

Lecture 6. Perceptron.

◦ online learning:

(1) make predictions while learning.

$$(x_1, y_1) \dots (x_m, y_m)$$

total number of errors.

(2) perceptron:

$$h_{\theta}(x) = g(\theta^T x) \quad g(x) = \begin{cases} 1 & z \geq 0 \\ -1 & z < 0 \end{cases}$$

$$\theta := \theta + yx \quad (y \in \{-1, 1\}) \quad (\text{online})$$

(3) Theorem:

sequence: $(x_1, y_1) \dots (x_m, y_m)$

$$\|x_i\| \leq D \quad \exists u: \|u\|=1. \quad (\text{margin})$$

$$y_i \cdot (u^T x_i) \geq \gamma.$$

$$\sum_{i=1}^m \xi_i \leq \left(\frac{D}{\gamma}\right)^2$$

Proof: θ_k : k -th mistake.

$$x_i^T \theta_k y_i \leq 0$$

Considering the updating rule:

$$(\theta_{k+1})^T u = \theta_k^T u + y_i x_i^T u \geq \theta_k^T u + \gamma$$

$$\geq \theta_{k-1}^T u + 2\gamma \dots \geq k\gamma.$$

$$\|\theta_{k+1}\|^2 \leq \|\theta_k\|^2 + D^2 \Rightarrow \|\theta_{k+1}\|^2 \leq kD^2$$

$$\Leftrightarrow k^2 \gamma^2 \leq kD^2 \Rightarrow k \leq \left(\frac{D}{\gamma}\right)^2 \quad \square$$