

# *The Basics of Neural Networks*

**TM Quest**

# Overview

## What Will we Learn in This Module?

- What are **neural networks**?
  - The idea behind neural networks.
  - What do we mean when we talk about **deep** neural networks?
- What are **activation functions** and **weights**?
  - How neural networks update their behavior.
  - How activation functions make neural networks interesting.

*What are neural networks?*

# Motivation

## Conceptual Description

An (artificial) neural network is based on a collection of connected units or nodes called *artificial neurons*, which loosely model the neurons in a biological brain.

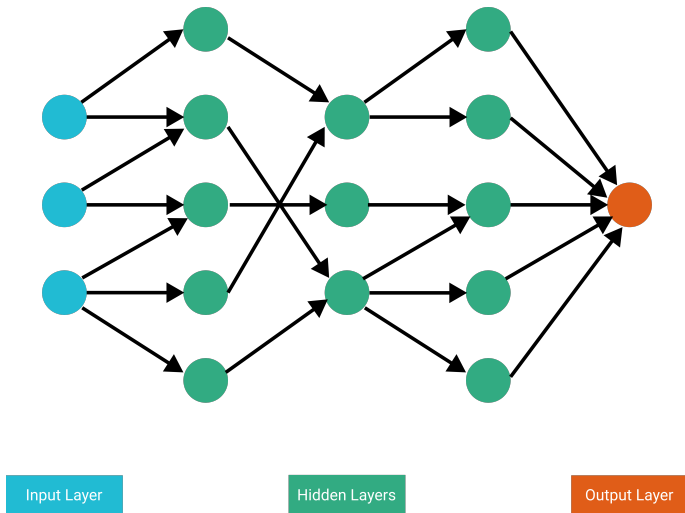
## Recent Popularity

Neural nets have become very popular in recent years do to the increase in data volume. They perform well on large amounts of data.

## In Practice?

Neural nets are still just machine learning models (for regression and classification) that combine simple pieces into something complicated.

# *Overview of a Neural Network*



# *Weights and activation functions*

# Weights in Neural Networks

## Weighted Input

Each node after the input layer takes its input from the previous layer. It uses **weights** to put different emphasis on the input based on their importance.

## Example

If the input to a node is numbers  $x$  and  $y$ , then the node create the weighted average

$$w_1x + w_2y + b,$$

where  $w_1$ ,  $w_2$  are weights and  $b$  is a **bias term**. It is the weights  $w_1$ ,  $w_2$  and the bias term  $b$  that are the **parameters** of the neural network.

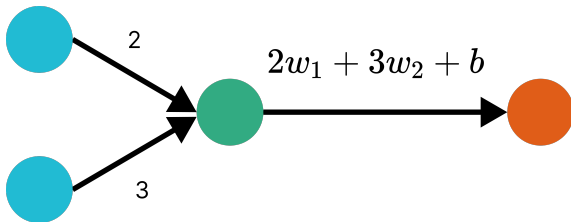
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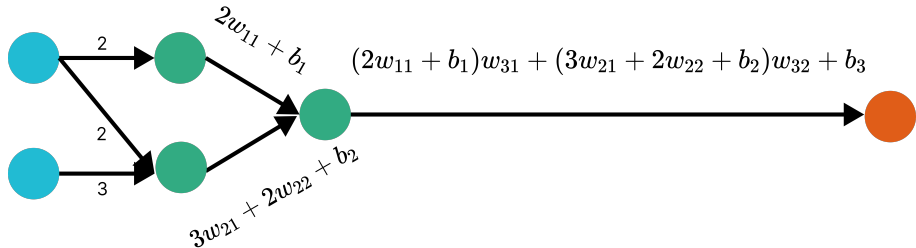
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# Why are Weights not Enough?



## *Summary of the Problem*

### Problems

- By only using weights (and bias terms) in all the nodes, we are creating linear functions.
- The composition (passing the output of one function into the next one) of linear functions is still linear.
- Deep neural networks for classification in this way can only generate linear decision boundaries.

We need to introduce something that makes the neural nets "non-linear"!

# Activation Functions

## Motivation

We introduce **activation function** between the nodes that introduce non-linearity. Activation functions take the weighted combination from a node and apply a special function to spice things up.

## Example

The **sigmoid function**

$$g(x) = \frac{1}{1 + e^{-x}}$$

is an example of an activation function that is used in practice.

## *In Action*

### Note!

The activation functions are chosen before the training begins. Hence the weights and bias terms are the only parameter that the model trains on.

