Once a solution for  $\lambda$  and  $\mu$  is obtained, we have

$$w = -X \binom{\lambda}{\mu} = \sum_{i=1}^{p} \lambda_i u_i - \sum_{j=1}^{q} \mu_j v_j$$
$$b = -\sum_{i=1}^{p} \lambda_i + \sum_{j=1}^{q} \mu_j.$$

Note that the constraint

$$\sum_{i=1}^{p} \lambda_i - \sum_{j=1}^{q} \mu_j = 0$$

occurring in the dual of Program (SVM<sub>s2'</sub>) has been traded for the equation

$$b = -\sum_{i=1}^{p} \lambda_i + \sum_{j=1}^{q} \mu_j$$

determining b.

If  $\nu > (p_f + q_f)/(p + q)$ , then  $\eta$  is determined by expressing that the duality gap is zero. We obtain

$$\begin{split} ((p+q)\nu - p_f - q_f)\eta &= (p_f - q_f)b + w^\top \bigg( \sum_{j \in K_\mu} v_j - \sum_{i \in K_\lambda} u_i \bigg) \\ &+ \frac{1}{K_s} \begin{pmatrix} \lambda^\top & \mu^\top \end{pmatrix} \bigg( X^\top X + \begin{pmatrix} \mathbf{1}_p \mathbf{1}_p^\top & -\mathbf{1}_p \mathbf{1}_q^\top \\ -\mathbf{1}_q \mathbf{1}_p^\top & \mathbf{1}_q \mathbf{1}_q^\top \end{pmatrix} \bigg) \begin{pmatrix} \lambda \\ \mu \end{pmatrix}. \end{split}$$

In practice another way to compute  $\eta$  is to assume the Standard Margin Hypothesis for (SVM<sub>s3</sub>). Under the **Standard Margin Hypothesis** for (SVM<sub>s3</sub>), either some  $u_{i_0}$  is a support vector of type 1 or some  $v_{j_0}$  is a support vector of type 1. By the complementary slackness conditions  $\epsilon_{i_0} = 0$  or  $\xi_{j_0} = 0$ , so we have

$$w^{\top} u_{i_0} - b = \eta$$
, or  $-w^{\top} v_{j_0} + b = \eta$ ,

and we can solve for  $\eta$ . As in (SVM<sub>s2'</sub>) we get more numerically stable formulae by averaging over the sets  $I_{\lambda}$  and  $I_{\mu}$ .

Proposition 54.5 gives bounds  $\nu$ , namely

$$\frac{p_f + q_f}{p + q} \le \nu \le \frac{p_m + q_m}{p + q}.$$

In Section 54.11 we investigate conditions on  $\nu$  that ensure that either there is some blue support vector  $u_{i_0}$  or there is some red support vector  $v_{j_0}$ . Theorem 54.7 shows