

Johns Hopkins Engineering

Applied Machine Learning for Mechanical Engineers

Machine Learning Fundamentals, Part 1, C



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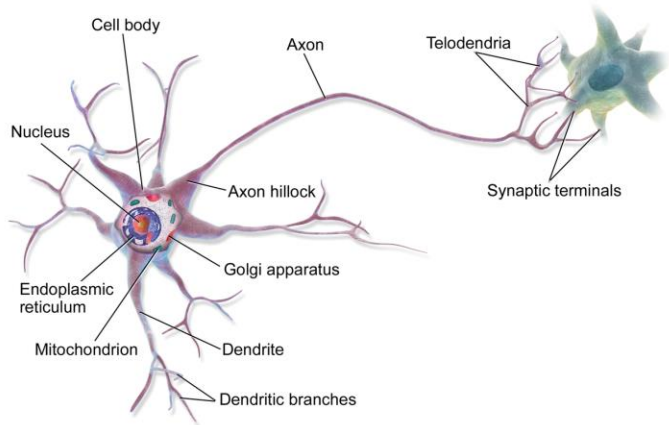
Perceptron and Neural Networks

- 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

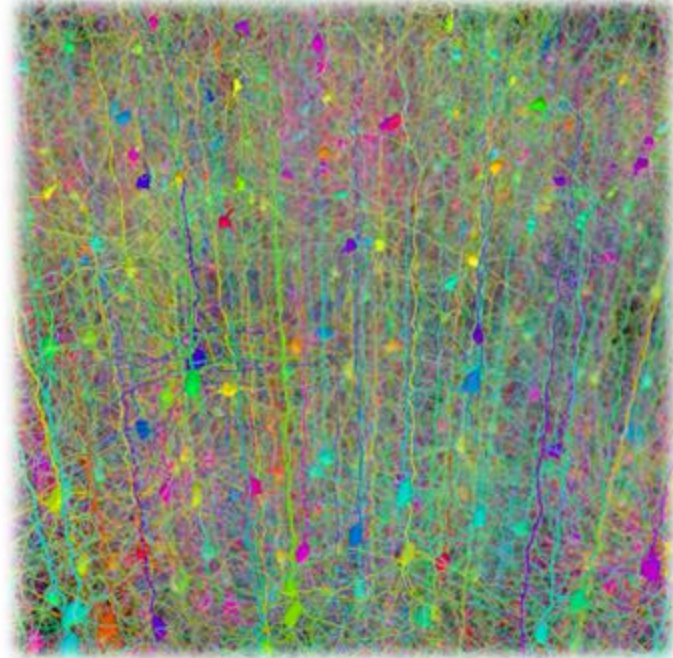
Perceptron and Neural Networks

■ Perceptron

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



Simplified biological model of a neuron in brain

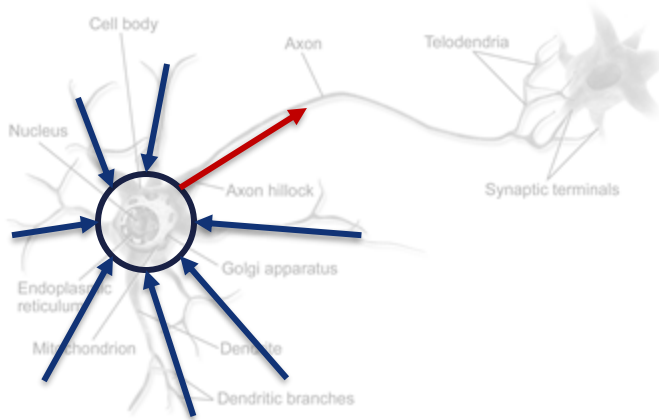


Computer simulation of dendrites of pyramidal network of neurons (biological neural networks)

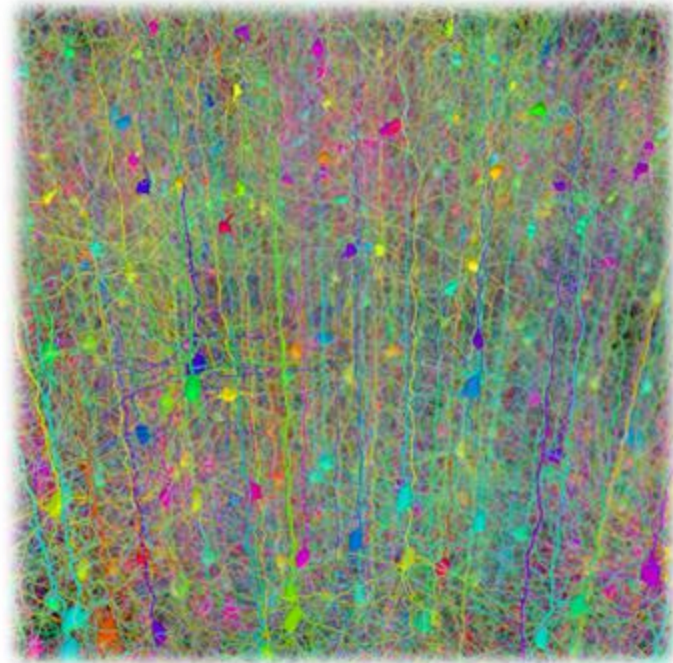
Perceptron and Neural Networks

■ Perceptron

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



Simplified biological model of a neuron in brain

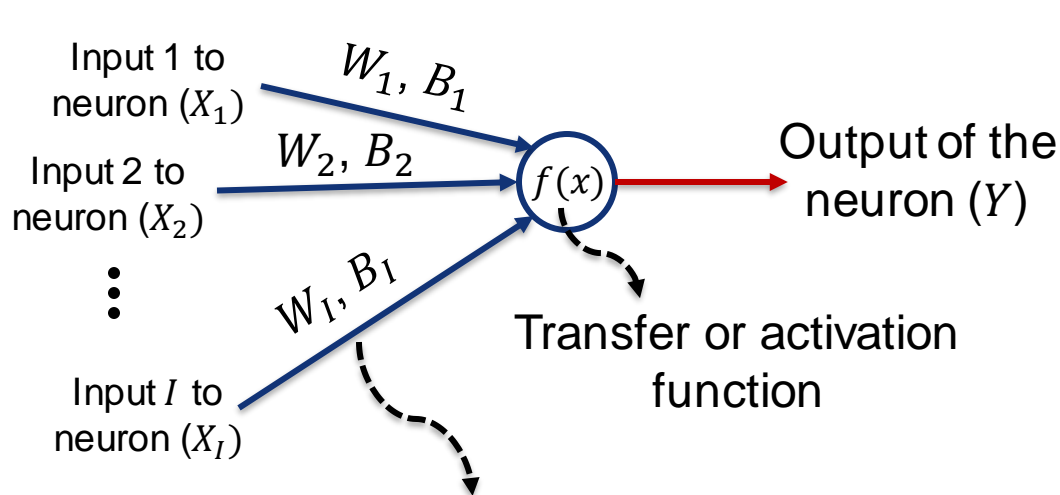


Computer simulation of dendrites of pyramidal network of neurons (biological neural networks)

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



$$x = \sum_{i=1}^I X_i W_i + B_i$$

Weight Bias

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

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

$$x = \sum_{i=1}^I X_i W_i + B_i$$

Weight Bias


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

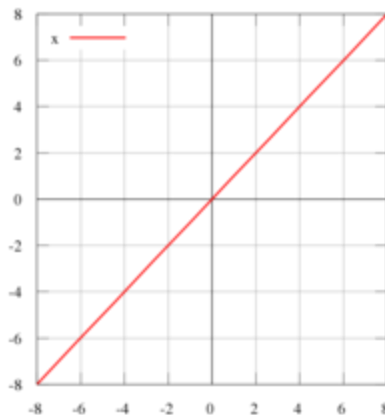
Perceptron and Neural Networks

- Activation/transfer functions

- Popular activation/transfer functions

- Linear or identity function

$$f(x) = x$$



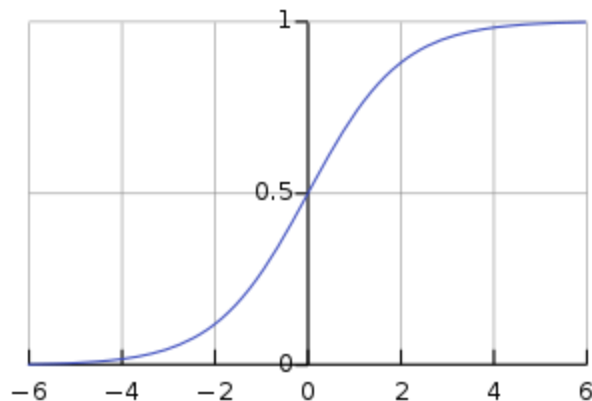
Perceptron and Neural Networks

- Activation/transfer functions

- Popular activation/transfer functions

- Sigmoid or logistic function

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

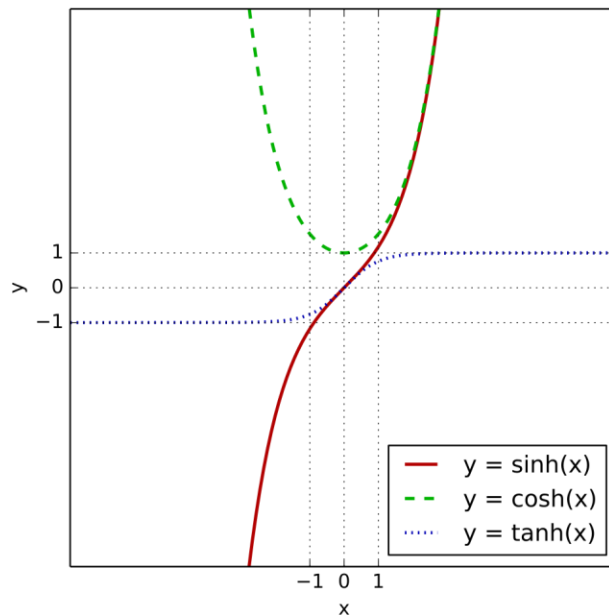


Perceptron and Neural Networks

■ Activation/transfer functions

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

$$f(x) = \tanh(x)$$



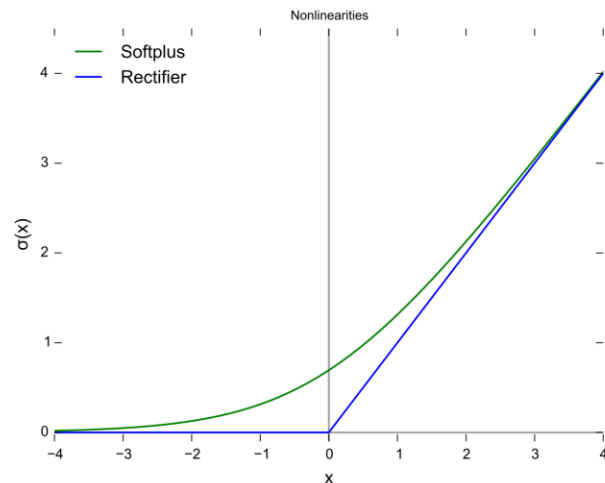
Perceptron and Neural Networks

■ Activation/transfer functions

○ Popular activation/transfer functions

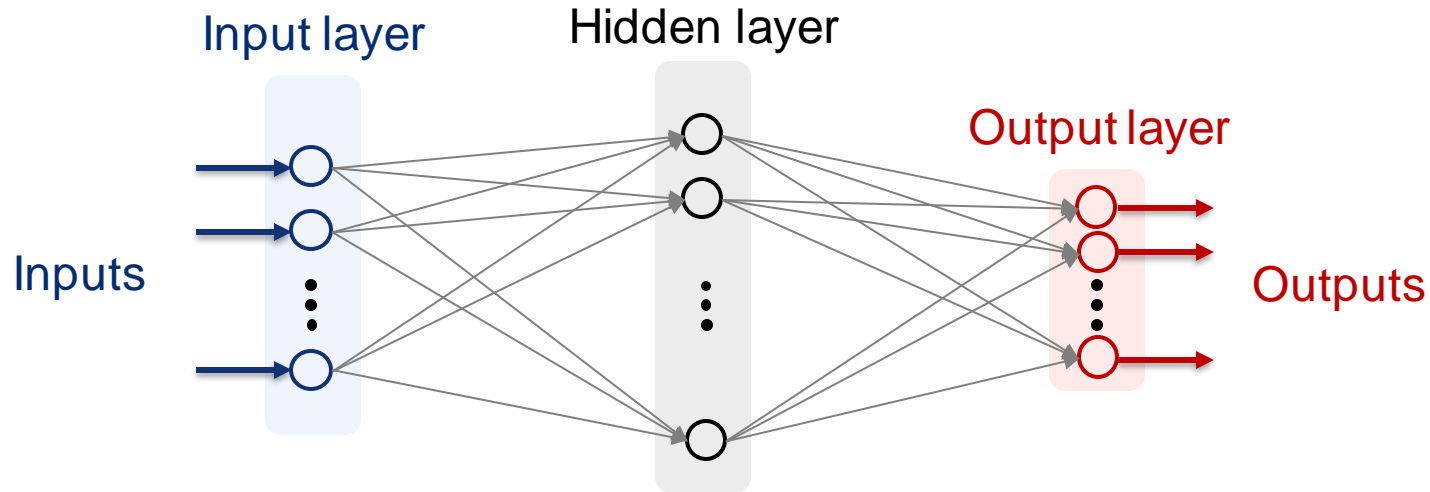
● Rectified linear unit (ReLU)

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



Perceptron and Neural Networks

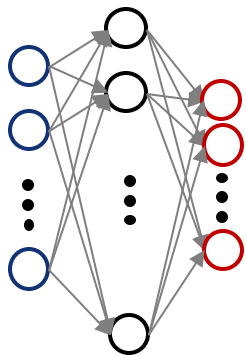
- Multi-Layer perceptron (feed-forward artificial neural network)
 - Increase memory by increasing complexity for complex problems



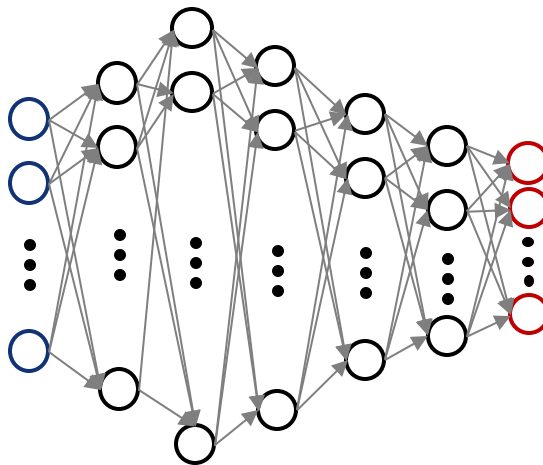
Perceptron and Neural Networks

- 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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