



December 6, 2025

Learning Goal

Understand how biological neurons inspire the mathematical structure of artificial neurons.

This slide establishes the learning objective for this topic

Key Concept (1/3)

The human brain contains approximately 86 billion neurons, each connected to thousands of others. These biological neurons receive signals through tree-like structures called **dendrites**, process them in the cell body (**soma**), and transmit outputs through a long fiber called the **axon** to other neurons via **synapses**.

Understanding this concept is crucial for neural network fundamentals

Key Concept (2/3)

Artificial neurons mimic this process mathematically. Instead of electrical signals traveling through dendrites, we have numerical inputs. Instead of synaptic strengths that change with learning, we have **weights** that adjust during training. The soma's signal integration becomes a **weighted sum**, and the axon's firing decision becomes an **activation function**.

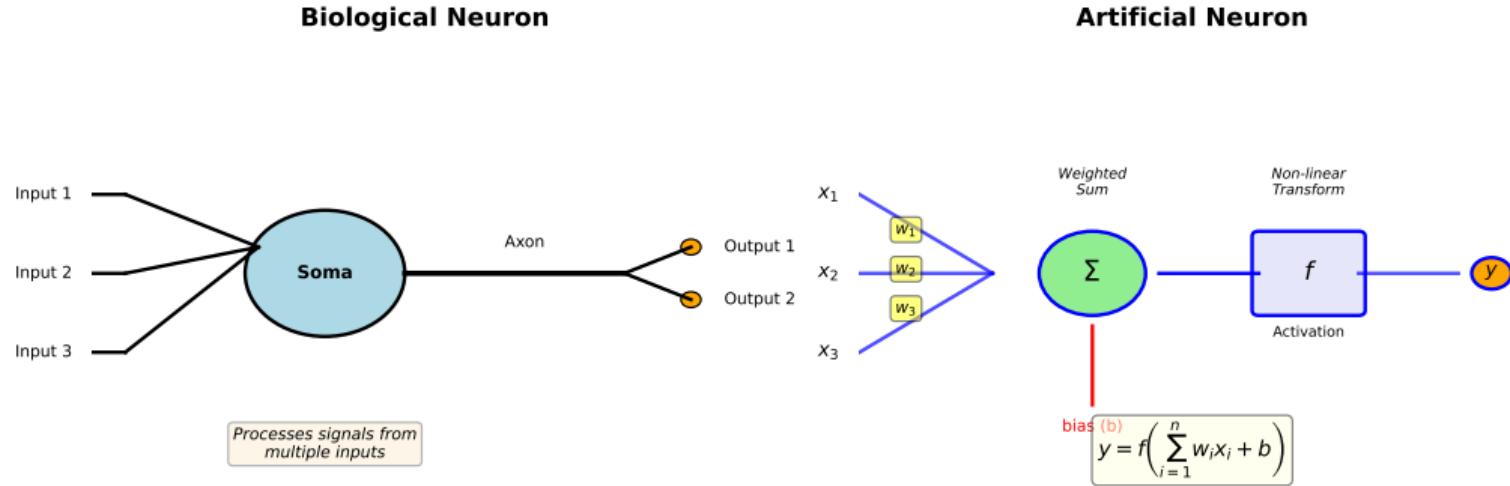
Understanding this concept is crucial for neural network fundamentals

Key Concept (3/3)

This biological analogy is not just poetic - it guided the development of neural networks and helps us understand why they work. Just as biological neurons strengthen connections that lead to successful outcomes, artificial neurons adjust their weights to minimize prediction errors.

Understanding this concept is crucial for neural network fundamentals

From Biology to Artificial Intelligence



Visual representations help solidify abstract concepts

The artificial neuron computes:

$$y = f \left(\sum_{i=1}^n w_i x_i + b \right)$$

Where: - x_i = input values (like dendrite signals) - w_i = weights (like synaptic strengths) - b = bias (baseline activation threshold) - f = activation function (like the firing decision) - y = output (like the axon signal)

Mathematical formalization provides precision

Think of a neuron as a voting system. Each input casts a vote (x_i), but not all votes are equal - some have more influence (w_i). The neuron tallies the weighted votes, adds a baseline preference (b), and then decides whether to "fire" based on the total.

If the weighted sum exceeds a threshold, the neuron produces a strong output. If it falls below, the output is weak. This simple mechanism, repeated across millions of neurons, creates the complex behaviors we see in neural networks.

Intuitive explanations bridge theory and practice

Practice Problem 1

Problem 1

A neuron receives three inputs: $x_1 = 0.5$, $x_2 = 0.8$, $x_3 = 0.2$. The weights are $w_1 = 0.4$, $w_2 = 0.3$, $w_3 = 0.5$, and the bias is $b = -0.1$. Calculate the weighted sum z .

Solution

$$\begin{aligned} z &= w_1x_1 + w_2x_2 + w_3x_3 + b \\ z &= (0.4)(0.5) + (0.3)(0.8) + (0.5)(0.2) + (-0.1) \\ z &= 0.20 + 0.24 + 0.10 - 0.10 \\ z &= 0.44 \end{aligned}$$

Practice problems reinforce understanding

Practice Problem 2

Problem 2

In the biological neuron, what structure is analogous to the weights in an artificial neuron? Explain why this analogy makes sense.

Solution

The **synaptic strengths** are analogous to weights. This makes sense because:

- Synapses can be strong or weak, just like weights can be large or small
- Synaptic strengths change with learning (long-term potentiation/depression)
- A strong synapse means that input has more influence on the neuron's output
- Both determine how much each input contributes to the final decision

Practice problems reinforce understanding

- Biological neurons inspire artificial neuron design
- Dendrites - \downarrow Inputs, Synapses - \downarrow Weights, Soma - \downarrow Summation, Axon - \downarrow Output
- The weighted sum aggregates all inputs before the activation decision
- Learning adjusts weights, similar to how synapses strengthen or weaken

These key points summarize the essential learnings