

Dual Form of SVM

Soft-margin SVM

$$\min_{w, b} \frac{1}{2} \|w\|^2 + c \sum_{i=1}^n \xi_i$$

$$\text{s.t. } y_i (w^T x_i + b) \geq 1 - \xi_i, \forall i$$
$$\xi_i \geq 0$$

Primal
Problem

The Dual Form of SVM is

$$\max_{\alpha_i} \sum_{i=1}^n \alpha_i - \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n \alpha_i \alpha_j y_i y_j x_i^T x_j$$

$$\text{s.t. } \alpha_i \geq 0 \text{ and } \sum_{i=1}^n \alpha_i y_i = 0$$

In SVM instead of solving the Primal Problem, we solve Dual ~~Problem~~ Form

By comparing ~~the~~ both the forms we

can say that:

① For every x_i we have corresponding α_i $x_i \rightarrow \alpha_i$

② x_i and x_j , α_i 's occur only in the form of $\alpha_i^T x_j$

$$③ f(x_q) = \sum_{i=1}^n \alpha_i y_i x_i^T x_q + b$$

$$\therefore x_q \rightarrow (w^T x_q + b) = f(x_q)$$

\downarrow query point \Downarrow y_q \downarrow +ve / -ve

④ $\alpha_i > 0$, only for support vectors.

and $\alpha_i = 0$, for non-support vectors.

$$f(x) = \sum_{i=1}^n \alpha_i y_i x_i^T x + b$$

for SVS $\alpha_i > 0$

for NSVS $\alpha_i = 0$

\therefore to compute $f(x_q)$ the only point that matters is your support vectors.

Hence the behaviour of SVM, the only points that matter is Support Vectors

This is the Reason the whole algorithm is named as support vector machine

Now, let's look over the formulation one more time.

$$\max_{\alpha_i} \sum_{i=1}^n \alpha_i - \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n \alpha_i \alpha_j y_i y_j x_i^T x_j$$

$$\text{s.t. } \alpha_i \geq 0 \rightarrow \alpha_i \geq 0 \text{ for SVs}$$

$$\alpha_i = 0 \text{ for NSVs}$$

$$\sum_{i=1}^n \alpha_i y_i = 0$$