

September 16 2024.

$$\text{Pg 18 } J_{\text{reg}} = J(w) + \lambda R(w) = J(w) + \frac{1}{2} \sum_j w_j^2$$

$\frac{1}{2} \sum_j w_j^2$   $\hookrightarrow$  L2 norm of  $w$ , or  $\|w\|_2^2$

$$\frac{\partial J}{\partial w_j} > 0 \Rightarrow \text{increase } w_j \Rightarrow \text{increase } J \Rightarrow \text{you should decrease } w_j$$
$$\frac{\partial J}{\partial w_j} < 0 \Rightarrow \text{increase } w_j \Leftarrow \text{decrease } J \Rightarrow \text{you should increase } w_j$$

$\Rightarrow$  when gradient  $\nabla J(w_j) < 0 \Rightarrow$  increase  $w_j$  } going towards the opposite  
 $> 0 \Rightarrow$  decrease  $w_j$  } direction of the gradient  
until we find the critical point ( $\nabla J(w_j) = 0$ )

*opposite direction of the gradient*

$$\text{Update rule: } w_j \leftarrow w_j - \partial \frac{\partial J}{\partial w_j} \text{ or } w_j \leftarrow w_j - \partial \cdot \nabla J(w_j)$$

$$w \leftarrow w - \frac{1}{N} \sum_i \partial \cdot (y^i - t^i) \cdot x^i$$

*derived in previous lectures*

$$w \leftarrow w - \frac{2}{N} \sum_i (y^i - t^i) x^i$$

$$w \leftarrow w - \frac{\partial}{\partial w} (J + \lambda R(w)) \cdot \alpha = w - \partial \cdot \left( \frac{\partial J}{\partial w} + \lambda \cdot \frac{\partial R}{\partial w} \right)$$

$$= w - \partial \cdot \left( \frac{\partial J}{\partial w} + \lambda \cdot \frac{\partial}{\partial w} \cdot \frac{1}{2} \sum_j w_j^2 \right)$$

$$= w - \partial \cdot \left( \frac{\partial J}{\partial w} + \lambda \cdot \frac{1}{2} \cdot 2 \cdot \frac{[w_1, \dots, w_D]}{\hookrightarrow w} \right) = w - \partial \cdot \left( \frac{\partial J}{\partial w} + \lambda w \right)$$

$$= w - \partial \cdot \frac{\partial J}{\partial w} - \partial \cdot \lambda \cdot w = w - 2 \cdot w \cdot \frac{\lambda}{2} - \partial \cdot \frac{\partial J}{\partial w}$$

$$= (1 - \partial \lambda) w - \partial \cdot \frac{\partial J}{\partial w}$$