

$$\textcircled{1} \quad f(x, y) = xy$$

$$h(x, y) = x + y - 10$$

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

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

$$\Rightarrow \frac{\partial L}{\partial x} = y + \beta = 0 \quad \textcircled{1}$$

$$\Rightarrow \frac{\partial L}{\partial y} = x + \beta = 0 \quad \textcircled{2}$$

$$\Rightarrow \frac{\partial L}{\partial \beta} = x + y - 10 = 0 \quad \textcircled{3}$$

$$\text{from } \textcircled{1} \textcircled{2} \quad y = -\beta, x = -\beta$$

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

$$-2\beta - 10 = 0$$

$$\beta = -5$$

$$\Rightarrow \boxed{x = 5, y = 5}$$

$$(2) \quad f(x) = 7x_1^3 + 8x_2^2 + 9x_1$$

s.t.

$$h_1(x) \Rightarrow 6x_1 + 11x_2^2 = 13x_2$$

$$\Rightarrow 6x_1 + 11x_2^2 - 13x_2 = 0$$

$$g_1(x) \Rightarrow 5x_2^2 \leq 4x_1 - 7$$

$$\Rightarrow 5x_2^2 - 4x_1 + 7 \leq 0$$

$$g_2(x) \Rightarrow 20x_1 + 5 \geq x_2$$

$$\Rightarrow x_2 - 20x_1 - 5 \leq 0$$

$$L(\alpha, \beta, x) = f(x) + \alpha_1 g_1(x) + \alpha_2 g_2(x) + h(x)$$

$$= 7x_1^3 + 8x_2^2 + 9x_1 + \alpha_1 (5x_2^2 - 4x_1 + 7) + \alpha_2 (x_2 - 20x_1 - 5) + \beta_1 (6x_1 + 11x_2^2 - 13x_2)$$

(3)

$$f(x) = \frac{1}{2} \|w\|^2$$

s.t.

$$y_i (w^T x_i + b) \geq 1 \quad i = 1, 2, \dots, N$$

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

$$L(w, b, \alpha) = f(x) + \sum_{i=1}^N \alpha_i g_i(x)$$

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

$$\frac{\partial L}{\partial w} = w + \sum_{i=1}^N (-\alpha_i y_i x_i) = 0$$

$$w = \sum_{i=1}^N \alpha_i y_i x_i \quad \text{--- (1)}$$

$$\frac{\partial L}{\partial b} = \sum_{i=1}^N \alpha_i y_i = 0 \quad \text{--- (2)}$$

use w from equation (1) and plug into L

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

$$= \frac{1}{2} w \cdot w^T + \sum_{i=1}^N \alpha_i - \sum_{i=1}^N \alpha_i y_i w^T x_i - \sum_{i=1}^N \alpha_i y_i b$$

$$\sum_{i=1}^N \alpha_i + \frac{1}{2} \left(\sum_{i=1}^N \alpha_i y_i x_i \right) \left(\sum_{i=1}^N \alpha_i y_i (x_i)^T \right) - \sum_{i=1}^N \alpha_i y_i \left(\sum_{i=1}^N \alpha_i y_i (x_i)^T \right) x_i$$

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

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

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

$$\mathcal{O}_p(\omega) = \max_{\alpha, \beta: \alpha_i \geq 0} L(\omega, \alpha, \beta)$$

$$= \max_{\alpha, \beta: \alpha_i \geq 0} L(\omega, b, \alpha)$$

$$\text{also s.t. } \sum_{i=1}^N \alpha_i y_i = 0 \quad (\text{from Eq(2)})$$

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

$$\text{s.t. } \alpha_i \geq 0, i=1, 2, \dots, N$$

$$\sum_{i=1}^N \alpha_i y_i = 0, i=1, 2, \dots, N$$

$$(4) (a) \phi(x) = [1, \sqrt{2}x, x^2]^T$$

$$\phi(x_1) = [1, \sqrt{2} \times 0, (0)^2]^T$$

$$= [1, 0, 0]^T$$

$$\phi(x_2) = [1, \sqrt{2} \times \sqrt{2}, (\sqrt{2})^2]^T$$

$$= [1, 2, 2]^T$$

(b) Since there are only 2 points in the dataset, those points are support vectors and to maximize the width of the street, w vector will pass through those 2 points.

hence, $\text{width} = 2 \times \text{margin}$

width is distance between those 2 points.

$(1, 0, 0)$ and $(1, 2, 2)$

$$\begin{aligned}\text{distance} &= \sqrt{0^2 + 2^2 + 2^2} \\ &= \sqrt{8}\end{aligned}$$

$$\text{and margin} = \frac{\sqrt{8}}{2} = \sqrt{2}$$

$$\underline{\text{margin} = \sqrt{2}}$$

$$(c). \quad w = (1-1)t\hat{i} + (2-0)t\hat{j} + (2-0)t\hat{k}$$

$$w = 2t\hat{j} + 2t\hat{k}$$

$$\Rightarrow \boxed{w_1 = 0}$$

$$\text{margin} = \frac{1}{\|w\|}$$

$$\sqrt{2} = \frac{1}{\sqrt{(2t)^2 + (2t)^2}} = \frac{1}{\sqrt{8t^2}}$$

$$\sqrt{16t^2} = 1 \Rightarrow t = \frac{1}{4} \quad \boxed{\therefore w = \frac{1}{2}\hat{j} + \frac{1}{2}\hat{k}}$$

$$\begin{aligned}
 (d) \quad & y_1 (\omega^T \phi(x_1) + \omega_0) = 1 \quad \text{--- ①} \\
 & y_2 (\omega^T \phi(x_2) + \omega_0) = 1 \quad \text{--- ②}
 \end{aligned}
 \quad \left. \vphantom{\begin{aligned} y_1 (\omega^T \phi(x_1) + \omega_0) = 1 \\ y_2 (\omega^T \phi(x_2) + \omega_0) = 1 \end{aligned}} \right\} \text{on decision boundary}$$

for ①

$$\Rightarrow -1 \left(\begin{bmatrix} 0 & \frac{1}{2} & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + \omega_0 \right) = 1$$

$$\Rightarrow -1 (0 + \omega_0) = 1$$

$$\boxed{\omega_0 = -1}$$

for ②

$$\Rightarrow 1 \left(\begin{bmatrix} 0 & \frac{1}{2} & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} + \omega_0 \right) = 1$$

$$\Rightarrow 2 \times \frac{1}{2} + 2 \times \frac{1}{2} + \omega_0 = 1$$

$$\Rightarrow 2 + \omega_0 = 1$$

$$\boxed{\omega_0 = -1}$$

$$(e) \quad f(x) = \omega_0 + \omega^T \phi(x)$$

$$f(x) = -1 + \begin{bmatrix} 0 & \frac{1}{2} & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 1 \\ \sqrt{2}x \\ x^2 \end{bmatrix}$$

$$f(x) = -1 + \frac{\sqrt{2}x}{2} + \frac{x^2}{2} = \frac{x^2 + \sqrt{2}x - 2}{2}$$