

HW-8

- ① Find the solution (x^*, y^*) to the following problem:
optimize xy ; subject to $x+y=10$

→ $x+y-10=0$

Lagrangian $L(x, y, \beta) = xy + \beta(x+y-10)$

Taking partial derivatives;

$\nabla_x L(x, y, \beta) = y + \beta = 0 \rightarrow \textcircled{1}$

$\nabla_y L(x, y, \beta) = x + \beta = 0 \rightarrow \textcircled{2}$

$\nabla_\beta L(x, y, \beta) = x+y-10=0 \rightarrow \textcircled{3}$

Solving above equation for $x+y$:-

$y = -\beta$ (from ①)

$x = -\beta$ (from ②)

Substituting in ③,

$-\beta - \beta - 10 = 0$

$-2\beta - 10 = 0$

$-2\beta = 10$

$\therefore \boxed{\beta = -5}$

ie. $x^* = 5, y^* = 5$

- ② The SVM optimization can be defined by the primal form:
$$\min_w \frac{1}{2} \|w\|^2 \text{ subject to } y_i(w^T x_i + b) \geq 1, \quad i=1, \dots, N$$

i) Write the primal problem in standard form.

$g_i(w) = -y_i(w^T x_i + b) + 1 \leq 0$

ii) Form the Lagrangian function $L(w, b, \alpha)$

$L(w, \alpha, \beta) = f(w) + \sum_{i=1}^m \alpha_i g_i(w) + \sum_{i=1}^k \beta_i h_i(w)$

here, $L(w, b, \alpha) = \frac{1}{2} \|w\|^2 + \sum_{i=1}^m \alpha_i [-y_i(w^T x_i + b) + 1]$

as there are no equality constraints $h_i(w)$,
there is no β_i term

iii) Find w and b that minimize $L(w, b, \alpha)$

$$= \frac{1}{2} \|w\|^2 + \sum_{i=1}^N \alpha_i [-y_i (w^T x_i + b) + 1]$$

Taking derivative,

$$\nabla_w L(w, b, \alpha) = w - \sum_{i=1}^N \alpha_i y_i x_i = 0$$

$$\therefore w = \sum_{i=1}^N \alpha_i y_i x_i$$

$$\therefore \nabla_b L(w, b, \alpha) = \sum_{i=1}^N \alpha_i y_i = 0$$

iv) Plug the results back into $L(w, b, \alpha)$

$$\|z\|^2 = z^T z$$

$$\therefore L(w, b, \alpha) = \frac{1}{2} w^T w + \sum_{i=1}^N \alpha_i [-y_i (w^T x_i + b) + 1]$$

$$\begin{aligned} \therefore L(w, b, \alpha) &= \frac{1}{2} \sum_{i=1}^N \alpha_i y_i x_i^T \cdot \sum_{j=1}^N \alpha_j y_j x_j \\ &\quad + \sum_{i=1}^N \alpha_i [-y_i ((\sum_{j=1}^N \alpha_j y_j x_j^T) x_i + b) + 1] \\ &= \frac{1}{2} \sum_{i=1}^N \sum_{j=1}^N \alpha_i \alpha_j y_i y_j (x_i^T x_j) - \sum_{i=1}^N \sum_{j=1}^N \alpha_i \alpha_j y_i y_j (x_j^T x_i) \\ &\quad - \sum_{i=1}^N \alpha_i y_i b + \sum_{i=1}^N \alpha_i \end{aligned}$$

But, $\sum_{i=1}^N \sum_{j=1}^N \alpha_i \alpha_j y_i y_j (x_j^T x_i) = \sum_{i=1}^N \sum_{j=1}^N \alpha_i \alpha_j y_i y_j (x_i^T x_j)$

also, from subequation (3), $\sum_{i=1}^N \alpha_i y_i b = 0$

\therefore The objective function $J(\alpha) = L(w, b, \alpha)$

$$= \sum_{i=1}^N \alpha_i - \frac{1}{2} \sum_{i=1}^N \sum_{j=1}^N \alpha_i \alpha_j y_i y_j (x_i^T x_j)$$