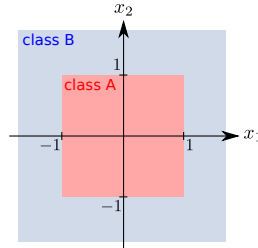


Exercise Sheet 12

Exercise 1: Designing a Neural Network (25 P)

We would like to implement a neural network that classifies data points in \mathbb{R}^2 according to the following decision boundary:



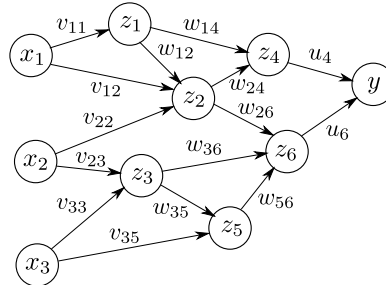
Note that the area for class B stretches to the infinity. We consider as an elementary computation the *threshold neuron* whose relation between inputs $\{z_i\}$ and output z_j is given by

$$z_j = \begin{cases} 1 & \text{if } \sum_i z_i w_{ij} + b_j > 0 \\ 0 & \text{if } \sum_i z_i w_{ij} + b_j \leq 0. \end{cases}$$

In a similar way the XOR problem was solved in the slides, design at hand a neural network that takes x_1 and x_2 as input and produces the output “1” if the input belong to class A, and “0” if the input belongs to class B. Draw the neural network model and write down its corresponding weight and bias parameters.

Exercise 2: Backward Computations (25 P)

We consider a neural network with the following structure:



The elementary computation of this network is the *sigmoid neuron* defined as:

$$g(a_j) = \frac{e^{a_j}}{1 + e^{a_j}} \quad \text{where} \quad a_j = \sum_i z_i w_{ij}.$$

Examples of forward computations are:

$$\begin{aligned} z_5 &= g(x_3 \cdot v_{35} + z_3 \cdot w_{35}) \\ z_6 &= g(z_2 \cdot w_{26} + z_3 \cdot w_{36} + z_5 \cdot w_{56}) \\ y &= g(z_4 \cdot u_4 + z_6 \cdot u_6) \end{aligned}$$

Assuming that we have computed the activation of each neuron and the output, and knowing the error gradient $\frac{\partial E}{\partial y}$, write the sequence of computations that lead to the evaluation of the error gradient $\frac{\partial E}{\partial w_{12}}$. *Hint:* The derivative of the sigmoid function $g(a_j)$ can be expressed in terms of neuron activations as

$$\frac{\partial}{\partial a_j} g(a_j) = \underbrace{g(a_j)}_{z_j} (1 - \underbrace{g(a_j)}_{z_j}).$$

Exercise 3: Programming (50 P)

Download the programming files on ISIS and follow the instructions.