

# **Introduction to Machine Learning**

Chapter 0: Deep Learning- XOR Problem

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Suppose we have four data points

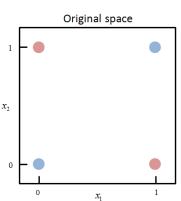
$$X = \{(0,0)^{\top}, (0,1)^{\top}, (1,0)^{\top}, (1,1)^{\top}\}$$

 The XOR gate (exclusive or) returns true, when an odd number of inputs are true:

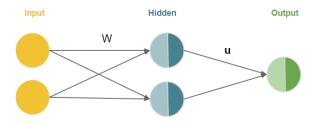
<i>X</i> <sub>1</sub>	<i>X</i> <sub>2</sub>	XOR = y
0	0	0
0	1	1
1	0	1
1	1	0

• Can you learn the target function with a logistic regression model?

- Logistic regression can not solve this problem. In fact, any model using simple hyperplanes for separation can not (including a single neuron).
- A small neural net can easily solve the problem by transforming the space!



Consider the following model:



**Figure:** A neural network with two neurons in the hidden layer. The matrix  $\mathbf{W}$  describes the mapping from  $\mathbf{x}$  to  $\mathbf{z}$ . The vector  $\mathbf{u}$  from  $\mathbf{z}$  to  $\mathbf{y}$ .

• Let use ReLU  $\sigma(z) = \max\{0,z\}$  as activation function and a simple thresholding function  $\tau(z) = [z>0] = \begin{cases} 1 & \text{if } z>0 \\ 0 & \text{otherwise} \end{cases}$  as output transformation function. We can represent the architecture of the model by the following equation:

$$f(\mathbf{x} \mid \boldsymbol{\theta}) = f(\mathbf{x} \mid \mathbf{W}, \mathbf{b}, \mathbf{u}, c) = \tau \left( \mathbf{u}^{\top} \sigma(\mathbf{W}^{\top} \mathbf{x} + \mathbf{b}) + c \right)$$
$$= \tau \left( \mathbf{u}^{\top} \max\{0, \mathbf{W}^{\top} \mathbf{x} + \mathbf{b}\} + c \right)$$

- So how many parameters does our model have?
  - In a fully connected neural net, the number of connections between the nodes equals our parameters:

$$\underbrace{(2\times2)}_{W} + \underbrace{(2\times1)}_{b} + \underbrace{(2\times1)}_{u} + \underbrace{(1)}_{c} = 9$$

Let 
$$\mathbf{W} = \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix}$$
,  $\mathbf{b} = \begin{pmatrix} 0 \\ -1 \end{pmatrix}$ ,  $\mathbf{u} = \begin{pmatrix} 1 \\ -2 \end{pmatrix}$ ,  $c = -0.5$ 

$$\mathbf{X} = \begin{pmatrix} 0 & 0 \\ 0 & 1 \\ 1 & 0 \\ 1 & 1 \end{pmatrix}, \ \mathbf{XW} = \begin{pmatrix} 0 & 0 \\ 1 & 1 \\ 1 & 1 \\ 2 & 2 \end{pmatrix}, \ \mathbf{XW} + \mathbf{B} = \begin{pmatrix} 0 & -1 \\ 1 & 0 \\ 1 & 0 \\ 2 & 1 \end{pmatrix}$$

Note: **X** is a  $(n \times p)$  design matrix in which the *rows* correspond to the data points. **W**, as usual, is a  $(p \times m)$  matrix where each *column* corresponds to a single (hidden) neuron. **B** is a  $(n \times m)$  matrix with **b** duplicated along the rows.

$$X = \begin{pmatrix} 0 & 0 \\ 0 & 1 \\ 1 & 0 \\ 1 & 1 \end{pmatrix} \qquad W = \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix}$$

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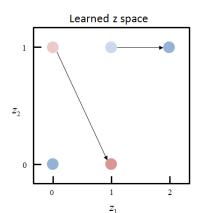
$$\mathbf{X} = \begin{pmatrix} 0 & 0 \\ 0 & 1 \\ 1 & 0 \\ 1 & 1 \end{pmatrix}, \ \mathbf{XW} = \begin{pmatrix} 0 & 0 \\ 1 & 1 \\ 1 & 1 \\ 2 & 2 \end{pmatrix}, \ \mathbf{XW} + \mathbf{B} = \begin{pmatrix} 0 & -1 \\ 1 & 0 \\ 1 & 0 \\ 2 & 1 \end{pmatrix}$$

$$Z = \max\{0, XW + B\} = \begin{pmatrix} 0 & 0 \\ 1 & 0 \\ 1 & 0 \\ 2 & 1 \end{pmatrix}$$

• Note that we computed all examples at once.

 The input points are mapped into transformed space to

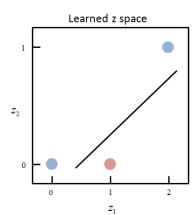
$$\mathbf{Z} = egin{pmatrix} 0 & 0 \ 1 & 0 \ 1 & 0 \ 2 & 1 \end{pmatrix}$$



 The input points are mapped into transformed space to

$$oldsymbol{Z} = egin{pmatrix} 0 & 0 \ 1 & 0 \ 1 & 0 \ 2 & 1 \end{pmatrix}$$

which is easily separable.



In a final step we have to multiply the activated values of matrix Z with the vector u and add the bias term c:

$$f(\mathbf{x} \mid \mathbf{W}, \mathbf{b}, \mathbf{u}, c) = \begin{pmatrix} 0 & 0 \\ 1 & 0 \\ 1 & 0 \\ 2 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ -2 \end{pmatrix} + \begin{pmatrix} -0.5 \\ -0.5 \\ -0.5 \end{pmatrix} = \begin{pmatrix} -0.5 \\ 0.5 \\ 0.5 \\ -0.5 \end{pmatrix}$$

• And then apply the step function  $\tau(z) = [z > 0]$ . This solves the XOR problem perfectly!

<i>X</i> <sub>1</sub>	<i>X</i> <sub>2</sub>	XOR = y
0	0	0
0	1	1
1	0	1
1	1	0