NEURAL NETWORKS LECTURE 1

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AIMS

- Brief introduction to neural networks concepts and architectures
- Capabilities and possible applications
- Java Neural Network Framework-om Neuroph
- Recognize typical neural network problems and implement neural network based solutions
- Neuroph framework development

BRIEF OVERVIEW

Lectures

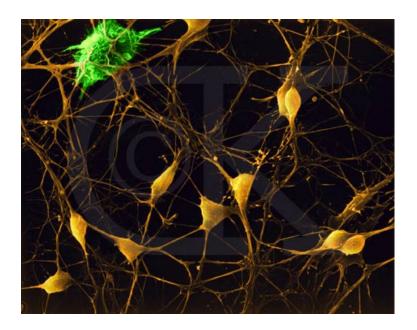
- 1. Basic concepts, architecture and components
- 2. Multi Layer Perceptron and Backpropagation
- 3. Problem solving using neural networks
- 4. Backpropagation improvements

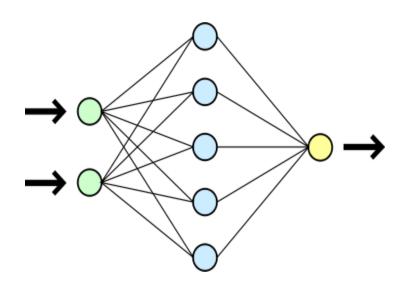
Labs

- 1. and 2. Neural network framewrok Neuroph and basic usage for image recognition and OCR
- 3. and 4. Neuroph framework architecture and extensions

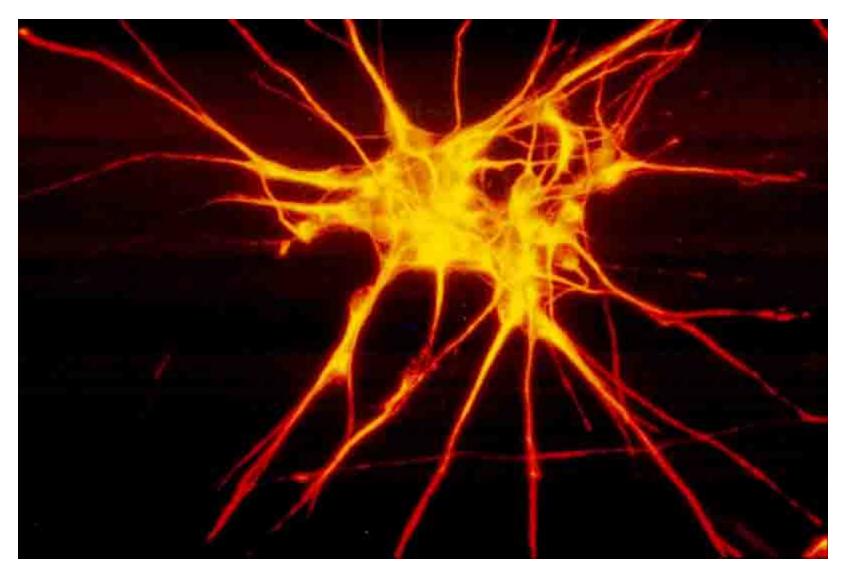
WHAT IS NEURAL NETWORK

- Brain inspired mathematical models
- Biological and artificial NN

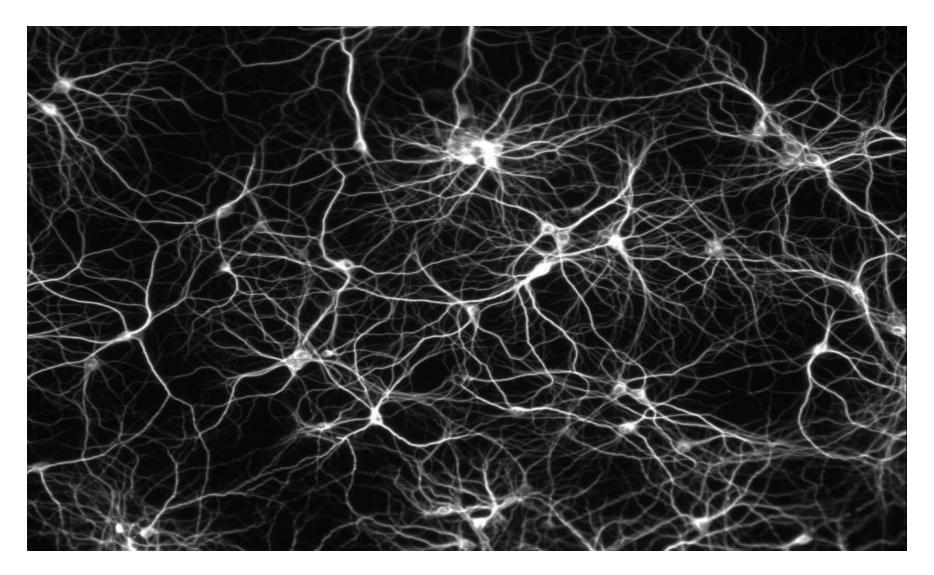




BIOLOGICAL NEURON



Biological Neural Network



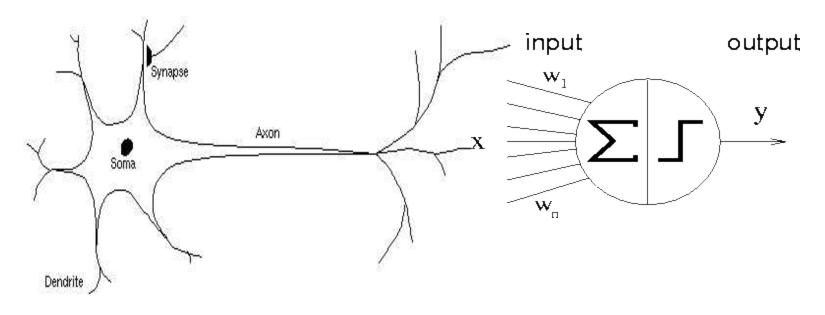
BRAIN AND ANN

- Brain:
 - 10¹⁰ neurons
 - 10¹³ connections
 - speed at milisec
 - Fully parallel

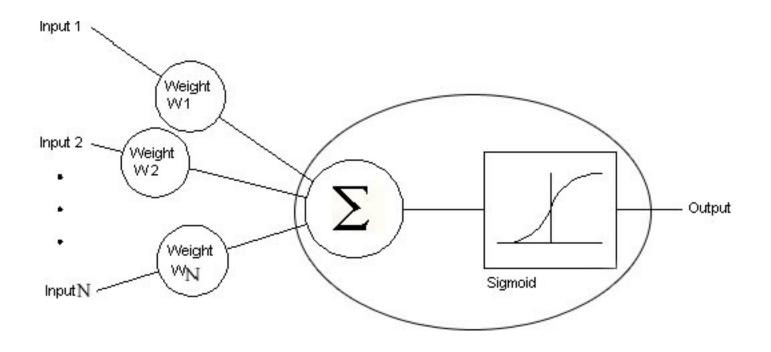
- ANN:
 - -20 000 neurons
 - Speed at nanosec
 - Simulating parallel computation

Biological and artificial neuron

 Basic parts: body(soma), dendrites(inputs), axons (outputs), synapses (connections)



Artificial Neuron



output = $f(w_1in_1 + ... + w_nin_n)$

Basic parts of artificial neuron

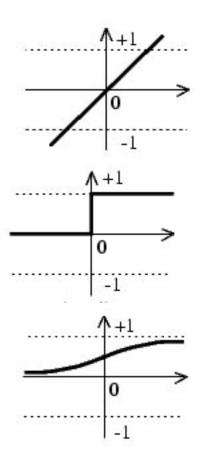
- Input summing function
- Transfer function
- Weighted inputs
- Output

TRANSFER FUNCTION

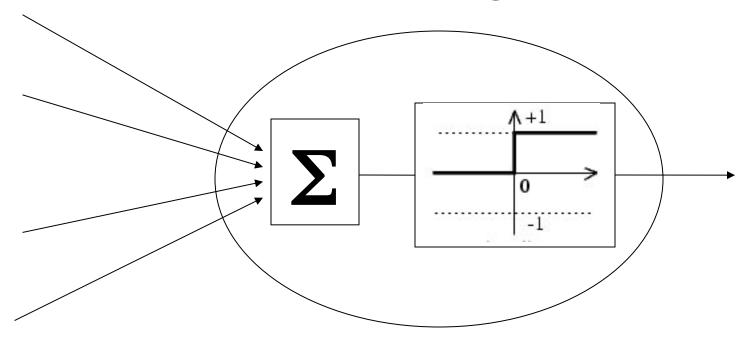
Linear

Step

Sigmoid

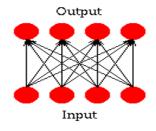


McCulloch Pits Neuron Threshold Logic Unit

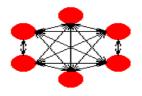


 $y = STEP(w_1u_1 + ... + w_nu_n)$

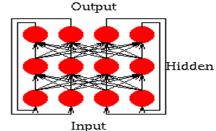
NEURAL NETWORK ARCHITECTURES



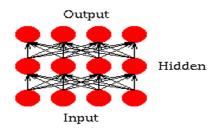
Single Layer Feedforward



Fully Recurrent Network

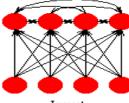


Jordan Network



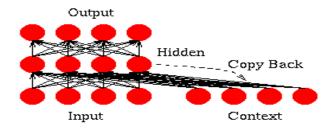
Multi Layer Feedforward

Output



Input

Competitive Network



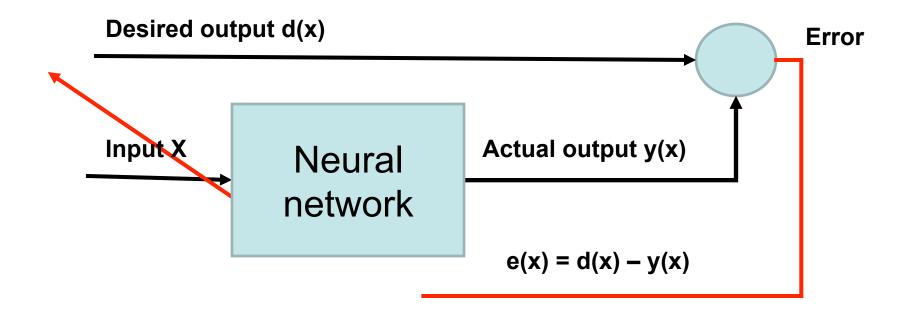
Simple Recurrent Network

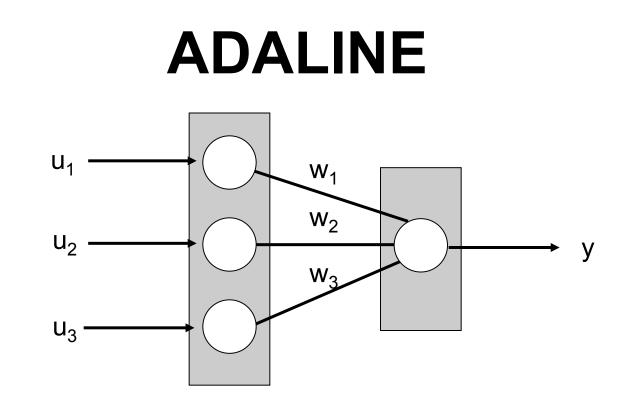
Neural network training/ learning

- Learning procedure: adjusting connection weights, until network gets desired behaviour
- Supervised Learning
- Unsupervised Learning

SUPERVISED LEARING

Basic principle: iterative error minimization





Linearna transfer function Linear combination of inputs

$$y = w_1 u_1 + w_2 u_2 + \dots + w_n u_n$$

Learning algorithm: Least Mean Squares

LMS LEARNING

LMS rule equations:

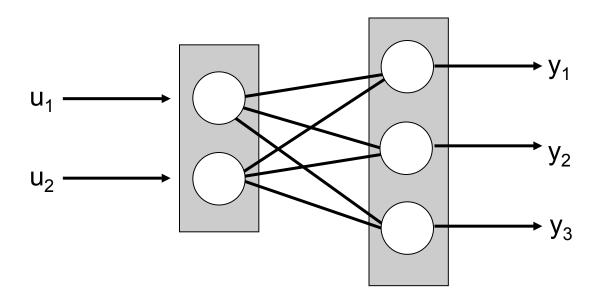
(1) output neuron error for pattern p

 $\varepsilon_{\rho}=d_{\rho}-y_{\rho}$

(2) weight change according to error

(3) total network error for all patterns from training set (training stops when this error is below some predefined value)

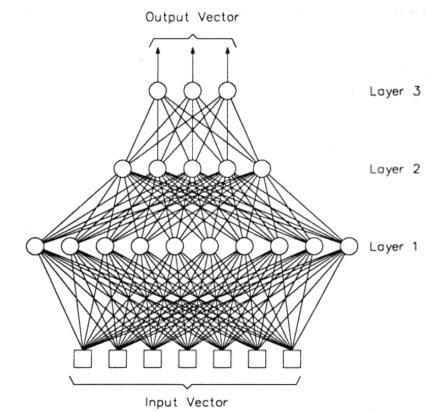
PERCEPTRON



- Step transfer function
- Perceptron learning first algorithm for learning nonlinear systems
- Only for linear separable problems

Multi Layer Perceptron

- Basic perceptrn extended with one or more layers of hidden neurons between input and output layer
- Differentiable neuron transfer functions (tanh, sigmoid)
- Using Backpropagation algorithm for learning, which is bassed on LMS algorithm
- Able to solve complex problems



Backpropagation algorithm

- For Multi Layer Perceptron training able to adjust weights in hidden layers
- Supervised learning algorithm based on LMS algorithm
- Multi Layer Perceptron with Backpropagation is universal aproximator

Backpropagation algorithm

Formula for adjusting hidden neurons

$$w_{ji}(k+1) = w_{ji}(k) + \mu f'(net_j(k)) \left(\sum_a \varepsilon_a(k) f'(net_a(k)) w_{aj}(k)\right) u_{ji}$$

Backpropagation settings

- Max error
- Max iterations
- Learning rate
- Momentum
- Batch mode

Backpropagation algoritam

- Local minimum
- Stopping conditions:
 - Max iterations
 - Max error
 - Error stall
 - -MSE for specific test set

LINKS AND BOOKS

- http://is.fon.rs/neuronske_mreze
- http://neuroph.sourceforge.net
- http://opensource.fon.bg.ac.rs

- Neural Networks A Systematic Introduction , free online book
- Introduction to Neural Computation