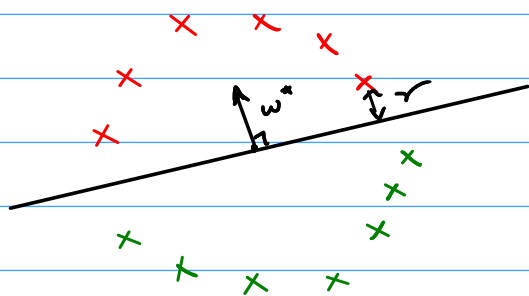


Correction TDS

Exercise 9: Perceptron with margin



$$\gamma = \sup_{\substack{w \in \mathbb{R}^n \\ \|w\|=1}} \min_{1 \leq i \leq n} \frac{y_i \langle w, x_i \rangle}{\|x_i\|}$$

1. By definition, $\exists (w_t)$ st $\forall t, \|w_t\|=1$, $\min_{1 \leq i \leq n} \frac{y_i \langle w_t, x_i \rangle}{\|x_i\|} \xrightarrow{n \rightarrow \infty} \gamma$

$\{w : \|w\|=1\}$ is compact. Indeed, it is bounded by definition and is equal to $\|\cdot\|^{-1}(\{1\})$ so it is a closed set.

Since we work in finite dimension, this proves the compactness.

Hence, $\exists p: \mathbb{N} \rightarrow \mathbb{N}$ strictly increasing such that $w_{p(t)} \rightarrow w^*$ and $\|w^*\|=1$

Furthermore, $w \mapsto \min_{1 \leq i \leq n} \frac{y_i \langle w, x_i \rangle}{\|x_i\|}$ is continuous,

$$\text{So, } \gamma = \min_{1 \leq i \leq n} \frac{y_i \langle w^*, x_i \rangle}{\|x_i\|}$$

$$\text{So, } \forall 1 \leq i \leq n, \frac{y_i \langle w^*, x_i \rangle}{\|x_i\|} \geq \gamma$$

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Perceptron with margin γ :

Input: Margin γ

Data: training set $(x_1, y_1), \dots, (x_n, y_n)$

$w_0 \leftarrow (0, \dots, 0)$

$t \geq 0$

while $\exists i_t : y_{i_t} \langle w_t, x_{i_t} \rangle \leq \frac{\gamma}{2} \|x_{i_t}\| \|w_t\|$

$$\left[\begin{array}{l} w_{t+1} = w_t + y_{i_t} \frac{x_{i_t}}{\|x_{i_t}\|} \\ t \leftarrow t+1 \end{array} \right.$$

return w_t

Let's prove that $\forall t, \langle w^*, w_t \rangle \geq \gamma t$

• $t=0$ $\langle w^*, w_0 \rangle = 0 = 0\gamma$

• Let's suppose that $\langle w^*, w_{t-1} \rangle \geq \gamma(t-1)$ for a given $t \geq 1$ and w_t passed the test.

$$w_t = w_{t-1} + y_{i_{t-1}} \frac{x_{i_{t-1}}}{\|x_{i_{t-1}}\|}$$

$$S_1, \langle w^*, w_t \rangle = \langle w^*, w_{t-1} \rangle + y_{i_{t-1}} \langle w^*, x_{i_{t-1}} \rangle / \|x_{i_{t-1}}\|$$

$$\geq \underbrace{\gamma(t-1)}_{\text{rec}} + \underbrace{\gamma}_{w^*}$$

$$\geq \gamma t$$

which proves that $\forall t$ that passes the test, $\langle w^*, w_t \rangle \geq \gamma t$

$$\begin{aligned}
 \underline{3.} \quad \|\omega_{t+1}\|^2 &= \left\| \omega_t + y_{i_t} \frac{x_{i_t}}{\|x_{i_t}\|} \right\|^2 \\
 &= \|\omega_t\|^2 + 2 y_{i_t} \underbrace{(\omega_t, x_{i_t}) / \|x_{i_t}\|}_{\leq \|\omega_t\| \frac{\sigma}{2} \text{ by test}} + y_{i_t}^2 \frac{\|x_{i_t}\|^2}{\|x_{i_t}\|^2}
 \end{aligned}$$

$$\leq \|\omega_t\|^2 + \sigma \|\omega_t\| + 1$$

$$\underline{\text{So, } \|\omega_{t+1}\|^2 \leq \|\omega_t\|^2 + \sigma \|\omega_t\| + 1}$$

$$\underline{4.} \quad \text{So, } \|\omega_{t+1}\|^2 \leq \left(\|\omega_t\| + \frac{3\sigma}{\gamma} \right)^2 + 1 - \frac{\sigma}{2} \|\omega_t\|$$

$$\underline{\text{And if } \|\omega_t\| \geq 2/\sigma, \|\omega_{t+1}\|^2 \leq \left(\|\omega_t\| + \frac{3\sigma}{\gamma} \right)^2}$$

5 Let's prove it by recurrence:

$$\underline{* t=0} \quad \|\omega_0\| = 0 \leq 1 + \frac{2}{\sigma} + \frac{3\sigma t}{\gamma}$$

* Let's suppose the result true for a given t:

$$* \text{ if } \|\omega_t\| < 2/\sigma,$$

$$\begin{aligned}
 \|\omega_{t+1}\| &= \left\| \omega_t + y_{i_t} \frac{x_{i_t}}{\|x_{i_t}\|} \right\| \leq \|\omega_t\| + |y_{i_t}| \frac{\|x_{i_t}\|}{\|x_{i_t}\|} \\
 &\leq 2/\sigma + 1 \leq 1 + \frac{2}{\sigma} + \frac{3\sigma(t+1)}{\gamma}
 \end{aligned}$$

• if $\|w_t\| \geq 2/\sigma$

$$\|w_{t+1}\| \leq \|w_t\| + \frac{3\sigma}{4}$$

$$\leq 1 + \frac{2}{\sigma} + \frac{3\sigma t}{4} + \frac{3\sigma}{4} \quad \text{by rec}$$

$$\leq 1 + \frac{2}{\sigma} + \frac{3\sigma(t+1)}{4}$$

$$\text{So } \forall t, \|w_t\| \leq 1 + \frac{2}{\sigma} + \frac{3\sigma t}{4}$$

6. So, As long as we stay in the loop, we have,

$$\begin{cases} \sigma t \leq (w^*, w_t) \leq \|w^*\| \|w_t\| = \|w_t\| \\ \|w_t\| \leq 1 + \frac{2}{\sigma} + \frac{3\sigma t}{4} \end{cases}$$

And so we are sure to have a break before

$$\sigma t > 1 + \frac{2}{\sigma} + \frac{3\sigma t}{4}$$

Furthermore, $0 < \sigma \leq 1$ so $t > 12/\sigma^2$
is a sufficient condition.

Conclusion: The algorithm achieves margin at least $\sigma/2$
in at most $12/\sigma^2$ iterations

7. Condition becomes

$$\text{while } \exists i_t : g_{i_t} \langle w_t, x_{i_t} \rangle \leq (1-\gamma) \gamma \|x_{i_t}\| \|w_t\|$$

$$\bullet \langle w^*, w_t \rangle \geq \gamma t \text{ still true}$$

$$\bullet \|w_{t+1}\|^2 \leq \|w_t\|^2 + 2(1-\gamma) \gamma \|w_t\| + 1$$

$$\bullet \text{ if } \|w_t\| \geq \frac{1}{\gamma \eta}, \quad \|w_{t+1}\|^2 \leq \left(\|w_t\| + \frac{2-\gamma}{2} \gamma \right)^2$$

$$\bullet \|w_t\| \leq 1 + \frac{1}{\gamma \eta} + \frac{(2-\gamma)}{2} \gamma t$$

So we know that it stops when $\gamma t \geq 1 + \frac{1}{\gamma \eta} + \frac{(2-\gamma)}{2} \gamma t$

So $t \geq \frac{1}{\gamma^2} \left[\frac{2}{\eta} \left(1 + \frac{1}{\eta} \right) \right]$ is enough.

The algorithm produces a marginal point $(1-\gamma) \gamma$ in at most $K(\gamma) / \gamma^2$ operations with

$$K(\gamma) = \frac{2}{\eta} \left(1 + \frac{1}{\eta} \right)$$