

A close-up photograph of a hummingbird with iridescent green feathers and a bright orange throat patch, hovering near a large, vibrant red flower. The background is a soft-focus green.

rethinking insurance



June 2020

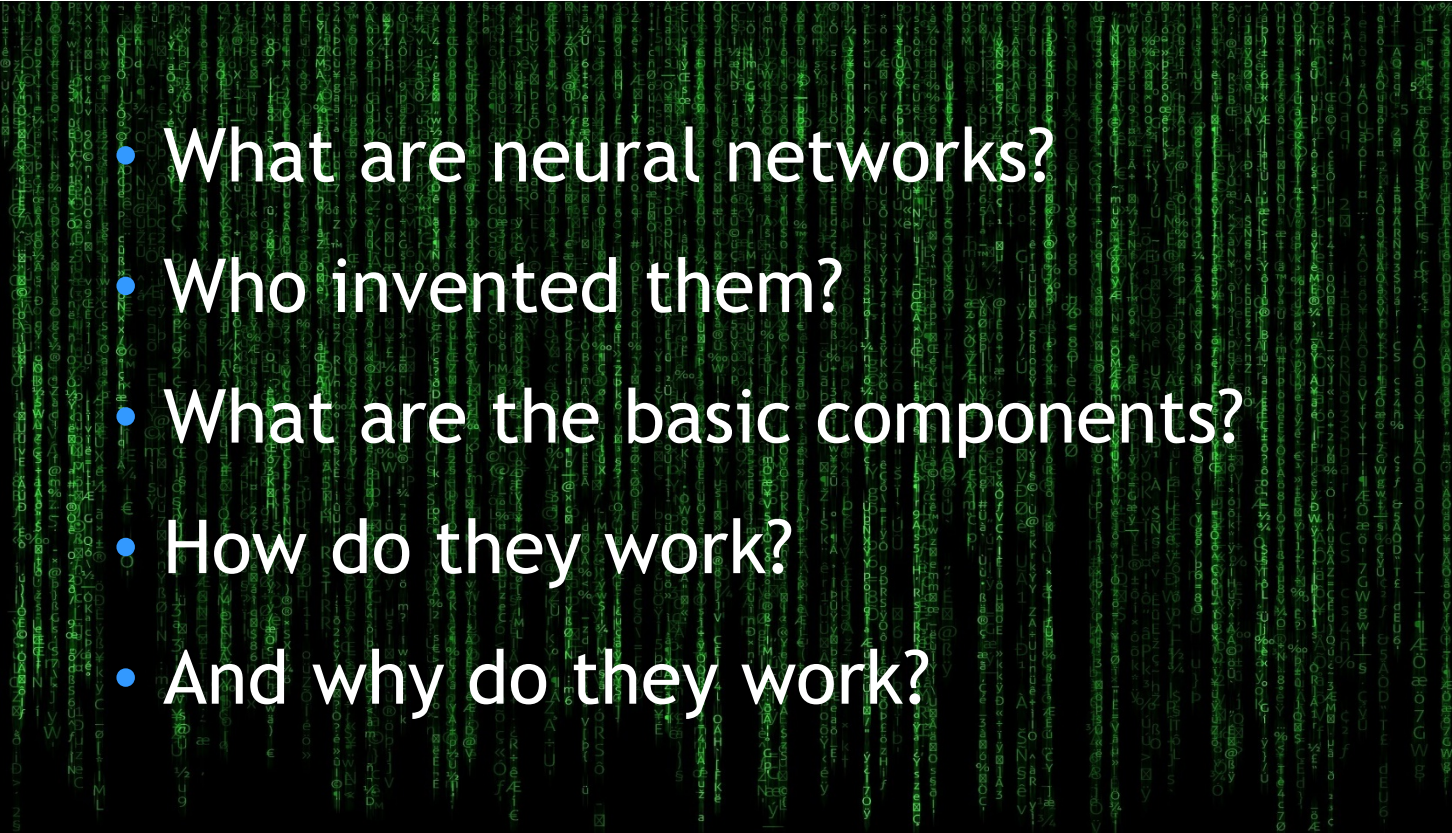
Artificial neural networks basics, part 1 - forwardpass

An elementary introduction

Dr. Stefan Nörtemann, msg life central europe

Introduction

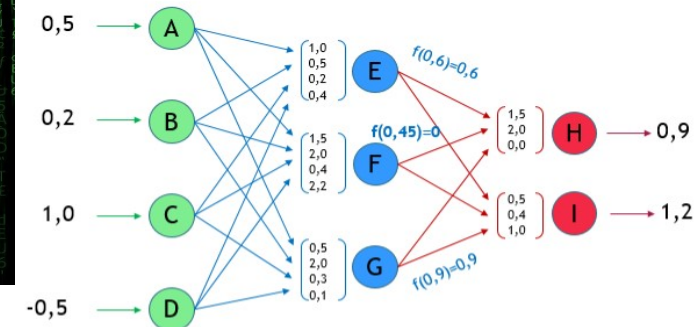
What's this about?

- 
- The background of the slide is a dark image showing a computer screen with green text, resembling a terminal or code editor, with various symbols and characters visible.
- What are neural networks?
 - Who invented them?
 - What are the basic components?
 - How do they work?
 - And why do they work?

Introduction

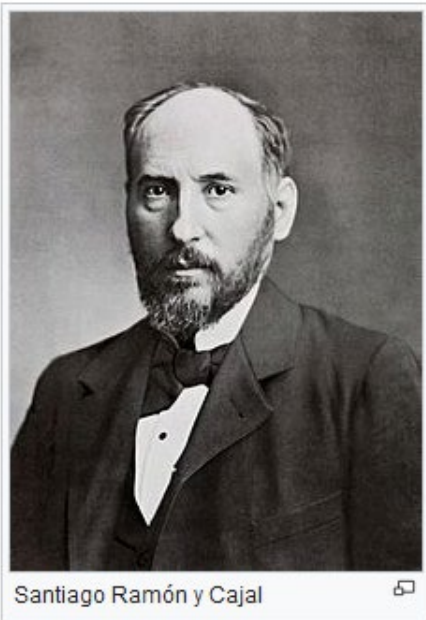
What's this about?

- What are neural networks?
- Who invented them?
- What are the basic components?
- How do they work?
- And why do they work?

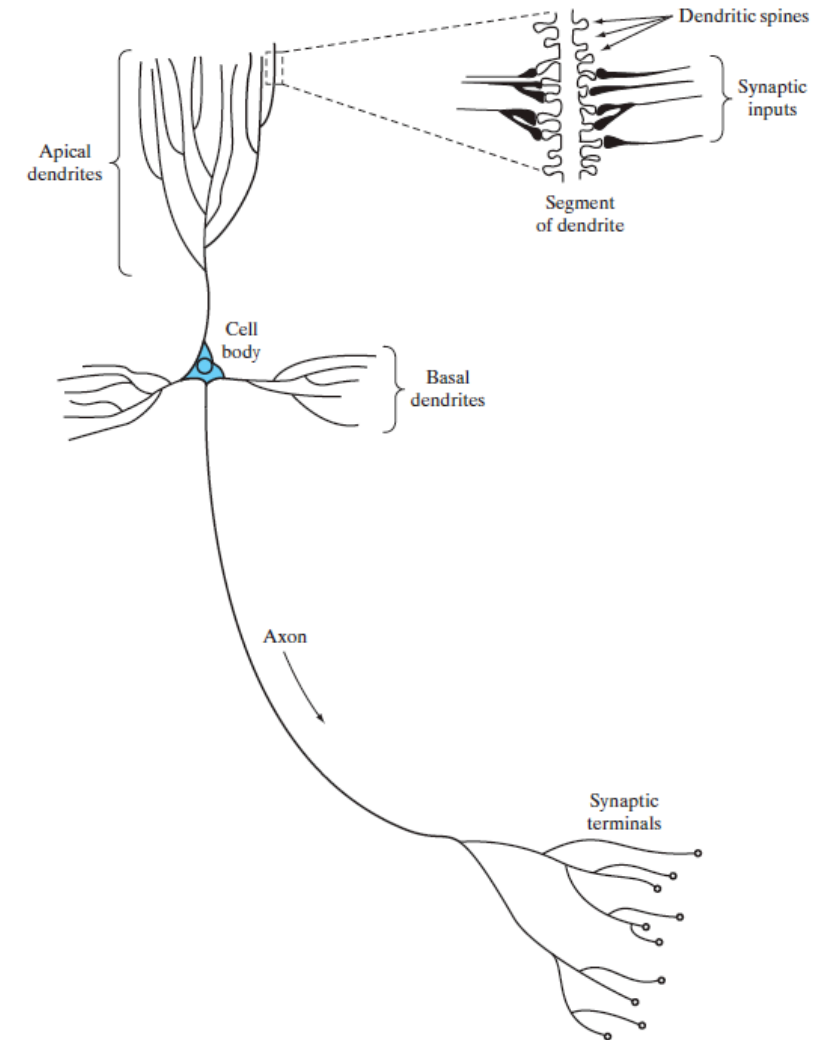


Neuronal networks

Motivation from brain research



- Ramón y Cajál (1911): The idea of neurons
- The brain as an extremely efficient computer
- Imitation of intelligence



Neural Networks & Deep Learning

Warren McCulloch & Walter Pitts, 1943

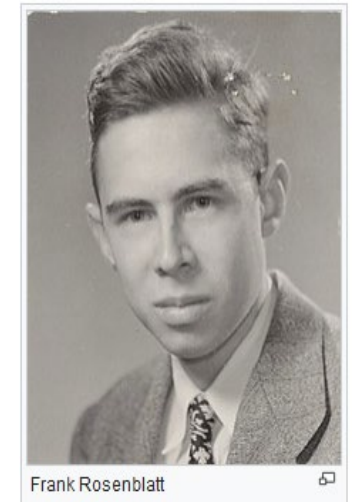
- first idea for an artificial neural network, consisting of linked elementary units
- for the calculation of logical and arithmetic functions



Warren McCulloch and Walter Pitts

Frank Rosenblatt, 1958

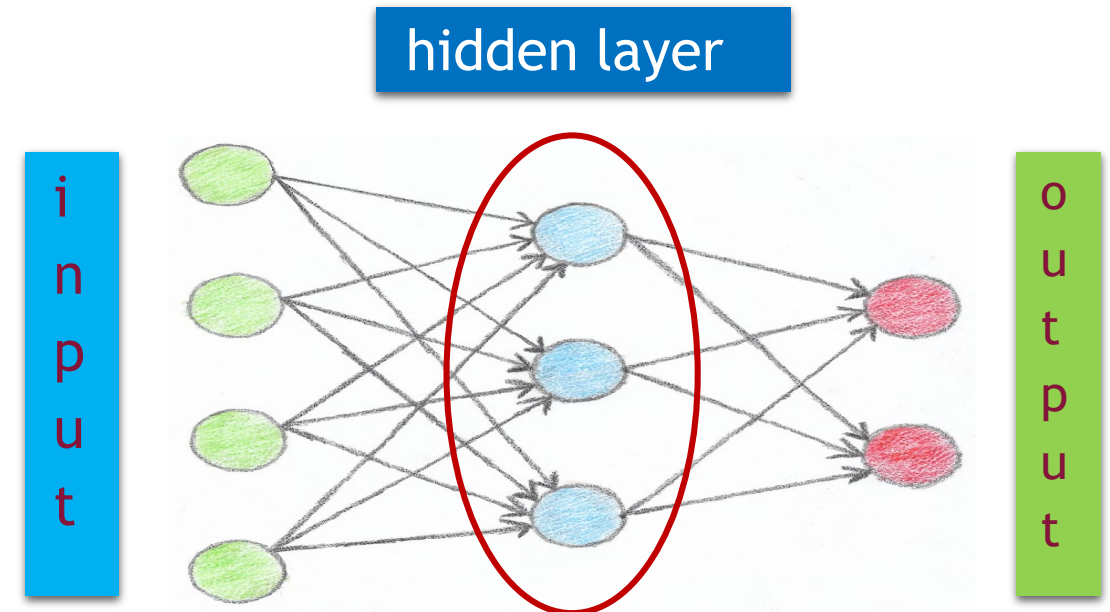
- Neurocomputer *Marc I Perceptron*: Multi-Layer Perceptron
- **Idea:** Very many calculation units that only become "intelligent" through their interaction
- Since then a constant up and down: hypes and disillusion
- Boom for about 10 years due to the enormous increase in computing capacity



Frank Rosenblatt

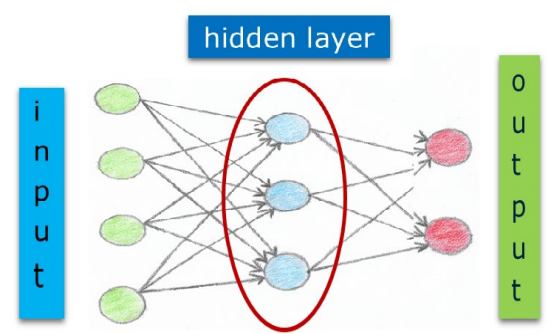
kNN: Input - Hidden layers - Output

- **idea:** "make it like the human brain"
- **method:** Linking of so-called artificial neurons (units)
- **components:** input units, hidden layer, output units
- **connections:** exchange of data (= numbers) between the units



- each connection has a **weight** (which can "change")
- in each unit an **input** is processed and an **output** is generated, which can again be an input for a downstream unit.

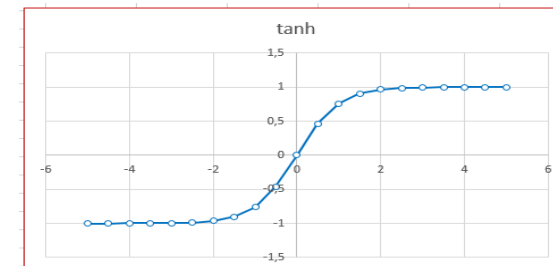
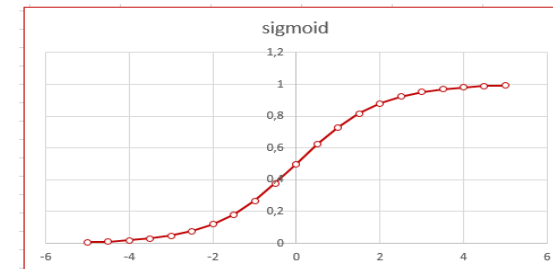
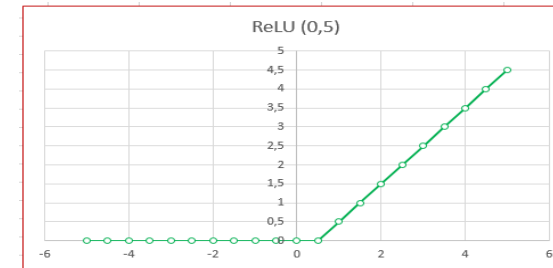
kNN: Activation function(s)



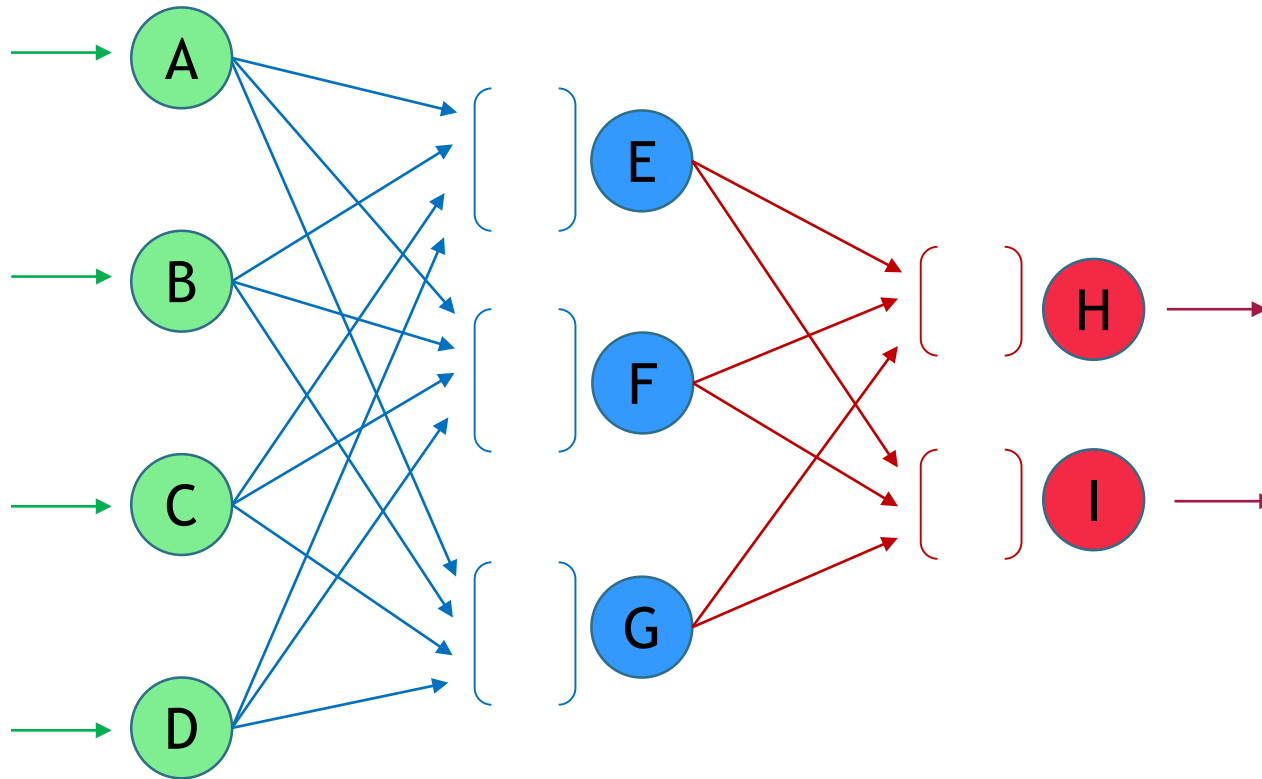
The **output** of a unit is calculated from the sum of the weighted input values, possibly a bias b and a so-called **activation function** f .

Examples of activation functions

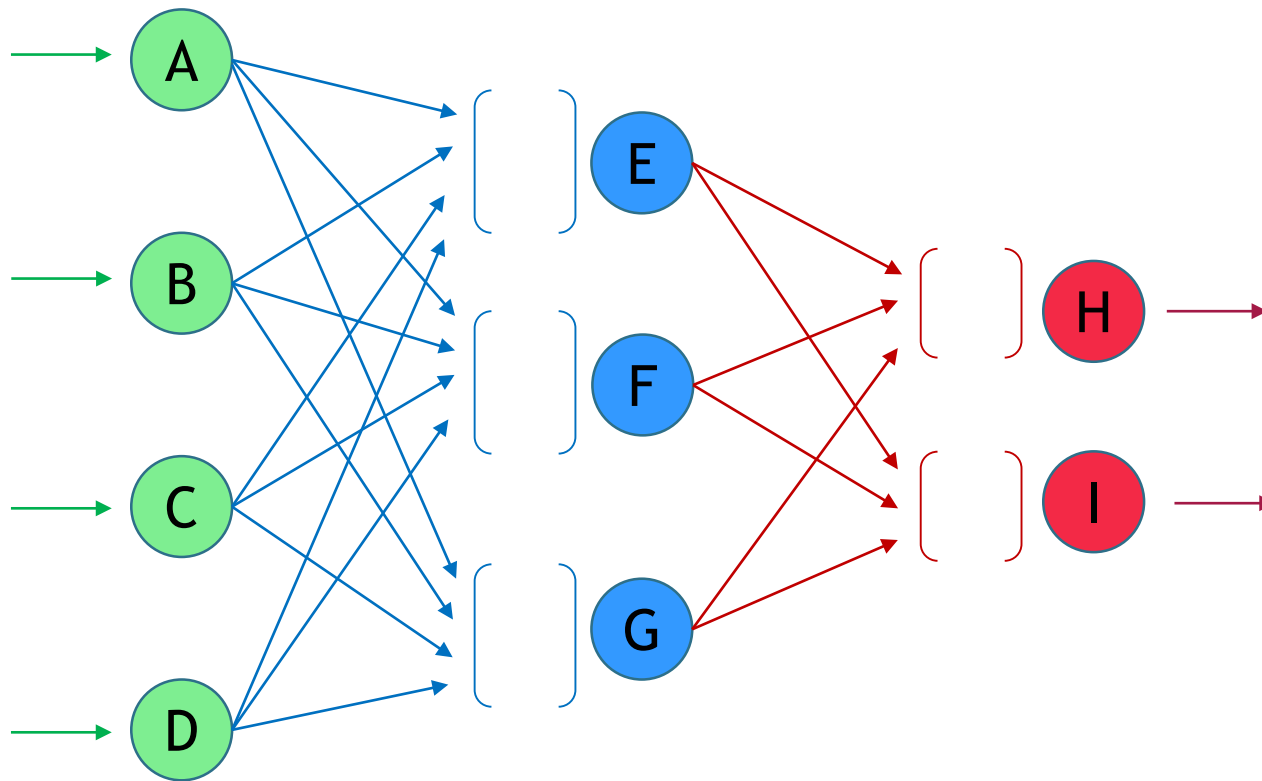
- Rectified Linear Unit (ReLU) $f(x, \theta) = \max(0, x - \theta)$
- Sigmoid function $f(x) = \frac{1}{1+e^{-x}}$
- hyperbolic tangent $f(x) = \tanh(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}}$
- Logistic $f(x, \theta) = \frac{1}{1+e^{-(x-\theta)}}$
- Leaky ReLU $f(x, \theta) = \max(\alpha(x - \theta), x - \theta), \alpha > 0$ small
- ...



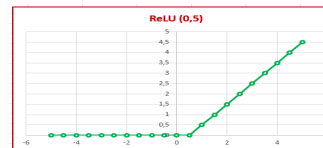
A simple example



A simple example

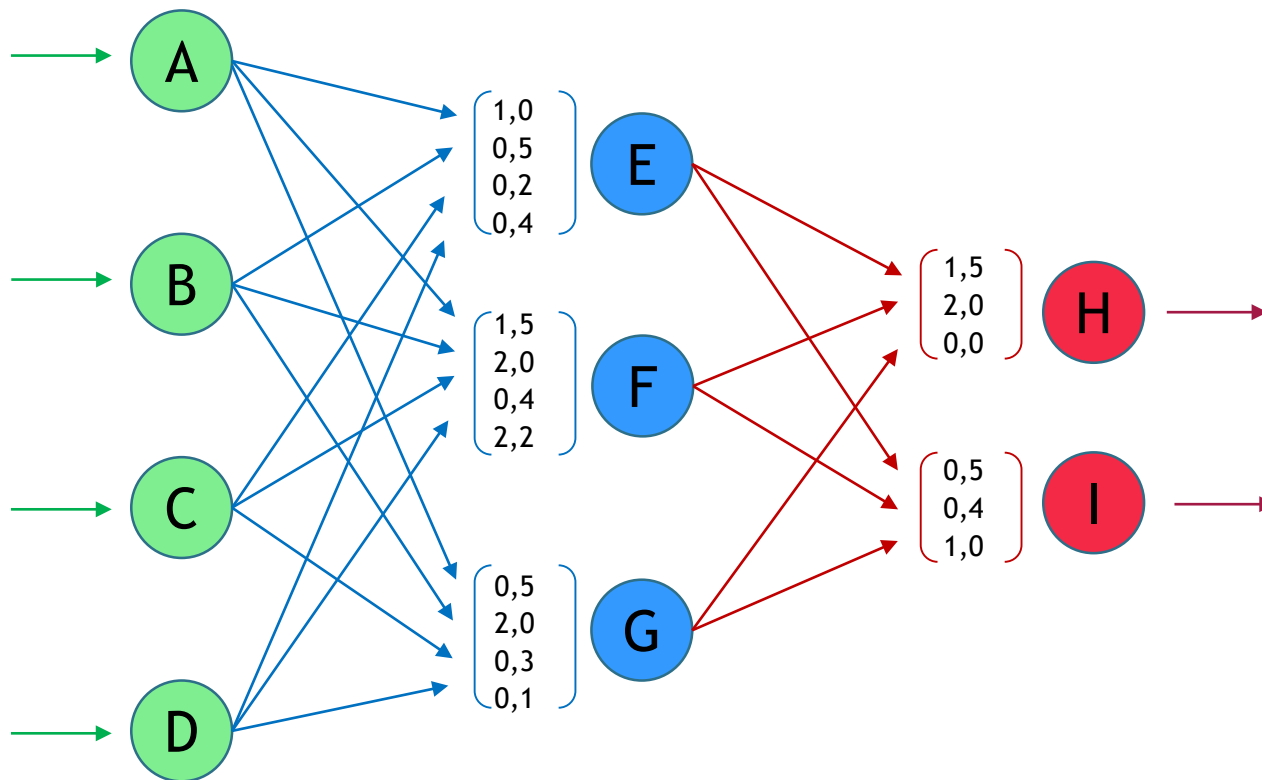


Activation function: $f = \text{ReLU} (0.5)$

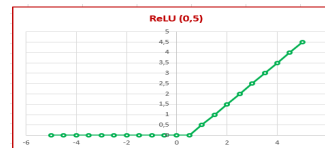


,no bias (so that the graphic does not become too confusing).

A simple example

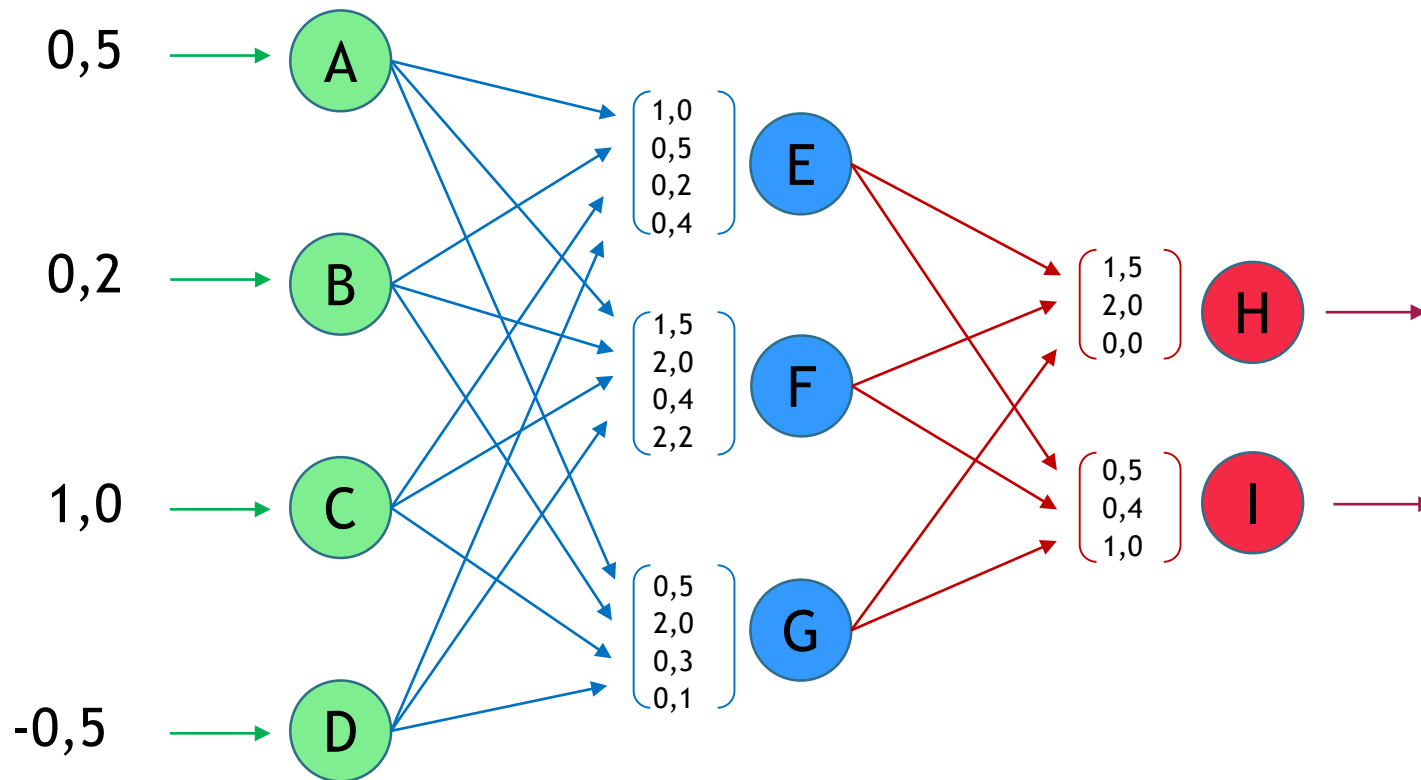


Activation function: $f = \text{ReLU} (0.5)$

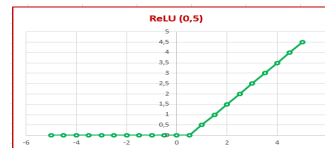


,no bias (so that the graphic does not become too confusing).

A simple example

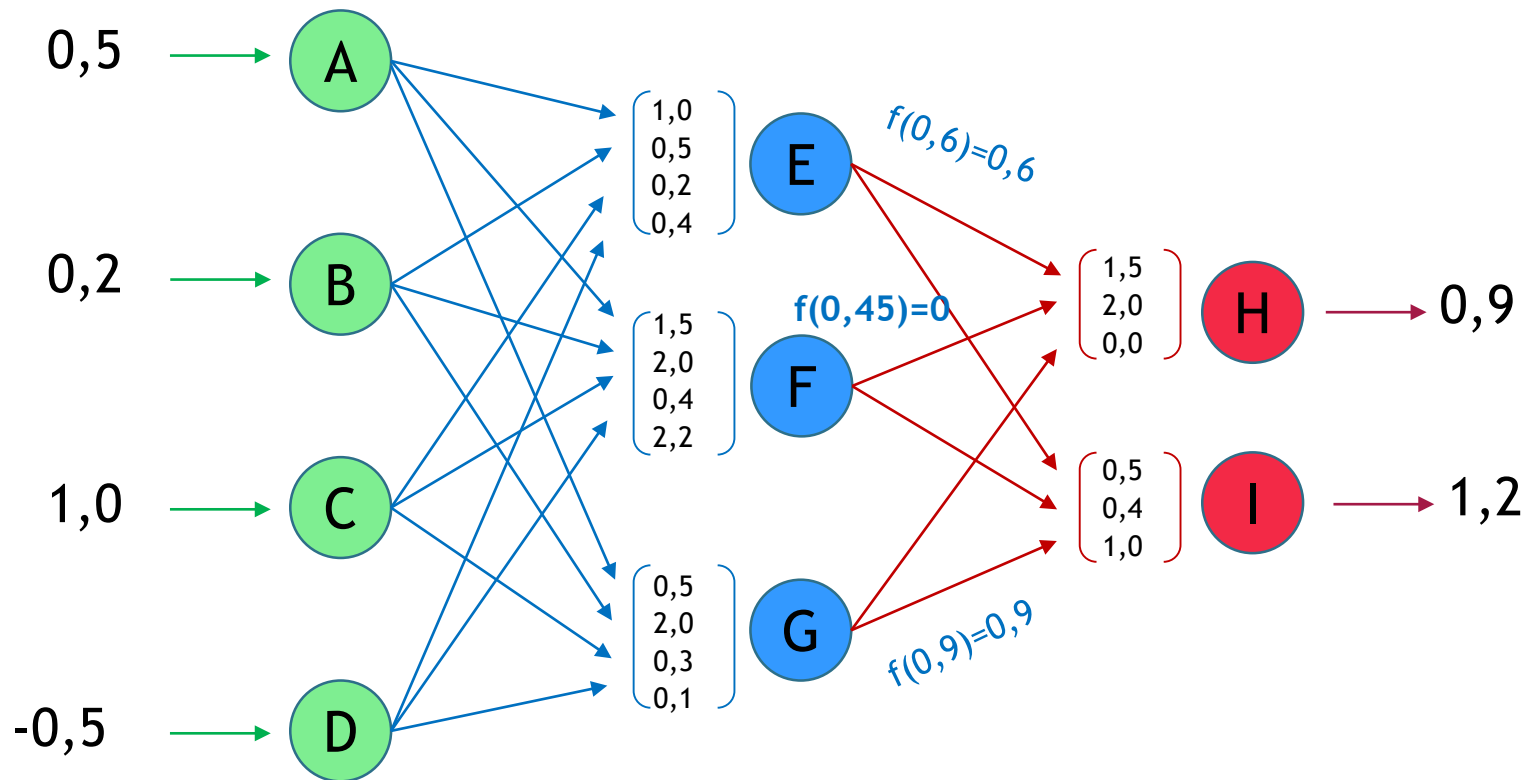


Activation function: $f = \text{ReLU} (0.5)$

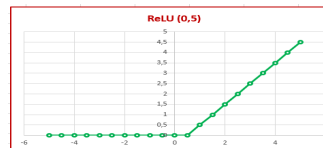


,no bias (so that the graphic does not become too confusing).

Forward propagation



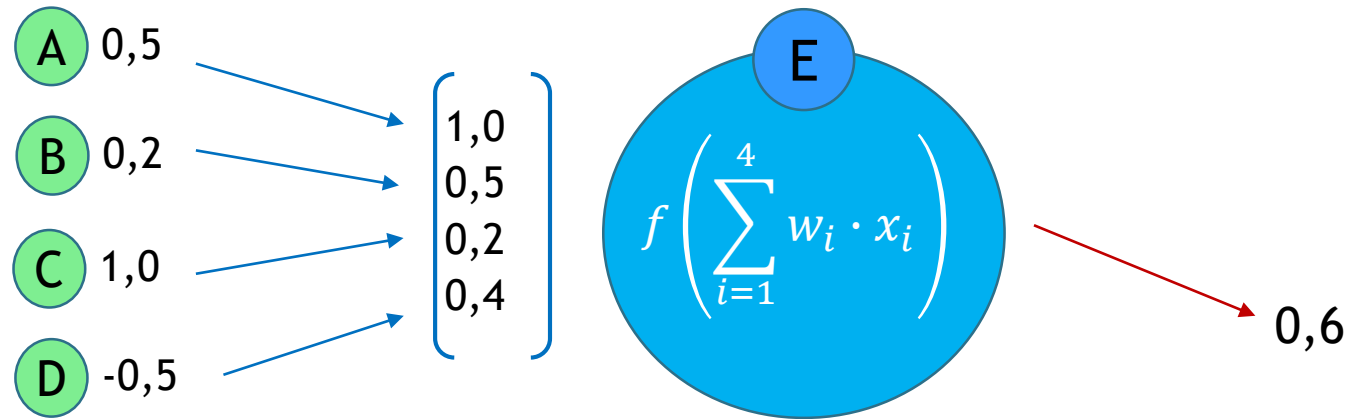
Activation function: $f = \text{ReLU} (0.5)$



,no bias (so that the graphic does not become too confusing).

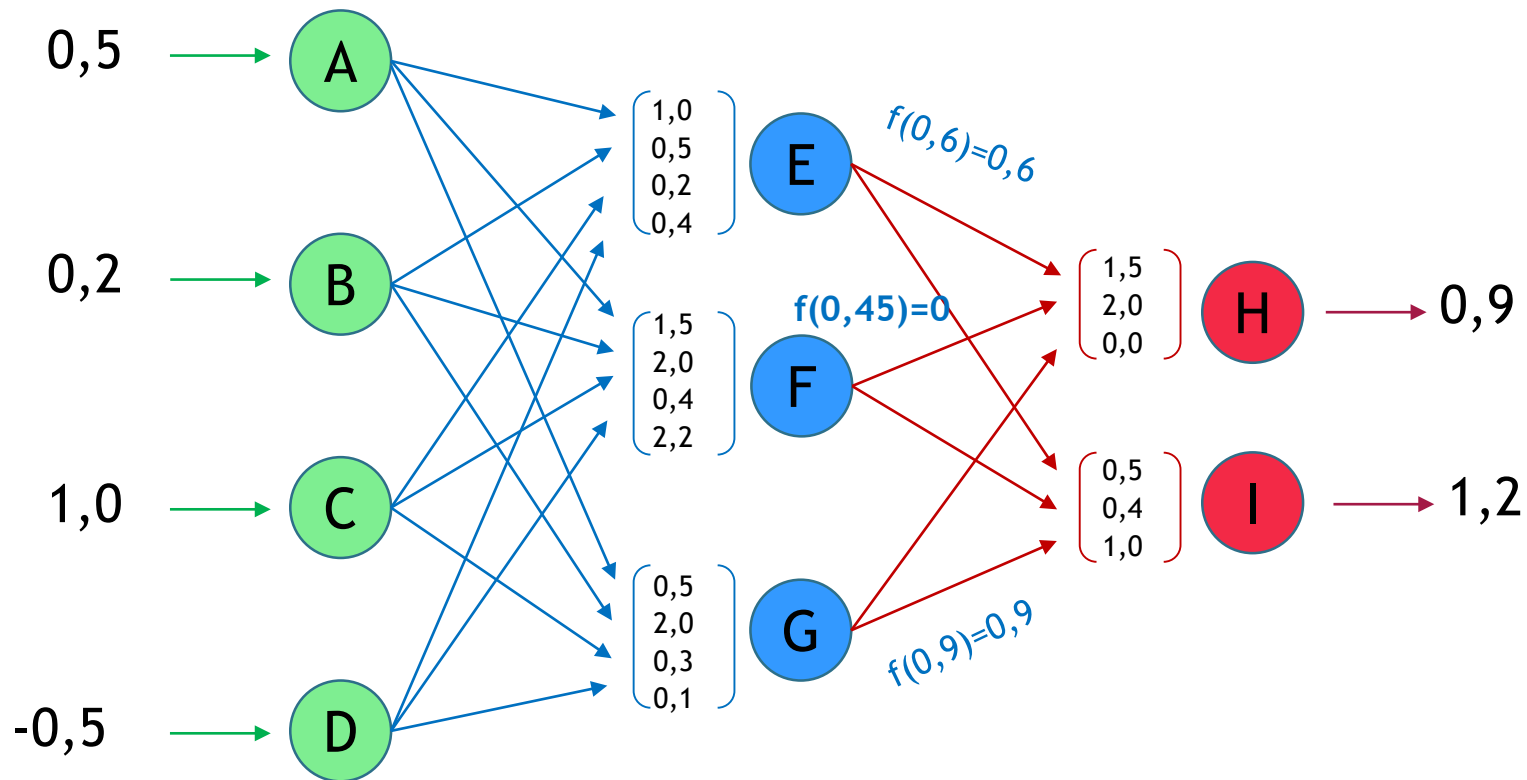
Zoom into the unit E

Activation function: $f = \text{ReLU} (0.5)$

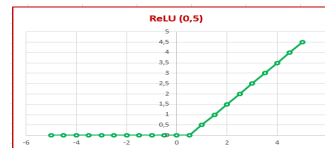


$$f\left(\sum_{i=1}^4 w_i \cdot x_i\right) = f(0,5 \cdot 1 + 0,2 \cdot 0,5 + 1 \cdot 0,2 + (-0,5) \cdot 0,4) = f(0,6) = \mathbf{0,6}$$

Forward propagation



Activation function: $f = \text{ReLU} (0.5)$



,no bias (so that the graphic does not become too confusing).

Excursus: Matrix notation - forward propagation

$$I = \begin{pmatrix} 0,5 \\ 0,2 \\ 1,0 \\ -0,5 \end{pmatrix}$$

$$W_{\text{Input,hidden}} = \begin{pmatrix} 1,0 & 0,5 & 0,2 & 0,4 \\ 1,5 & 2,0 & 0,4 & 2,2 \\ 0,5 & 2,0 & 0,3 & 0,1 \end{pmatrix}$$

$$W_{\text{hidden,Output}} = \begin{pmatrix} 1,5 & 2,0 & 0 \\ 0,5 & 0,4 & 1,0 \end{pmatrix}$$

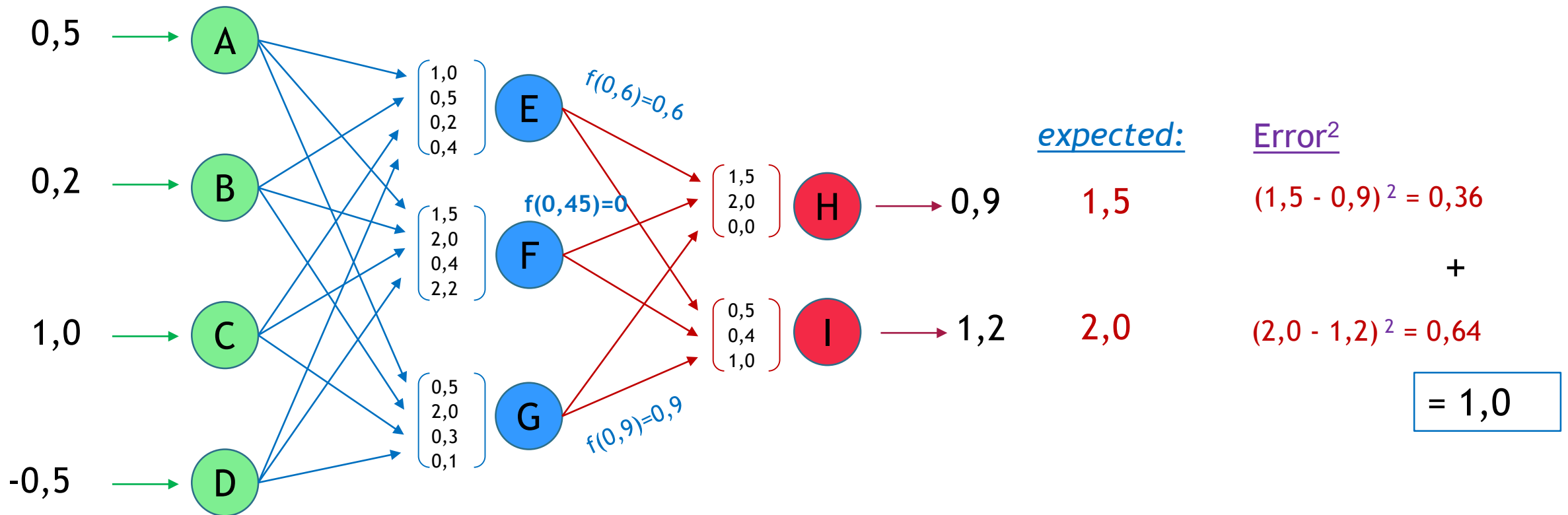
$$X_{\text{hidden}} = W_{\text{Input,hidden}} \cdot I = \begin{pmatrix} 1,0 & 0,5 & 0,2 & 0,4 \\ 1,5 & 2,0 & 0,4 & 2,2 \\ 0,5 & 2,0 & 0,3 & 0,1 \end{pmatrix} \cdot \begin{pmatrix} 0,5 \\ 0,2 \\ 1,0 \\ -0,5 \end{pmatrix} = \begin{pmatrix} 0,6 \\ 0,45 \\ 0,9 \end{pmatrix}$$

$$\text{Out}_{\text{hidden}} = f(X_{\text{hidden}}) = f\left(\begin{pmatrix} 0,6 \\ 0,45 \\ 0,9 \end{pmatrix}\right) = \begin{pmatrix} 0,6 \\ 0 \\ 0,9 \end{pmatrix}$$

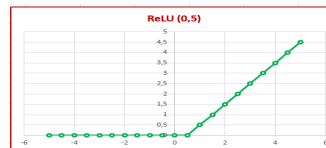
$$X_{\text{Output}} = W_{\text{hidden,Output}} \cdot \text{Out}_{\text{hidden}} = \begin{pmatrix} 1,5 & 2,0 & 0 \\ 0,5 & 0,4 & 1,0 \end{pmatrix} \cdot \begin{pmatrix} 0,6 \\ 0 \\ 0,9 \end{pmatrix} = \begin{pmatrix} 0,9 \\ 1,2 \end{pmatrix}$$

$$\text{Output} = f(X_{\text{Output}}) = f\left(\begin{pmatrix} 0,9 \\ 1,2 \end{pmatrix}\right) = \begin{pmatrix} 0,9 \\ 1,2 \end{pmatrix}$$

Error analysis



Activation function: $f = \text{ReLU} (0.5)$



,no bias (so that the graphic does not become too confusing).

So what now?

- Now we know how to process an input and generate an output from it.
 - We also know the output we would like to have.
 - What can we do so that the generated output matches or at least comes close to the expected one?
-
- This will occupy the topic in my next lecture.
 - Also here on actuvview.

Thank you for
your attention

Let us talk to one another



Dr. Stefan Noertemann

(Aktuar DAV)

Stefan.Noertemann@msg-life.com

Tel.: +49 (0)711 949581201

www.msg-life.com