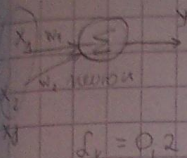


NEURONAL NETWORK

1. LMS ALGORITHM

učebná jednotka neuronu

$$W^{(k+1)} = W^{(k)} - \delta_k \cdot X_{d,k} - e_k$$



$$f(x) = 2x_1 - x_2$$

$$e_k = x_{d,k}^T \cdot W^{(k)} - y_{d,k}$$

	x_1	x_2	x_d	
$[0.5 \ 1.5]$	0.5	1.5	-0.5	0
	1	0.5	1.5	1
	1.5	1	2	2
	2	2	2	3

$$W_1^0 = 0$$

$$W_2^0 = 0$$

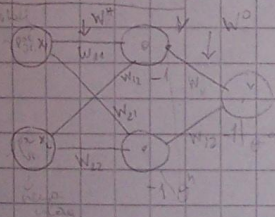
$$W = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$W^{(1)} = W^0 - \delta_0 \cdot X_{d,0} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} - 0.2 \cdot \begin{bmatrix} 0.5 \\ 1.5 \end{bmatrix} = \begin{bmatrix} -0.1 \\ -0.3 \end{bmatrix}$$

$$W^{(2)} = \begin{bmatrix} -0.05 \\ -0.15 \end{bmatrix} - 0.2 \cdot \begin{bmatrix} 1 \\ 0.5 \end{bmatrix} = \begin{bmatrix} -0.05 \\ -0.15 \end{bmatrix} + 1.3 \cdot \begin{bmatrix} 0.2 \\ 0.4 \end{bmatrix}$$

$W^5 \rightarrow$ uvažujeme $X_{d,0}$

učebná vrstva



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

$$\delta_k = \delta_k^{(h)} \cdot f'(v_k)$$

$$W = \begin{bmatrix} W_{11}^n & W_{12}^n \\ W_{21}^n & W_{22}^n \end{bmatrix}$$

$$W^0 = [W_{11}^0 \ W_{12}^0]$$

1. unapřední množina

2. back propagation

$$V = W^H \cdot X - \theta^h$$

$$z = \text{sigmoid}(v)$$

$$u = W^0 \cdot z + (1) \cdot \theta^0$$

$$y = \text{sigmoid}(u)$$

$$\delta^0 = (y - y_d) \cdot y \cdot (1 - y)$$

$$\delta^h = (W^0)^T \cdot \delta^0 \cdot z \cdot (1 - z)$$

$$W^0 = V^0 \cdot \eta \cdot \delta^0 \cdot z$$

$$\theta^0 = \theta^0 + \eta \cdot \delta^0$$

$$W^h = W^h + \eta \cdot \delta^h \cdot x$$

$$\theta^h = \theta^h + \eta \cdot \delta^h$$

x_1	x_2	y_d
0	0	0
0	1	1
1	0	1
1	1	0

$$W_0 = 0$$

$$W^h = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

$$\theta^0 = \begin{bmatrix} 0 & 0 \end{bmatrix}$$

$$\theta^h = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\theta^0 = \begin{bmatrix} 0 \end{bmatrix}$$

$$V = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \cdot \begin{bmatrix} 0 \\ 1 \end{bmatrix} - \begin{bmatrix} 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$z = \text{sig}(u) = \frac{1}{1 + e^{-u}} = \frac{1}{1 + e^{-1/2}} = \frac{1}{1/2}$$

$$u = W^0 z + (-1) \cdot A = \begin{bmatrix} 0 & 0 \end{bmatrix} \cdot \begin{bmatrix} 1/2 \\ 1/2 \end{bmatrix} - 0 = 0 \quad y = \text{sig}(u) = \frac{1}{2}$$

unfolding (back propagation)

$$\delta^0 = (y - y_d) \cdot y \cdot (1 - y) = \left(\frac{1}{2} - 0\right) \cdot \frac{1}{2} \cdot \left(1 - \frac{1}{2}\right) = \frac{1}{8}$$

$$\delta^h = \left((W^0)^T \cdot \delta^0 \right) \cdot z \cdot (1 - z) = \left(\begin{bmatrix} 0 \\ 0 \end{bmatrix} \cdot \frac{1}{8} \right) \cdot \begin{bmatrix} 1/2 \\ 1/2 \end{bmatrix} \cdot \begin{bmatrix} 1/2 \\ 1/2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$W_0^{(2)} = W_0 - \eta \cdot \delta^0 \cdot z^T = \begin{bmatrix} 0 & 0 \end{bmatrix} - 0.1 \cdot \frac{1}{8} \cdot \begin{bmatrix} 1/2 & 1/2 \end{bmatrix} = \begin{bmatrix} -\frac{1}{160} & -\frac{1}{160} \end{bmatrix}$$

$$\theta^{(2)} = \theta^0 - \eta \cdot \delta^0 \cdot (-1) = \begin{bmatrix} 0 \end{bmatrix} + 0.1 \cdot \frac{1}{8} = \frac{1}{80}$$

$$W^{(2)} = W^h - \eta \cdot \delta^h \cdot x_d = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} - \begin{bmatrix} 0 \\ 0 \end{bmatrix} \cdot \begin{bmatrix} 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

$$\theta^{(2)} = \theta^h - \eta \cdot \delta^h \cdot (-1) = \begin{bmatrix} 0 \\ 0 \end{bmatrix} - 0 \cdot \begin{bmatrix} 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

task 2

unaprior

$$\begin{matrix} x_1 & x_2 & y \\ 0 & 0 & 0 \\ 0 & 1 & 1 \end{matrix}$$

$$V = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \cdot \begin{bmatrix} 0 \\ 1 \end{bmatrix} - \begin{bmatrix} 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad z = \begin{bmatrix} 1/2 \\ 1/2 \end{bmatrix}$$

$$u = \begin{bmatrix} -\frac{1}{160} & -\frac{1}{160} \end{bmatrix} \cdot \begin{bmatrix} 1/2 \\ 1/2 \end{bmatrix} - \frac{1}{80} = A$$

$$y = \text{sig}(A) = \frac{1}{1 + e^{-A}}$$

back propagation

$$\delta^0 = (y - y_d) \cdot y \cdot (1 - y) = (\text{sig}(A) - 1) \cdot \text{sig}(A) \cdot (1 - \text{sig}(A)) = B$$

$$\delta^h = \begin{bmatrix} -\frac{1}{160} \\ -\frac{1}{160} \end{bmatrix} \cdot \delta^0 \cdot \begin{bmatrix} 1/2 \\ 1/2 \end{bmatrix} \cdot \begin{bmatrix} 1/2 \\ 1/2 \end{bmatrix} = \begin{bmatrix} D \\ E \end{bmatrix}$$

$$W^{(2)} = W^{(2)} - \eta \cdot \delta^0 \cdot z^T = \begin{bmatrix} -\frac{1}{160} & -\frac{1}{160} \end{bmatrix} - \delta^0 \cdot \begin{bmatrix} 1/2 & 1/2 \end{bmatrix}$$

$$\theta^{(2)}$$

$$W^{(2)}$$

$$\theta^{(2)}$$

$$\begin{matrix} x_1 & x_2 & y \\ 0 & 0 & 0 \\ 0 & 1 & 1 \end{matrix}$$

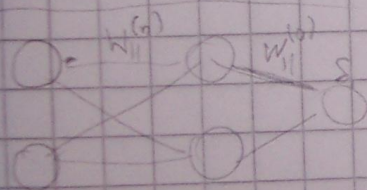
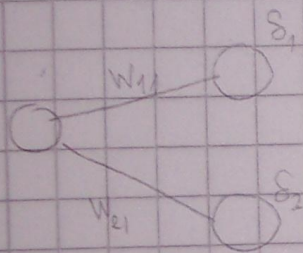
DELTA PRAVILO

$$(W_{ji}) = (W_{ji})_{\text{trenina}} - \eta \cdot \delta_j \cdot z_i$$

$\downarrow f'(x)$ klas i-tog prehodnika

$$\delta_j = y(1-y)(y-y_0) \text{ za izlaznu sloj}$$

$$\delta_j = z \cdot (1-z) \cdot \sum_k \delta_k \cdot W_{kj} \text{ za skrytnu sloj}$$



$$W_{11}^{(1)} = W_{11}^{(0)} - \eta \cdot \delta_1 \cdot x_1 = W_{11}^{(0)} - 0.1 \cdot x_1 \cdot \delta_1$$

$$\delta_1 = z \cdot (1-z) \cdot \sum_k \delta_k \cdot W_{k1} = \delta_1 \cdot W_{11}^{(0)}$$

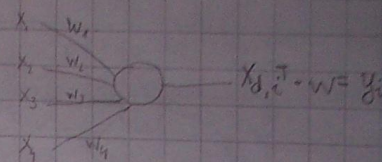
Imamo neuron s 4 ulaza treba riješiti se sljedećim niza podataka, (2/2/2)

ulaz 1	2	3	4	izlaz
1	0	0	0	1
0	0	0	1	1
0	0	1	0	0

$$X_d = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \end{bmatrix}$$

Je li rješenje jedinstveno ili ne?

3 para



$$X_d^T \cdot w = y_d$$

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \end{bmatrix} \cdot \begin{bmatrix} w_1 \\ w_2 \\ w_3 \\ w_4 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}$$

$$w_1 \cdot \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} + w_2 \cdot \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix} + w_3 \cdot \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix} + w_4 \cdot \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}$$

$$w_1 = 1$$

$$w_2 = \infty$$

$$w_3 = 0$$

$$w_4 = 1$$

$$w = \begin{bmatrix} 1 \\ \infty \\ 0 \\ 1 \end{bmatrix}$$

Ima više para od parametara rješenje nije jedinstveno. Rješenje ima beskonačno.

Bravo rješenje nejednake norme. :-

$$A^T (AA^T)^{-1}$$

$$A = X_d^T : A \cdot A^T = I_3 \Rightarrow (AA^T)^{-1} = I_3$$

$$\text{optimalno rješenje } w = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix}$$

$$w = A^T (AA^T)^{-1} \cdot \underset{y_d}{b} = A^T \cdot I_3 \cdot y_d = A^T \cdot y_d = X_d \cdot y_d = \dots = \text{projeke}$$

← optimalni parametar

2. način preko korekcijskog algoritma (naučivanje mreže)

$$w_0 = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\mu = 1$$

$$w_{k+1} = w_k - \mu \frac{e_k X_{d,k}}{\|X_{d,k}\|^2}$$

$$\text{korak 1: } w_1 = w_0 - 1 \cdot \frac{e_0 X_{d,0}}{\|X_{d,0}\|^2} \quad X_{d,0} = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} \quad e_0 = X_{d,0}^T \cdot w_0 - y_{d,0} = -1$$

$$= \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} - 1 \cdot \frac{1 \cdot X_{d,0}}{1} = X_{d,0} = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\text{korak 2: } w_2 = w_1 - 1 \cdot \frac{e_1 X_{d,1}}{\|X_{d,1}\|^2} = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} - 1 \cdot \frac{[0 \ 0 \ 0 \ 1]^T}{1} = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix}$$

$$e_1 = X_{d,1}^T \cdot w_1 - y_{d,1} = [0 \ 0 \ 0 \ 1] \cdot \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} - 1 = -1$$

3. način: izračunavanje direktnih projekcija

→ 1. algoritam

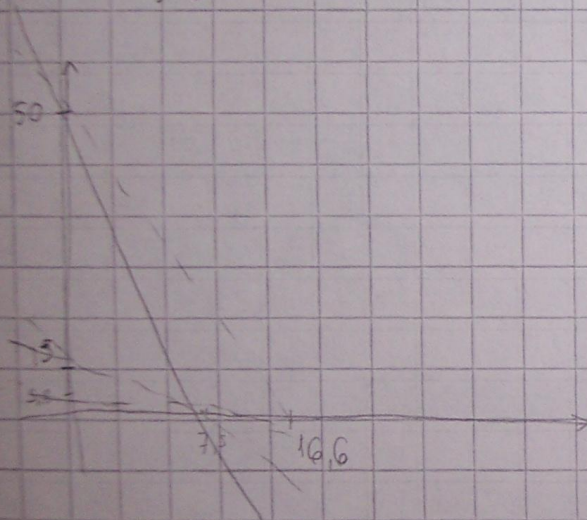
$$EA_o = \left(\frac{y_o}{y_o} - \frac{y_{dp}}{y_o} \right) = \frac{1}{2} - 1 = -\frac{1}{2}$$

Simplex algorithm

max $5x_1 + x_2$

subject $\begin{bmatrix} 3 & 1 \\ 4 & 2 \\ 3 & 4 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \geq \begin{bmatrix} 50 \\ 30 \\ 40 \end{bmatrix}$

$x_1 \geq 0$



$$3x_1 + x_2 = 50$$
$$x_2 = 50 - 3x_1$$

$$4x_1 + 2x_2 = 30$$
$$x_2 = 15 - 2x_1$$

W/e dobrze będzie
zadatek

$$3x_1 + 4x_2 = 40$$
$$x_2 = \frac{40 - 3x_1}{4}$$