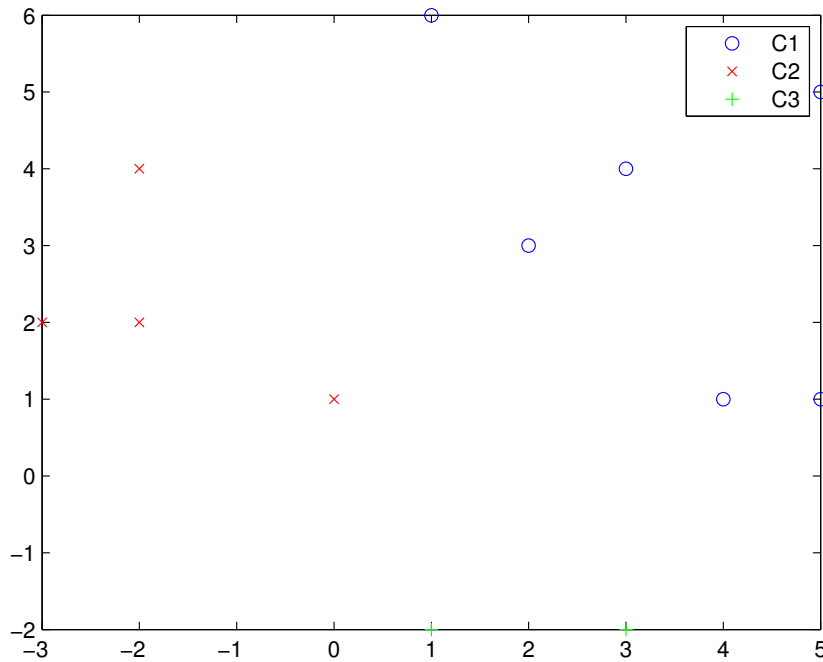


Problem 1

Can you argue (without explicitly solving for the weights) whether or not your perceptron can learn a hyperplane to separate the examples?

Yes, it is possible. Just look at the plot.



Now, what happens if you add (0, 7) to C2? Can the perceptron still learn a hyperplane to separate the examples? What happens if you add (7, 7) to C2?

Yes. No. Just add them to the plot.

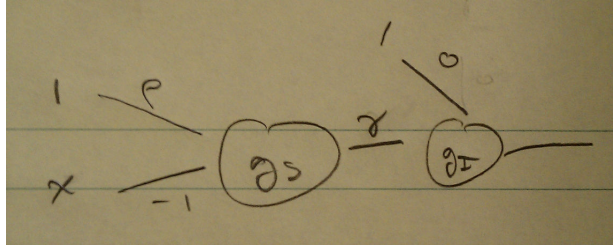
What kind of modifications/extensions to your implementation will you consider to address such situations?

Add slack variables or use nonlinear transformation.

Problem 2

(a) For user-provided/given integers γ and ρ (i.e., these values are fixed beforehand), can you use the activation functions above to construct a neural network to give the following function $out(\cdot)$: $out(x) = \gamma$ if $x < \rho$ and 0 otherwise.

I.e. we want $out(x) = \gamma g_S(\rho - x) = g_I(0 \cdot 1 + \gamma \cdot g_S(\rho \cdot 1 + (-1) \cdot x))$. Just read off the network from the preceding expression and get:



(b) Assume now that there are two input units: x and y (each is a 1 bit binary value), construct a neural network that performs the operation of exclusive-OR. In other words, the neural network should output 1 if $x \neq y$ and 0 otherwise.

This map works:

$$f(x, y) = g_S(-1 + g_S(-1 - x + y) + g_S(-1 + x - y))$$

Check:

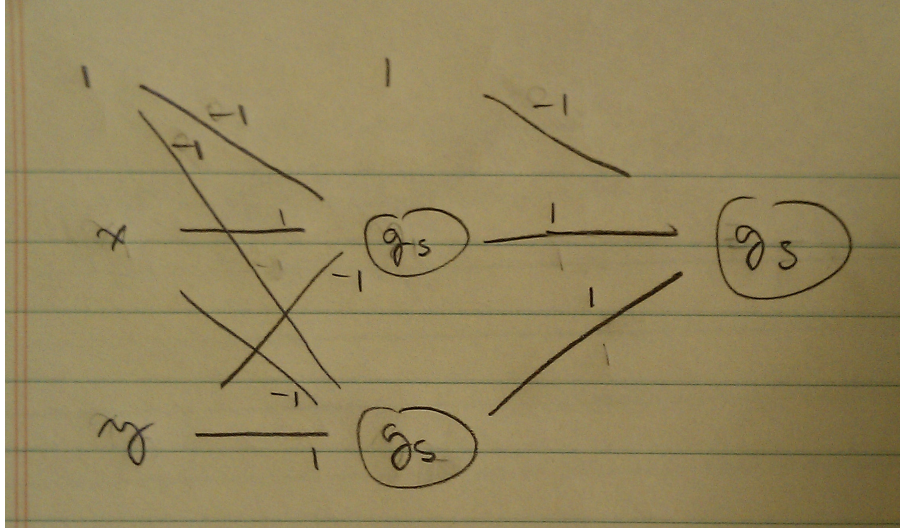
$$\begin{aligned} f(0, 0) &= g_S(-1 + g_S(-1 - 0 + 0) + g_S(-1 + 0 - 0)) \\ &= g_S(-1 + 0 + 0) \\ &= 0 \end{aligned}$$

$$\begin{aligned} f(1, 0) &= g_S(-1 + g_S(-1 - 1 + 0) + g_S(-1 + 1 - 0)) \\ &= g_S(-1 + 0 + 1) \\ &= 1 \end{aligned}$$

$$\begin{aligned} f(0, 1) &= g_S(-1 + g_S(-1 - 0 + 1) + g_S(-1 + 0 - 1)) \\ &= g_S(-1 + 1 + 0) \\ &= 1 \end{aligned}$$

$$\begin{aligned} f(1, 1) &= g_S(-1 + g_S(-1 - 1 + 1) + g_S(-1 + 1 - 1)) \\ &= g_S(-1 + 0 + 0) \\ &= 0 \end{aligned}$$

Get structure:

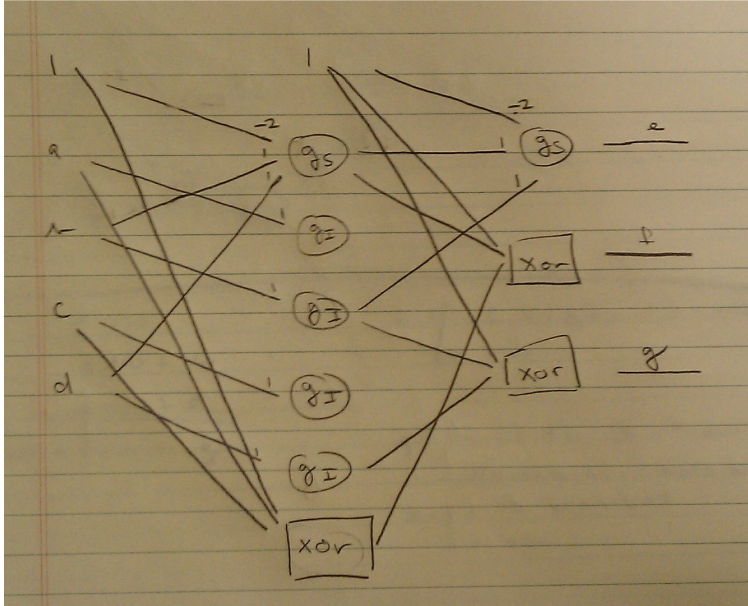


(c) Now, we have two binary numbers: ab and cd . Also, efg is the binary number (of 3 bits) which is the addition of ab and cd . Here a, c, e are the higher-order bits and b, d, g are the lower-order bit. That is, if ab is 01 and cd is 11 , then efg should be 100 . Similarly, if ab is 01 , and cd is 10 , then efg should be 011 . Can you design a neural network to implement this add operation?

Want:

$$\begin{aligned}
 g(a, b, c, d) &= \text{mod}(b + d, 2) \\
 &= \text{xor}(b, d) \\
 x(a, b, c, d) &= \begin{cases} 1 & \text{if } b + d = 2 \\ 0 & \text{o.w.} \end{cases} \\
 &= g_s(-2 + b + d) \\
 f(a, b, c, d) &= \text{mod}(x + a + c, 2) \\
 &= \text{xor}(x, \text{xor}(a, c)) \\
 e(a, b, c, d) &= \begin{cases} 1 & \text{if } x + a + c \geq 2 \\ 0 & \text{o.w.} \end{cases} \\
 &= g_s(-2 + x + a + c)
 \end{aligned}$$

Read off the following network:



Here XOR denotes a module of the form in part (b).