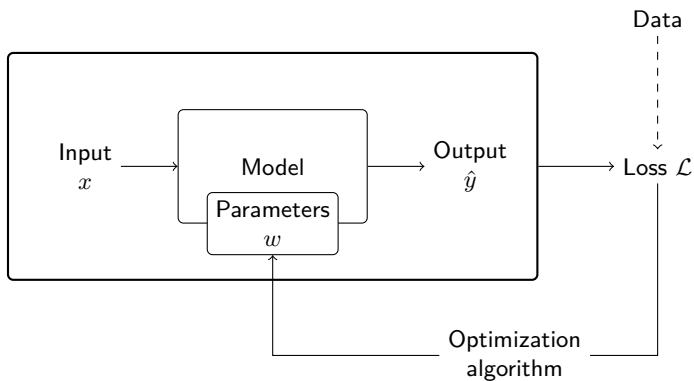


# Introduction to deep learning

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## General Idea



# Summary

① Toy examples

② Gradient descent

③ Neural networks

④ Complete example

## Toy examples

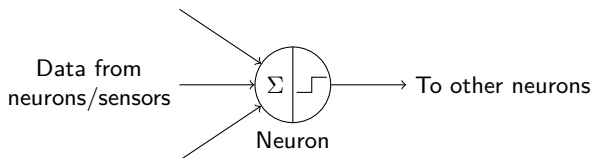
cf Jupyter notebook.

# Gradient descent

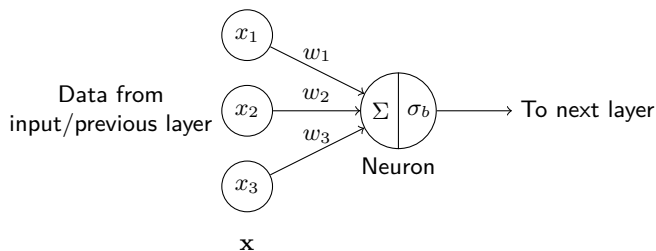
cf Jupyter notebook.

# Neural networks

Neural networks are "vaguely inspired by the biological neural networks that constitute animal brains". (Wikipedia)



In machine learning:



Operation implemented:

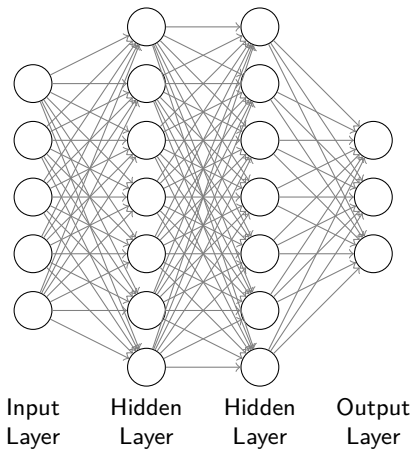
$$\mathbf{x} \mapsto \sigma(\mathbf{w} \cdot \mathbf{x} + b), \quad \mathbf{w} \text{ weights, } b \text{ bias}$$

$\sigma$  is called the activation function. It is a smooth non-linear function, for instance the sigmoid:

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

## Neural network

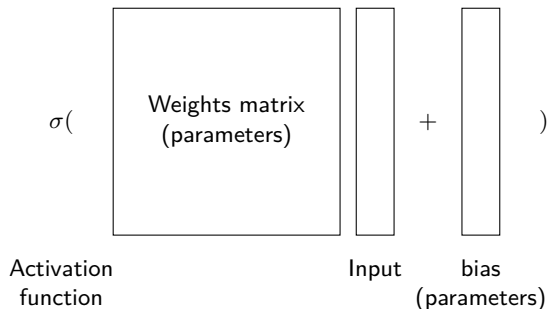
A complete neural network looks like this:





## Hidden layers

Each hidden layer performs a linear transformation followed by a non-linear activation function:



Such a layer is called *dense* (by opposition to e.g a convolutional layer). A network with only dense layers is called a multilayer perceptron (MLP).

# Backpropagation

For the gradient descent algorithm to work, we have to compute the gradient of the loss with respect to each parameter of the neural network.

As a neural network is a composition of functions, the gradient of the loss with respect to the parameters is computed using the chain rule.

This computation goes from right to left, hence the name *backpropagation*.

In practice, the computation of the gradient is performed by automatic differentiation.

## Complete example

cf Jupyter notebook.