# Soft margins

Hard margins:

for 
$$i = 1, ..., m, y_i(w'x_i + b) \ge 1$$

Soft margins:

for 
$$i = 1, ..., m$$
,  $y_i(w'x_i + b) \ge 1 - \zeta_i$   $\zeta_i \ge 0$ 

#### The primal problem:

Let C be a constant that corresponds to the "amount of allowed softness". The function to be minimized and the linear inequality constraints are augmented to:

Minimize 
$$\frac{1}{2}|w|^2 + C\sum_{i=1}^m \zeta_i$$

subject to the 2m linear inequality constraints:

for 
$$i = 1, ..., m$$
,  $y_i(w'x_i + b) \ge 1 - \zeta_i$ ,  $\zeta_i \ge 0$ 

Intuitively, large values of C would emphasize the requirement that the  $\zeta_i$  are small, and thus decrease the softness.

# Derivation of the dual problem

The Lagrangian of the primal problem:

$$L(w, b, \zeta_1, \dots, \zeta_m, \alpha_1, \dots, \alpha_m, r_1, \dots, r_m) = \frac{1}{2} |w|^2 + C \sum_{i=1}^m \zeta_i + \sum_{i=1}^m \alpha_i (1 - \zeta_i - y_i(w'x_i + b)) - \sum_{i=1}^m r_i \zeta_i$$
 (1)

To compute the dual problem we need to minimize L with respect to  $w, b, \zeta_i$  so that it is a function of only  $\alpha_i, r_i$ .

The derivative of 
$$L$$
 w.r.t.  $w$  gives: 
$$w = \sum_{i=1}^{m} \alpha_i y_i x_i$$
The derivative of  $L$  w.r.t.  $b$  gives: 
$$\sum_{i=1}^{m} \alpha_i y_i = 0$$
The derivative of  $L$  w.r.t.  $\zeta_i$  gives: 
$$C - \alpha_i - r_i = 0$$
(2)

Substituting these values in L and simplifying we get::

$$L(\alpha_1, \dots, \alpha_m, r_1, \dots, r_m) = -\frac{1}{2} \sum_{i=1}^m \sum_{j=1}^m \alpha_i \alpha_j y_i y_j x_i' x_j + \sum_{i=1}^m \alpha_i$$

This is exactly the same dual function as in the hard-margins case. For the dual problem we also need the last two constraints in (2), and  $\alpha_i \geq 0, r_i \geq 0$ . The difference between the hard and the soft case is that from the third equation in (2) and the condition  $r_i \geq 0$  we have:  $\alpha_i \leq C$ .

#### The dual problem:

Maximize 
$$L(\alpha_1, \dots, \alpha_m) = \sum_{i=1}^m \alpha_i - \frac{1}{2} \sum_{i=1}^m \sum_{j=1}^m \alpha_i \alpha_j y_i y_j x_i' x_j$$
  
subject to:  $0 \le \alpha_i \le C$ ,  $\sum_{i=1}^m \alpha_i y_i = 0$ 

This is a quadratic programming problem and we assume that there is a black-box that solves it. The solution gives the values of the  $\alpha_i$ .

### The Karush-Kuhn-Tucker Complementary Conditions

In this case the KKT condition gives:

$$\alpha_i (y_i(w'x_i + b) - 1 + \zeta_i) = 0$$
  
$$\zeta_i(\alpha_i - C) = 0$$

From the second condition it follows that either  $\zeta_i = 0$ , or  $\alpha_i = C$ . Therefore:

$$\begin{array}{lll} \alpha_i = 0 & \to & \text{not support vector} \\ 0 < \alpha_i < C & \to & y_i(w'x_i + b) = 1 & \text{point on hard margin} \\ \alpha_i = C & \to & y_i(w'x_i + b) = 1 - \zeta_i & \text{point on soft margin} \end{array}$$

## Recovering w, b

From (2) it follows that w can be recovered from the support vectors in the same way as in the hard-margins case:

$$w = \sum_{j=1}^{k} \alpha_j y_j x_j \tag{3}$$

Once w is determined the value of b can be computed from any one of the hard margins support vectors (with  $\alpha_i < C$ ), using the same formulas as in the hard-margins case:

$$0 < \alpha_s < C \quad \to \quad b = \frac{1}{y_s} - w' x_s \tag{4.1}$$

As in the hard-margins case it is also possible to compute the value of b from all support vectors on the hard margins (satisfying  $0 < \alpha_s < C$ ). Since the formulas for w, b are the same as in the hard-margins case we can also use kernels.

# The value of $\zeta$

In the hard-margins case the dual optimization problem can give infinite values, indicating that the primal problem has no solution (the data is not linearly separable.) This cannot happen in the soft-margins case. If the point i is wrongly classified by the hyperplane then we can always choose  $\zeta_i = 1 - y_i(w'x_i + b)$ , since this gives  $\zeta \geq 0$  (in fact it gives  $\zeta \geq 1$ ). If the point i is correctly classified but with distance from the margins that is too short, we can still choose  $zeta_i = 1 - y_i(w'x_i + b)$ , since we would still have  $\zeta \geq 0$ . The case in which  $\zeta < 0$  corresponds to points that are correctly classified with  $y_i(w'x_i + b) \geq 1$ , and they are not inside the soft margins.

# Example

i	0	1	2	3	4
$x_i$	0	1	2	3	4
$y_i$	-1	-1	1	-1	1
Lagrangian multiplier	$\alpha_0$	$\alpha_1$	$\alpha_2$	$\alpha_3$	$\alpha_4$

The dual problem:

$$\begin{aligned} \text{maximize} \quad & L(\alpha_0,\dots,\alpha_4) = \alpha_0 + \alpha_1 + \alpha_2 + \alpha_3 + \alpha_4 \\ & -\frac{1}{2}(\alpha_1^2 + 4\alpha_2^2 + 9\alpha_3^2 + 16\alpha_4^2 \\ & - 4\alpha_1\alpha_2 + 6\alpha_1\alpha_3 - 8\alpha_1\alpha_4 - 12\alpha_2\alpha_3 + 16\alpha_2\alpha_4 - 24\alpha_3\alpha_4) \\ & = \alpha_0 + \alpha_1 + \alpha_2 + \alpha_3 + \alpha_4 - \frac{1}{2}(-\alpha_1 + 2\alpha_2 - 3\alpha_3 + 4\alpha_4)^2 \\ \text{subject to: } & 0 \leq \alpha_0 \leq C, \quad 0 \leq \alpha_1 \leq C, \quad 0 \leq \alpha_2 \leq C, \quad 0 \leq \alpha_3 \leq C, \quad 0 \leq \alpha_4 \leq C, \\ & -\alpha_0 - \alpha_1 + \alpha_2 - \alpha_3 + \alpha_4 = 0 \end{aligned}$$

With C=10 the solution (computed by the black box quadratic optimizer) is:  $\alpha_0=0, \ \alpha_1=\alpha_4=3.55, \ \alpha_2=\alpha_3=10$ . Therefore, the support vectors are  $x_1,x_2,x_3,x_4$ . We can now compute w from (3):

$$w = -3.55 + 20 - 30 + 4 * 3.55 = 0.66666$$

The value of b can be computed, for example, from  $x_1$ , the first support vector, using (4.1):

$$b = -1 - 0.666 = -1.666$$

It cannot be computed from  $x_2, x_3$  since they satisfy  $\alpha_i = C$ . It can be computed from  $x_4$ : using (4.1):

$$b = 1 - 0.6666 \times 4 = -1.6666$$

Observe that in this case  $x_2, x_3$  are wrongly classified.

#### **Distances**

In our case the "hyperplane" is the point satisfying w'x+b=0, which is x=2.5. The distance of the hard-margins support vectors from that hyperplane is 1.5. Observe that 1.5/|w|=1, as expected. The  $\zeta$  value for  $x_2$  is 1-(-1/3)=4/3. Its distance from the hyperplane is  $(1-\zeta)/|w|=-1/2$ .