

Deep Learning

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YSU, Krisp

September 11, 2019

Outline

- 1 Course Schedule
- 2 References
- 3 What is Supervised Learning?
- 4 What is a neural network?
- 5 Gradient Descent

Course Schedule

1. Intro to supervised learning.
2. Overfitting, underfitting
3. What is a neural network?
4. Gradient Descent
5. Linear Regression, Logistic Regression
6. Activation functions
7. Softmax classifier
8. Data normalization
9. Back propagation
10. Random Initialization
11. Regularization
12. Dropout
13. Batch normalization
14. Data augmentation
15. Vanishing / Exploding gradients
16. Mini-batch gradient descent

Course Schedule

17. Gradient descent with momentum
18. RMSprop
19. Adam optimization algorithm
20. Learning rate tricks
21. Batch normalization
22. Introduction to TensorFlow
23. Convolutional neural networks
24. ResNets
25. Inceptions
26. VGG
27. Transfer learning
28. Autoencoders
29. GANs
30. Multitask learning
31. Basic Recurrent Neural Networks
32. GRU

- 33. LSTM
- 34. Bidirectional RNN
- 35. Multicell RNNs
- 36. Attention models
- 37. Bayesian neural networks
- 38. Word2Vec
- 39. Sequence to sequence models
- 40. GPU optimisations for Neural Networks

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1. Ian Goodfellow, Yoshua Bengio and Aaron Courville - Deep Learning
2. Original Papers
3. Blog Posts

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What is Supervised Learning?

Definition 1

*In machine learning, **supervised learning** refers to a class of systems and algorithms that determine a predictive model using data points with known outcome.*

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Supervised learning is often used to create machine learning models for two types of problems:

- 1 Regression

What is Supervised Learning?

Definition 1

*In machine learning, **supervised learning** refers to a class of systems and algorithms that determine a predictive model using data points with known outcome.*

Supervised learning is often used to create machine learning models for two types of problems:

- 1 Regression
- 2 Classification

Example of regression problem

Example of regression problem

Size of House	Price of House
950	\$123,325
1,535	\$156,570
1,605	\$158,895
1,905	\$200,025
2,057	\$230,384
2,227	\$233,835
3,150	\$261,420
3,620	\$433,500

Example of classification problem

Example of classification problem



Our data consist of pairs

$$(x_i, y_i)_{i=1}^n, \text{ where } x_i \in \mathbb{R}^k, y_i \in \mathbb{R}^m, i = 0, 1, \dots, n.$$

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We have to find a function f for which

$$f(x_i) \approx y_i$$

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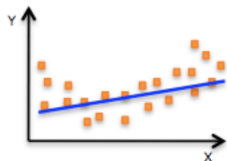
$$f(x_i) \approx y_i$$

and not only for pairs in our data.

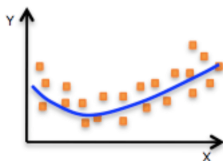
Train, Validation, Test



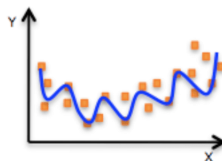
Overfitting, Underfitting



Underfitting

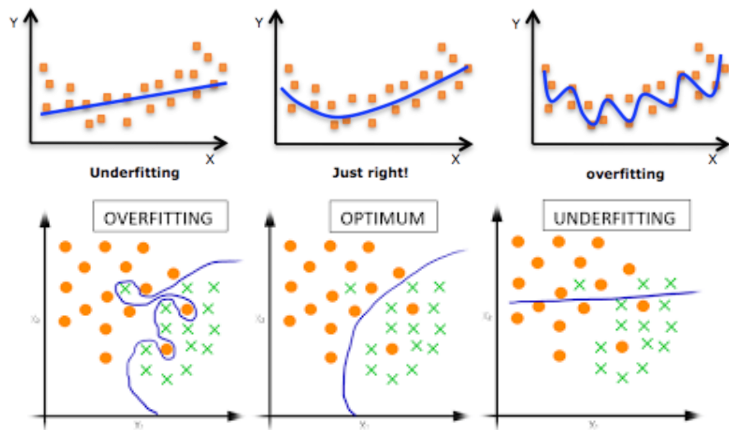


Just right!



overfitting

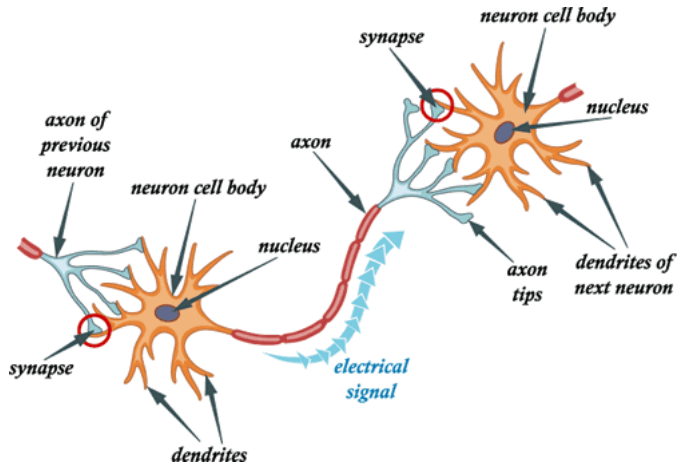
Overfitting, Underfitting



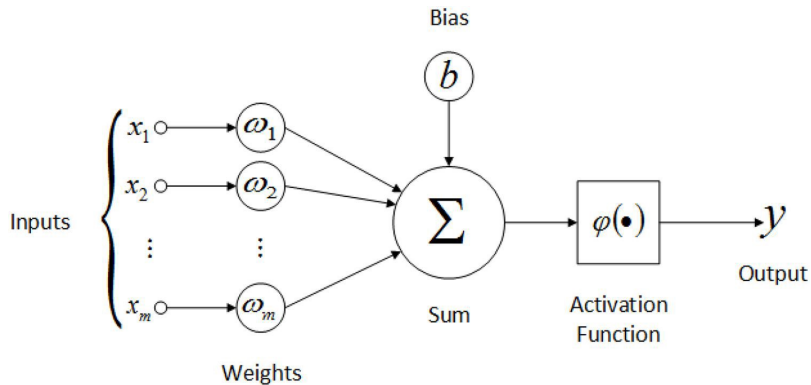
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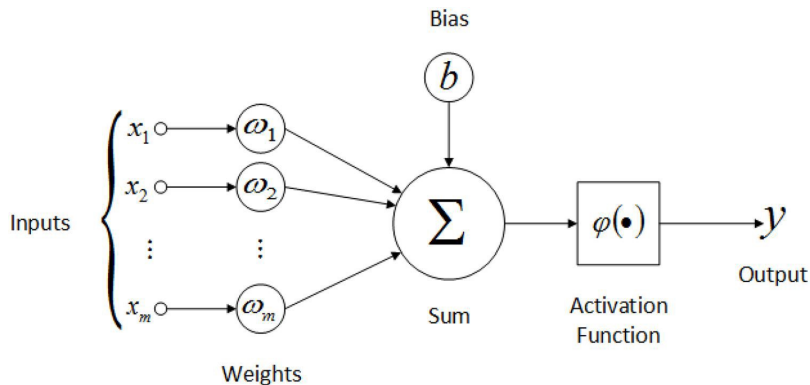
Human Brain



Artificial neuron



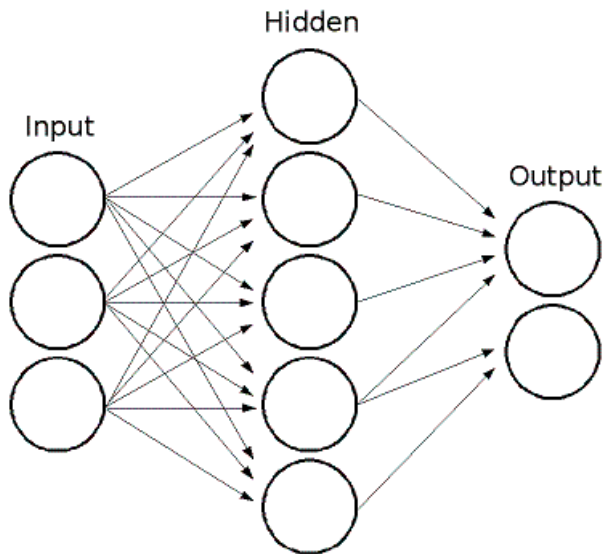
Artificial neuron



So artificial neuron is a function from \mathbb{R}^m to \mathbb{R} , where m is the dimension of input:

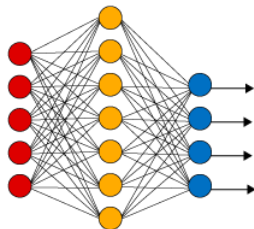
$$f_{w,b}(x_1, x_2, \dots, x_m) = \phi(w_1x_1 + w_2x_2 + \dots + w_mx_m + b) = \phi(w^T x + b)$$

Hidden Layer



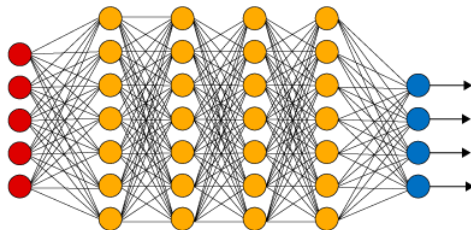
Deep neural network

Simple Neural Network



● Input Layer

Deep Learning Neural Network



● Hidden Layer

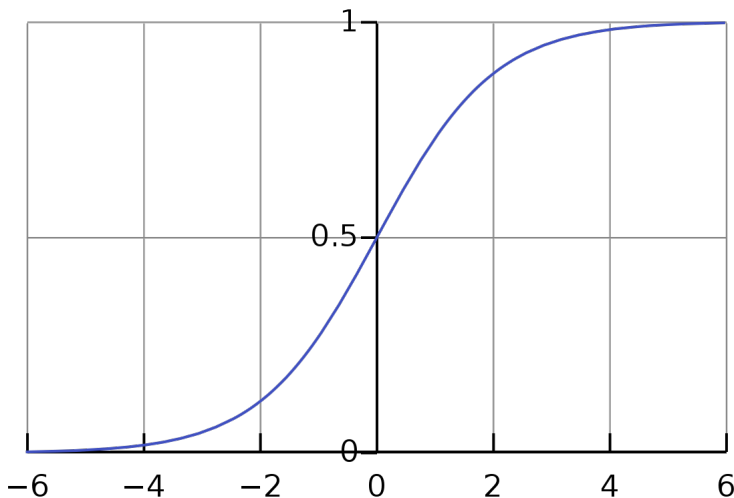
● Output Layer

Activation functions

1. Sigmoid: $\sigma(x) = \frac{1}{1 + e^{-x}}$

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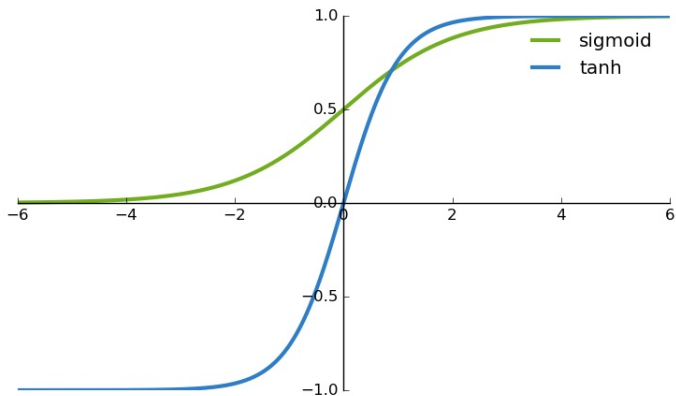


Activation functions

2. Tanh: $\tanh(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}}$

Activation functions

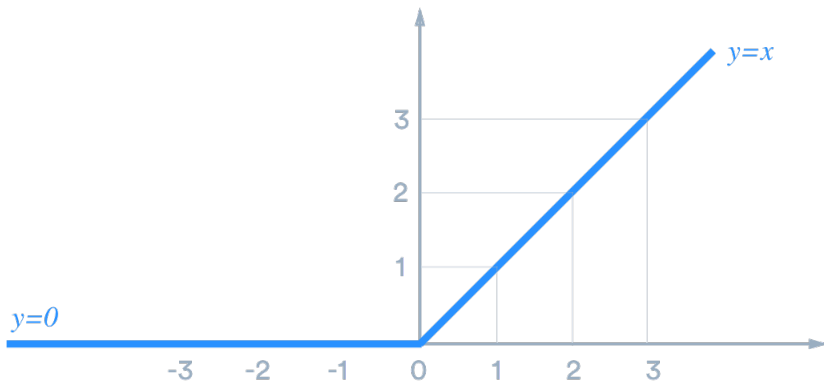
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3. Rectified linear unit: $ReLU(x) = \max(0, x)$

Activation functions

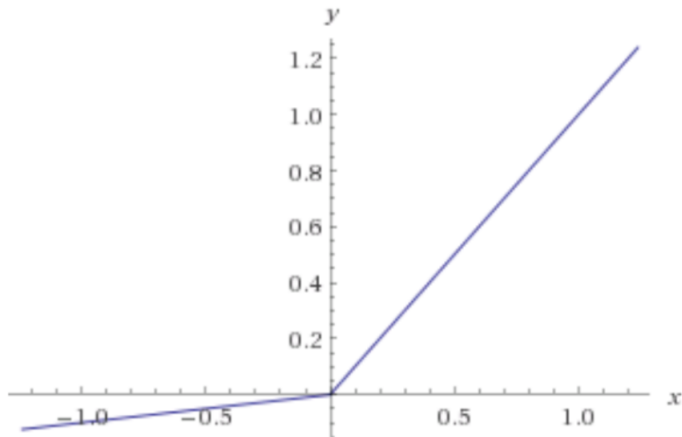
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4. Leaky ReLU: $LR(x) = \begin{cases} 0.01x, & \text{for } x < 0 \\ x, & \text{for } x \geq 0 \end{cases}$

Activation functions

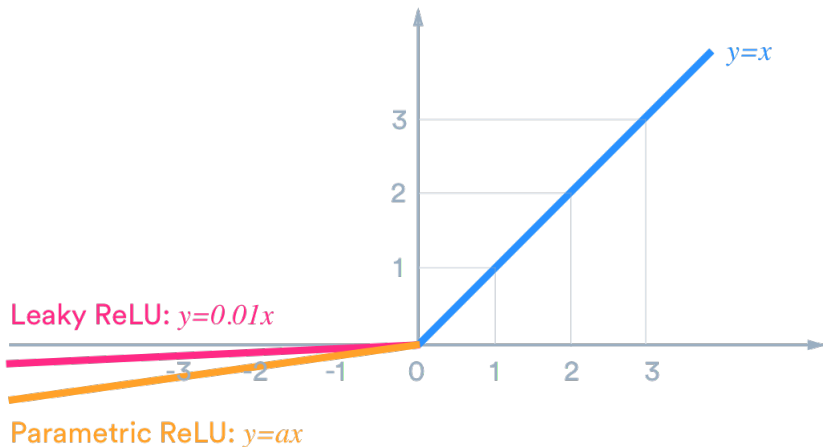
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5. Parametric ReLU: $PR(x) = \begin{cases} ax, & \text{for } x < 0 \\ x, & \text{for } x \geq 0 \end{cases}$

Activation functions

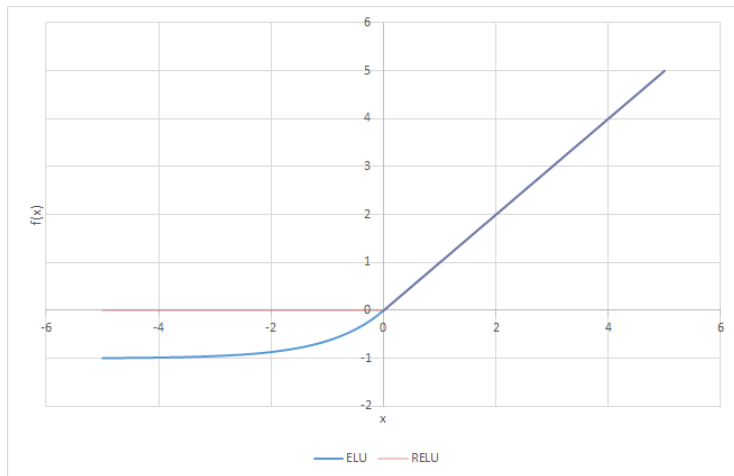
5. Parametric ReLU: $PR(x) = \begin{cases} ax, & \text{for } x < 0 \\ x, & \text{for } x \geq 0 \end{cases}$



6. Exponential linear unit: $ELU(x) = \begin{cases} a(e^x - 1), & \text{for } x < 0 \\ x, & \text{for } x \geq 0 \end{cases}$

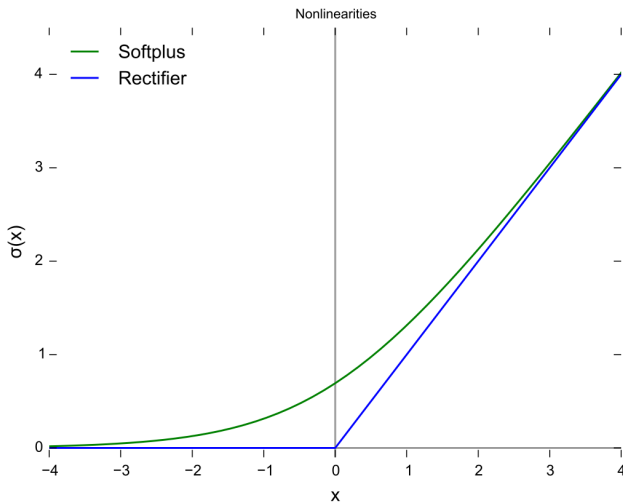
Activation functions

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Activation functions

7. SoftPlus: $SP(x) = \log(1 + e^x)$



8. Softmax: $S(x_1, x_2, \dots, x_n) = \left(\frac{e^{x_1}}{\sum_{i=1}^n e^{x_i}}, \frac{e^{x_2}}{\sum_{i=1}^n e^{x_i}}, \dots, \frac{e^{x_n}}{\sum_{i=1}^n e^{x_i}} \right)$

- 1 Why do we need activation functions?

- 1 Why do we need activation functions?
- 2 How should we define activation functions, for a layer or for a neuron?

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Gradient Descent

Let $f : \mathbb{R}^k \rightarrow \mathbb{R}$ be a convex function and we want to find its global minimum.

Gradient Descent

Let $f : \mathbb{R}^k \rightarrow \mathbb{R}$ be a convex function and we want to find its global minimum. This optimization algorithm is based on the fact that the fastest decreasing direction of the function is the opposite direction of gradient:

$$x_{n+1} = x_n - \alpha \nabla f(x_n)$$

and $x_0 \in \mathbb{R}^k$ is a arbitrary point.