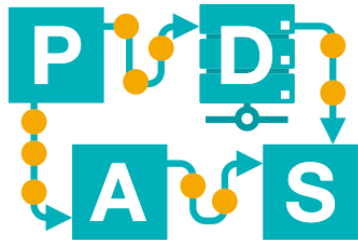


# Neural Network

## Lecture 7 instruction

# IDS-I-L7

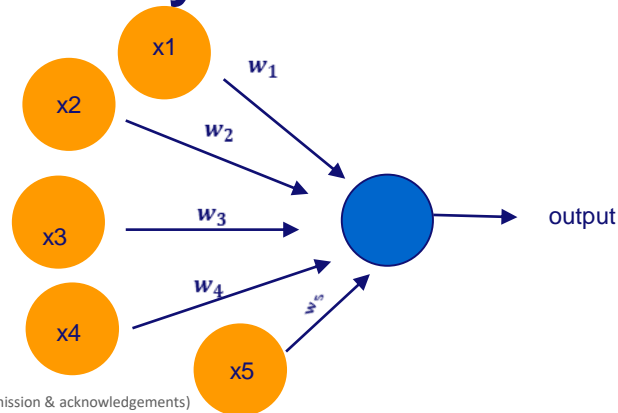


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# Exercise.1

- Imagine a simple single neuron
    - If it has 5 input and 1 output
    - Each input can have value of 0 or 1
- a. Calculate the number of different input pattern ?
- b. Can you formulize the pattern?



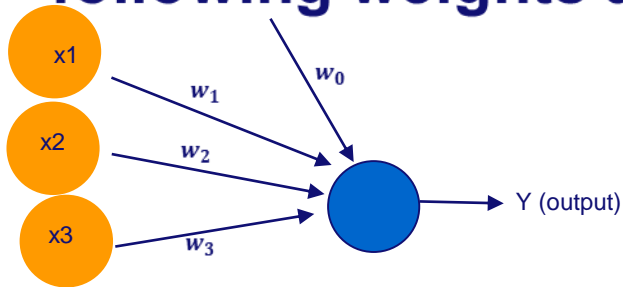
# Solution.1

- As you can see the input has binary pattern, so the pattern of input is:  $2^n$  where  $n$  is the number of inputs

|    | 1 | 2 | 3 | 4 | 5 | 6 | ... | 16 |
|----|---|---|---|---|---|---|-----|----|
| X1 | 0 | 1 | 0 | 0 | 0 | 1 | ... | 1  |
| X2 | 0 | 0 | 1 | 0 | 0 | 1 | ... | 1  |
| X3 | 0 | 0 | 0 | 1 | 0 | 0 | ... | 1  |
| x4 | 0 | 0 | 0 | 0 | 1 | 0 | ... | 1  |

# Exercise.2

- For the previous neuron, consider three input, following weights and activation function, for simplification ignore  $w_0$ :



$$\begin{aligned} w_1 &= 2 \\ w_2 &= -4 \\ w_3 &= 1 \end{aligned}$$

$$f(x) = \begin{cases} 0, & x < 0 \\ 1, & x \geq 0 \end{cases}$$

- Calculate the output of neurons:

| pattern | $p_1$ | $p_2$ | $p_3$ | $p_4$ |
|---------|-------|-------|-------|-------|
| $x_1$   | 1     | 0     | 1     | 1     |
| $x_2$   | 0     | 1     | 0     | 1     |
| $x_3$   | 0     | 1     | 1     | 1     |

# Solution.2

- Use this formula for sum of the inputs:

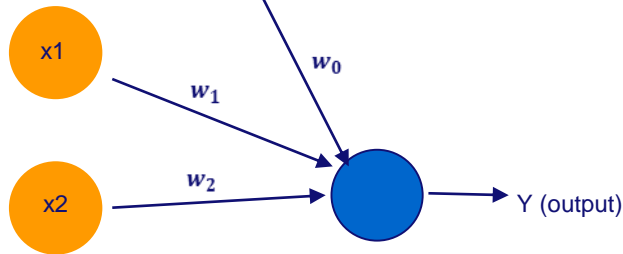
$$a = \sum_i w_i x_i = w_1 \cdot x_1 + w_2 \cdot x_2 + w_3 \cdot x_3$$

- Apply the activation function to result:
- Answer is:

- $p_1: a = 2 * 1 - 4 * 0 + 1 * 0 = 2, f(2) = 1$
- $p_2: a = 2 * 0 - 4 * 1 + 1 * 1 = -3, f(-3) = 0$
- $p_3: a = 2 * 1 - 4 * 0 + 1 * 1 = 3, f(3) = 1$
- $p_4: a = 2 * 1 - 4 * 1 + 1 * 1 = -1, f(-1) = 0$

# Exercise.3a

- The main building blocks of any computational device are logical operators.
- You have already seen representation of operator AND and OR in the lecture by a single neuron:
  - With the new activation function how AND works?
  - Imagine two weights are 1

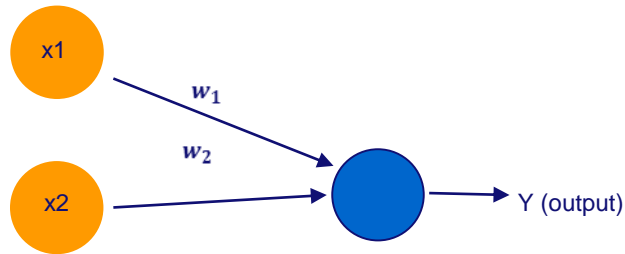


$$f(x) = \begin{cases} 0, & x < 2 \\ 1, & x \geq 2 \end{cases}$$

# Solution.3a

- $p_1: a = 1 * 0 + 1 * 0 = 0, f(0) = 0$
- $p_2: a = 1 * 1 + 1 * 0 = 1, f(1) = 0$
- $p_3: a = 1 * 0 + 1 * 1 = 1, f(1) = 0$
- $p_4: a = 1 * 1 + 1 * 1 = 2, f(2) = 1$

$$f(x) = \begin{cases} 0, & x < 2 \\ 1, & x \geq 2 \end{cases}$$



# Exercise.3b

- **How to change the neuron parameter to have logical OR function?**
  - Which parameters can be changed?
  - How to change them?
  - What are the possible solutions?



# Solution.3b

- **Activation function and weights can be changed**

- **First solution:**

- $p_1: a = 2 * 0 + 2 * 0 = 0, f(0) = 0$
- $p_2: a = 2 * 1 + 2 * 0 = 2, f(1) = 1$
- $p_3: a = 2 * 0 + 2 * 1 = 2, f(1) = 1$
- $p_4: a = 2 * 1 + 2 * 1 = 4, f(2) = 1$

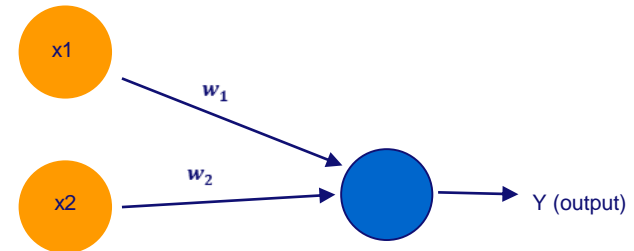
$$f(x) = \begin{cases} 0, & x < 2 \\ 1, & x \geq 2 \end{cases}$$

- **Second solution:**

- **Changing activation function:**

$$f(x) = \begin{cases} 0, & x < 1 \\ 1, & x \geq 1 \end{cases}$$

- **Check the functionality!**



# Exercise.4

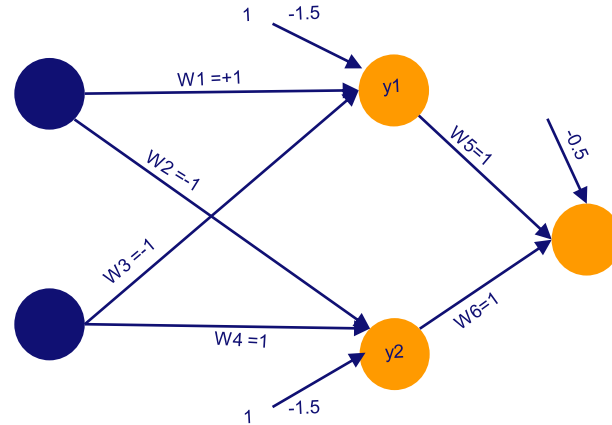
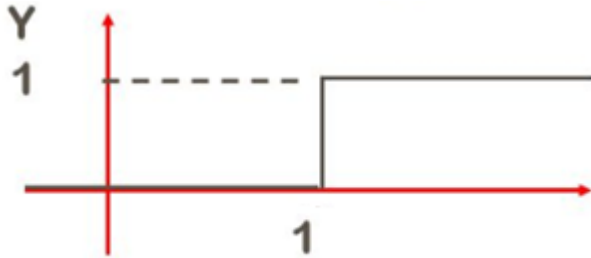


- You have already seen the logical XOR neural network in the lecture.
- XOR network can be created based on basic operators:
  - Such the one in the lecture
  - a) Try to create a neural network of XOR function(Not using first fact)
  - b)Take a way part! set different parameter for your network to act as XOR function

$$\text{XOR} = (x_1 \text{ OR } x_2) \text{ AND NOT}(x_1 \text{ AND } x_2)$$

# Solution.4

- $A \text{ xor } B = (A \wedge \neg B) \vee (\neg A \wedge B)$



# Exercise.5

- **Imagine a car renting company wants to deploy a new system for assessing worthiness of its customers.**
  - The new system is using feed forward neural network with supervised learning algorithms.
- **Suggest what should the company do before the system can be used?**

# Solution.5

- **First thing in order to have a supervised learning is some historical data about its customer and the one who already took cars in the past.**
  - **This data will be used as training set for neural network**
  - **The amount of data and the extensiveness of data is very important**
  - **The network is not be able to predict and works accurate for the customer that their similar situation were not in the historical data.**

# Exercise.6

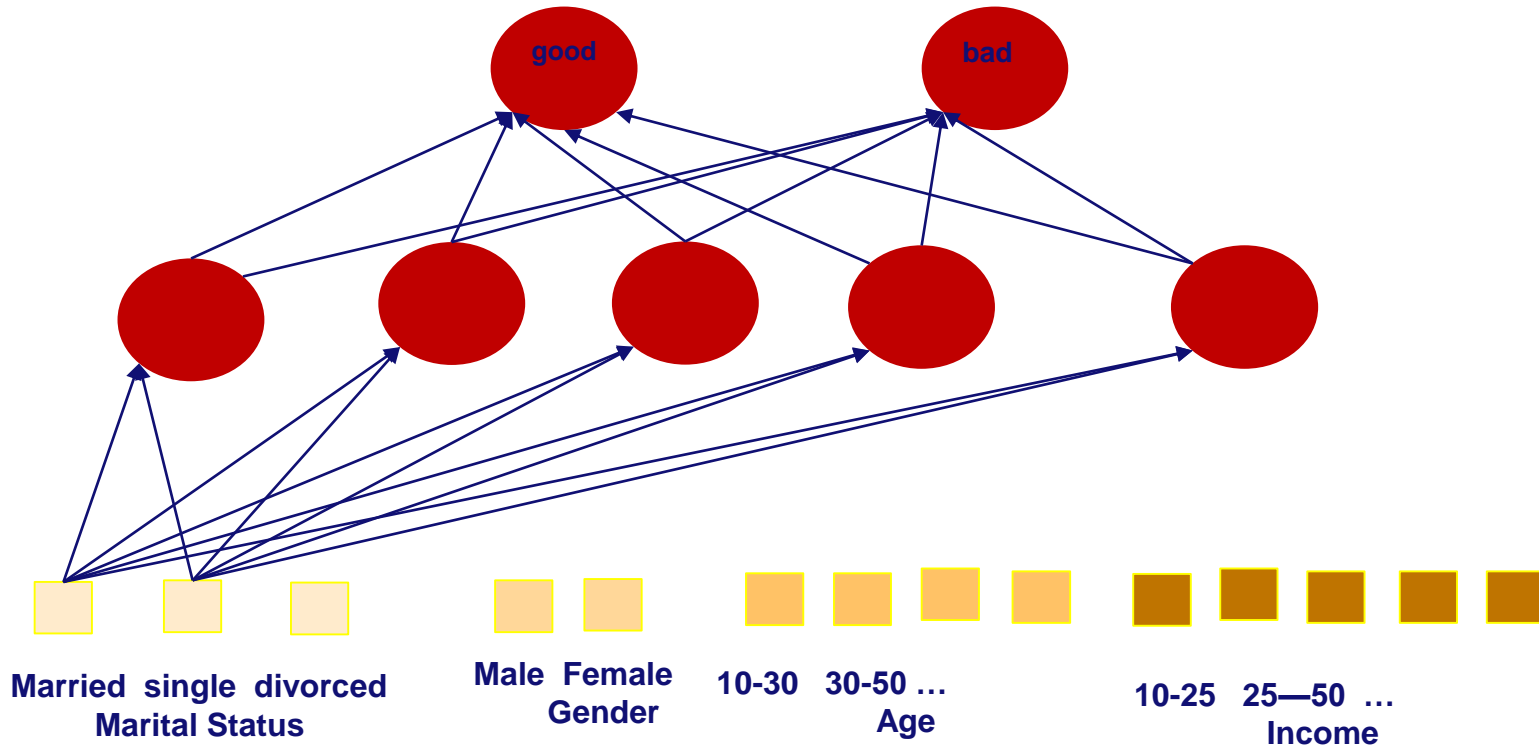
- Now imagine the previous car company
  - If we want to classify potential company customers as good customers or bad ones for renting high price car with special offers:
  - We have a training dataset describing past customers using the following attributes:
    - Age {[18..30[, [30..50[, [50..65[, [65+]}],
    - Marital status {married, single, divorced},
    - Gender {male, female},
    - Income {[10K..25K[, [25K..50K[, [50K..65K[, [65K..100K[, [100K+]}].
- Design a neural network in order to predict the good customers?



# Solution.6

- We have two output classes good and bad
  - Then two nodes in the output layer
  - We have 4 variables
  - And for each: {3,2,4,5} values
  - We would have 14 neurons in the input layer
  - In the hidden layer we probably can have less number than input
    - However there is no predefined answer

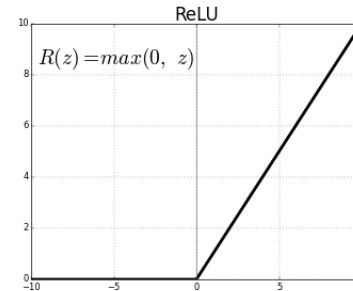
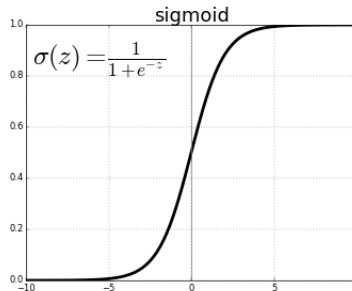
# Solution.6





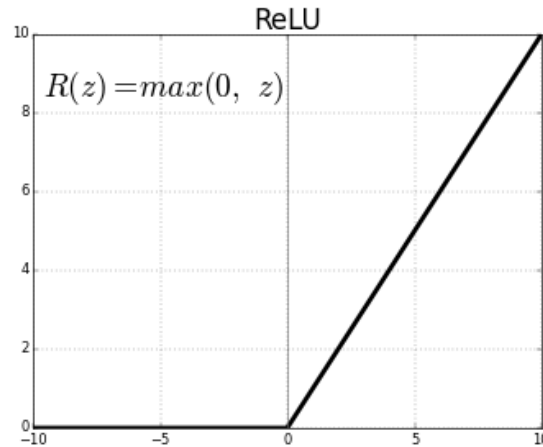
# Exercise.7

- Assume that we have a 2-layer net
  - one hidden layer
  - with weights  $W_0^1, w_1^1, W_0^2$  and  $w_1^2$ .
  - Assume that we use activation function named Relu in the hidden layer, and no activation on the output layer.
- Write down an equation for the output of the j'th node in the hidden layer.



# Solution.7

*output of  $j$ th neuron in the hidden layer =  $\max(W_1^1 * X + w_0^1, 0)$*



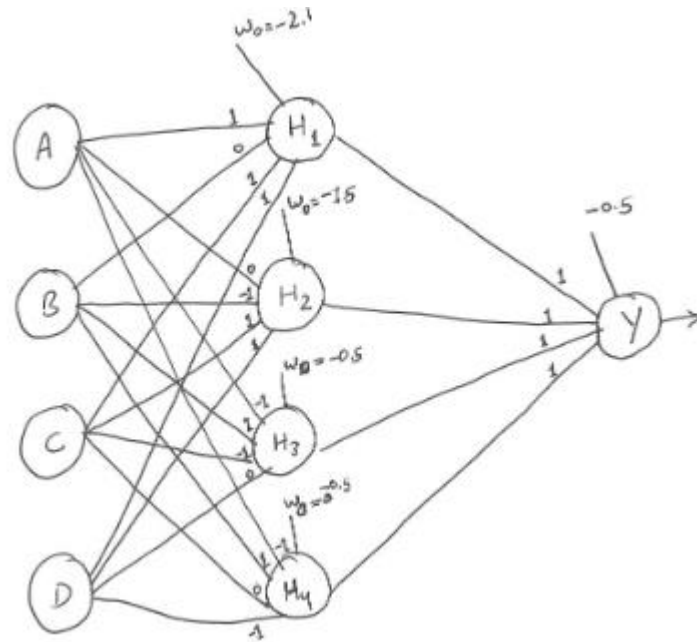
# Exercise.8

- Create a neural network with only one hidden layer (of any number of units) that implements  $(A \vee \neg B) \oplus (\neg C \vee \neg D)$ .
- Draw your network, and show all weights of each unit.

# Solution.8

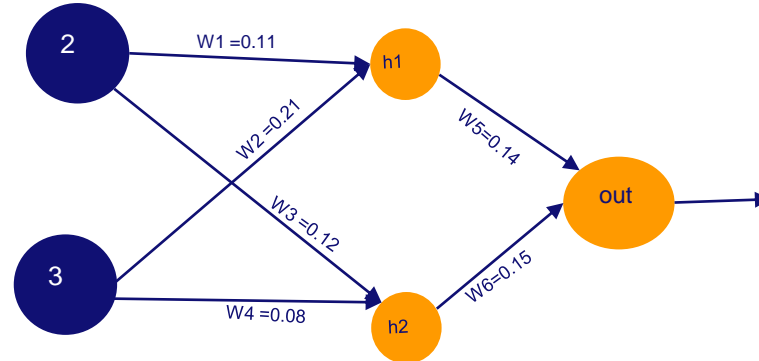
- XOR can be written in terms of AND and OR:
  - $p \text{ xor } q = (p \wedge \neg q) \vee (\neg p \wedge q)$
  - Given this, we can rewrite the formula as
  - $(A \wedge C \wedge D) \vee (\neg B \wedge C \wedge D) \vee (\neg A \wedge B \wedge \neg C) \vee (\neg A \wedge B \wedge \neg D)$
- This formula can be represented by a neural network with one hidden layer and
- four nodes in the hidden layer
  - one unit for each parenthesis

# Solution.8

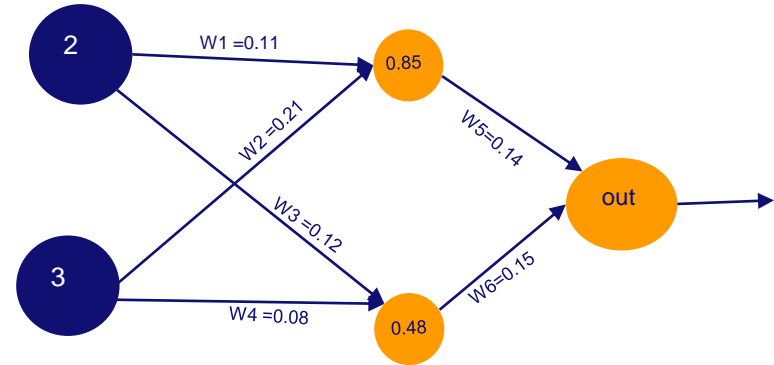


# Exercise.9

- There is a feedforward network with specified weights and inputs, calculate the value of the hidden layer neurons and the output layer neuron.



# Solution.9



$$H1 : 2 \times 0.11 + 3 \times 0.21 = 0.85$$

$$\text{Out: } 0.85 \times 0.14 + 0.48 \times 0.15 = 0.191$$

$$H2: 2 \times 0.12 + 3 \times 0.08 = 0.48$$