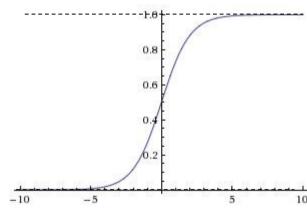
$$f(z) = 1 - \frac{1}{1 + e^{-x}}$$

Sigmoid

Complies with the interpretation of a firing neuron, between zero and one

Saturates and vanishes gradients

Outputs are not zero-centered



http://cs231n.github.io/neural-networks-1/

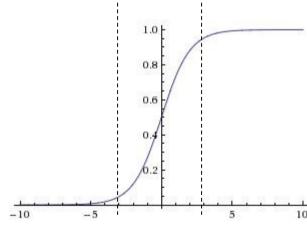
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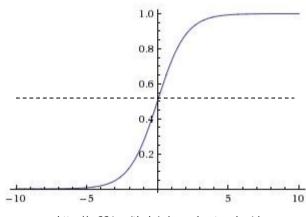
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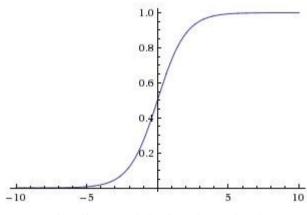
Sigmoid

Complies with the interpretation of a firing neuron, between zero and one

Saturates and vanishes gradients

Outputs are not zero-centered

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



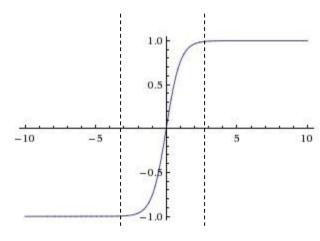
http://cs231n.github.io/neural-networks-1/

$f(z) = 1 - \frac{2}{e^{2z} + 1}$

Tanh

Saturates and vanishes gradients

Outputs are zero-centered in a range between minus one and one



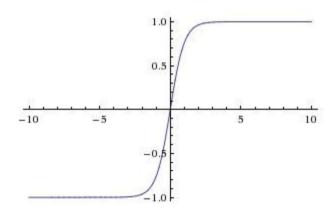
http://cs231n.github.io/neural-networks-1/

Tanh

Saturates and vanishes gradients

Outputs are zero-centered in a range between minus one and one

$$f(z) = 1 - \frac{2}{e^{2z} + 1}$$



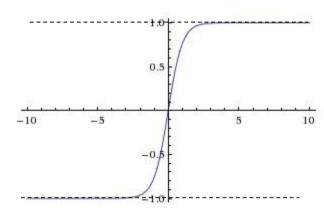
http://cs231n.github.io/neural-networks-1/

Tanh

Saturates and vanishes gradients

Outputs are zero-centered in a range between minus one and one

$$f(z) = 1 - \frac{2}{e^{2z} + 1}$$
$$f'(z) = 1 - f(z)^2$$



http://cs231n.github.io/neural-networks-1/

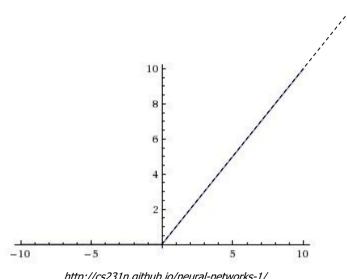
ReLU (Rectified Linear Unit)

Does not saturate in the positive domain and thus the gradients do not vanish in the positive direction and learning is accelerated

Cheap operation of thresholding at zero

ReLUs can be fragile and "die" during training when the weights are updated too far into the negative domain. Fixed by leaky ReLU

$$f(z) = \max\{0, z\}$$



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$$f(z) = \max\{0, z\}$$

$$f'(z) = \begin{cases} 0, & \text{if } z < 0. \\ 1, & \text{if } z > 0. \end{cases}$$

http://cs231n.github.io/neural-networks-1/

ReLU (Rectified Linear Unit)

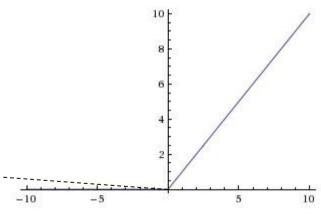
Does not saturate and thus the gradients do not vanish and learning is accelerated

Cheap operation of thresholding at zero

ReLUs can be fragile and "die" during training when the weights are updated too far into the negative domain. Fixed by leaky ReLUs and an adjusted learning rate.

$$f(z) = \max\{0, z\}$$

$$f'(z) = \begin{cases} 0, & \text{if } z < 0. \\ 1, & \text{if } z > 0. \end{cases}$$



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