


# Introduction to Deep Learning

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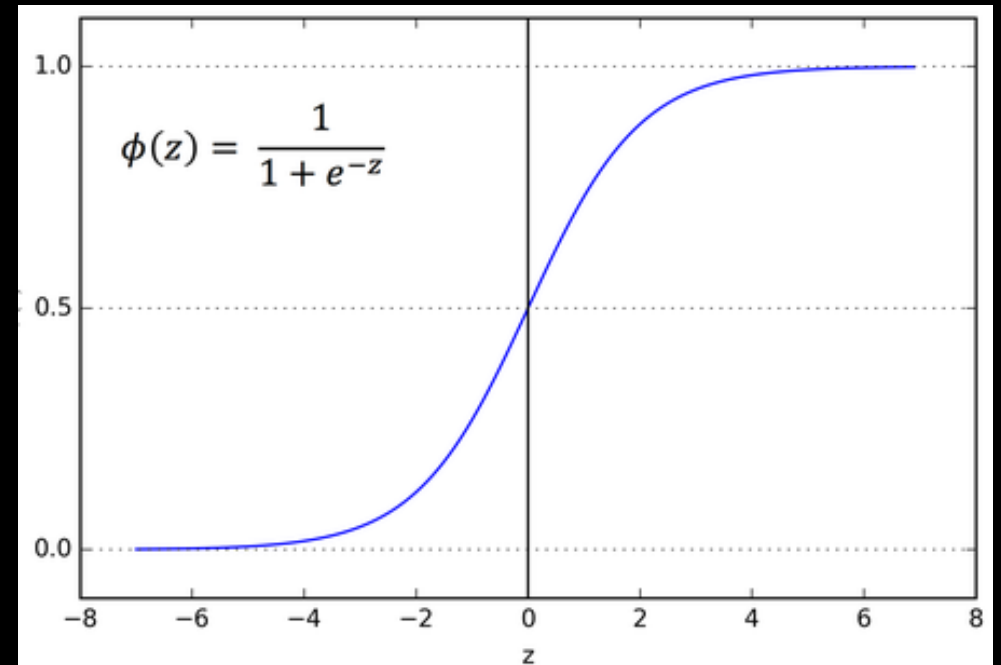
 @oaugereau

# Activation functions

Defines the output of a neuron given a set of inputs.

Constraints:

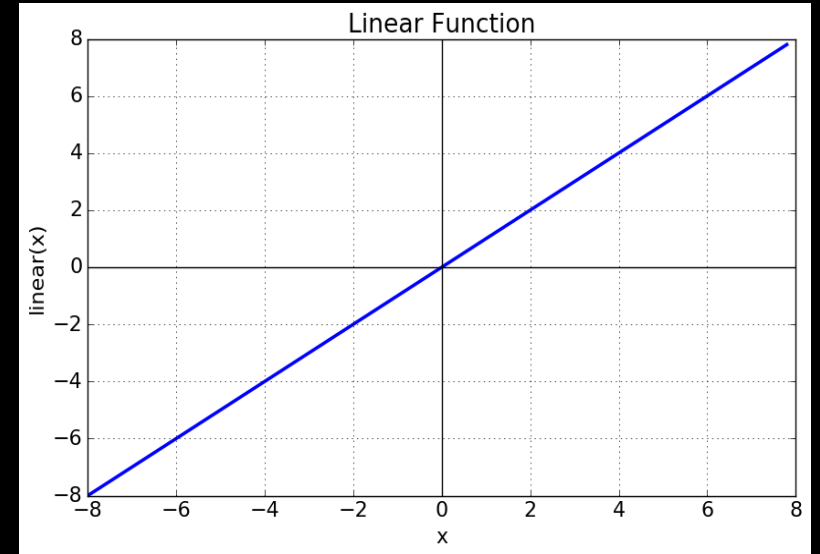
- continuous and infinite in domain,
- monotonic,
- nonlinear.



sigmoid

# Linear Activation functions

- a neural network with only linear activations performs as well as a linear regression
- backpropagation cannot be applied to find how to change the neural weights based on the errors found: the gradient is a constant and do not depend on the input values

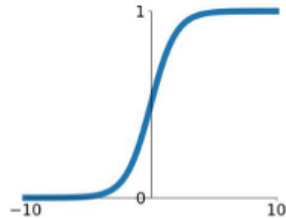


# Activation functions

## Activation Functions

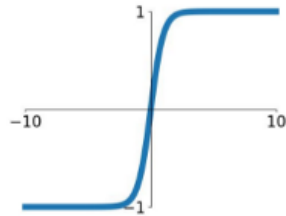
### Sigmoid

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



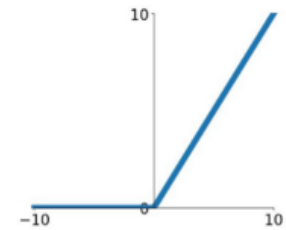
### tanh

$$\tanh(x)$$



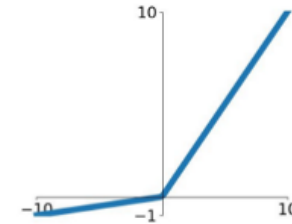
### ReLU

$$\max(0, x)$$



### Leaky ReLU

$$\max(0.1x, x)$$

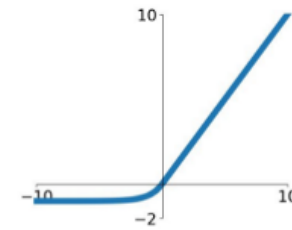


### Maxout

$$\max(w_1^T x + b_1, w_2^T x + b_2)$$

### ELU

$$\begin{cases} x & x \geq 0 \\ \alpha(e^x - 1) & x < 0 \end{cases}$$



# Sigmoid, Softmax et ReLU

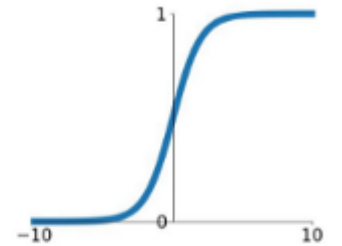
- smooth gradient, output values bound between 0 and 1.
- for high & low values of X, there is almost no change to the prediction, computationally expensive.

(softmax for multi classes)

- computationally efficient, allows the network to converge quickly.
- Pb of « dying » neuron if often  $<0$

## Sigmoid

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



## ReLU

$$\max(0, x)$$

