Artificial Intelligence: Modeling Human Intelligence with Networks

Jeová Farias Sales Rocha Neto jeova_farias@brown.edu The Perceptron Algorithm!!

Algorithms

■ What is an algrithm?

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

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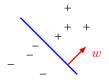


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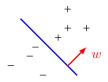


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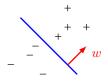
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- \blacksquare How do we find w automatically?
- The Perceptron Algorithm!

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:

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

Set up

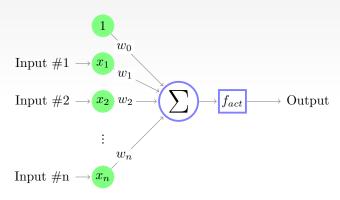
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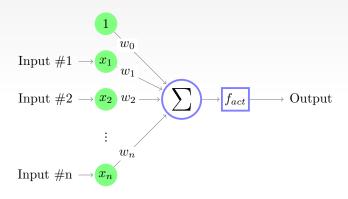
■ We then break it up in two variables: a matrix X of points and a vector *classes* of classes:

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

Remember the neuron model?



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■ We need to add a bias...

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

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().

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
    x = X[i,:]
 3:
     true\_class = classes[i],
    predicted\_class = f_{act}(x \cdot w),
 5:
     if predicted\_class == 1 and true\_class == -1 then
 6:
 7:
         w = w - x
      end if
 8:
      if predicted\_class == -1 and true\_class == 1 then
 9:
10:
         m = m + x
      end if
11:
12: end for
```

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: \mathbf{for}\ epoch = 1\ \mathbf{to}\ num\_epochs\ \mathbf{do}
       for i = 1 to m do
 3.
          x = X[i,:]
 4:
          true\_class = classes[i],
 5:
         predicted\_class = f_{act}(x \cdot w),
 6:
 7:
          if predicted\_class == 1 and true\_class == -1 then
8:
             w = w - x
          end if
 g.
          if predicted\_class == -1 and true\_class == 1 then
10:
11:
             w = w + x
          end if
12:
       end for
13:
14: end for
```

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,:],
          true\_class = classes[i],
 5:
         predicted\_class = f_{act}(x \cdot w),
6:
          if predicted\_class \times true\_class == -1 then
 7:
            w = w + true\_class \times x
 8:
          end if
9:
       end for
10:
11: end for
```