

This is the classical problem discussed in all books on machine learning or pattern analysis, for instance Vapnik [182], Bishop [23], and Shawe–Taylor and Christianini [159]. The trivial solution where all variables are 0 is ruled out because of the presence of the 1 in the inequalities, but it is not clear that if  $(w, b, \epsilon, \xi)$  is an optimal solution, then  $w \neq 0$ .

We prove that if the primal problem has an optimal solution  $(w, \epsilon, \xi, b)$  with  $w \neq 0$ , then  $w$  is determined by any optimal solution  $(\lambda, \mu)$  of the dual. We also prove that there is some  $i$  for which  $\lambda_i > 0$  and some  $j$  for which  $\mu_j > 0$ . Under a mild hypothesis that we call the **Standard Margin Hypothesis**,  $b$  can be found.

Note that this framework is still somewhat sensitive to outliers because the penalty for misclassification is linear in  $\epsilon$  and  $\xi$ .

First we write the constraints in matrix form. The  $2(p+q) \times (n+p+q+1)$  matrix  $C$  is written in block form as

$$C = \begin{pmatrix} X^\top & -I_{p+q} & \begin{smallmatrix} \mathbf{1}_p \\ -\mathbf{1}_q \end{smallmatrix} \\ 0_{p+q,n} & -I_{p+q} & 0_{p+q} \end{pmatrix},$$

where  $X$  is the  $n \times (p+q)$  matrix

$$X = \begin{pmatrix} -u_1 & \cdots & -u_p & v_1 & \cdots & v_q \end{pmatrix},$$

and the constraints are expressed by

$$\begin{pmatrix} X^\top & -I_{p+q} & \begin{smallmatrix} \mathbf{1}_p \\ -\mathbf{1}_q \end{smallmatrix} \\ 0_{p+q,n} & -I_{p+q} & 0_{p+q} \end{pmatrix} \begin{pmatrix} w \\ \epsilon \\ \xi \\ b \end{pmatrix} \leq \begin{pmatrix} -\mathbf{1}_{p+q} \\ 0_{p+q} \end{pmatrix}.$$

The objective function  $J(w, \epsilon, \xi, b)$  is given by

$$J(w, \epsilon, \xi, b) = \frac{1}{2} w^\top w + K \begin{pmatrix} \epsilon^\top & \xi^\top \end{pmatrix} \mathbf{1}_{p+q}.$$

The Lagrangian  $L(w, \epsilon, \xi, b, \lambda, \mu, \alpha, \beta)$  with  $\lambda, \alpha \in \mathbb{R}_+^p$  and with  $\mu, \beta \in \mathbb{R}_+^q$  is given by

$$\begin{aligned} L(w, \epsilon, \xi, b, \lambda, \mu, \alpha, \beta) &= \frac{1}{2} w^\top w + K \begin{pmatrix} \epsilon^\top & \xi^\top \end{pmatrix} \mathbf{1}_{p+q} \\ &\quad + \begin{pmatrix} w^\top & \epsilon^\top & \xi^\top & b \end{pmatrix} C^\top \begin{pmatrix} \lambda \\ \mu \\ \alpha \\ \beta \end{pmatrix} + \begin{pmatrix} \mathbf{1}_{p+q}^\top & 0_{p+q}^\top \end{pmatrix} \begin{pmatrix} \lambda \\ \mu \\ \alpha \\ \beta \end{pmatrix}. \end{aligned}$$