## SUPPORT VECTOR MACHINES CONTINUED.

## 1. SVM FOR OVERLAPPING CLASSES

- In the previous case,  $t_n(\mathbf{w}^t\phi(\mathbf{x}_n) + b) \ge 1$  is a constraint. It is required. What if the classes are not linearly separable in the kernel space? No solution may exist. So, we must allow for misclassification.
- We do this through slack variables,  $\xi$
- The slack variables are  $\xi_n \geq 0$
- If  $\xi_n = 0$  if correctly classified on the boundary or farther away
- $\xi_n = |t_n y(\mathbf{x}_n)|$  for other points. Increases as the data point is more mis-classified.
- So, the new constraints are:

(1) 
$$t_n y(\mathbf{x}_n) \ge 1 - \xi_n \quad \text{where } \xi_n \ge 0$$

• So the objective function become:

(2) 
$$C\sum_{n=1}^{N} \xi_n + \frac{1}{2} \|\mathbf{w}\|^2$$

where C > 0 is a trade-off parameter

• So, our Lagrangian becomes:

(3) 
$$\mathscr{L}(\mathbf{w}, b, \mathbf{a}) = C \sum_{n=1}^{N} \xi_n + \frac{1}{2} \|\mathbf{w}\|^2 - \sum_{n=1}^{N} a_n \{t_n y(x_n) - 1 + \xi_n\} - \sum_{n=1}^{N} \mu_n \xi_n$$

where  $a_n \geq 0$ ,  $\mu_n \geq 0$  are Lagrange multipliers

- What is the meaning of each term?
- So, what are the resulting KKT conditions?
- We now need to optimize with respect to  $\mathbf{w}$ , b, and  $\xi_n$ . How do we do this?

(4) 
$$\tilde{\mathscr{L}}(\mathbf{a}) = \sum_{n=1}^{N} a_n - \frac{1}{2} \sum_{n} \sum_{m} a_n a_m t_n t_m K(\mathbf{x}_n, \mathbf{x}_m)$$

where  $0 \le a_n \le C$  and  $\sum_n a_n t_n = 0$ 

• The constraints are:

(5) 
$$t_n y(\mathbf{x}_n) \ge 1 - \xi_n \quad \text{where } \xi_n \ge 0$$

• So the objective function becomes:

(6) 
$$C\sum_{n=1}^{N} \xi_n + \frac{1}{2} \|\mathbf{w}\|^2$$

where C > 0 is a trade-off parameter

• Our Lagrangian is then:

(7) 
$$\mathscr{L}(\mathbf{w}, b, \mathbf{a}) = C \sum_{n=1}^{N} \xi_n + \frac{1}{2} \|\mathbf{w}\|^2 - \sum_{n=1}^{N} a_n \{t_n y(x_n) - 1 + \xi_n\} - \sum_{n=1}^{N} \mu_n \xi_n$$

where  $a_n \geq 0$ ,  $\mu_n \geq 0$  are Lagrange multipliers

- What is the meaning of each term?
- So, what are the resulting KKT conditions?

(8) 
$$a_n \ge 0$$
  
(9)  $t_n y(x_n) - 1 + \xi_n \ge 0$   
(10)  $a_n (t_n y(x_n) - 1 + \xi_n) = 0$   
(11)  $\xi_n \ge 0$ 

$$\mu_n \ge 0$$

$$\mu_n \xi_n = 0$$

• We now need to optimize with respect to  $\mathbf{w}$ , b, and  $\xi_n$ . How do we do this?

(14) 
$$\frac{\partial \mathcal{L}}{\partial \mathbf{w}} = 0 \to \mathbf{w} = \sum_{n=1}^{N} a_n t_n \phi(\mathbf{x}_n)$$

(15) 
$$\frac{\partial \mathcal{L}}{\partial \mathbf{b}} = 0 \to \sum_{n=1}^{N} a_n t_n = 0$$

(16) 
$$\frac{\partial \mathcal{L}}{\partial \xi} = 0 \to a_n = C - \mu_n$$

• Plug these into the Lagrangian to get the Dual form:

$$\mathscr{L}(\!\!(\mathbf{w},\!\!\!),\mathbf{a}) = C \sum_{n=1}^{N} \xi_n + \frac{1}{2} \|\mathbf{w}\|^2 - \sum_{n=1}^{N} a_n \{t_n y(x_n) - 1 + \xi_n\} - \sum_{n=1}^{N} \mu_n \xi_n$$

(18) 
$$= \sum_{n=1}^{N} \xi_n(C - \mu_n) + \frac{1}{2} \|\mathbf{w}\|^2 - \sum_{n=1}^{N} a_n \left\{ t_n \left( \mathbf{w}^T \phi(\mathbf{x}_n) + b \right) - 1 + \xi_n \right\}$$

(19) 
$$= \sum_{n=1}^{N} \xi_n a_n + \frac{1}{2} \|\mathbf{w}\|^2 - \sum_{n=1}^{N} a_n \left\{ t_n \left( \mathbf{w}^T \phi(\mathbf{x}_n) \right) \right\} - b \sum_{n=1}^{N} a_n t_n + \sum_{n=1}^{N} a_n - \sum_{n=1}^{N} a_n \xi_n \right\}$$

(20) 
$$= \frac{1}{2} \|\mathbf{w}\|^2 - \sum_{n=1}^{N} a_n \left\{ t_n \left( \mathbf{w}^T \phi(\mathbf{x}_n) \right) \right\} + \sum_{n=1}^{N} a_n$$

(21) 
$$= \sum_{n=1}^{N} a_n + \frac{1}{2} \left( \sum_{n=1}^{N} a_n t_n \phi(\mathbf{x}_n) \right)^T \left( \sum_{n=1}^{N} a_n t_n \phi(\mathbf{x}_n) \right)$$

(22) 
$$-\sum_{n=1}^{N} a_n \left\{ t_n \left( \left( \sum_{m=1}^{N} a_m t_m \phi(\mathbf{x}_m) \right)^T \phi(\mathbf{x}_n) \right) \right\}$$

(23) 
$$= \sum_{n=1}^{N} a_n - \frac{1}{2} \sum_{n} \sum_{m} a_n a_m t_n t_m K(\mathbf{x}_n, \mathbf{x}_m)$$

(24) 
$$\tilde{\mathscr{L}}(\mathbf{a}) = \sum_{n=1}^{N} a_n - \frac{1}{2} \sum_{n} \sum_{m} a_n a_m t_n t_m K(\mathbf{x}_n, \mathbf{x}_m)$$

where  $0 \le a_n \le C$  and  $\sum_n a_n t_n = 0$ • Why did we want it in this form?