

# Machine Learning

## INF2008

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Lecture 02: The Single Layer Perceptron

Donny Soh

Singapore Institute  
of Technology

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# SIT Open House 2023 – The Winners (Judges' Choice)



## Most Attractive Booth (Judges' Choice)

### Champion: Naval Architecture and Marine Engineering



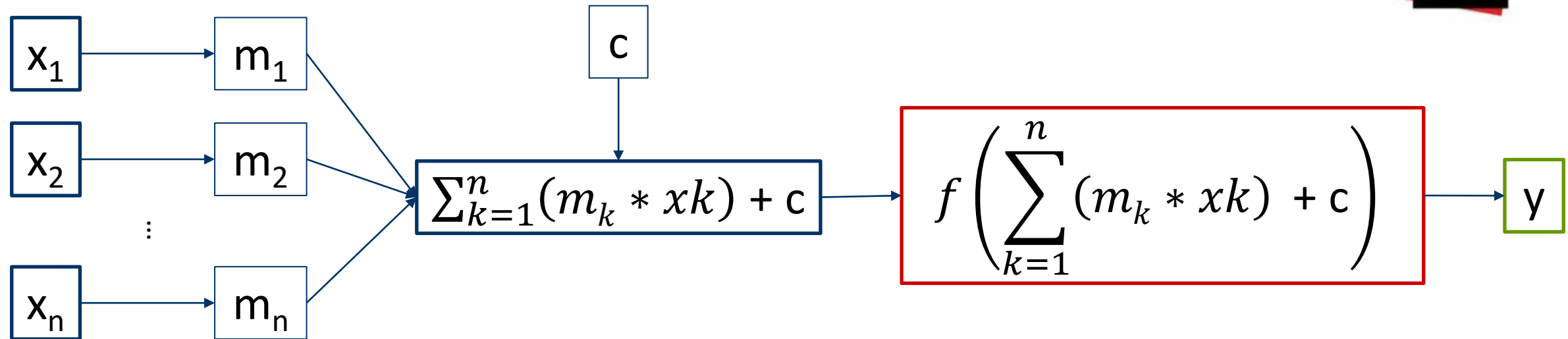
### 1<sup>st</sup> Runner-up: Applied Artificial Intelligence



### 2<sup>nd</sup> Runner-up: Physiotherapy



# Linear Models 3: Neural Network (Single Layer Perceptron)

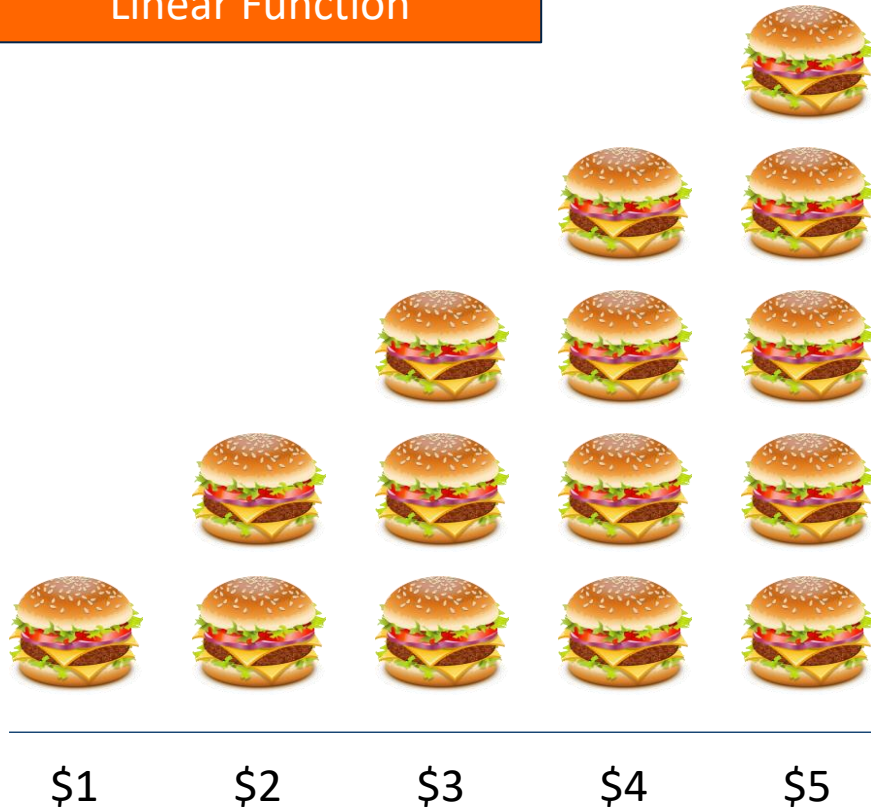


$$y = f\left(\sum_{k=1}^n m_k * x_k + c\right)$$

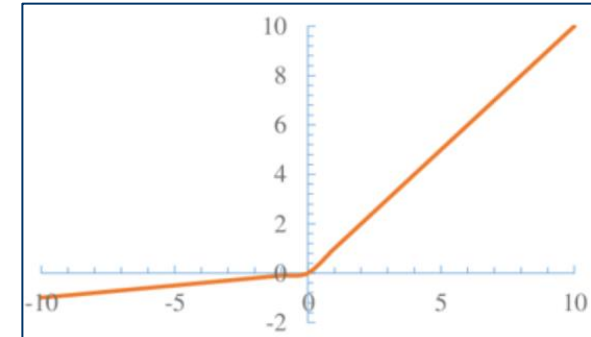
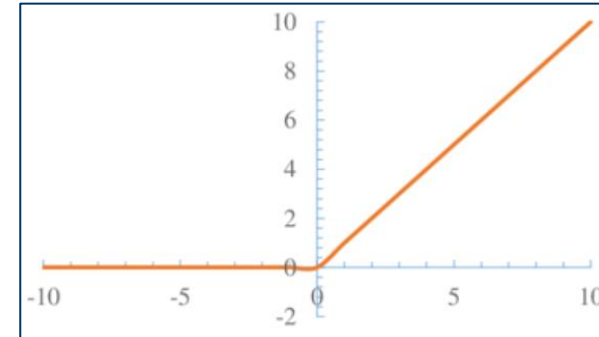
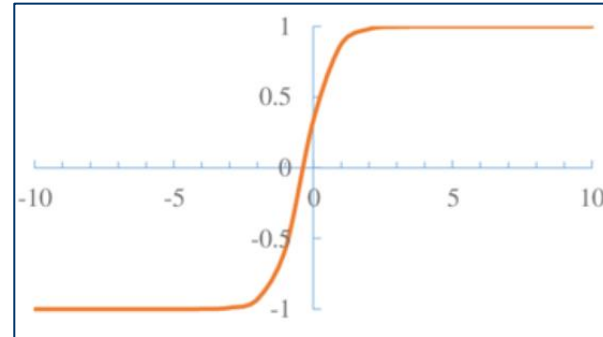
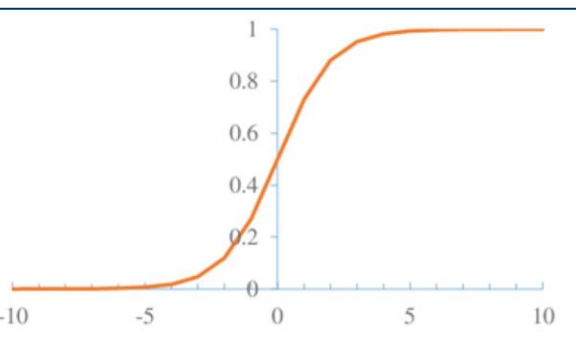
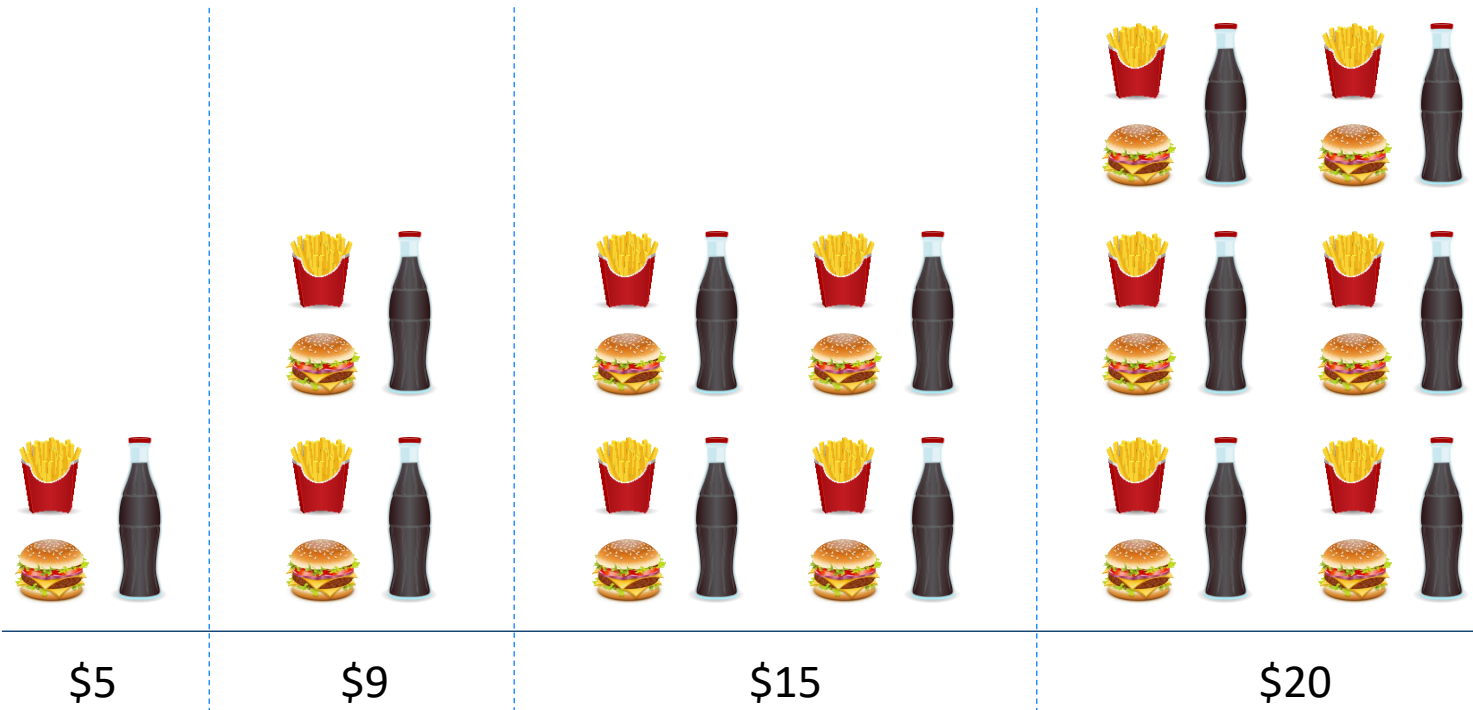
We “learn” the variables  $m_i$  and  $c$

# What is a non-linear function?

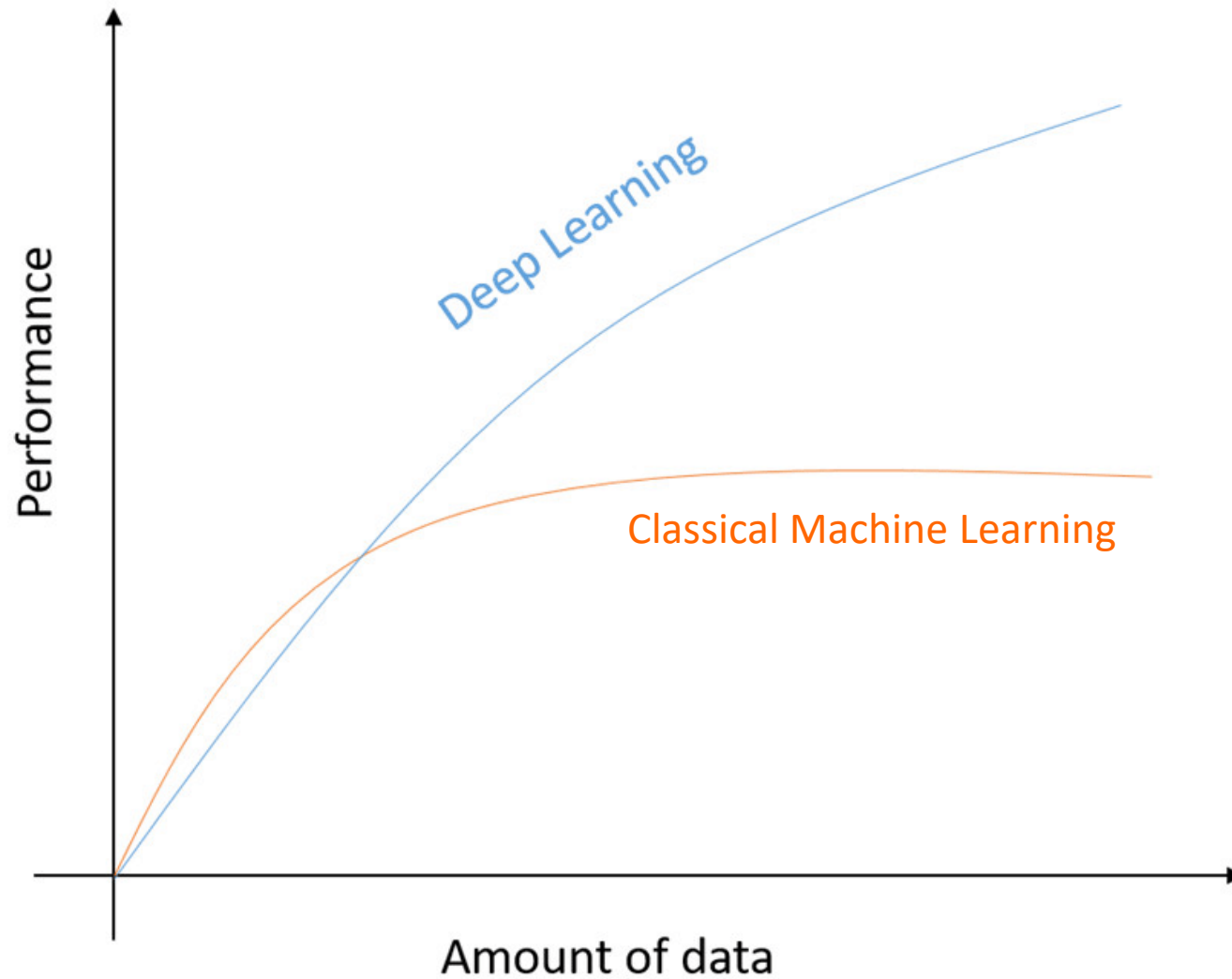
Linear Function



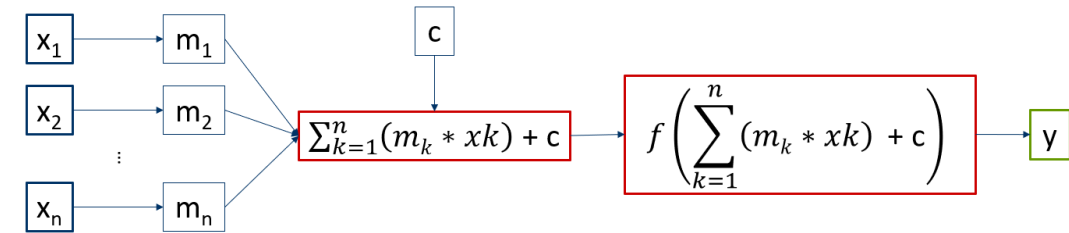
Non-Linear Function



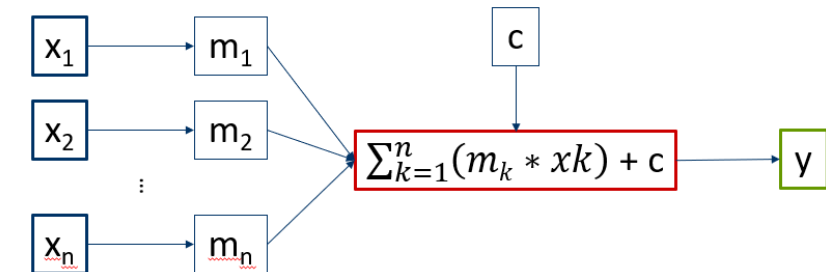
# Classical Machine Learning



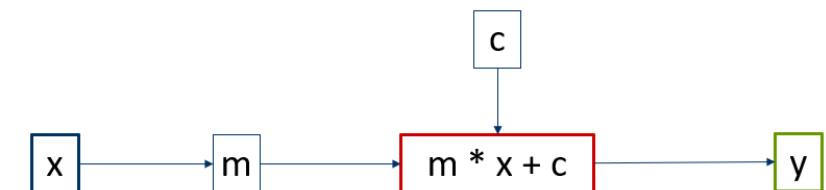
## Neural Networks (Single Layer Perceptron)



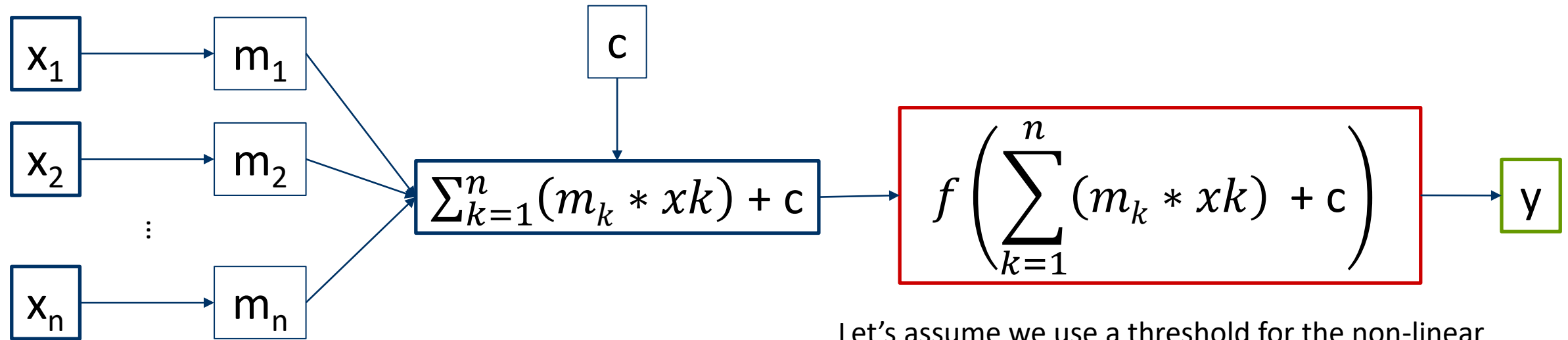
## Multiple Linear Regression



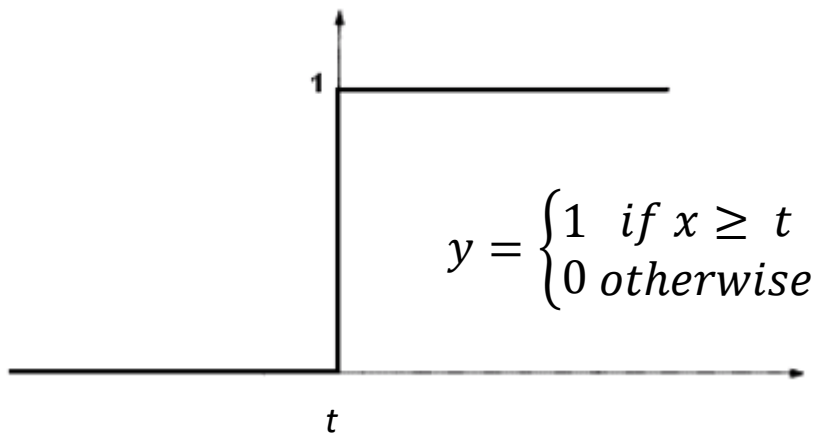
## Linear Regression



# Single Layer Perceptron



Let's assume we use a threshold for the non-linear function.

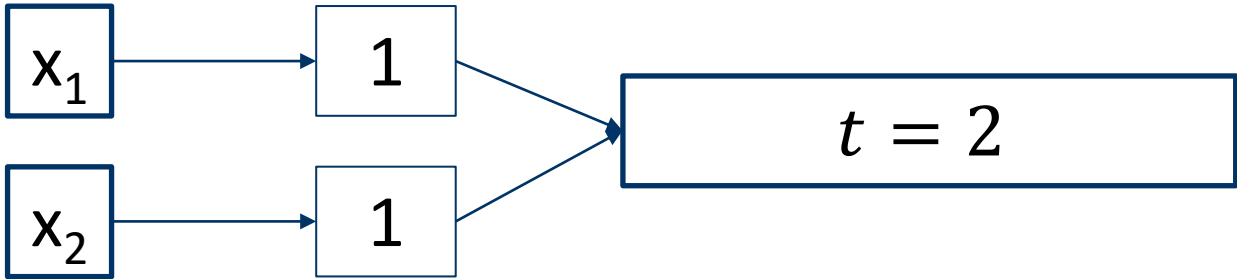


Essentially, this means that for some threshold value of  $t$ ,  $f(x)$  is 0 for  $x < t$  and  $f(x)$  is 1 for  $x \geq t$



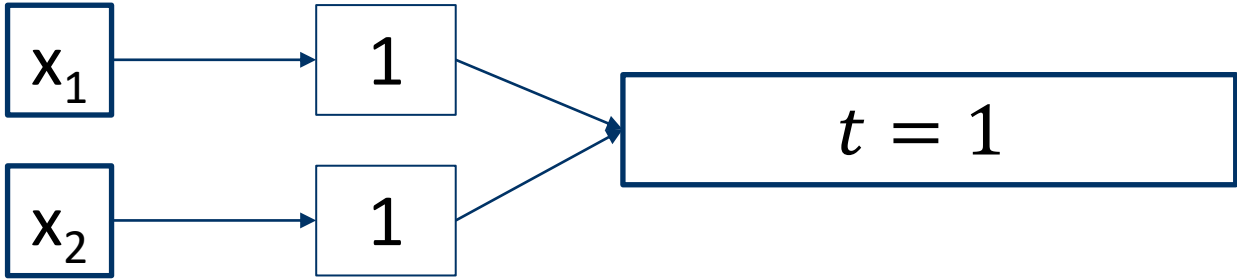
Let's draw the perceptron to mimic Boolean gates

AND Gate



x1	x2	output	value
0	0	0	False
1	0	1	False
0	1	1	False
1	1	2	True

OR Gate



x1	x2	output	value
0	0	0	False
1	0	1	True
0	1	1	True
1	1	2	True

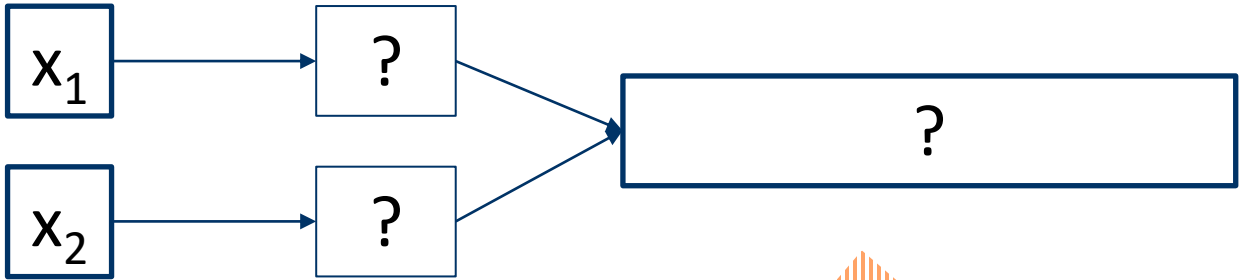
NOT Gate



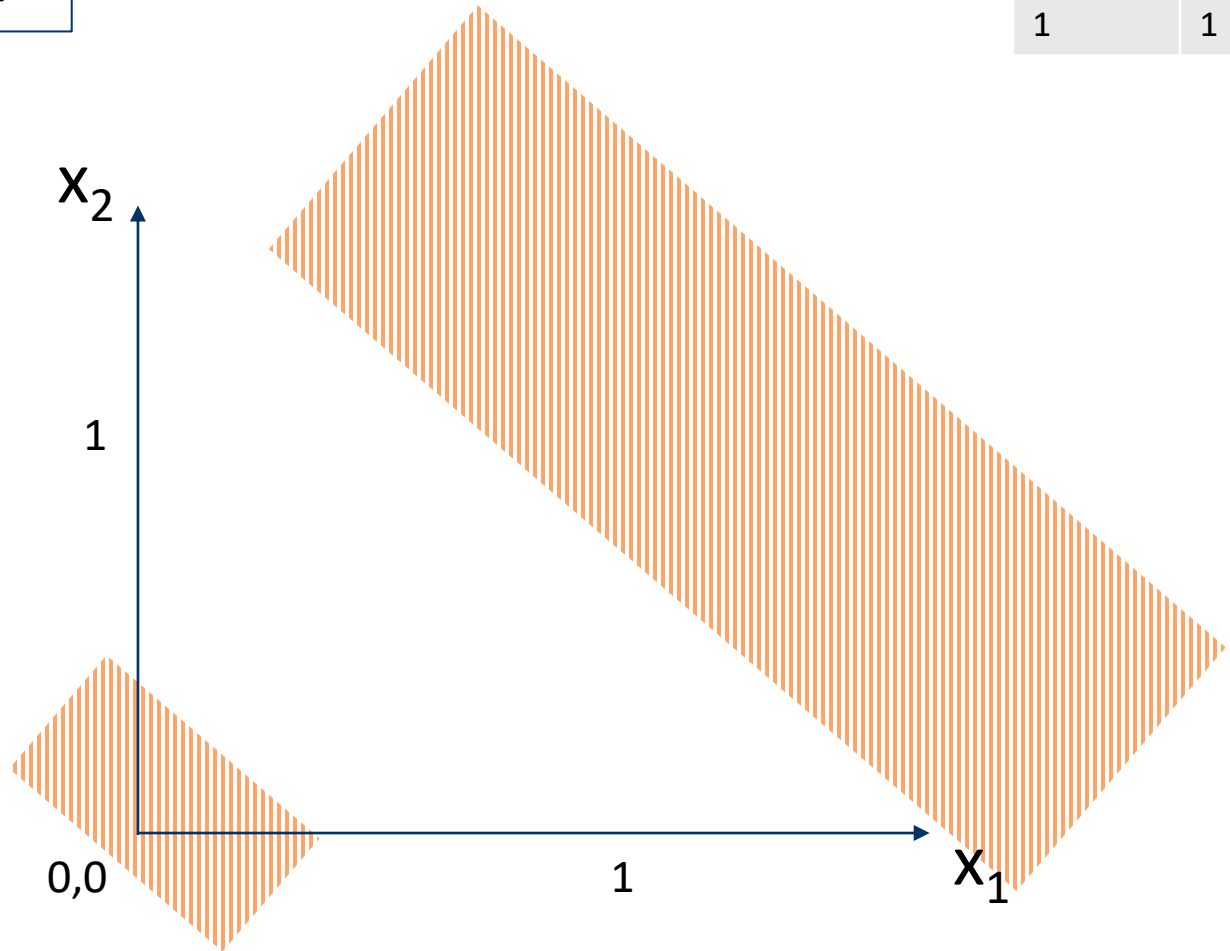
x1	output	value
0	0	True
1	-1	False

# How about the XOR gate?

XOR Gate

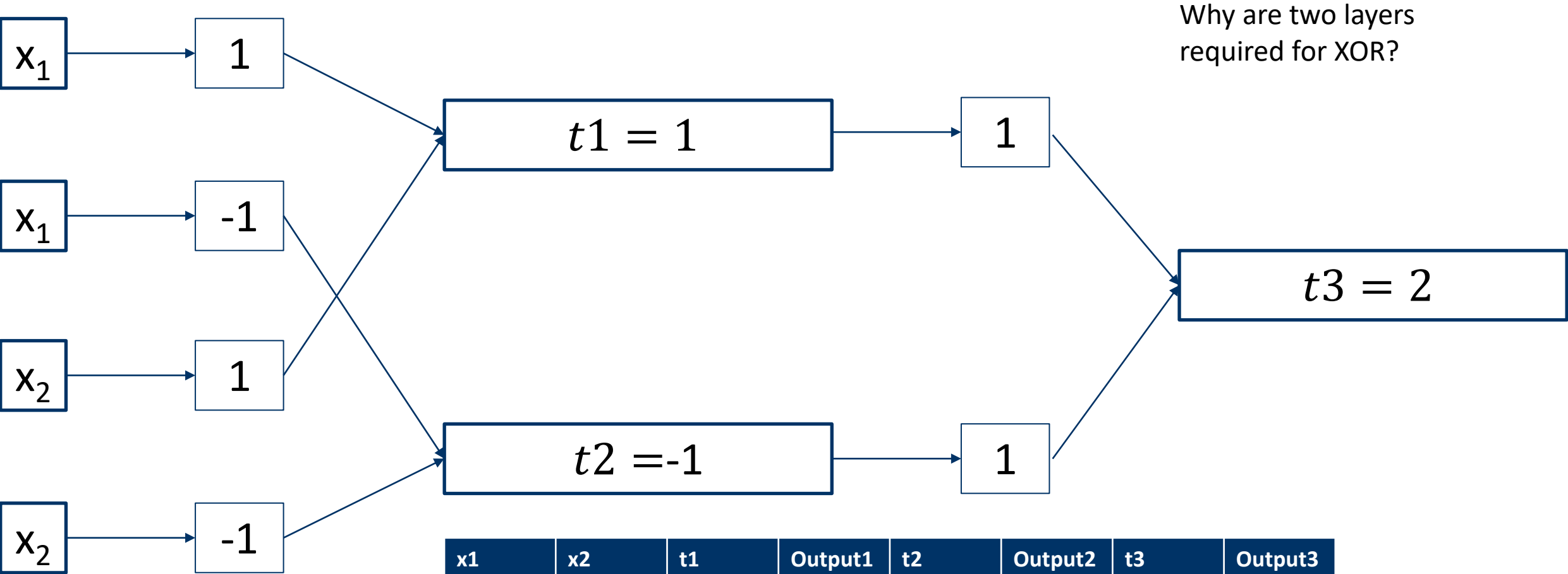


x1	x2	output	value
0	0	0	False
1	0	1	True
0	1	1	True
1	1	2	False





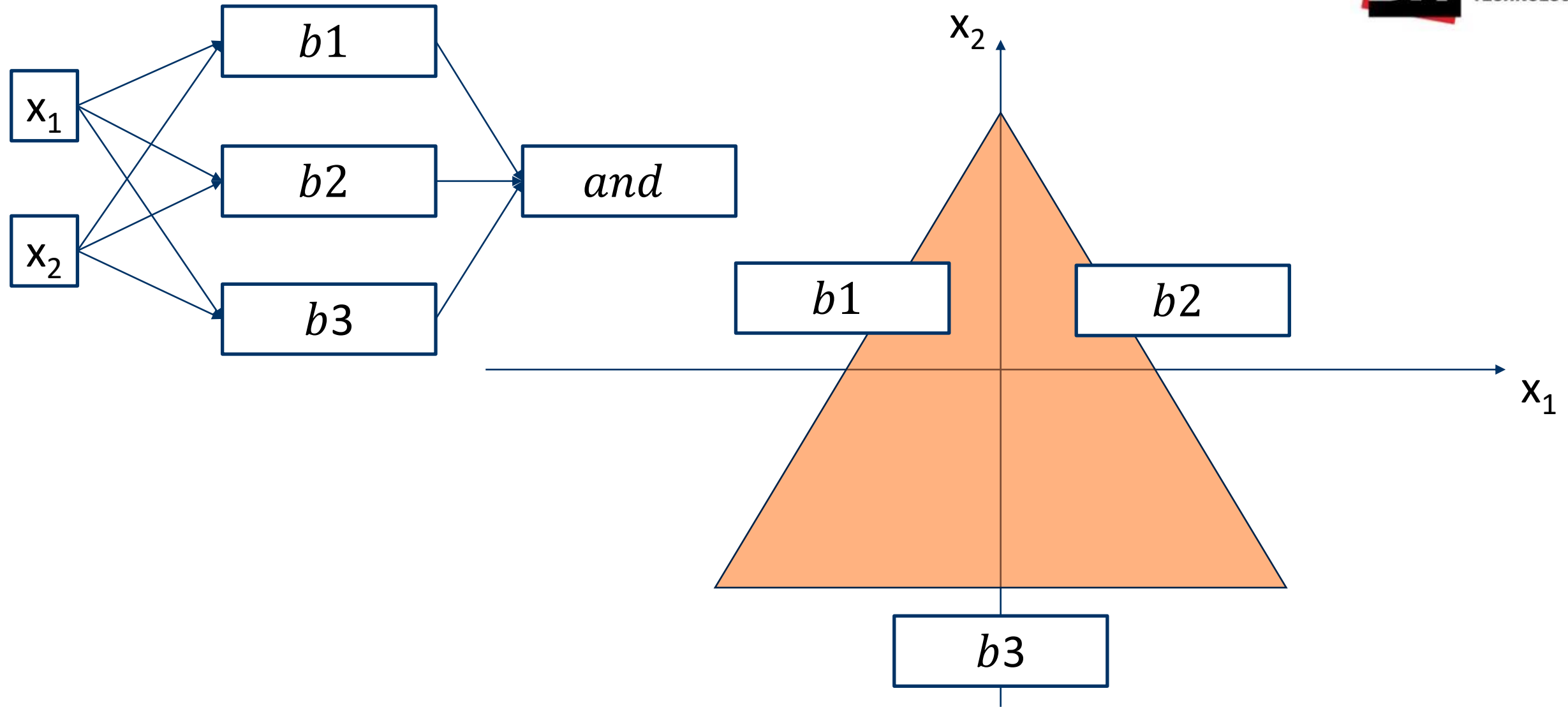
# Multi Layer Perceptron is required



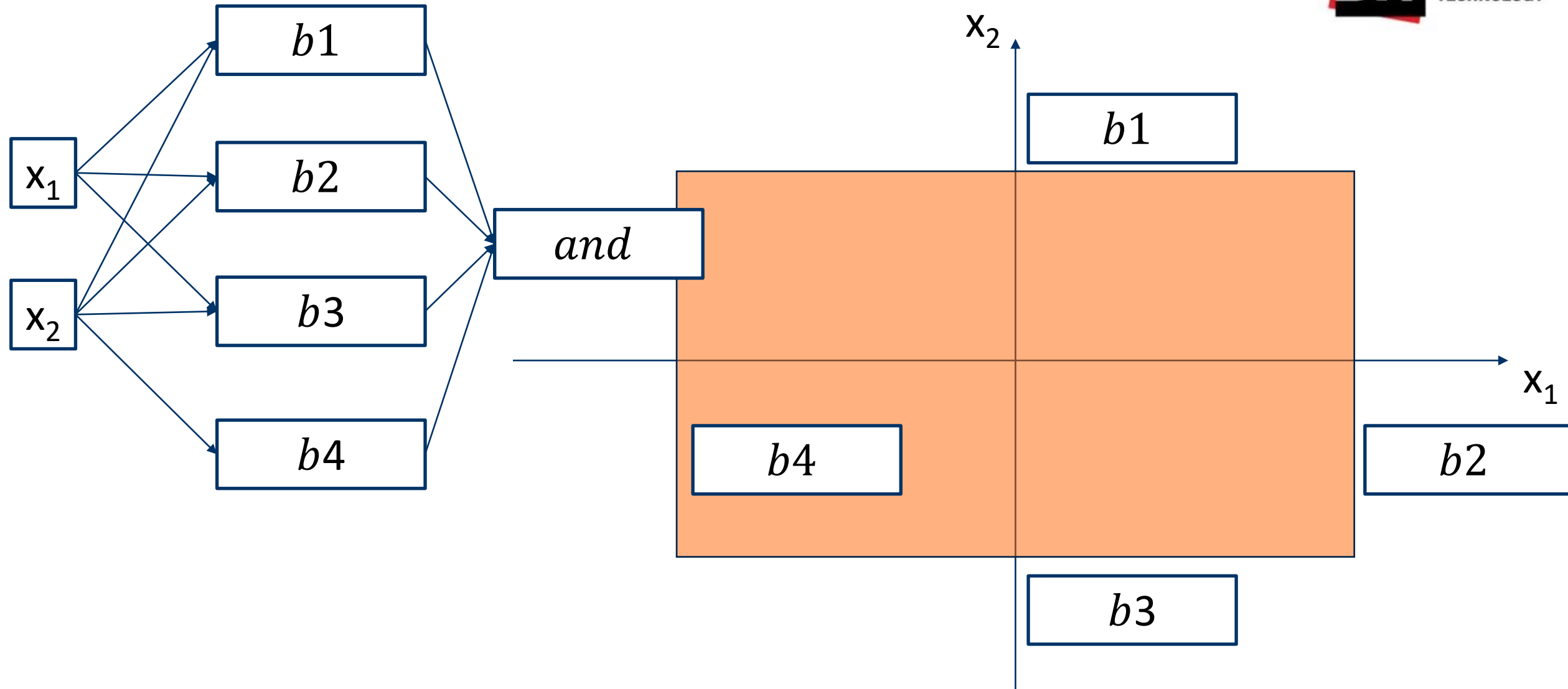
Why are two layers required for XOR?

x1	x2	t1	Output1	t2	Output2	t3	Output3
0	0	0	0 (False)	0	1 (True)	1	False
1	0	1	1 (True)	-1	1 (True)	2	True
0	1	1	1 (True)	-1	1 (True)	2	True
1	1	2	1 (True)	-2	0 (False)	1	False

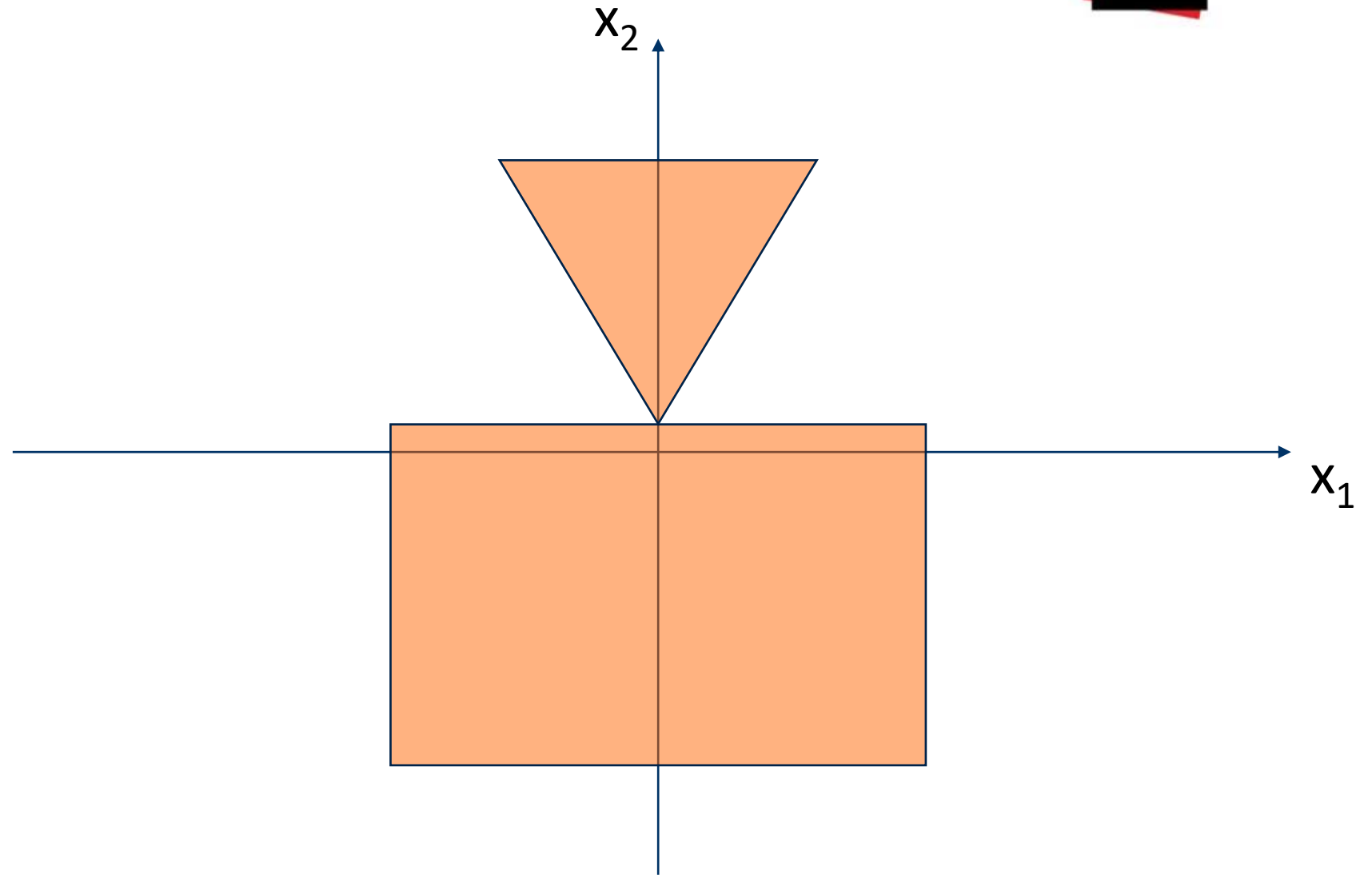
# Complex Decision Boundaries: How would the representation look like?



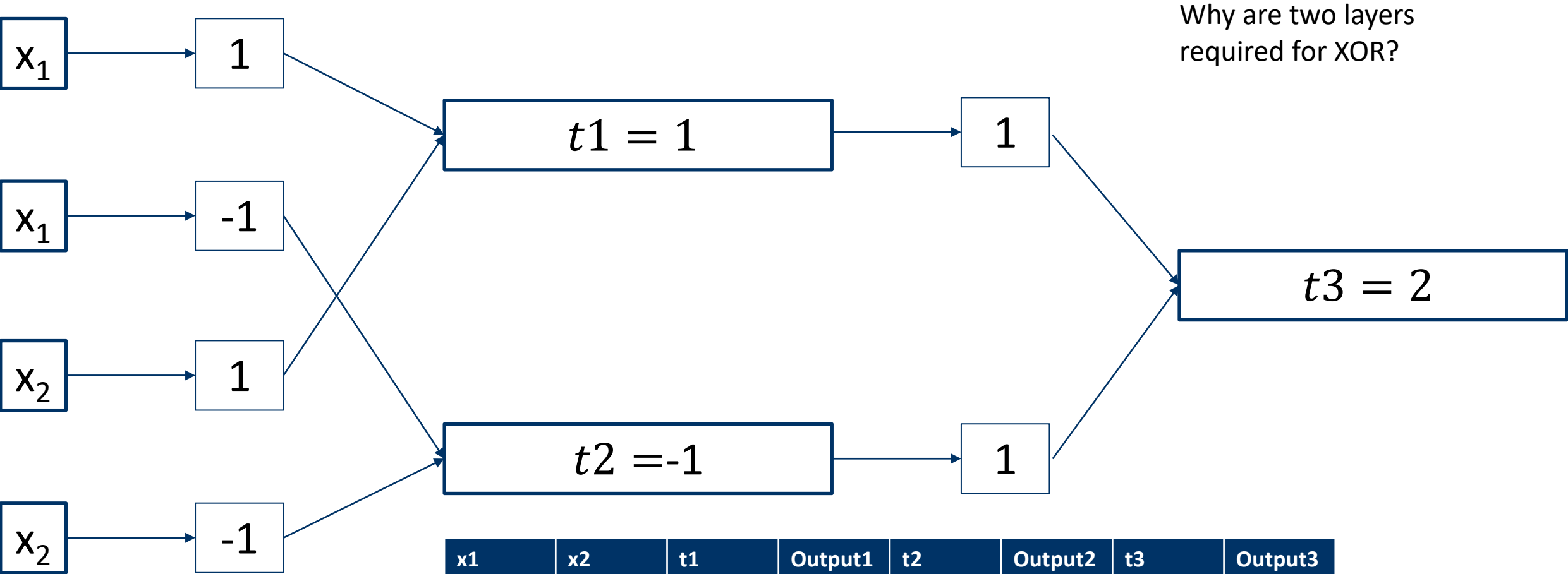
# Complex Decision Boundaries: How would the representation look like?



Complex Decision Boundaries: How would the representation look like with two hidden layers?



# Multi Layer Perceptron is required

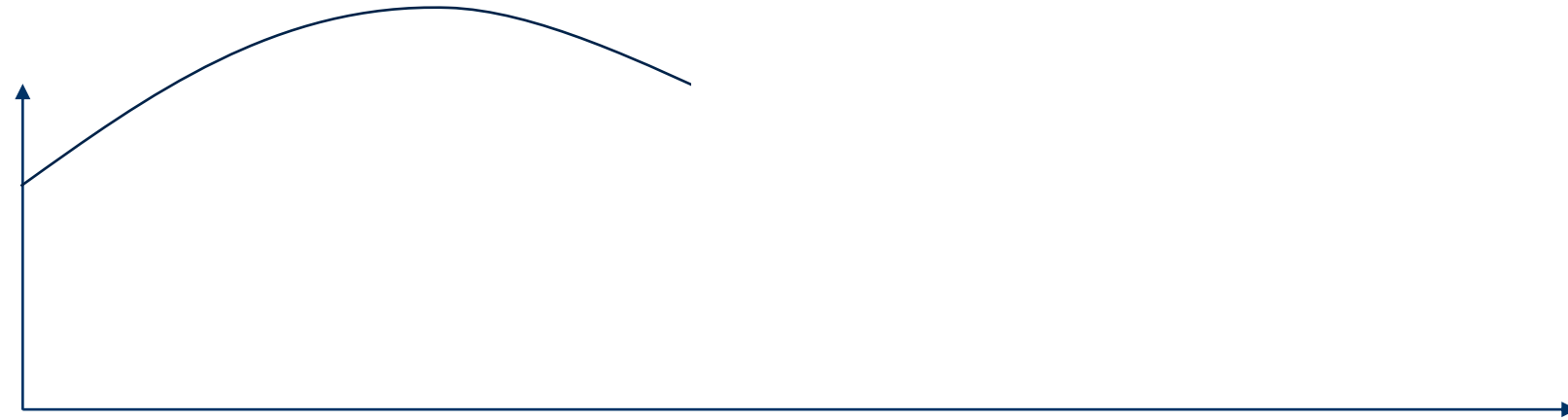


Why are two layers required for XOR?

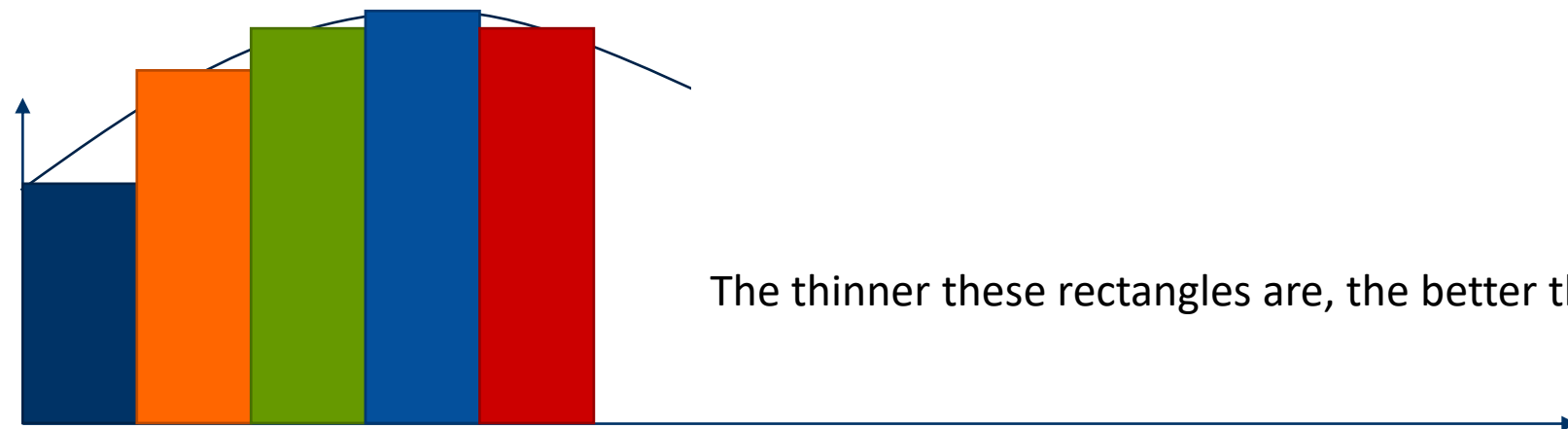
x1	x2	t1	Output1	t2	Output2	t3	Output3
0	0	0	0 (False)	0	1 (True)	1	False
1	0	1	1 (True)	-1	1 (True)	2	True
0	1	1	1 (True)	-1	1 (True)	2	True
1	1	2	1 (True)	-2	0 (False)	1	False

## What if the output value is a continuous real value?

Intuition: Suppose we have a continuous 2-d real value we wish to model with a neural network.



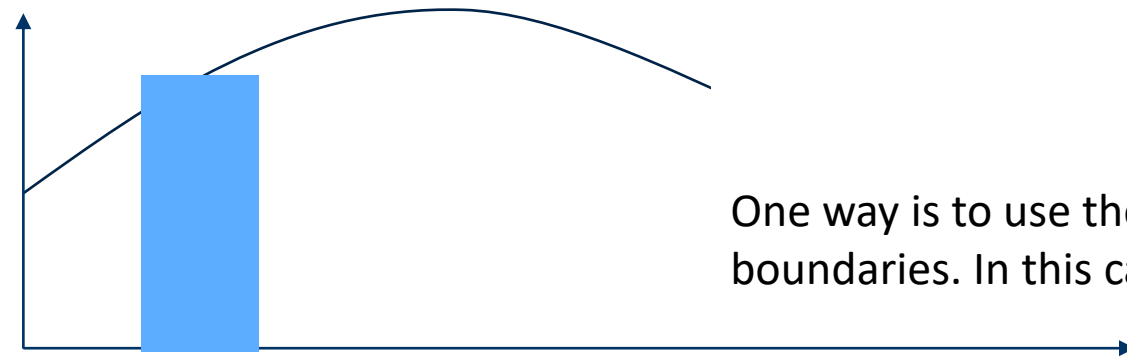
Intuition: Any continuous real value can be broken up into approximate rectangles.



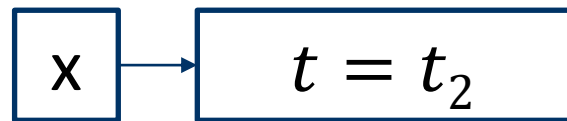
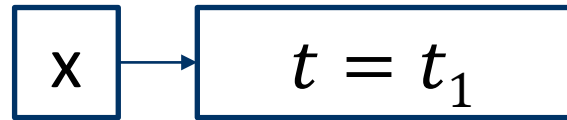
The thinner these rectangles are, the better the approximation.

# What if the output value is a continuous real value?

Challenge: Can we model one rectangle using a MLP?



One way is to use the technique we did earlier, having four boundaries. In this case, we can do a little bit better.

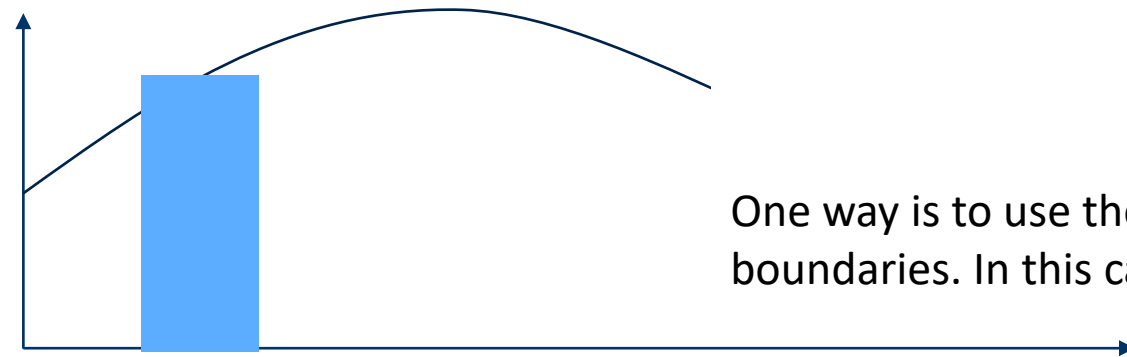


subtract

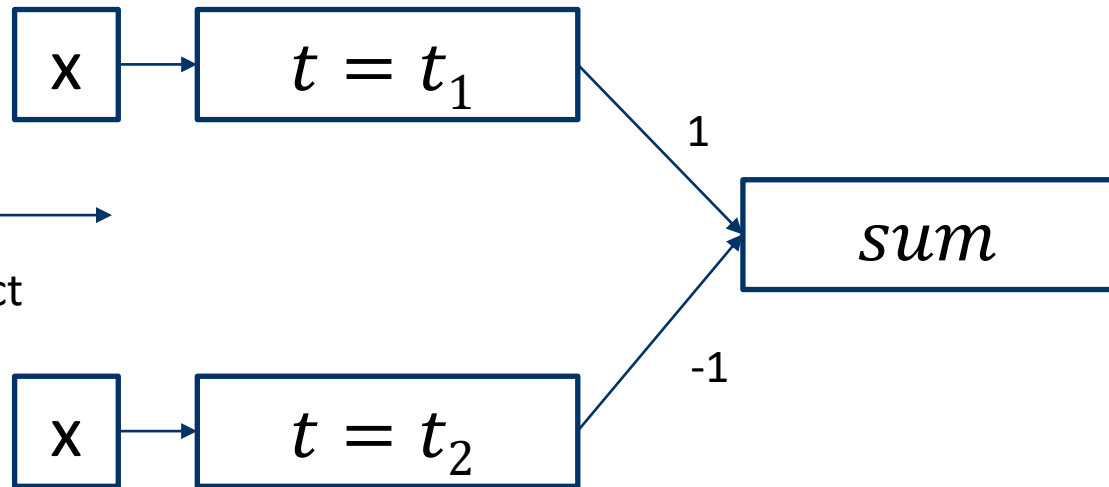


# What if the output value is a continuous real value?

Challenge: Can we model one rectangle using a MLP?

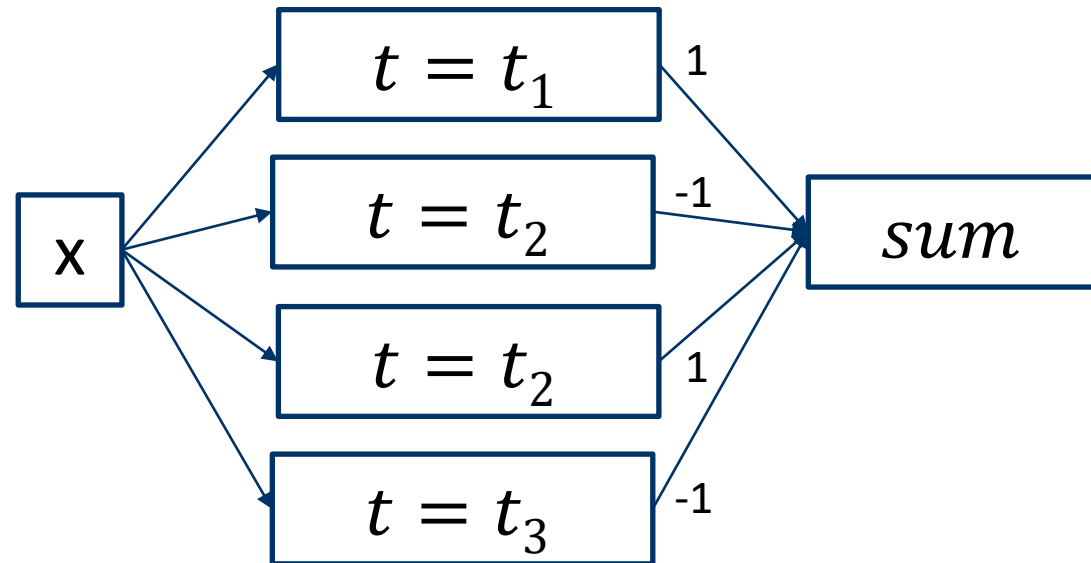
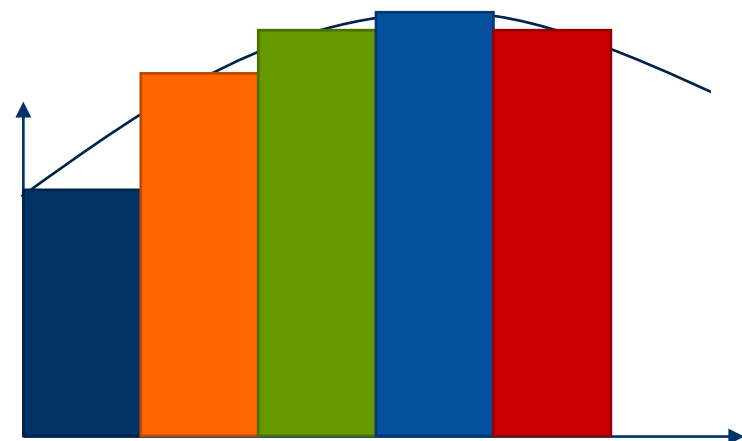
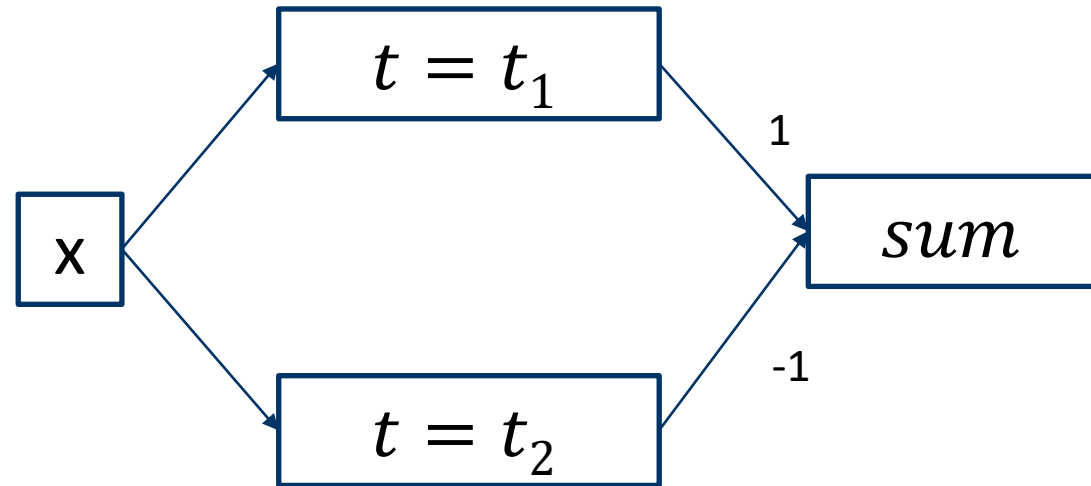
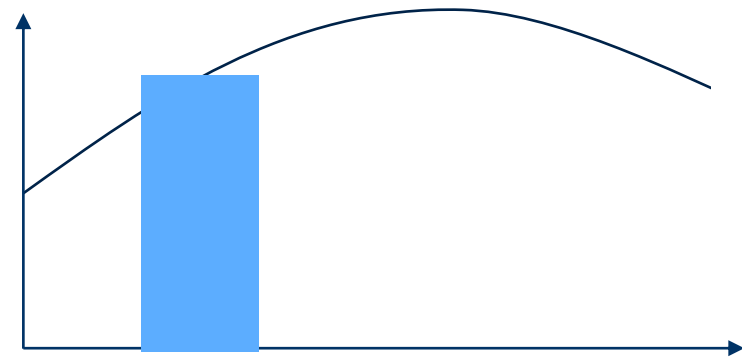


One way is to use the technique we did earlier, having four boundaries. In this case, we can do a little bit better.

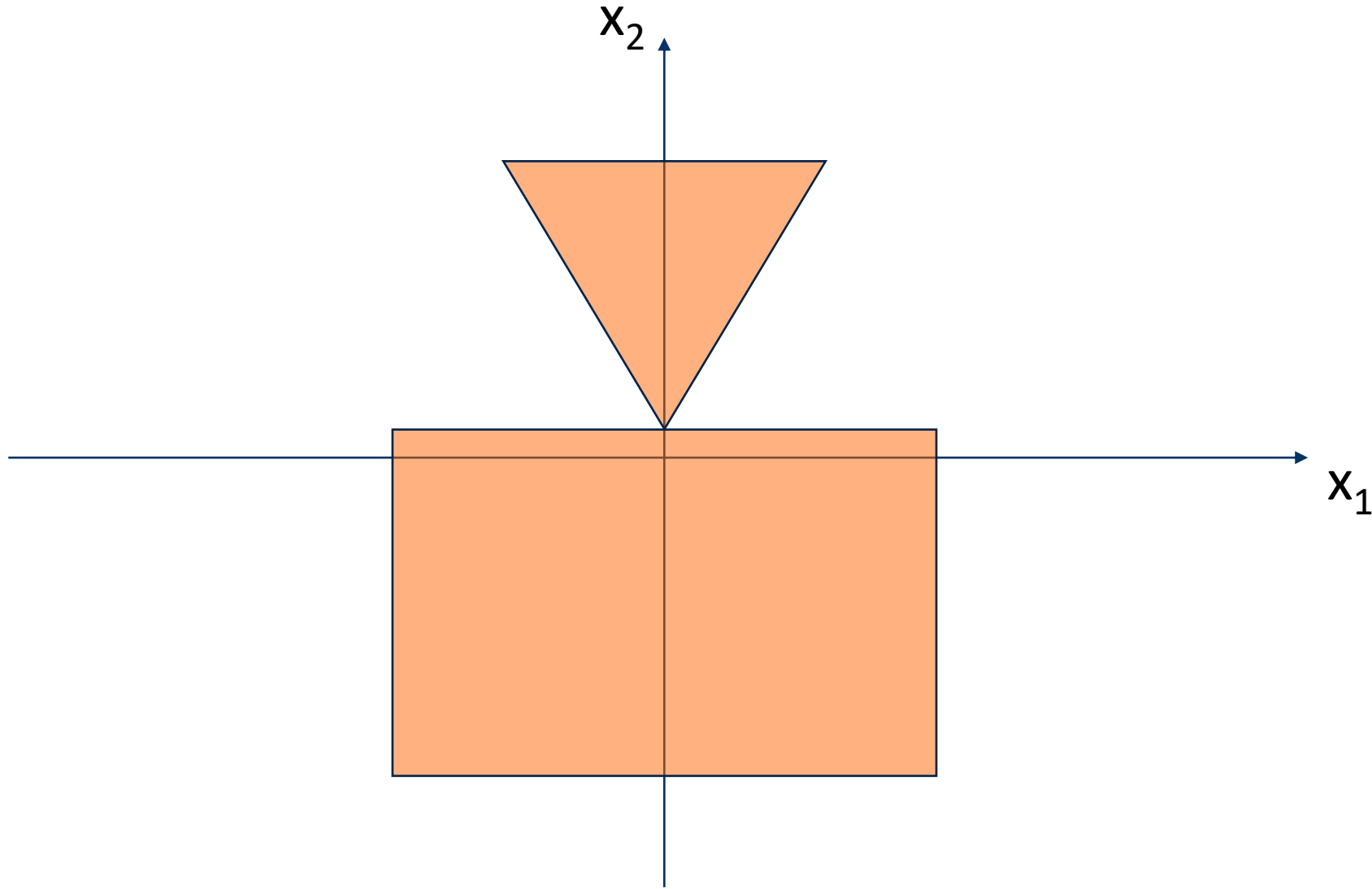


# What if the output value is a continuous real value?

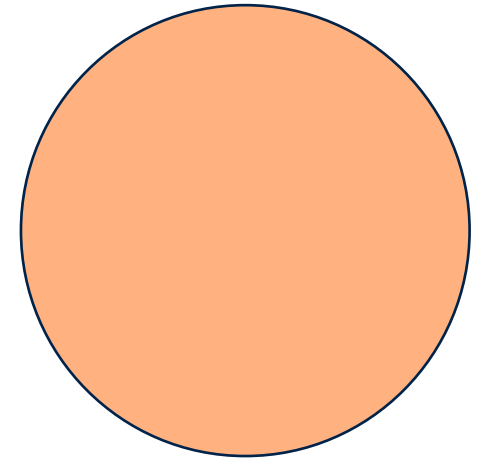
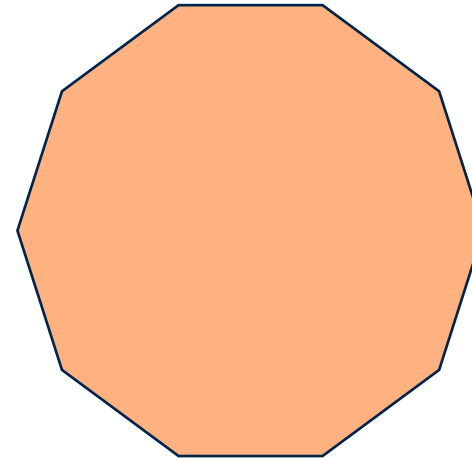
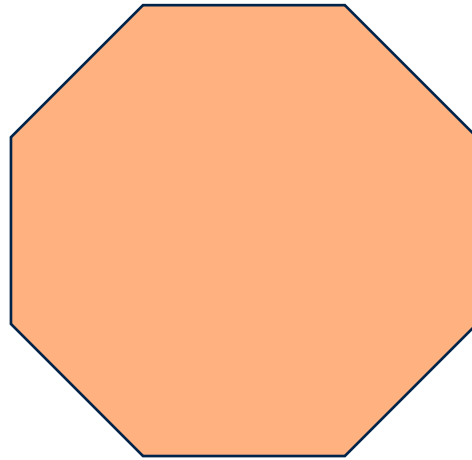
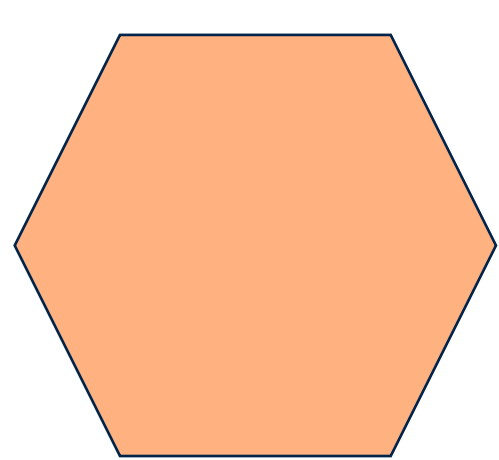
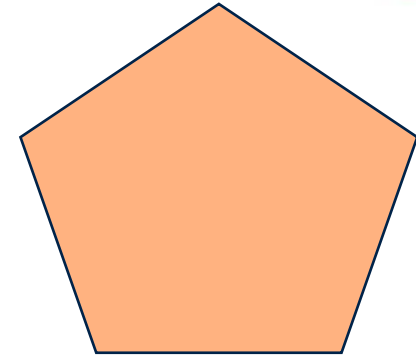
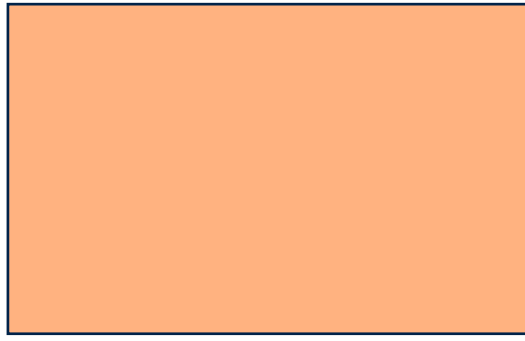
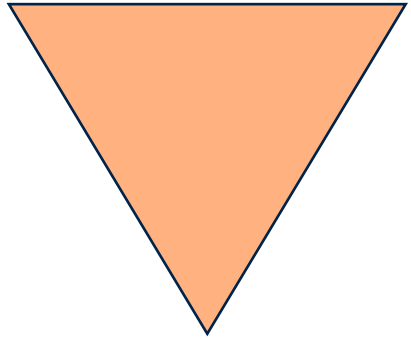
Challenge: Can we model one rectangle using a MLP?



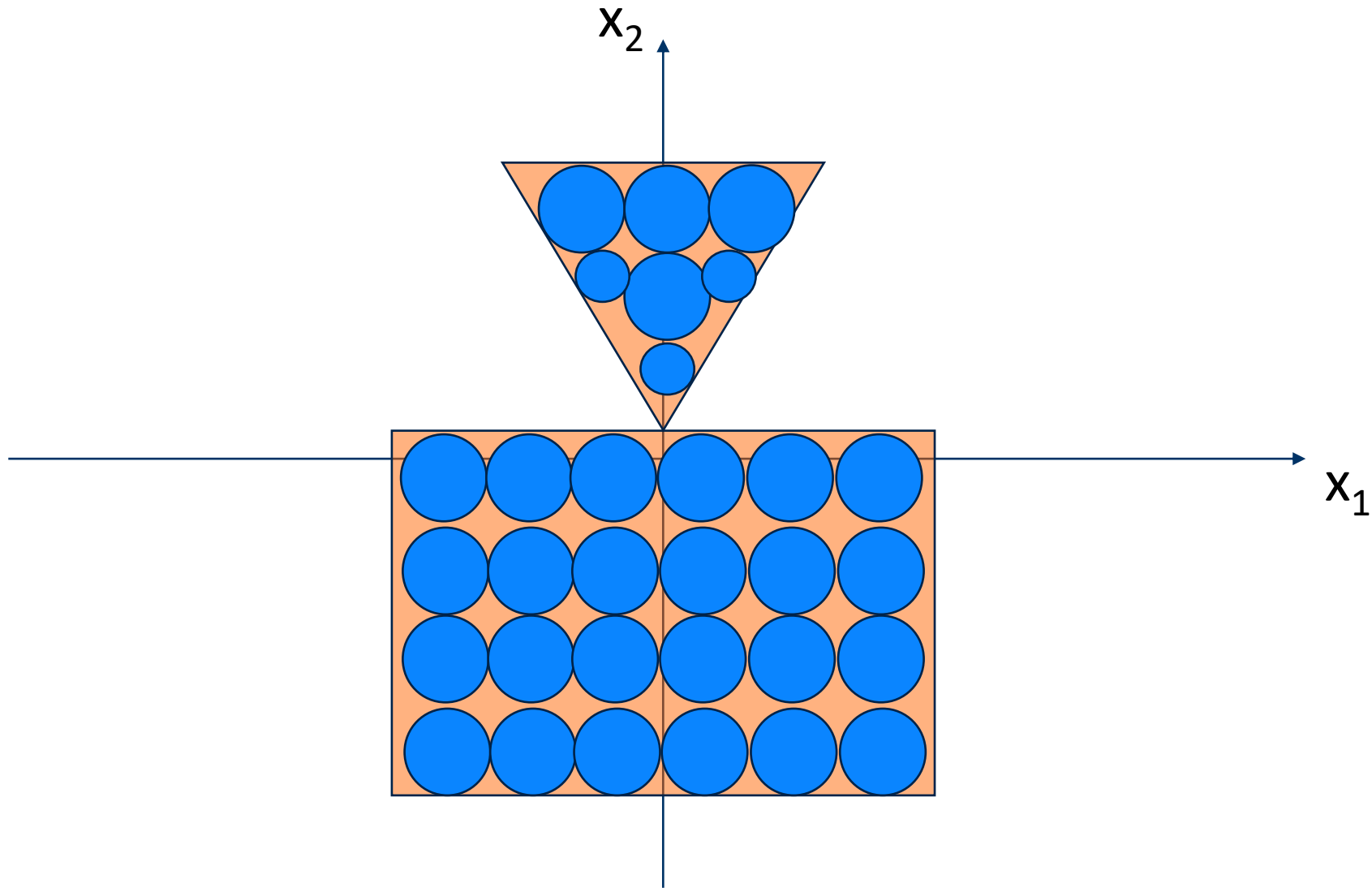
Complex Decision Boundaries: How would the representation look like with just one hidden layer?



Complex Decision Boundaries: How many neurons would we need if we have only one hidden layer?



# Complex Decision Boundaries: How would the representation look like?



What are the steps in training a single layer perceptron?