

Artificial Intelligence: Modeling Human Intelligence with Networks

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The Perceptron Algorithm!!

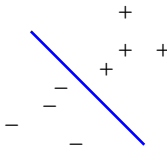
The perceptron

Algorithms

- Set of instructions, typically to solve a class of problems or perform a computation.

From last class

- We had a rule: separate the data using a line,



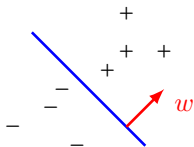
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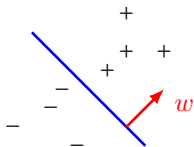
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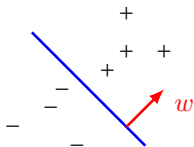
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- The Perceptron Algorithm!

The perceptron

Set up

- In our exercises, we begin with a matrix *data* that contains the data: the points and the classes (negatives or positives).
- The matrix *data* will look like this:

$$data = \begin{bmatrix} 2.1 & 5.2 & 1 \\ 1.1 & -2.7 & -1 \\ 1.4 & 2.2 & -1 \\ \vdots & \vdots & \vdots \\ 3.5 & 1.7 & 1 \end{bmatrix}$$

The perceptron

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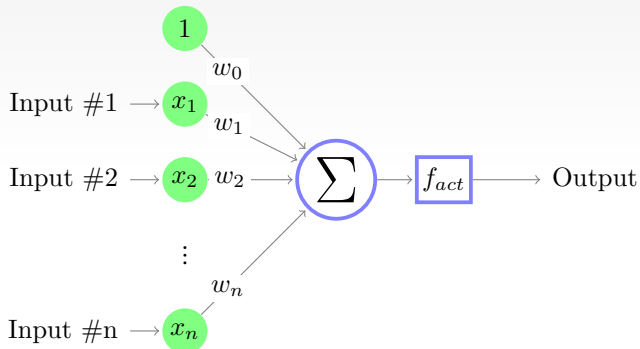
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- We then break it up in two variables: a matrix *X* of points and a vector *classes* of classes:

$$X = \begin{bmatrix} 2.1 & 5.2 \\ 1.1 & -2.7 \\ \vdots & \vdots \\ 3.5 & 1.7 \end{bmatrix}, \quad classes = [1, -1, \dots, 1]$$

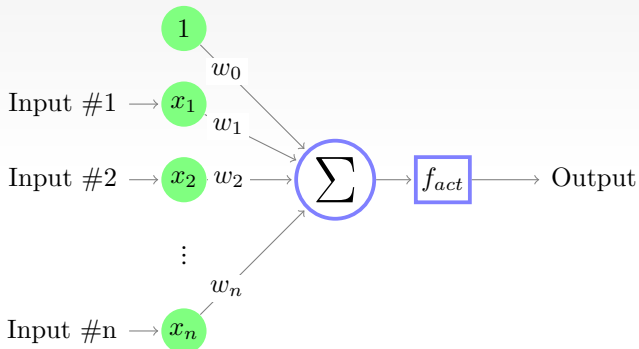
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Remember the neuron model?



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- We need to add a bias...

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Adding the bias

- Each row of X is an input point $x = [x_1, x_2]$ and there are m data points.

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Adding the bias

- Each row of X is an input point $x = [x_1, x_2]$ and there are m data points. We need to add a 1 before them:

$$X = \begin{bmatrix} 2.1 & 5.2 \\ 1.1 & -2.7 \\ \vdots & \vdots \\ 3.5 & 1.7 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 2.1 & 5.2 \\ 1 & 1.1 & -2.7 \\ \vdots & \vdots & \vdots \\ 1 & 3.5 & 1.7 \end{bmatrix}$$

- This can be done in python by a function `np.concatenate()`.

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The Algorithm

Algorithm 1 Perceptron - v2

Input: $X \in \mathbb{R}^{m \times 3}$ and $classes \in \mathbb{R}^m$

Output $w \in \mathbb{R}^3$

```
1: Initialize  $w$ 
2: for  $i = 1$  to  $m$  do
3:    $x = X[i, :]$ 
4:   if  $x \cdot w > 0$  and  $classes[i] == -1$  then
5:      $w = w - x$ 
6:   end if
7:   if  $x \cdot w \leq 0$  and  $classes[i] == 1$  then
8:      $w = w + x$ 
9:   end if
10: end for
```

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Algorithm 2 Perceptron - v2

Input: $X \in \mathbb{R}^{m \times 3}$, $classes \in \mathbb{R}^m$, $num_epochs \in \mathbb{Z}$

Output $w \in \mathbb{R}^3$

```
1: Initialize  $w$ 
2: for  $epoch = 1$  to  $num\_epochs$  do
3:   for  $i = 1$  to  $m$  do
4:      $x = X[i, :]$ 
5:     if  $x \cdot w > 0$  and  $classes[i] == -1$  then
6:        $w = w - x$ 
7:     end if
8:     if  $x \cdot w \leq 0$  and  $classes[i] == 1$  then
9:        $w = w + x$ 
10:    end if
11:  end for
12: end for
```

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Algorithm 3 Perceptron - Final

Input: $X \in \mathbb{R}^{m \times 3}$, $classes \in \mathbb{R}^m$, $num_epochs \in \mathbb{Z}$

Output $w \in \mathbb{R}^3$

- 1: Initialize w
 - 2: **for** $epoch = 1$ to num_epochs **do**
 - 3: **for** $i = 1$ to m **do**
 - 4: $x = X[i, :]$,
 - 5: **if** $classes[i] \times (x \cdot w) < 0$ **then**
 - 6: $w = w + classes[i] \times x$
 - 7: **end if**
 - 8: **end for**
 - 9: **end for**
-