



$$y(x) = \sum_{i=1}^n a_i t_i k(x, x_i) + b$$

finding  
the decision  
boundary  
hyperplane

you need  
to solve

a function defined on  $x$ -testing  
with training example

$$\max_x \sum_{i=1}^m a_i - \frac{1}{2} \sum_{i=1}^m \sum_{j=1}^m a_i a_j k_i k_j k(x_i, x_j)$$

$$\text{s.t. } a_i \geq 0$$

$$\sum_{i=1}^m a_i t_i = 0$$

$$k(x_i, x_j) = \phi(x_i)^T \phi(x_j)$$

$$b = \frac{1}{N_S} \sum_{i \in S} \left( t_i - \sum_{j \in Z(x_i, x_i)} a_j k_j \right)$$

$$e^{-\frac{\|x_i - x_j\|^2}{2\sigma^2}}$$

$k$  can be any famous kernel

disadvantage: Too many training examples - you  
need to compute  $O(n^2)$   
no. of training examples