

Primal Problem:

$$\min_w \frac{1}{2} \|w\|^2$$

$$\text{s.t. } y^{(i)} (w^T x^{(i)} + b) \geq 1 \quad \forall i \in \{1, 2, \dots, m\}$$

Dual Problem:

$$\max_{\alpha} \theta_D(\alpha) = \sum_{i=1}^m \alpha_i - \frac{1}{2} \sum_{i=1}^m \sum_{j=1}^m \alpha_i \alpha_j y^{(i)} y^{(j)} \langle x^{(i)}, x^{(j)} \rangle$$

$$\text{s.t. } \alpha_i \geq 0 \text{ and } \sum_{i=1}^m \alpha_i y^{(i)} = 0$$

Replace $\langle x^{(i)}, x^{(j)} \rangle$ with kernels

Kernels:

$$\text{linear: } k(x, z) = x^T z + b$$

$$\text{Polynomial: } K(x, z) = (x^T z + b)^d$$

$$\text{Gaussian or rbf: } k(x, z) = \exp\left(-\frac{\|x - z\|^2}{2\sigma^2}\right)$$

$$\text{All } k(x, z) \in \mathbb{R}$$

tells how close given two feature vectors are

SMO: use SMO algo to solve dual Problem

find α 's

$$\max_{\alpha} \theta_D(\alpha_1, \alpha_2, \dots, \alpha_m)$$

~~fix two~~ vary two α 's and fix rest and update α 's.

$$\sum_{i=1}^m \alpha_i y^{(i)} = 0$$

Fix ^{Vary} two alphas at a time.

$$\alpha_i y^{(i)} + \alpha_j y^{(j)} = - \sum_{\substack{k=1 \\ k \neq i, j}}^m \alpha_k y^{(k)} = \text{const} \rightarrow \textcircled{1}$$

$$\text{Let } S = y^{(i)} y^{(j)}$$

Multiplying eqn ① with $y^{(i)}$

$$\alpha_j y^{(i)} y^{(j)} + \alpha_i y^{(i)} y^{(i)} = \gamma$$

$$\alpha_i = \gamma - \alpha_j$$

Dual objective function written as :

$$\begin{aligned} \theta_p(d) = & \alpha_i + \alpha_j + \text{const} - \frac{1}{2} \left(y_i y_i x_i^T x_i \alpha_i \alpha_i + \right. \\ & y_j y_j x_j^T x_j \alpha_j \alpha_j + \\ & 2 y_i y_j x_i^T x_j \alpha_i \alpha_j + \\ & \left. 2 \left(\sum_{\substack{k=1 \\ k \neq i, j}}^m \alpha_k y_k x_k^T \right) (y_i x_i \alpha_i + \right. \\ & \left. y_j x_j \alpha_j) \right. \\ & \left. + \text{const} \right) \end{aligned}$$

Solving :

$$\theta_D(d) = \frac{1}{2} \eta \chi_j^2 + (y_j (E_i^{\text{old}} - E_j^{\text{old}}) - \eta d_j^{\text{old}}) d_j + \text{const}$$

where

$$\eta = 2(k_{ii} - k_{ii} - k_{jj})$$

$$k_{ii} = x_i^T x_i, \quad k_{jj} = x_j^T x_j, \quad k_{ij} = x_i^T x_j$$

$$E_i^{\text{old}} = \frac{1}{2} (x_i^T w^{\text{old}} - b^{\text{old}} - y_i)^2$$

$$E_j^{\text{old}} = \frac{1}{2} (x_j^T w^{\text{old}} - b^{\text{old}} - y_j)^2$$

So the update rule:

$$d_j := d_j + \frac{y_j (E_j^{\text{old}} - E_i^{\text{old}})}{\eta}$$

$$w = \sum_{i=1}^m d_i y_i x_i$$

$$b = \frac{-\max_{i: y_i = -1} w^T x^{(i)} + \min_{i: y_i = 1} w^T x^{(i)}}{2}$$

$$\text{Prediction: } w^T x + b = \sum_{i=1}^m d_i y^{(i)} k(x^{(i)}, x) + b$$