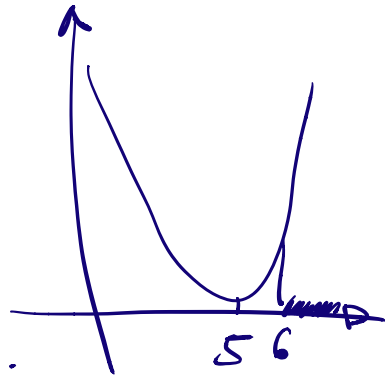


Examples

$$\textcircled{1} f_0(x) = (x-5)^2$$

$$f_1(x) = 6-x$$

(Convex) \wedge (Affine) holds.



KKT: (I)

$$f'_0(x^*) + z_1 f'_1(x^*) = 0$$

$$\Leftrightarrow 2(x^*-5) - z_1 = 0$$

$$\text{(II)} \quad z_1 f_1(x^*) = 0, \quad z_1 \geq 0$$

Case $z_1 > 0 \Rightarrow f_1(x) = 0 \Rightarrow x = 6$ ✓

Case $z_1 = 0 \Rightarrow 2(x^*-5) = 0$

$$\Rightarrow x^* = 5$$

$$\text{but } f_1(x^*=5) = 1 \quad \text{✗}$$

$$\textcircled{2} f_0(x) = (x-5)^2$$

$$f_1(x) = 3-x$$

$$\text{KKT (I)} \quad 2(x^*-5) - z_1 = 0$$

$$\text{KKT (II)} \quad z_1 > 0:$$

Case $f_1(x^*) = 0 \Rightarrow x^* = 3$

$$\Rightarrow 2(3-5) - z_1 = 0$$

$$\Rightarrow z_1 = -4 \quad \text{✗}$$

Case $z_1 = 0 \Rightarrow x = 5$ ✓

③ Hard margin case of SVM

In this case

$$f_0(w) = \frac{1}{2} \|w\|^2$$

$$f_i(w) = 1 - y^{(i)} w \cdot x^{(i)} \quad i = 1 \dots n$$

and no equality constraints

We know already

(Convex) holds

$\exists x^*$ optimal solution

and of training data lin. sep. dec. (Affine)

KKT (I)

$$d_w f_0(w^*) + \sum_{i=1}^n \alpha_i d_w f_i(w^*) = 0$$

$$\Rightarrow w^* = \sum_{i=1}^n \alpha_i y^{(i)} x^{(i)}$$

$$\text{KKT (II)} \quad \alpha_i f_i(w^*) = 0 \quad \text{for } \alpha_i \geq 0$$

KKT (III) empty

$\Rightarrow w^*$ determined by x_i for
i such that $f_i(w^*) = 0$
otherwise $\alpha_i = 0$!