

①

### 1. logical OR

$$Z_{in} = (0 \ 1 \ 1 \ 0 \ 1) \text{ for example}$$

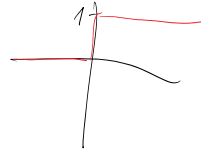
$$\beta = (1 \ 1 \ 1 \ 1 \ 1) = \{1\}^0$$

$$b = -0.5 \text{ (to avoid exactly 0)}$$

activation function  $f = \text{step function } \Theta$

$$\text{eg } Z' = Z_{in} \cdot \beta + b = (0 + 1 + 1 + 0 + 1) - 0.5 = 2.5$$

$$f(Z') = 1$$



### 2. masked logical OR

$$Z_{in} = (1 \ 0 \ 1)$$

$$\beta = 2$$

$$c = (1 \ 0 \ 1)$$

$$b = -0.5$$

activation function  $f = \Theta$

$$Z' = Z_{in} \cdot \beta + b = (1 + 0 + 1) - 0.5 = 1.5$$

$$\Theta(1.5) = 1$$

### 3. perfect match

$$Z_{in} = (1 \ 1 \ 0) \text{ for example}$$

$$c = (1 \ 1 \ 0) \text{ for example}$$

activation function  $f = \text{ReLU}$

$$Z' = Z_{in} \cdot \beta + b = (1 + 1 + 0) - 1 = 1 \quad f(1) = 1$$

$$\beta_i = \begin{cases} -1 & \text{if } c_i \leq 0 \\ +1 & \text{if } c_i = 1 \end{cases} = (1 \ 1 \ -1)$$

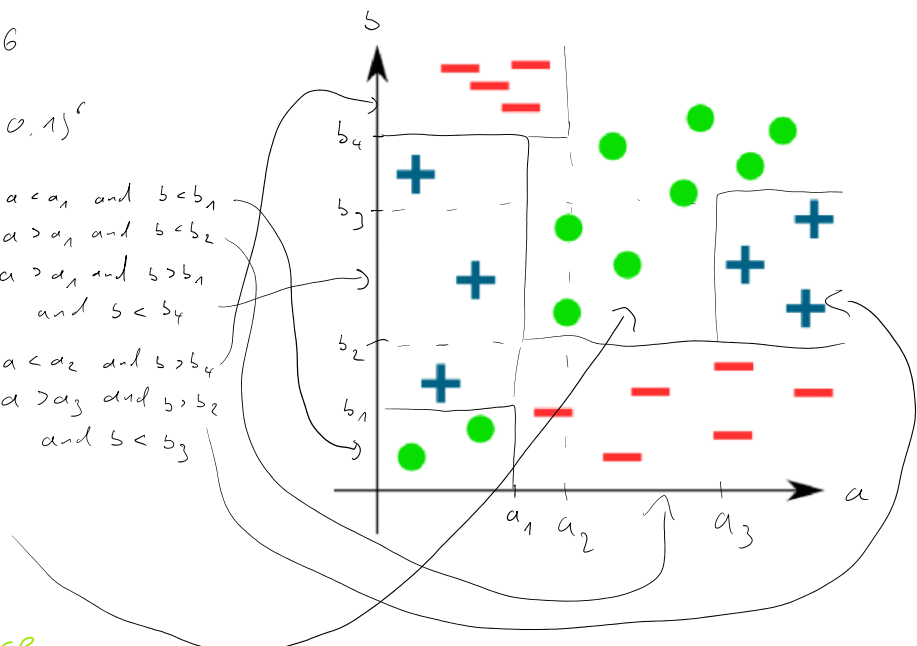
$$b = -\sum_{i=1}^n c_i + 1$$

I think this solution should work as well.

$$X = (a, b) \quad M = 6$$

first layer  $f(x) \rightarrow \{0, 1\}^6$

$$f(a, b) = \begin{cases} (100000) & \text{if } a < a_1 \text{ and } b < b_1 \\ (010000) & \text{if } a > a_1 \text{ and } b < b_2 \\ (001000) & \text{if } a > a_1 \text{ and } b > b_1 \text{ and } b < b_4 \\ (000100) & \text{if } a < a_2 \text{ and } b > b_4 \\ (000010) & \text{if } a > a_3 \text{ and } b > b_2 \text{ and } b < b_3 \\ (000001) & \text{else} \end{cases}$$



$M=3$  is enough since

a hypercube with  $M=3$  has 8 corners. Theoretically  $f(x) \rightarrow \{0, 1\}^8$ , however appending 00 to each vector is redundant

Second layer not necessary.  
third layer has 3 Neurons

Given the input vectors above this should work

1.  $z_{out1} = \text{masked OR } (z_{in}, 010100)$  red minus
2.  $z_{out2} = \text{masked OR } (z_{in}, 001010)$  blue plus
3.  $z_{out3} = \text{masked OR } (z_{in}, 100001)$  green circle

other option  $M=7$  7 neurons : one for  $a > a_1$ , one for  $a > a_2 \dots$

$$f_1(a, b) = \Theta(a - a_1) \quad \text{so } \beta = (1, 0) : z_{in} = x = (1, b) \quad z' = z_{in} \beta + b = a - a_1$$

$$b = -a_1 \quad f = \Theta$$

$$f_2(a, b) = \Theta(a - a_2)$$

$$f_3(a, b) = \Theta(a - a_3)$$

$$f_4(a, b) = \Theta(b - b_1)$$

$$f_5(a, b) = \Theta(b - b_2)$$

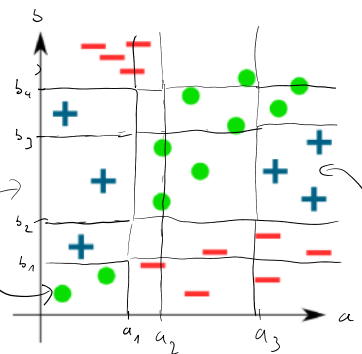
$$f_6(a, b) = \Theta(b - b_3)$$

$$f_7(a, b) = \Theta(b - b_4)$$

examples:  
this would be  
(0000000)

this would be  
(1111100)

this would be  
(0001100)



(perfect match)

Second layer : as many Neurons as regions (4.5 = 20 here)

each neuron represents one Region

1.  $z_{out} = \text{perfect match } (z_{in}, 10000000)$
2. ...
3. ...

third layer : 3 masked OR Neurons:

$$z_{out1} = \text{masked OR } (z_{in}, \{0,1\}^{20}) \quad \text{with } 1 \text{ where region contains red minus}$$

$$z_{out2} = \text{masked OR } (z_{in}, \{0,1\}^{20}) \quad \text{with } 1 \text{ where region contains blue plus}$$

$$z_{out3} = \text{masked OR } (z_{in}, \{0,1\}^{20}) \quad \text{with } 1 \text{ where region contains green circle}$$

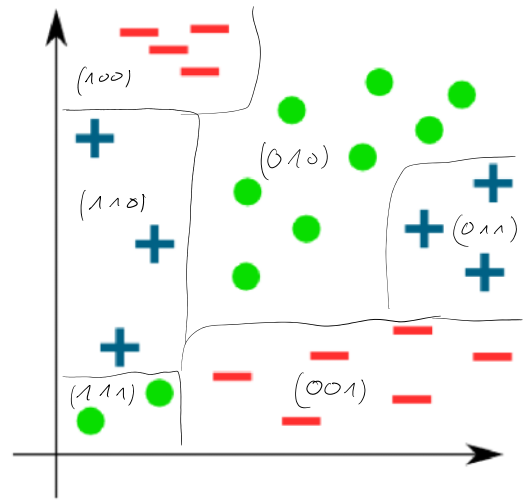
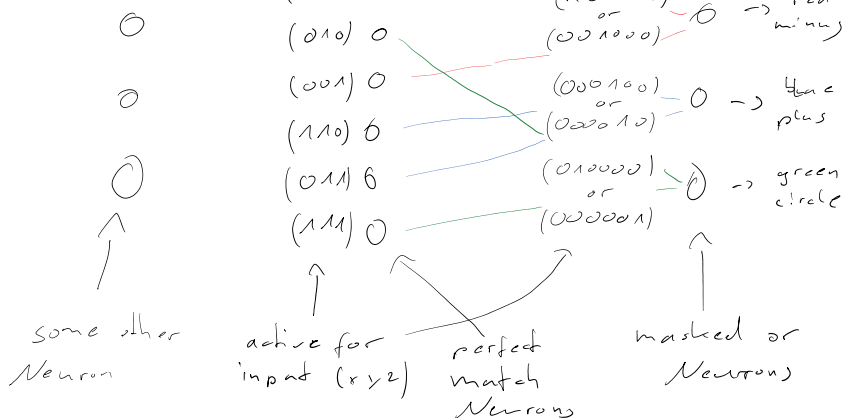
third option

$$M = 3 = \text{ceil}(\log_2(\# \text{ Regions}))$$

first layer  
M Neurons

second layer  
as many Neurons as  
regions

third layer  
3 Neurons



For more dimensions you need more decision Regions and  $M$  gets larger. You always need  $M$  Neurons in the first layer mapping Features to distinct  $M$ -dimensional vectors. The second layer has as many Neurons as decision Regions and each  $M$ -dim vector gets mapped to a specific Neuron representing that region. The last layer has as many Neurons as there are classes. Using the masked OR all Neurons corresponding to a class are mapped to one Neuron representing that class.

This gets difficult because there will be a lot of decision Regions and a lot of necessary Neurons. Also overfitting will be a problem.

② a network with  $L$  layers

$$Z_0 = X$$

$$\tilde{Z}_l = Z_{l-1} \cdot B_l + b_l$$

$$Z_l = \phi_l(\tilde{Z}_l)$$

$$Z_L = Z_{L-1} \cdot B_L + b_L$$

$$= \underbrace{(Z_{L-2} \cdot B_{L-1} + b_{L-1})}_{\dots} \cdot B_L + b_L$$

$$= Z_{L-2} \cdot \underbrace{B_{L-1} \cdot B_L}_{\text{new } B} + \underbrace{b_{L-1} \cdot B_L + b_L}_{\text{new } b}$$

$$= Z_{L-2} B + b$$

... repeat for all  $L$  layers

$$= Z_0 B' + b'$$

same as one layer

