

# The Perceptron

A Brief and Semi-Technical Introduction for Non Computer Scientists

James Skripchuk

## 1 Motivation

One of the most common problems faced in Computer Science is *classification*. For example, if we give a program all of the attributes of a furry four-legged creature we found, we want the program to tell us what kind of creature this is.

Generalizing, given an object that can be represented as a finite list of numerical features  $x = [x_1, x_2, \dots, x_n]$ , we would like to categorize it as an element of a finite predetermined set of classes  $\{A, B, C \dots\}$ . A *binary classifier* is a type of classifier where there are only two classes to choose from  $\{A, B\}$ .

## 2 The Perceptron

**The perceptron** is one of the first developed and simplest examples of a binary classifier. In order to classify an object  $x = [x_1, x_2, \dots, x_n]$  as one of the two classes  $\{+, -\}$ , the perceptron algorithm is "tuned" by two parameters that are set before classification starts:

- a list of numerical weights  $w = [w_1, w_2, \dots, w_n]$
- a threshold value  $t$

The weight  $w_i$  determines how important feature  $x_i$  is in making the decision. Large positive numbers means the associated feature is important, while numbers close to zero means the feature doesn't affect the overall decision much. A negative weight means that the associated feature  $x_i$  is undesirable. The threshold value  $t$  is used to determine how confident our classifier needs to be in order to say that  $x$  is indeed a member of one class. A high threshold value  $t$  means we need more evidence in order to classify the object as positive.

First, we multiply pairwise each elements of  $x$  and  $w$  and add them together (also known as the *dot product* between  $x$  and  $w$ ).

$$z = x_1w_1 + x_2w_2 + x_3w_3 + \dots x_nw_n \tag{1}$$

Then to classify the object, we simply compare  $z$  to the threshold value  $t$

$$f(z) = \begin{cases} + & \text{if } z \geq t \\ - & \text{if } z < t \end{cases} \tag{2}$$

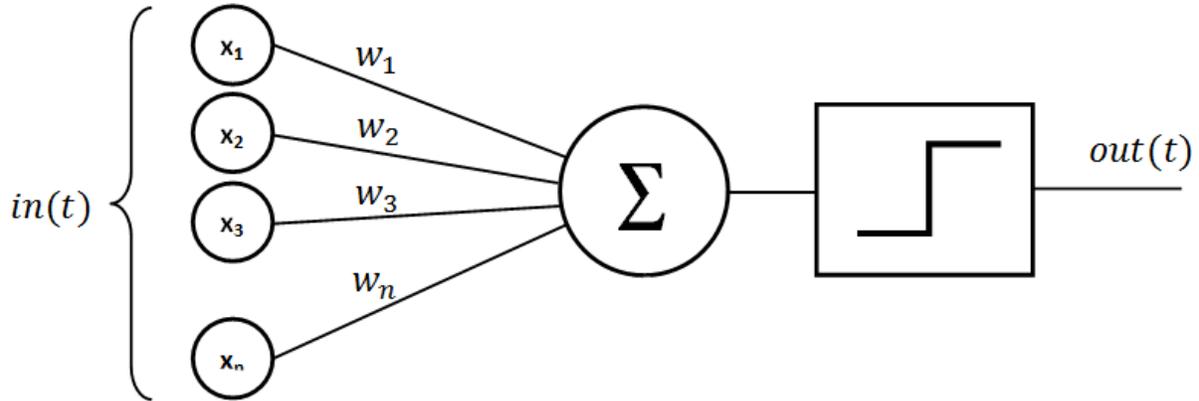


Figure 1: A Simple Perceptron [3]

### 3 Example

Suppose we have a group consisting of dogs and cats and we want to classify if something is a dog based off of a couple features: whether or not it has fur (represented by -1 or 1), its height, and its time spent napping. If our object's dot product exceeds the threshold, we classify it as a dog. We can pick our weights using the following reasoning.

$$w = [1, 4, -2] \tag{3}$$

A lot of animals have fur, so if something we're looking at has fur, it doesn't tell us much. Therefore, we can assign  $w_{fur}$  a relatively small number. Dogs tend to be much taller than cats, and thus height seems to be an important attribute, so we give  $w_{height}$  a larger number. Dogs tend to spend much less time napping than cats, so if something naps a lot, then it's probably *not* a dog. Because of this, we can assign  $w_{napping}$  to be a negative number (so the more something naps, the more it brings the total sum down). Finally, we need to set a threshold value; this can be obtained by checking the results of the dot product on data that you already know the answer to. Let's make  $t = 100$ .

Now, let's try to classify the following animals:

- Fido:  $f = [1, 50, 2]$
- Mittens:  $m = [1, 10, 10]$

Performing the dot product, we see that  $z_{Fido} = 117$  and  $z_{Mittens} = 21$ . Fido passes the  $t = 100$  threshold, and thus we classify it is as a dog. On the other hand, we can't say Mittens is a dog since it doesn't pass the threshold.

## 4 Conclusion

Of course, perceptrons aren't perfect. I am sure you can come up with examples that can fool the classifier based on the assumptions that were made in the beginning (maybe you have a very lazy dog, or a very large cat). This is only an introduction of the most primitive form of the perceptron; other resources (such as those sourced below) describe more modern research.

## References

- [1] Akshay Chandra Lagandula <https://towardsdatascience.com/perceptron-learning-algorithm-d5db0deab975>.
- [2] Manuela Veloso *Perceptrons and Neural Networks* Carnegie Mellon Fall 2001.
- [3] Perceptron by Mayranna is licensed under CC3.0  
[https://commons.wikimedia.org/wiki/File:Perceptron\\_moj.png](https://commons.wikimedia.org/wiki/File:Perceptron_moj.png)
- [4] Perceptrons - the most basic form of a neural network  
<https://appliedgo.net/perceptron/>