

Consequently, if we set  $\nabla L_{w,\epsilon,b,\xi,\xi'} = 0$ , we obtain the equations

$$\begin{aligned} w &= \sum_{i=1}^m (\mu_i - \lambda_i) x_i = X^\top (\mu - \lambda), \\ C\nu - \gamma - \sum_{i=1}^m (\lambda_i + \mu_i) &= 0 \\ \sum_{i=1}^m (\lambda_i - \mu_i) &= 0 \\ \frac{C}{m} - \lambda - \alpha &= 0, \quad \frac{C}{m} - \mu - \beta = 0. \end{aligned} \tag{*}_w$$

Substituting the above equations in the second expression for the Lagrangian, we find that the dual function  $G$  is independent of the variables  $\gamma, \alpha, \beta$  and is given by

$$G(\lambda, \mu) = -\frac{1}{2} \sum_{i,j=1}^m (\lambda_i - \mu_i)(\lambda_j - \mu_j) x_i^\top x_j - \sum_{i=1}^m (\lambda_i - \mu_i) y_i$$

if

$$\begin{aligned} \sum_{i=1}^m \lambda_i - \sum_{i=1}^m \mu_i &= 0 \\ \sum_{i=1}^m \lambda_i + \sum_{i=1}^m \mu_i + \gamma &= C\nu \\ \lambda + \alpha &= \frac{C}{m}, \quad \mu + \beta = \frac{C}{m}, \end{aligned}$$

and  $-\infty$  otherwise.

The dual program is obtained by maximizing  $G(\alpha, \mu)$  or equivalently by minimizing  $-G(\alpha, \mu)$ , over  $\alpha, \mu \in \mathbb{R}_+^m$ . Taking into account the fact that  $\alpha, \beta \geq 0$  and  $\gamma \geq 0$ , we obtain the following dual program:

**Dual Program for  $\nu$ -SV Regression:**

$$\begin{aligned} &\text{minimize} \quad \frac{1}{2} \sum_{i,j=1}^m (\lambda_i - \mu_i)(\lambda_j - \mu_j) x_i^\top x_j + \sum_{i=1}^m (\lambda_i - \mu_i) y_i \\ &\text{subject to} \\ &\quad \sum_{i=1}^m \lambda_i - \sum_{i=1}^m \mu_i = 0 \\ &\quad \sum_{i=1}^m \lambda_i + \sum_{i=1}^m \mu_i \leq C\nu \\ &\quad 0 \leq \lambda_i \leq \frac{C}{m}, \quad 0 \leq \mu_i \leq \frac{C}{m}, \quad i = 1, \dots, m, \end{aligned}$$