

1.2.3

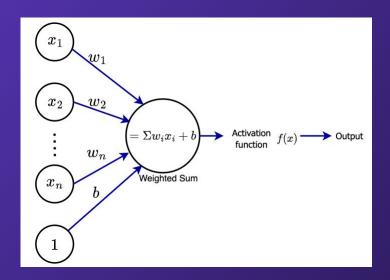
Introduction to Deep Learning

Activation Functions



What are Activation Functions?

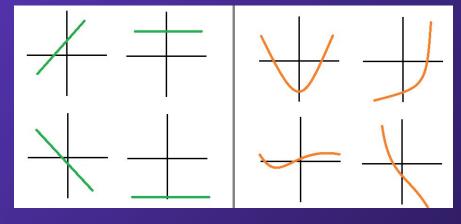
- Activation functions are mathematical functions applied to the output of a neuron (or node) in a neural network.
- They determine whether a neuron should 'fire' (activate), and to what degree, based on the weighted sum of its inputs.
- This non-linear behavior empowers neural networks to learn complex patterns that would be impossible with linear models.





Why Non-linearity Matters

- Most real-world data and relationships do not follow simple linear patterns.
- Linear models are restricted in their ability to learn complex patterns.
- Non-linear activation functions allow neural networks to approximate nearly any function, making them universal approximators.



Linear Functions

Non-linear Functions



Common Activation Functions

| Sigmoid | Tanh | ReLU | Leaky ReLU |
|---------------------------------------------------------------------------------|-------------------------------------|--------------------|----------------------------------------------------|
| $g(z) = \frac{1}{1+e^{-z}}$ | $g(z)=rac{e^z-e^{-z}}{e^z+e^{-z}}$ | $g(z) = \max(0,z)$ | $g(z) = \max(\epsilon z, z)$ with $\epsilon \ll 1$ |
| $\begin{array}{c c} 1 \\ \hline \\ \frac{1}{2} \\ \hline \\ -4 & 0 \end{array}$ | | | |



Choosing an Activation Function

- Task: The type of problem (classification, regression) influences the choice.
- Output Layer: Sigmoid for binary classification, softmax for multi-class, linear for regression
- Hidden Layers: ReLU is a good default, experiment with Leaky ReLU, explore others if needed.
- Experimentation & Research: The field evolves, and best practices change, so stay updated.



Activation Functions

