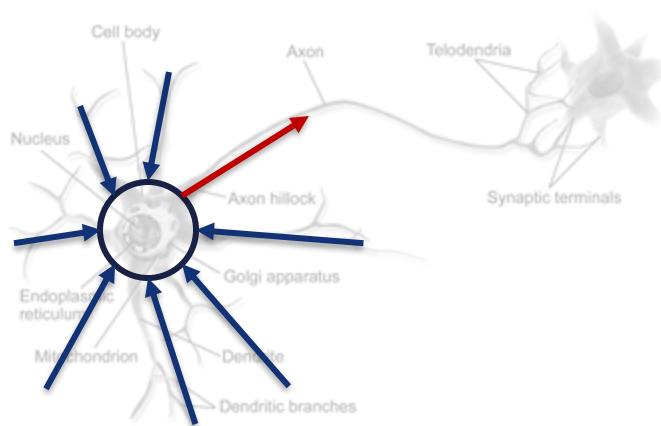


Computer simulation of dendrites of pyramidal network of neurons (biological neural networks)

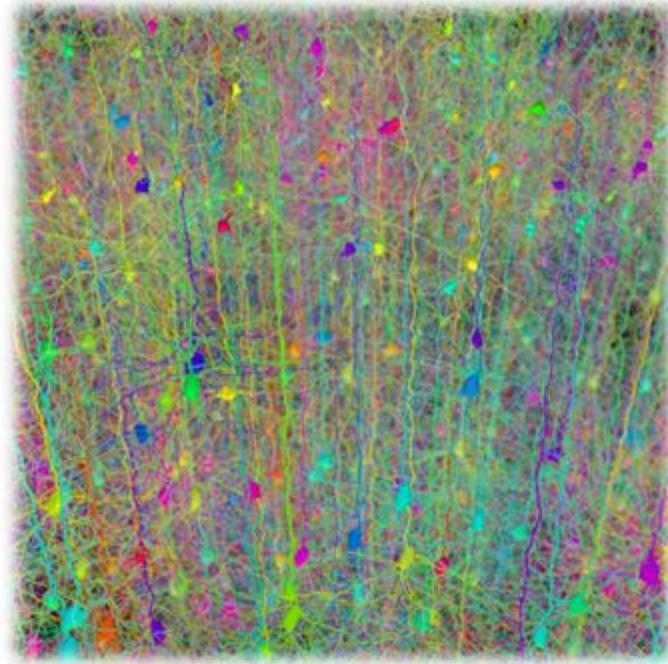
# Perceptron and Neural Networks

## ■ Perceptron

- Brain is a giant network of neurons connected to one another



Simplified biological model of a neuron in brain

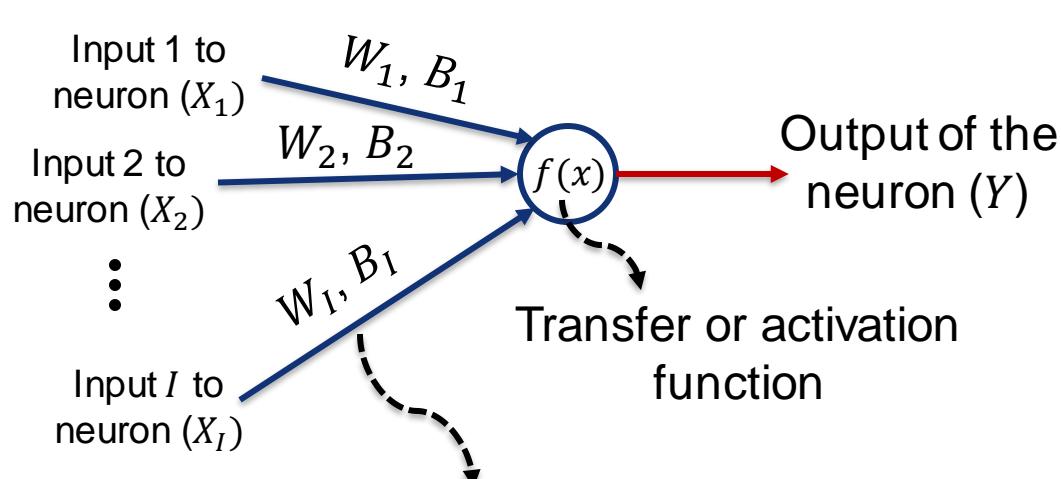


Computer simulation of dendrites of pyramidal network of neurons (biological neural networks)

# Perceptron and Neural Networks

## ■ Perceptron

- Simplified perceptron as a simple neural network by Rosenblatt (1958) and Minsky and Seymour (1969), so-called artificial neural networks



$$x = \sum_{i=1}^I X_i W_i + B_i$$
$$Y = f(x) = f\left(\sum_{i=1}^I X_i W_i + B_i\right)$$

Connection weights (also have an associated bias value)

# Perceptron and Neural Networks

- Perceptron
  - Simplified perceptron as a simple neural network by Rosenblatt (1958) and Minsky and Seymour (1969), so-called artificial neural networks

Optimization problem: what are the optimum values of all  $W_n$  and  $B_n$  such minimize (maximize) error (accuracy) measurement of  $Y_{estimated}$  versus  $Y_{real}$ ?

Fitting one or very few datapoints

$$x = \sum_{i=1}^I X_i W_i + B_i$$
$$Y = f(x) = f\left(\sum_{i=1}^I X_i W_i + B_i\right)$$

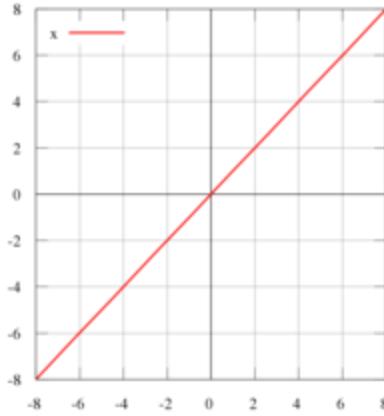
Weight      Bias

The diagram illustrates the perceptron model. It shows inputs  $X_i$  (represented by dashed arrows) being multiplied by weights  $W_i$  (also represented by dashed arrows). These weighted inputs are then summed with a bias  $B_i$  (represented by a dashed arrow). The final result of this summation is passed through an activation function  $f$ , which is shown as a curved bracket.

# Perceptron and Neural Networks

- Activation/transfer functions
  - Popular activation/transfer functions
    - Linear or identity function

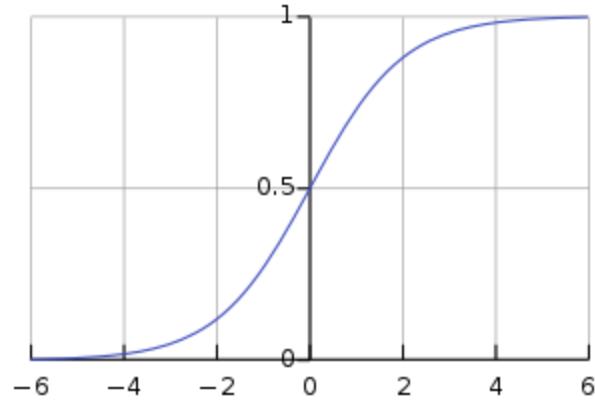
$$f(x) = x$$



# Perceptron and Neural Networks

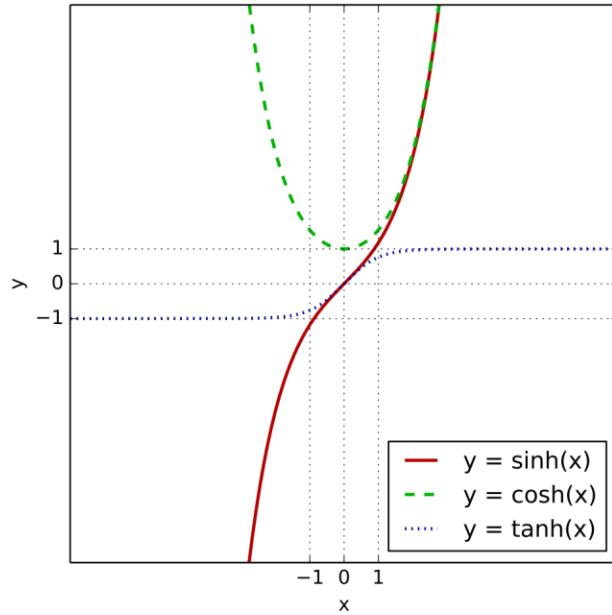
- Activation/transfer functions
  - Popular activation/transfer functions
    - Sigmoid or logistic function

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



# Perceptron and Neural Networks

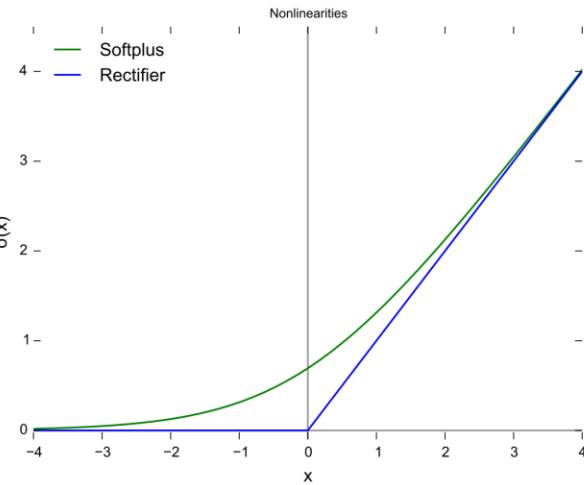
- Activation/transfer functions
  - Popular activation/transfer functions
    - Hyperbolic functions such as  $\tanh(x)$
- $f(x) = \tanh(x)$



# Perceptron and Neural Networks

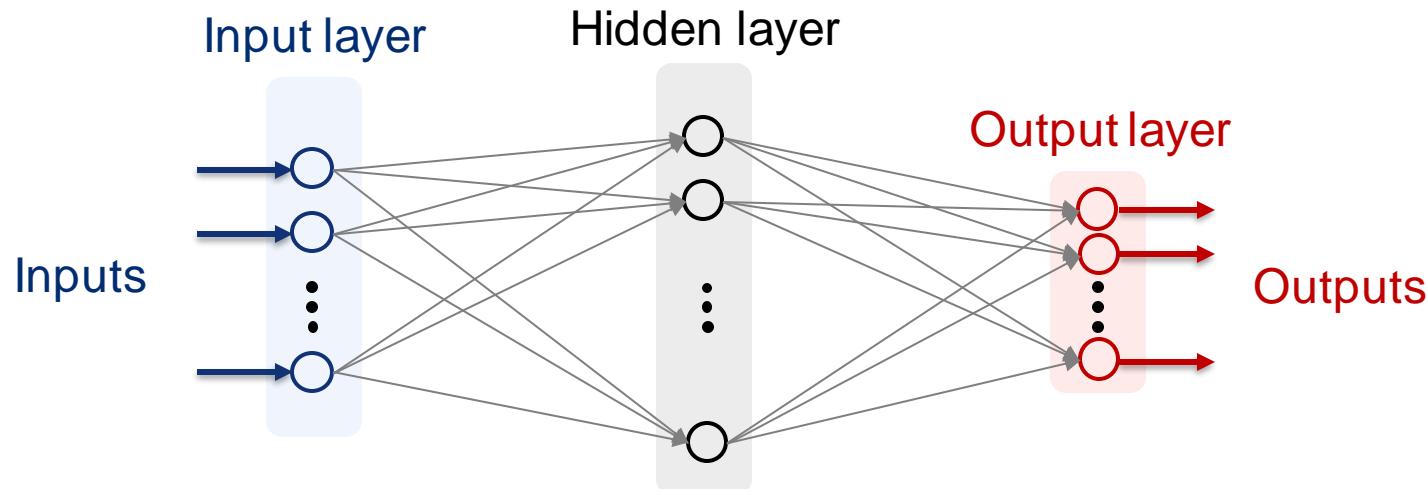
- Activation/transfer functions
  - Popular activation/transfer functions
    - Rectified linear unit (ReLU)

$$f(x) = \begin{cases} 0 & x \leq 0 \\ 1 & x > 0 \end{cases}$$



# Perceptron and Neural Networks

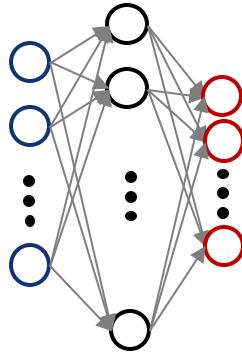
- Multi-Layer perceptron (feed-forward artificial neural network)
  - Increase memory by increasing complexity for complex problems



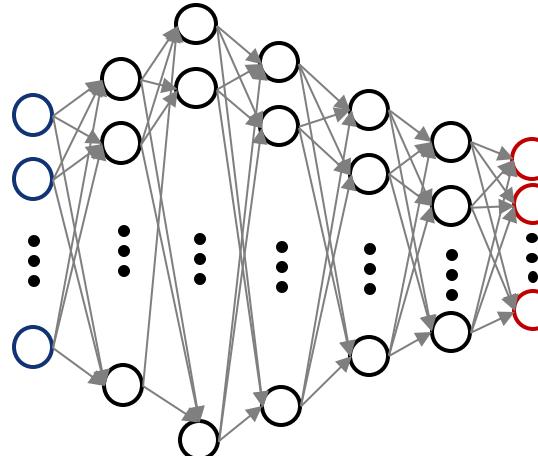
# Perceptron and Neural Networks

- Multi-Layer perceptron (feed-forward artificial neural network)
  - Increase memory by increasing complexity for complex problems

Shallow  
Neural Network  
(1 hidden layer)



Deep  
Neural Network  
(2 or more hidden layers)



# Perceptron and Neural Networks

- Neural network optimization
  - What are the optimum values of weights and biases of all layers of a neural network such when being fed (i.e., inputted and outputted) by every training datapoint in a repository, the estimated output of every training datapoint becomes very close to the actual output of that training datapoint?
  - What are the optimum values of weights and biases of all layers of a neural network such when being fed (i.e., inputted and outputted) by every training datapoint, the overall estimation error becomes minimized (or the accuracy becomes maximized)?
  - Fitness function?

# Perceptron and Neural Networks

## ■ Neural network optimization

- The fitness function in neural network usually called **loss function** (sometimes is referred to as the cost function), and has a variety of forms depending on the problem (e.g., regression, classification, etc.) or the machine learning algorithm (e.g., supervised, unsupervised, classifier, regressor, etc.)
- Example (with respect to notations used in slide 3 of lecture B of this module)

$$l(p) = \frac{1}{2N} \sum_{n=1}^N \|Y_n - \tilde{Y}_n\|^2$$

where  $p$  is a set of optimization variables including weights and biases of different layers of neural networks,  $\|\cdot\|$  is the norm function, and  $Y_n$  and  $\tilde{Y}_n$  are the actual and estimated outputs of  $n^{th}$  training datapoint. In this example, loss function computes the average of mean squared error of estimations.

# Perceptron and Neural Networks

- In this lecture, you learned about:
  - Perceptron
  - Activation/transfer functions
  - Multi-layer perceptron
  - Neural network optimization
- In the next lecture, we will talk about learning algorithms



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