

Welcome to the introduction  
to computational cognitive  
modelling workshop!

## **Part 2: Introduction to artificial neural networks**

# Part 2: Introduction to artificial neural networks

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*Chris Brand*

*Nick Sexton*

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# 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

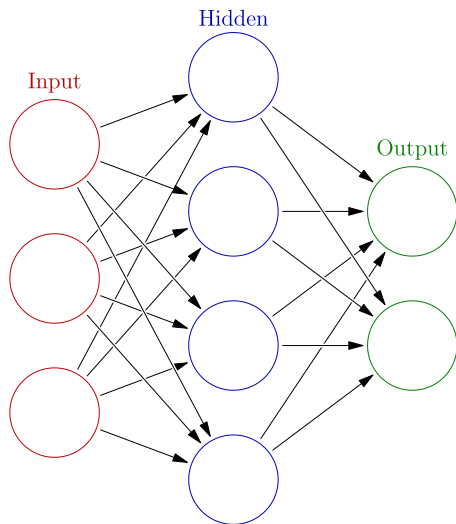


Figure: Glosser.ca / CC-BY-SA-3.0

# Why use artificial neural networks for modelling?

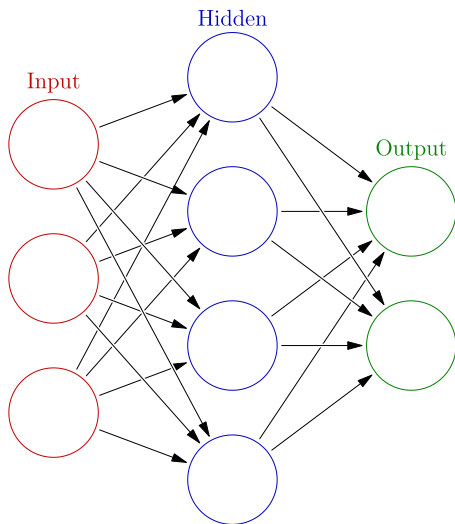
Some aspects of their behaviour are like their namesake!

- ▶ 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

# How does an artificial neural network run?

By using maths, predictably!

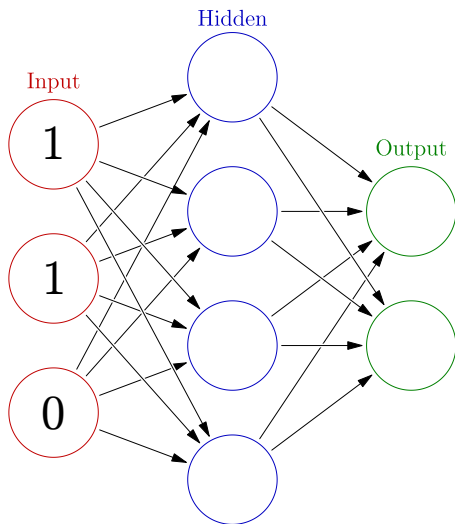
1. **Input units** are set to a *pattern*



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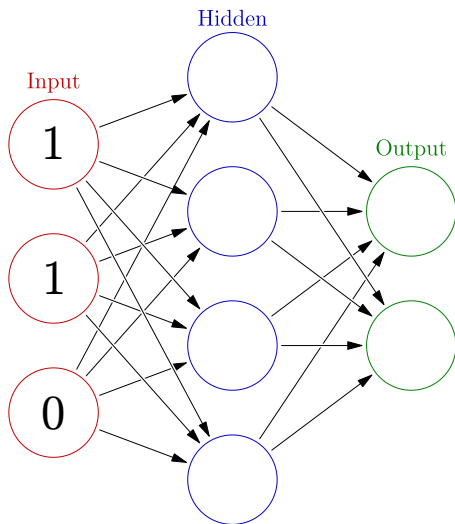
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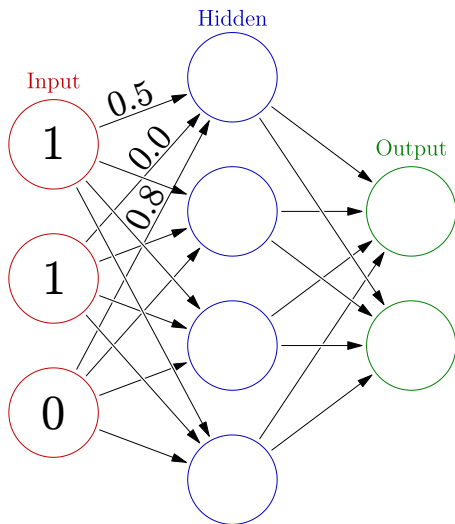
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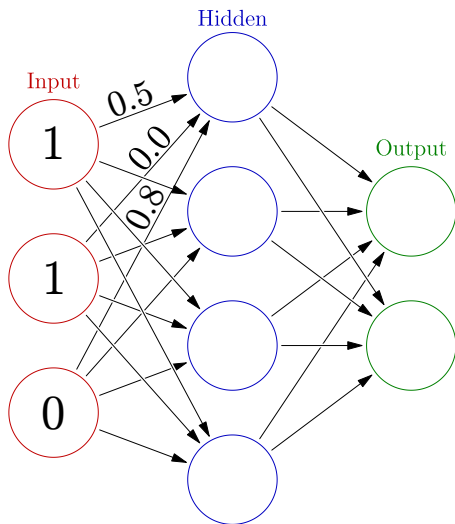


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$$1 \times 0.5 = 0.5$$



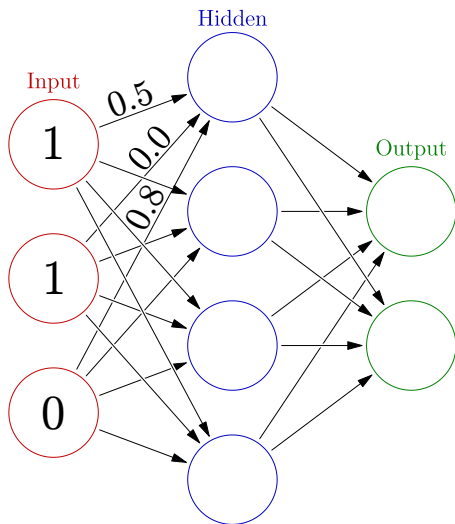
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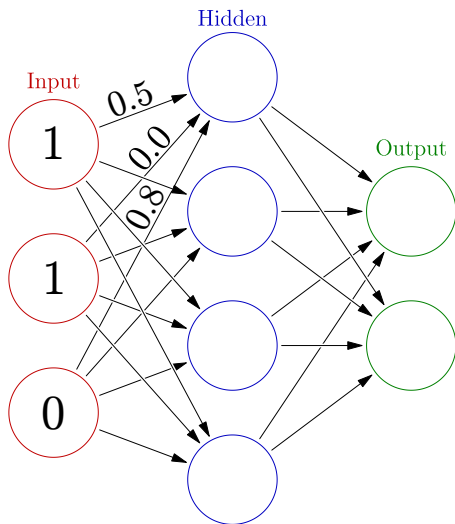
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$$1 \times 0.5 = 0.5$$

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$$0 \times 0.8 = 0.0$$

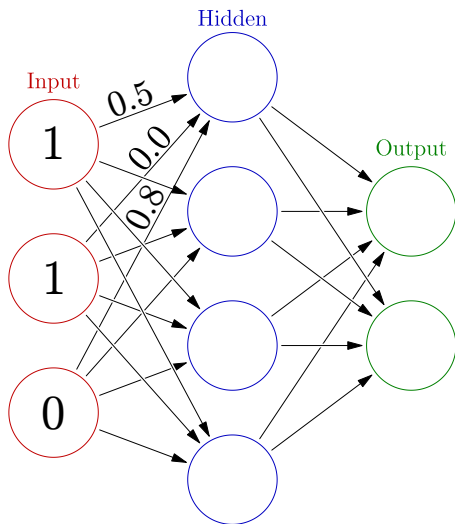


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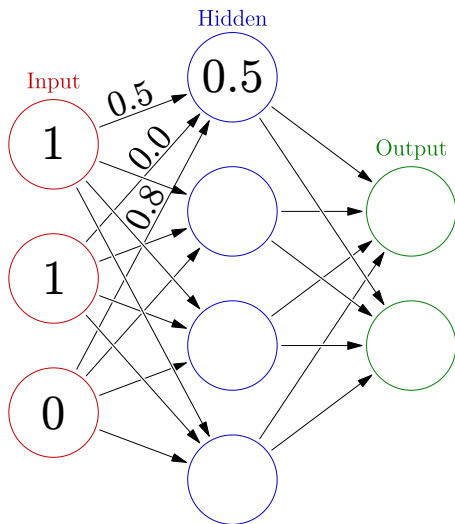


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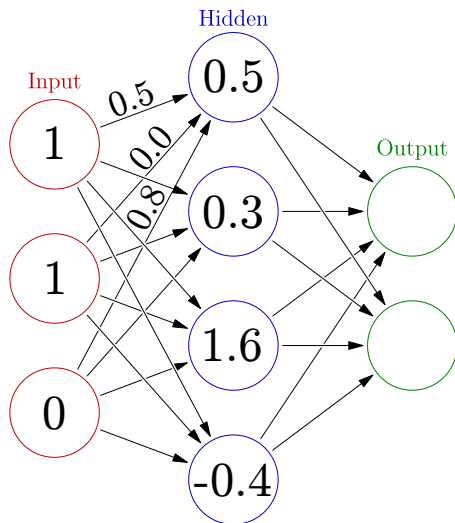
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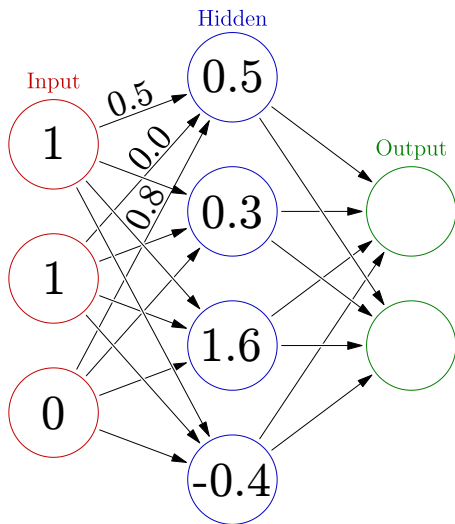
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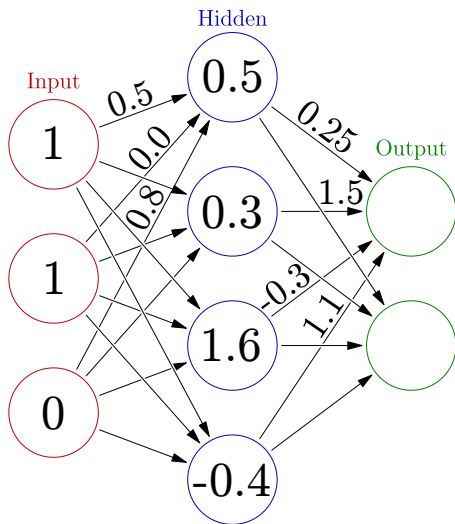
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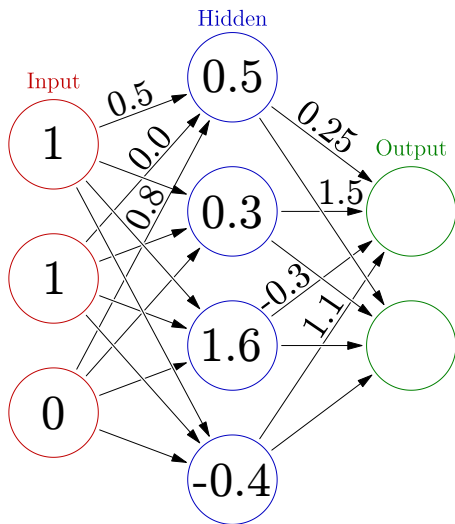


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$$0.5 \times 0.25 = 0.125$$

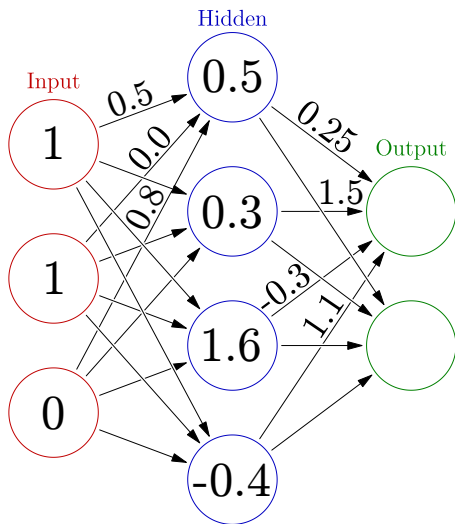


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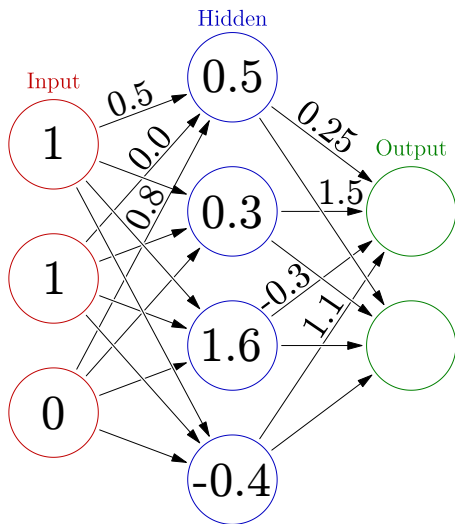


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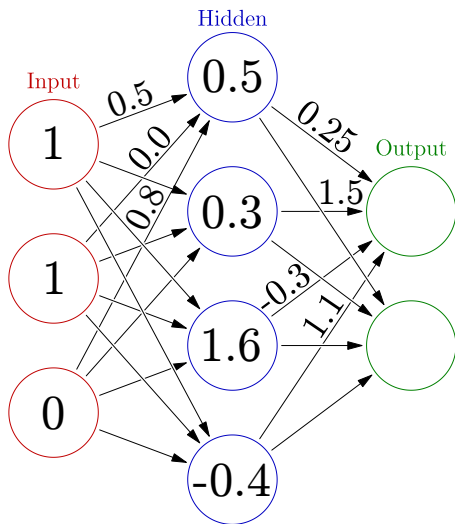


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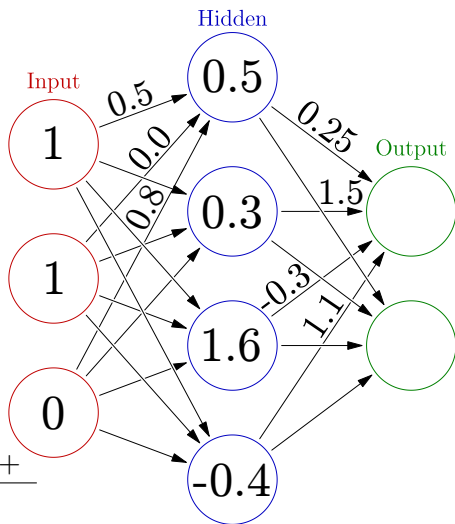


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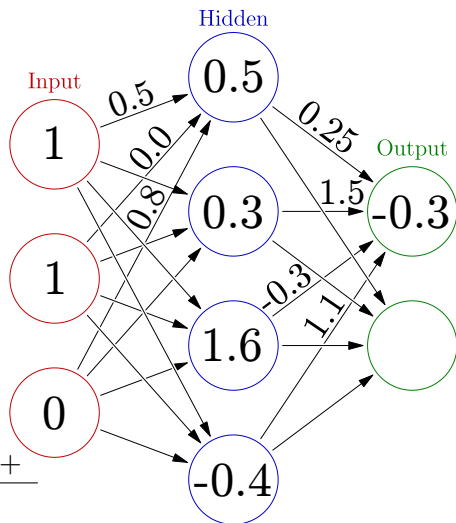


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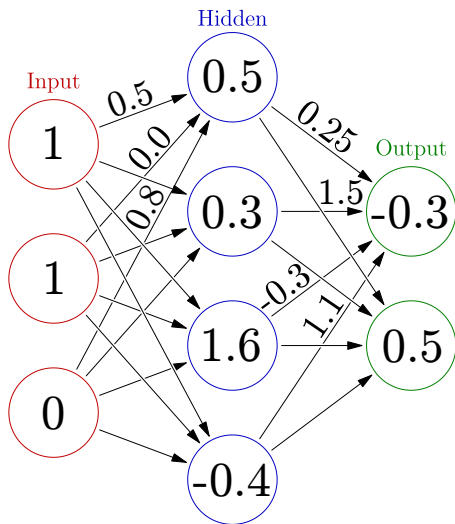
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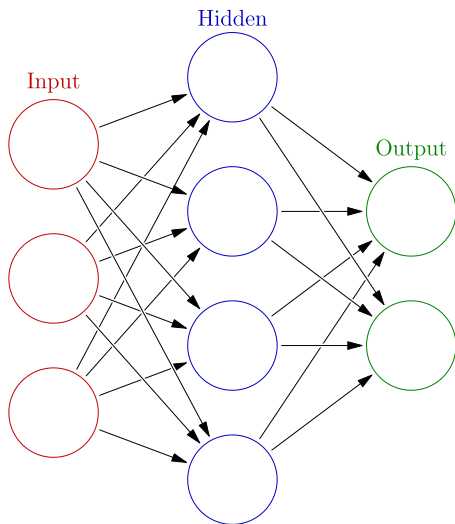
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- ▶ But programmers are *lazy*!

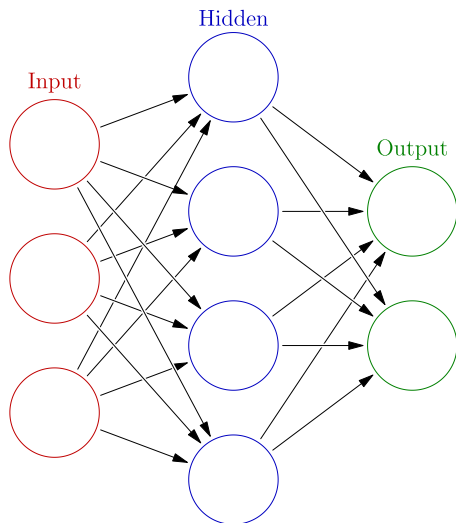




# How does an artificial neural network run?

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- General names save time

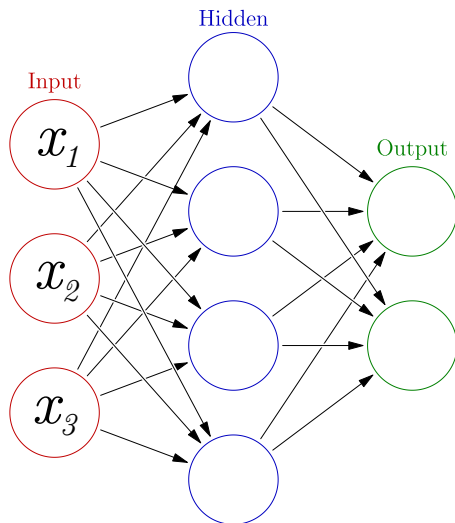


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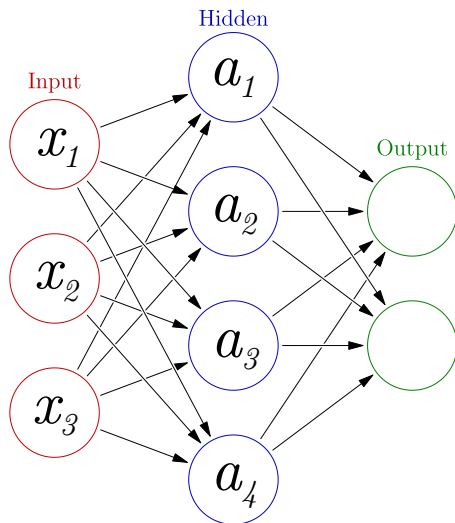
- ▶ **input units:**  $x_i$



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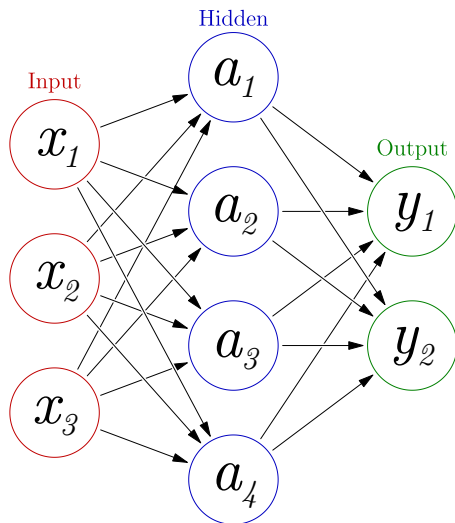
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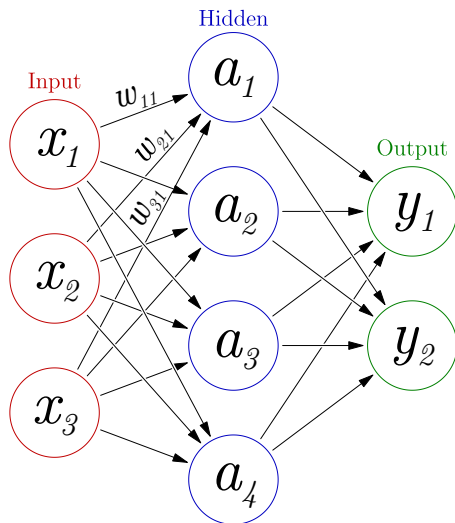
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- ▶ connection weights:  $w_{ij}$



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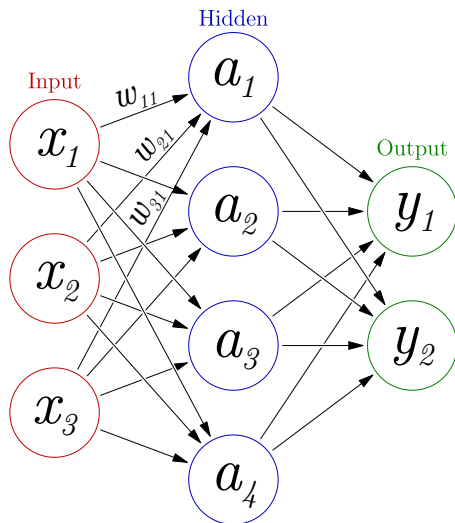
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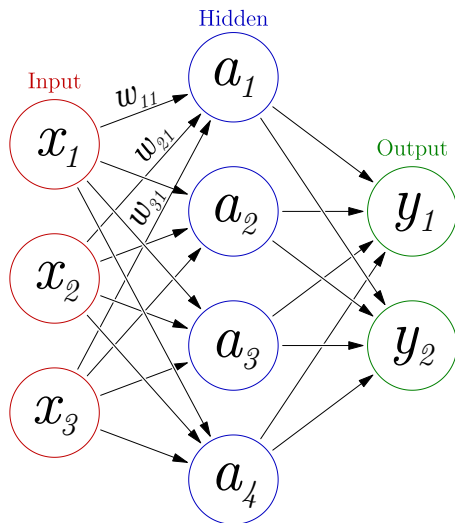
- ▶ subscripts  
general:  $ijklm\dots$   
specific: 12345...



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We use general names to write a general equation:

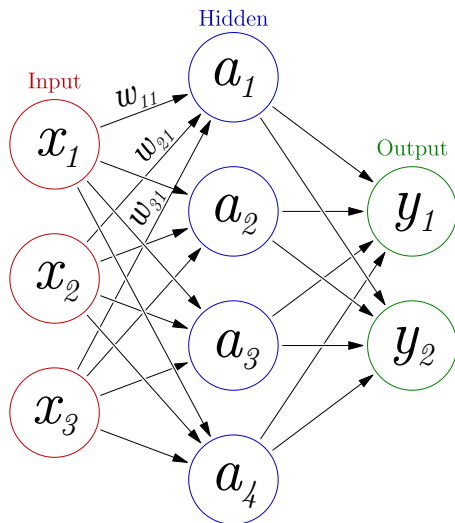


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We use general names to write a general equation:

$$a_i =$$



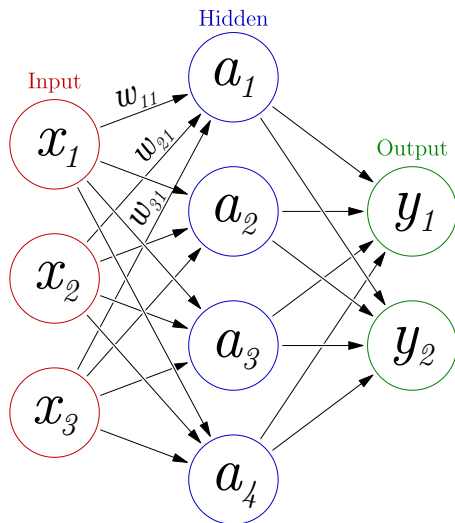


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We use general names to write a general equation:

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

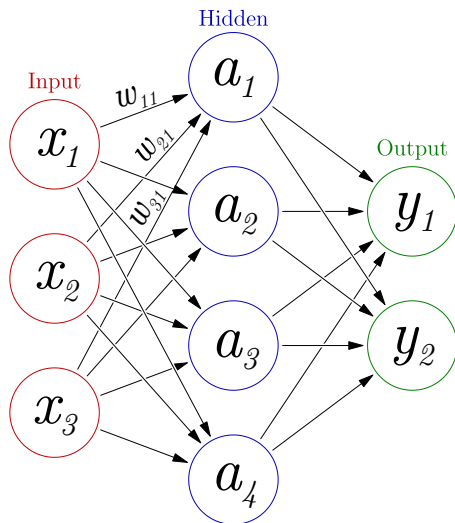


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We use general names to write a general equation:

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

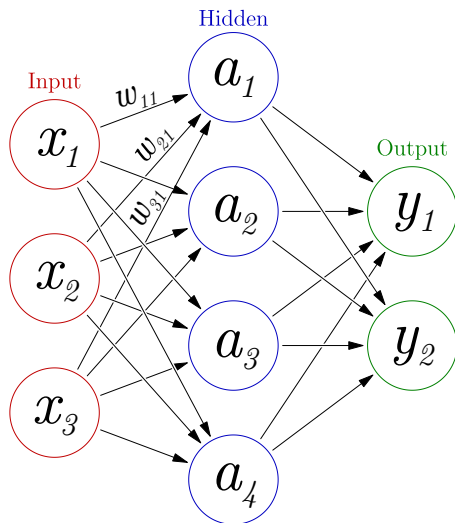


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$$a_i = f \left( \sum_{j=1}^N x_j \times w_{ji} \right)$$

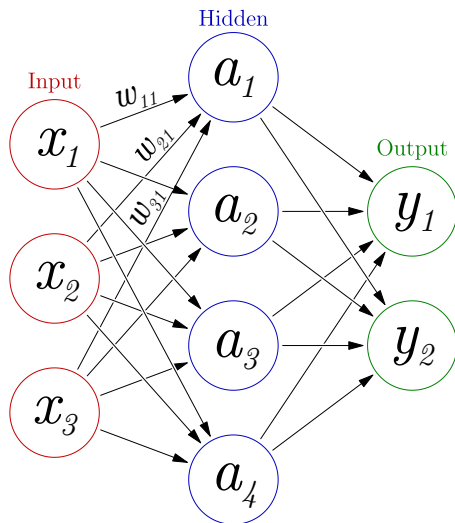


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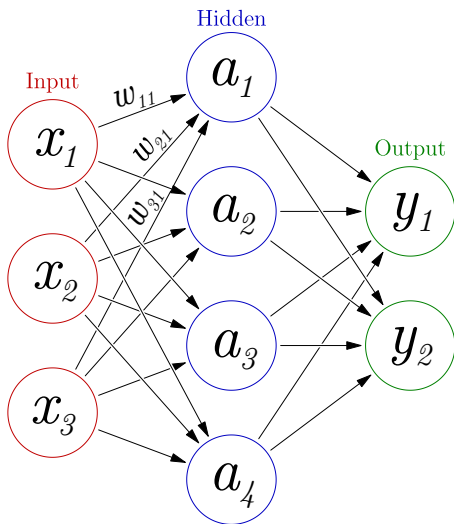
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where  $a_i$  is the unit whose state we want to calculate,  $N$  is the number of units on the previous layer,  $w_{ji}$  is the weight on the connection between  $i$  and  $j$ , and  $f$  is a function that the unit applies.

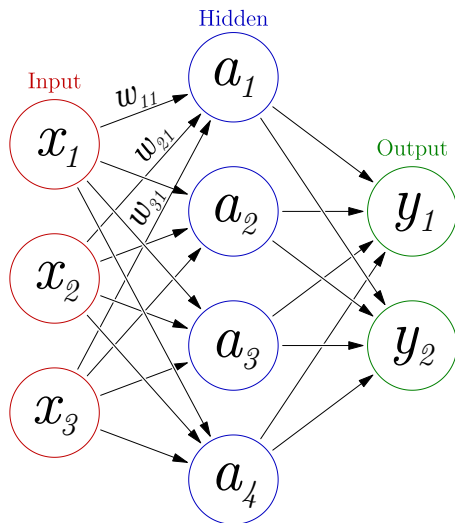


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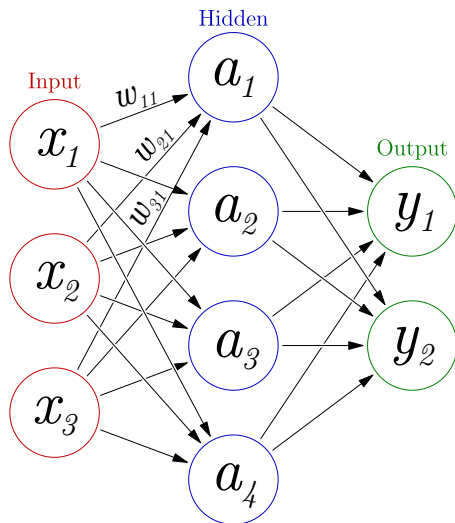
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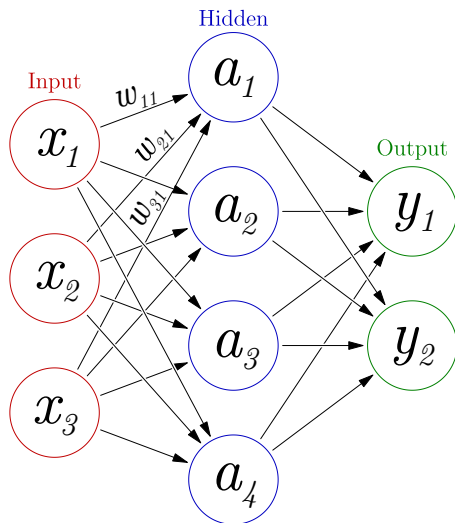
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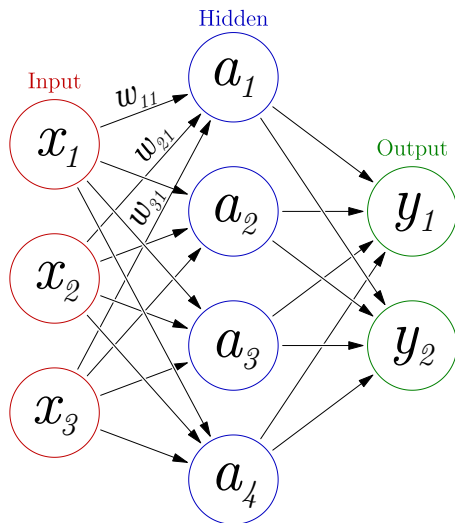
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$$x_{\textcolor{red}{1}} \times w_{\textcolor{blue}{1}\textcolor{blue}{1}}$$



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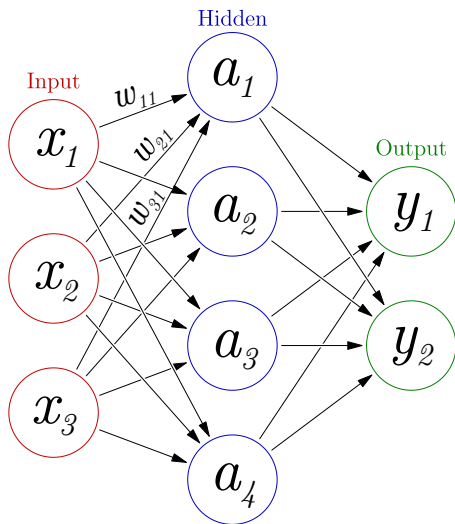
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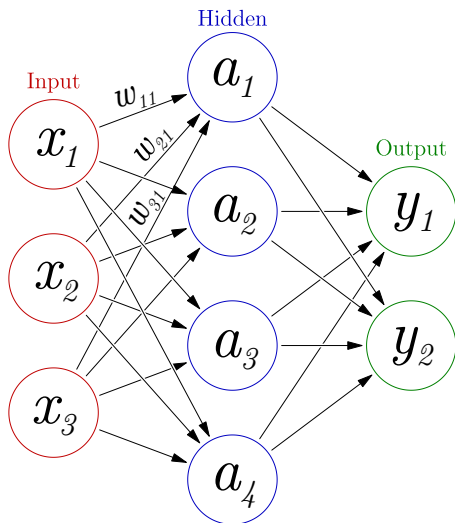
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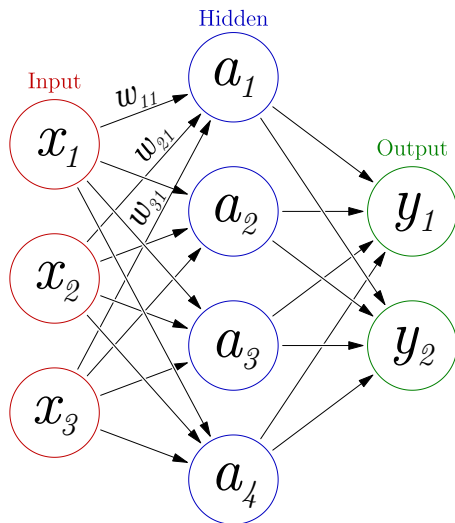
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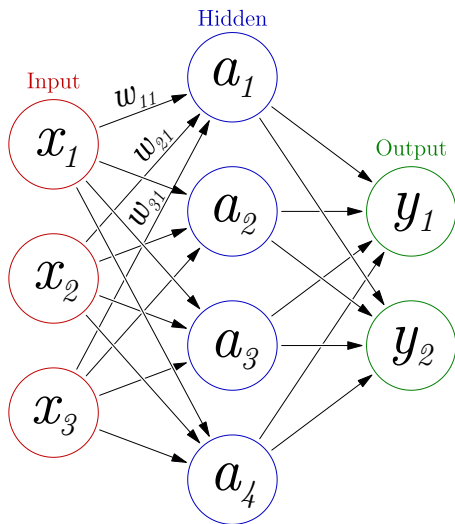
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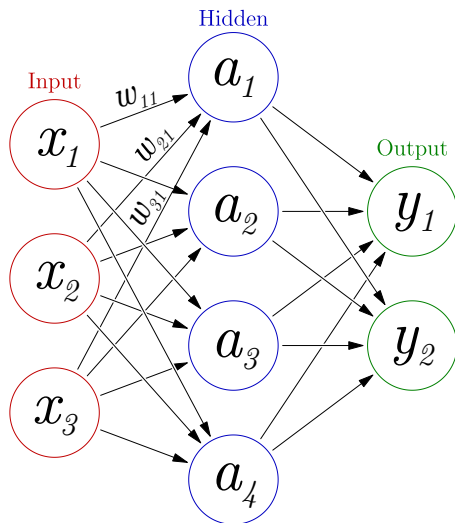
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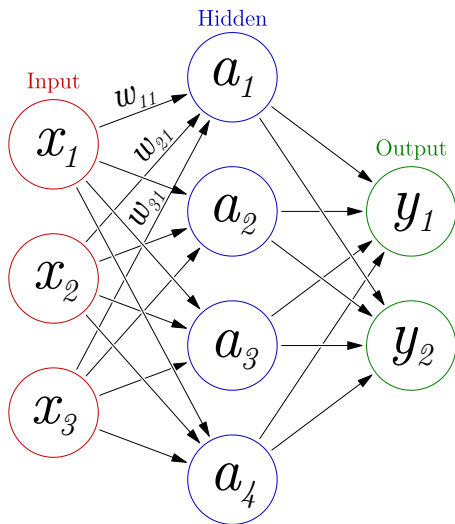
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Cunning!

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



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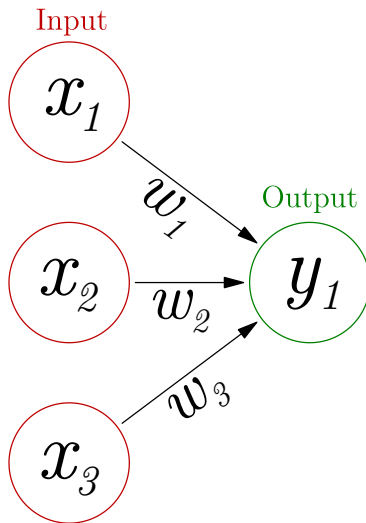
- ▶ Many options: Hebbian learning, back-propagation of error, Boltzmann machine learning, self-organising map algorithm, etc.
- ▶ All learning algorithms work by changing the connection weights
- ▶ Learning can be divided into *supervised*, *unsupervised*, and *reinforcement*

# Hebbian learning

A very simple learning rule

“Cells that fire together, wire together”

— Carla Shatz



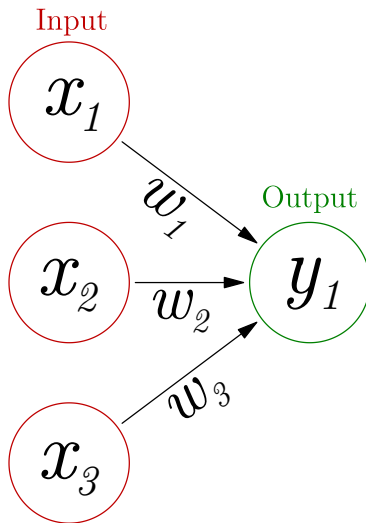
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$w_i =$



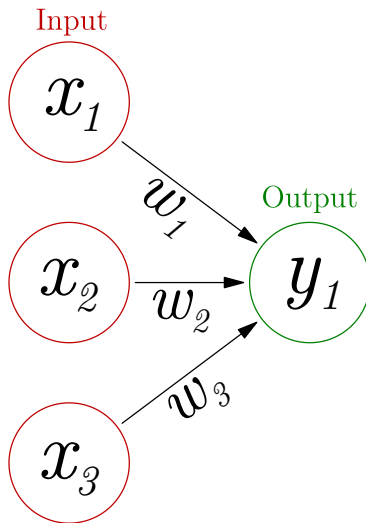
# Hebbian learning

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$$w_i = x_i \times y_j$$



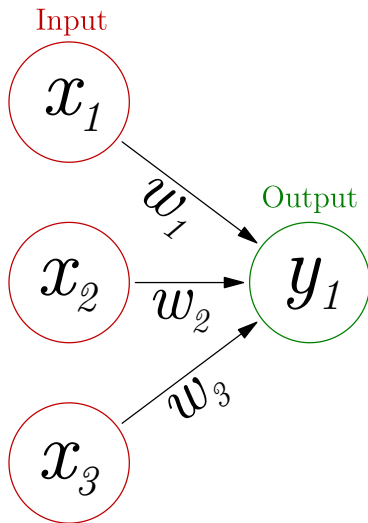
# Hebbian learning

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$$w_i = \eta \times x_i \times y_j$$



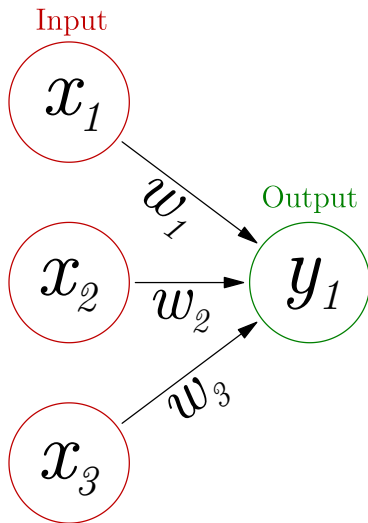
# Hebbian learning

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$$\Delta w_i = \eta \times \textcolor{red}{x}_i \times \textcolor{green}{y}_j$$



# Hebbian learning

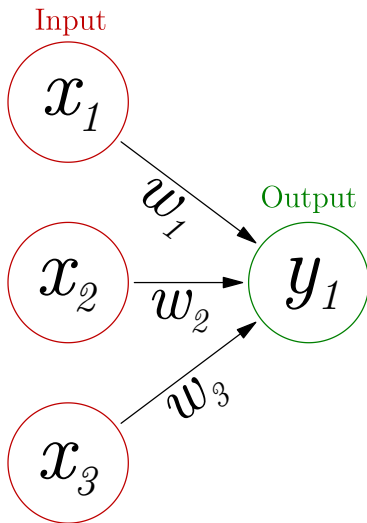
## A very simple learning rule

“Cells that fire together, wire together”

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$$\Delta w_i = \eta \times x_i \times y_j$$

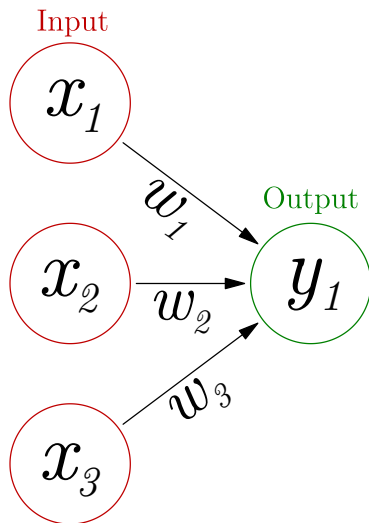
which means each weight is changed by a small in/decrement for every pattern





# Hebbian learning

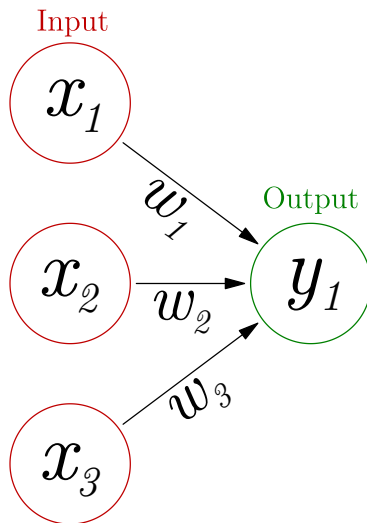
Hebb's rule is simple, but *very unstable!*



# Hebbian learning

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

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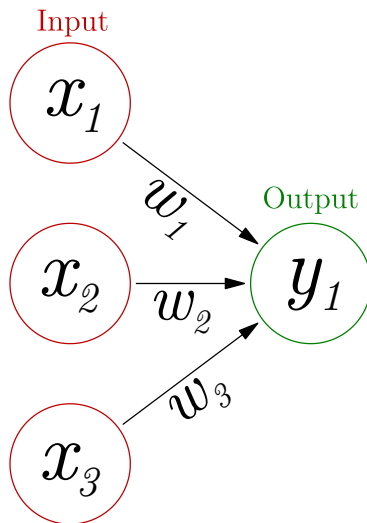


# Hebbian learning

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$$\Delta w_i = \eta \times x_i \times y_j$$

$$\Delta w_1 = \eta \times x_i \times y_j$$

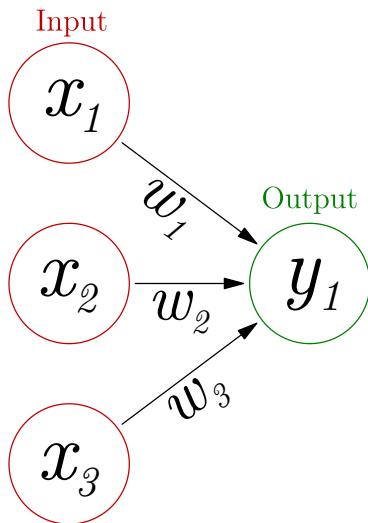


# Hebbian learning

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

$$\Delta w_i = \eta \times x_i \times y_j$$

$$\Delta w_1 = 0.5 \times x_i \times y_j$$

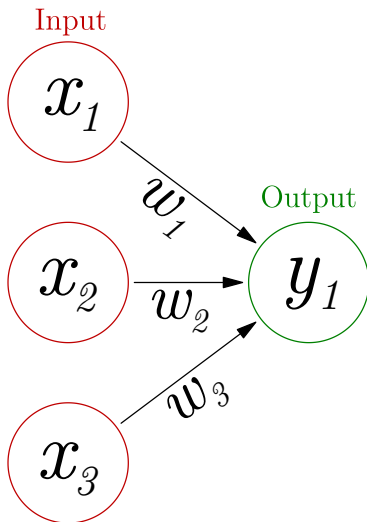


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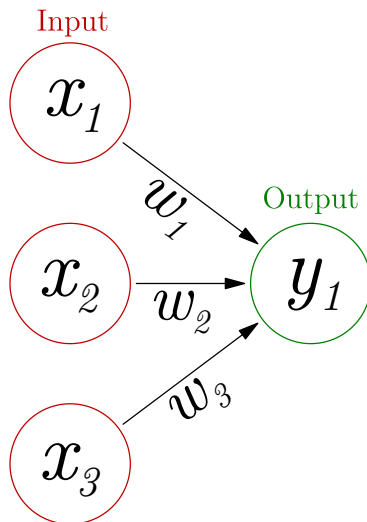


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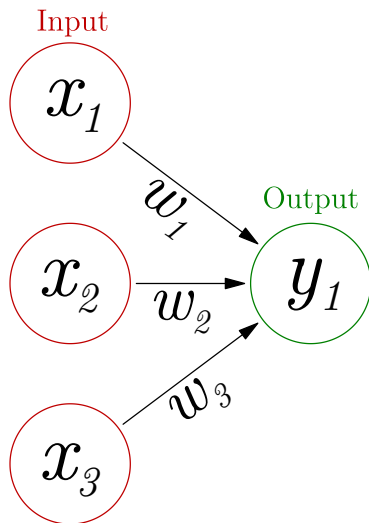


# Hebbian learning

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$$\Delta w_i = \eta \times x_i \times y_j$$

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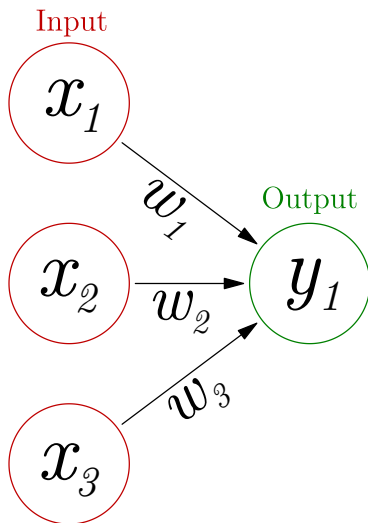


# Hebbian learning

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

$$\Delta w_i = \eta \times x_i \times y_j$$

$$\Delta w_1 = 0.5 \times 1.0 \times 0.3$$



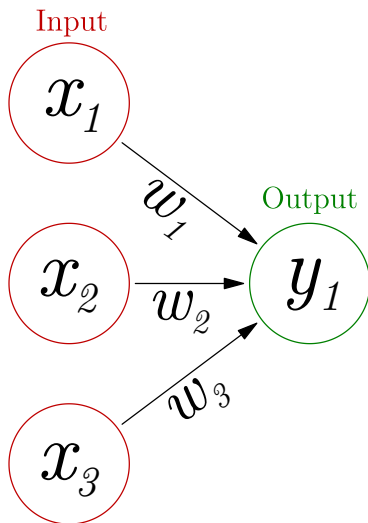


# Hebbian learning

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

$$\Delta w_i = \eta \times x_i \times y_j$$

$$\Delta w_1 = 0.15$$



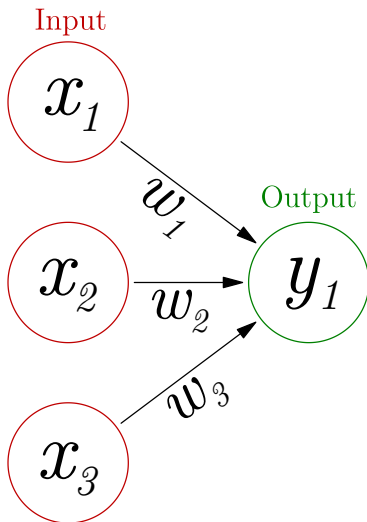
# Hebbian learning

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

$$\Delta w_i = \eta \times x_i \times y_j$$

$$\Delta w_1 = 0.15$$

$$\mathbf{new} \ w_1 = \mathbf{old} \ w_1 + \Delta w_1$$



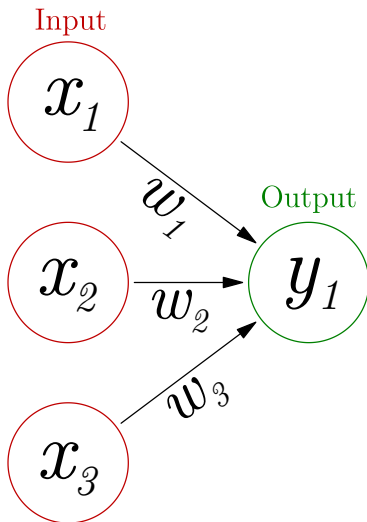
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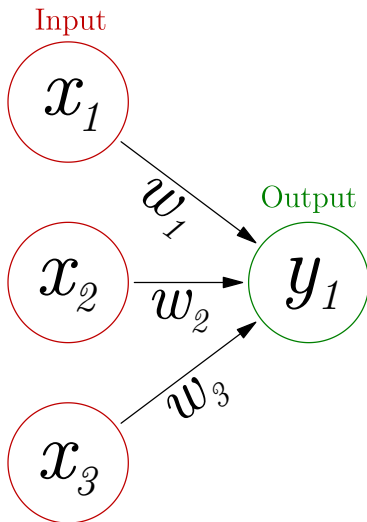
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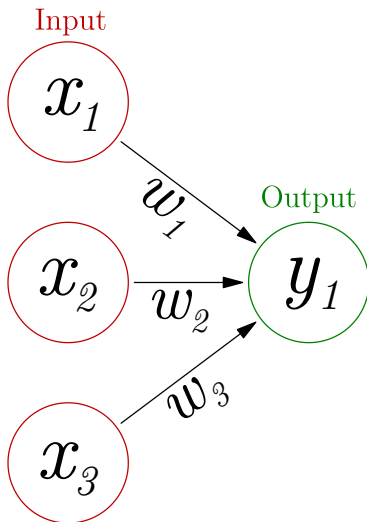
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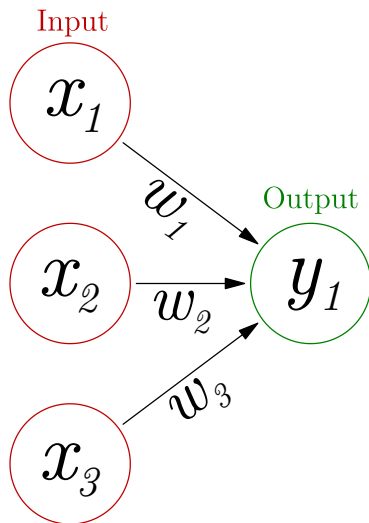
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$$w_1 = 0.15$$



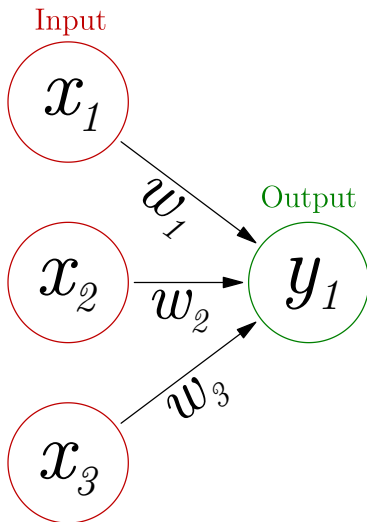
# Hebbian learning

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

$$\Delta w_i = \eta \times x_i \times y_j$$

$$\Delta w_1 = 0.15$$

$w_1 = 0.15 + \text{something}$   
positive



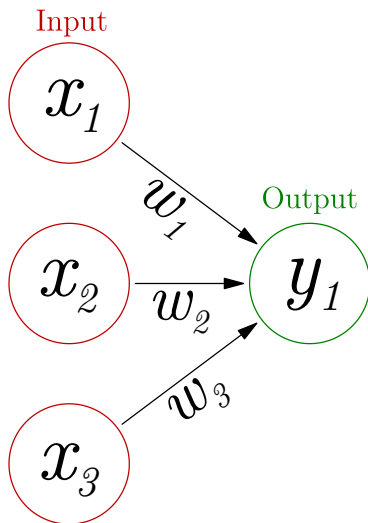
# Hebbian learning

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

$$\Delta w_i = \eta \times x_i \times y_j$$

$$\Delta w_1 = 0.15$$

$w_1 = 0.15 + \text{something}$   
positive + something else  
positive +





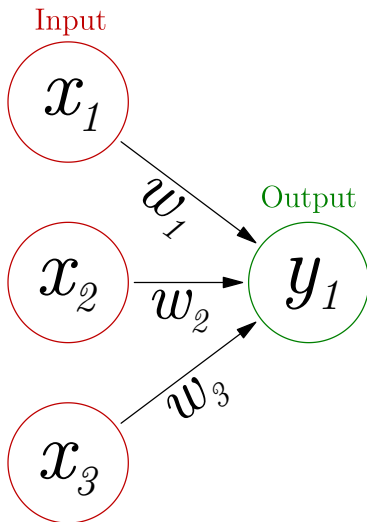
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Hebb's rule is simple, but *very unstable!*

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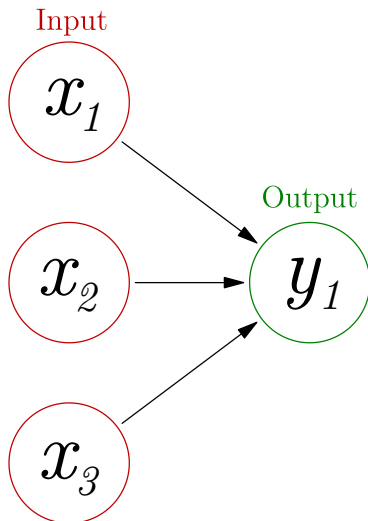
$$\Delta w_1 = 0.15$$

$w_1 = 0.15 + \text{something}$   
positive + something else  
positive + another positive  
value + ...



# The perceptron

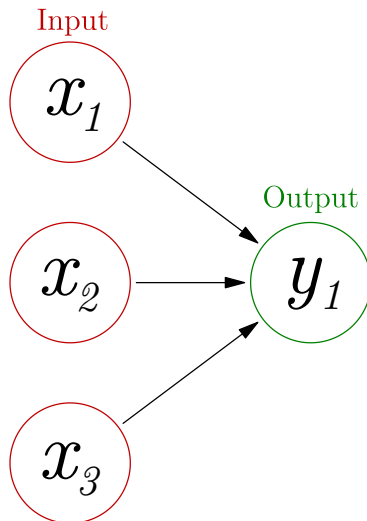
A simple classifier



# The perceptron

A simple classifier

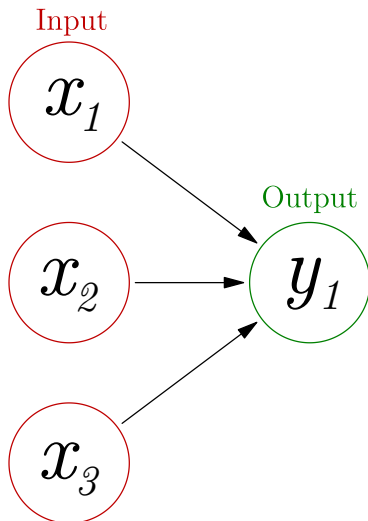
- Created in 1957 at the Cornell Aeronautical Laboratory by Frank Rosenblatt



# The perceptron

## A simple classifier

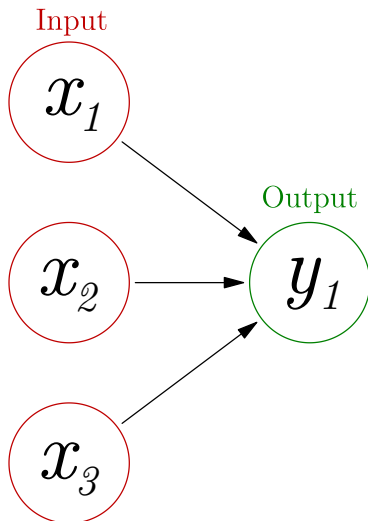
- ▶ Created in 1957 at the Cornell Aeronautical Laboratory by Frank Rosenblatt
- ▶ Linear classifier



# The perceptron

## A simple classifier

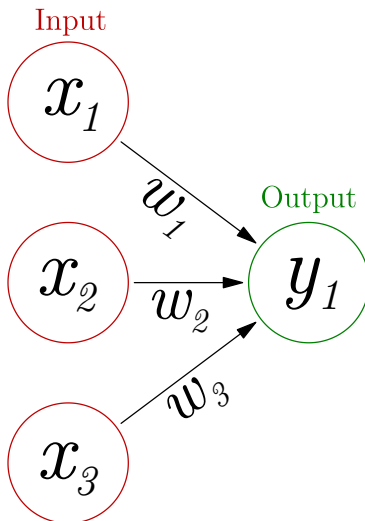
- ▶ Created in 1957 at the Cornell Aeronautical Laboratory by Frank Rosenblatt
- ▶ Linear classifier
- ▶ Simplest form of feedforward network



# How does the perceptron learn?

Maths again!

1. Initialise weights



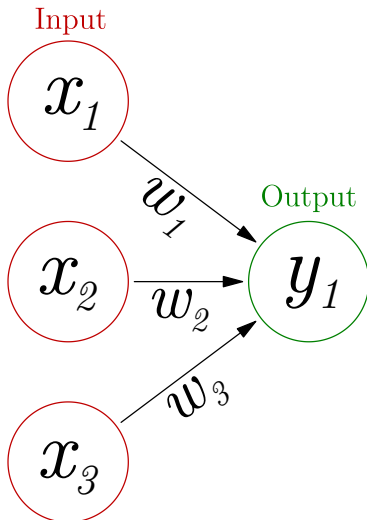
# How does the perceptron learn?

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!

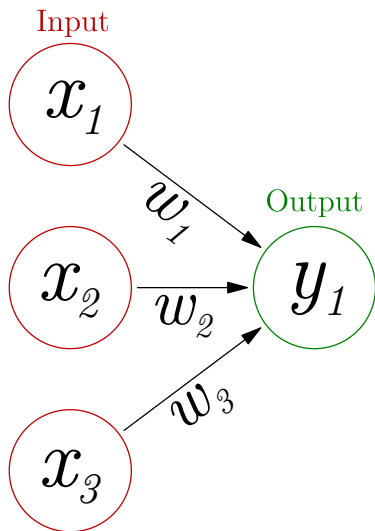


# How does the perceptron learn?

Maths again!

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

$$\Delta w_i = \eta \quad y_j \times x_i$$



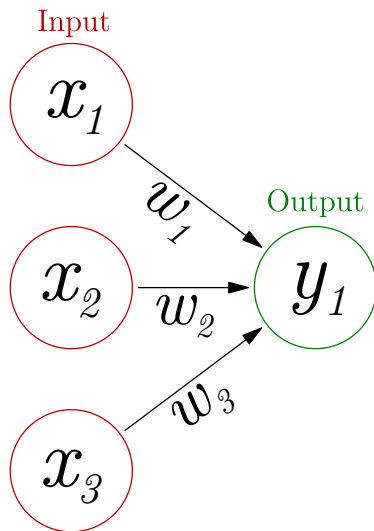


# How does the perceptron learn?

Maths again!

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

$$\Delta w_i = \eta (d_j - y_j) \times x_i$$



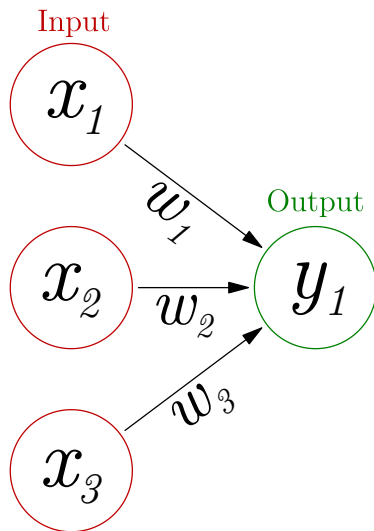
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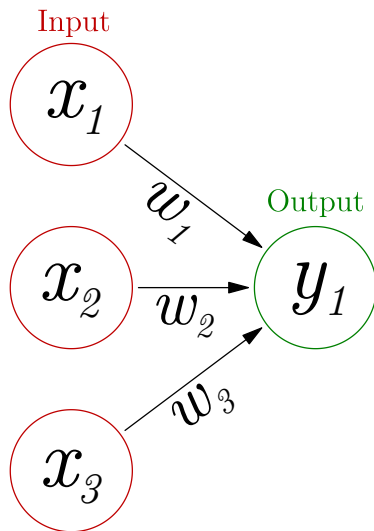
where  $d$  is what we want  $y$  to be given  $x$ , and  $\eta$  is the learning rate.



# How does the perceptron learn?

Maths again!

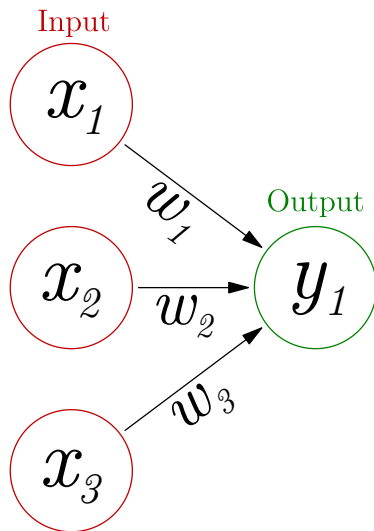
1. Initialise weights
2. Run network
3. Update weights
4. Repeat 2 and 3



# 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!