

Soft - Margin SVM

$$\min \frac{1}{2} \|w\|_2^2 + \frac{c}{n} \sum_{i=1}^n s_i$$

$$w^* = \sum_{i=1}^n \alpha_i^* y_i x_i$$

$$y_i (w^T x_i + b) \geq 1 - s_i \quad \forall i$$

$$s_i \geq 0 \quad \forall i$$

Dual Problem

$$\max_{\alpha, \beta} -\frac{1}{2} \sum_{i,j} \alpha_i \alpha_j y_i y_j x_i^T x_j + \sum \alpha_i$$

$$\sum_i \alpha_i y_i = 0$$

$$\alpha_i + \beta_i = \frac{c}{n}$$

$$\alpha_i \geq 0 \quad \beta_i \geq 0 \quad \forall i$$

primal problem

$$\min f(x)$$

$$f_i(x) \leq 0 \quad \forall i$$

Lagrangian

$$L(x, \lambda) = f(x) + \sum_i \lambda_i f_i(x)$$

Dual

$$\max_{\lambda \geq 0} L_0(\lambda)$$

$$\text{Lagrange Dual function } L_D(\lambda) = \min_x L(x, \lambda)$$

Eliminate $\beta_i \Rightarrow \beta_i = \frac{c}{n} - \alpha_i, \beta_i \geq 0 \Rightarrow 0 \leq \alpha_i \leq \frac{c}{n}$

Define $K_{ij} = \langle x_i, x_j \rangle$

$$\text{Diag}(y) = \begin{bmatrix} y_1 & & \\ & \ddots & \\ & & y_n \end{bmatrix}$$

Simplify the dual:

$$\max -\frac{1}{2} \alpha^T \text{Diag}(y) \cdot K \text{Diag}(y) \cdot \alpha + \sum_{i=1}^n \alpha_i$$

$$0 \leq \alpha_i \leq \frac{c}{n} \quad \forall i$$

$$\sum \alpha_i y_i = 0$$

Test sample x

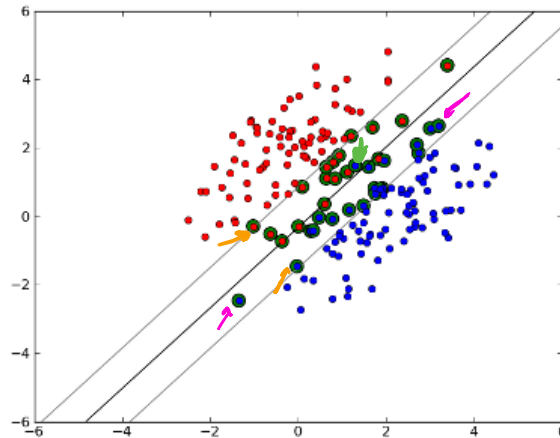
$$\hat{y} \rightarrow \text{sign}(w^{*T} x + b^*) = \text{sign}\left(\sum_i \alpha_i^* y_i \langle x_i, x \rangle + b^*\right)$$

Complementary slackness:

$$(y_i(x^*) - \lambda_i^* = 0)$$

$$\alpha_i^* \cdot (1 - s_i^* - y_i \cdot (w^{*T} x_i + b^*)) = 0$$

If $y_i (w^{\dagger T} x_i + b^{\dagger}) + s_i^{\dagger} = 1$, we call x_i a support vector.



If x_i is not a support vector, we have $\alpha_i^{\dagger} = 0$

$$w^{\dagger} = \sum_{i=1}^n \alpha_i^{\dagger} y_i \underline{x}_i = \sum_{\text{support vectors}} \alpha_i^{\dagger} y_i \underline{x}_i$$

- (a) If $y_i (w^{\dagger T} x_i + b_i) > 1$, then $s_i^{\dagger} = 0$ and x_i is not a SV.
- (b) If $y_i (w^{\dagger T} x_i + b_i) = 1$, then $s_i^{\dagger} = 0$ and x_i is a SV
- (c) If $0 \leq y_i (w^{\dagger T} x_i + b_i) < 1$, then $s_i^{\dagger} > 0$ and x_i is a SV
- (d) If $y_i (w^{\dagger T} x_i + b_i) < 0$ then $s_i^{\dagger} > 0$ and x_i is a SV

How to find b^{\dagger} ?

Find an i s.t. $0 < \alpha_i^{\dagger} < \frac{C}{n}$ which means $\beta_i = \frac{C}{n} - \alpha_i > 0$

Comp. slackness $s_i^{\dagger} = 0$ and $y_i (w^{\dagger T} x_i + b^{\dagger}) = 1 - s_i^{\dagger} = 1$

$$\text{Solve for } b^{\dagger} = \frac{1}{y_i} - w^{\dagger T} x_i = y_i - w^{\dagger T} x_i$$

since $y_i \in \{-1, +1\}$