

FODA L26

the Perceptron Algorithm for Linear Classification

Nov 29, 2022

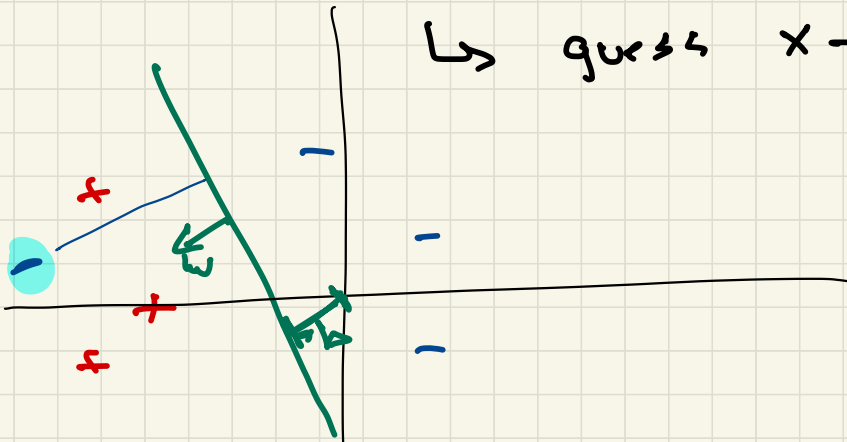
Linear Classification

Input $(X, y) \subset \mathbb{R}^d \times \{-1, +1\}$
data point (x_i, y_i) ← label $\{-1, +1\}$

Output linear classifier $(w, b) \in \mathbb{R}^d \times \mathbb{R}$

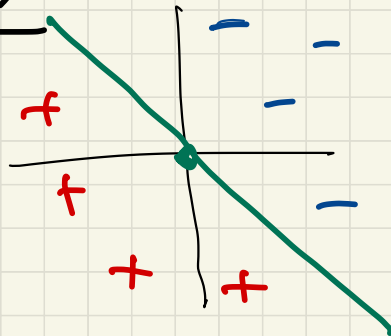
$$g_{w,b}(x) = b + \langle w, x \rangle$$

↳ guess $x \rightarrow +1$ if $g_{w,b}(x) > 0$
 $\rightarrow -1$ if $g_{w,b}(x) < 0$



Simplification for Perceptron

① Assume decision boundary passes through origin



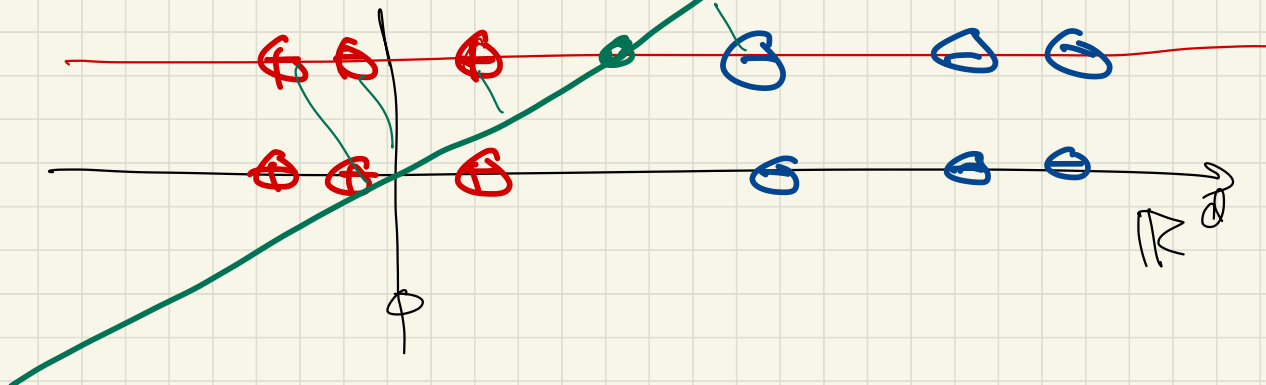
$$b = 0$$

$$g_w(x) = \langle x, w \rangle$$

convert

$$x \rightarrow (1; x) \in \mathbb{R}^{d+1}$$

$$(b, w) = (\alpha_0, \alpha_1 \dots \alpha_d) = \alpha \in \mathbb{R}^{d+1}$$

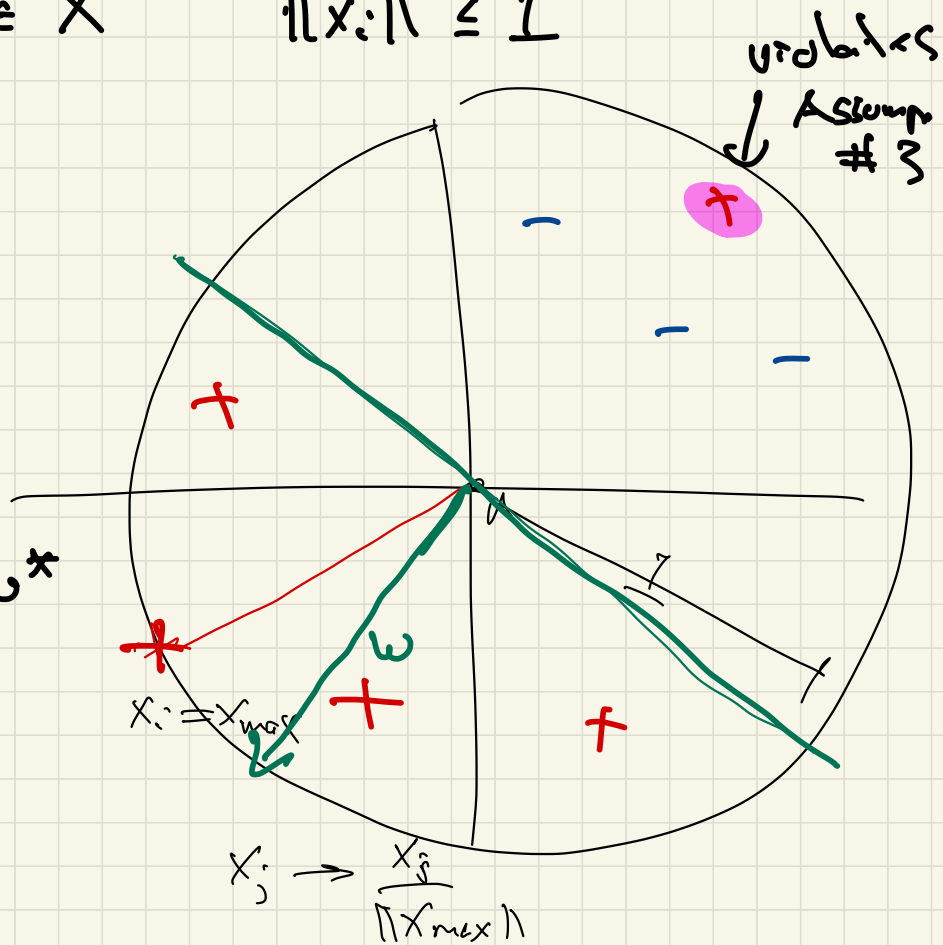


② Assume $\forall x_i \in X \quad \|x_i\| \leq 1$

③ Assume there exist a "perfect" classifier w^*

s. that $\forall (x_i, y_i) \in (X, Y)$

$$\text{sign}(g_{w^*}(x_i)) = y_i$$



Perceptron Algorithm

0. Initialize : Set $w = y_i x_i$ for any $(x_i, y_i) \in (X, y)$ $\in \mathbb{R}^d$

1. repeat

for any $(x_i, y_i) \in (X, y)$ s.t. $y_i \langle x_i, w \rangle < 0$
misclassified

update $w \leftarrow w + y_i x_i$

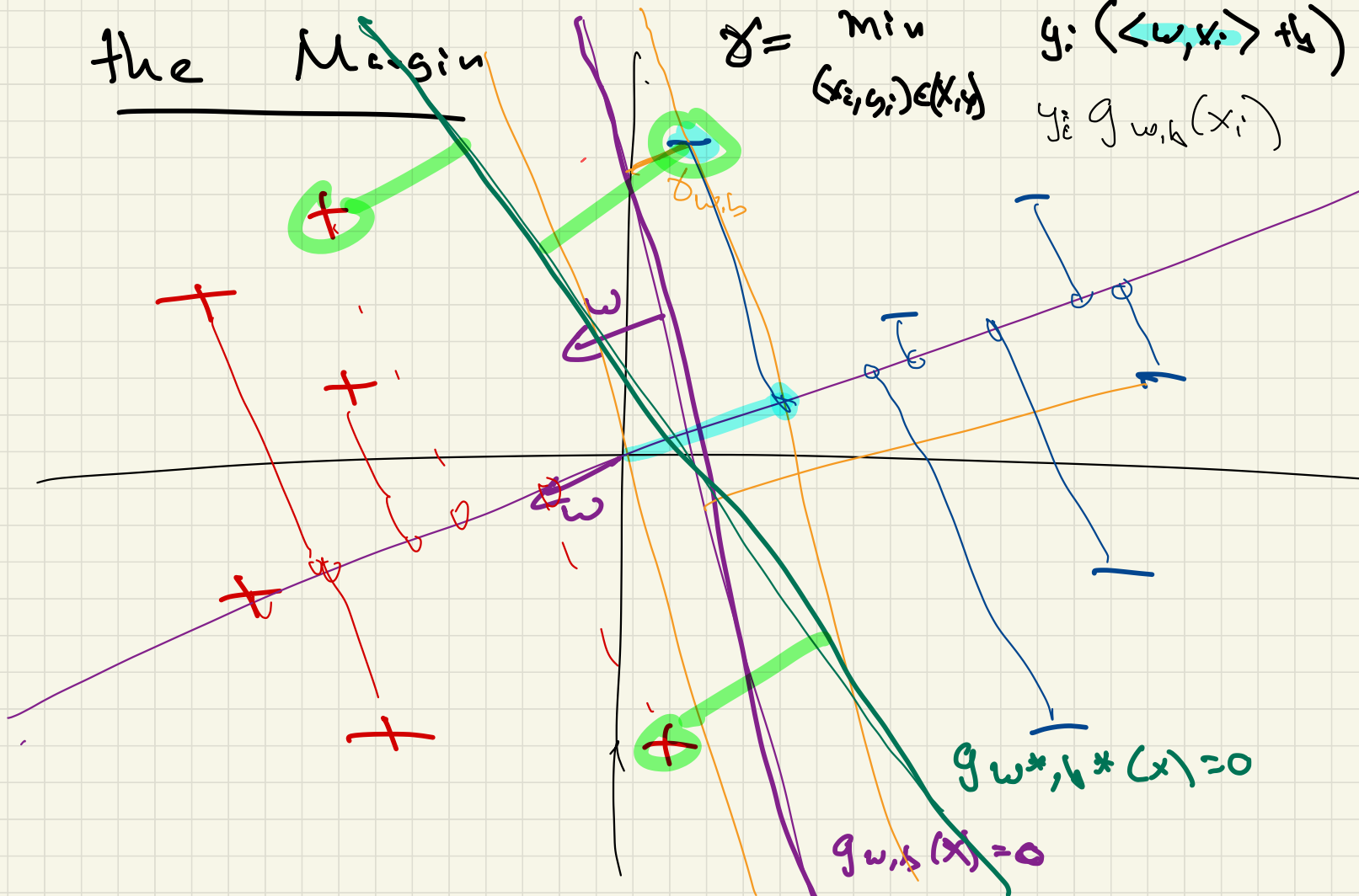
until (no point misclassified)
or T steps

2. return $w \leftarrow \frac{w}{\|w\|}$

the Margin

$$J = \min_{(w, b)} \sum_{i=1}^n \ell(x_i, y_i)$$

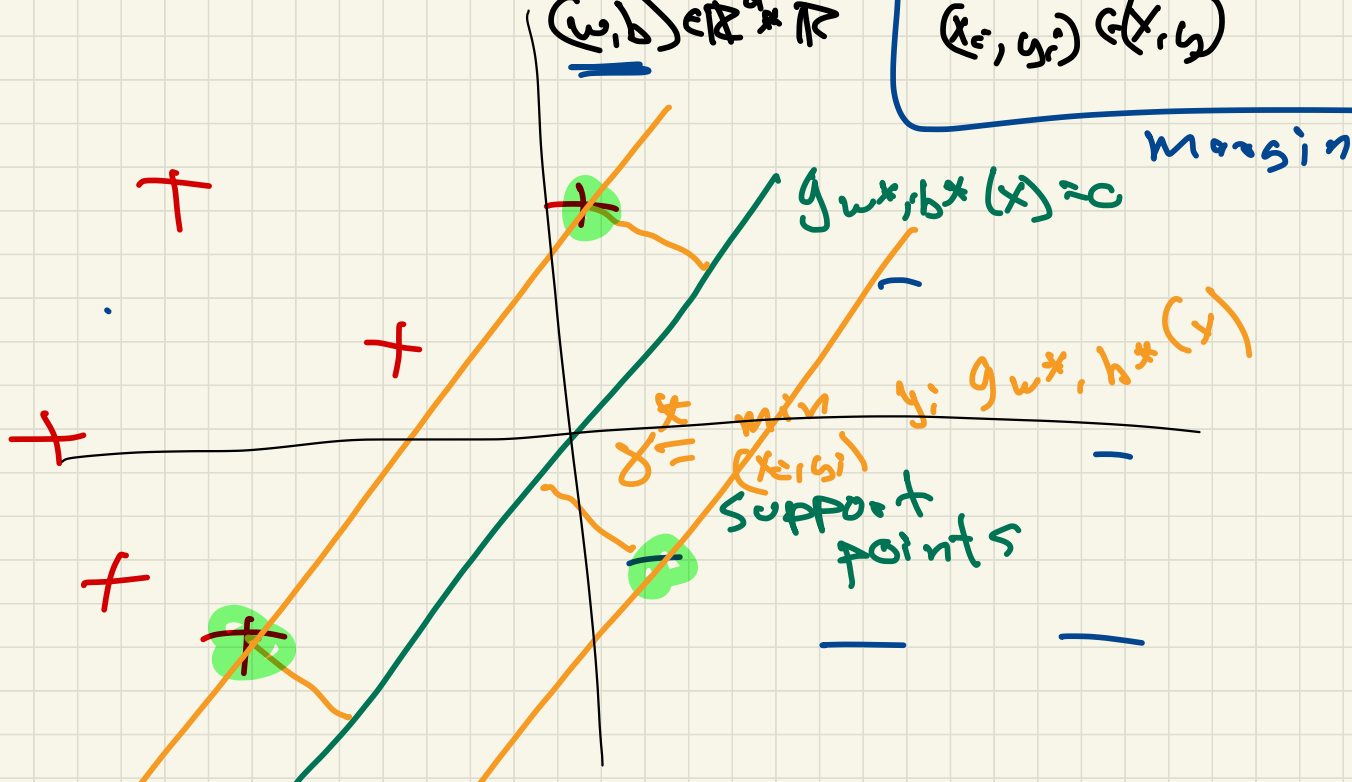
$$y_i = \langle w, x_i \rangle + b$$
$$y_i \in \{g_{w,b}(x_i)\}$$

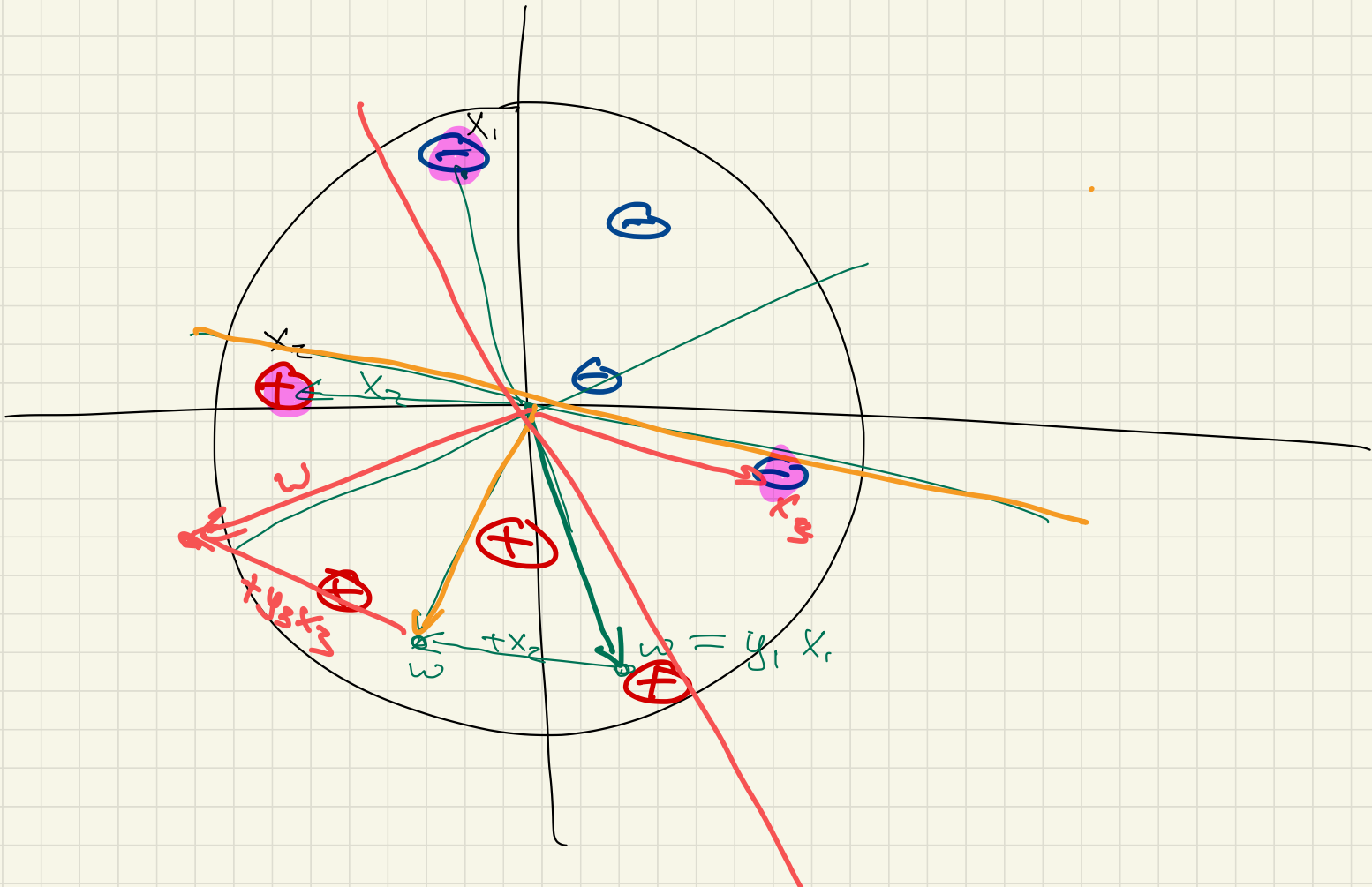


Max-Margin Separator

$$\omega^*, b^* = \arg \max_{(\omega, b) \in \mathbb{R}^d \times \mathbb{R}}$$

$$\min_{(x_i, y_i) \in \mathcal{X} \times \mathcal{Y}} y_i g_{\omega, b}(x_i)$$





Why does Perceptron work?

uses at most $T = (1/\gamma^*)^2$ steps

Max-margin

① $\langle w, w^* \rangle$

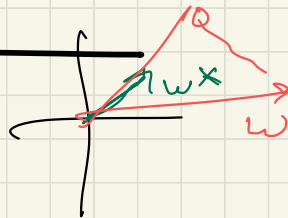
② $\|w\|^2 \leq t \Rightarrow \|w\| \leq \sqrt{t}$

$$\langle w + y_i x_i, w + y_i x_i \rangle = \langle w, w \rangle + \underbrace{\left(\frac{1}{\gamma^*}\right)^2}_{\leq 1} \langle x_i, x_i \rangle + \underbrace{2 y_i \langle w, x_i \rangle}_{\leq 0}$$
$$\leq \langle w, w \rangle + 1$$

$\|w\|^2 \leq t$ after t step

$\langle w, w^* \rangle > t \gamma^*$

$$\langle w + y_i x_i, w^* \rangle = \langle w, w^* \rangle + \frac{2 \gamma^*}{1}$$



$t \gamma^* < \langle w, w^* \rangle \leq \langle w, \frac{w}{\|w\|} \rangle = \|w\| \leq \sqrt{t}$

$t \leq (1/\gamma^*)^2 \Rightarrow$ Most for min after $(1/\gamma^*)^2$ steps