

ECE/CS 559 - Fall 2016 - Midterm Part #1.

Full Name:

ID Number:

- **Q1 (24 pts):** In the following, we follow the convention that $x_0 = 1$.

Recall that the step-activation function $u : \mathbb{R} \rightarrow \{0, 1\}$ is defined as $u(v) = 1$ if $v \geq 0$, and $u(v) = 0$, otherwise. Then, for n inputs x_1, \dots, x_n , a **perceptron** can be defined via the input-output relationship

$$y' = u\left(\sum_{i=0}^n w_i x_i\right) = u(w_0 + w_1 x_1 + \dots + w_n x_n),$$

where y' is the perceptron output, w_1, \dots, w_n are the synaptic weights, and w_0 is the bias term.

We define a new type of neuron, namely, a **sauron**, via the input-output relationship

$$y = u\left(\prod_{i=0}^n (w_i + x_i)\right) = u((w_0 + 1)(w_1 + x_1) \cdots (w_n + x_n)),$$

where y is called the sauron output.

Let the real number 1 represent a **TRUE**, and the real number 0 represent a **FALSE**.

- (a) **(8 pts):** Let $n = 1$. Does there exist w_0, w_1 such that $y = 1 - x_1$ for $x_1 \in \{0, 1\}$? In other words, can a single sauron implement the **NOT** gate? If your answer is “Yes,” find specific w_0, w_1 such that the sauron implements the **NOT** gate. If your answer is “No,” prove that no choice for w_0, w_1 can result in a sauron that implements the **NOT** gate.
 - (b) **(8 pts):** Let $n = 2$. Does there exist w_0, w_1, w_2 such that $y = x_1 x_2$ for $x_1, x_2 \in \{0, 1\}$? In other words, can a single sauron implement the **AND** gate? Justify your answer as in (a).
 - (c) **(8 pts):** Let $n = 2$. Does there exist w_0, w_1, w_2 such that $y = (x_1 + x_2) \bmod 2$ for $x_1, x_2 \in \{0, 1\}$? In other words, can a single sauron implement the **XOR** gate? Justify your answer as in (a).
- **Q2 (26 pts):** In the following, consider only neurons with the step-activation function $u(\cdot)$.
 - (a) **(13 pts):** Let $\mathcal{C}_0 = \left\{\begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 4 \end{bmatrix}\right\}$ and $\mathcal{C}_1 = \left\{\begin{bmatrix} 0 \\ 2 \end{bmatrix}, \begin{bmatrix} 0 \\ -1 \end{bmatrix}, \begin{bmatrix} 0 \\ -2 \end{bmatrix}\right\}$ as illustrated in Figure (i). Members of classes \mathcal{C}_0 and \mathcal{C}_1 are represented by black disks and crosses, respectively.
 - [I] **(7 pts):** We wish to design a perceptron $y = u(w_0 + w_1 x_1 + w_2 x_2)$ such that $y = 0$ if $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \in \mathcal{C}_0$ and $y = 1$ if $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \in \mathcal{C}_1$. Suppose that we use the perceptron training algorithm for this purpose with initial weights $w_0 = 1, w_1 = 0, w_2 = 1$ and learning rate $\eta = 1$. Either prove that the algorithm will converge, or prove that it will not converge. You may use the perceptron convergence theorem.
 - [II] **(6 pts):** Design a neural network (single-layer or multi-layer) such that the network provides an output of 0 if $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \in \mathcal{C}_0$ and an output of 1 if $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \in \mathcal{C}_1$.
 - (b) **(13 pts)** Repeat (a) for classes $\mathcal{C}_0 = \left\{\begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 4 \end{bmatrix}\right\}$ and $\mathcal{C}_1 = \left\{\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ -1 \end{bmatrix}, \begin{bmatrix} 0 \\ -2 \end{bmatrix}\right\}$ illustrated in Figure (ii). Note that the only difference is that now, instead of the point $\begin{bmatrix} 0 \\ 2 \end{bmatrix}$, we have the point $\begin{bmatrix} 0 \\ 0 \end{bmatrix}$ in class \mathcal{C}_1 .

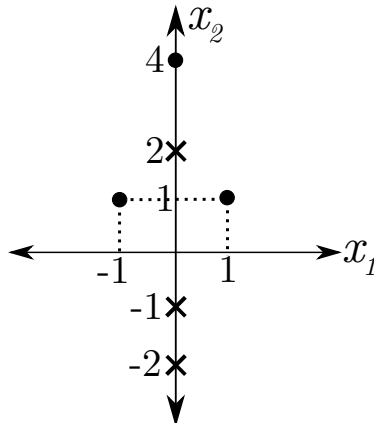


Figure (i)

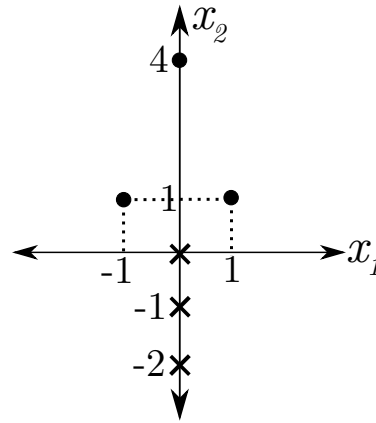


Figure (ii)