Then it is easy to see that we can compute b and η using the following averaging formulae:

$$b = w^{\top} \left(\left(\sum_{i \in I_{\lambda}} u_i \right) / |I_{\lambda}| + \left(\sum_{j \in I_{\mu}} v_j \right) / |I_{\mu}| \right) / 2$$
$$\eta = w^{\top} \left(\left(\sum_{i \in I_{\lambda}} u_i \right) / |I_{\lambda}| - \left(\sum_{j \in I_{\mu}} v_j \right) / |I_{\mu}| \right) / 2.$$

The "kernelized" version of Problem (SVM_{s2'}) is the following:

Soft margin kernel SVM (SVM $_{s2'}$):

minimize
$$\frac{1}{2}\langle w, w \rangle - K_m \eta + K_s \left(\epsilon^{\top} \quad \xi^{\top} \right) \mathbf{1}_{p+q}$$

subject to $\langle w, \varphi(u_i) \rangle - b \geq \eta - \epsilon_i, \quad \epsilon_i \geq 0 \qquad i = 1, \dots, p$
 $-\langle w, \varphi(v_j) \rangle + b \geq \eta - \xi_j, \quad \xi_j \geq 0 \qquad j = 1, \dots, q$
 $\eta > 0.$

Tracing through the derivation of the dual program we obtain **Dual of the Soft margin kernel SVM** (SVM_{s2'}):

minimize
$$\frac{1}{2} \begin{pmatrix} \lambda^{\top} & \mu^{\top} \end{pmatrix} \mathbf{K} \begin{pmatrix} \lambda \\ \mu \end{pmatrix}$$
subject to
$$\sum_{i=1}^{p} \lambda_{i} - \sum_{j=1}^{q} \mu_{j} = 0$$
$$\sum_{i=1}^{p} \lambda_{i} + \sum_{j=1}^{q} \mu_{j} \geq K_{m}$$
$$0 \leq \lambda_{i} \leq K_{s}, \quad i = 1, \dots, p$$
$$0 \leq \mu_{i} \leq K_{s}, \quad j = 1, \dots, q,$$

where \mathbf{K} is the kernel matrix of Section 54.1.

As in Section 54.3, we obtain

$$w = \sum_{i=1}^{p} \lambda_i \varphi(u_i) - \sum_{j=1}^{q} \mu_j \varphi(v_j),$$