

Perceptron: Toy Example

Perceptron Learning Algorithm in action

1. Dataset

x_1	x_2	y
-1	-1	0
-5	-2.5	0
-7.5	7.5	0
10	7.5	1
-2.5	12.5	0
5	10	1
5	5	1

2. $P \Rightarrow$ Green points, $N \Rightarrow$ Red points

3. Decision boundary line is given by $w_1x_1 + w_2x_2 - b \geq 0$

a. Or $x_2 = -(w_1/w_2)x_1 + (b/w_2)$

b. Can be rewritten as $x_2 = mx + c$

c. Where $m = -(w_1/w_2)$ and $c = (b/w_2)$

4. Initialize w randomly

a. $w_1 = 1.00, w_2 = 1.00, b = 5.00 \Rightarrow m = -1.00, c = 5.00$

5. The line is $x_2 = -x_1 + 5$

6. While !convergence do:

a. Pick random $x \in P \cup N$

b. If $x \in P$ and $\sum_{i=0}^n w_i x_i < 0$ then, $w = w + x$; **end**

c. If $x \in N$ and $\sum_{i=0}^n w_i x_i \geq 0$ then, $w = w - x$; **end**

d. Consider $x = [x_0; x_1; x_2]$ and $w = [w_0; w_1; w_2]$, where x_0 bias term is always 1

e. On the 5th training example, condition c isn't satisfied, so we recalculate $w = w - x$

f. $w_1 = 3.5, w_2 = -11.5, b = 4.00 \Rightarrow m = 0.3, c = -0.35$

g. The line changes, causing a new error on the 6th training example, so we calculate $w = w + x$

h. $w_1 = 8.5, w_2 = -1.5, b = 5.00 \Rightarrow m = 5.67, c = -3.33$

i. The resulting line predicts all examples perfectly, thus convergence is reached