

**CSE - 5526**

**Homework 1**

*Submitted by*

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## 1 Question 1

Give the weights and bias for a McCulloch-Pitts (M-P) neuron that has the following:

- Inputs:  $x$ ,  $y$  and  $z$
- Output: is equal to  $z$  if  $x = -1$  and  $y = 1$ , and is equal to -1 otherwise

**Solution**

Table 1: Truth Table

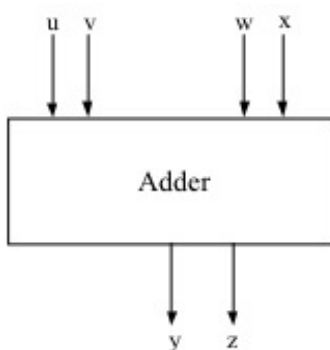
$x$	$y$	$z$	Output
-1	-1	-1	-1
-1	-1	1	-1
-1	1	-1	-1
-1	1	1	1
1	-1	-1	-1
1	-1	1	-1
1	1	-1	-1
1	1	1	-1

Approach: For the given set of three inputs;  $x$ ,  $y$ , and  $z$ , I created a truth table which satisfied the given output constraints. Then by plugging in various values of three weights and biases, I found the combination that satisfied each input-output combination in the truth table. For inputs  $x$ ,  $y$ , and  $z$ , final weights taken were -1, 1, 1, and the bias was -2.

## 2 Question 2

For this problem, change the definition of an M-P neuron so that both its inputs and outputs are binary (0 or 1). View  $uv$  and  $wx$  each as two-bit binary (00, 01, 10, or 11) numbers, and  $yz$  as the 2 low-order bits of the numerical addition of  $uv$  and  $wx$  (see the figure below).

- Give weights and biases for an M-P network which generates  $z$ .
- Give weights and biases for an M-P network which generates  $y$ .



**Solution**

Table 2: Truth Table

$u$	$v$	$w$	$x$	$y$	$z$
0	0	0	0	0	0
0	0	0	1	0	1
0	0	1	0	1	0
0	0	1	1	1	1
0	1	0	0	0	1
0	1	0	1	1	0
0	1	1	0	1	1
0	1	1	1	0	0
1	0	0	0	1	0
1	0	0	1	1	1
1	0	1	0	0	0
1	0	1	1	0	1
1	1	0	0	1	1
1	1	0	1	0	0
1	1	1	0	0	1
1	1	1	1	1	0

z	w,x			
	00	01	11	10
u,v 00	0	1	1	0
01	1	0	0	1
11	1	0	0	1
10	0	1	1	0

Figure 1: Karnaugh Map for part (a)

From above K-Map, it can be inferred that the relationship between inputs and outputs for an MP network that generates  $z$  is given by:

$$z = v'x + vx' \quad (1)$$

Following neural network can be drawn that satisfies the above equation, with weights and biases mentioned:

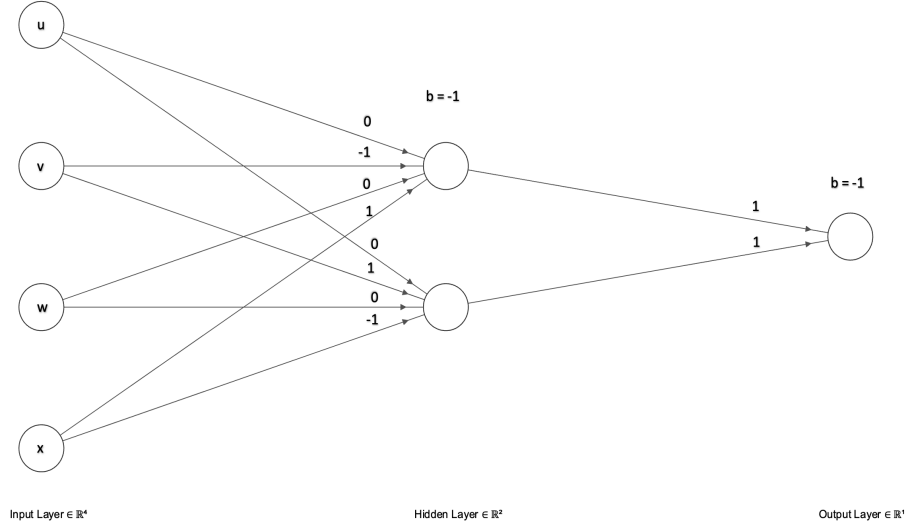


Figure 2: M-P Network for part (a)

y	w,x			
	00	01	11	10
u,v 00	0	0	1	1
01	0	1	0	1
11	1	0	1	0
10	1	1	0	0

Figure 3: Karnaugh Map for part (b)

From above K-Map, it can be inferred that the relationship between inputs and outputs for an MP network that generates  $y$  is given by:

$$y = uw'x' + uv'w' + u'v'w + u'wx' + u'vw'x + uvwx \quad (2)$$

Following neural network can be drawn that satisfies the above equation, with weights and biases mentioned:

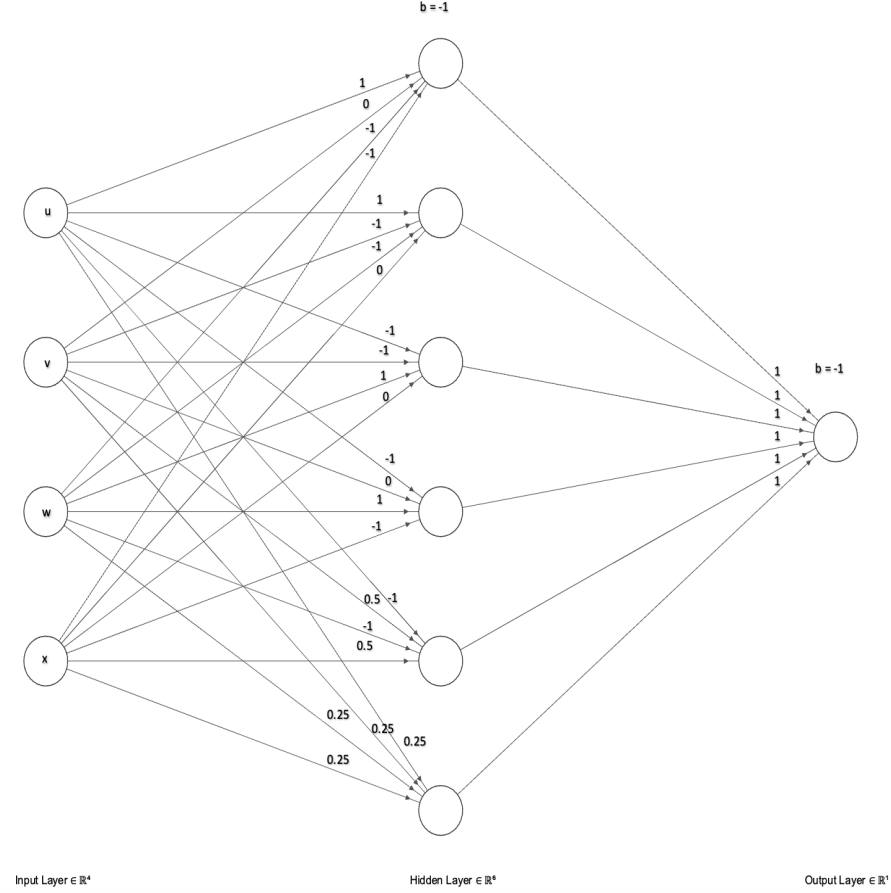


Figure 4: M-P Network for part (b)

Approach: For the given set of binary inputs;  $uv$  and  $wx$ , I created a truth table upon changing the definition of an M-P neuron so that both the inputs and outputs are binary (0 or 1). For both parts (a) and (b), I generated a Karnaugh Map to draw out a relationship between input bits and outputs ( $z$  for part (a) and  $y$  for part (b)), as shown in Figure 1 & 3. Once the input-output relationship was known, I generated M-P networks with weights and biases mentioned, as shown in Figure 2 & 4.

### 3 Question 3

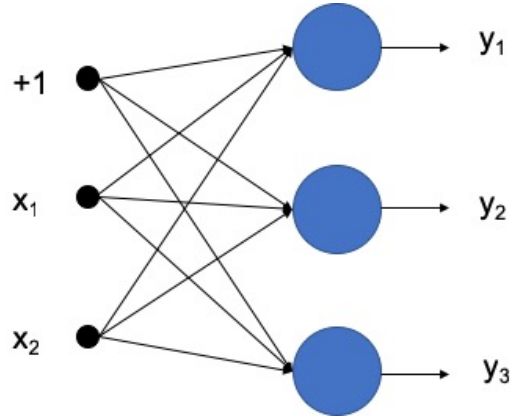
Given the following 3-class classification problem:

$C_1: \{(4,1), (2,3), (3,5), (5,4), (1,6), (-1,6)\}$

$C_2: \{(0,2), (-2,2), (-3,2), (-2,4)\}$

$C_3: \{(1,-2), (3,-2)\}$

and the following single layer perceptron:



(a) Can the net learn to separate the samples, given that you want: if  $\mathbf{x} \in C_i$  then  $y_i = 1$  and  $y_j = -1$  for  $j \neq i$ . No need to solve for the weights, but justify your answer.

(b) Remove data point  $(-1,6)$  from  $C_1$ . Repeat part (a) with this data set.

**Solution**

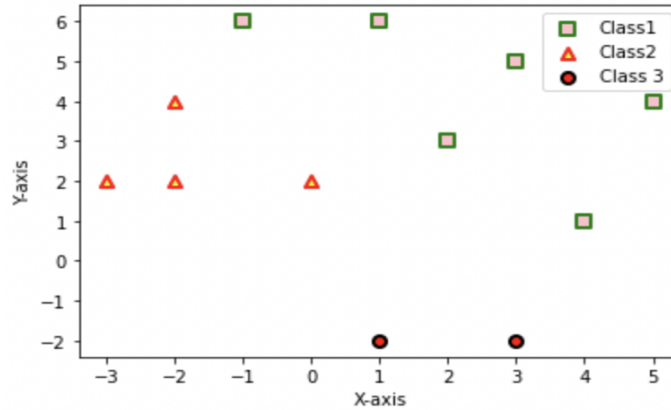


Figure 5: Scatter plot as per (a)

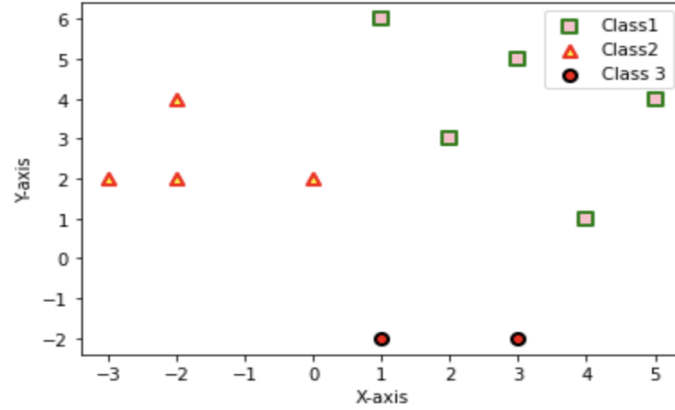


Figure 6: Scatter plot as per (b)

Approach: I plotted a graph representing all the different points from given three classes, as shown in Figure 5 & 6. For part (a), it can be seen that points in the classes  $C_1$  and  $C_3$  are linearly separable, hence a single layer perceptron can separate the samples in  $C_1$  and  $C_3$ .

For part (b), as per Figure 6, after removing point  $(-1,6)$  from  $C_1$ , even the samples in  $C_2$  become linearly separable. In this case, a single layer perceptron can separate the samples from all given classes, one at a time.