One hidden layer is enough to represent (not learn) an approximation of any function to an arbitrary degree of accuracy.

# XOR, perceptron, and Multilayer Neural Networks

Why can't a perceptron model handle XOR, but multi-layer neural networks can?

ENLP 2025 Spring Xiulin Yang

#### XOR

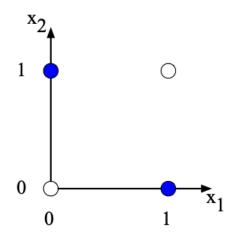
XOR (exclusive OR) returns true if and only if one of the two conditions is true, but not both.

OR				XOR			
x1	x2	у	_	x1	x2	у	
0	0	0	_	0	0	0	
	1	1		0	1	1	
1	0	1		1	0	1	
1	1	1		1	1	0	

#### **XOR**

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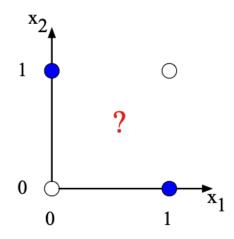
	AND			OR			XOR	
<b>x</b> 1	x2	у	<b>x</b> 1	x2	у	<b>x</b> 1	_x2	у
0	0	0		0			0	
0	1	0	0	1	1	0	1	1
1	0	0		0			0	
1	1	1	1	1	1	1	1	0



#### **XOR**

XOR (exclusive OR) returns true if and only if one of the two conditions is true, but not both.

AND				OR			XOR		
X	1 x2	у	x1	x2	у		<b>x</b> 1	x2	у
0	0	0		0			0	0	0
0	1	0	0	1	1			1	
1	0	0		0			1	0	1
1	1	1	1	1	1		1	1	0



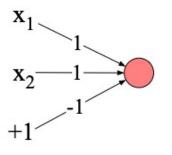
#### Why perceptron fails?

- A perceptron can handle and & or
- This is because a perceptron is essentially a *linear* classifier.

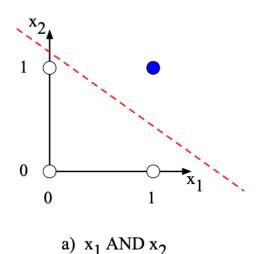
AND			
<b>x</b> 1	x2	у	y = a(X1*w1+x2*w2+b)
0	0	0	0*1+0*1-1 =-1 →0
0	0 1	0	0*0+1*1-1 =0 <b>→</b> 0
1	0	0	1*1+0*1-1 =0 →0
1	1	1	1*1+1*1-1 =1 →1

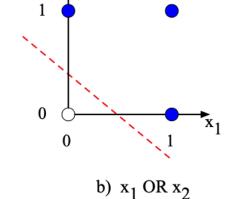
#### perceptron

step function



$$y = \begin{cases} 0, & \text{if } \mathbf{w} \cdot \mathbf{x} + b \le 0 \\ 1, & \text{if } \mathbf{w} \cdot \mathbf{x} + b > 0 \end{cases}$$





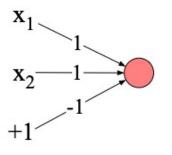
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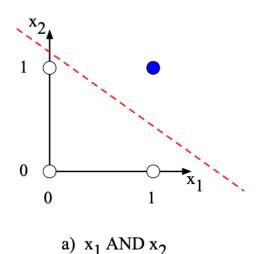
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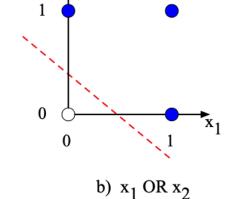
#### perceptron

step function



$$y = \begin{cases} 0, & \text{if } \mathbf{w} \cdot \mathbf{x} + b \le 0 \\ 1, & \text{if } \mathbf{w} \cdot \mathbf{x} + b > 0 \end{cases}$$





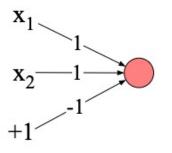
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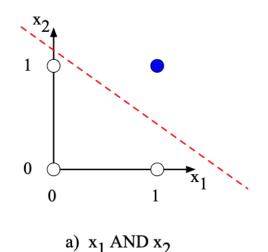
AND			
	x2	_	y = a(X1*w1+x2*w2+b)
0	0 1 0 1	0	0*1+0*1-1 =-1 <b>→</b> 0
0	1	0	0*0+1*1-1 =0 <b>→</b> 0
1	0	0	1*1+0*1-1 =0 →0
1	1	1	1*1+1*1-1 =1 →1

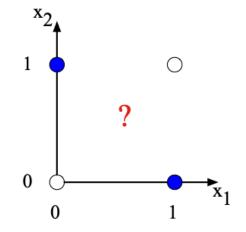
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step function



$$y = \begin{cases} 0, & \text{if } \mathbf{w} \cdot \mathbf{x} + b \le 0 \\ 1, & \text{if } \mathbf{w} \cdot \mathbf{x} + b > 0 \end{cases}$$

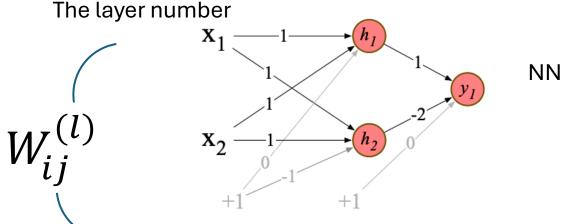




problem?

	XOR					
x1	x2	у				
0	0	0				
0	1	1				
1	0	1				
1	1	0				

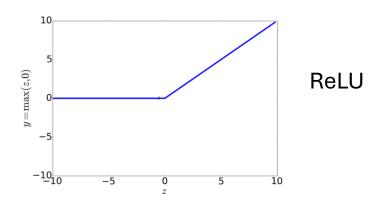
$$\begin{aligned} &\text{h1 = ReLU}(\mathbf{x}_1 * W_{11}^{(1)} + \mathbf{x}_2 \, W_{21}^{(1)} + b_1^{(1)}) \\ &\text{h2 = ReLU}(\mathbf{x}_1 * W_{12}^{(1)} + \mathbf{x}_2 \, W_{22}^{(1)} + b_2^{(1)}) \\ &\text{y = ReLU}(\mathbf{h}_1 * W_{21}^{(2)} + \mathbf{h}_2 \, W_{22}^{(2)} + b_1^{(2)}) \end{aligned}$$



i: the neuron index from the previous layer; j the neuron index of the current layer

Rectified linear unit, also called the ReLU

$$y = ReLU(z) = max(z, 0)$$

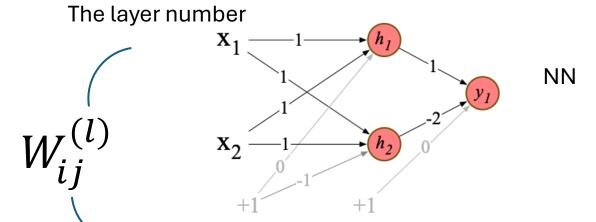


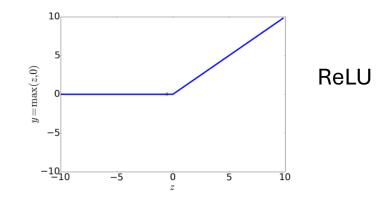
problem?

2	XOR						
x1	x2	у					
0	0	0					
0	1	1					
1	0	1					
1	1	0					

h1 = ReLU(
$$x_1*W_{11}^{(1)} + x_2 W_{21}^{(1)} + b_1^{(1)}$$
)  
h2 = ReLU( $x_1*W_{12}^{(1)} + x_2 W_{22}^{(1)} + b_2^{(1)}$ )  
y = ReLU( $h_1*W_{21}^{(2)} + h_2 W_{22}^{(2)} + b_1^{(2)}$ )  
h1 = ReLU(0\*1+0\*1+0)=0  
h2 = ReLU(0\*1+0\*1-1)=0

y = ReLU(0\*1+0\*(-2)+0) = 0





problem?

XOR

0

0

x1 x2

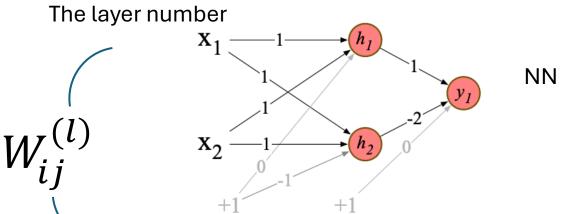
h1 = ReLU(
$$x_1*W_{11}^{(1)} + x_2 W_{21}^{(1)} + b_1^{(1)}$$
)

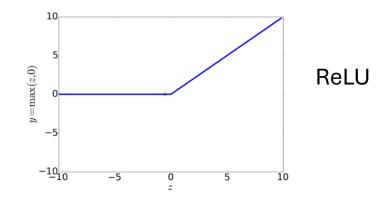
h2 = ReLU(
$$x_1*W_{12}^{(1)}+x_2W_{22}^{(1)}+b_2^{(1)}$$
)

y = ReLU(
$$h_1*W_{21}^{(2)}+h_2W_{22}^{(2)}+b_1^{(2)}$$
)

$$y = ReLU(0*1+0*(-2)+0) = 0$$

$$y = ReLU(1*1+0*(-2)+0) = 1$$





problem?

XOR

0

0

x1 x2

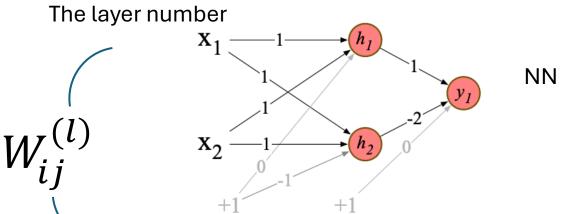
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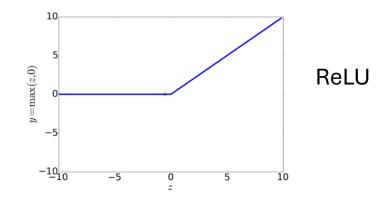
h2 = ReLU(
$$x_1*W_{12}^{(1)}+x_2W_{22}^{(1)}+b_2^{(1)}$$
)

y = ReLU(
$$h_1*W_{21}^{(2)}+h_2W_{22}^{(2)}+b_1^{(2)}$$
)

$$y = ReLU(0*1+0*(-2)+0) = 0$$

$$y = ReLU(1*1+0*(-2)+0) = 1$$

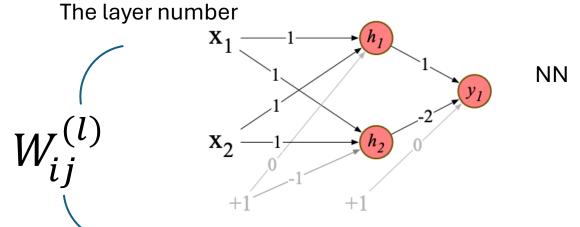


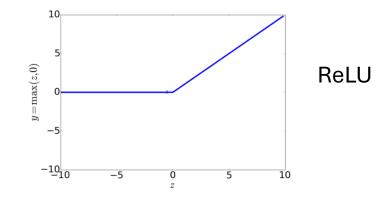


problem?

h1 = ReLU(
$$x_1*W_{11}^{(1)} + x_2 W_{21}^{(1)} + b_1^{(1)}$$
)  
h2 = ReLU( $x_1*W_{12}^{(1)} + x_2 W_{22}^{(1)} + b_2^{(1)}$ )  
y = ReLU( $h_1*W_{21}^{(2)} + h_2 W_{22}^{(2)} + b_1^{(2)}$ )

y = ReLU(0\*1+0\*(-2)+0) = 0





problem?

XOR

0

0

0

x1 x2

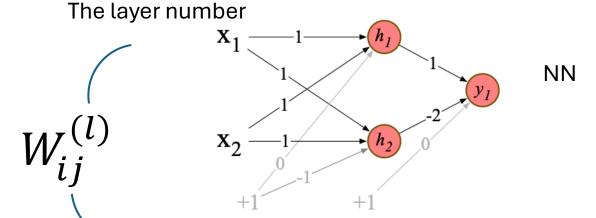
h1 = ReLU(
$$x_1*W_{11}^{(1)} + x_2 W_{21}^{(1)} + b_1^{(1)}$$
)  
h2 = ReLU( $x_1*W_{12}^{(1)} + x_2 W_{22}^{(1)} + b_2^{(1)}$ )

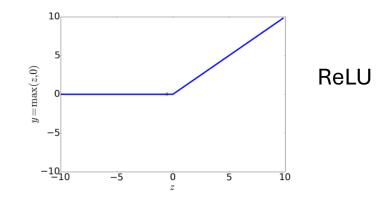
y = ReLU(
$$h_1*W_{21}^{(2)}+h_2W_{22}^{(2)}+b_1^{(2)}$$
)

$$y = ReLU(0*1+0*(-2)+0) = 0$$

$$y = ReLU(1*1+0*(-2)+0) = 1$$

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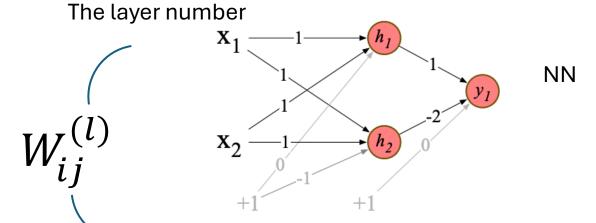
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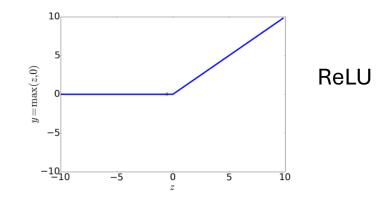
			h2 = ReLU(
2	XOR		y = ReLU(h
<b>x</b> 1	x2	у	,
0	0	0	y = ReLU(0
0	1	1	y = ReLU(1
1	0	1	y = ReLU(1
1	1	•	-

$$\begin{aligned} &\text{h1} = \text{ReLU}(\mathbf{x}_1 * W_{11}^{(1)} + \mathbf{x}_2 \, W_{21}^{(1)} + b_1^{(1)}) \\ &\text{h2} = \text{ReLU}(\mathbf{x}_1 * W_{12}^{(1)} + \mathbf{x}_2 \, W_{22}^{(1)} + b_2^{(1)}) \\ &\text{y} = \text{ReLU}(\mathbf{h}_1 * W_{21}^{(2)} + \mathbf{h}_2 \, W_{22}^{(2)} + b_1^{(2)}) \end{aligned}$$

$$y = ReLU(0*1+0*(-2)+0) = 0$$
  
 $y = ReLU(1*1+0*(-2)+0) = 1$ 

$$y = ReLU(1*1+0*(-2)+0) = 1$$





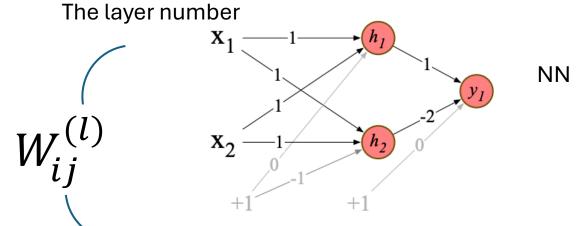
problem?

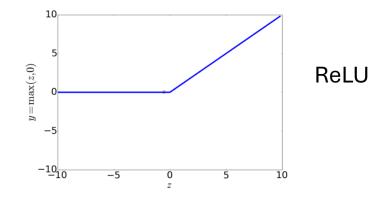
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)  
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y = ReLU( $h_1*W_{21}^{(2)} + h_2 W_{22}^{(2)} + b_1^{(2)}$ )  
y = ReLU( $0*1+0*(-2)+0$ ) = 0

XOR

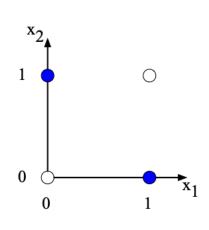
$$y = ReLU(1*1+0*(-2)+0) = 1$$
  
 $y = ReLU(2*1+1*(-2)+0) = 0$ 

y = ReLU(1\*1+0\*(-2)+0) = 1

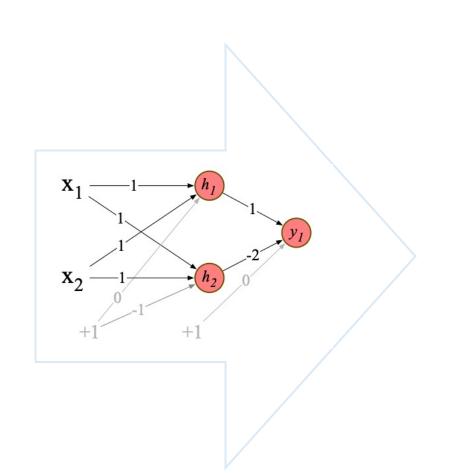


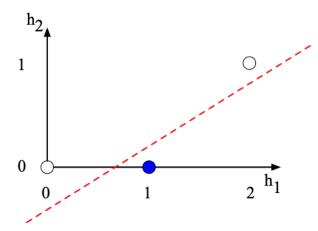


# Why can a multilayer NNs solve the XOR problem?



a) The original x space





b) The new (linearly separable) h space

Code

https://colab.research.google.com/drive/1Dkr6GSFfD6klrdpkH5bw0YeO5H9FzeAT?usp=sharing