

Exercise 1 : Perceptron Learning

In this exercise, you design a single perceptron with two inputs  $x_1$  and  $x_2$ . This perceptron shall implement the boolean formula  $A \wedge \neg B$  with a suitable function  $y(x_1, x_2)$ . Use the values 0 for *false* and 1 for *true*.

- (a) Draw all possible examples and a suitable decision boundary in a coordinate system.
- (b) Draw the graph of the perceptron. The schematic must include  $x_1$ ,  $x_2$ , and all model weights.
- (c) Manually determine a set of suitable weights  $\mathbf{w} = (w_0, w_1, w_2)$  from your drawings.

Exercise 2 : Perceptron Learning

Why can the boolean formula  $A \text{ XOR } B$  not be learned by a single perceptron? Justify your answer with a drawing.

Exercise 3 : Parameters of the Multilayer Perceptrons

In this exercise, you analyze the number of weights (parameters) of multilayer perceptrons. We use the notation from the lecture (e.g., slide ML:IV-104), where multilayer perceptrons have  $d$  layers,  $p$  attributes, hidden layer  $i$  with  $l_i$  units, and an output layer with  $k$  units.

- (a) Let  $d = 4$ ,  $p = 7$ ,  $l_1 = 5$ ,  $l_2 = 3$ ,  $l_3 = 3$ , and  $k = 4$ . Draw the graph of the multilayer perceptron.
- (b) Calculate the number of weights in the multilayer perceptron of (a).
- (c) Calculate the number of weights in the multilayer perceptron of (a) but with each  $l_i$  doubled, i.e.,  $l_1 = 10$ ,  $l_2 = 6$ ,  $l_3 = 6$ . Has the number of weights doubled as well?
- (d) Let  $f(p, l_1, \dots, l_{d-1}, k)$  be a function that computes the number of weights in the general case. Write down an expression for  $f$ .