Dual of the Soft margin kernel SVM (SVM $_{s4}$):

minimize
$$\frac{1}{2} \begin{pmatrix} \lambda^{\top} & \mu^{\top} \end{pmatrix} \begin{pmatrix} \mathbf{K} + \frac{1}{2K_s} I_{p+q} \end{pmatrix} \begin{pmatrix} \lambda \\ \mu \end{pmatrix}$$
subject to
$$\sum_{i=1}^{p} \lambda_i - \sum_{j=1}^{q} \mu_j = 0$$
$$\sum_{i=1}^{p} \lambda_i + \sum_{j=1}^{q} \mu_j \ge \nu$$
$$\lambda_i \ge 0, \quad i = 1, \dots, p$$
$$\mu_j \ge 0, \quad j = 1, \dots, q,$$

where **K** is the kernel matrix of Section 54.1. Then w, b, and f(x) are obtained exactly as in Section 54.5.

54.14 Solving SVM (SVM_{s4}) Using ADMM

In order to solve (SVM_{s4}) using ADMM we need to write the matrix corresponding to the constraints in equational form,

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

with $K_m = (p+q)K_s\nu$. This is the $2 \times (p+q+1)$ matrix A given by

$$A = \begin{pmatrix} \mathbf{1}_p^\top & -\mathbf{1}_q^\top & 0 \\ \mathbf{1}_p^\top & \mathbf{1}_q^\top & -1 \end{pmatrix}.$$

We leave it as an exercise to prove that A has rank 2. The right-hand side is

$$c = \begin{pmatrix} 0 \\ K_m \end{pmatrix}.$$

The symmetric positive semidefinite $(p+q)\times(p+q)$ matrix P defining the quadratic functional is

$$P = X^{\top}X + \frac{1}{2K_c}I_{p+q}$$
, with $X = \begin{pmatrix} -u_1 & \cdots & -u_p & v_1 & \cdots & v_q \end{pmatrix}$,

and

$$q = 0_{p+q}.$$