#### **Foundations of Artificial Neural Networks**

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## What are artifical neural networks (ANN)?

- A new paradigm of computing.
- A simplified implementation of biologically inspired neural networks.

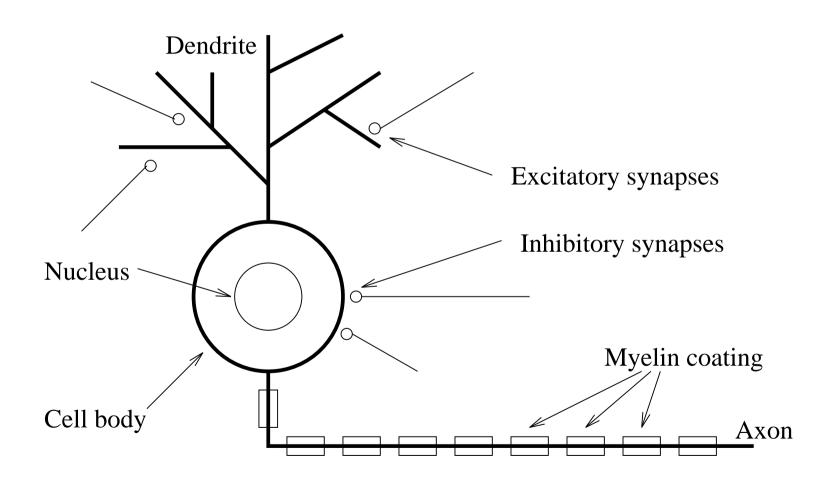
#### Where ANNs are better than traditional computers

- Adapting to new environments by *learning*.
- Processing *fuzzy* (imprecise) data.
- Working with noisy or erroneous data.
- Performing classification tasks very quickly.

#### **Biological neural networks – the basics**

- A human brain contains about  $10^{11}$  neurons.
- Each neuron has inputs (synapses connected to *dendrites*) and an output (*axon* ending in synapses).
- $\bullet$  One neuron generates input for  $\approx 10^3 \dots 10^4$  other neurons.
- **④** Each neuron receives data from  $\approx 10^3 \dots 10^4$  neurons.
- Synapses connect neurons to each other.
- G Signals are transmitted by releasing chemicals (*neurotransmitters*) from a sender neuron causing an electric potential in the receiving cell, etc.

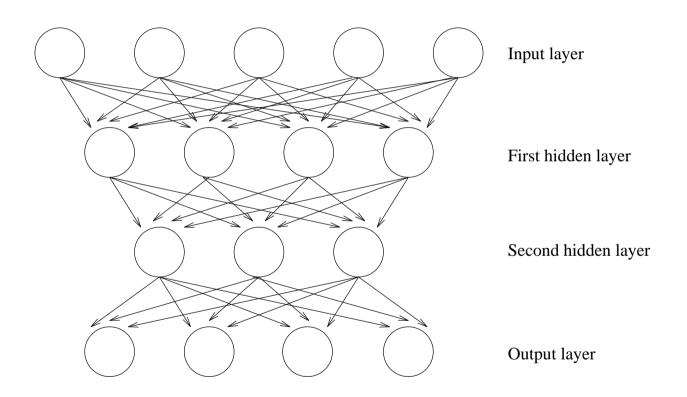
# **Biological neurons**



#### Simplified model of a neuron

- A neuron has a current *activation*  $a_j(t)$ .
- **2** All inputs get numeric data  $o_i(t)$  from other neurons.
- $\bullet$  Inputs are multiplied by *weights*  $w_{i,j}$ .
- **④** Generate a *net input* by employing a *propagation rule*:  $n_j(t) = \sum_i o_i(t) w_{i,j}.$
- **6** Calculated new activation ( $\Theta$  is a threshold value)  $a_j(t+1) = f_{act} (a_j(t), n_j(t), \Theta_j).$
- **6** A new output is calculated using an output function:  $o_j(t+1) = f_{out}(a_j(t+1)).$

## A simple feed forward ANN



## Basic types of ANNs

- Feed forward ANNs
  - Several layers of neurons are connected unidirectionally.
  - Connections can skip layers.
- Peedback ANNs
  - Direct feedback networks allow a neuron to have a feedback connection to itself  $\rightarrow$  self amplification/attenuation.
  - Indirect feedback networks have only feedback connections between layers  $\rightarrow$  a method to control attention of the ANN.
  - Lateral feedback networks have feedback connections only within a layer  $\rightarrow$  winner takes all network.

# Implementing ANNs

• Synchronous activation: All neurons change their states simultaneously  $\rightarrow$  very useful for SIMD parallel computers.

**2** Asynchronous activation:

*Fixed order activation*: Order of activation is fixed.

**Random order activation:** Order is (pseudo)random – not every neuron is calculated, some are calculated multiply.

**Random permutation activation:** As before but every neuron is calcuted once and only once.

Topological order activation: Ideal for feedforward networks.

# Training (teaching) ANNs by

• creating new connections/neurons,

deleting existing connections/neurons,

- $\boldsymbol{\Theta}$  changing the weights  $w_{i,j}$ ,
- $\bullet$  changing the threshold  $\Theta_j$  or by
- **6** changing the functions  $f_{act}()$  and/or  $f_{out}()$ .

In most cases, the weights  $w_{i,j}$  will be modified – this can also be used to simulate additional or deleted connections/neurons.

## **Basic learning strategies**

- Supervised learning: The learning algorithm receives the output generated by the ANN as well as the desired output and performs changes on the weights  $w_{i,j}$ .  $\rightarrow$  Desired result has to be known in advance, sometimes not possible/realistic.
- Reinforcement learning: The ANN receives results in form of reward/punishment from actions based on its generated output.
- Our Consupervised learning: The only goal is to build representations of the input data which can in turn be used for prediction/classification, etc.

## The backpropagation learning rule

**Idea:** Change the weights  $w_{i,j}$  according to the difference between the actual output  $o_j(t)$  and the teaching value  $t_j(t)$  with a somewhat complex mechanism as follows:

$$\Delta w_{i,j} = \eta o_i \delta_j$$
 with

$$\delta_j = \begin{cases} f_{\rm act}(n_j(t))(t_j(t) - o_j(t)) \text{ for output neurons} \\ f_{\rm act}(n_j(t)) \sum_k (\delta_k w_{j,k}) \text{ for hidden neurons.} \end{cases}$$

## Input data sets

- Training set: Examples used to train the ANN (for example by changing the weights  $w_{i,j}$  according to the backpropagation rule).
- Validation set: Data set for tuning the ANN (for example by changing the number of hidden layers, etc.).
- Test set: Data set used to assess the performance of the ANN (obviously never to be used as a training set).

Sometimes the difference between the training and the validation set is not quite clear and depends on the applied learning rule.

#### References

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