

# The Perceptron Algorithm

( STOP in Finite Steps )

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Theorem (Novikoff)

Let  $S$  be a non-trivial training set, and let  $R = \max_{1 \leq i \leq \ell} \|x^i\|_2$ .

Suppose that there exists a vector  $w_{opt} \in R^n$ ,  $\|w_{opt}\| = 1$

and  $y_i(\langle w_{opt} \cdot x^i \rangle + b_{opt}) \geq \gamma, \forall 1 \leq i \leq \ell$ . Then the number of mistakes made by the on-line perceptron algorithm

on  $S$  is at most  $\left(\frac{2R}{\gamma}\right)^2$ .

# Proof of Finite Termination

Proof: Let

$$\bar{x}^i = \begin{bmatrix} x^i \\ R \end{bmatrix}, \bar{w} = \begin{bmatrix} w \\ \frac{b}{R} \end{bmatrix}.$$

The algorithm starts with an augmented weight vector  $\bar{w}^0 = \mathbf{0}$  and updates it at each mistake.

Let  $\bar{w}^{t-1}$  be the augmented weight vector prior to the  $t$  th mistake. The  $t$  th update is performed when

$$y_i \langle \bar{w}^{t-1}, \bar{x}^i \rangle = y_i (\langle w^{t-1}, x^i \rangle + b_{t-1}) \leq 0$$

where  $(x^i, y_i) \in S$  is the point incorrectly classified by  $\bar{w}^{t-1}$ .

# Update Rule of Perceptron

$$w^t \leftarrow w^{t-1} + \eta y_i x^i ; b_t \leftarrow b_t + \eta y_i R^2$$

$$\bar{w}^t = \begin{bmatrix} w^t \\ \frac{b_t}{R} \end{bmatrix} = \begin{bmatrix} w^{t-1} \\ \frac{b_{t-1}}{R} \end{bmatrix} + \eta y_i \begin{bmatrix} x^i \\ R \end{bmatrix} = \bar{w}^{t-1} + \eta y_i \bar{x}^i.$$

$$\langle \bar{w}^t, \bar{w}_{opt} \rangle = \langle \bar{w}^{t-1}, \bar{w}_{opt} \rangle + \eta y_i \langle \bar{x}^i, \bar{w}_{opt} \rangle$$

$$\geq \langle \bar{w}^{t-1}, \bar{w}_{opt} \rangle + \eta \gamma \geq \langle \bar{w}^{t-2}, \bar{w}_{opt} \rangle + 2\eta \gamma \dots \geq t\eta \gamma$$

(NOTE :  $y_i(\langle w_{opt} \cdot x^i \rangle + b_{opt}) \geq \gamma, \forall 1 \leq i \leq \ell$ )

Similarly,  $\|\bar{w}^t\|_2^2 = \|\bar{w}^{t-1}\|_2^2 + 2\eta y_i \langle \bar{w}^{t-1}, \bar{x}^i \rangle + \eta^2 \|\bar{x}^i\|_2^2$

$$\leq \|\bar{w}^{t-1}\|_2^2 + \eta^2 \|\bar{x}^i\|_2^2 = \|\bar{w}^{t-1}\|_2^2 + \eta^2 (\|x^i\|_2^2 + R^2)$$
$$\leq \|\bar{w}^{t-2}\|_2^2 + 2\eta^2 R^2 \leq \dots \leq 2t\eta^2 R^2$$

# Update Rule of Perceptron

$$w^t \leftarrow w^{t-1} + \eta y_i x^i ; b_t \leftarrow b_t + \eta y_i R^2$$

$$\langle \bar{w}^t, \bar{w}_{opt} \rangle \geq t\eta\gamma \text{ and } \|\bar{w}^t\|_2^2 \leq 2t\eta^2 R^2$$

$$\|\bar{w}_{opt}\|_2 \sqrt{2t} \eta R \geq \|\bar{w}_{opt}\|_2 \|\bar{w}^t\|_2 \geq \langle \bar{w}^t, \bar{w}_{opt} \rangle \geq t\eta\gamma$$

$$\Rightarrow t \leq 2 \left( \frac{R}{\gamma} \right)^2 \|\bar{w}_{opt}\|_2^2 \leq \left( \frac{2R}{\gamma} \right)^2$$

*Note :*  $b_{opt} \leq R, \|\bar{w}_{opt}\|_2^2 \leq \|w_{opt}\|_2^2 + 1 = 2$