

Tanh

- **Introduction:**

Tanh, short for hyperbolic tangent, is a commonly used activation function in **neural networks**. It is a scaled version of the logistic sigmoid function and is popular for its ability to squish input values to the **range** $[-1, 1]$, allowing for easier training of deep neural networks.

The **Tanh** activation function is defined as:

$$\text{Tanh}(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}}$$

*** Characteristics:**

- **Range:** Tanh squashes input values to the range $[-1, 1]$, which helps in centering the activations around zero.
- **Symmetry:** Tanh function is symmetric around the origin, which can aid in training deep networks by making the outputs zero-centered.
- **Smoothness:** Tanh is a smooth, continuous function that is differentiable everywhere.
- **Sensitivity to Input:** Tanh is more sensitive to small changes in input compared to the sigmoid function.

*** Advantages:**

1. **Zero Centering:** Tanh helps center the output activations around zero, which can help in optimizing the neural network.
2. **Gradient Saturation:** Tanh mitigates the vanishing gradient problem better than sigmoid by mapping the input values to a wider range.

* Disadvantages:

1. *Vanishing Gradient:* Tanh is still susceptible to the vanishing gradient problem, especially for deeper networks.
2. *Computation Cost:* Tanh function involves exponential operations, which can be computationally expensive compared to functions like ReLU.

- Conclusion:

Tanh **activation function** is a valuable tool in **deep learning** due to its ability to squash input values to a range that aids in optimizing neural networks. While it has advantages such as **zero-centering** and alleviating gradient saturation to some extent, it is important to consider its potential drawbacks like vanishing gradients and computational cost when choosing it for a model architecture.

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