

SMO Algorithm

Initialize $\alpha_i \leftarrow 0$ and $b \leftarrow 0$

Let $f_a(\mathbf{x}) = b + \sum_{i=1}^m y_i \alpha_i k(\mathbf{x}, \mathbf{x}_i)$

Let τ be the tolerance.

Loop

Find two exs. (\mathbf{x}_p, y_p) and (\mathbf{x}_q, y_q) such that:

$$\begin{aligned} & (f_a(\mathbf{x}_p) - y_p + \tau < f_a(\mathbf{x}_q) - y_q - \tau) \wedge \\ & ((\alpha_p < C \wedge y_p = 1) \vee (\alpha_p > 0 \wedge y_p = -1)) \wedge \\ & ((\alpha_q > 0 \wedge y_q = 1) \vee (\alpha_q < C \wedge y_q = -1)) \end{aligned}$$

exit loop if no such examples can be found

$$\eta \leftarrow \frac{(f_a(\mathbf{x}_q) - y_q) - (f_a(\mathbf{x}_p) - y_p)}{k(\mathbf{x}_p, \mathbf{x}_p) - 2k(\mathbf{x}_p, \mathbf{x}_q) + k(\mathbf{x}_q, \mathbf{x}_q)}$$

if needed, reduce η so that:

$$\begin{aligned} & (0 \leq \alpha_p + y_p \eta \leq C) \wedge \\ & (0 \leq \alpha_q - y_q \eta \leq C) \end{aligned}$$

$$\alpha_p \leftarrow \alpha_p + y_p \eta$$

$$\alpha_q \leftarrow \alpha_q - y_q \eta$$

End Loop

$$b_1 \leftarrow \max \{f_a(x_i) \mid y_i = 1 \wedge \alpha_i > 0\}$$

$$b_{-1} \leftarrow \min \{f_a(x_i) \mid y_i = -1 \wedge \alpha_i > 0\}$$

$$b \leftarrow -(b_1 + b_{-1})/2$$

SMO Explanation

SMO means “sequential minimal optimization”.

The conditions on \mathbf{x}_p and \mathbf{x}_q mean:

1. Relative to each other, \mathbf{x}_p is below its margin boundary, and \mathbf{x}_q is above its boundary.
2. The tolerance is a tradeoff between accuracy and efficiency.
3. α_p can be changed to increase $f_a(\mathbf{x}_p)$.
4. α_q can be changed to decrease $f_a(\mathbf{x}_q)$.

To maintain $\sum_{i=1}^m \alpha_i y_i = 0$, an increase in $\alpha_p y_p$ is matched by a decrease in $\alpha_q y_q$.

η optimizes the objective function.

$0 \leq \alpha_i \leq C$ must be maintained.

One way to set the bias weight b is shown.

Many details to make this more efficient and stable have been omitted.