

Question 1

Let  $f(\cdot) = \|\cdot\|: \mathbb{R}^n \rightarrow \mathbb{R}^+$ .

So,  $f$  is a norm function on  $\mathbb{R}^n$ , which by definition is non-negative and homogeneous.

$f$  is a convex function iff it is sub-additive.

Proof.

$$\begin{aligned} f(\theta x + (1-\theta)y) &\leq f(\theta x) + f((1-\theta)y) \quad \text{|| triangle ineq.} \\ &= \theta f(x) + (1-\theta)f(y) \quad \text{|| abs. homogeneity} \end{aligned}$$

Question 2

Claim:

A set  $C$  is convex and  $x_i \in C$ , for  $i = \{1, 2, 3\}$ .  
 $\theta_i \geq 0$ , for  $i = \{1, 2, 3\}$  and since  $\theta_1 + \theta_2 + \theta_3 = 1$ ,  
 $\theta_i \leq 1$ , for  $i = \{1, 2, 3\}$ .

Then also  $\sum_{i=1}^3 \theta_i x_i \in C$ .

Proof.

We can prove the above claim by induction.

$$\begin{aligned} \sum_{i=1}^3 \theta_i x_i &= \theta_1 x_1 + \theta_2 x_2 + \theta_3 x_3 \\ &= \theta_1 x_1 + \sum_{i=2}^3 \theta_i \left( \frac{\theta_2}{\theta_2 + \theta_3} x_2 + \frac{\theta_3}{\theta_2 + \theta_3} x_3 \right) \end{aligned}$$

$$= \theta_1 x_1 + \sum_{i=2}^3 \theta_i \bar{x}$$

|| By induction  $\bar{x} \in C$

$$= \theta_1 x_1 + (1 - \theta_1) \bar{x}$$



### Question 3

$$\begin{aligned} \text{a) } L_P &= \frac{1}{2} w^T w + C \sum_i \xi_i - \sum_i \alpha_i (y_i (w^T \phi(x_i) + b) - 1 + \xi_i) - \sum_i \mu_i \xi_i \\ &= \frac{1}{2} w^T w + C \sum_i \xi_i - \sum_i \alpha_i y_i w^T \phi(x_i) - \sum_i \alpha_i y_i b + \sum_i \alpha_i \\ &\quad - \sum_i \alpha_i \xi_i - \sum_i \mu_i \xi_i \\ &= \frac{1}{2} w^T w - \sum_i \alpha_i y_i w^T \phi(x_i) - \sum_i \alpha_i y_i b + \sum_i (C - (\alpha_i + \mu_i)) \xi_i + \sum_i \alpha_i \end{aligned}$$

$$\text{b) } \frac{\partial L_P}{\partial w} = w - \sum_i \alpha_i y_i \phi(x_i) = 0$$

$$w = \sum_i \alpha_i y_i \phi(x_i)$$

$$\frac{\partial L_P}{\partial b} = - \sum_i \alpha_i y_i = 0$$

$$\sum_i \alpha_i y_i = 0$$

$$\frac{\partial L_P}{\partial \xi_i} = \sum_i (C - (\alpha_i + \mu_i)) = 0$$

$$C = (\alpha_i + \mu_i)$$

c) We substitute the K.T. - K. conditions into primal Lagrangian, to get the dual form of L-SVM as follows:

$$\begin{aligned} L_D &= \frac{1}{2} w^T w - \underbrace{w^T \cdot \sum_i \alpha_i y_i \phi(x_i)}_{=w} - \underbrace{\sum_i \alpha_i y_i b}_{=0} + \sum_i \underbrace{(C - (\alpha_i + \mu_i)) \xi_i}_{=0} \\ &= + \sum_i \alpha_i \\ &= \frac{1}{2} w^T w - w^T w + \sum_i ((\alpha_i + \mu_i) - (\alpha_i + \mu_i)) \xi_i + \sum_i \alpha_i \\ &= \sum_i \alpha_i - \frac{1}{2} w^T w \end{aligned}$$

$$= \sum_i \alpha_i - \frac{1}{2} \sum_i \sum_j y_i y_j \alpha_i \alpha_j \phi(x_i) \phi(x_j)$$

$$= \sum_i \alpha_i - \frac{1}{2} \sum_i \sum_j y_i y_j \alpha_i \alpha_j K(x_i, x_j)$$

→ The dual form of the C-SVM problem is posed as

$$\max_{0 \leq \alpha_i \leq C} \sum_i \alpha_i - \frac{1}{2} \sum_i \sum_j y_i y_j \alpha_i \alpha_j K(x_i, x_j)$$


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