

$$\Rightarrow w_i = \begin{cases} 0, & -\lambda < z_i < \lambda \\ z_i, & \text{otherwise} \end{cases}$$

We have:

$$E(w) = \frac{1}{2} \sum_i (d_i - w^T x_i)^2 + \frac{\lambda}{2} \|w\|^2$$

Proximal gradient descent:

$$z^{(k)} = w^{(k)} - \frac{1}{2} x_i^T (x_i w^{(k)} - d) \quad (\text{least squares gradient descent})$$

$$\Rightarrow w^{(k+1)} = \arg \min_w |z^{(k)} - w_i|^2 + \frac{1}{2} \lambda |w| \quad (1)$$

take derivative (1):

$$\bullet w_i > 0: \quad -2(z^{(k)} - w_i) + \frac{\lambda}{2} = 0$$

$$\Leftrightarrow w_i = z_i^{(k)} - \frac{\lambda}{4} \quad \text{only when } z_i^{(k)} > \frac{\lambda}{4} \quad (w_i > 0)$$

$$w_i = 0 \quad \text{when } z_i < \frac{\lambda}{4}$$

$$\bullet w_i < 0: \quad \frac{d}{dw_i} \left[(z^{(k)} - w_i)^2 + \frac{\lambda}{2} (-w) \right] = 0 \quad (\text{for it to be min})$$

$$\Leftrightarrow -2(z^{(k)} - w_i) - \frac{\lambda}{2} = 0$$

$$\Leftrightarrow w_i = z_i^{(k)} + \frac{\lambda}{4}, \quad \text{when } z_i^{(k)} < -\frac{\lambda}{4}$$

$$w_i = 0 \quad \text{when } z_i > -\frac{\lambda}{4}$$

Thứ

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$$\Rightarrow w_i = \begin{cases} 0, & -\frac{\lambda}{4} < z_i < \frac{\lambda}{4} \\ z_i - \frac{\lambda}{4}, & z_i > \frac{\lambda}{4} \\ z_i + \frac{\lambda}{4}, & z_i < -\frac{\lambda}{4} \end{cases}$$

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