

In Section 54.7 we investigate conditions on  $\nu$  that ensure that some point  $u_{i_0}$  and some point  $v_{j_0}$  is a support vector. Theorem 54.3 shows that for every optimal solution  $(w, b, \eta, \epsilon, \xi)$  of Problem (SVM<sub>s2'</sub>) with  $w \neq 0$  and  $\eta > 0$ , if

$$\max\{2p_f/(p+q), 2q_f/(p+q)\} < \nu < \min\{2p/(p+q), 2q/(p+q)\},$$

then some  $u_{i_0}$  and some  $v_{j_0}$  is a support vector. Under the same conditions on  $\nu$  Proposition 54.4 shows that  $\eta$  and  $b$  can always be determined in terms of  $(\lambda, \mu)$  using a single support vector.

- (3) **Soft margin  $\nu$ -SVM Problem (SVM<sub>s3</sub>).** This is the variation of Problem (SVM<sub>s2'</sub>) obtained by adding the term  $(1/2)b^2$  to the objective function. The result is that in minimizing the Lagrangian to find the dual function  $G$ , not just  $w$  but also  $b$  is determined. We also suppress the constraint  $\eta \geq 0$  which turns out to be redundant. If  $\nu > (p_f + q_f)/(p + q)$ , then  $\eta$  is also determined. The fact that  $b$  and  $\eta$  are determined by the dual seems to be an advantage of Problem (SVM<sub>s3</sub>).

The optimization problem is

$$\begin{aligned} & \text{minimize} \quad \frac{1}{2}w^\top w + \frac{1}{2}b^2 + (p+q)K_s \left( -\nu\eta + \frac{1}{p+q} (\epsilon^\top \quad \xi^\top) \mathbf{1}_{p+q} \right) \\ & \text{subject to} \\ & \quad w^\top u_i - b \geq \eta - \epsilon_i, \quad \epsilon_i \geq 0 \quad i = 1, \dots, p \\ & \quad -w^\top v_j + b \geq \eta - \xi_j, \quad \xi_j \geq 0 \quad j = 1, \dots, q. \end{aligned}$$

Theoretically it is convenient to assume that  $K_s = 1/(p+q)$ . Otherwise,  $\nu$  needs to be replaced by  $(p+q)K_s\nu$  in all the formulae below.

It is shown in Section 54.13 that the dual is given by

**Dual of the Soft margin  $\nu$ -SVM Problem (SVM<sub>s3</sub>):**

$$\begin{aligned} & \text{minimize} \quad \frac{1}{2}(\lambda^\top \quad \mu^\top) \left( X^\top X + \begin{pmatrix} \mathbf{1}_p \mathbf{1}_p^\top & -\mathbf{1}_p \mathbf{1}_q^\top \\ -\mathbf{1}_q \mathbf{1}_p^\top & \mathbf{1}_q \mathbf{1}_q^\top \end{pmatrix} \right) \begin{pmatrix} \lambda \\ \mu \end{pmatrix} \\ & \text{subject to} \\ & \quad \sum_{i=1}^p \lambda_i + \sum_{j=1}^q \mu_j = \nu \\ & \quad 0 \leq \lambda_i \leq K_s, \quad i = 1, \dots, p \\ & \quad 0 \leq \mu_j \leq K_s, \quad j = 1, \dots, q. \end{aligned}$$