

① for KKT conditions :

$$\text{st. } \lambda_i(y_i - (w\varphi(x) + w_0) + \xi_i) = 0$$

$$\xi_i \geq 0$$

$$\lambda_i \geq 0$$

$$y_i - (w\varphi(x) + w_0) \leq \varepsilon$$

$$\text{for } \xi_i = 0 \Rightarrow \lambda_i(y_i - (w\varphi(x) + w_0)) = 0$$

$$\text{if } \lambda_i > 0, \text{ then } y_i - (w\varphi(x) + w_0) = \varepsilon$$

(support vectors)

$$w\varphi(x) + w_0 = y_i - \varepsilon$$

$$w_0 = y_i - \varepsilon - w\varphi(x)$$

for support vector at the upper edge of ε-table, which is presented by x_j

$$w_0 = y_j - \varepsilon - w\varphi(x_j)$$

~~$$w\varphi(x_j) = \sum \lambda_i y_i k(x_i, x_j)$$~~

~~$$w_0 = y_j - \varepsilon - \sum \lambda_i y_i k(x_i, x_j)$$~~

$$\text{SVR: } L(w, w_0, \xi, \lambda^+, \lambda^-) = 0.5 \|w\|^2 + C(\sum \lambda_i^+ + \sum \lambda_i^-) - \sum \lambda_i^+ (y_i - w\varphi(x_i) + w_0) - \sum \lambda_i^- (y_i - w\varphi(x_i) + w_0) + \sum \lambda_i^+ \xi_i^+$$

$$\frac{\partial L}{\partial w} = 0 \Rightarrow w + \sum \lambda_i^+ \varphi(x_i) - \sum \lambda_i^- \varphi(x_i) = 0$$

$$w = \sum (\lambda_i^+ - \lambda_i^-) \varphi(x_i)$$

$$\frac{\partial L}{\partial w_0} = 0$$

$$\sum (\lambda_i^+ - \lambda_i^-) = 0$$

substitute w into the equation with w_0 ,

$$w_0 = y_j - \varepsilon - \sum (\lambda_i - \lambda^*) \varphi(x_i) \cdot \varphi(x_j)$$

$$K(x_i, x_j) = \varphi(x_i) \cdot \varphi(x_j)$$

$$\therefore w_0 = y_j - \varepsilon - \sum (\lambda_i - \lambda^*) K(x_i, x_j)$$

$$= y_j - \varepsilon - \sum (\lambda^+ - \lambda^-) K(x_i, x_j)$$

② a. learning rate $\alpha = 0.1\tau = 0.5$

2 epochs only

$$\begin{cases} x_1 = c_1 = [1, -1, -1] \\ x_2 = c_2 = [-1, 1, -1] \\ x_3 = c_3 = [-1, -1, 1] \end{cases}$$

Start with $\begin{bmatrix} 0 & 2 \\ 3 & 0 \\ 1 & -1 \end{bmatrix}$

for x_1 target $[1, -1, -1]$

$$\begin{aligned} e_1 &= y_1 - \text{sign}(w(0)^T x(1)) = [1, -1, -1] - \text{sign}[0, 3, 1, 2, -1] \\ &= [0, -2, -2] \end{aligned}$$

$$w(1) = w(0) + \alpha (1) * x(1) + e_1^T = \begin{bmatrix} 0 & 1 \\ 3 & -1 \\ 1 & -2 \end{bmatrix}$$

$$e_2 = y_2 - \text{sign}(w(1)^T x(2)) = [0, 2, 0]$$

$$w(1) = w(0) + \eta(1) * x_1 \cdot e_1^T = \begin{bmatrix} 0 & 0 \\ 3 & -2 \\ 1 & -1 \end{bmatrix}$$

$$e_1 = [0 \ 0 \ 2]$$

$$w(2) = w(1) + \eta(2) * x_2 \cdot e_2^T = \begin{bmatrix} 0 & 0 \\ 3 & -2 \\ 1 & -3 \end{bmatrix}$$

not converges yet, and it is keeping changing

$$(b) w(0) = \begin{bmatrix} 0 & 0 \\ 3 & 0 \\ 1 & -1 \end{bmatrix}$$

$$e_1 = y_1 - \text{sign}(w(0)^T \cdot x(1)) = [0 \ -2 \ -2]$$

$$w_1(1) = w_1(0) + \eta e_1 \cdot x(1) = 0 \cdot 5 \cdot 0 \cdot 1 + 0 \cdot 5 \cdot 0$$

$$w_2(1) = 1$$

$$w_3(1) = -2$$

$$\therefore w(1) = [0 \ 1 \ -2]$$

$$e_2 = y_2 - \text{sign}(w(1)^T \cdot x(2)) = [0 \ 0 \ -2]$$

$$w_1(2) = 0$$

$$w_2(2) = 1$$

$$w_3(2) = -1$$

$$\therefore w(2) = [0 \ 1 \ -1]$$

$$e_3 = y_3 - \text{sign}(w(2)^T \cdot x(3)) = [0 \ -1 \ 1]$$

$$h(\vec{z}) = 0$$

$$\mathcal{M}_2(\vec{z}) = 1.8$$

$$\mathcal{W}_3(\vec{z}) = 1.5$$

$$h(\vec{z}) = [0 \ 1.5 \ -1.5]$$

Mt converge either

③

x_1	x_2
x_2	x_4

class1 : $x_1 \& x_2$ or $x_3 \& x_4$

class2 : $x_1 \& x_3$ or $x_2 \& x_4$

$$D = \text{sigmoid}(\text{sigmoid}(x_1 * x_2) + \text{sigmoid}(x_3 * x_4))$$

If $D > 0.5$ class 1

$D < 0.5$ class 2