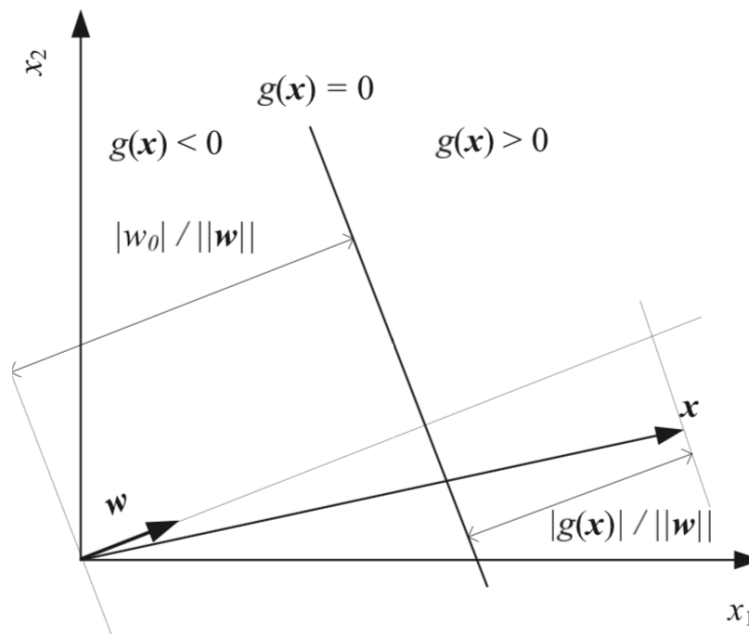


1. For the lecture of linear discriminant, we have:

$$\mathbf{x} = \mathbf{x}_p + r \frac{\mathbf{w}}{\|\mathbf{w}\|}$$

and point  $\mathbf{x}$  is in the corresponding figure.



Point  $\mathbf{x}$  and its projection distance on  $\mathbf{w}$  is  $r$ , we have

$$(10.4) \quad r = \frac{g(\mathbf{x})}{\|\mathbf{w}\|}$$

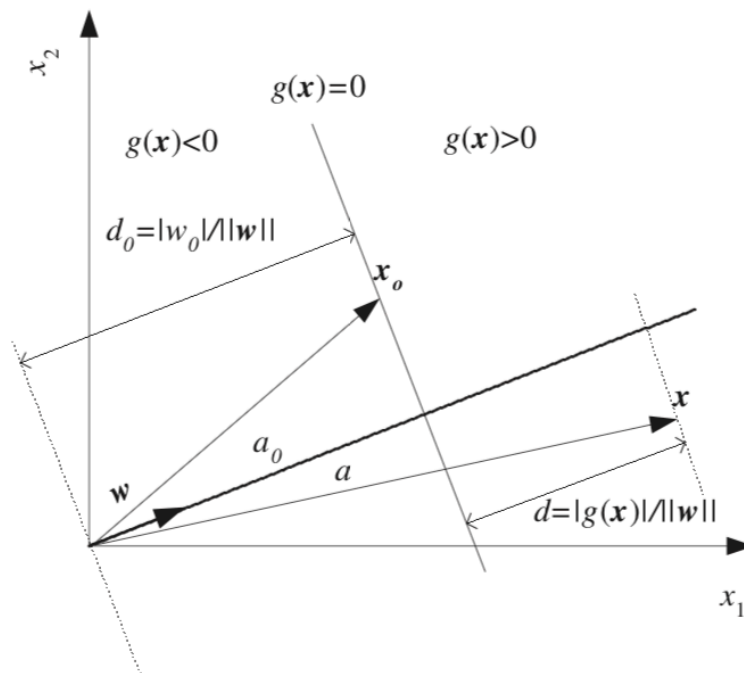
We see then that the distance to origin is

$$(10.5) \quad r_0 = \frac{w_0}{\|\mathbf{w}\|}$$

Prove the equalities of (10.4) and (10.5).

**Answer:**

Given figure 10.1, first let us take input  $\mathbf{x}_0$  on the hyperplane. The angle between  $\mathbf{x}_0$  and  $\mathbf{w}$  is  $a_0$  and because it is on the hyperplane



**Figure 10.1** The geometric interpretation of the linear discriminant.

$g(\mathbf{x}_0) = 0$ . Then

$$g(\mathbf{x}_0) = \mathbf{w}^T \mathbf{x}_0 + w_0 = \|\mathbf{w}\| \|\mathbf{x}_0\| \cos a_0 + w_0 = 0$$

$$d_0 = \|\mathbf{x}_0\| \cos a_0 = \frac{|w_0|}{\|\mathbf{w}\|}$$

For any  $\mathbf{x}$  with angle  $a$  to  $\mathbf{w}$ , similarly we have

$$g(\mathbf{x}) = \mathbf{w}^T \mathbf{x} + w_0 = \|\mathbf{w}\| \|\mathbf{x}\| \cos a + w_0$$

$$d = \|\mathbf{x}\| \cos a - \frac{w_0}{\|\mathbf{w}\|} = \frac{g(\mathbf{x}) - w_0 + w_0}{\|\mathbf{w}\|} = \frac{g(\mathbf{x})}{\|\mathbf{w}\|}$$

2. Show that the derivative of the softmax,  $y_i = \exp(a_i) / \sum_j \exp(a_j)$ , is  $\partial y_i / \partial a_j = y_i(\delta_{ij} - y_j)$ , where  $\delta_{ij}$  is 1 if  $i = j$  and 0 otherwise.

**Answer:**

Given that

$$y_i = \frac{\exp a_i}{\sum_j \exp a_j}$$

for  $i = j$ , we have

$$\begin{aligned}\frac{\partial y_i}{\partial a_i} &= \frac{\exp a_i \left( \sum_j \exp a_j \right) - \exp a_i \exp a_j}{\left( \sum_j \exp a_j \right)^2} \\ &= \frac{\exp a_i}{\sum_j \exp a_j} \left( \frac{\sum_j \exp a_j - \exp a_i}{\sum_j \exp a_j} \right) \\ &= y_i (1 - y_i)\end{aligned}$$

and for  $i \neq j$ , we have

$$\begin{aligned}\frac{\partial y_i}{\partial a_j} &= \frac{-\exp a_i \exp a_j}{\left( \sum_j \exp a_j \right)^2} \\ &= - \left( \frac{\exp a_i}{\sum_j \exp a_j} \right) \left( \frac{\sum_j \exp a_j}{\sum_j \exp a_j} \right) \\ &= y_i (0 - y_j)\end{aligned}$$

which we can combine in one equation as

$$\frac{\partial y_i}{\partial a_i} = y_i (\delta_{ij} - y_j)$$