

### Solution-3

(a) let  $\alpha_i = 1 \quad \forall i = 1, 2, \dots, m$  and  $b = 0$ .

for  $\{x^i, y^i\} \in \text{training set}$ , we have

$$|f(x^i) - y^i| = \left| \sum_{j=1}^m y^j k(x^i, x^j) - y^i \right|$$

$$\left( \because k(x, z) = \exp(-\|x - z\|^2 / \tau^2) \right)$$

$$= \left| \sum_{j=1}^m y^j \exp(-\|x^i - x^j\|^2 / \tau^2) - y^i \right|$$

$$= \left| y^i + \sum_{j \neq i} y^j \exp(-\|x^i - x^j\|^2 / \tau^2) - y^i \right|$$

$$= \left| \sum_{j \neq i} y^j \exp(-\|x^i - x^j\|^2 / \tau^2) \right|$$

$$(\because |a+b| \leq |a| + |b|)$$

$$\leq \sum_{j \neq i} |y^j| \exp(-\|x^i - x^j\|^2 / \tau^2)$$

$$= \sum_{j \neq i} |y^j| \cdot \exp(-\|x^i - x^j\|^2 / \tau^2)$$

$$= \sum_{j \neq i} \exp(-\|x^i - x^j\|^2 / \tau^2)$$

$$\leq \sum_{j \neq i} \exp(-\varepsilon^2 / \tau^2) = (m-1) \exp(-\varepsilon^2 / \tau^2)$$

assuming  $\|x^i - x^j\| \geq \varepsilon \quad \forall i \neq j$

$$(m-1) \exp(-\varepsilon^2 / \tau^2) < 1$$

$$\tau < \frac{\varepsilon}{\sqrt{\log(m-1)}}$$

(b) Yes we will obtain zero training error.

Because if the SVM without slack variables is able to find a

Solution it will always give zero training error.

We will now show that there exists at least one feasible point.

Let  $y^i(\omega^T x^i + b)$  for some  $i$  and let  $b=0$  then.

$$y^i(\omega^T x^i + b) = y^i(\omega^T x^i) = y^i f(x^i) > 0$$

Since,  $y^i, f(x^i)$  have same sign. Choosing all  $x^i$ 's which are large enough such that  $y^i(\omega^T x^i + b) > 1$ .

hence the opt. problem is feasible.

## Solution-2

a) Consider a solution of the given problem with  $\epsilon < 0$ .

2) then  $y^i(w^T x^i + b) \geq 1 - \epsilon_i$  will be satisfied for  $\epsilon_i = 0$ . and then the objective function will be lower

3) which implies this can not be an optimal solution.

b)

$$L(w, b, \epsilon, \alpha) = \frac{1}{2} w^T w + \frac{c}{2} \sum_{i=1}^m \epsilon_i^2 - \sum_{i=1}^m \alpha_i (y^i (w^T x^i + b) - 1 + \epsilon_i).$$

where  $\alpha_i \geq 0 \quad i = 1, 2, 3, \dots, m$ .

c)

$$w(\alpha) = \min_{w, b, \epsilon} L(w, b, \epsilon, \alpha)$$

$$= \frac{1}{2} \sum_{i=1}^m \sum_{j=1}^m (\alpha_i y^i x^i)^T (\alpha_j y^j x^j) + \frac{1}{2} \sum_{i=1}^m \frac{\alpha_i}{\epsilon_i} \epsilon_i^2 - \sum_{i=1}^m \alpha_i \left( y^i \left( \sum_{j=1}^m \alpha_j y^j x^j \right)^T x^i + b \right) - 1 + \epsilon_i$$

$$= -\frac{1}{2} \sum_{i=1}^m \sum_{j=1}^m \alpha_i \alpha_j y^i y^j (x^i)^T x^j + \frac{1}{2} \sum_{i=1}^m \alpha_i \epsilon_i + \sum_{i=1}^m \alpha_i - \left( \sum_{i=1}^m \alpha_i y^i \right) b - \sum_{i=1}^m \alpha_i \epsilon_i$$

$$= \sum_{i=1}^m \alpha_i - \frac{1}{2} \sum_{i=1}^m \sum_{j=1}^m \alpha_i \alpha_j y^i y^j (x^i)^T x^j - \frac{1}{2} \sum_{i=1}^m \alpha_i \epsilon_i$$

$$= \sum_{i=1}^m \alpha_i - \frac{1}{2} \sum_{i=1}^m \sum_{j=1}^m \alpha_i \alpha_j y^i y^j (x^i)^T x^j - \frac{1}{2} \sum_{i=1}^m \frac{\alpha_i^2}{c}$$

# Dual formulation

$$\max_{\alpha} \sum_{i=1}^m \alpha_i - \frac{1}{2} \sum_{i=1}^m \sum_{j=1}^m \alpha_i \alpha_j y^i y^j (x^i)^T x^j - \frac{1}{2} \sum_{i=1}^m \frac{\alpha_i^2}{c}$$

Such that  $\alpha_i \geq 0, i=1, 2, \dots, m$ .

$$\sum_{i=1}^m \alpha_i y^i = 0.$$