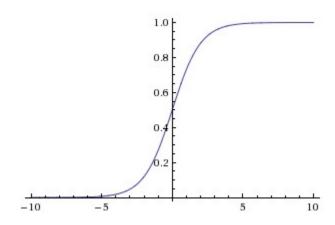
# **Activation Functions**

## **Sigmoid**

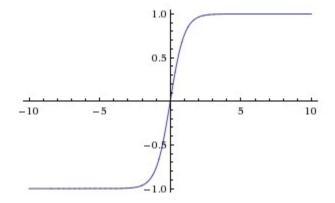


Ranges from (0, 1)

### **Disadvantages**

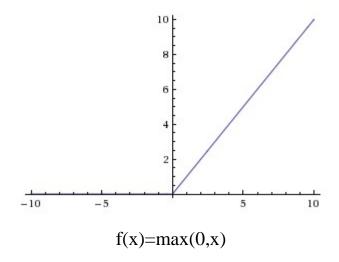
- 1) Saturates and Kills gradients
  - a. If activation values reaches the tails of zero or one, then the gradient becomes very close to zero.
  - b. If the gradient value is close to zero, the learning is very low and stops eventually.
- 2) Not Zero centered
  - a. This causes all the weights to be either +ve or -ve
  - b. Can cause over fitting/under fitting

#### TanH



Ranges from (-1, 1). It is zero-centered. Thus, it is preferred over sigmoid.

### **RELU (REctified Linear Unit)**



#### **Advantages:**

- 1) Faster convergence of stochastic gradient descent compared to the sigmoid/tanh functions.
- 2) Less expensive operations compared to sigmoid/tanh.

### **Disadvantages:**

1) In RELU, A large gradient can update weight of a neuron in a way that it can never be activated again. It is said that if the learning rate is high, 40% of the network can be dead (never activated again).

### Reference

• <a href="http://cs231n.github.io/neural-networks-1/">http://cs231n.github.io/neural-networks-1/</a>