

4.2.2. Theory

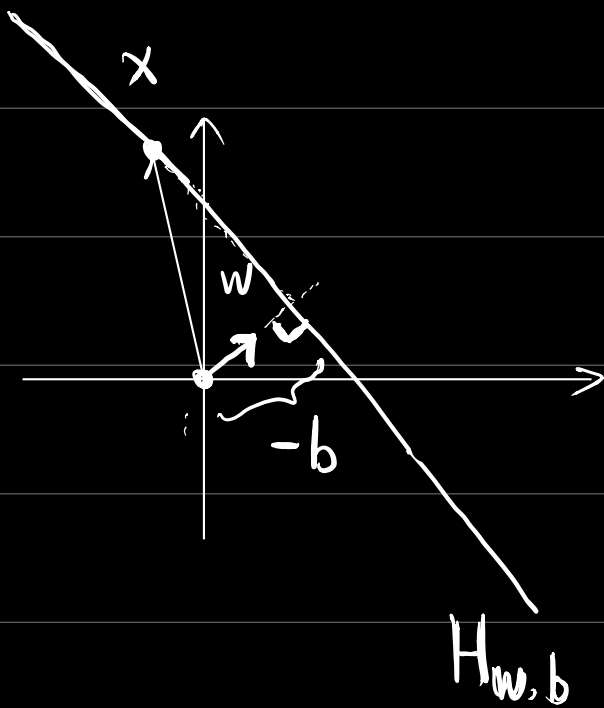
Recall that given a unit vector $w \in \mathbb{R}^p$ with $\|w\|=1$ and $b \in \mathbb{R}$. Then

$$H_{w,b} \stackrel{\text{def}}{=} \{x \in \mathbb{R}^p : x \cdot w + b = 0\}$$

$$= \{x \in \mathbb{R}^p : \underbrace{x \cdot w}_{\text{magnitude of proj of } x \text{ onto } w} = -b\}$$

magnitude of proj of x onto w .

represents a hyperplane that is orthogonal to w and of distance $|b|$ to the origin.



Remark:

(i) For any const $C \neq 0$,

$$H_{cw, cb} = \{x \in \mathbb{R}^p : cw \cdot x + cb = 0\}$$

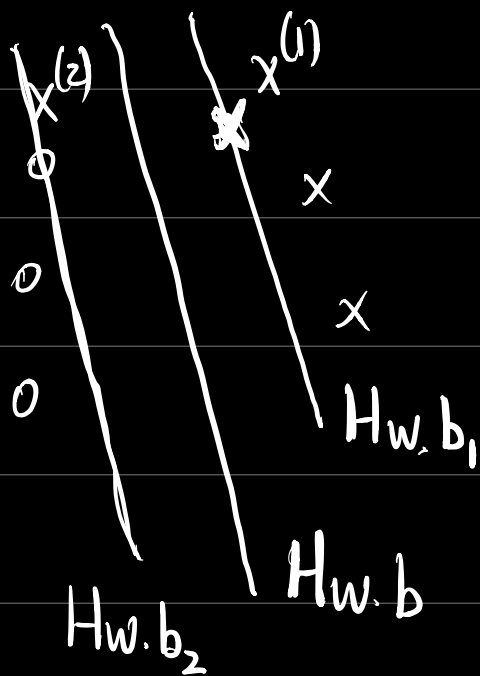
$$= \{x \in \mathbb{R}^T : w \cdot x + b = 0\} = H_{w,b}$$

In particular, $H_{-w,-b} = H_{w,b}$.

(2) By varying b , we get a series of parallel hyperplanes. By varying w , we get hyperplanes with different orientations.

(3) $H_{w,b}$ correctly classifies the sample pts

\Leftrightarrow for each i ,



$$\begin{cases} x^{(i)} \cdot w + b \geq 0 & \text{if } y_i = 1 \\ x^{(i)} \cdot w + b < 0 & \text{if } y_i = -1 \end{cases}$$

Without loss of generality, suppose

$$x^{(1)} \cdot w + b = \min_{\{i: y_i = 1\}} x^{(i)} \cdot w + b$$

i.e. $x^{(1)}$ is the "x" closest to $H_{w,b}$.

$$x^{(2)} \cdot w + b = \max_{\{i: y_i = -1\}} x^{(i)} \cdot w + b$$

i.e. $x^{(2)}$ is the "o" closest to $H_{w,b}$.

By varying b , $\exists b_1$ such that $x^{(1)} \in H_{w,b_1}$

$$(\text{i.e. } x^{(1)} \cdot w + b_1 = 0)$$

and $x^{(i)}$'s with $y_i = 1$ are "above" H_{w,b_1} .

$$\exists b_2 \text{ such that } x^{(2)} \in H_{w,b_2}$$

$$(\text{i.e. } x^{(2)} \cdot w + b_2 = 0)$$

and $x^{(i)}$'s with $y_i = -1$ are "below" H_{w,b_2} .

Note in the picture $-b_1 > -b_2$, i.e. $b_1 < b_2$.

Consider the hyperplane $H_{w, \frac{b_1+b_2}{2}} = H_{\tilde{w}, \tilde{b}}$

where $\tilde{w} = \frac{2}{b_2-b_1} w$, $\tilde{b} = \frac{2}{b_2-b_1} \cdot \frac{b_1+b_2}{2} = \frac{b_1+b_2}{b_2-b_1}$

Then (1) it is most stable with respect to perturbations of inputs.

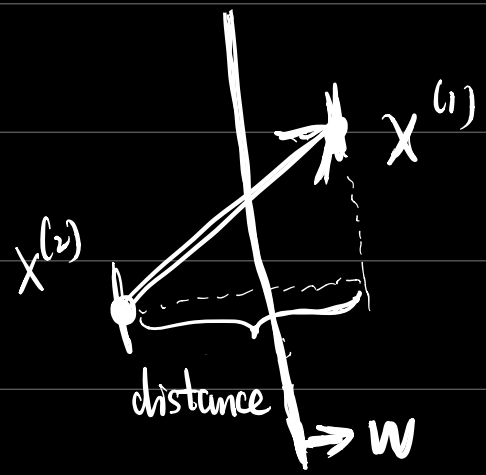
(2) it is equally distant from H_{w, b_1} and H_{w, b_2} .

(3) The distance between H_{w, b_1} and H_{w, b_2} ("margin")

$$= | (x^{(1)} - x^{(2)}) \cdot w |$$

eqns of H_{w, b_1} and H_{w, b_2}

$$= |-b_1 + b_2| = b_2 - b_1$$



$$\|w\| = \left\| \frac{z}{b_2 - b_1} w \right\| = \frac{z}{b_2 - b_1}$$

$$\frac{z}{\|w\|}$$