Lectures

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1 Rosenblatt's perceptron

1.1 Introduction

Recall: for an input $x \in \mathbb{R}^n$, the parametrized function describing the mapping computed by a perceptron is:

$$F(\omega, x) = F((\omega_1, \omega_2, \dots, \omega_n, \theta), (x_1, x_2, \dots, x_n)) = sgn(\sum_{i=1}^n \omega_i x_i - \theta)$$

Or in the vector form:

$$F(\boldsymbol{\omega}, \boldsymbol{x}) = sgn(\boldsymbol{\omega^T x})$$

This perceptron can only classify patterns which are linearly seprable.

Theorem 1. Suppose $C_1 \bigcup C_2 = C$ are linearly separable classes over the training set $z \in Z^T$ with the assumption $z_t = \{x_t, y_t \in C_1\}$, and perceptron's response r_t with mistake $e_t = r_t - y_t \neq 0$ can be corrected by applying the learning rule to its current state $\omega \in \Omega^T$:

$$\omega_{t+1} = \omega_t + e_t x_t$$

Then perceptron's error correction algorithm converges in k number of steps with following assumptions: training input is bounded by Euclidean norm $\|\mathbf{x}_t\| \leq R$ and $\boldsymbol{\omega}_k^T \mathbf{x}_t \geq \gamma$ for t = 1..T, where $\gamma > 0$. Initial state $\boldsymbol{\omega}_0 = 0$. Note, that γ uses to be sure that some example is classified correctly.

Proof: Expand equation for learning rule at t steps as:

$$\boldsymbol{\omega_{t+1}} = x_1 + x_2 + \ldots + x_t$$

Multiply each side by $\boldsymbol{\omega_k^T}$:

$$\omega_k^T \omega_{t+1} = \omega_k^T x_1 + \omega_k^T x_2 + \ldots + \omega_k^T x_t$$

From the assumption $\boldsymbol{\omega}_{k}^{T} \boldsymbol{x}_{t} \geq \gamma$:

$$\boldsymbol{\omega_k^T \omega_{t+1}} \ge k \gamma$$

From Cauchy-Schwarz inequality:

$$\left|\omega_{k}^{T}\omega_{t+1}
ight|\leq\left\|\omega_{k}^{T}
ight\|\cdot\left\|\omega_{t+1}
ight\|$$