

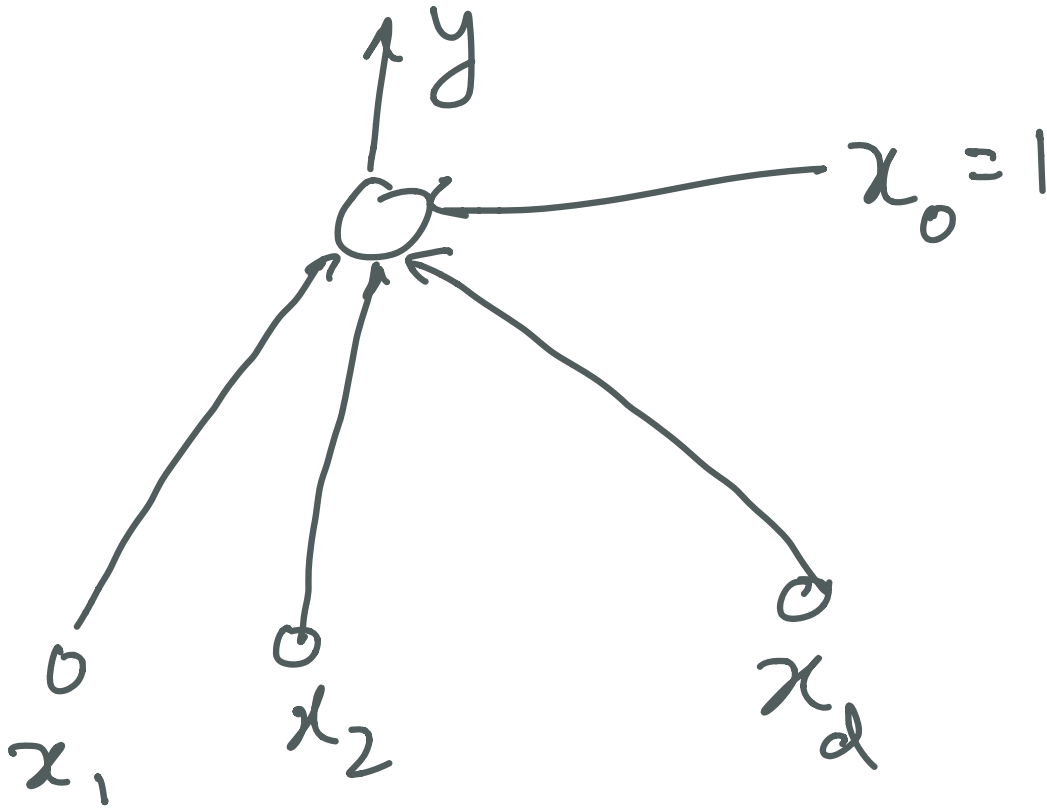
References

1. Pattern Classification

Duda, Hart, Stork

(Artificial Neural Networks

McCulloch-Pitts Neuron



$$y = \text{sign}\left(\sum_{j=0}^d \omega_j x_j\right)$$

$$\mathcal{D} = \left\{ (x^{(i)}, y^{(i)}) \mid \begin{array}{l} x^{(i)} \in \mathcal{X}, y^{(i)} \in \{-1, 1\} \\ i \in [N] \end{array} \right\}$$

Assume that there exist
 ω^* such that

$$\text{sign}(\omega^*{}^T x^{(i)}) = y^{(i)},$$

$$i \in [N].$$

$$\|x^{(i)}\| \leq R$$

$$\omega^{(i)} \geq 0$$

$$i = 1, \dots, N$$

$$\Rightarrow y^{(i)} (\omega^*{}^T x^{(i)}) > 0$$

$$\gamma = \min_i \frac{y^{(i)} (\omega^*{}^T x^{(i)})}{\|\omega^*\|}.$$

$\omega^{(n)}$ be the current estimate.

Let $(x^{(i)}, y^{(i)}) \in \mathcal{D}$ such that

$$\text{sign}((\omega^{(n)})^T x^{(i)}) \neq y_i$$

$$\Rightarrow y^{(i)} (\omega^{(n)T} x^{(i)}) < 0$$

$$\omega^{(n+1)} = \begin{cases} \omega^{(n)} + y^{(i)} x^{(i)} [\text{update}] \\ \omega^{(n)} & \text{otherwise} \end{cases}$$

$$\|\omega^{(n+1)}\|^2 - \|\omega^{(n)}\|^2$$

$$= (\omega^{(n+1)} - \omega^{(n)})^T (\omega^{(n+1)} + \omega^{(n)})$$

$$= (y^{(n)} x^{(n)})^T (2\omega^{(n)} + y^{(n)} x^{(n)})$$

$$= 2 y^{(n)} \omega^{(n)T} x^{(n)} + \|x^{(n)}\|^2$$

$$\|\omega^{(n+1)}\|^2 - \|\omega^{(n)}\|^2 \leq \|x^{(n)}\|^2$$

$$y^{(n)} \omega^{(n)T} x^{(n)} \leq \mathcal{O}$$

$$\begin{aligned}
& (\omega^*)^T (\omega^{(n+1)} - \omega^{(n)}) \\
&= (\omega^*)^T (y^{(n)} x^{(n)}) \\
&\geq \gamma \|\omega^*\|
\end{aligned}$$

Let M be the Number
of updates till there
are no mistakes

After M updates

$$\begin{aligned}
& \sum_{n=1}^M (\omega^*)^T (\omega^{(n+1)} - \omega^{(n)}) \\
&\geq M\gamma \|\omega^*\|
\end{aligned}$$

$$\omega^{*T} (\omega^{(n+1)} - \omega^{(1)}) \geq Mv \|\omega^*\|$$

if we use $\omega^{(1)} = 0$

$$(\omega^*)^T (\omega^{(n+1)}) \geq Mv \|\omega^*\|$$

$$Mv \leq \|\omega^{(n+1)}\| \quad \left(\begin{array}{l} \text{Because} \\ \text{of Cauchy} \\ \text{Schwarz} \end{array} \right)$$

$$\|\omega^{(n+1)}\|^2 \leq MR^2$$

$$M^2 v^2 \leq MR^2$$

$$M \leq \frac{R^2}{\gamma^2} = \frac{R^2 \|w^*\|^2}{\min_i (w^{*T} x^{(i)})}$$

Adapt the proof to $w^{(1)} \neq 0$

Perceptron \rightarrow DHS.