Welcome to the computational cognitive modelling workshop!

Part 2: Artificial neural networks

Part 2: **Artificial neural networks**

Olivia Guest Chris Brand

Nick Sexton Nicole Cruz De Echeverria Loebell

What is a neural network?

A mathematical model

- Inspired by the nervous system
- ► A set of *units*, connected by *weights*
- ► The network runs by passing activations from the input (to the hidden) to the output units

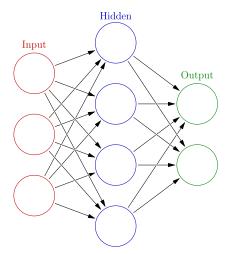


Figure: Glosser.ca / CC-BY-SA-3.0

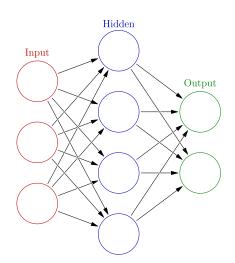
Why use artificial neural networks for modelling?

Some aspects of their behaviour are like their namesake!

- Learn pretty much any input-output data
- Uncover rules on their own about data
- Generalise from what they have learnt
- Cope with noise and damage

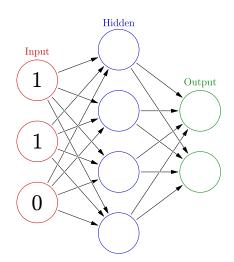
By using maths, predictably!

1. Input units are set to a pattern

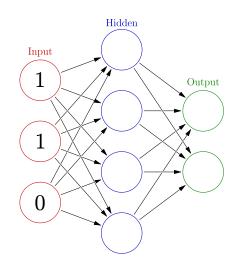


By using maths, predictably!

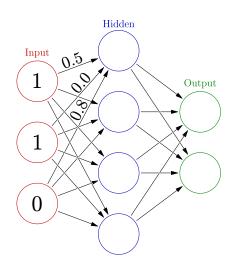
1. Input units are set to a pattern



- 1. Input units are set to a pattern
- 2. Calculate hidden units' states

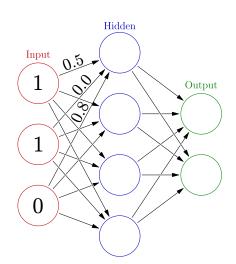


- 1. Input units are set to a pattern
- 2. Calculate hidden units' states



- 1. Input units are set to a pattern
- 2. Calculate hidden units' states:

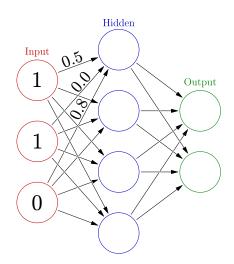
$$1 \times 0.5 = 0.5$$



- 1. Input units are set to a pattern
- 2. Calculate hidden units' states:

$$1 \times 0.5 = 0.5$$

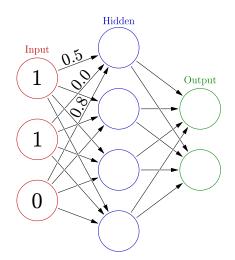
 $1 \times 0.0 = 0.0$



- 1. Input units are set to a pattern
- 2. Calculate hidden units' states:

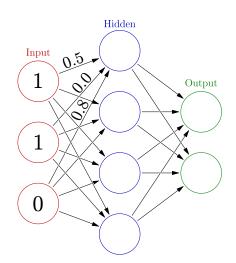
$$1 \times 0.5 = 0.5$$

 $1 \times 0.0 = 0.0$
 $0 \times 0.8 = 0.0$



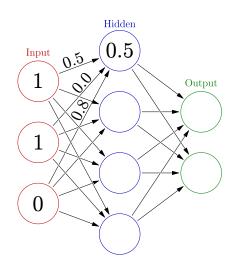
- 1. Input units are set to a pattern
- 2. Calculate hidden units' states:

$$\begin{array}{rcl}
1 \times 0.5 & = & 0.5 \\
1 \times 0.0 & = & 0.0 \\
0 \times 0.8 & = & 0.0 + \\
\hline
0.5
\end{array}$$

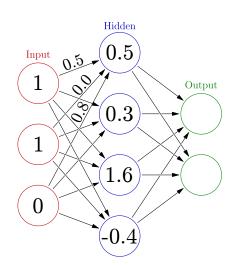


- 1. Input units are set to a pattern
- 2. Calculate hidden units' states:

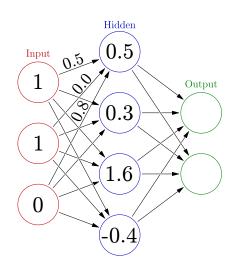
$$\begin{array}{rcl}
1 \times 0.5 & = & 0.5 \\
1 \times 0.0 & = & 0.0 \\
0 \times 0.8 & = & 0.0 + \\
\hline
0.5
\end{array}$$



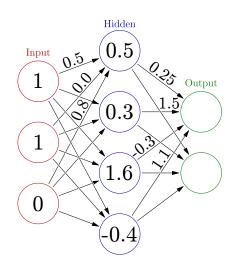
- 1. Input units are set to a pattern
- 2. Calculate hidden units' states



- 1. Input units are set to a pattern
- 2. Calculate hidden units' states
- 3. Same for output units

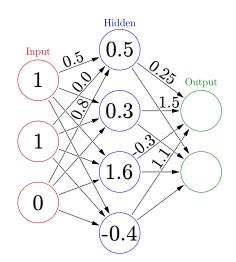


- 1. Input units are set to a pattern
- 2. Calculate hidden units' states
- 3. Same for output units



- 1. Input units are set to a pattern
- 2. Calculate hidden units' states
- 3. Same for output units:

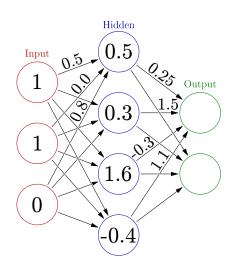
$$0.5 \times 0.25 = 0.125$$



- 1. Input units are set to a pattern
- 2. Calculate hidden units' states
- 3. Same for output units:

$$0.5 \times 0.25 = 0.125$$

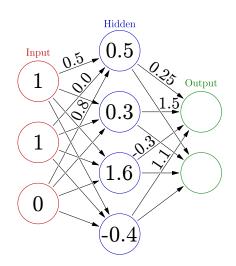
 $0.3 \times 1.5 = 0.45$



- 1. Input units are set to a pattern
- 2. Calculate hidden units' states
- 3. Same for output units:

$$0.5 \times 0.25 = 0.125$$

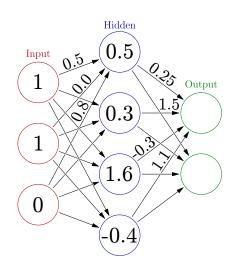
 $0.3 \times 1.5 = 0.45$
 $1.6 \times -0.3 = -0.48$



- 1. Input units are set to a pattern
- 2. Calculate hidden units' states
- 3. Same for output units:

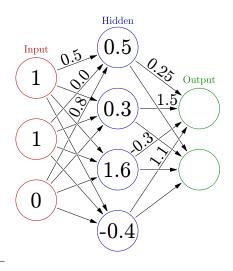
$$0.5 \times 0.25 = 0.125$$

 $0.3 \times 1.5 = 0.45$
 $1.6 \times -0.3 = -0.48$
 $-0.4 \times 1.1 = -0.44$



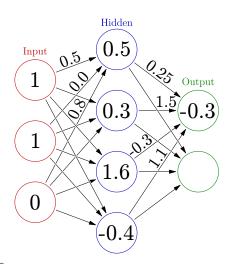
- 1. Input units are set to a pattern
- 2. Calculate hidden units' states
- 3. Same for output units:

$$\begin{array}{rcl}
0.5 \times 0.25 & = & 0.125 \\
0.3 \times 1.5 & = & 0.45 \\
1.6 \times -0.3 & = & -0.48 \\
-0.4 \times 1.1 & = & -0.44 \\
-0.4 \times 1.1 & = & -0.44 + \\
\hline
& & & & & & + \\
\hline
& & & & & & + \\
\end{array}$$

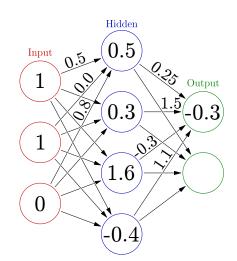


- 1. Input units are set to a pattern
- 2. Calculate hidden units' states
- 3. Same for output units:

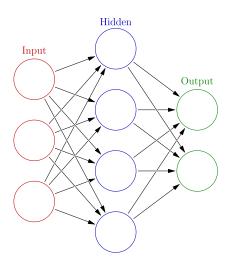
$$\begin{array}{rcl}
0.5 \times 0.25 & = & 0.125 \\
0.3 \times 1.5 & = & 0.45 \\
1.6 \times -0.3 & = & -0.48 \\
-0.4 \times 1.1 & = & -0.44 \\
-0.4 \times 1.1 & = & -0.44 + \\
\hline
& & & & & & + \\
\hline
& & & & & & + \\
\end{array}$$



- 1. Input units are set to a pattern
- 2. Calculate hidden units' states
- 3. Same for output units

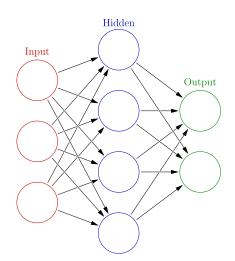


By using maths, predictably!



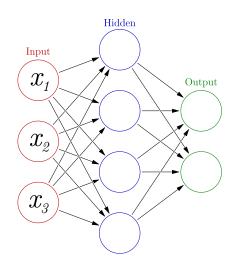
By using maths, predictably!

But we/programmers are lazy:



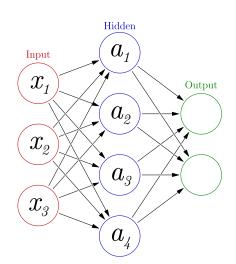
By using maths, predictably!

But we/programmers are lazy:



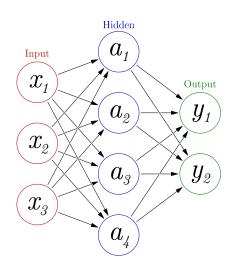
By using maths, predictably!

But we/programmers are lazy:



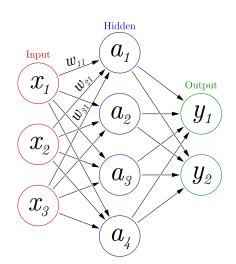
By using maths, predictably!

But we/programmers are lazy:



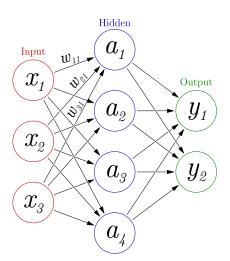
By using maths, predictably!

But we/programmers are lazy:



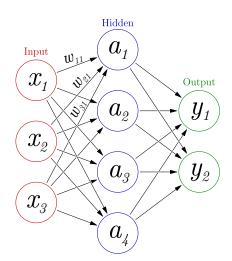
By using maths, predictably!

$$a_i =$$



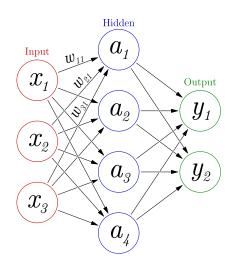
By using maths, predictably!

$$a_i = x_j \times w_{ji}$$



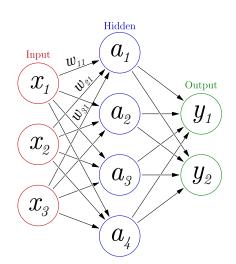
By using maths, predictably!

$$a_i = \sum x_j \times w_{ji}$$



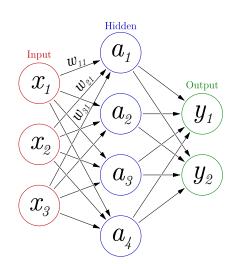
By using maths, predictably!

$$a_i = \sum_{j=1}^{N} x_j \times w_{ji}$$



By using maths, predictably!

$$a_i = f\left(\sum_{j=1}^N x_j \times w_{ji}\right)$$

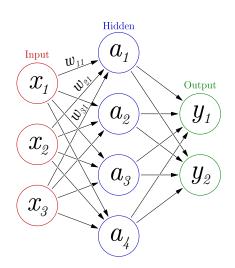


By using maths, predictably!

But we/programmers are lazy:

$$a_i = f\left(\sum_{j=1}^N x_j \times w_{ji}\right)$$

where a_i is the unit whose state we want to calculate, N is the number of units on the previous layer, w_{ji} is the weight on the connection between i and j, and f is a function, commonly the logistic step function.



How do networks learn?

Cunning!

Many options: Hebbian learning, back-propagation of error, Boltzmann machine learning, self-organising map algorithm, etc.

How do networks learn?

Cunning!

- Many options: Hebbian learning, back-propagation of error, Boltzmann machine learning, self-organising map algorithm, etc.
- All learning algorithms work by changing the connection weights

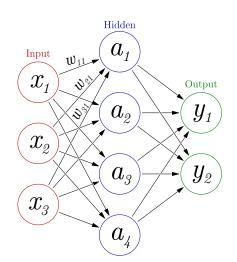
How do networks learn?

Cunning!

- Many options: Hebbian learning, back-propagation of error, Boltzmann machine learning, self-organising map algorithm, etc.
- All learning algorithms work by changing the connection weights
- Learning can be divided into supervised, unsupervised, and reinforcement

A very simple learning rule

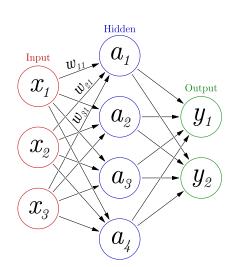
"Cells that fire together, wire together"



A very simple learning rule

"Cells that fire together, wire together"

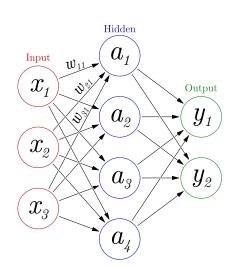
$$w_{ij} = x_i \times a_j$$



A very simple learning rule

"Cells that fire together, wire together"

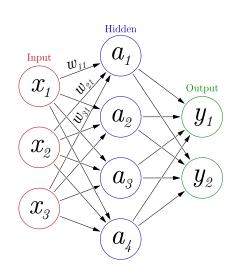
$$w_{ij} = \boldsymbol{\eta} \times x_i \times a_j$$



A very simple learning rule

"Cells that fire together, wire together"

$$\Delta w_{ij} = \boldsymbol{\eta} \times \boldsymbol{x}_i \times \boldsymbol{a}_j$$



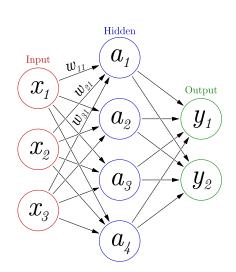
A very simple learning rule

"Cells that fire together, wire together"

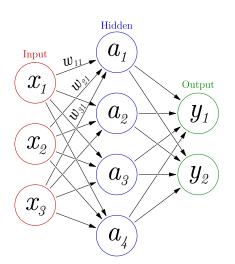
— Carla Shatz

$$\Delta w_{ij} = \eta \times x_i \times a_j$$

which means each weight, w_{ij} is changed by a small in/decrement for every pattern

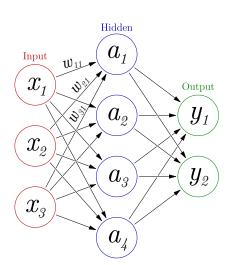


"Cells that fire together, wire together" — Carla Shatz



"Cells that fire together, wire together" — Carla Shatz

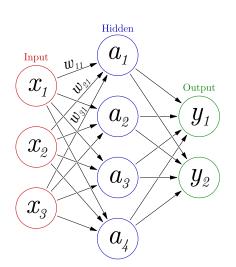
$$\Delta w_{ij} = \boldsymbol{\eta} \times x_i \times a_j$$



"Cells that fire together, wire together" — Carla Shatz

$$\Delta w_{ij} = \boldsymbol{\eta} \times \boldsymbol{x}_i \times \boldsymbol{a}_j$$

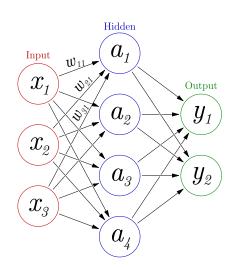
$$\Delta w_{11} = \boldsymbol{\eta} \times x_i \times a_j$$



"Cells that fire together, wire together" — Carla Shatz

$$\Delta w_{ij} = \boldsymbol{\eta} \times x_i \times a_j$$

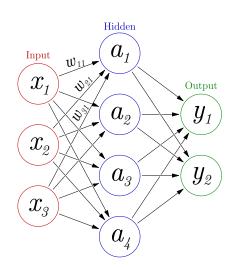
$$\Delta w_{11} = 0.5 \times x_i \times a_j$$



"Cells that fire together, wire together" — Carla Shatz

$$\Delta w_{ij} = \boldsymbol{\eta} \times \boldsymbol{x}_i \times \boldsymbol{a}_j$$

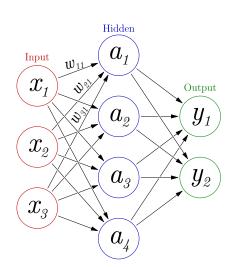
$$\Delta w_{11} = 0.5 \times x_1 \times a_j$$



"Cells that fire together, wire together" — Carla Shatz

$$\Delta w_{ij} = \boldsymbol{\eta} \times \boldsymbol{x}_i \times \boldsymbol{a}_j$$

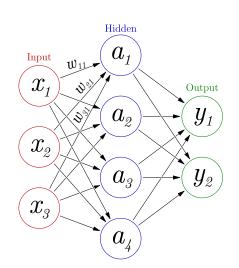
$$\Delta w_{11} = 0.5 \times x_1 \times a_1$$



"Cells that fire together, wire together" — Carla Shatz

$$\Delta w_{ij} = \boldsymbol{\eta} \times x_i \times a_j$$

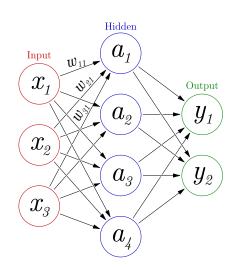
$$\Delta w_{11} = 0.5 \times 1.0 \times a_1$$



"Cells that fire together, wire together" — Carla Shatz

$$\Delta w_{ij} = \boldsymbol{\eta} \times \boldsymbol{x}_i \times \boldsymbol{a}_j$$

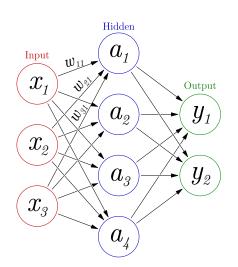
$$\Delta w_{11} = 0.5 \times 1.0 \times 0.3$$



"Cells that fire together, wire together" — Carla Shatz

$$\Delta w_{ij} = \boldsymbol{\eta} \times x_i \times a_j$$

$$\Delta w_{11} = 0.15$$

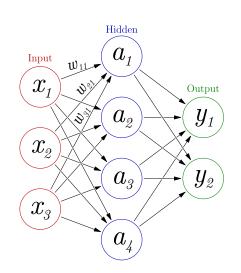


"Cells that fire together, wire together" — Carla Shatz

$$\Delta w_{ij} = \boldsymbol{\eta} \times x_i \times a_j$$

$$\Delta w_{11} = 0.15$$

new
$$w_{11} =$$
old $w_{11} + \Delta w_{11}$

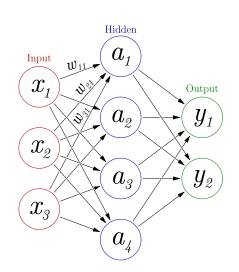


"Cells that fire together, wire together" — Carla Shatz

$$\Delta w_{ij} = \boldsymbol{\eta} \times x_i \times a_j$$

$$\Delta w_{11} = 0.15$$

new
$$w_{11} = 0.0 + \Delta w_{11}$$

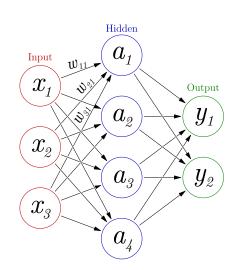


"Cells that fire together, wire together" — Carla Shatz

$$\Delta w_{ij} = \boldsymbol{\eta} \times \boldsymbol{x}_i \times \boldsymbol{a}_j$$

$$\Delta w_{11} = 0.15$$

new
$$w_{11} = 0.0 + 0.15$$

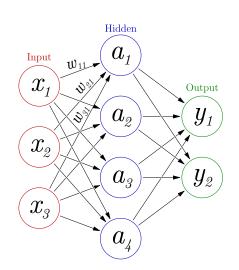


"Cells that fire together, wire together" — Carla Shatz

$$\Delta w_{ij} = \boldsymbol{\eta} \times x_i \times a_j$$

$$\Delta w_{11} = 0.15$$

new
$$w_{11} = 0.15$$

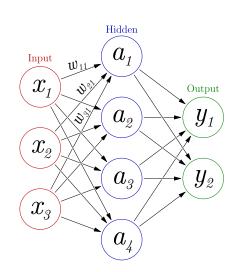


"Cells that fire together, wire together" — Carla Shatz

$$\Delta w_{ij} = \boldsymbol{\eta} \times x_i \times a_j$$

$$\Delta w_{11} = 0.15$$

$$w_{11} = 0.15$$



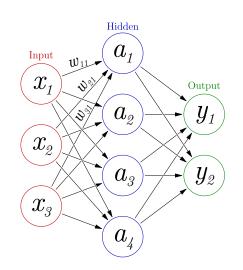
"Cells that fire together, wire together" — Carla Shatz

Hebb's rule is simple, but *very unstable*!

$$\Delta w_{ij} = \boldsymbol{\eta} \times x_i \times a_j$$

$$\Delta w_{11} = 0.15$$

 $w_{11} = 0.15 + \text{something positive}$



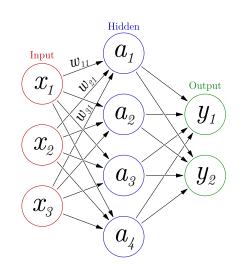
"Cells that fire together, wire together" — Carla Shatz

Hebb's rule is simple, but *very unstable*!

$$\Delta w_{ij} = \boldsymbol{\eta} \times x_i \times a_j$$

$$\Delta w_{11} = 0.15$$

 $w_{11} = 0.15 + \text{something positive} + \text{something else positive} +$



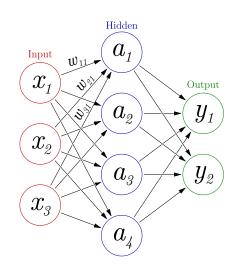
"Cells that fire together, wire together" — Carla Shatz

Hebb's rule is simple, but *very unstable*!

$$\Delta w_{ij} = \boldsymbol{\eta} \times x_i \times a_j$$

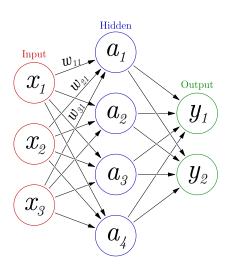
$$\Delta w_{11} = 0.15$$

 $w_{11} = 0.15 + \text{something positive} + \text{something else positive} + \text{another positive value} + \dots$

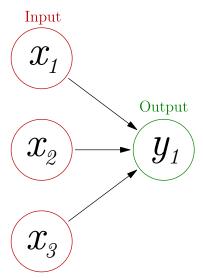


"Cells that fire together, wire together" — Carla Shatz

$$\Delta w_{ij} = \boldsymbol{\eta} \times x_i \times a_j$$

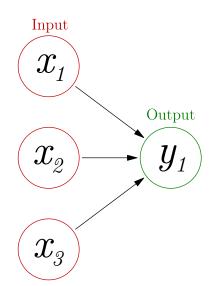


A simple classifier



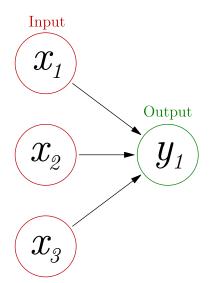
A simple classifier

 Created in 1957 at the Cornell Aeronautical Laboratory by Frank Rosenblatt



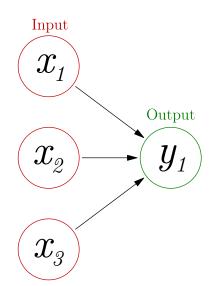
A simple classifier

- Created in 1957 at the Cornell Aeronautical Laboratory by Frank Rosenblatt
- Linear classifier



A simple classifier

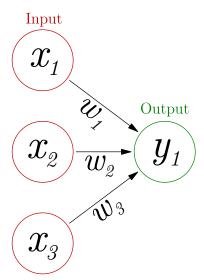
- Created in 1957 at the Cornell Aeronautical Laboratory by Frank Rosenblatt
- Linear classifier
- Simplest form of feedforward network



How does the perceptron learn?

Maths again!

1. Initialise weights

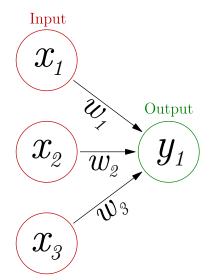


Maths again!

- 1. Initialise weights
- 2. Run network using:

$$y_j = f\left(\sum_{1}^{N} w_i \times x_i\right)$$

same as always!

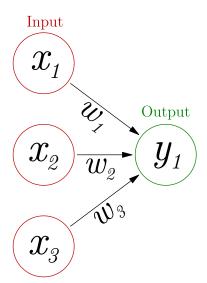


Maths again!

- 1. Initialise weights
- 2. Run network
- 3. Update weights using:

$$\Delta w_i = \eta (d_i - y_i) \times x_i$$

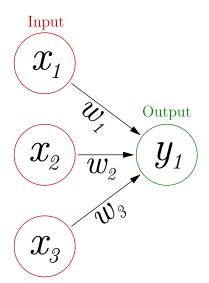
where d is what we want y to be given x_i , and η is the learning rate.



How does the perceptron learn?

Maths again!

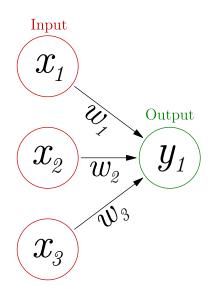
- 1. Initialise weights
- 2. Run network
- 3. Update weights
- 4. Repeat 2 and 3



How does the perceptron learn?

Maths again!

- 1. Initialise weights
- 2. Run network
- 3. Update weights
- 4. Repeat 2 and 3
- 5. When do we stop?



The end

Time to program a perceptron!