

Neural network

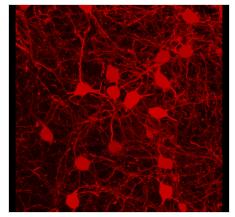
A **neural network** is a group of interconnected units called <u>neurons</u> that send signals to one another. Neurons can be either <u>biological cells</u> or <u>mathematical models</u>. While individual neurons are simple, many of them together in a network can perform complex tasks. There are two main types of neural network.

- In neuroscience, a *biological neural network* is a physical structure found in <u>brains</u> and complex nervous systems a population of nerve cells connected by synapses.
- In machine learning, an *artificial neural network* is a mathematical model used to approximate nonlinear functions. Artificial neural networks are used to solve artificial intelligence problems.

In biology

In the context of biology, a neural network is a population of biological <u>neurons</u> chemically connected to each other by <u>synapses</u>. A given neuron can be connected to hundreds of thousands of synapses. [1] Each neuron sends and receives <u>electrochemical</u> signals called <u>action potentials</u> to its connected neighbors. A neuron can serve an <u>excitatory</u> role, amplifying and propagating signals it receives, or an <u>inhibitory</u> role, suppressing signals instead. [1]

Populations of interconnected neurons that are smaller than neural networks are called <u>neural circuits</u>. Very large interconnected networks are called <u>large scale brain networks</u>, and many of these together form <u>brains</u> and nervous systems.



Animated <u>confocal micrograph</u> of part of a biological neural network in a mouse's striatum

Signals generated by neural networks in the brain eventually travel through the nervous system and across <u>neuromuscular junctions</u> to <u>muscle cells</u>, where they cause contraction and thereby motion. [2]

In machine learning

In machine learning, a neural network is an artificial mathematical model used to approximate nonlinear functions. While early artificial neural networks were physical machines, [3] today they are almost always implemented in software.

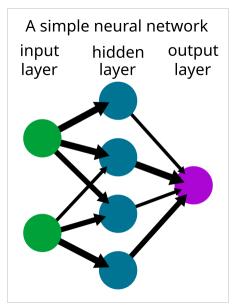
Neurons in an artificial neural network are usually arranged into layers, with information passing from the first layer (the input layer) through one or more intermediate layers (the hidden layers) to the final layer (the output layer). The "signal" input to each neuron is a number, specifically a linear combination of the outputs of the connected neurons in the previous layer. The signal each neuron outputs is calculated from this number, according to its activation function. The behavior of the

network depends on the strengths (or *weights*) of the connections between neurons. A network is trained by modifying these weights through <u>empirical risk minimization</u> or <u>backpropagation</u> in order to fit some preexisting dataset. [5]

Neural networks are used to solve problems in <u>artificial</u> <u>intelligence</u>, and have thereby found applications in many disciplines, including predictive modeling, <u>adaptive control</u>, <u>facial</u> recognition, handwriting recognition, general game playing, and generative AI.

History

The theoretical base for contemporary neural networks was independently proposed by <u>Alexander Bain</u> in 1873^[6] and <u>William James</u> in 1890.^[7] Both posited that human thought emerged from interactions among large numbers of neurons inside the brain. In 1949, Donald Hebb described Hebbian learning, the idea that



Schematic of a simple feedforward artificial neural network

neural networks can change and learn over time by strengthening a synapse every time a signal travels along it. [8]

Artificial neural networks were originally used to model biological neural networks starting in the 1930s under the approach of connectionism. However, starting with the invention of the perceptron, a simple artificial neural network, by Warren McCulloch and Walter Pitts in 1943, [9] followed by the implementation of one in hardware by Frank Rosenblatt in 1957, [3] artificial neural networks became increasingly used for machine learning applications instead, and increasingly different from their biological counterparts.

See also

- Emergence
- Biological cybernetics
- Biologically-inspired computing

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