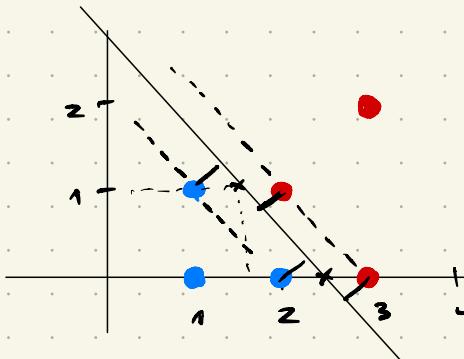


2



• '+, class

• '- , class

- b) Inspection: The line above, passing from $(2.5, 0)$ and $(1.5, 1)$ satisfies $x_1 + x_2 = 2.5$. Thus, parametrically our separating hyperplane is:

$$\omega^T z + b = 0, \text{ where}$$

$$\omega_0 = \omega_1 = 1, \quad b = -2.5 \quad \Rightarrow \quad \|\omega\| = \sqrt{2}$$

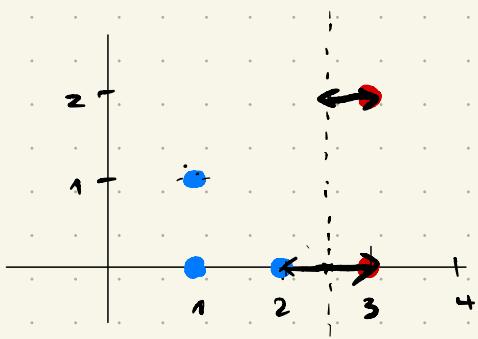
$$\tilde{\gamma}^{((2,1))} = \tilde{\gamma}^{((2,0))} = \tilde{\gamma}^{((1,1))} = \tilde{\gamma}^{((3,0))} = \frac{0.5}{\sqrt{2}}$$

$$\tilde{\gamma}^{((1,0))} = \frac{1.5}{\sqrt{2}}$$

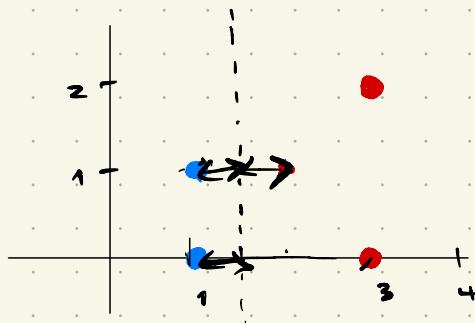
$$\tilde{\gamma}^{((3,1))} = \frac{2.5}{\sqrt{2}}$$

Support vectors: $(2,0), (1,1), (2,1), (3,0)$, Margin = $\frac{0.5}{\sqrt{2}}$

- c) Remove $(2,1), (2,0)$ margin increases
 " $(1,1), (3,0)$ " stays the same

without $(2,1)$

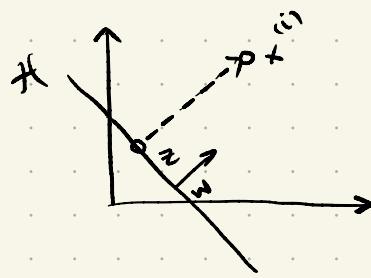
$$\text{margin} = 1 > \frac{0.5}{\sqrt{2}}$$

without $(2,0)$

$$\text{margin} = 1 > \frac{0.5}{\sqrt{2}}$$

3

Let $\mathcal{H} = \{x \mid w^T x + b = 0\}$ denote the hyperplane defined through w and b .



Let $x^{(i)}$ a training example. Let z be the projection of $x^{(i)}$ onto \mathcal{H} . Since $x^{(i)} - z \perp \mathcal{H}$ and $w \perp \mathcal{H}$ $\forall \alpha \in \mathbb{R}$:

$$x^{(i)} - z = \alpha w. \text{ Note that}$$

$$|\alpha| = \frac{|x^{(i)} - z|}{\|w\|} \Leftrightarrow \alpha = \pm \frac{|x^{(i)} - z|}{\|w\|} \Leftrightarrow \alpha = \pm \frac{\delta^{(i)}}{\|w\|}$$

$$\pm \frac{(x^{(i)} - z)}{\delta^{(i)}} = \frac{w}{\|w\|}, \text{ Note that } \alpha = \frac{\delta^{(i)}}{\|w\|} \text{ when } y^{(i)} = +1$$

$$\text{and } \alpha = -\frac{\delta^{(i)}}{\|w\|} \text{ when } y^{(i)} = -1.$$

$$\frac{y^{(i)}(x^{(i)} - z)}{\delta^{(i)}} = \frac{w}{\|w\|} \stackrel{w^T}{\Leftrightarrow} \frac{y^{(i)}(w^T x^{(i)} - w^T z)}{\delta^{(i)}} = \frac{w^T w}{\|w\|}$$

$$\frac{y^{(i)}(w^T x^{(i)} + b)}{\delta^{(i)}} = \frac{\cancel{\|w\|^2}}{\cancel{\|w\|}}, \text{ where we used } z \in \mathcal{H}.$$

$$\text{Thus, } \delta^{(i)} = \frac{y^{(i)}(w^T x^{(i)} + b)}{\|w\|}$$

□

$$\textcircled{4} \quad \text{Hard margin: } \min_{w,b} w^T w$$

(*)

$$\text{s.t. } \forall i \quad y_i(w^T x_i + b) \geq 1$$

a) If w^*, b^* optimal for (*):

Claim: $\exists i \text{ s.t. } y_i(w^* T x_i + b^*) = 1$

Proof: Suppose that $y_i(w^* T x_i + b^*) > 1, \forall i$. Then, $\exists \varepsilon > 0$:

$$y_i(w^* T x_i + b^*) \geq 1 + \varepsilon, \forall i$$

$$y_i \left[\frac{w^* T x_i + b^*}{1 + \varepsilon} \right] \geq 1, \forall i. \quad \text{Denote } \tilde{w} = \frac{w^*}{1 + \varepsilon}, \tilde{b} = \frac{b^*}{1 + \varepsilon}$$

$y_i(\tilde{w}^T x_i + \tilde{b}) \geq 1, \forall i$. But since (w^*, b^*) optimal

$$\text{for (*), then } w^* T w^* = |w^*|^2 \leq \tilde{w}^T \tilde{w} = |\tilde{w}|^2, \text{ or}$$

$$|w^*| \leq |\tilde{w}| = \frac{|w^*|}{1 + \varepsilon},$$

which entails, $1 + \varepsilon \leq 1$ which cannot hold as $\varepsilon > 0$.

By contradiction the claim holds. So, $\exists i_0$ s.t.

$$y_{i_0}(w^* T x_{i_0} + b^*) = 1 \Rightarrow \frac{y_{i_0}(w^* T x_{i_0} + b^*)}{|w^*|} = \frac{1}{|w^*|} \Rightarrow$$

$$s^{(i_0)} = \frac{1}{|w^*|}, \text{ where } s^{(i_0)} \text{ is the dist of } x_{i_0}$$

from the optimal hyperplane $\{z | w^* T z + b^* = 0\}$.

④ b) (z, d) separating hyperplane with margin M

$$z' = \frac{z}{|z| \cdot M}, \quad d' = \frac{d}{|z| \cdot M}$$

$$|z'| = \frac{1}{M}$$

$$\forall i \quad \frac{y_i(z^T x_i + d)}{|z|} \geq M \Leftrightarrow \forall i \quad \frac{y_i(z'^T x_i + d')}{|z'|} \geq 1$$

$$\forall i \quad y_i \left(\frac{z^T}{|z| \cdot M} x_i + \frac{d}{|z| \cdot M} \right) \geq 1 \Leftrightarrow \forall i \quad y_i (z'^T x_i + d') \geq 1,$$

so (z', d') provides a feasible solution.

c) Since (w^*, b^*) is optimal for (★).

$$|w^*|^2 \leq |z'|^2 \Leftrightarrow |w^*| \leq |z'| \Leftrightarrow \frac{1}{|w^*|} \geq \frac{1}{|z'|} = M$$

$$\frac{1}{|w^*|} \geq M$$

$\frac{1}{|w^*|}$ was the margin for (w^*, b^*) from a) and

M is the margin for (z, d) , so we have proven

the statement.

5) a) Suppose there is a solution (w_0, w_1, b) . Then, it must satisfy:

$$\begin{array}{l} (1) \quad w_0 + b \geq 1 \\ (2) \quad 3w_0 + 2w_1 + b \geq 1 \\ (3) \quad -w_0 - 2w_1 - b \geq 1 \\ (4) \quad -3w_0 - b \geq 1 \end{array}$$

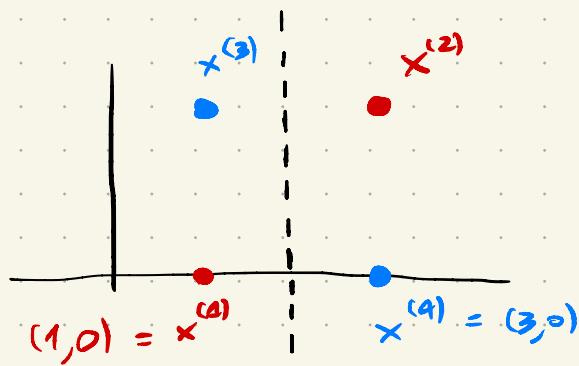
$$(1) + (4): -2w_0 \geq 2 \Rightarrow w_0 \leq -1$$

$$(2) + (3): 2w_0 \geq 2 \Rightarrow w_0 \geq 1$$

*

b) $\min_{w, b, \xi} w^T w + C \sum_{i=1}^4 \xi^{(i)}$ s.t. $\xi_i \geq 0, \forall i$ (***)

$$y_i(w^T x_i + b) \geq 1 - \xi_i, \forall i$$



This is a possible approach (not optimal). Let

$$\left. \begin{array}{l} w_0 = 1 \\ w_1 = 0 \\ b = -2 \end{array} \right\} \text{(see the dotted line above)}$$

Constraints:

$$\left. \begin{array}{l} w_0 + b \geq 1 - \xi^{(1)} \\ 3w_0 + 2w_1 + b \geq 1 - \xi^{(2)} \\ -w_0 - 2w_1 - b \geq 1 - \xi^{(3)} \\ -3w_0 - b \geq 1 - \xi^{(4)} \end{array} \right\}$$

$$\begin{aligned} -1 \geq 1 - \xi^{(1)} &\Rightarrow \xi^{(1)} \geq 2 \\ 1 \geq 1 - \xi^{(2)} &\Rightarrow \xi^{(2)} \geq 0 \\ 1 \geq 1 - \xi^{(3)} &\Rightarrow \xi^{(3)} \geq 0 \\ -1 \geq 1 - \xi^{(4)} &\Rightarrow \xi^{(4)} \geq 2 \end{aligned}$$

Thus, for $C = 1$ in (***) , we have that the min occurs at :

$$(w, b, \xi) = (1, 0, -2, 2, 0, 0, 2)$$