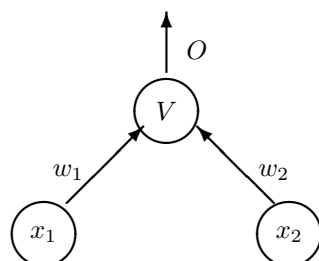


## Homework-2

### Question 1



	$x_1$	$x_2$	$y$
$e_1$	1	0	1
$e_2$	1	1	1
$e_3$	2	0	0
$e_4$	2	1	0

You are given a perceptron implemented with a sigmoid, with  $\beta = 1$ . There are NO bias connections. The initial values of the weights are  $w_1 = 0$ ,  $w_2 = 1$ .

#### Part 1

Give explicit expressions to the way the weights change if the network is given the example  $e_3$ . Use  $\epsilon = 0.1$ . You may use temporary variables in your answer, but make sure that they are all specified in terms of the given values. You may use the notation  $S(\cdot)$  instead of explicitly computing sigmoid values.

## Part 2

- a. If you could choose  $\epsilon$  to be as small as you like, and run back propagation as many epochs as you like with the four examples  $e_1, e_2, e_3, e_4$ , do you expect the computed output of the perceptron to be within 0.001 of the desired output (the value of  $y$ ) for all four examples?

**Answer:** YES / NO / IMPOSSIBLE-TO-TELL

- b. Will your answer to a. remain the same if bias connection is added (everything else stays as in a.)?

**Answer:**

- YES. My answer is exactly the same as in Part a.
- NO. My answer changes. With this new condition
  - my new answer to a. is YES.
  - my new answer to a. is NO.
  - my new answer to a. is IMPOSSIBLE-TO-TELL

## Question 2

### Part I

We would like to use the perceptron with a sigmoid unit to determine weights  $w_1, w_2, w_3$  which are to be used to compute  $O$ , an approximation of the desired response  $y$ , according to:

$$\begin{aligned} h &= w_1\phi_1 + w_2\phi_2 + w_3\phi_3 \\ O &= \frac{1}{1 + \exp\{-2\beta h\}} \end{aligned}$$

with

$$\phi_1 = x_1, \quad \phi_2 = x_2, \quad \phi_3 = 1.$$

The 4 training examples are given by:

	$x_1$	$x_2$	$y$
$e_1$	0	0	0
$e_2$	0	1	1
$e_3$	1	0	1
$e_4$	1	1	0

- a. With  $\beta = 1, \epsilon = 0.5$ , show that when the algorithm is given  $e_1$  the values of  $w_1, w_2, w_3$  change according to:

$$\begin{aligned} O &\leftarrow \frac{1}{1 + \exp\{-2w_3\}} \\ \delta &\leftarrow O^2(O - 1) \\ w_1 &\leftarrow w_1 \\ w_2 &\leftarrow w_2 \\ w_3 &\leftarrow w_3 + \frac{\delta}{2} \end{aligned}$$

- b. Give explicit formulas to how  $w_1, w_2, w_3$  change when the algorithm is given  $e_2$ .
- c. Give explicit formulas to how  $w_1, w_2, w_3$  change when the algorithm is given  $e_3$ .
- d. Give explicit formulas to how  $w_1, w_2, w_3$  change when the algorithm is given  $e_4$ .

## Part II

Write a program that implements these equations. Start with  $w_1 = w_2 = w_3 = 0$  and compute their value after (a) 1 training epoch, (b) 2 epochs, (c) 100 epochs, (d) 1000 epochs, (e) 2000 epochs. (An epoch is a pass through the entire training set which in our case is  $e_1, e_2, e_3, e_4$ .)

## Part III

The results obtained in Part II are used to classify examples by labeling a pair  $(x_1, x_2)$  of real numbers as a positive example when  $O > 0.5$ , and as a negative example when  $O \leq 0.5$ .

- a. for the values of  $w_1, w_2, w_3$  that were obtained in (a) show in a diagram of the  $x_1, x_2$  plane what area is labeled as positive by shadowing that area.
- b. Same as (a) for the values of  $w_1, w_2, w_3$  that were obtained in (b).
- c. Same as (a) for the values of  $w_1, w_2, w_3$  that were obtained in (c).
- d. Same as (a) for the values of  $w_1, w_2, w_3$  that were obtained in (d).
- e. Same as (a) for the values of  $w_1, w_2, w_3$  that were obtained in (e).