## 54.13 Soft Margin SVM; (SVM $_{s4}$ )

In this section we consider the version of Problem (SVM<sub>s2'</sub>) in which instead of using the function  $K\left(\sum_{i=1}^{p} \epsilon_i + \sum_{j=1}^{q} \xi_j\right)$  as a regularizing function we use the quadratic function  $K(\|\epsilon\|_2^2 + \|\xi\|_2^2)$ .

Soft margin SVM (SVM $_{s4}$ ):

minimize 
$$\frac{1}{2}w^{\top}w + (p+q)K_s \left(-\nu\eta + \frac{1}{p+q}(\epsilon^{\top}\epsilon + \xi^{\top}\xi)\right)$$
subject to 
$$w^{\top}u_i - b \ge \eta - \epsilon_i, \qquad i = 1, \dots, p$$
$$-w^{\top}v_j + b \ge \eta - \xi_j, \qquad j = 1, \dots, q$$
$$\eta \ge 0,$$

where  $\nu$  and  $K_s$  are two given positive constants. As we saw earlier, theoretically, it is convenient to pick  $K_s = 1/(p+q)$ . When writing a computer program, it is preferable to assume that  $K_s$  is arbitrary. In this case  $\nu$  needs to be replaced by  $(p+q)K_s\nu$  in all the formulae obtained with  $K_s = 1/(p+q)$ .

The new twist with this formulation of the problem is that if  $\epsilon_i < 0$ , then the corresponding inequality  $w^{\top}u_i - b \geq \eta - \epsilon_i$  implies the inequality  $w^{\top}u_i - b \geq \eta$  obtained by setting  $\epsilon_i$  to zero while reducing the value of  $\|\epsilon\|^2$ , and similarly if  $\xi_j < 0$ , then the corresponding inequality  $-w^{\top}v_j + b \geq \eta$  obtained by setting  $\xi_j$  to zero while reducing the value of  $\|\xi\|^2$ . Therefore, if  $(w, b, \epsilon, \xi)$  is an optimal solution of Problem (SVM<sub>s4</sub>), it is not necessary to restrict the slack variables  $\epsilon_i$  and  $\xi_j$  to the nonnegative, which simplifies matters a bit. In fact, we will see that for an optimal solution,  $\epsilon = \lambda/(2K_s)$  and  $\xi = \mu/(2K_s)$ . The variable  $\eta$  can also be determined by expressing that the duality gap is zero.

One of the advantages of this methods is that  $\epsilon$  is determined by  $\lambda$ ,  $\xi$  is determined by  $\mu$ , and  $\eta$  and b are determined by  $\lambda$  and  $\mu$ . This method *does not* require support vectors to compute b. We can omit the constraint  $\eta \geq 0$ , because for an optimal solution it can be shown using duality that  $\eta \geq 0$ ; see Section 54.14.

A drawback of Program (SVM<sub>s4</sub>) is that for fixed  $K_s$ , the quantity  $\delta = \eta/||w||$  and the hyperplanes  $H_{w,b}$ ,  $H_{w,b+\eta}$  and  $H_{w,b-\eta}$  are independent of  $\nu$ . This will be shown in Theorem 54.8. Thus this method is less flexible than (SVM<sub>s2</sub>) and (SVM<sub>s3</sub>).