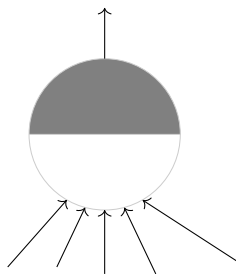
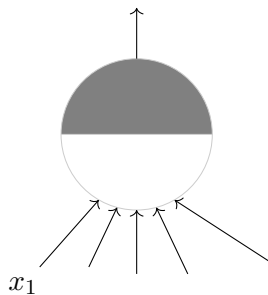


Module 2.2: McCulloch Pitts Neuron

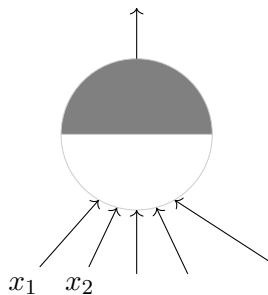
- McCulloch (neuroscientist) and Pitts (logician) proposed a highly simplified computational model of the neuron (1943)



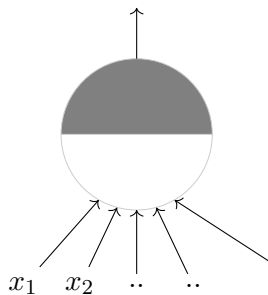
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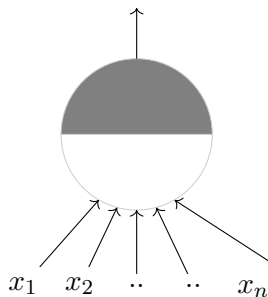
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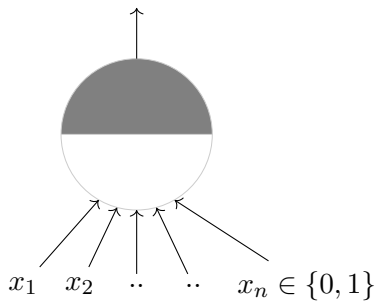
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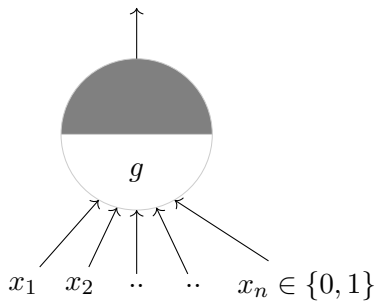


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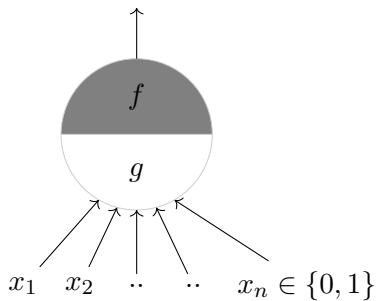


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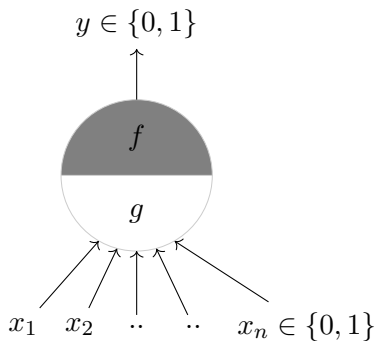




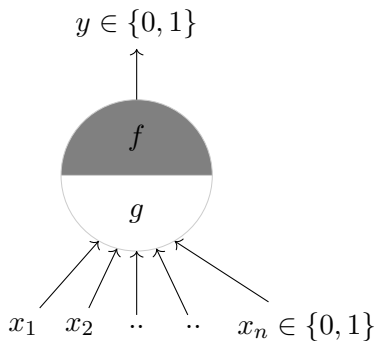
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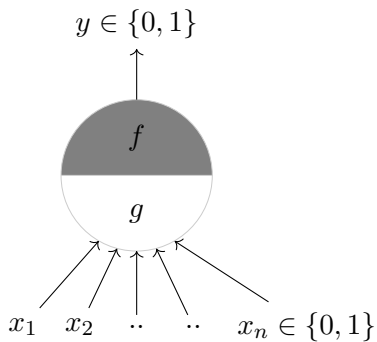
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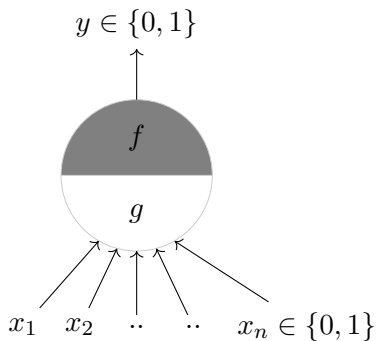
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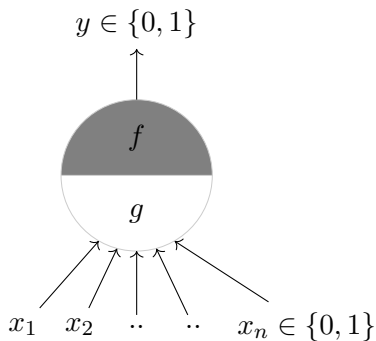


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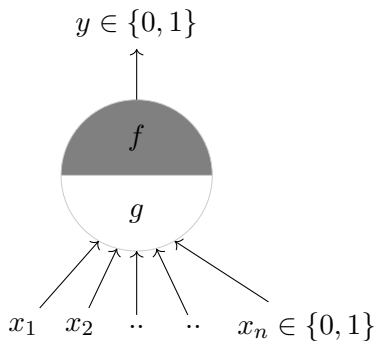
$$g(x_1, x_2, \dots, x_n) = g(\mathbf{x}) = \sum_{i=1}^n x_i$$



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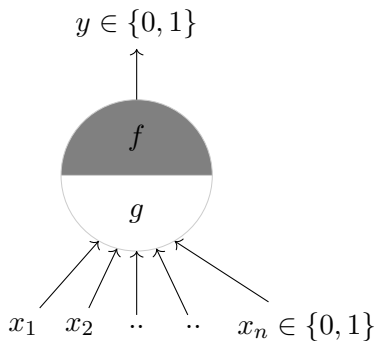
$$y = f(g(\mathbf{x})) = 1 \quad \text{if} \quad g(\mathbf{x}) \geq \theta$$



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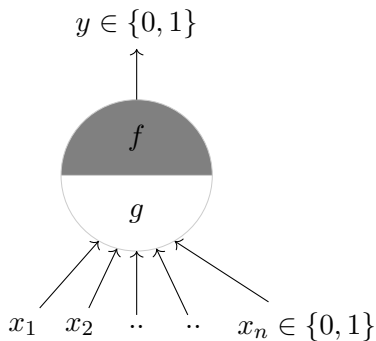


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- θ is called the thresholding parameter



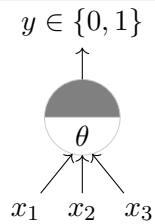
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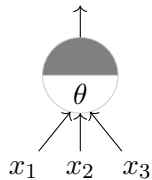
- θ is called the thresholding parameter
- This is called Thresholding Logic

Let us implement some boolean functions using this McCulloch Pitts (MP) neuron
...



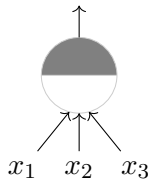
A McCulloch Pitts unit

$$y \in \{0, 1\}$$



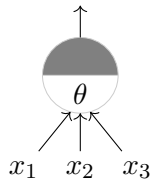
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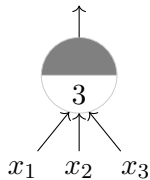
AND function

$$y \in \{0, 1\}$$



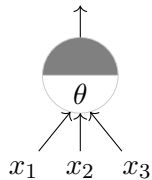
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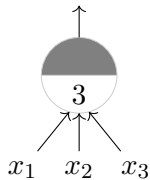
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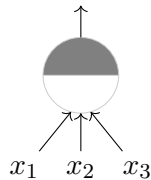
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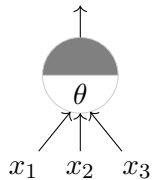
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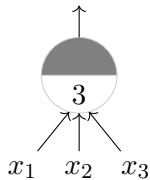
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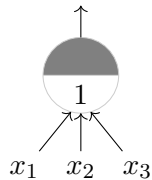
A McCulloch Pitts unit

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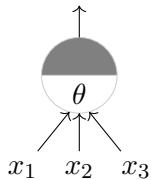
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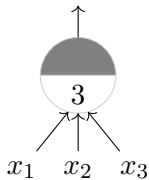
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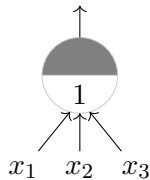
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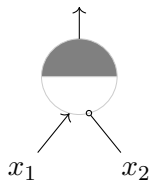
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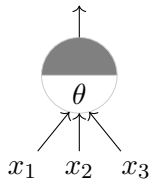
$$y \in \{0, 1\}$$



x_1 AND $!x_2$ *

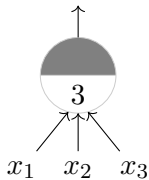
*circle at the end indicates inhibitory input: if any inhibitory input is 1 the output will be 0

$$y \in \{0, 1\}$$



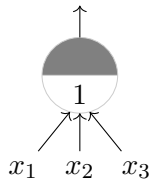
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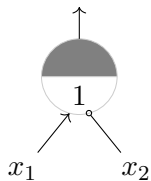
AND function

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OR function

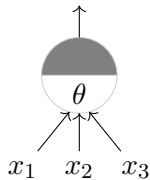
$$y \in \{0, 1\}$$



x_1 AND $\neg x_2$ *

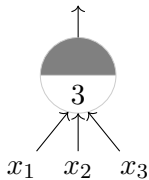
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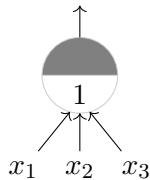
A McCulloch Pitts unit

$$y \in \{0, 1\}$$



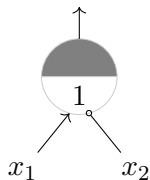
AND function

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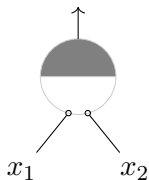
OR function

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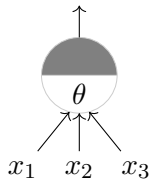
$$y \in \{0, 1\}$$



NOR function

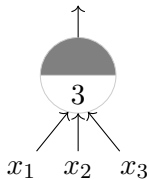
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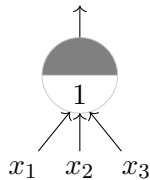
A McCulloch Pitts unit

$$y \in \{0, 1\}$$



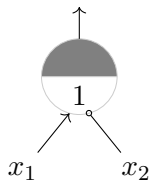
AND function

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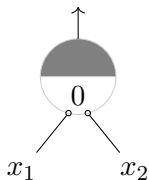
OR function

$$y \in \{0, 1\}$$



x_1 AND $!x_2^*$

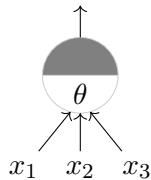
$$y \in \{0, 1\}$$



NOR function

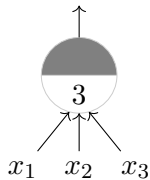
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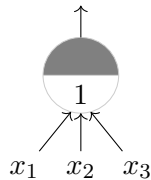
A McCulloch Pitts unit

$$y \in \{0, 1\}$$



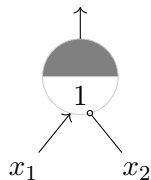
AND function

$$y \in \{0, 1\}$$



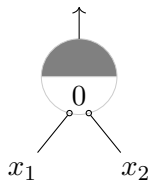
OR function

$$y \in \{0, 1\}$$



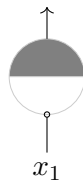
x_1 AND $!x_2^*$

$$y \in \{0, 1\}$$



NOR function

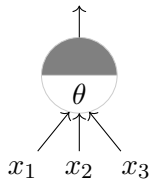
$$y \in \{0, 1\}$$



NOT function

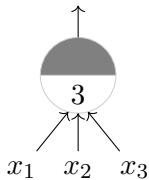
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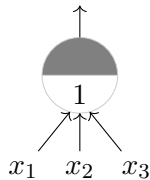
A McCulloch Pitts unit

$$y \in \{0, 1\}$$



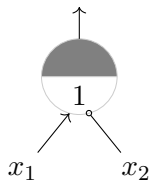
AND function

$$y \in \{0, 1\}$$



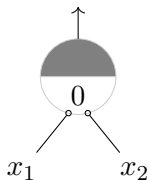
OR function

$$y \in \{0, 1\}$$



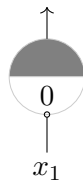
x_1 AND $!x_2^*$

$$y \in \{0, 1\}$$



NOR function

$$y \in \{0, 1\}$$

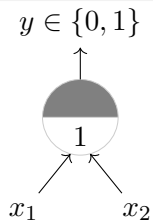


NOT function

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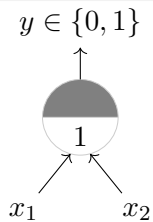
- Can any boolean function be represented using a McCulloch Pitts unit ?

- Can any boolean function be represented using a McCulloch Pitts unit ?
- Before answering this question let us first see the geometric interpretation of a MP unit ...



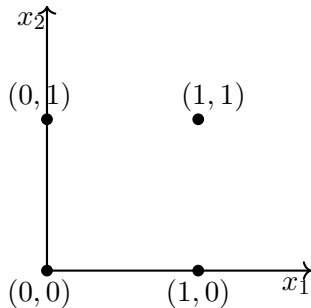
OR function

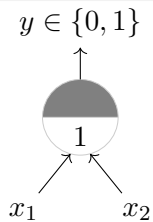
$$x_1 + x_2 = \sum_{i=1}^2 x_i \geq 1$$



OR function

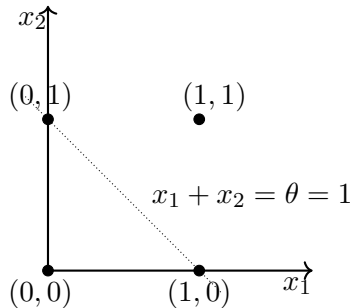
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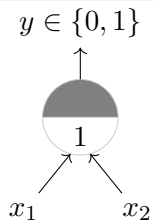




OR function

$$x_1 + x_2 = \sum_{i=1}^2 x_i \geq 1$$

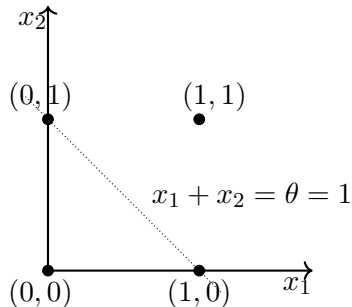


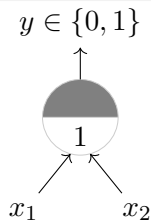


- A single MP neuron splits the input points (4 points for 2 binary inputs) into two halves

OR function

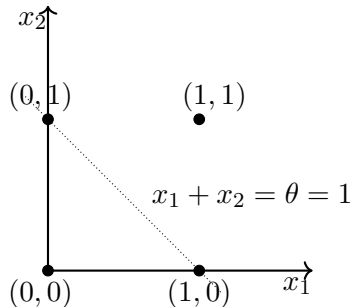
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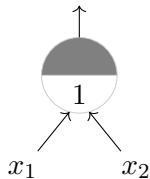
OR function

$$x_1 + x_2 = \sum_{i=1}^2 x_i \geq 1$$



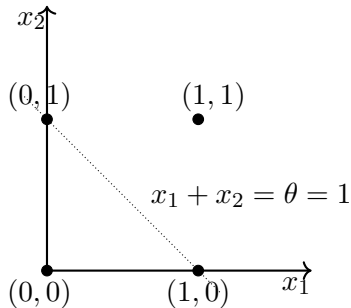
- A single MP neuron splits the input points (4 points for 2 binary inputs) into two halves
- Points lying on or above the line $\sum_{i=1}^n x_i - \theta = 0$ and points lying below this line

$$y \in \{0, 1\}$$



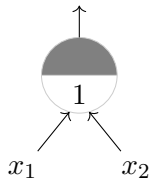
OR function

$$x_1 + x_2 = \sum_{i=1}^2 x_i \geq 1$$



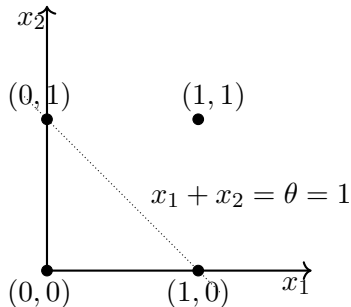
- A single MP neuron splits the input points (4 points for 2 binary inputs) into two halves
- Points lying on or above the line $\sum_{i=1}^n x_i - \theta = 0$ and points lying below this line
- In other words, all inputs which produce an output 0 will be on one side ($\sum_{i=1}^n x_i < \theta$) of the line and all inputs which produce an output 1 will lie on the other side ($\sum_{i=1}^n x_i \geq \theta$) of this line

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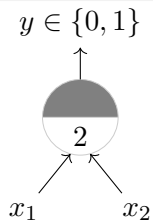


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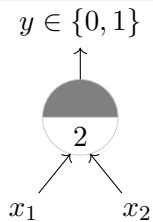


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- Let us convince ourselves about this with a few more examples (if it is not already clear from the math)



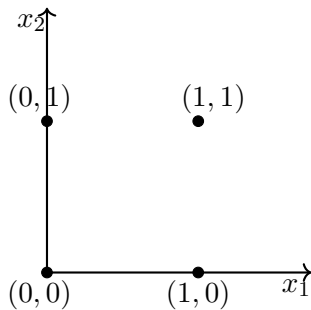
AND function

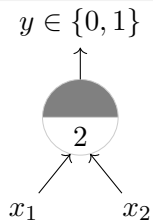
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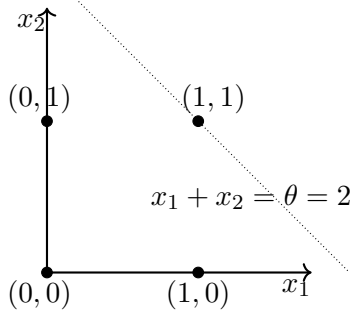
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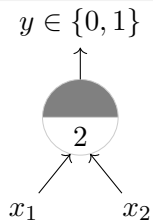




AND function

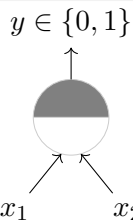
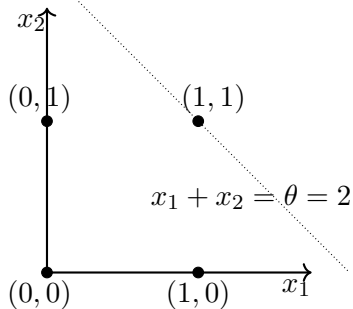
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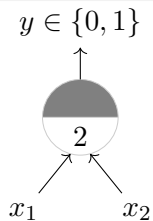


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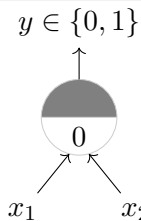
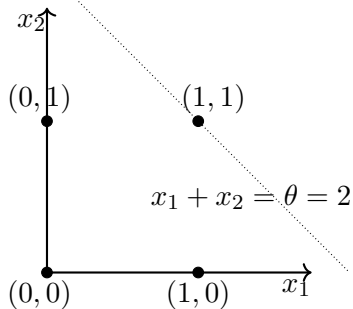


Tautology (always ON)

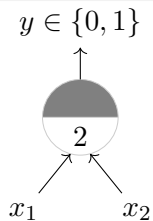


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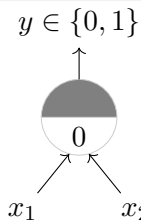
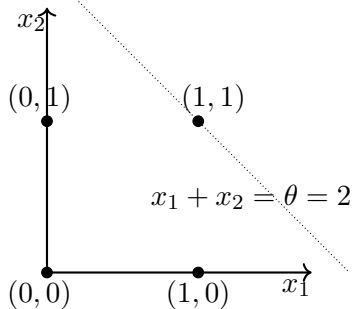


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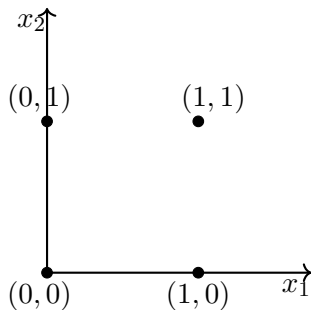


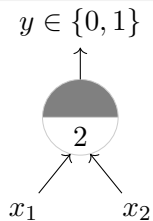
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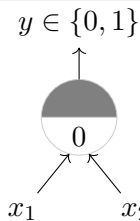
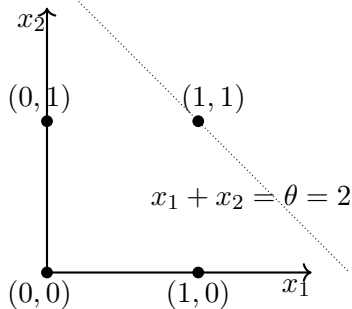
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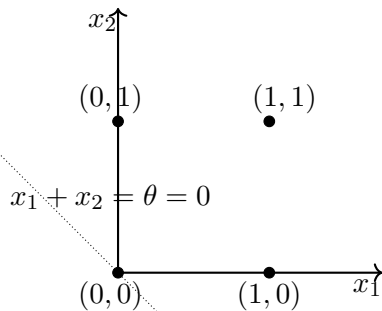


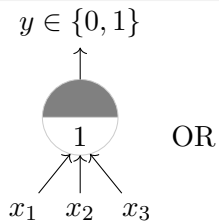
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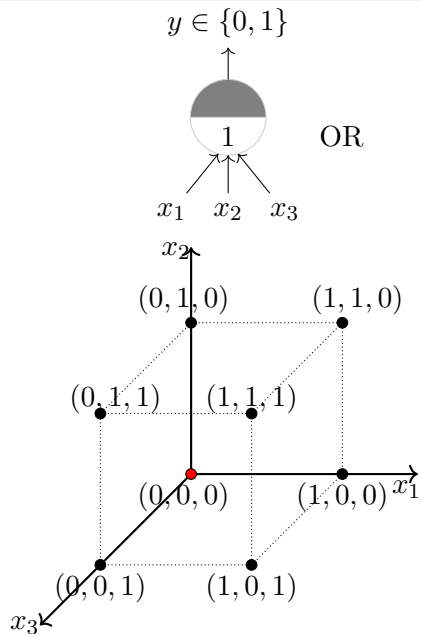
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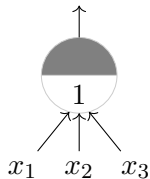


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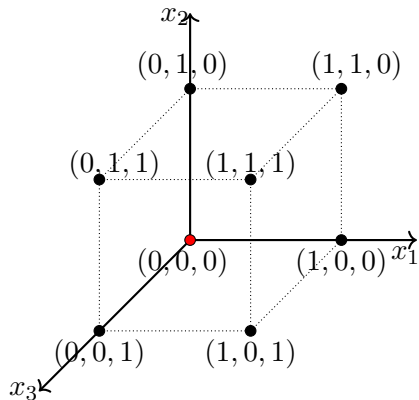
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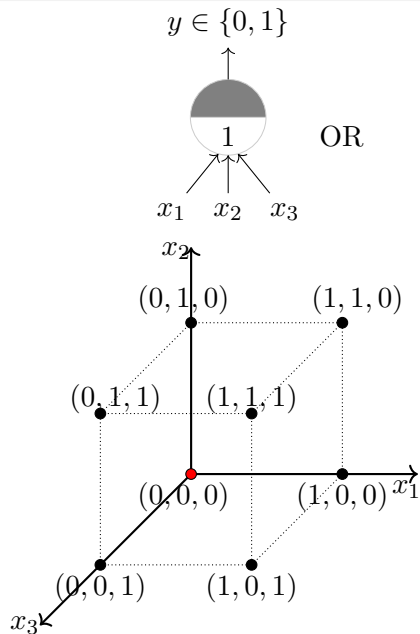
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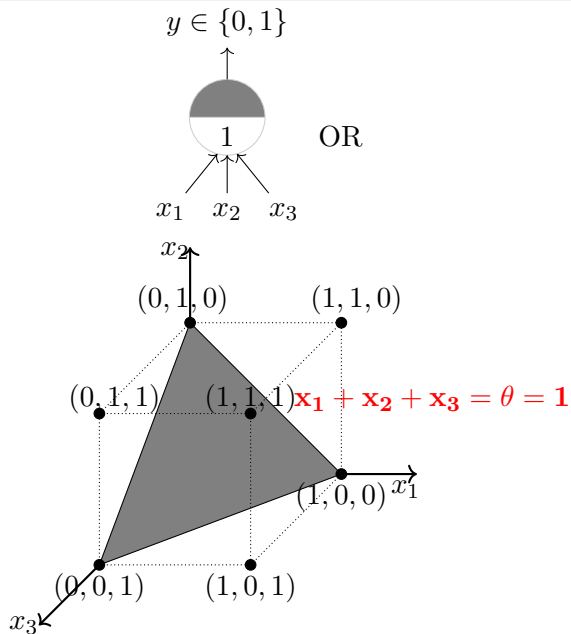
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- Linear separability (for boolean functions) : There exists a line (plane) such that all inputs which produce a 1 lie on one side of the line (plane) and all inputs which produce a 0 lie on other side of the line (plane)