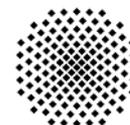


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# Neural Networks - I

**Thang Vu**  
**20.11.2025**



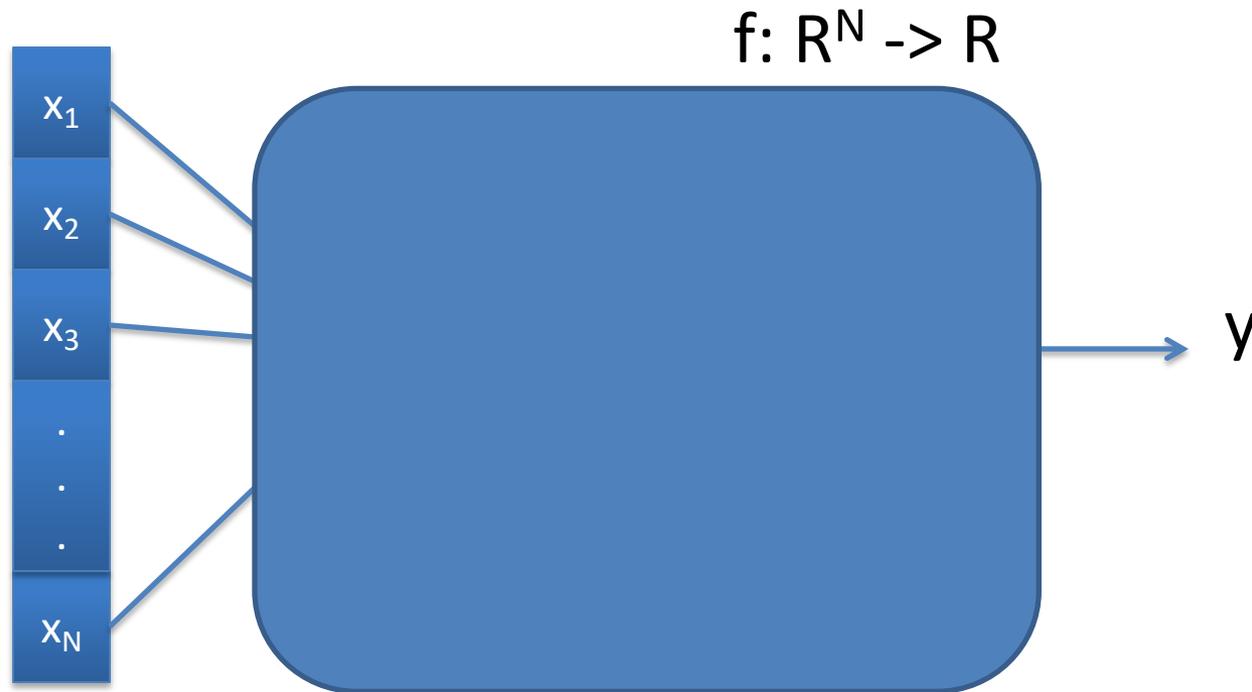
# Questionnaire

- Please go to Ilias
- Open the questionnaire ‚ML Basics‘
- Note that the questionnaire is anonymous, meaning we only receive the final statistics and responses, not the identities of the individuals who submitted them.

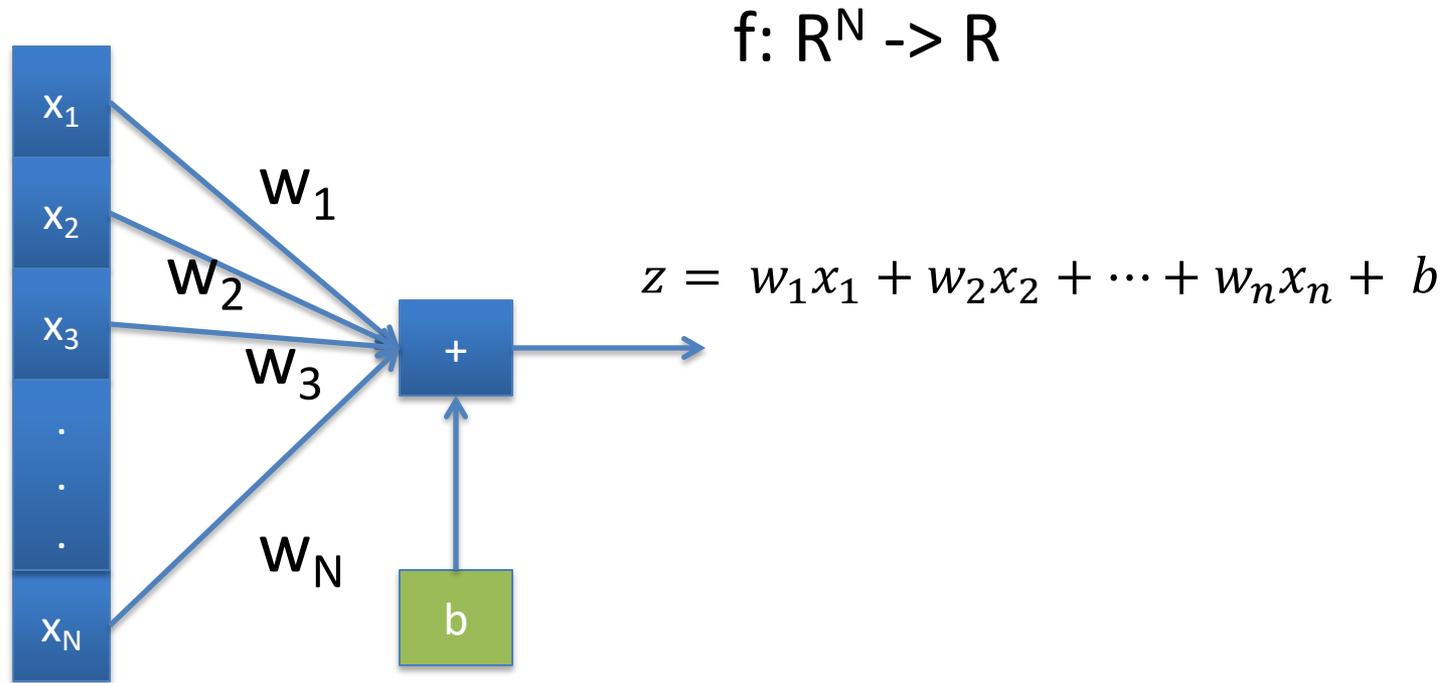
# Outline

- Neuron
- Neural network
- Computing the output
- Training the network
- Backpropagation

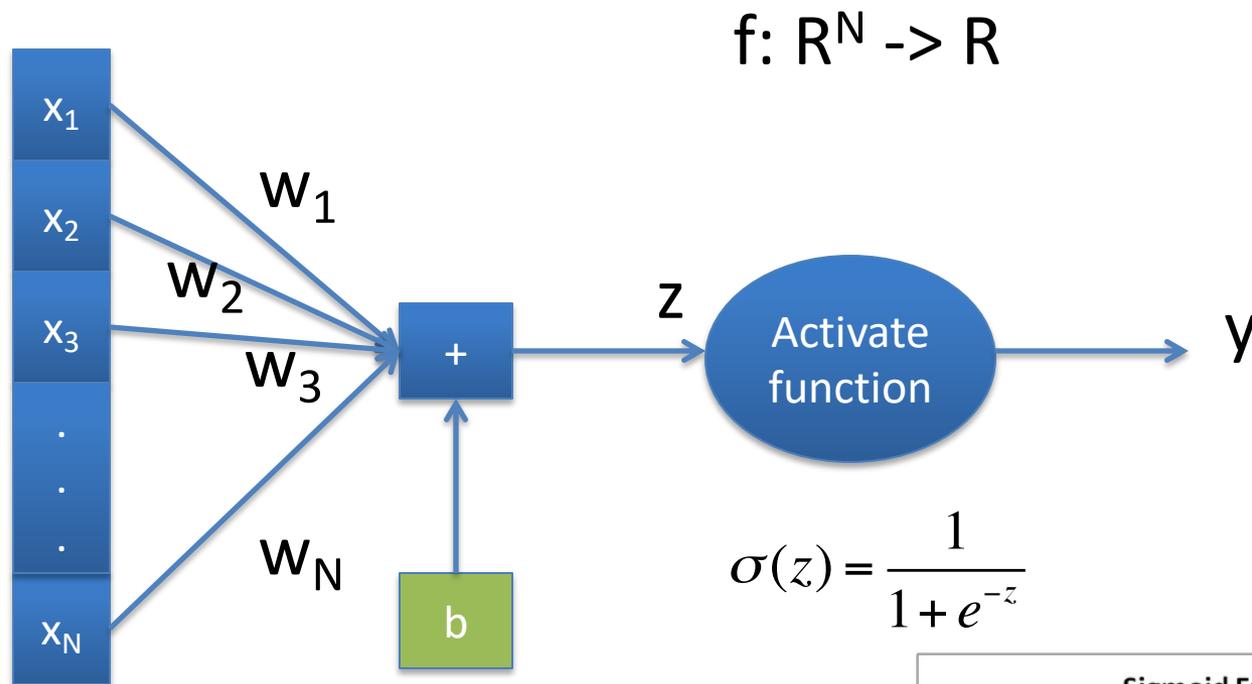
# Model: A single neuron



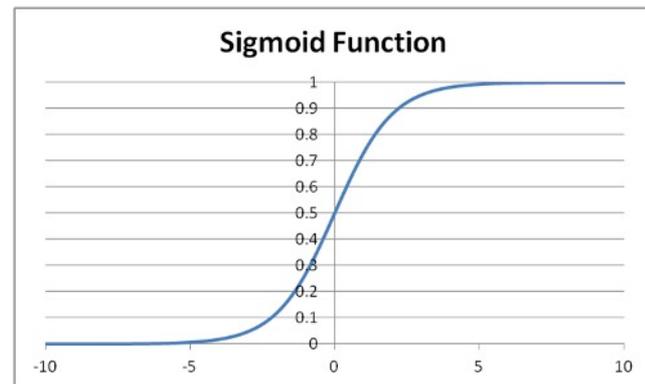
# Model: A single neuron



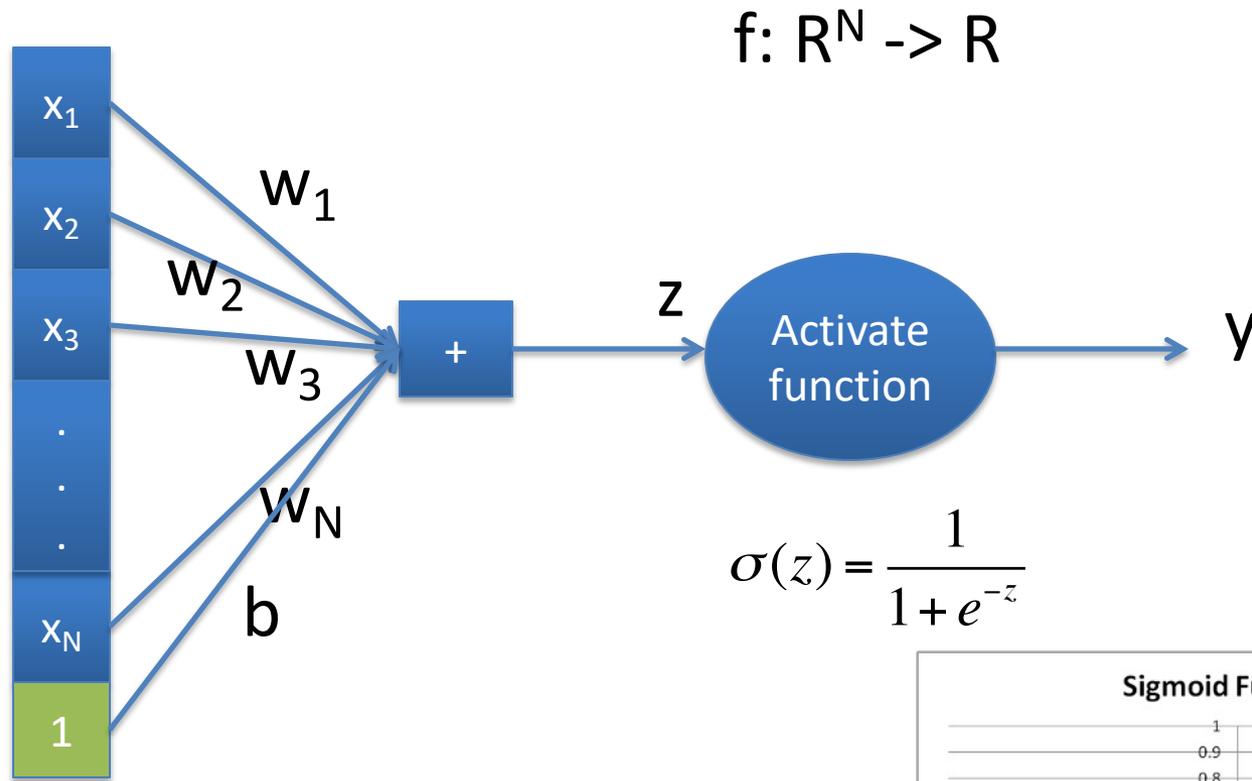
# Model: A single neuron



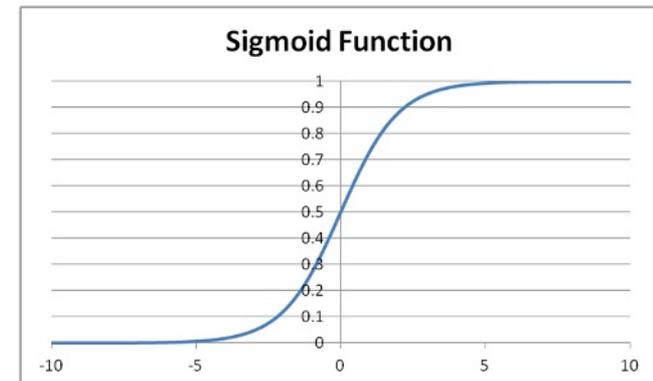
$$\sigma(z) = \frac{1}{1 + e^{-z}}$$



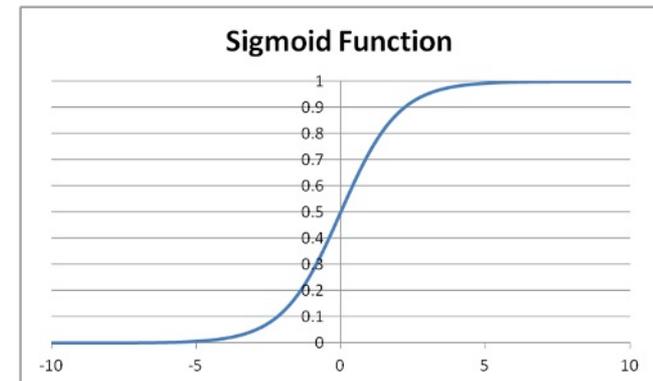
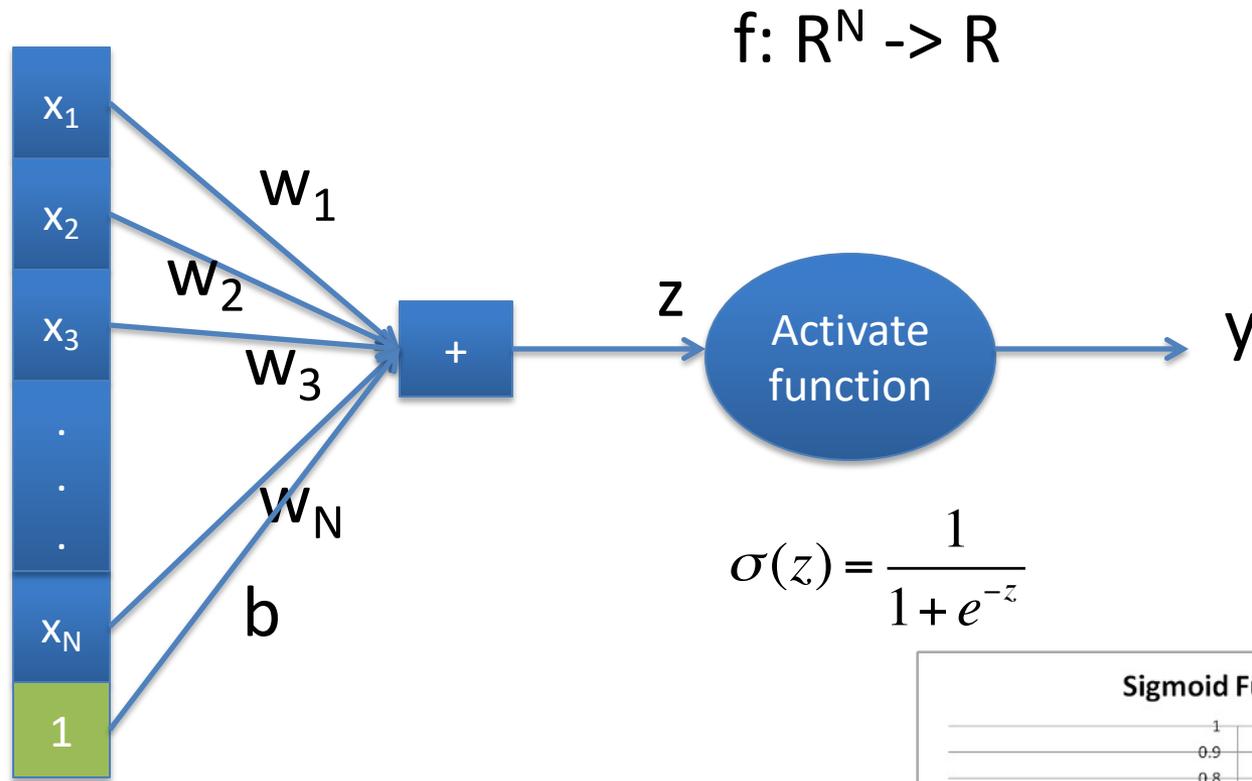
# Model: A single neuron



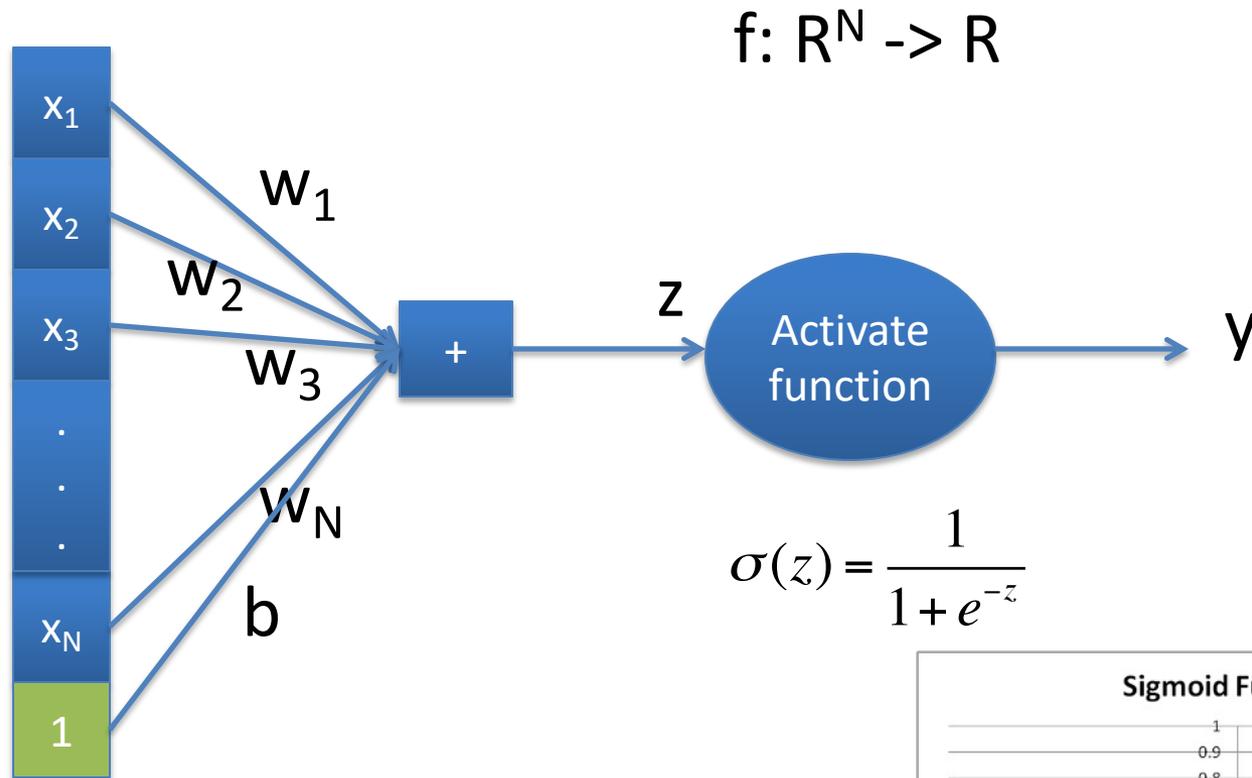
$$\sigma(z) = \frac{1}{1 + e^{-z}}$$



# Model: A single neuron

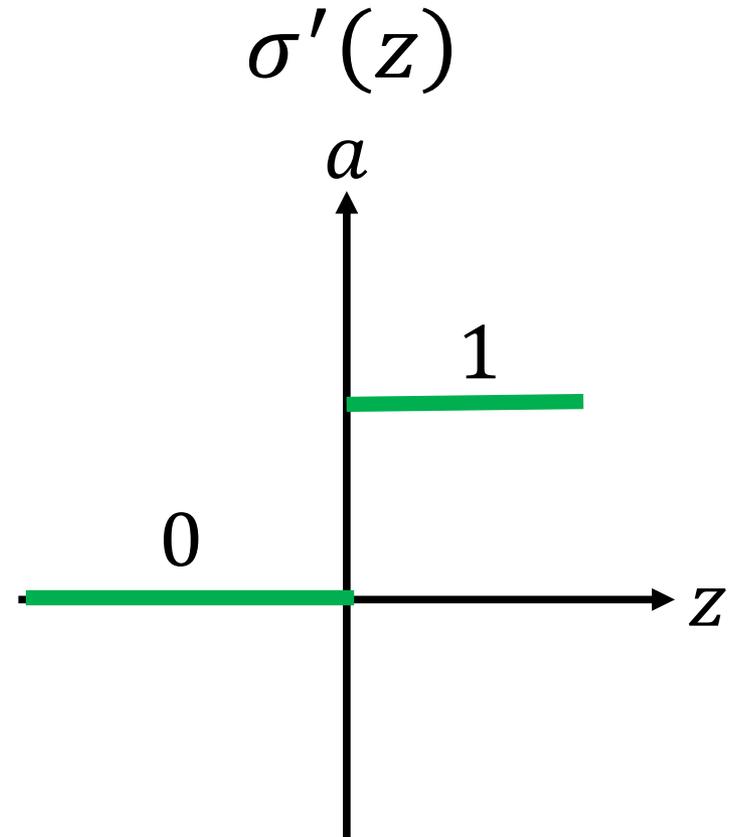
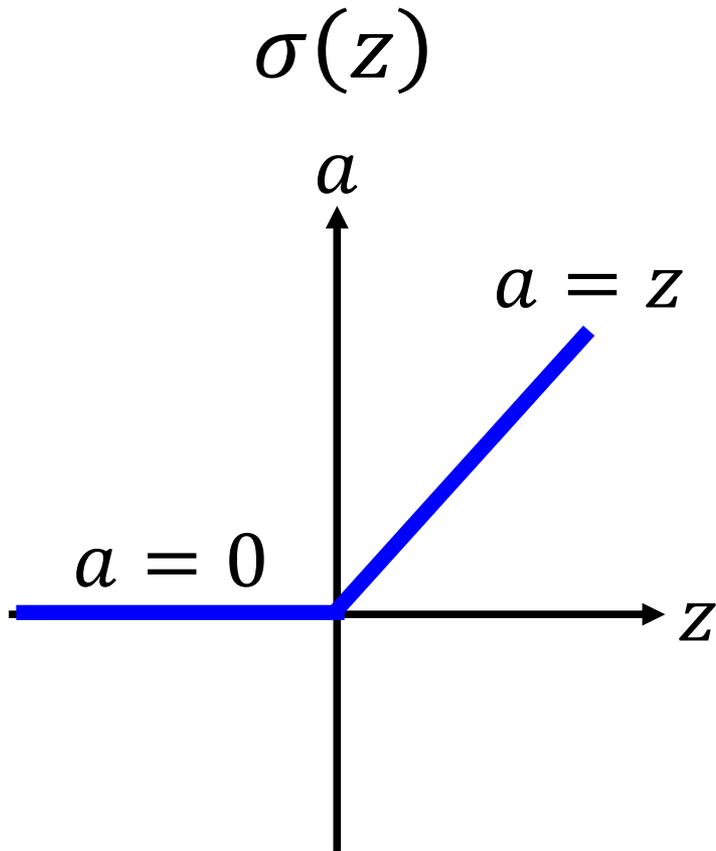


# Model: A single neuron



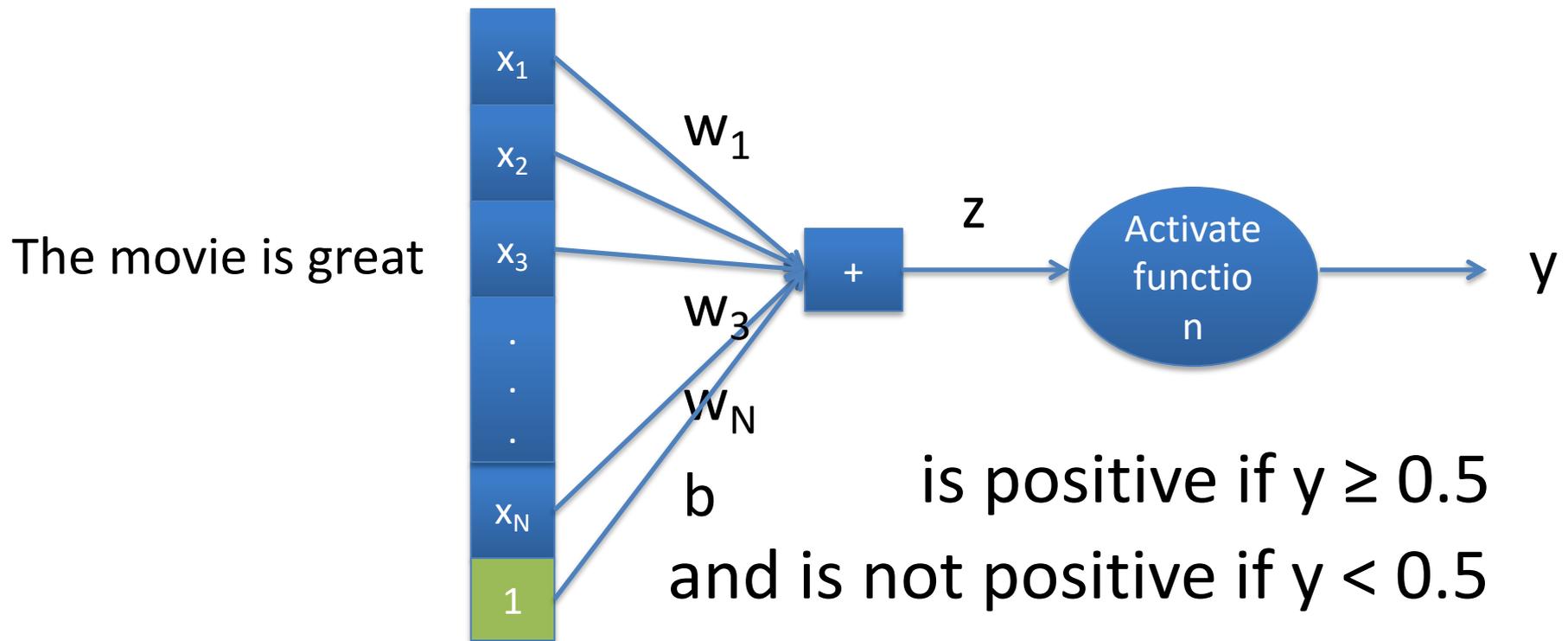
Logistic regression model? – Yes!

# Rectifier Linear Unit (ReLU)



# A single neuron

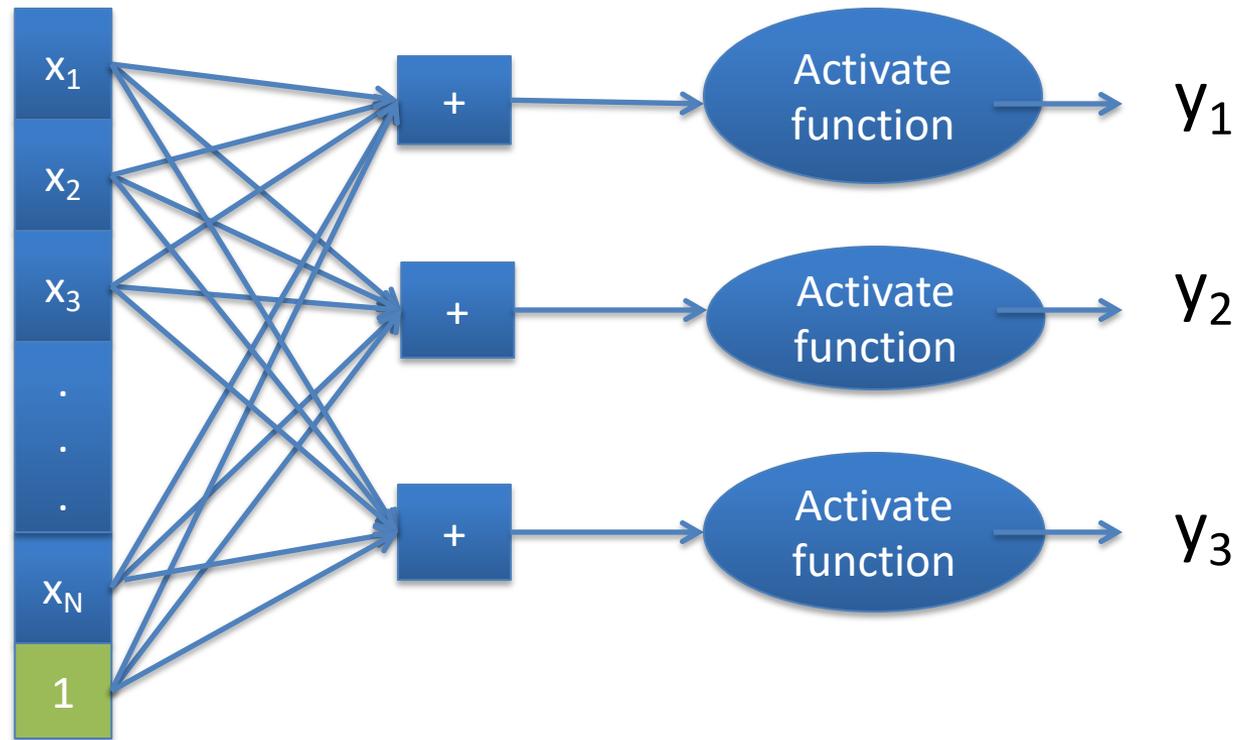
- A single neuron can handle binary classification task



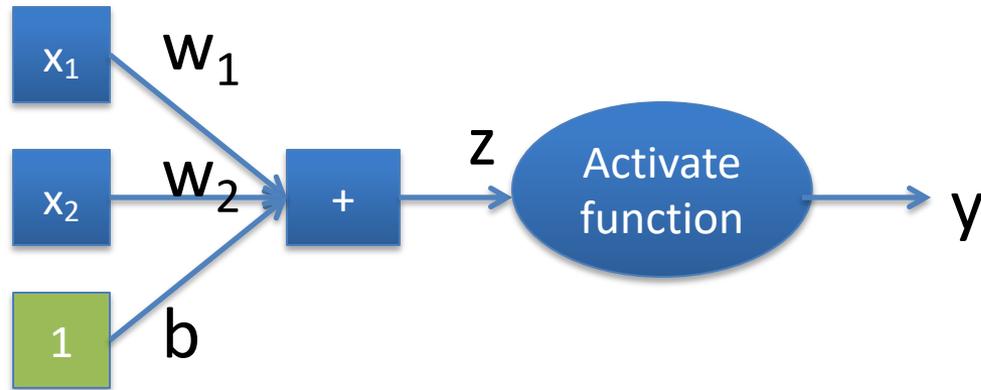
# Layer of neurons

- Sentiment analysis with 3 classes
  - Positive, negative, and neutral

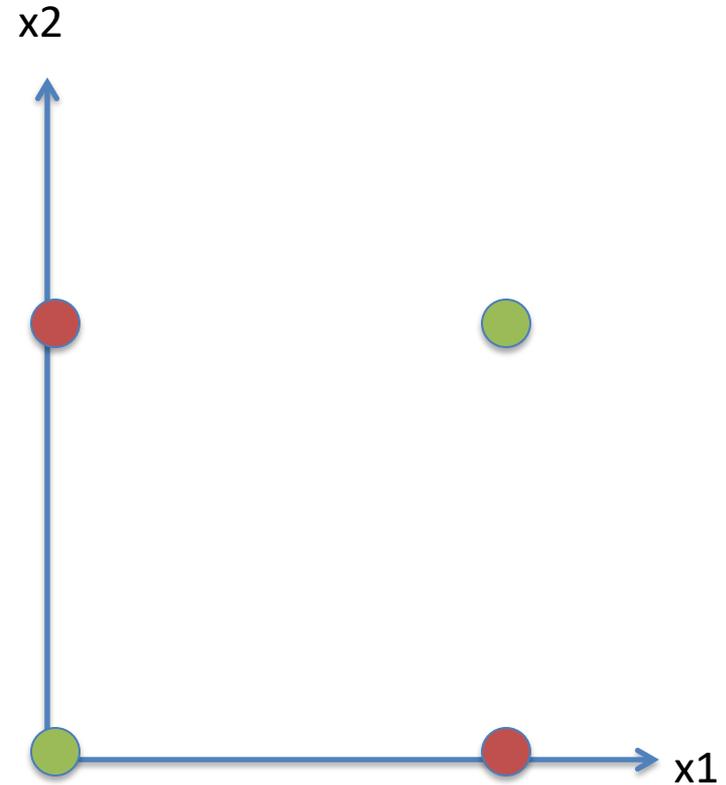
The movie is great



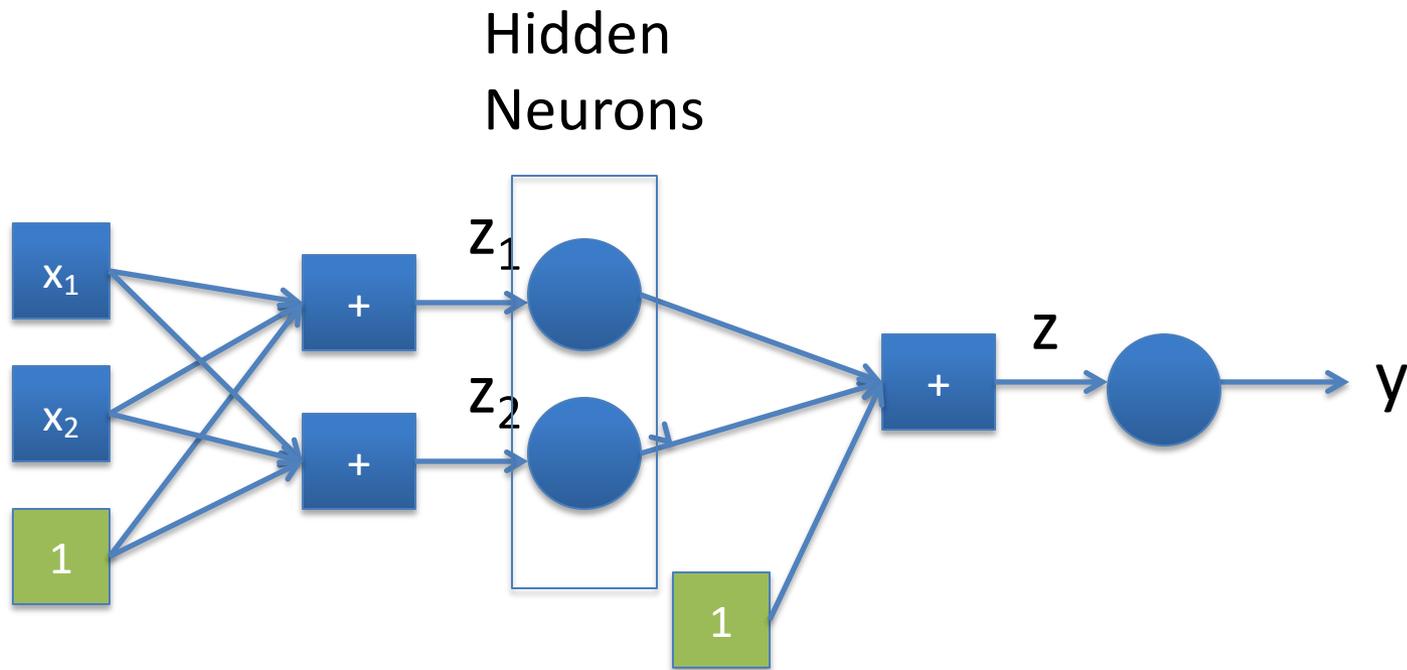
# Limitation of single layer



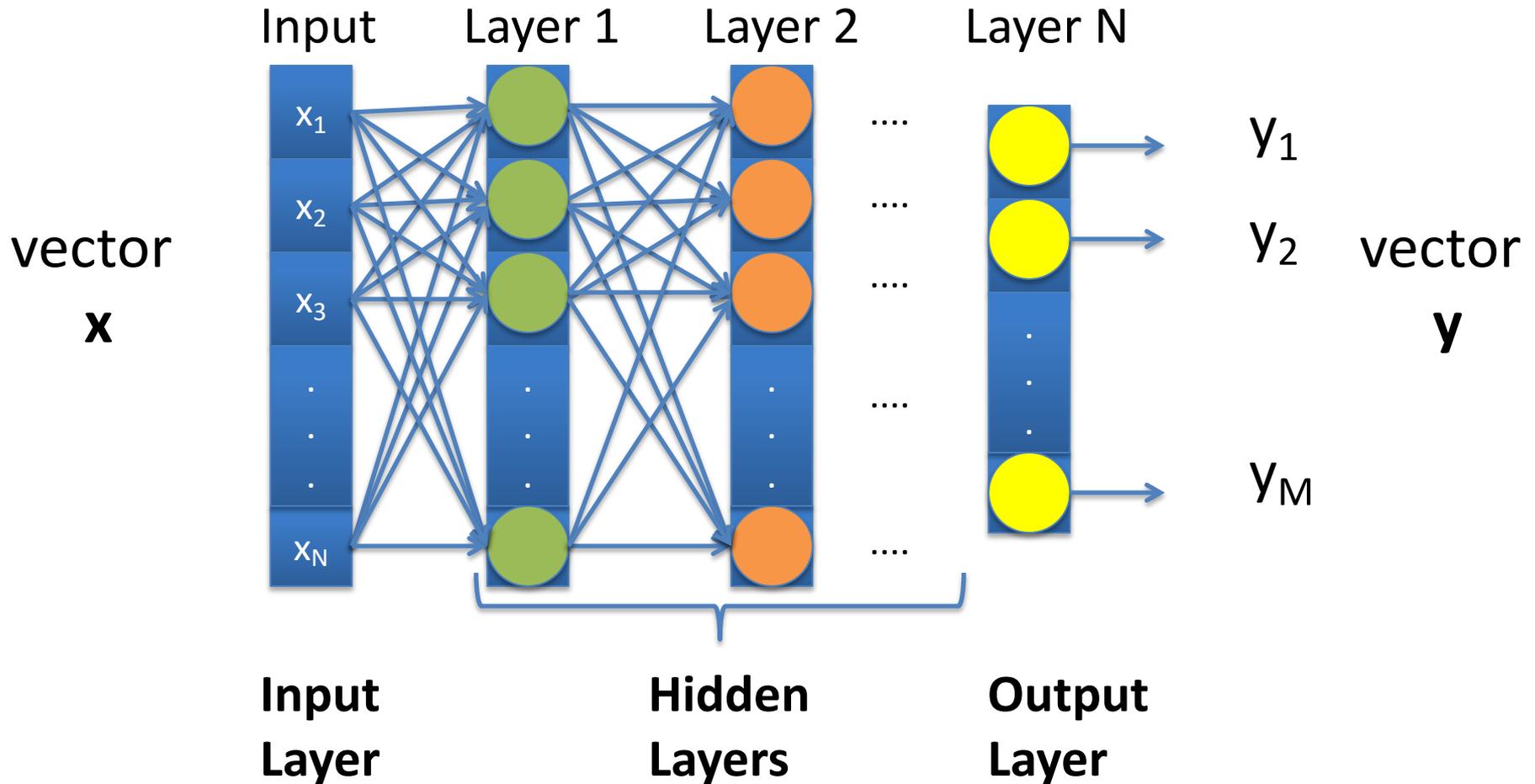
- A single layer of neurons can not handle XOR problem?



# Neural network



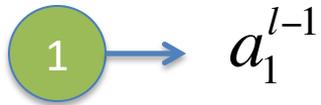
# Neural network with hidden layers



# Live Voting



# Notation



⋮



Layer  $l-1$   
 $N_{l-1}$  nodes

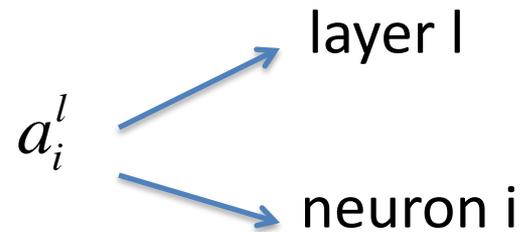


⋮



Layer  $l$   
 $N_l$  nodes

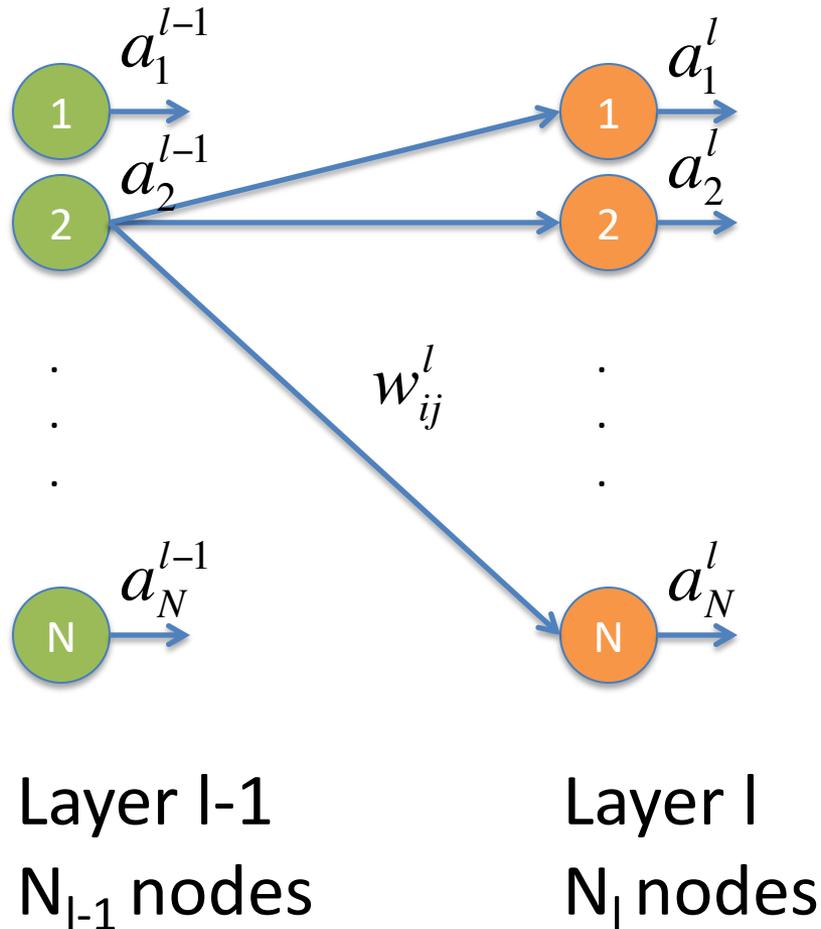
- Output of a neuron:



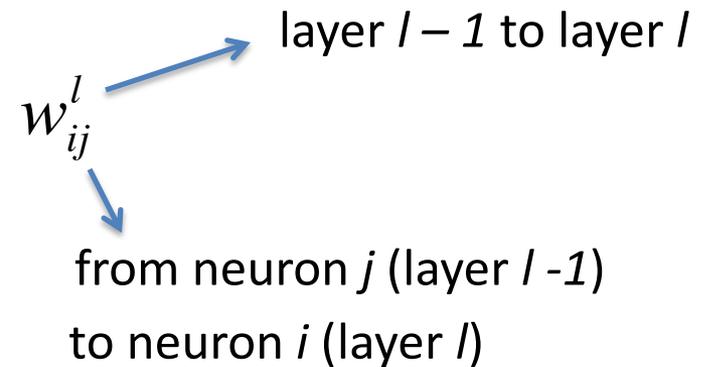
- Output of one layer:

$a^l$  is a vector

# Notation

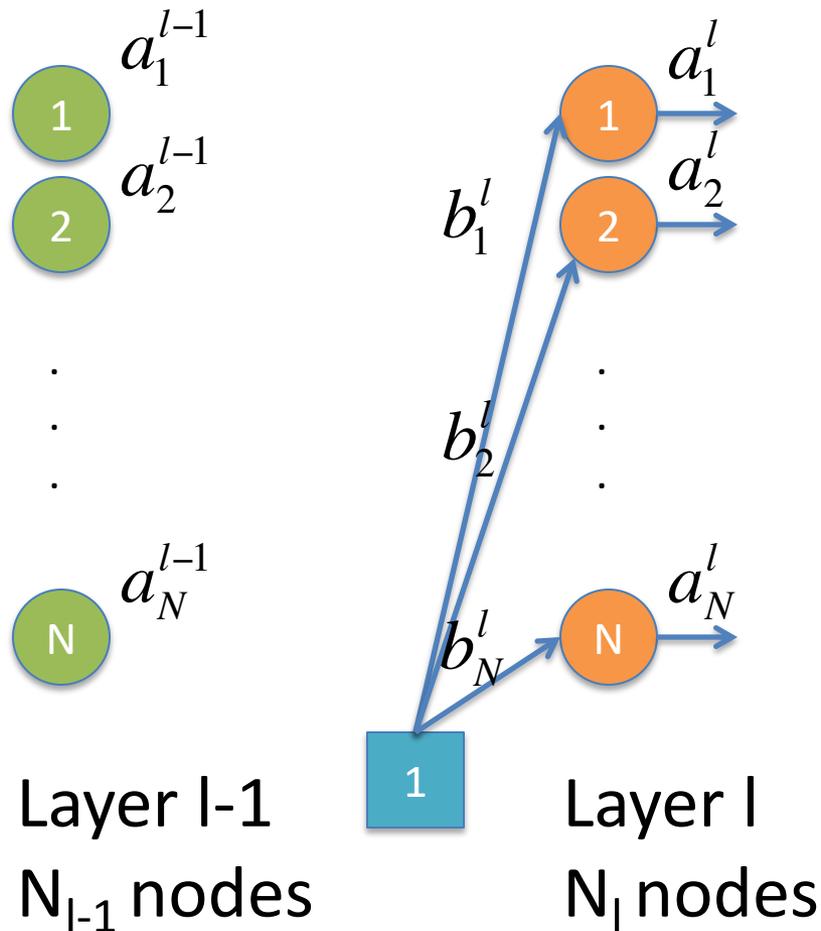


- Weights:

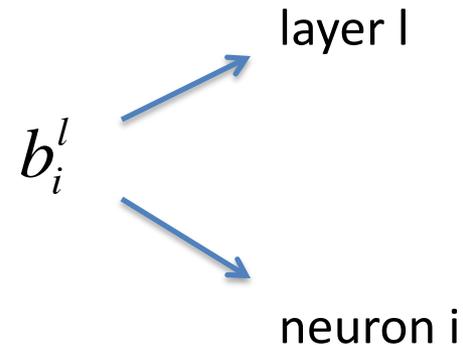


$$W^l = \begin{bmatrix} w_{11}^l & w_{12}^l & \dots \\ w_{21}^l & w_{22}^l & \dots \\ \vdots & & \end{bmatrix} \begin{matrix} N_{l-1} \\ N_l \end{matrix}$$

# Notation

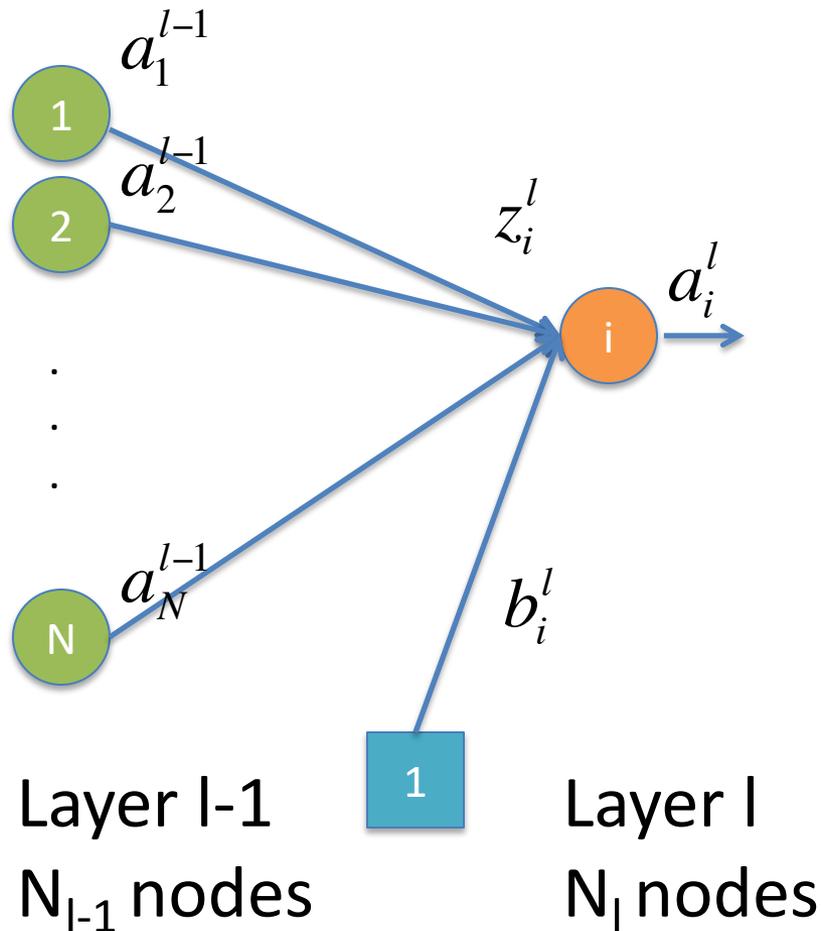


- Biases:



$$b^l = \begin{bmatrix} b_1^l \\ b_2^l \\ \vdots \end{bmatrix} \text{ Bias for all the neurons in layer l}$$

# Notation



$z_i^l$  : input of the activation function for neuron  $i$  at layer  $l$

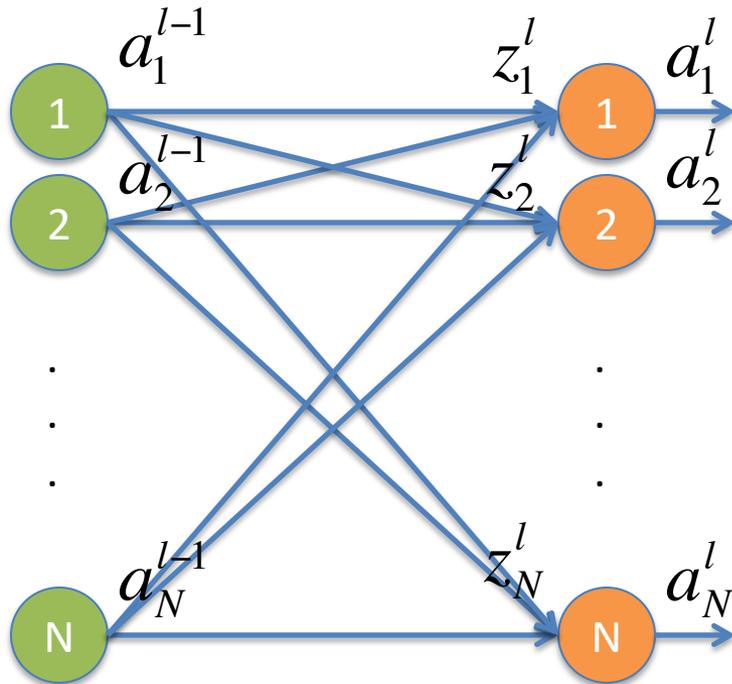
$z^l$  : input of the activation function of all the neurons in layer  $l$

$$z_i^l = w_{i1}^l a_1^{l-1} + w_{i2}^l a_2^{l-1} + \dots + b_i^l$$

or in another form

$$z_i^l = \sum_{j=1}^{N_{l-1}} w_{ij}^l a_j^{l-1} + b_i^l$$

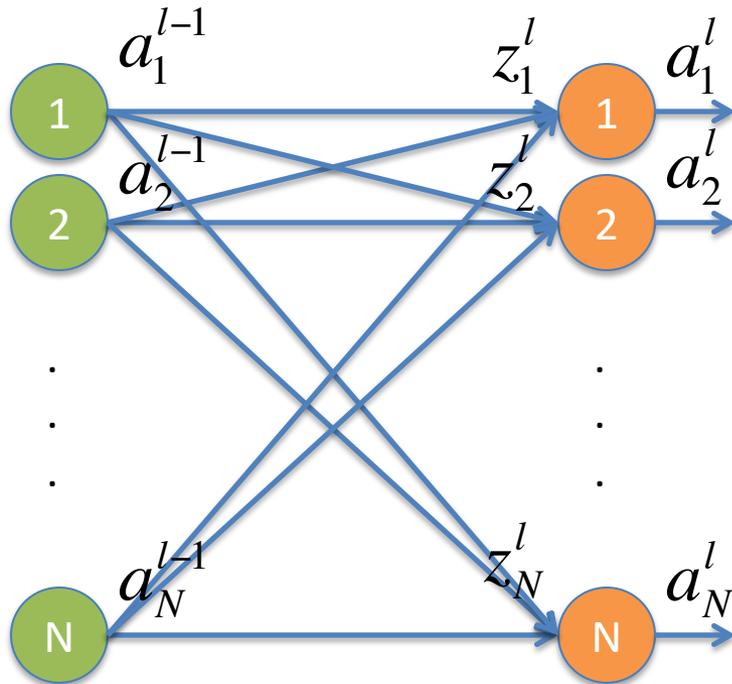
# Relations between layer outputs



Layer  $l-1$   
 $N_{l-1}$  nodes

Layer  $l$   
 $N_l$  nodes

# Relations between layer outputs



Layer l-1  
 $N_{l-1}$  nodes

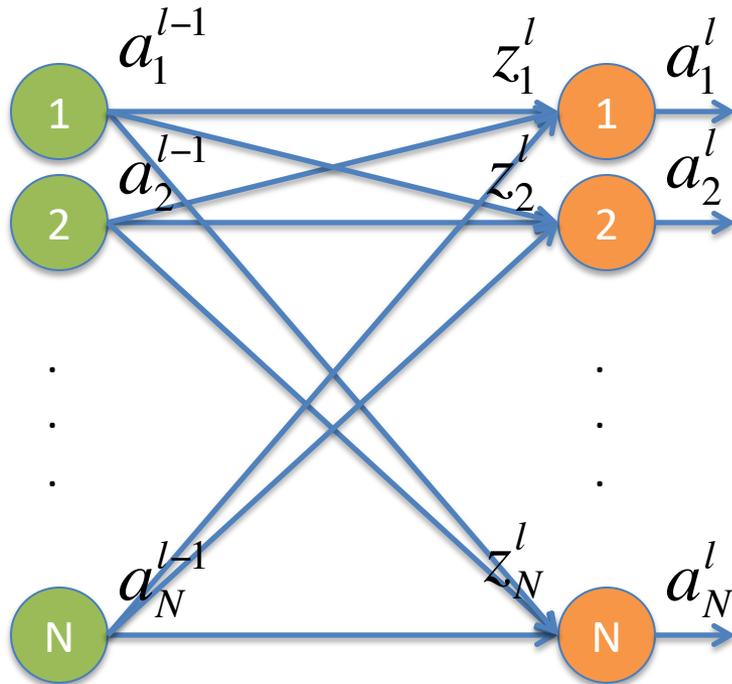
Layer l  
 $N_l$  nodes

$$\begin{aligned} z_1^l &= w_{11}^l a_1^{l-1} + w_{12}^l a_2^{l-1} + \dots + b_1^l \\ z_2^l &= w_{21}^l a_1^{l-1} + w_{22}^l a_2^{l-1} + \dots + b_2^l \\ &\dots \\ z_N^l &= w_{N1}^l a_1^{l-1} + w_{N2}^l a_2^{l-1} + \dots + b_N^l \end{aligned}$$



$$z^l = W^l a^{l-1} + b^l$$

# Relations between layer outputs



Layer l-1  
 $N_{l-1}$  nodes

Layer l  
 $N_l$  nodes

$$a_1^l = \sigma(z_1^l)$$
$$a_2^l = \sigma(z_2^l)$$

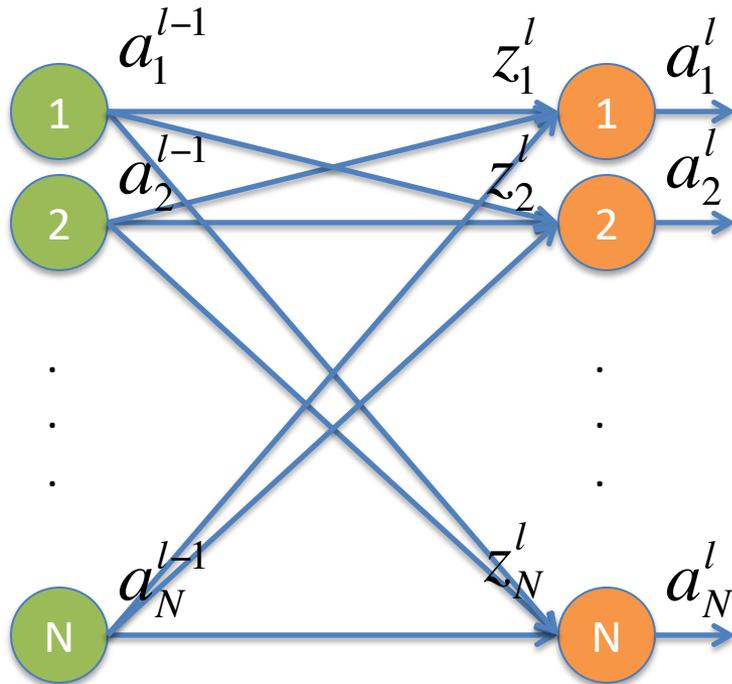
....

$$a_N^l = \sigma(z_N^l)$$



$$a^l = \sigma(z^l)$$

# Relations between layer outputs



Layer l-1  
 $N_{l-1}$  nodes

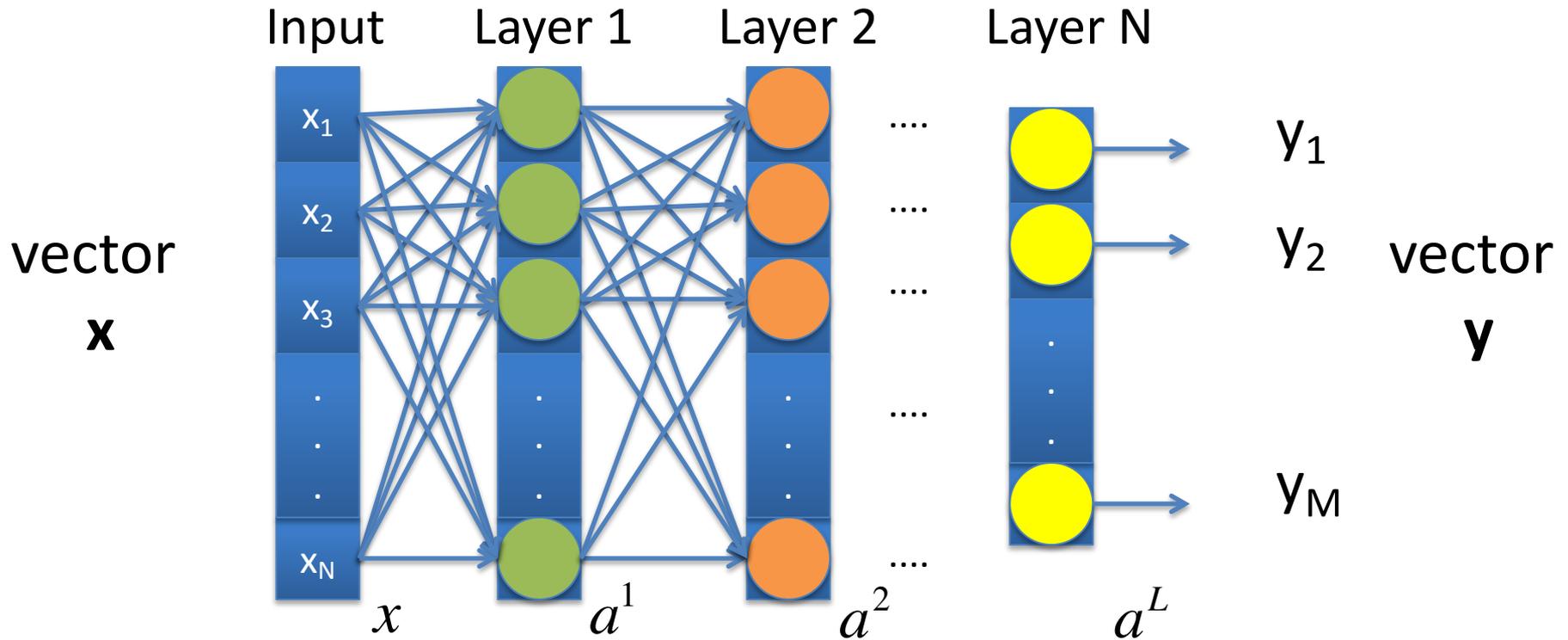
Layer l  
 $N_l$  nodes

$$z^l = W^l a^{l-1} + b^l$$

$$a^l = \sigma(z^l)$$

$$a^l = \sigma(W^l a^{l-1} + b^l)$$

# Computation of the final output

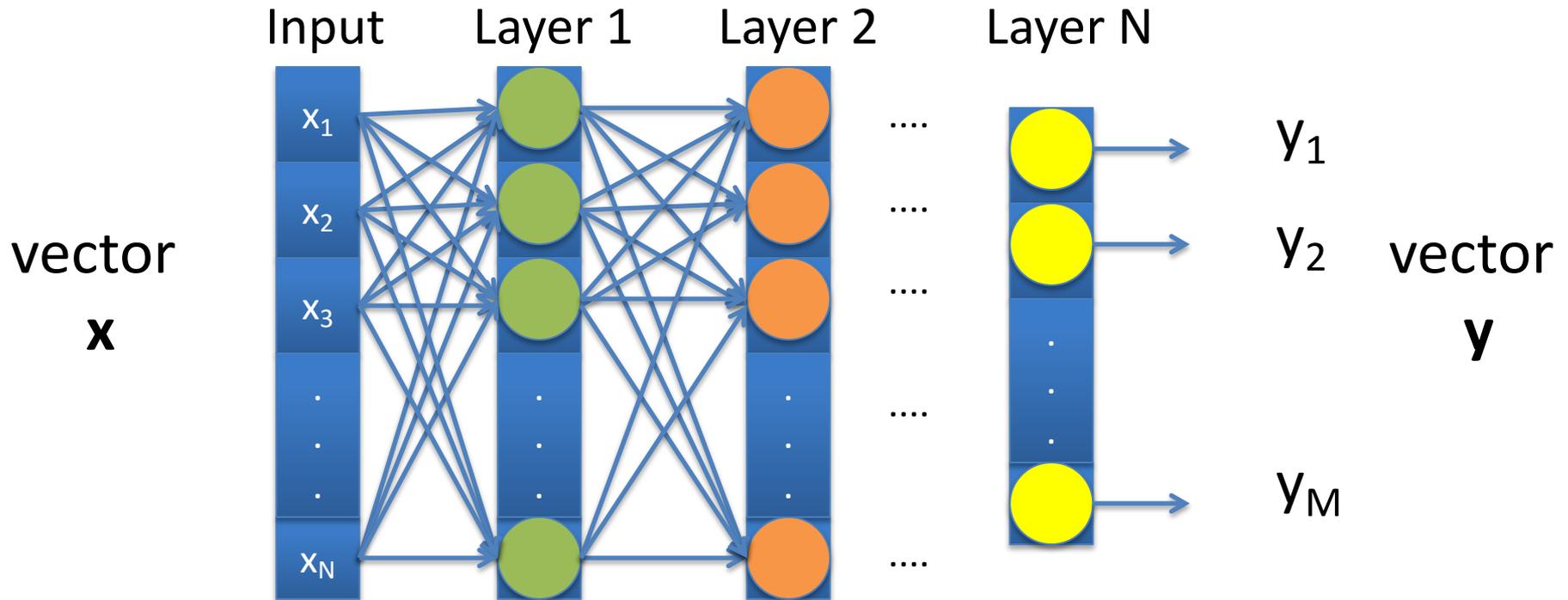


$$a^1 = \sigma(W^1 x + b^1)$$

$$a^2 = \sigma(W^2 a^1 + b^2)$$

$$a^L = \sigma(W^L a^{L-1} + b^L)$$

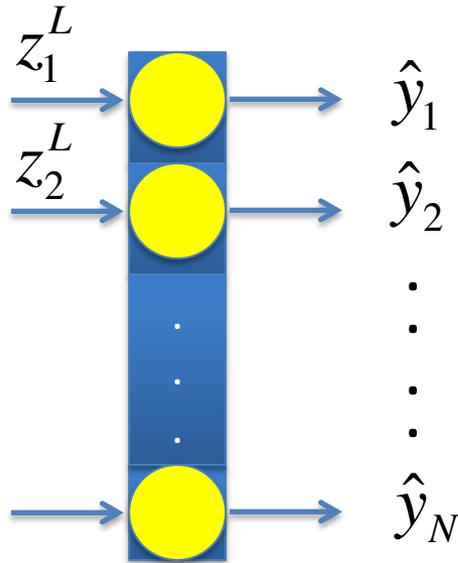
# Computation of the final output



$$y = f(x) = \sigma(W^L \dots \sigma(W^2 \sigma(W^1 x + b^1) + b^2) \dots + b^L)$$

# Softmax function

Output layer L



- Outputs are probabilities

$$0 < y_i < 1$$

$$\sum_i y_i = 1$$

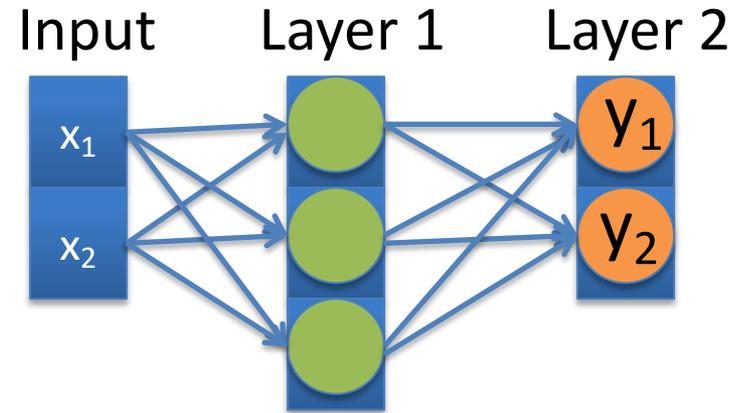
- Softmax function:

$$y_i = \frac{e^{z_i^L}}{\sum_j e^{z_j^L}}$$

$$\begin{aligned} \frac{\delta y_i}{\delta z_j} &= y_j(1 - y_j) & i = j \\ &= -y_i y_j & i \neq j \end{aligned}$$

# Exercise 1

- Compute the output  $\mathbf{y}$  of the following network
- Layer 1 uses ReLU
- Layer 2 uses Softmax



$$\mathbf{x} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$\mathbf{W}^1 = \begin{bmatrix} 1 & 1 \\ 0 & 1 \\ -1 & 0 \end{bmatrix}$$

$$\mathbf{b}^1 = \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}$$

$$\mathbf{W}^2 = \begin{bmatrix} 1 & 1 & -1 \\ 0 & 1 & 0 \end{bmatrix}$$

$$\mathbf{b}^2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

# Live Voting



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Thanks for listening!

