

## 10.2

1. Trainingsdaten  $D = \{(x_1, t_1), (x_2, t_2), \dots, (x_n, t_n)\}$

$$f_w(x, w) = w_0 x^0 + w_1 x^1 + w_2 x^2 + \dots + w_n x^n$$

$\hookrightarrow n=2$

$$2. E(w) = \frac{1}{2} \sum_{n=0}^3 [f(x, w) - t_n]^2 = \frac{1}{2} \sum_{n=0}^3 \left[ \sum_{m=0}^2 w_m x^m - t_n \right]^2$$

$$\begin{aligned} 3. \quad \Rightarrow \quad \frac{\partial E(w)}{\partial w_i} &= \frac{\partial}{\partial w_i} \frac{1}{2} \sum_{n=0}^3 \left[ \sum_{m=0}^2 w_m x^m - t_n \right]^2 = \\ &= \frac{1}{2} \cdot \sum_{n=0}^3 \left[ \frac{\partial}{\partial w_i} \left( \sum_{m=0}^2 w_m x^m - t_n \right)^2 \right] = \\ &= \frac{1}{2} \cdot \sum_{n=0}^3 \left[ 2 \cdot \left( \sum_{m=0}^2 w_m x^m - t_n \right) \cdot \frac{\partial}{\partial w_i} \left( \sum_{m=0}^2 w_m x^m - t_n \right) \right] = \\ &= \sum_{n=0}^3 \left( \sum_{m=0}^2 w_m x^m - t_n \right) \cdot \left( \sum_{m=0}^2 \frac{\partial}{\partial w_i} w_m x^m - \underbrace{\frac{\partial}{\partial w_i} t_n}_0 \right) = \\ &= \sum_{n=0}^3 \left[ \left( \sum_{m=0}^2 w_m x^m - t_n \right) \cdot x_n^i \right] \end{aligned}$$

$$\begin{aligned} 4. \quad \frac{\partial E(w)}{\partial w_0} &= x_0^0 (w_0 + x_0 w_1 + w_2 x_0^2 - t_0) + x_1^0 (w_0 + w_1 x_1 + w_2 x_1^2 - t_1) + \\ &\quad + x_2^0 (w_0 + w_1 x_2 + w_2 x_2^2 - t_2) + x_3^0 (w_0 + w_1 x_3 + w_2 x_3^2 - t_3) = \\ &= w_0 (x_0^0 + x_1^0 + x_2^0 + x_3^0) + w_1 (x_0^1 + x_1^1 + x_2^1 + x_3^1) + \\ &\quad + w_2 (x_0^2 + x_1^2 + x_2^2 + x_3^2) - (t_0 x_0^0 + t_1 x_1^0 + t_2 x_2^0 + t_3 x_3^0) = \\ &= w_0 (1+1+1+1) + w_1 (1+2+4+5) + w_2 (1+4+16+25) - \\ &\quad - (1+3+3+1) = 4w_0 + 12w_1 + 46w_2 - 8 \end{aligned}$$



$$\frac{\partial E(\omega)}{\partial \omega_1} = \omega_0(x_1 + x_2 + x_3 + x_0) + \omega_1(x_0^2 + x_1^2 + x_2^2 + x_3^2) + \omega_2(x_1^3 + x_2^3 + x_3^3 + x_0^3) - (t_0 x_0 + t_1 x_1 + t_2 x_2 + t_3 x_3) =$$

$$= 12\omega_0 + 46\omega_1 + 198\omega_2 - 24$$

$$\frac{\partial E(\omega)}{\partial \omega_2} = \omega_0(x_1^2 + x_2^2 + x_3^2 + x_0^2) + \omega_1(x_0^3 + x_1^3 + x_2^3 + x_3^3) + \omega_2(x_0^4 + x_1^4 + x_2^4 + x_3^4) - (t_0 x_0^2 + t_1 x_1^2 + t_2 x_2^2 + t_3 x_3^2) =$$

$$= 46\omega_0 + 198\omega_1 + 898\omega_2 - 86$$

5.  $\frac{\partial E(\omega)}{\partial \omega_i} = 0$  f.a.  $\omega_i$

$$\Rightarrow \frac{\partial E(\omega)}{\partial \omega_0} = 0 \Leftrightarrow 4\omega_0 + 12\omega_1 + 46\omega_2 = 8$$

$$\Rightarrow \frac{\partial E(\omega)}{\partial \omega_1} = 0 \Leftrightarrow 12\omega_0 + 46\omega_1 + 198\omega_2 = 24$$

$$\Rightarrow \frac{\partial E(\omega)}{\partial \omega_2} = 0 \Leftrightarrow 46\omega_0 + 198\omega_1 + 898\omega_2 = 86$$

Gaußsches Eliminationsverfahren:

$$\left( \begin{array}{ccc|c} 4 & 12 & 46 & 8 \\ 12 & 46 & 198 & 24 \\ 46 & 198 & 898 & 86 \end{array} \right)$$

II - 3·I

$$\downarrow$$

$$\left( \begin{array}{ccc|c} 4 & 12 & 46 & 8 \\ 0 & 10 & 60 & 0 \\ 46 & 198 & 898 & 86 \end{array} \right)$$

III - ~~11~~ I

$$\downarrow$$

$$\left( \begin{array}{ccc|c} 4 & 12 & 46 & 8 \\ 0 & 10 & 60 & 0 \\ 2 & 66 & 392 & -2 \end{array} \right)$$

III -  $\frac{1}{2}$  I

$$\downarrow$$

$$\left( \begin{array}{ccc|c} 4 & 12 & 46 & 8 \\ 0 & 10 & 60 & 0 \\ 0 & 60 & 369 & -6 \end{array} \right)$$

III - 6·II

$$\downarrow$$

$$\left( \begin{array}{ccc|c} 4 & 12 & 46 & 8 \\ 0 & 10 & 60 & 0 \\ 0 & 0 & 9 & -6 \end{array} \right)$$

$$\Rightarrow 9\omega_2 = -6 \Rightarrow \omega_2 = -\frac{2}{3}$$

$$\Rightarrow 10\omega_1 - \frac{60 \cdot 2}{3} = 0$$

$$\Rightarrow \omega_1 = 4$$

$$\Rightarrow 4\omega_0 + 4 \cdot 12 - \frac{2}{3} \cdot 46 = 8$$

$$\Rightarrow \omega_0 = -\frac{7}{3}$$



$$\Rightarrow y = -\frac{7}{3} + 4x - \frac{2}{3}x^2$$

$$f(x_0) = -\frac{7}{3} + 4 - \frac{2}{3} = 1 = t_0$$

$$f(x_1) = -\frac{7}{3} + 4 \cdot 2 - \frac{2}{3} \cdot 2^2 = 3 = t_1$$

$$f(x_2) = -\frac{7}{3} + 4 \cdot 4 - \frac{2}{3} \cdot 4^2 = 3 = t_2$$

$$f(x_3) = -\frac{7}{3} + 4 \cdot 5 - \frac{2}{3} \cdot 5^2 = 1 = t_3$$

10.1

$$\begin{pmatrix} 0,2 & 0,1 & 0,2 & 0,9 & 0,1 \\ 0,1 & i & 0,7 & j & 0,6 \\ 0,1 & 0,1 & 0,2 & 0,5 & 0,8 \end{pmatrix}$$

$$g_{\max} = 240$$

$$g_i = 120, \quad g_j = 180$$

$$c_s = 0,5, \quad c_d = -0,5$$

Initialisierung:

$$p^{(0)}(i) = \frac{g_i}{g_{\max}} = \frac{120}{240} = 0,5$$

$$p^{(0)}(j) = \frac{g_j}{g_{\max}} = \frac{180}{240} = 0,75$$

Betrag eines Nachbarpixels  $i_k$  zum Wert  $p(i)$ :

$$\begin{aligned} q(i) &= c_s \cdot p(i_k) + c_d \cdot (1 - p(i_k)) = \\ &= 0,5 \cdot p(i_k) + 0,5 \cdot (1 - p(i_k)) = \\ &= 0,5 \cdot p(i_k) + 0,5 p(i_k) - 0,5 = \\ &= p(i_k) - 0,5 \end{aligned}$$

Änderung von  $p(i)$  aufgrund der Beträge der Nachbarpixel:

$$\Delta p(i) = \frac{1}{8} \cdot \sum_{k=1}^8 q(i_k) = \frac{1}{8} \cdot \sum_{k=1}^8 (p(i_k) - 0,5)$$



Normalisierung der Wsl.  $p^{(r)}(i)$  nach Iteration  $r$ :

$$p^{(r)}(i) = \frac{p^{(r-1)}(i) \cdot (1 + \Delta p(i))}{p^{(r-1)}(i) \cdot (1 + \Delta p(i)) + (1 - p^{(r-1)}(i)) \cdot (1 - \Delta p(i))}$$

Iteration 1:

$$\Delta p(i) = \frac{1}{8} \cdot (-0,3 - 0,4 - 0,3 - 0,4 - 0,2 - 0,4 - 0,4 - 0,3) = \\ = \frac{1}{8} \cdot (-2,3) = -0,2875$$

$$\Delta p(j) = \frac{1}{8} \cdot (-0,3 + 0,4 - 0,4 + 0,2 + 0,1 - 0,3 + 0 + 0,3) = \\ = 0$$

$$\Rightarrow p^{(1)}(i) = \frac{p^{(0)}(i) \cdot (1 + \Delta p(i))}{p^{(0)}(i) \cdot (1 + \Delta p(i)) + (1 - p^{(0)}(i)) \cdot (1 - \Delta p(i))} = \frac{0,5 \cdot 0,7125}{0,5 \cdot 0,7125 + 0,5 \cdot 1,2875} =$$

$$= 0,35625$$

$$\Rightarrow p^{(1)}(j) = \frac{0,75 \cdot 1}{0,75 \cdot 1 + 0,25 \cdot 1} = 0,75$$

Iteration 2:

$\Delta p(i)$  und  $\Delta p(j)$  wie in Iteration 1

$$\Rightarrow p^{(2)}(i) = \frac{0,35625 \cdot 0,7125}{0,35625 \cdot 0,7125 + 0,64375 \cdot 1,2875} = 0,2344 \Rightarrow \text{Vordergrund}$$

$$\Rightarrow p^{(2)}(j) = 0,75 \Rightarrow \text{Hintergrund}$$