# ALPAYDIN'S BOOK: Ex. 11.14.2 NN

## Machine Learning 2024-25 Course Activity

Furno Francesco - francesco.furno@studenti.unipd.it - 2139507

December 18, 2024

#### Exercise 11.14.2

Show the perceptron that calculates NAND of its two inputs.

#### A NAND B

A NAND B is a boolean function that can be rewritten as  $\neg(A \land B)$ .

The following is the truth table for this boolean function:

| A | $\boldsymbol{B}$ | $A \wedge B$ | $\neg (A \land B)$ |
|---|------------------|--------------|--------------------|
| 0 | 0                | 0            | 1                  |
| 0 | 1                | 0            | 1                  |
| 1 | 0                | 0            | 1                  |
| 1 | 1                | 1            | 0                  |

Table 1:  $\neg(A \land B)$  truth table

### Perceptron implementation

We consider two inputs A and B. We know that the perceptron should return 0 only if both A and B are set to 1, 1 otherwise.

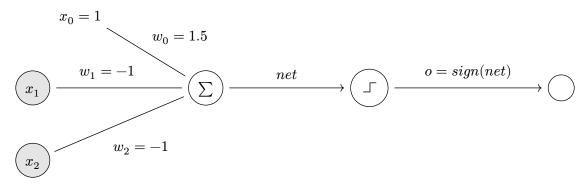
Note that it is possible to implement a perceptron for the NAND operator because it is a linearly separable function, making it perfectly suited for a perceptron

We set the input layer as follows:

- $x_0 = 1$
- $x_1 = A$
- $x_2 = B$

Now, let's assign the weights. We know that the function should output 0 when  $x_1 = 1 \land x_2 = 1$ , so when their sum is higher than 1.5. Then, the function should output 1 in all the other cases, so when  $x_1 + x_2$  is lower than 1.5. We can set the weights as follows:

- $w_0 = 1.5$
- $w_1 = -1$
- $w_2 = -1$



where  $net = \sum_{i=0}^{n} w_i x_i$  and  $o = \sigma(net) = sign(net)$ .

Now, let's consider all the possible values for inputs  $x_1$  and  $x_2$ .

$$\boldsymbol{x_1} = \boldsymbol{0}, \boldsymbol{x_2} = \boldsymbol{0}$$

The threshold expression is the following:

$$\begin{aligned} sign(net) &= sign(w_0 \cdot 1 + w_1 \cdot 0 + w_2 \cdot 0) \\ &= sign(1.5 \cdot 1 + (-1) \cdot 0 + (-1) \cdot 0) \\ &= sign(1.5) = 1 \end{aligned}$$

$$\boldsymbol{x_1} = \boldsymbol{0}, \boldsymbol{x_2} = \boldsymbol{1}$$

The threshold expression is the following:

$$\begin{split} sign(net) &= sign(w_0 \cdot 1 + w_1 \cdot 0 + w_2 \cdot 1) \\ &= sign(1.5 \cdot 1 + (-1) \cdot 0 + (-1) \cdot 1) \\ &= sign(0.5) = 1 \end{split}$$

$$x_1 = 1, x_2 = 0$$

The threshold expression is the following:

$$\begin{split} sign(net) &= sign(w_0 \cdot 1 + w_1 \cdot 1 + w_2 \cdot 0) \\ &= sign(1.5 \cdot 1 + (-1) \cdot 1 + (-1) \cdot 0) \\ &= sign(0.5) = 1 \end{split}$$

$$x_1 = 1, x_2 = 1$$

The threshold expression is the following:

$$\begin{split} sign(net) &= sign(w_0 \cdot 1 + w_1 \cdot 1 + w_2 \cdot 1) \\ &= sign(1.5 \cdot 1 + (-1) \cdot 1 + (-1) \cdot 1) \\ &= sign(-0.5) = -1 \Rightarrow 0 \end{split}$$

Hence, the perceptron correctly implements the NAND operator. This demonstrates that the NAND operator is linearly separable and can be represented by a single-layer perceptron.