

## Exercises

- Problem 1: Draw a perceptron that does A AND B; is it possible? (If possible, then specify the weights)
- Problem 2: Draw a perceptron that does A OR B; is it possible? (If possible, then specify the weights)
- Problem 3: Draw a perceptron that does A XOR B; is it possible? (If possible, then specify the weights)

- Problem 4: Write down the update rule for Stochastic Gradient Descent.

Let  $w \in R^n$  be the weight vector,  $x^{(i)}$  be the  $i$ th sample from the training data, and  $L(w, x)$  be a loss function (assume that it is continuous and differentiable). We observe example  $x^{(i)}$  and predict using the current hypothesis  $w$ . We then update the weight vector. Express the new weight vector as a function of  $w$ ,  $L(w, x^{(i)})$ , and the learning rate  $R$ .

$$w \leftarrow \dots$$

- Problem 5: Consider two perceptrons defined by the threshold expression  $w_0 + w_1x_1 + w_2x_2 > 0$ .

Perceptron A has weight values

$$w_0 = 1, w_1 = 2, w_2 = 1$$

and perceptron B has the weight values

$$w_0 = 0, w_1 = 2, w_2 = 1$$

True or false? Perceptron A is *more general than* perceptron B. (*more general than* is defined in Chapter 2.)

**Problem 6:** Design a two-input perceptron that implements the boolean function  $A \wedge \neg B$ , and a 2-layer network of perceptrons that implements  $A \text{ XOR } B$ .

**Problem 7:** Consider a two-dimensional space  $XY$ ;  $X$  and  $Y$  are inputs of a perceptron. We have seen that the decision surface of a perceptron is always a straight line. We also know that perceptrons can apply OR and AND operations to boolean values.

Now consider a neural network with two layers of perceptrons (a hidden layer with  $n$  perceptrons and an output layer of 1 perceptron). Based on the above observations, what kind of decision surface do you think such a network can at least form?