

Introdução ao TensorFlow para Redes Neurais

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Apresentação

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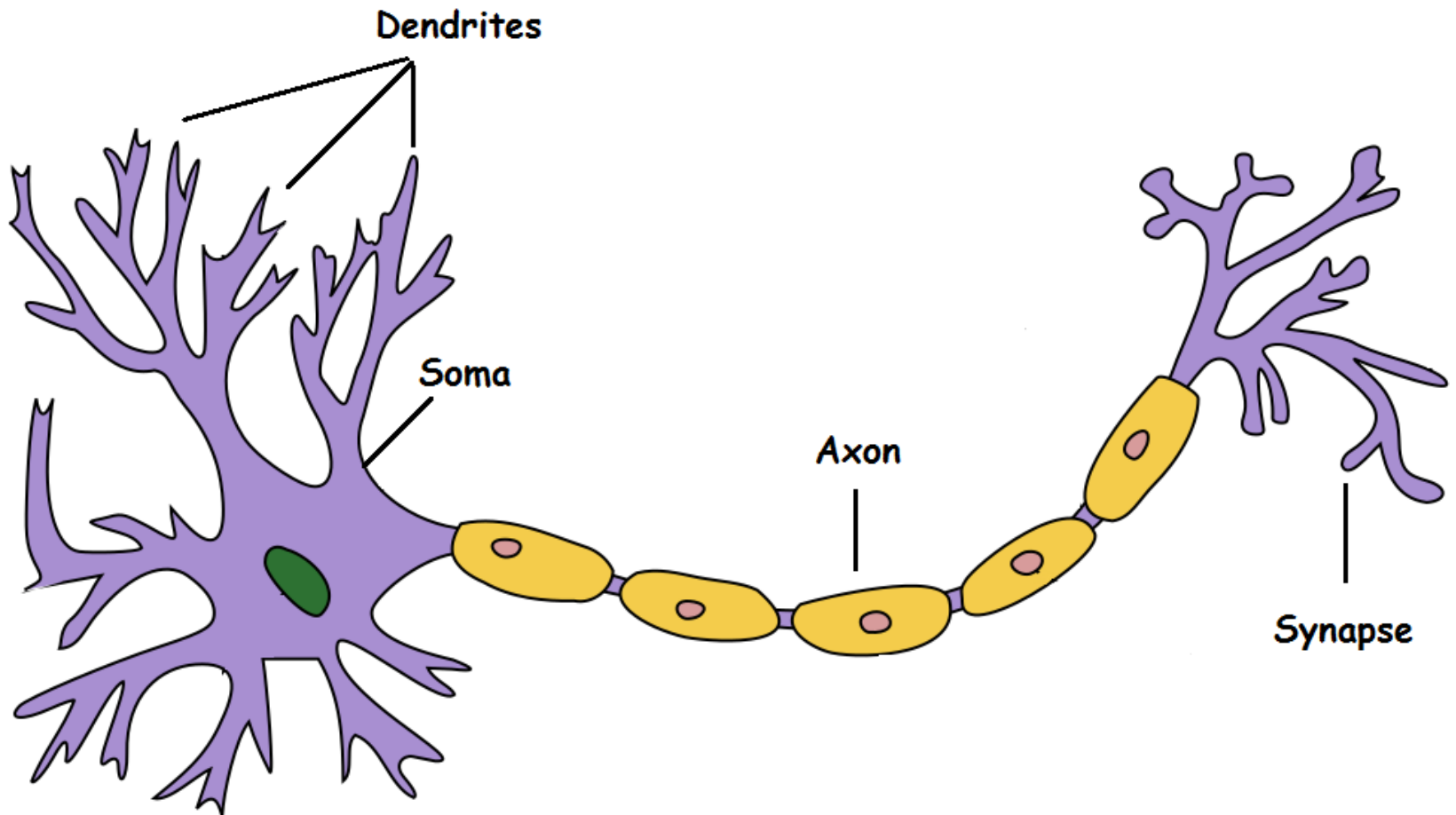
Introdução às Redes Neurais

Funcionamento inspirado no neurônio biológico

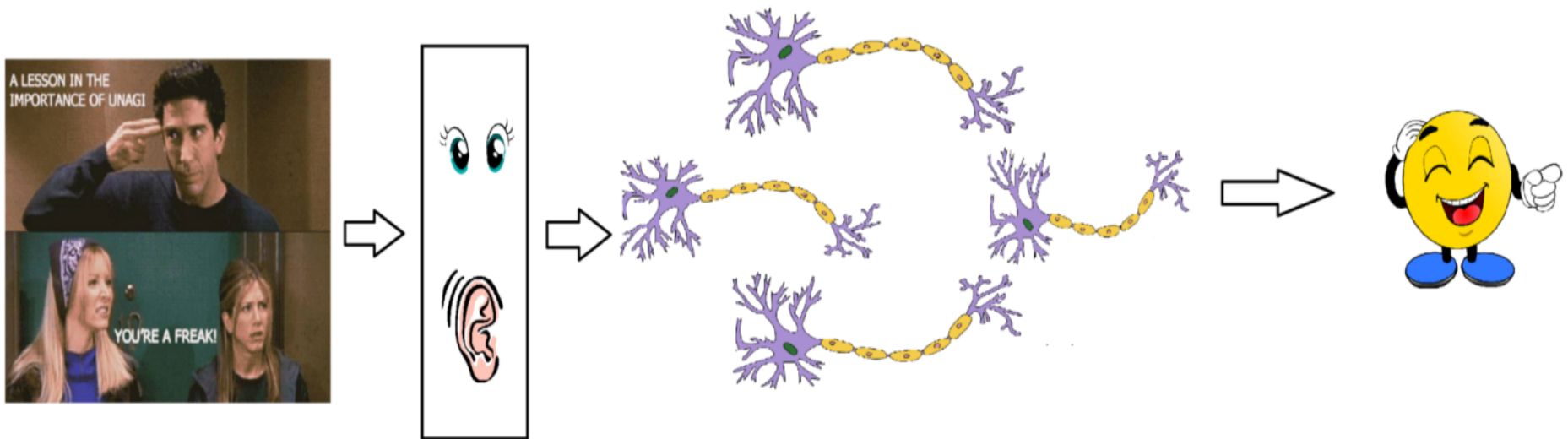
Neurônio Artificial de McCulloch-Pitts (1943)



Neurônio

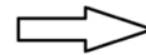
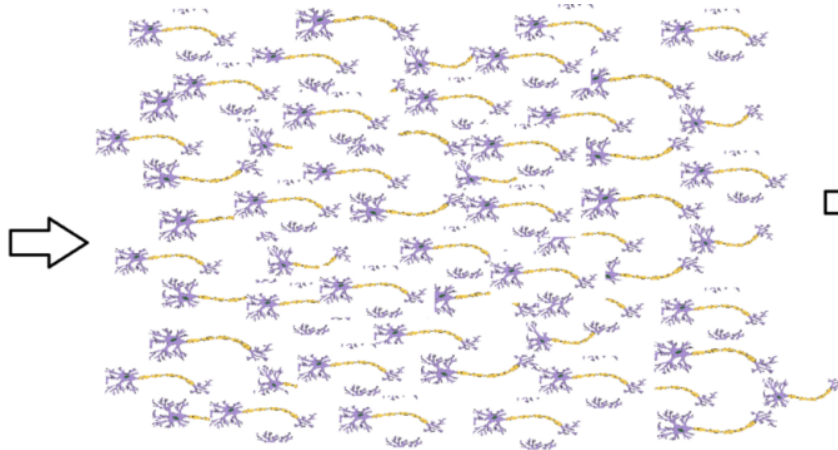
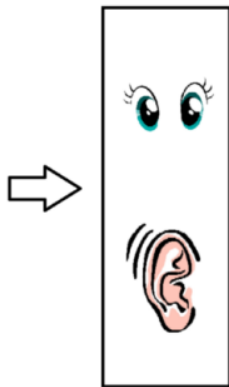


Funcionamento dos Neurônios

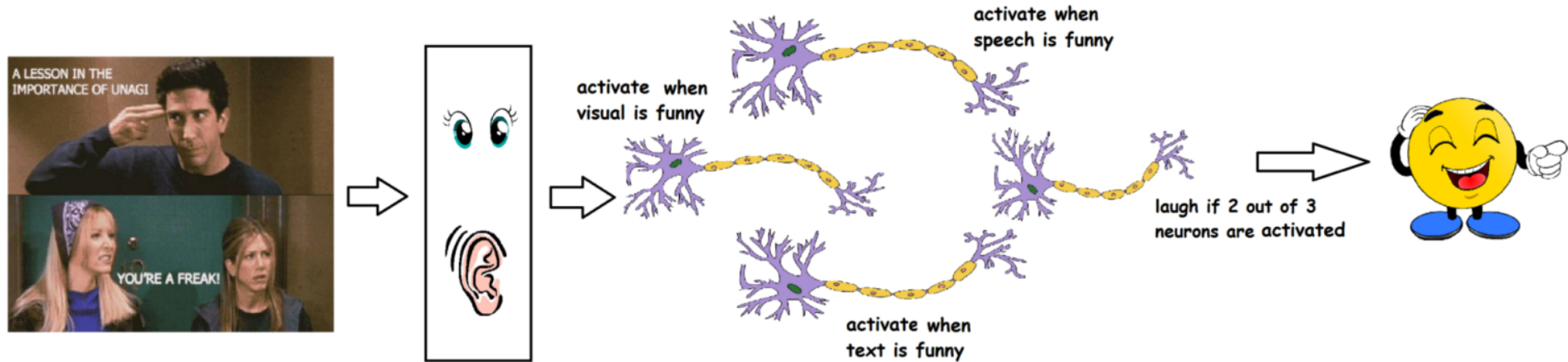


Funcionamento dos Neurônios

10^{11} neurônios (100 bilhões)



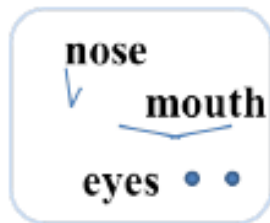
Funcionamento dos Neurônios



Funcionamento dos Neurônios



Layer 1: detect edges & corners



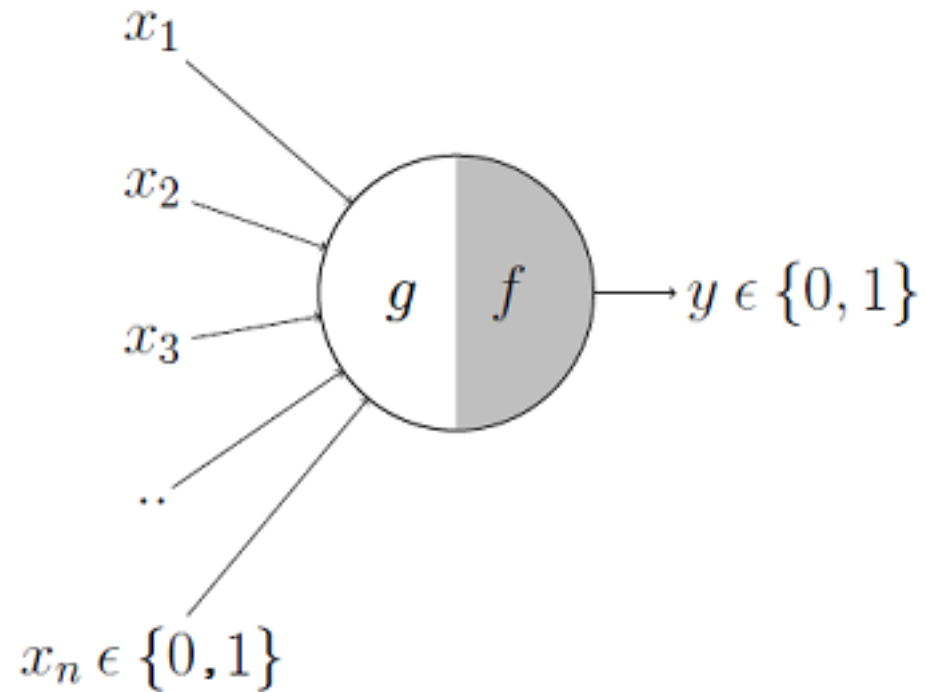
Layer 2: form feature groups



**Layer 3: detect high level
objects, faces, etc.**



Modelo de McCulloch-Pitts



Perceptron

McCulloch Pitts Neuron
(assuming no inhibitory inputs)

$$y = 1 \quad \text{if} \sum_{i=0}^n x_i \geq 0$$
$$= 0 \quad \text{if} \sum_{i=0}^n x_i < 0$$

Perceptron

$$y = 1 \quad \text{if} \sum_{i=0}^n w_i * x_i \geq 0$$
$$= 0 \quad \text{if} \sum_{i=0}^n w_i * x_i < 0$$



Predição de um Perceptron

x_1 - É um jogo do Brasileirão?

x_2 - É um amistoso?

x_3 - Estarei em casa?

x_4 - É jogo do Ceará?

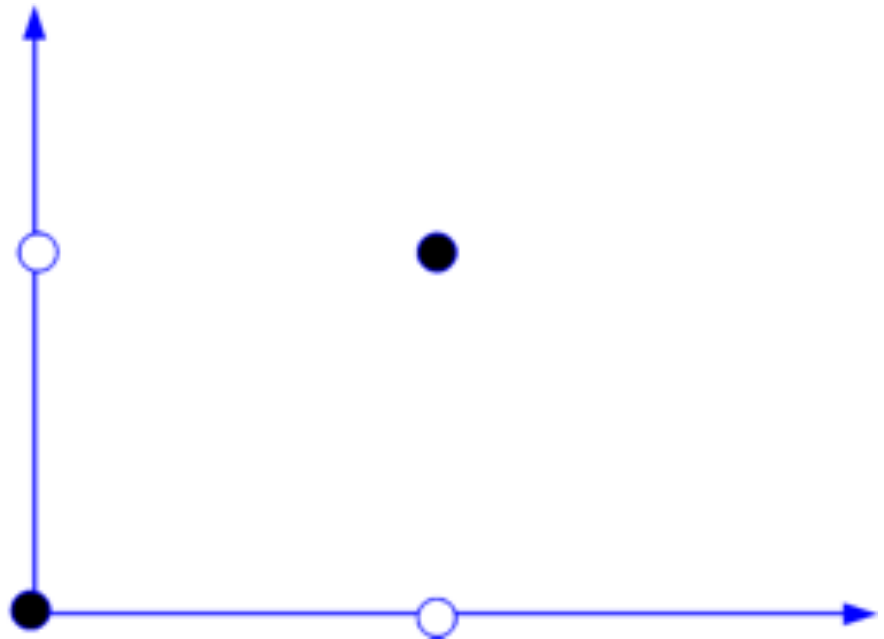
$$g(x_1, x_2, x_3, \dots, x_n) = g(\mathbf{x}) = \sum_{i=1}^n x_i$$

$$\begin{aligned} y = f(g(\mathbf{x})) &= 1 && \text{if } g(\mathbf{x}) \geq \theta \\ &= 0 && \text{if } g(\mathbf{x}) < \theta \end{aligned}$$

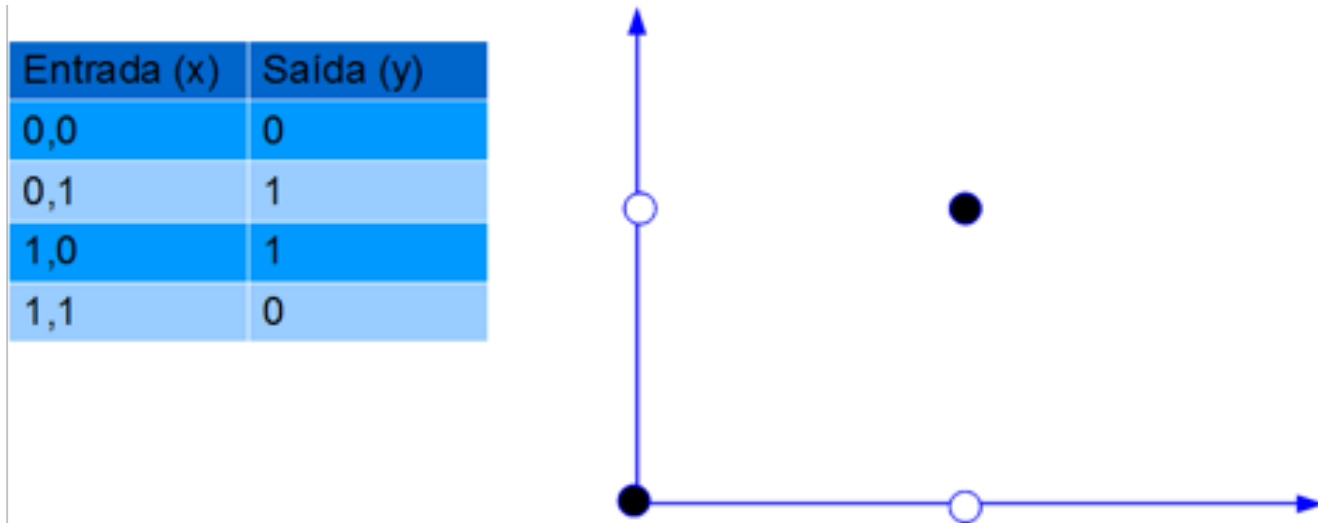


Impacto dos neurônios

Entrada (x)	Saída (y)
0,0	0
0,1	1
1,0	1
1,1	0



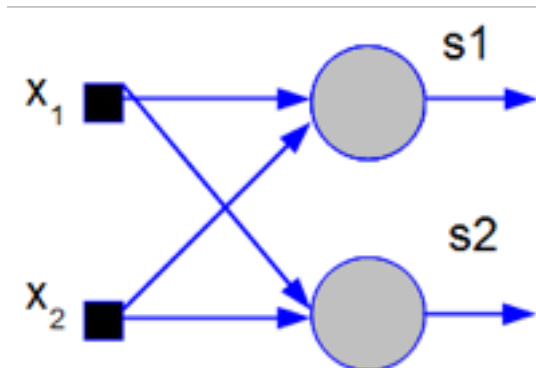
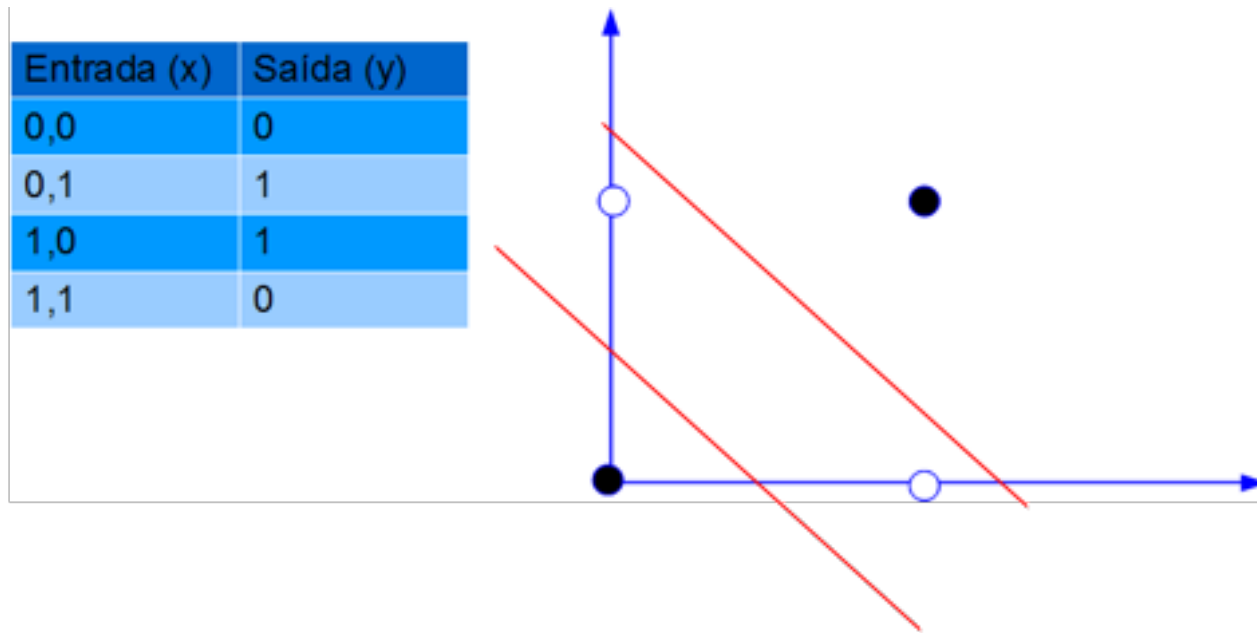
Impacto dos neurônios



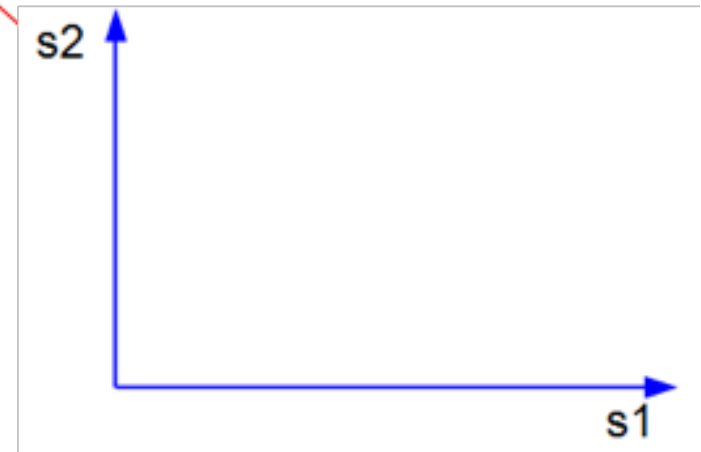
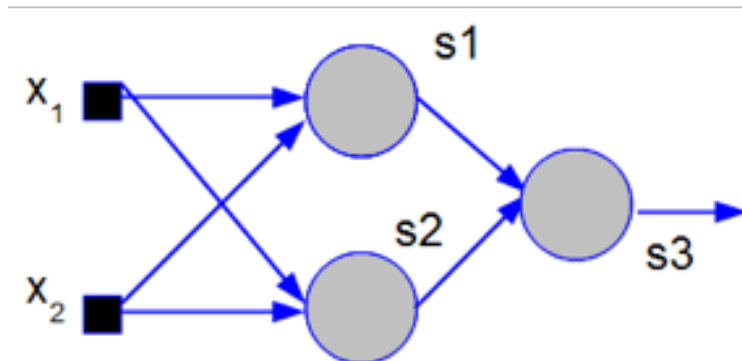
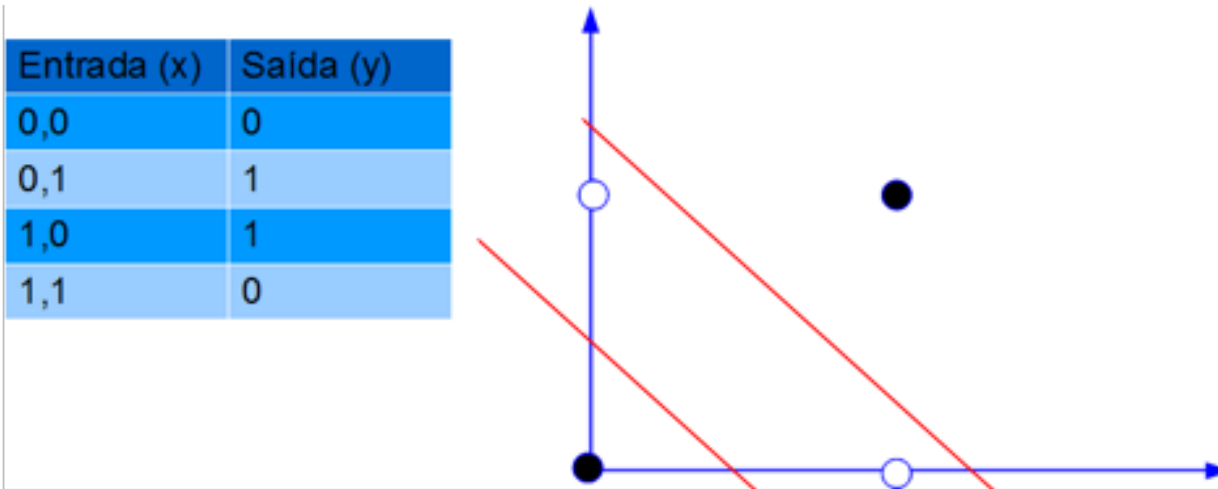
x_1 ■

x_2 ■

Impacto dos neurônios

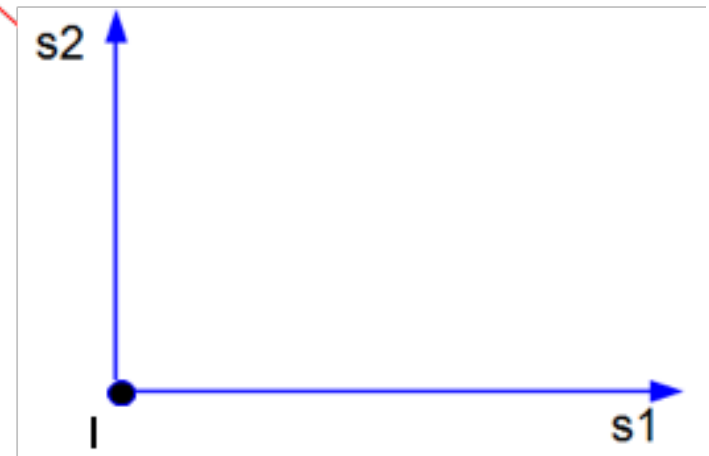
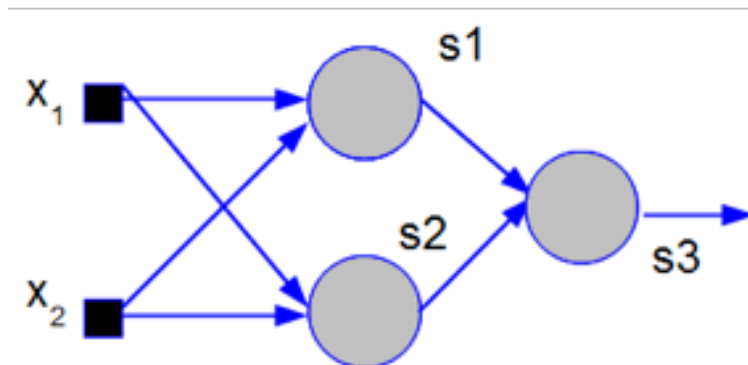
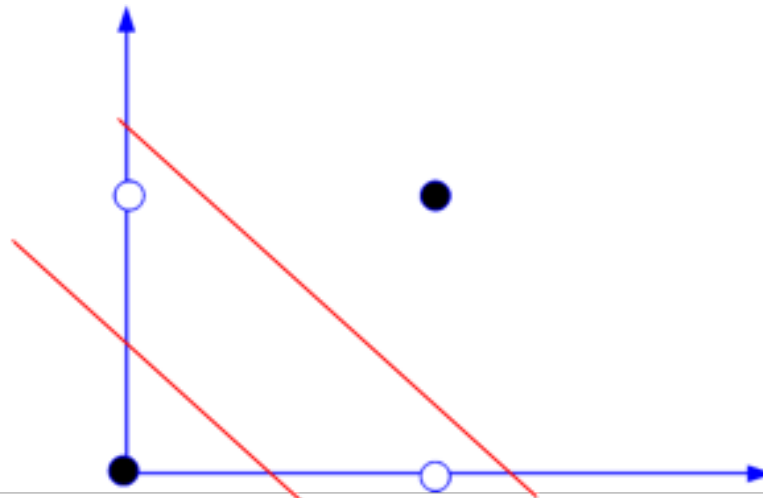


Impacto dos neurônios



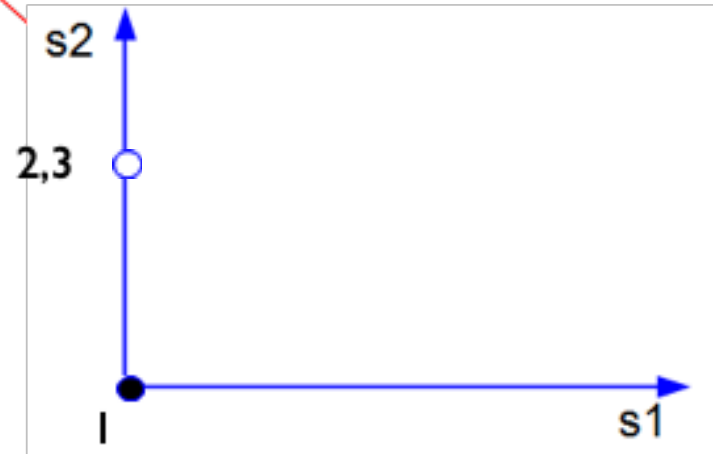
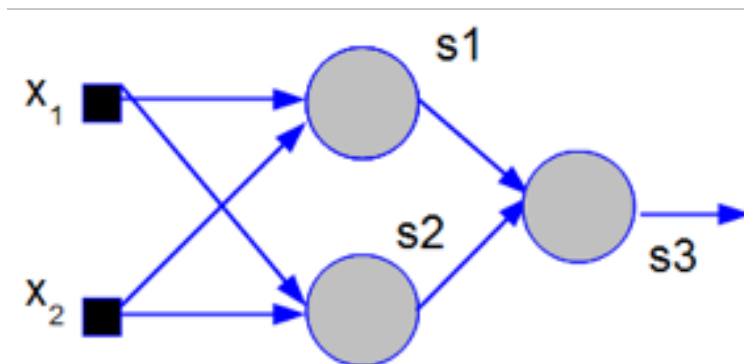
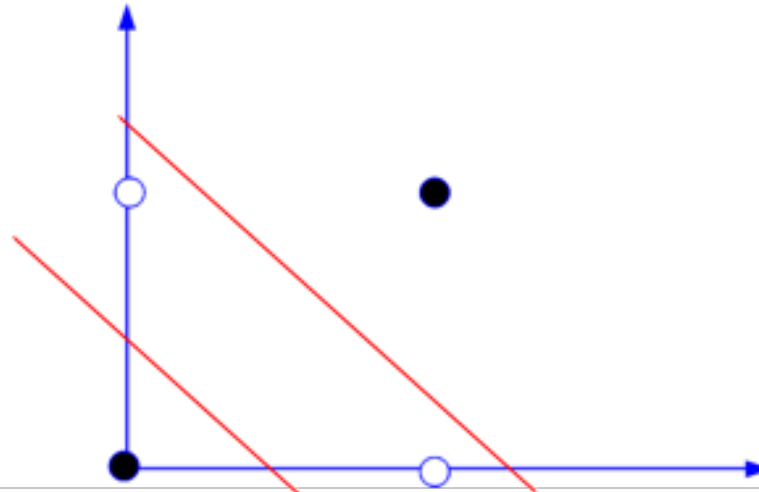
Impacto dos neurônios

Entrada (x)	Saída (y)
0,0	0
0,1	1
1,0	1
1,1	0

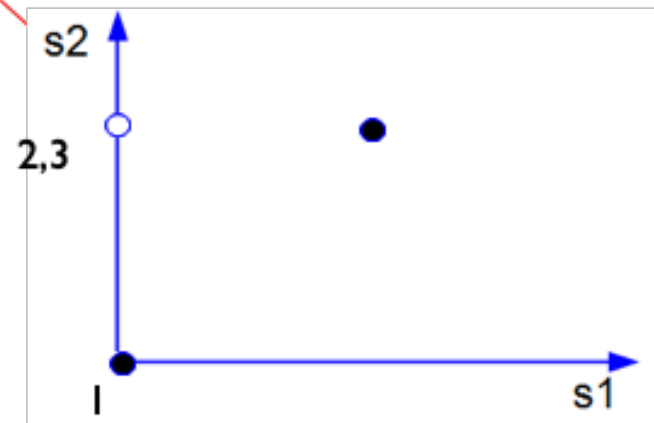
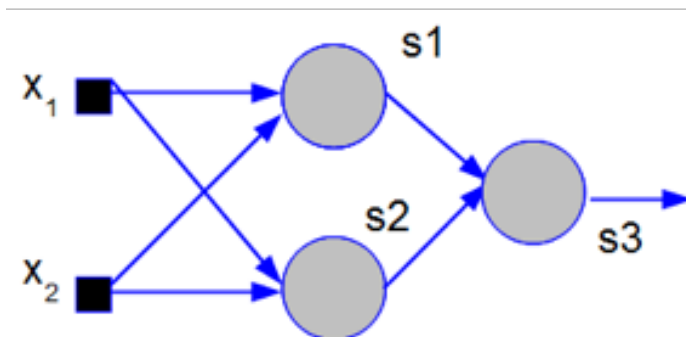
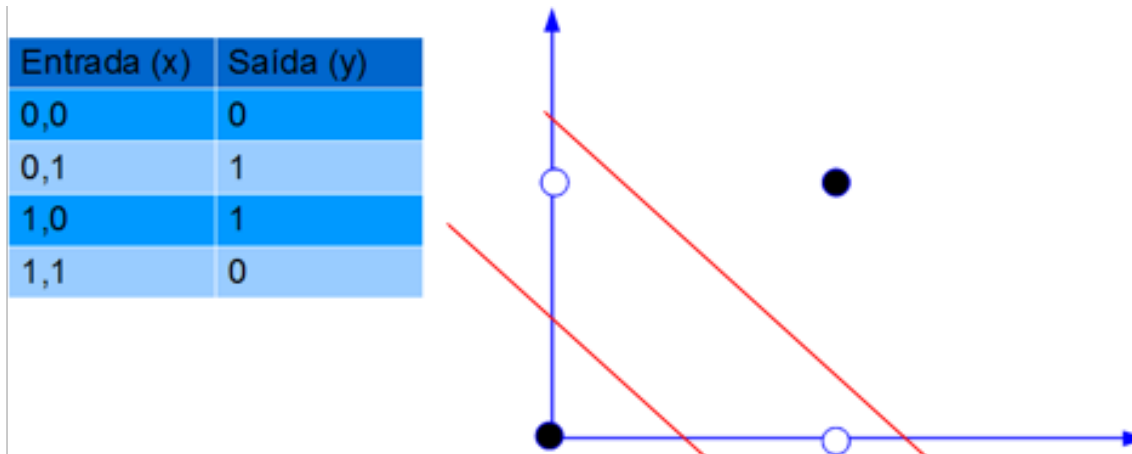


Impacto dos neurônios

Entrada (x)	Salda (y)
0,0	0
0,1	1
1,0	1
1,1	0

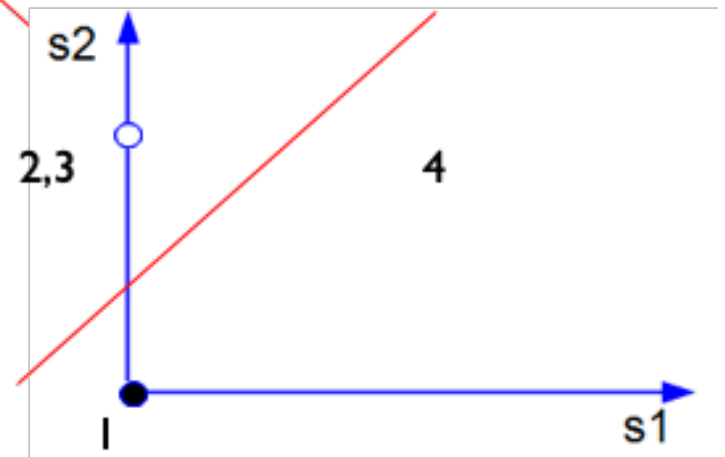
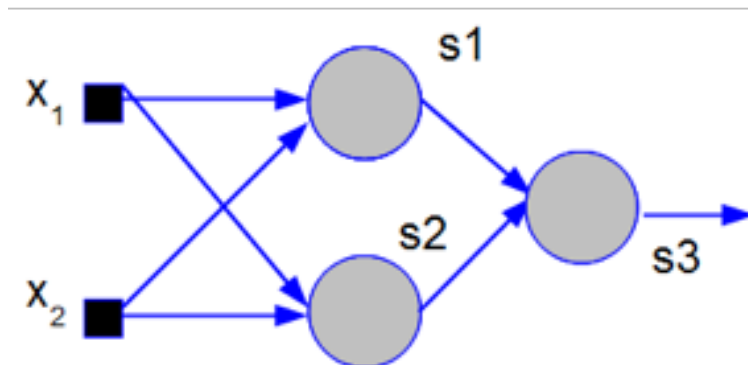
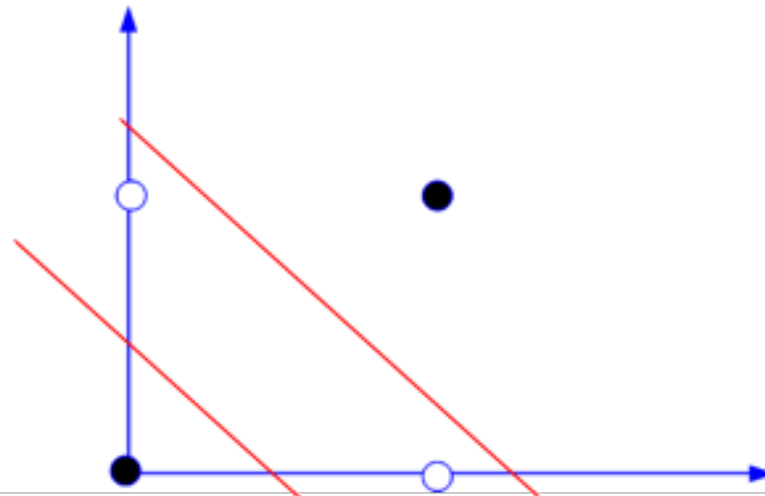


Impacto dos neurônios



Impacto dos neurônios

Entrada (x)	Saída (y)
0,0	0
0,1	1
1,0	1
1,1	0



TensorFlow

Google Brain

Biblioteca para computação numérica

CPUs, GPUs e dispositivos móveis

Desenvolvido para Aprendizado de Máquina e Deep Learning (e não somente isso)



TensorFlow

TensorFlow



Redes Convolucionais

Imagens são consideradas matrizes

Alta dimensionalidade

Grande número de pesos



Convolução

INPUT IMAGE

18	54	51	239	244	188
55	121	75	78	95	88
35	24	204	113	109	221
3	154	104	235	25	130
15	253	225	159	78	233
68	85	180	214	245	0

WEIGHT

1	0	1
0	1	0
1	0	1

429

Convolução

Input Volume (+pad 1) (7x7x3)

$x[:, :, 0]$

0	0	0	0	0	0	0
0	0	1	1	0	2	0
0	1	1	2	2	0	0
0	1	1	2	0	2	0
0	2	0	1	0	0	0
0	2	0	2	0	2	0
0	0	0	0	0	0	0

$x[:, :, 1]$

0	0	0	0	0	0	0
0	1	1	2	2	1	0
0	2	2	0	1	1	0
0	1	2	0	0	2	0
0	0	0	1	1	2	0
0	0	1	2	0	1	0
0	0	0	0	0	0	0

$x[:, :, 2]$

0	0	0	0	0	0	0
0	1	0	2	1	1	0
0	2	2	1	2	1	0
0	1	1	2	1	0	0
0	2	1	2	1	1	0
0	2	2	2	2	0	0
0	0	0	0	0	0	0

Filter W0 (3x3x3)

$w0[:, :, 0]$

-1	0	1
1	1	1
1	0	0

$w0[:, :, 1]$

1	1	-1
1	0	1
1	0	0

$w0[:, :, 2]$

0	-1	-1
-1	0	1
1	-1	0

Bias b0 (1x1x1)

$b0[:, :, 0]$

1

Filter W1 (3x3x3)

$w1[:, :, 0]$

-1	-1	-1
0	1	-1
0	0	0

$w1[:, :, 1]$

0	0	-1
-1	1	1
0	0	1

$w1[:, :, 2]$

0	-1	0
1	-1	-1
1	-1	-1

Bias b1 (1x1x1)

$b1[:, :, 0]$

0

Output Volume (3x3x2)

$o[:, :, 0]$



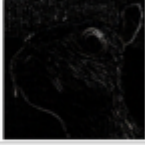


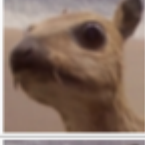
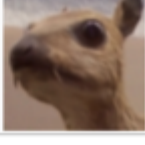
1	11	8
1	4	2
3	1	3

$o[:, :, 1]$

-2	1	2
-8	-10	2
-5	-3	4

toggle movement

Filtros

Operation	Filter	Convolved Image
Identity	$\begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$	
Edge detection	$\begin{bmatrix} 1 & 0 & -1 \\ 0 & 0 & 0 \\ -1 & 0 & 1 \end{bmatrix}$	
	$\begin{bmatrix} 0 & 1 & 0 \\ 1 & -4 & 1 \\ 0 & 1 & 0 \end{bmatrix}$	
	$\begin{bmatrix} -1 & -1 & -1 \\ -1 & 8 & -1 \\ -1 & -1 & -1 \end{bmatrix}$	
Sharpen	$\begin{bmatrix} 0 & -1 & 0 \\ -1 & 5 & -1 \\ 0 & -1 & 0 \end{bmatrix}$	
Box blur (normalized)	$\frac{1}{9} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$	
Gaussian blur (approximation)	$\frac{1}{16} \begin{bmatrix} 1 & 2 & 1 \\ 2 & 4 & 2 \\ 1 & 2 & 1 \end{bmatrix}$	

MaxPooling

Single depth slice

1	1	2	4
5	6	7	8
3	2	1	0
1	2	3	4

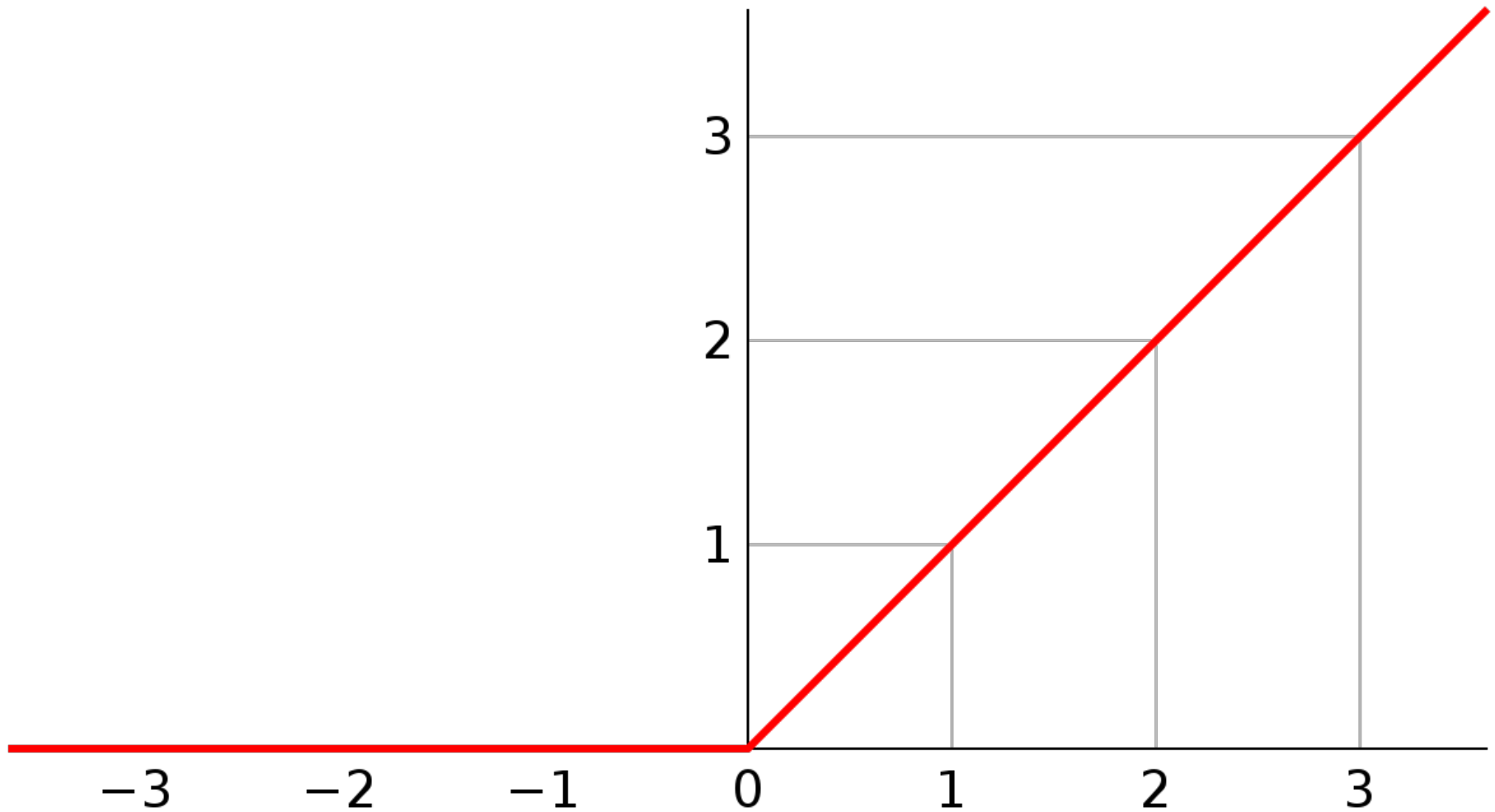
Max pool with 2x2 filters
and stride 2



6	8
3	4



Rectified Linear Unit - ReLU



Rectified Linear Unit - ReLU



Rectified Linear Unit - ReLU



Rede Neural Convucional - CNN

