

Optimization task for SVM (non-separable case)

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Consider the optimization task:

$$\begin{cases} \frac{1}{2} \|w\|^2 + C \sum_{n=1}^N \xi_n \rightarrow \min_{w, w_0, \xi} \\ y_n(w_0 + w^T x_n) \geq 1 - \xi_n \\ \xi_n \geq 0 \end{cases} \quad n = \overline{1, N} \quad (1)$$

Lagrangian becomes

$$L = \frac{1}{2} \|w\|^2 + C \sum_{n=1}^N \xi_n - \sum_{n=1}^N \alpha_n [y_n(w_0 + w^T x_n) - 1 + \xi_n] - \sum_{n=1}^N r_n \xi_n$$

with dual variables $\alpha_n \geq 0, r_n \geq 0 \ n = \overline{1, N}$. Objective function is convex, constraint functions are linear (so both convex and concave) \Rightarrow solution is a saddle point of L . Lagrangian is minimized w.r.t. primal variables w, w_0, ξ_n and minimized w.r.t. dual variables α_n, r_n .

Differentiate w.r.t. primal variables to obtain

$$\frac{\partial L}{\partial w} = w - \sum_{n=1}^N \alpha_n y_n x_n = 0 \quad (2)$$

$$\frac{\partial L}{\partial w_0} = - \sum_{n=1}^N \alpha_n y_n = 0 \quad (3)$$

$$\frac{\partial L}{\partial \xi_n} = C - \alpha_n - r_n = 0 \quad (4)$$

Substituting (2),(3),(4) into Lagrangian, we obtain:

$$\begin{aligned} L &= \frac{1}{2} \sum_{i,j} \alpha_i \alpha_j y_i y_j \langle x_i, x_j \rangle + C \sum_{n=1}^N \xi_n - \left(\sum_{n=1}^N \alpha_n y_n \right) w_0 - \sum_{n=1}^N \alpha_n \sum_{j=1}^N \alpha_j y_j \langle x_j, x_n \rangle + \sum_{n=1}^N \alpha_n (1 - \xi_n) - \sum_{n=1}^N r_n \xi_n \\ &= -\frac{1}{2} \sum_{i,j} \sum \alpha_i \alpha_j y_i y_j \langle x_i, x_j \rangle - \left(\sum_{n=1}^N \alpha_n y_n \right) w_0 + \sum_{n=1}^N (C - r_n - \alpha_n) \xi_n + \sum_{n=1}^N \alpha_n = \\ &= -\frac{1}{2} \sum_{i,j} \sum \alpha_i \alpha_j y_i y_j \langle x_i, x_j \rangle + \sum_{n=1}^N \alpha_n \end{aligned}$$

where we used equations (3) and (4).

Since primal task (in primal variables) was minimization, dual task (in dual variables) will be maximization:

$$\begin{cases} -\frac{1}{2} \sum \sum_{i,j} \alpha_i \alpha_j y_i y_j \langle x_i, x_j \rangle + \sum_{n=1}^N \alpha_n \rightarrow \max_{\alpha} \\ \sum_{n=1}^N \alpha_n y_n = 0 \\ 0 \leq \alpha_n \leq C \end{cases}$$

where first constraint comes from (3) and second - from non-negativity of dual variable α_n , (4) and non-negativity of dual variable r_n . Kernel version of this equation is

$$\begin{cases} -\frac{1}{2} \sum \sum_{i,j} \alpha_i \alpha_j y_i y_j K(x_i, x_j) + \sum_{n=1}^N \alpha_n \rightarrow \max_{\alpha} \\ \sum_{n=1}^N \alpha_n y_n = 0 \\ 0 \leq \alpha_n \leq C \end{cases}$$