



slope of max-margin hyperplane = slope of negative class support vector

$$w = \frac{1-0}{0-1} = -1.$$

y-intercept of max-margin hyperplane = halfway point between positive and negative class support vectors  $\rightarrow b = \frac{3}{2}$

max-margin hyperplane:  $x_2 = -x_1 + \frac{3}{2}$

2. support vector(s) for positive class:  $x^{(1)} = (1, 1)^T$

support vector(s) for negative class:  $x^{(3)} = (1, 0)$ ,  $x^{(4)} = (0, 1)$

3. Original optimization problem:  $\min_{\vec{w}, b} \frac{1}{2} \|\vec{w}\|^2$

constraint:  $(\vec{w} \cdot \vec{x}^{(j)} + b) y^{(j)} \geq 1 \quad \forall j$

rewrite constraint:  $(\vec{w} \cdot \vec{x}^{(j)} + b) y^{(j)} - 1 \geq 0$

Lagrangian:  $L(\vec{w}, b, \vec{\alpha}) = \frac{1}{2} \|\vec{w}\|^2 - \sum_j \alpha_j [(\vec{w} \cdot \vec{x}^{(j)} + b) y^{(j)} - 1]$

4.  $\min_{\vec{w}, b} L(\vec{w}, b, \vec{\alpha}) \rightarrow \begin{cases} \frac{\partial}{\partial \vec{w}} L(\vec{w}, b, \vec{\alpha}) = 0 & (1) \\ \frac{\partial}{\partial b} L(\vec{w}, b, \vec{\alpha}) = 0 & (2) \end{cases}$

Solving (1):

$$\frac{\partial}{\partial \vec{w}} \left[ \frac{1}{2} \|\vec{w}\|^2 - \sum_j \alpha_j [(\vec{w} \cdot \vec{x}^{(j)} + b) y^{(j)} - 1] \right] = 0$$

$$\|\vec{w}\| - \sum_j \alpha_j y^{(j)} \vec{x}^{(j)} = 0$$

$\vec{w} = \sum_j \alpha_j y^{(j)} \vec{x}^{(j)}$

Solving (2):

$$\frac{\partial}{\partial b} \left[ \frac{1}{2} \|\vec{w}\|^2 - \sum_j \alpha_j [(\vec{w} \cdot \vec{x}^{(j)} + b) y^{(j)} - 1] \right] = 0$$

$$0 - \sum_j \alpha_j y^{(j)} = 0$$

$$\boxed{\sum_j \alpha_j y^{(j)} = 0}$$

$$5. \max_{\vec{\alpha}} \left( \min_{\vec{w}, b} \left[ \frac{1}{2} \|\vec{w}\|^2 - \sum_j \alpha_j [(\vec{w} \cdot \vec{x}^{(j)} + b) y^{(j)} - 1] \right] \right)$$

From part 4,  $\vec{w} = \sum_j \alpha_j y^{(j)} \vec{x}^{(j)}$ . Substitute this into the equation above

$$\max_{\vec{\alpha}} \left[ \min_{\vec{w}, b} \left( \frac{1}{2} \|\vec{w}\|^2 - \sum_j \alpha_j y^{(j)} \vec{w} \cdot \vec{x}^{(j)} + \underbrace{\alpha_j y^{(j)} b}_{\rightarrow 0} - \alpha_j \right) \right]$$

from question 4,  $\sum_j \alpha_j y^{(j)} = 0$  so  $\sum_j \alpha_j y^{(j)} b = 0$  as well.

$$\Rightarrow \max_{\vec{\alpha}} \left[ \frac{1}{2} \sum_j \sum_k \alpha_j \alpha_k y^{(j)} y^{(k)} (\vec{x}^{(j)} \cdot \vec{x}^{(k)}) - \right]$$

$$\sum_j \sum_k \alpha_j \alpha_k y^{(j)} y^{(k)} (\vec{x}^{(j)} \cdot \vec{x}^{(k)}) + \sum_j \alpha_j \quad \text{rearrange.}$$

$$\Rightarrow \boxed{\max_{\vec{\alpha}} \left[ \sum_j \alpha_j - \frac{1}{2} \sum_j \sum_k \alpha_j \alpha_k y^{(j)} y^{(k)} (\vec{x}^{(j)} \cdot \vec{x}^{(k)}) \right]} \quad \square$$

6. From the question statement, the constraints in Eq. (1) can be removed for non-support vectors  
 $\hookrightarrow \text{ie } (\vec{w} \cdot \vec{x}^{(j)} + b) y^{(j)} \geq 1$

In Lagrangian format, Eq. (1):

$$\mathcal{L}(\vec{w}, b, \vec{\alpha}) = \frac{1}{2} \|\vec{w}\|^2 - \sum_j \alpha_j [(\vec{w} \cdot \vec{x}^{(j)} + b) y^{(j)} - 1]$$

this is the same constraint from Eq. 1.

If the constraint is disregarded, then  $\sum_j \alpha_j (\vec{w} \cdot \vec{x}^{(j)} + b) y^{(j)} - 1 = 0$  for some  $j^*$ .

Since  $j^*$  isn't a support vector,  $(\vec{w} \cdot \vec{x}^{(j^*)} + b) y^{(j^*)} - 1 \neq 0$ .

Thus the only way the constraint is 0 is if  $\alpha_{j^*} = 0$ .

$\alpha_{j^*} = 0$  when  $j^* = 2$ .

$\square$



7. From question 4:  $\sum_j a_j y^{(j)} = 0$ .

$$\rightarrow a_1 y_1 + a_2 y_2 + a_3 y_3 + a_4 y_4 = 0$$

↙ since  $a_2 = 0$  from question 6.

$$\rightarrow a_1 y_1 + a_3 y_3 + a_4 y_4 = 0$$

Plug in values of  $y_1 = 1, y_3 = -1, y_4 = -1$ :

$$\rightarrow a_1 - a_3 - a_4 = 0$$

$$\boxed{a_1 = a_3 + a_4}$$

8.  $\max_{\vec{a}} \left( \sum_j a_j - \frac{1}{2} \sum_j \sum_k a_j a_k y_j y_k (\vec{x}^{(j)} \cdot \vec{x}^{(k)}) \right)$

Expand the sum and simplify the terms by plugging in the values of  $y_j$  and  $\vec{x}^{(j)}$

Note that the  $a_2$  term was excluded because from part (6), we determined  $a_2 = 0$

$$\max_{\vec{a}} \left( a_1 + a_3 + a_4 - \frac{1}{2} (2a_1^2 + a_3^2 + a_4^2 - 2a_1 a_4 - 2a_1 a_3) \right)$$

$$= \max_{\vec{a}} \left( a_1 + a_3 + a_4 - a_1^2 - \frac{1}{2} a_3^2 - \frac{1}{2} a_4^2 + a_1 a_4 + a_1 a_3 \right)$$

Substitute  $a_1 = a_3 + a_4$  from part (7) and simplify:

$$= \max_{\vec{a}} \left( -\frac{1}{2} (a_3^2 + a_4^2) + 2a_3 + 2a_4 \right)$$

$$\frac{\partial}{\partial a_3} \left( \max_{\vec{a}} (\dots) \right) \Rightarrow -a_3 + 2 = 0 \rightarrow a_3 = 2$$

$$\frac{\partial}{\partial a_4} \left( \max_{\vec{a}} (\dots) \right) \Rightarrow -a_4 + 2 = 0 \rightarrow a_4 = 2$$

$$a_1 = a_3 + a_4 = 2 + 2$$

$$\boxed{\begin{matrix} a_1 = 4 \\ a_3 = 2 \\ a_4 = 2 \end{matrix}}$$

9. From (4),  $\vec{w} = \sum_j a_j y^{(j)} \vec{x}^{(j)}$

$$\begin{aligned}
 &= a_1 y^{(1)} \vec{x}^{(1)} + a_2 y^{(2)} \vec{x}^{(2)} + a_3 y^{(3)} \vec{x}^{(3)} + a_4 y^{(4)} \vec{x}^{(4)} \\
 &= (4 \times 1) \begin{bmatrix} 1 \\ 1 \end{bmatrix} + (2 \times -1) \begin{bmatrix} 1 \\ 0 \end{bmatrix} + (2 \times -1) \begin{bmatrix} 0 \\ 1 \end{bmatrix} \\
 &= \begin{bmatrix} 4 - 2 - 0 \\ 4 - 0 - 2 \end{bmatrix} \\
 &\boxed{\vec{w} = \begin{bmatrix} 2 \\ 2 \end{bmatrix}}
 \end{aligned}$$

10. Constraint:  $(\vec{w} \cdot \vec{x}^{(j)} + b) y^{(j)} \geq 1 \quad \forall j$   
 For support vectors:  $(\vec{w} \cdot \vec{x}^{(j)} + b) y^{(j)} = 1 \quad \forall j$   

$$b = \frac{1}{y^{(j)}} - \vec{w} \cdot \vec{x}^{(j)} \quad \forall j$$

Pick  $j=1$ :  $\vec{x}^{(1)} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ ,  $y^{(1)} = 1$ ,  $\vec{w} = \begin{bmatrix} 2 \\ 2 \end{bmatrix}$

$$\begin{aligned}
 b &= \frac{1}{1} - \begin{bmatrix} 2 \\ 2 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 1 \end{bmatrix} \\
 &= 1 - (2+2)
 \end{aligned}$$

$$\boxed{b = -3}$$

11. The decision boundary takes the form  $\vec{w}^T \vec{x} + b = 0$

Expanding:  $\begin{bmatrix} w_1 & w_2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + b = 0$

$$w_1 x_1 + w_2 x_2 + b = 0$$

$$\rightarrow x_2 = \frac{-1}{w_2} (w_1 x_1 + b)$$

$$\rightarrow x_2 = \frac{-1}{2} (2x_1 - 3)$$

$\rightarrow x_2 = -x_1 + \frac{3}{2}$  which is the same hyperplane determined in question 1.  $\square$