

# Deep Learning

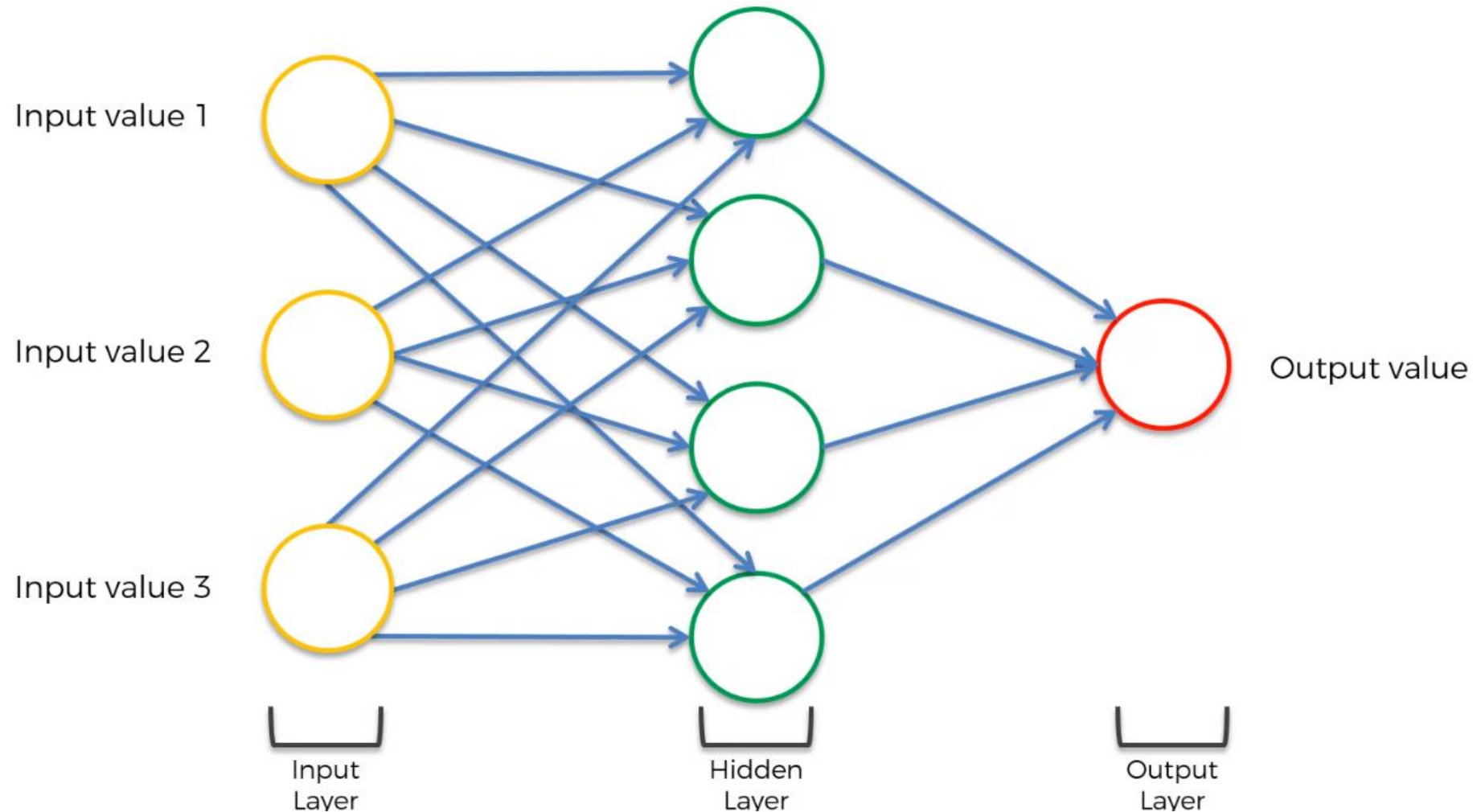
# Summary

- Overview
- The Neuron
- The Activation Function
- How do Neural Networks work? (example)
- How do Neural Networks learn?
- Gradient Descent
- Stochastic Gradient Descent
- Backpropagation
- Algorithm summary

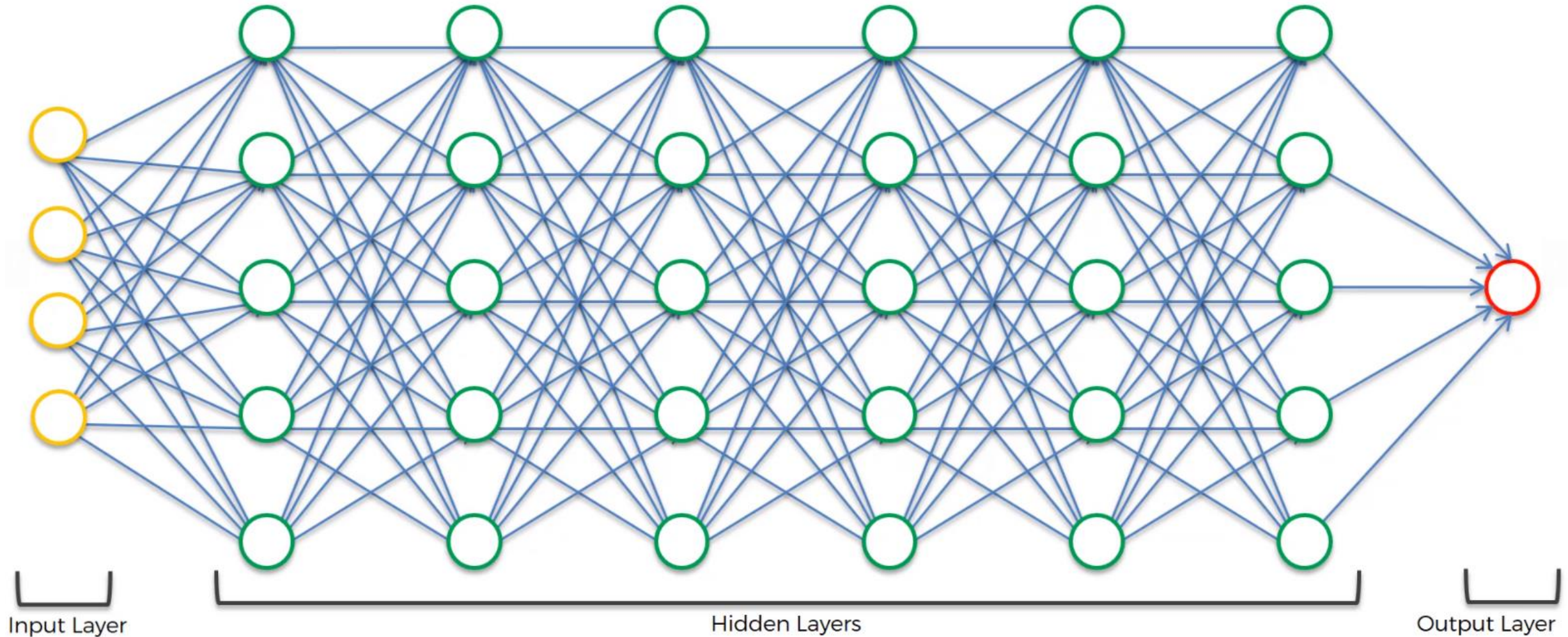
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# Overview



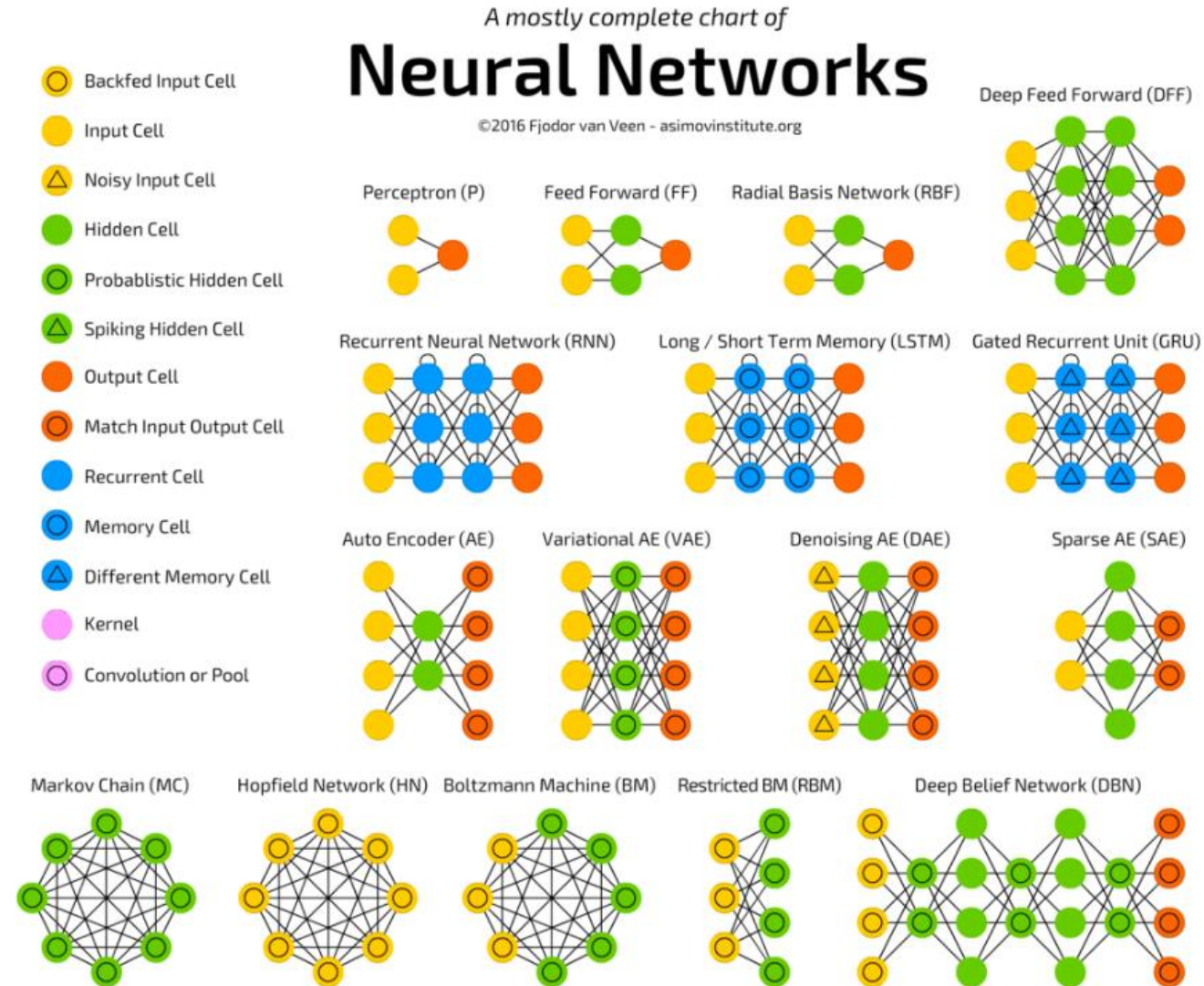
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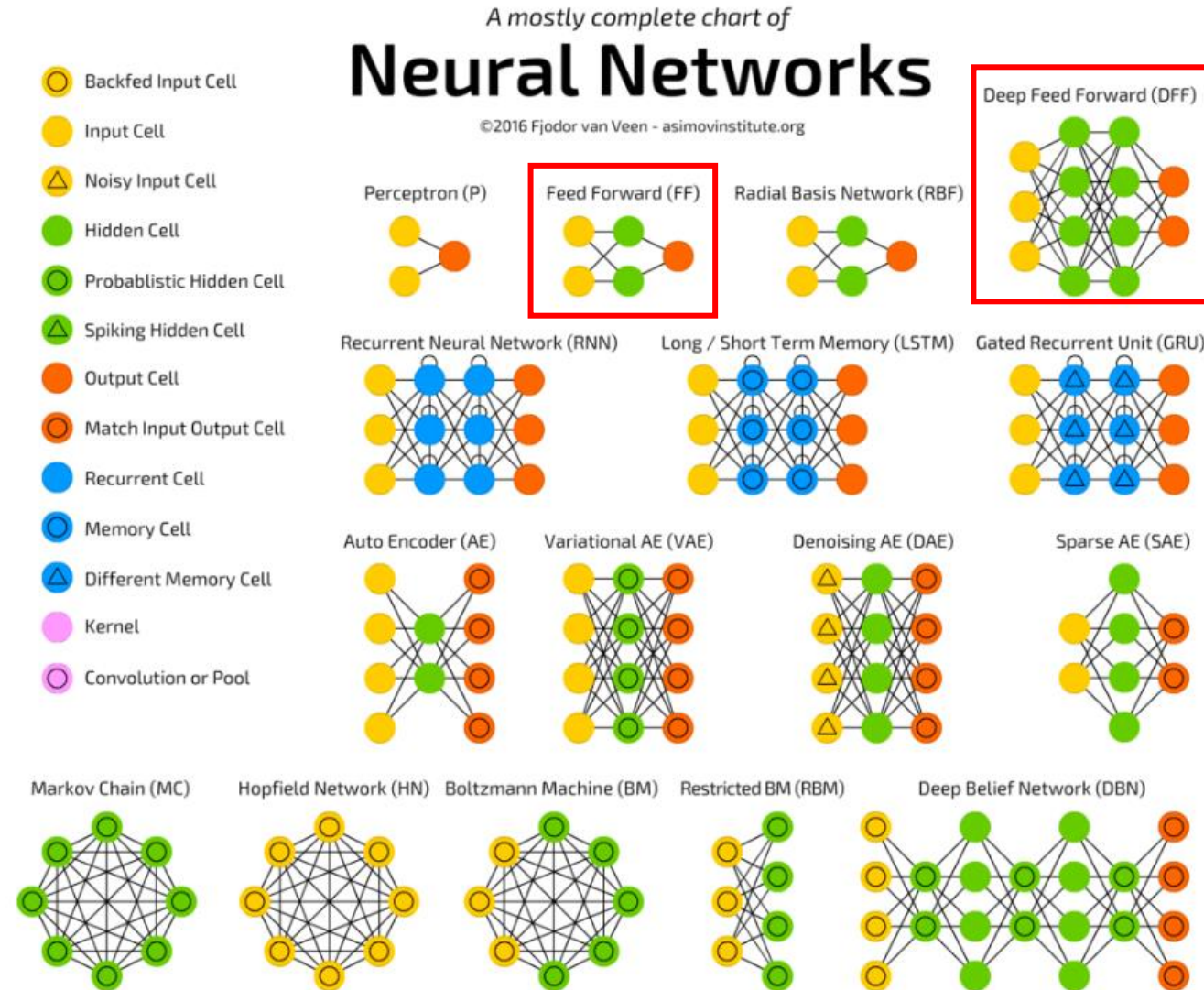
<https://www.udemy.com/course/machinelearning/>



# Overview

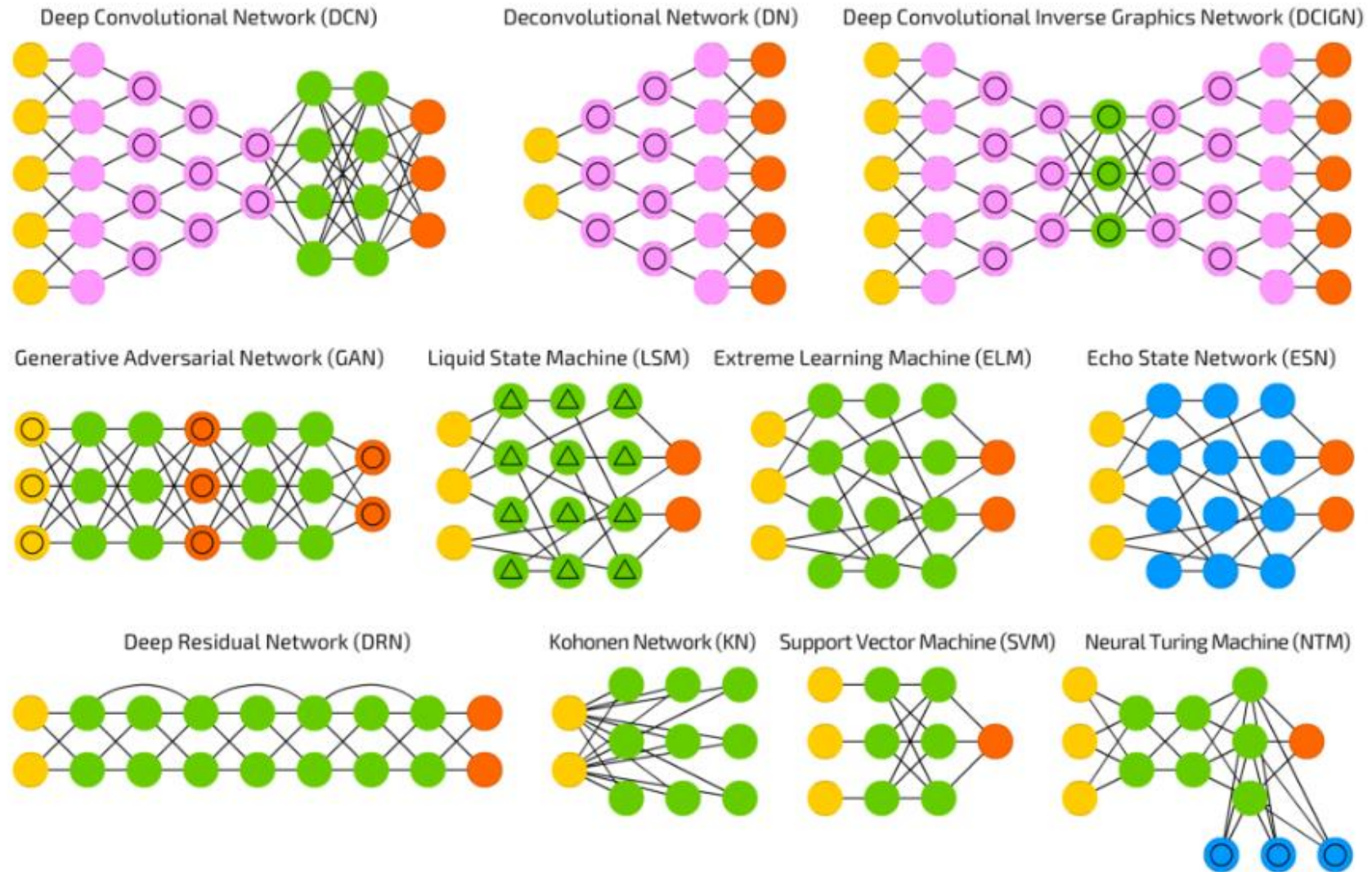


# Overview





# Overview





# Overview

- **Supervised Learning:** CNNs classifying images in imagenet.
- **Unsupervised Learning:** Boltzmann Machines, AutoEncoders, GANs, DC-GANS, VAE, SOMs, etc.
- **Reinforcement:** Deep Convolutional Q-Learning that plays videogames from pixel input, AlphaGO, etc.

# Overview

- Artificial Neural Networks for Regression and Classification
- Convolutional Neural Networks for Computer Vision
- Recurrent Neural Networks for Time Series Analysis
- Self Organizing Maps for Feature Extraction
- Deep Boltzmann Machines for Recommendation Systems
- Auto Encoders for Recommendation Systems

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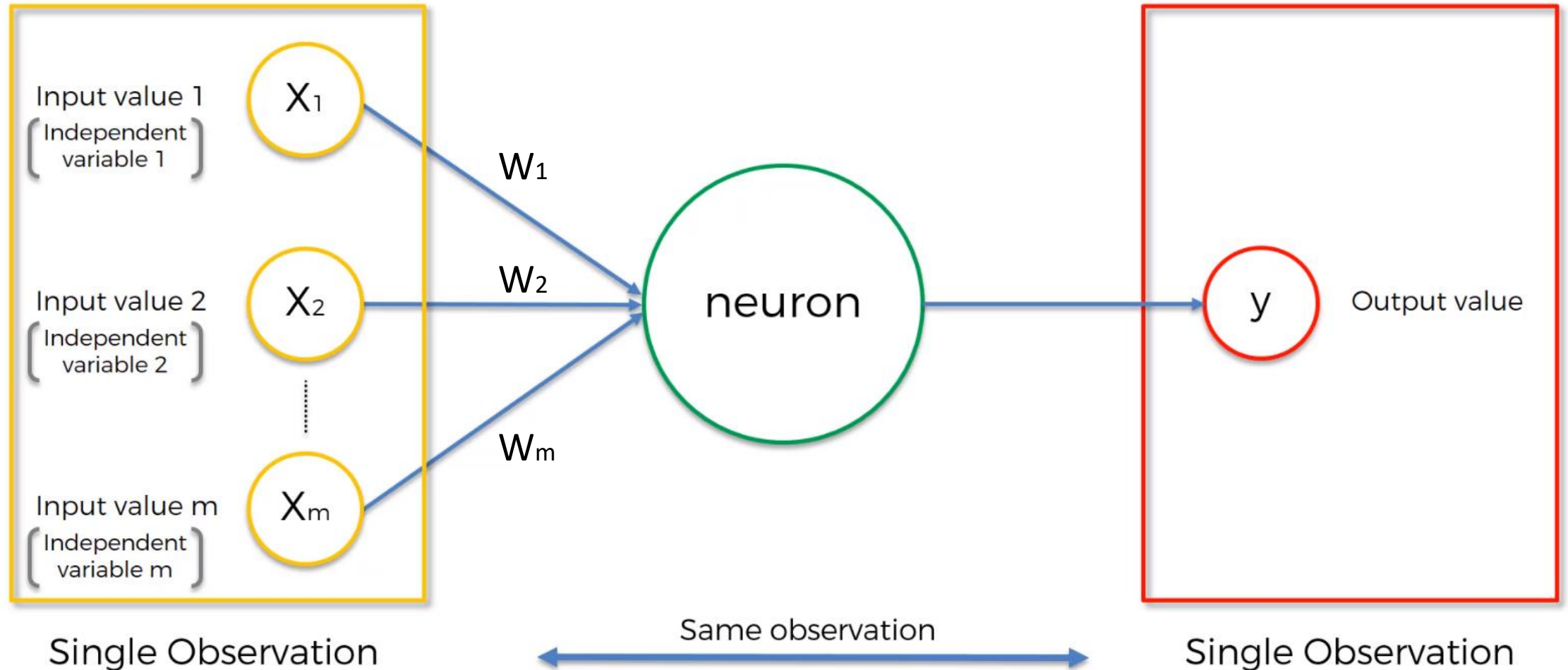
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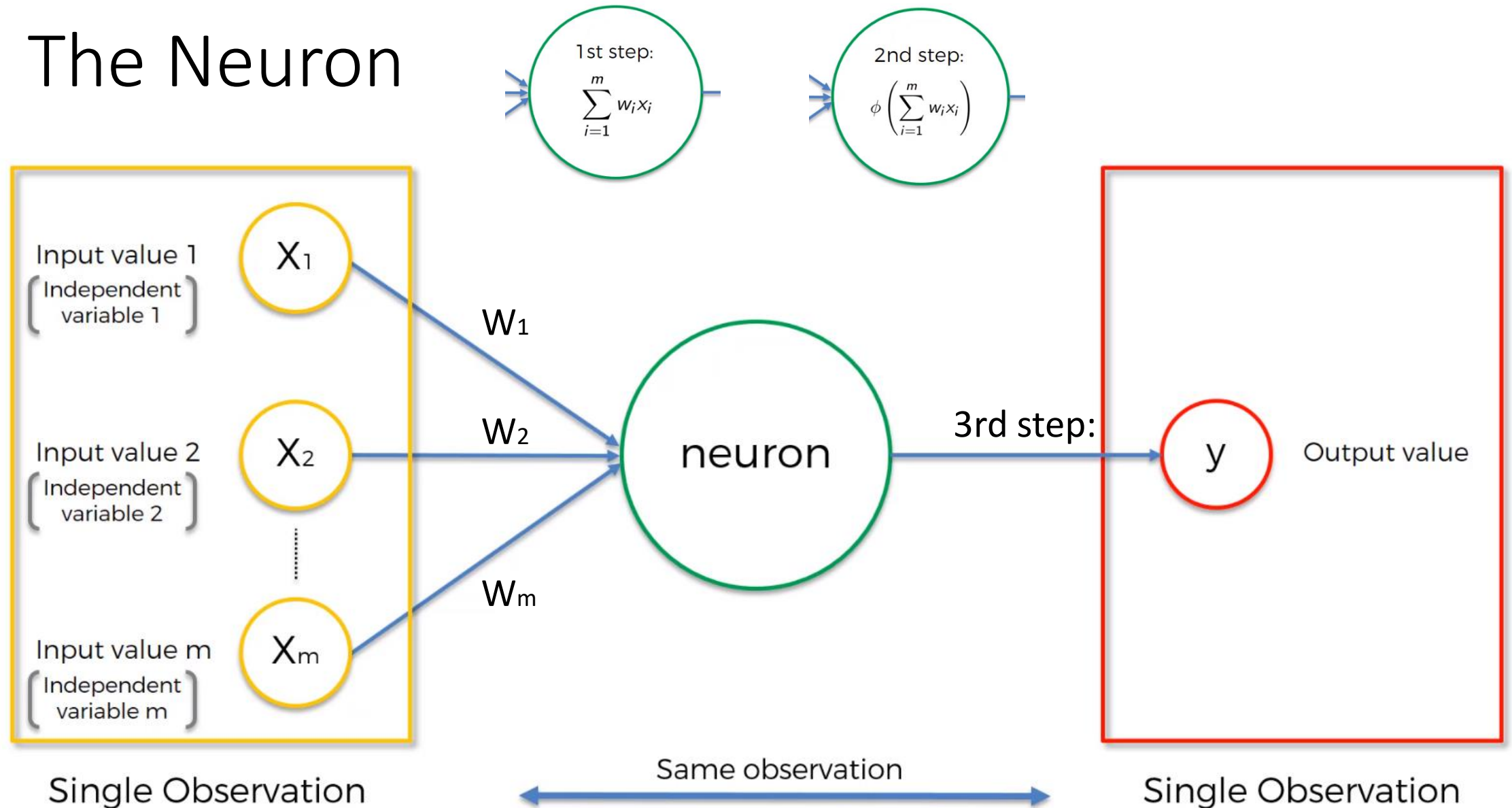
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# The Neuron



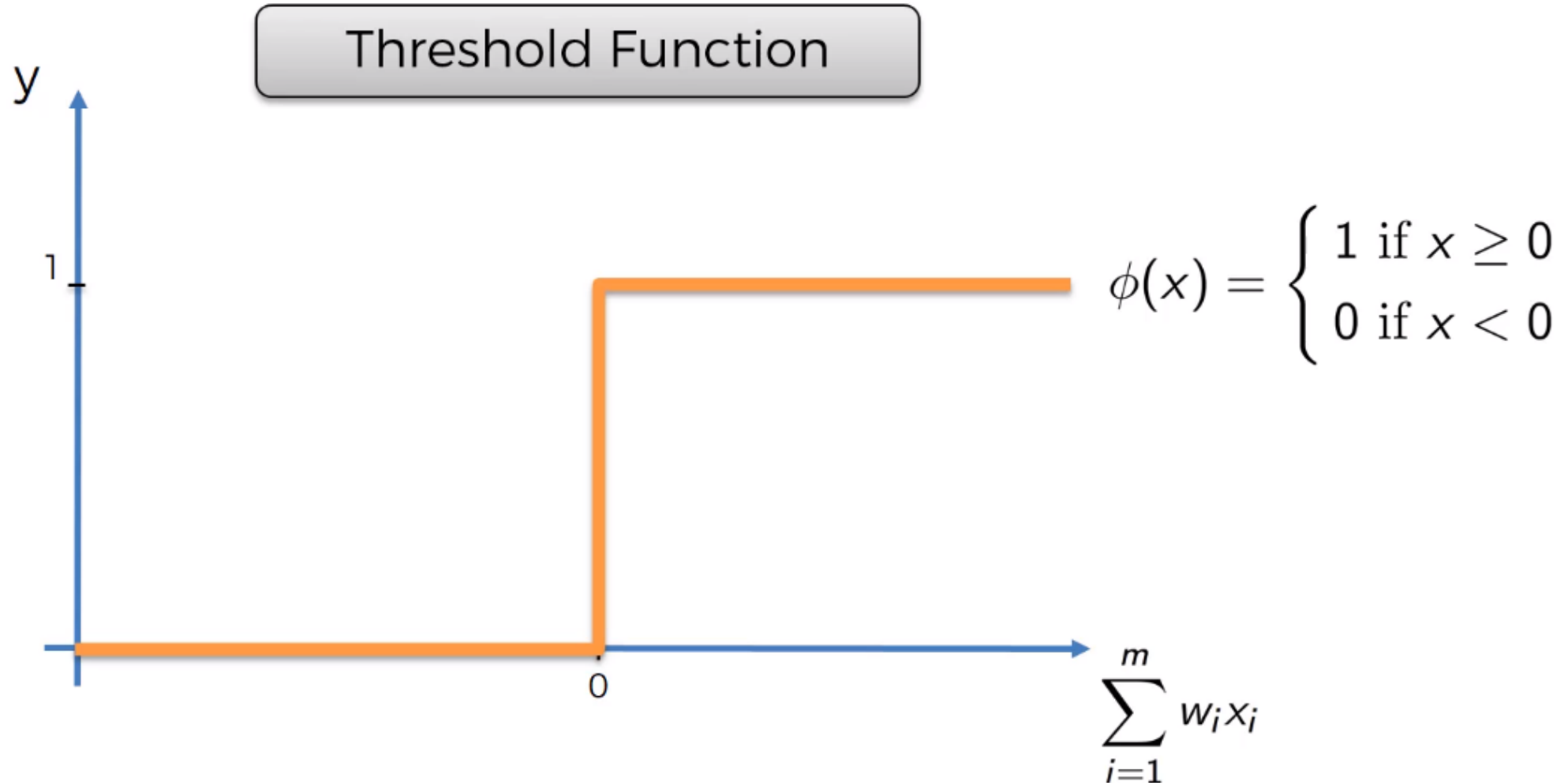
# The Neuron



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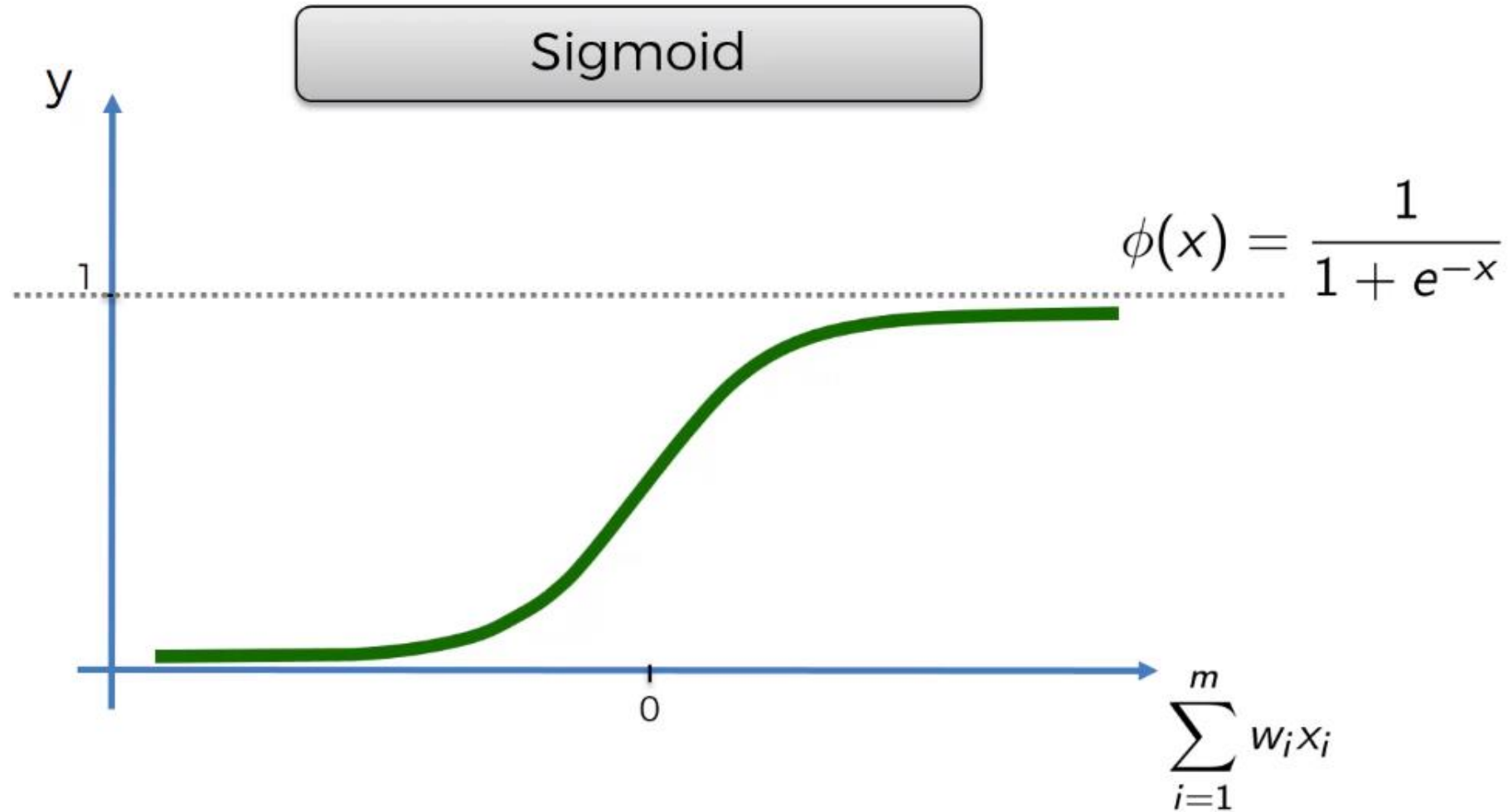
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# The Activation Function

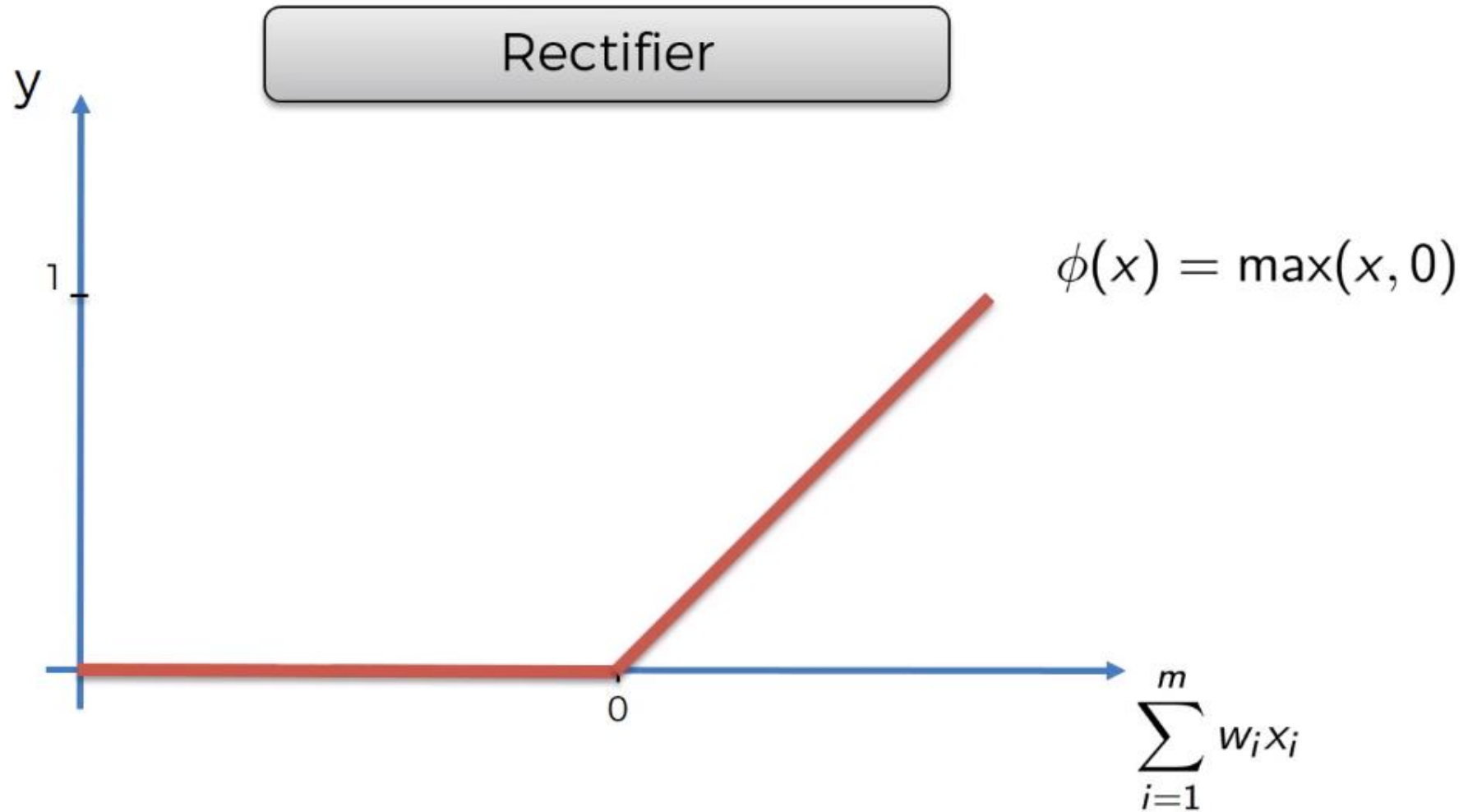




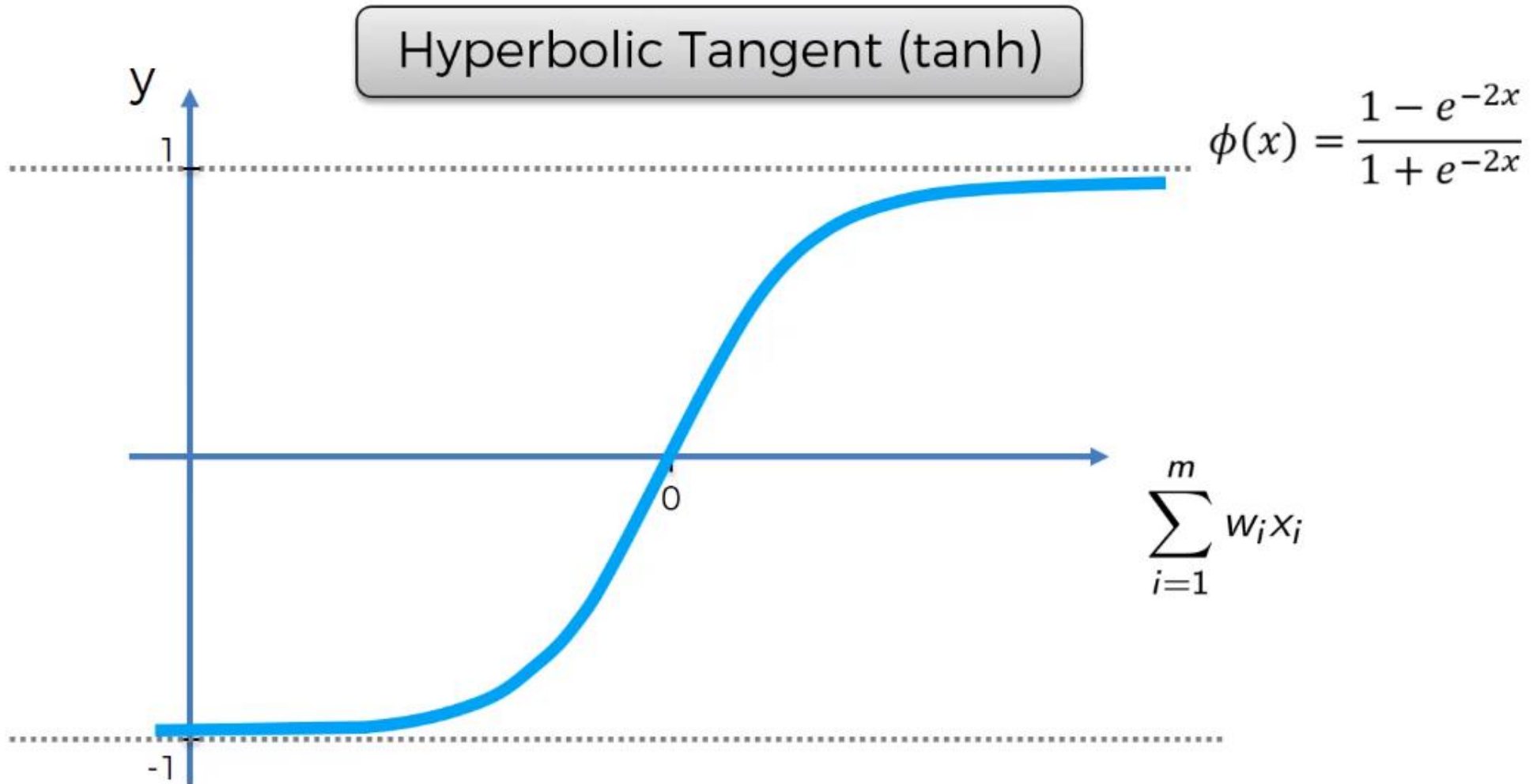
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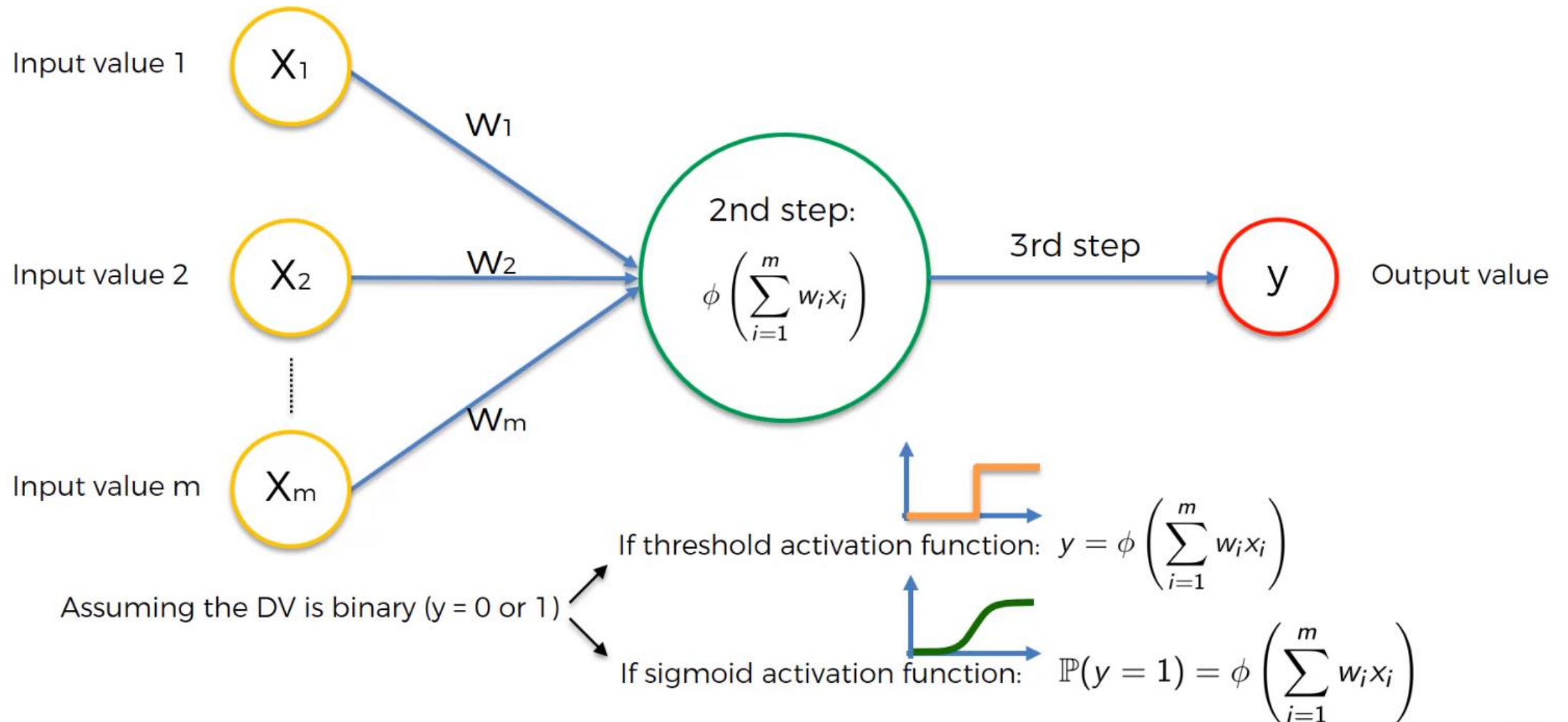
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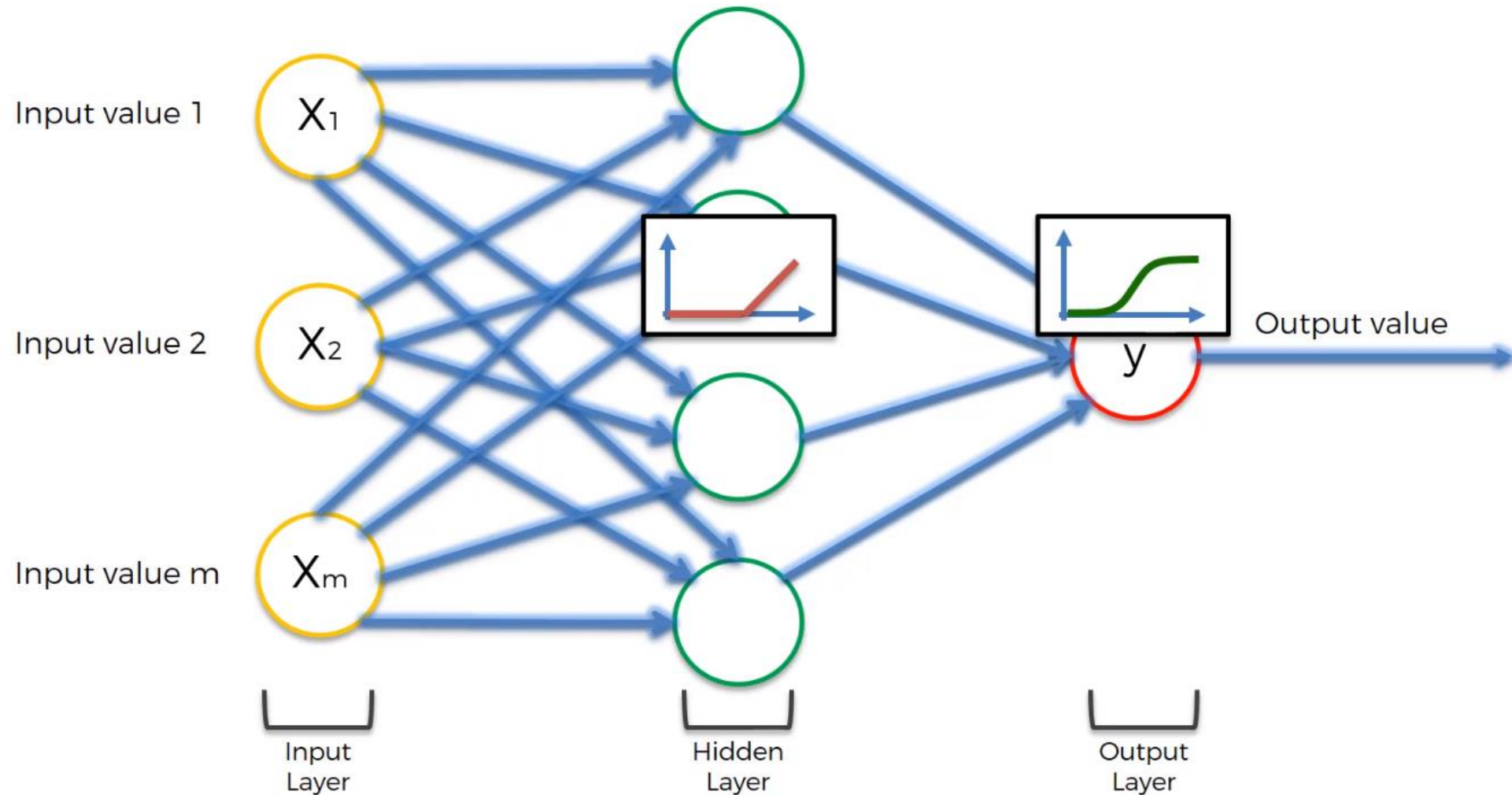


# The Activation Function





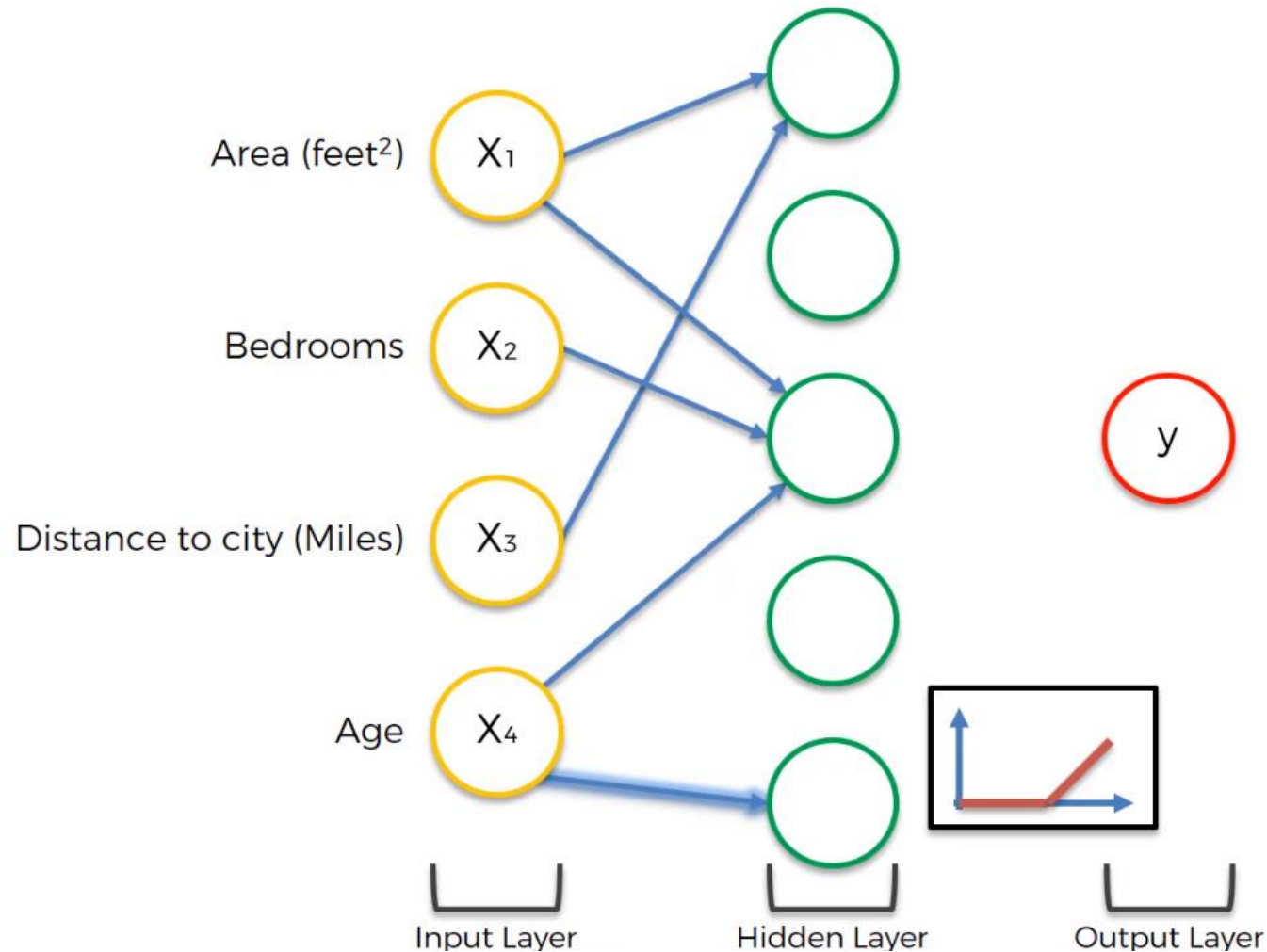
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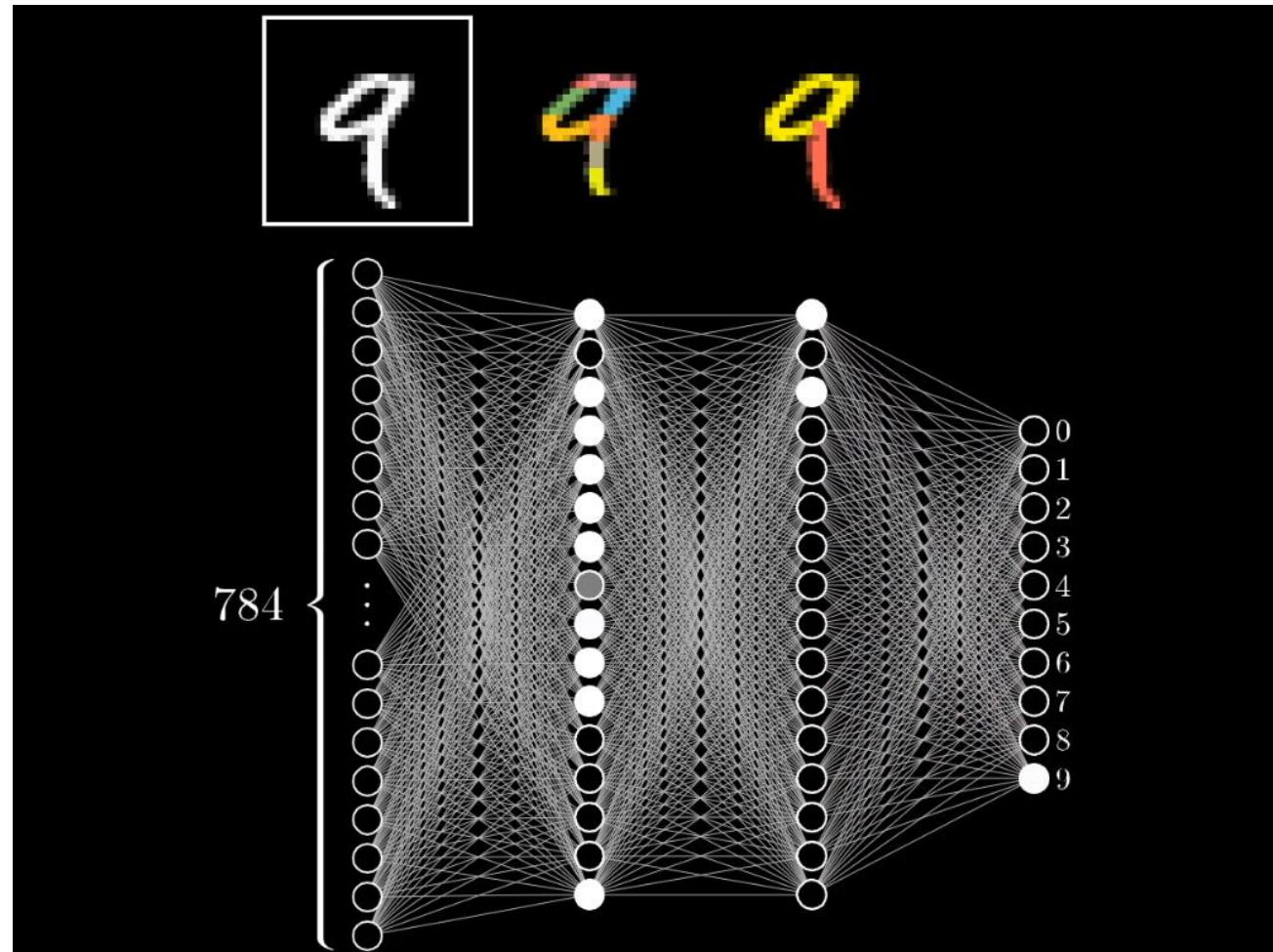
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# How do Neural Networks work?

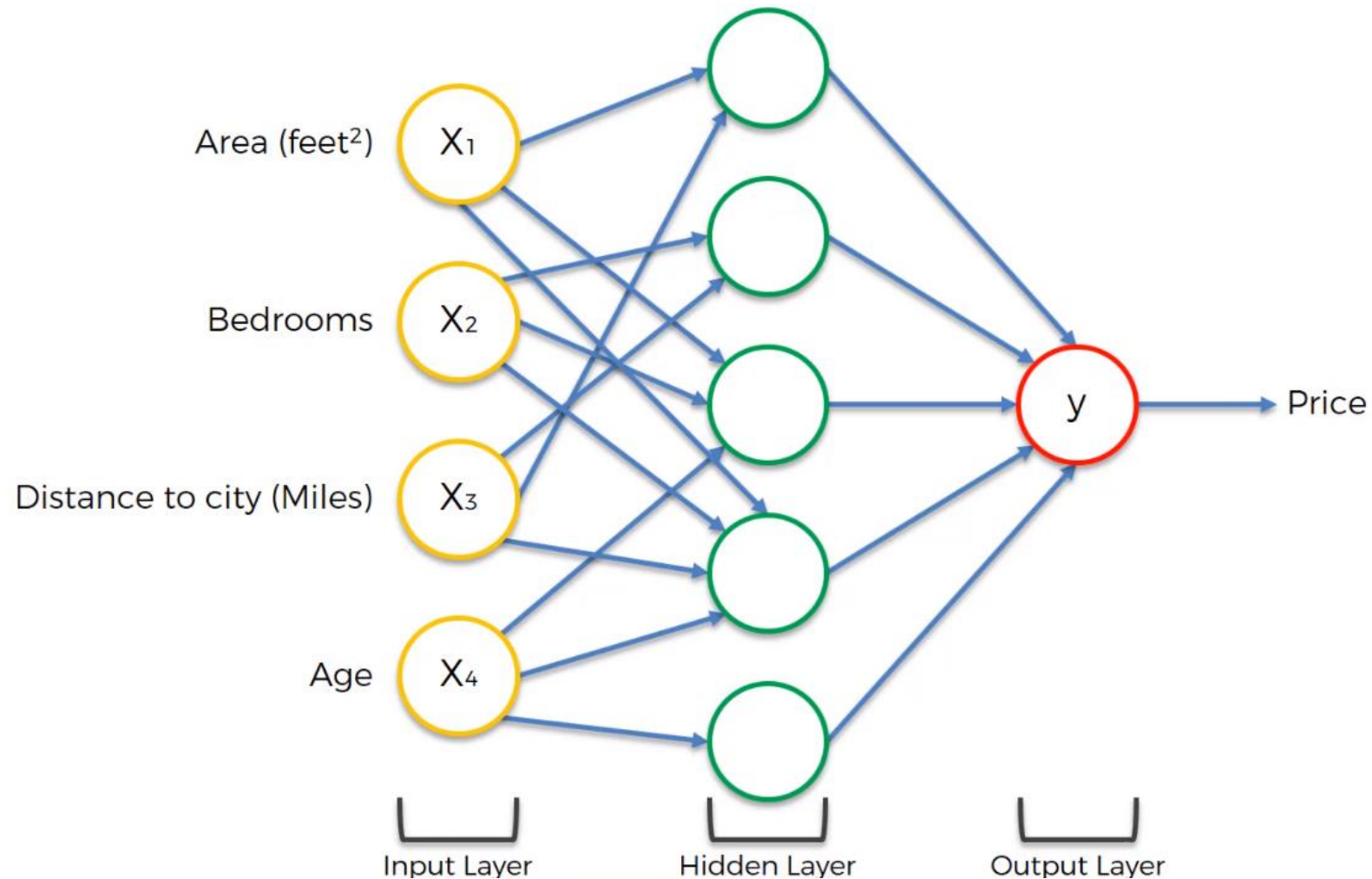


# How do Neural Networks work?





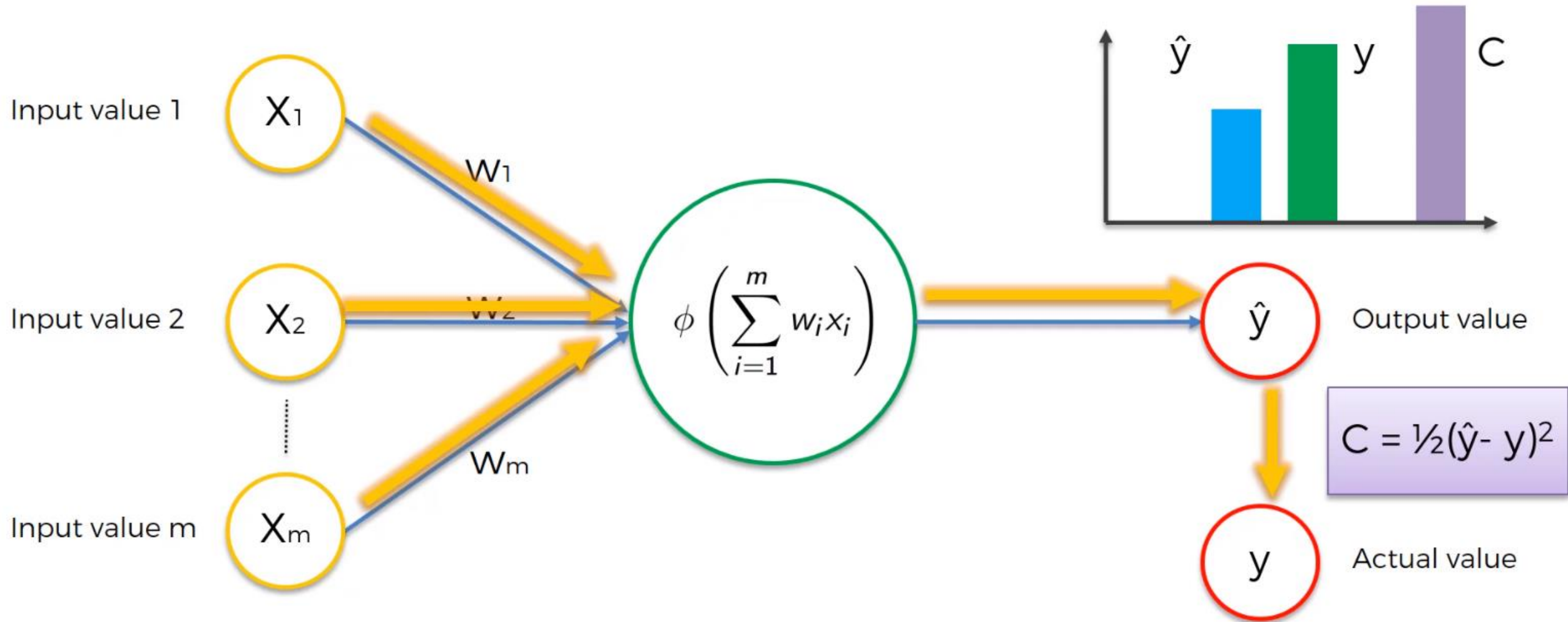
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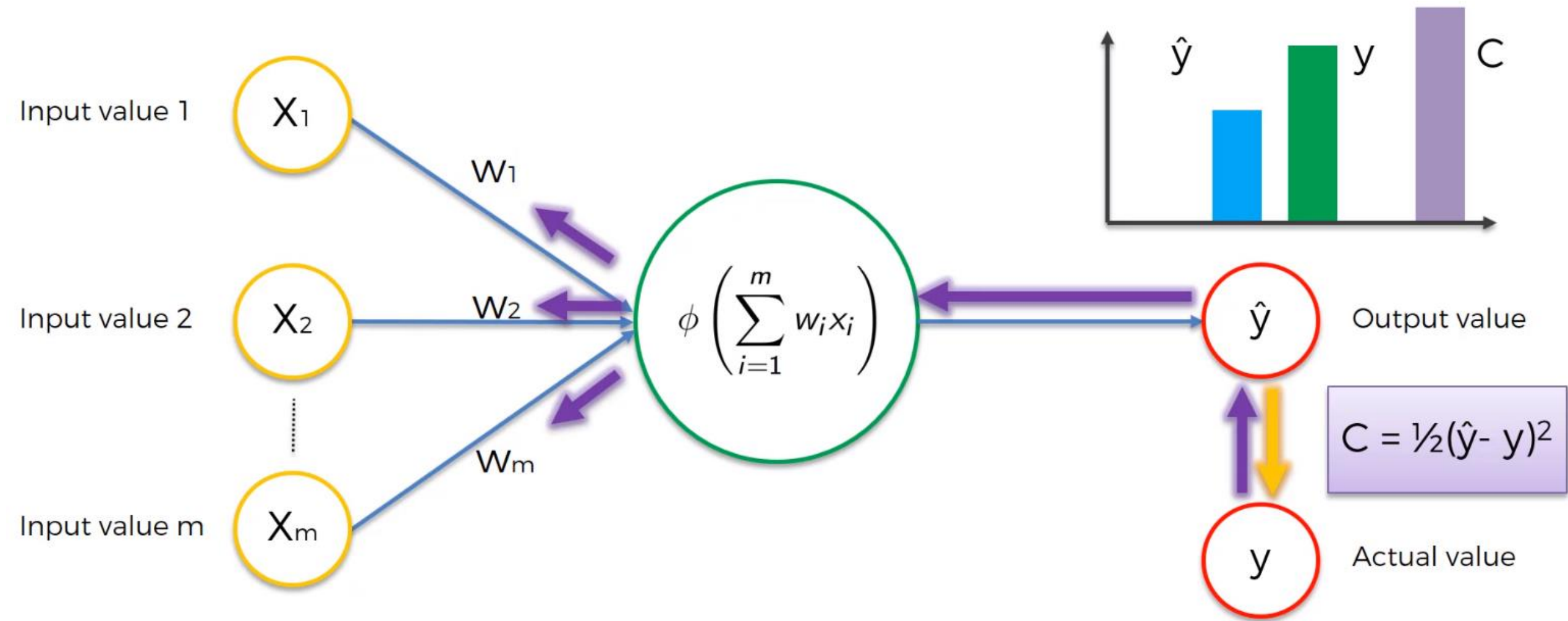
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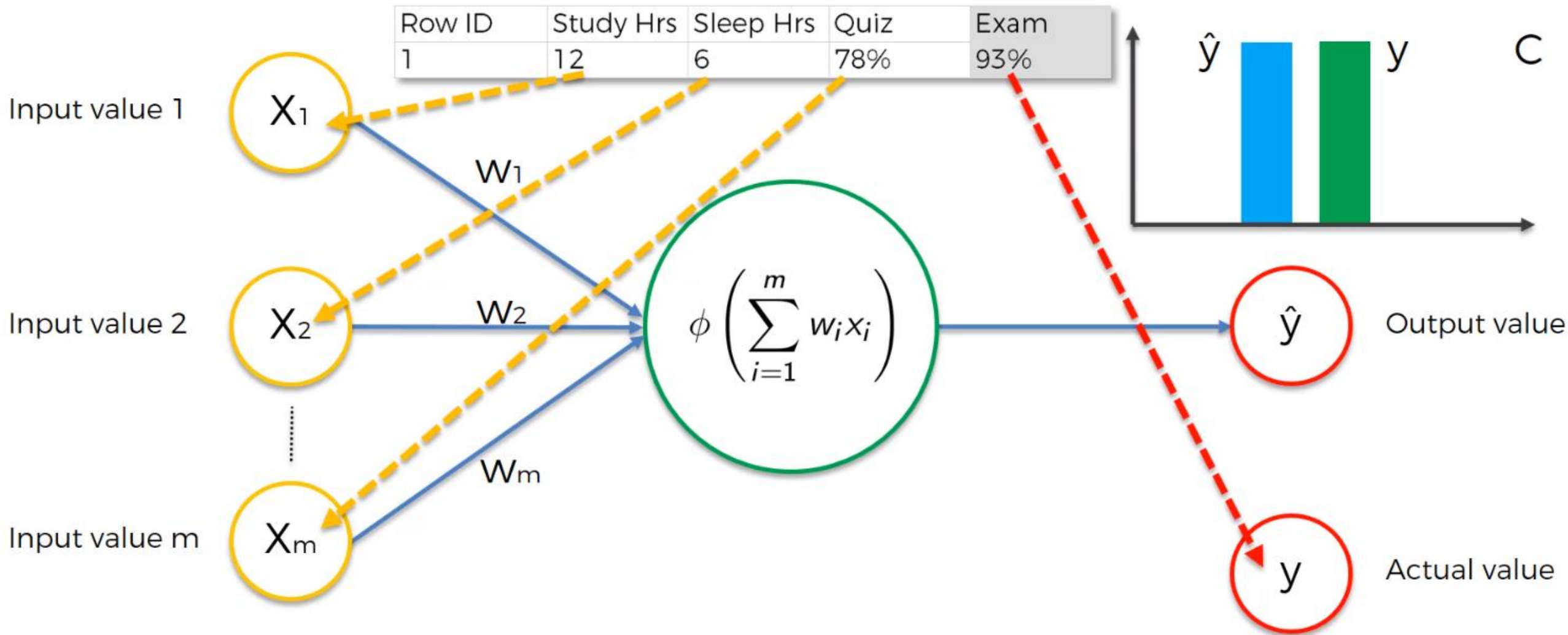
# How do Neural Networks learn?



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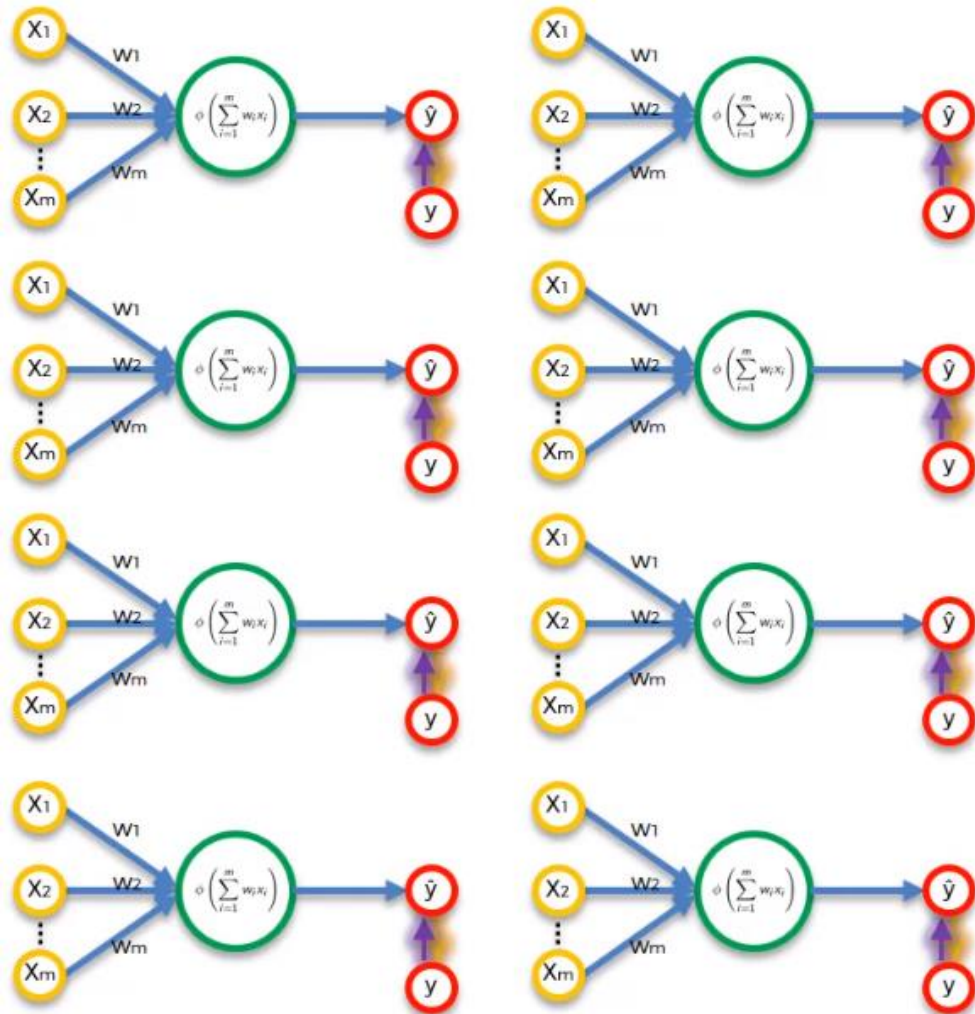


# How do Neural Networks learn?



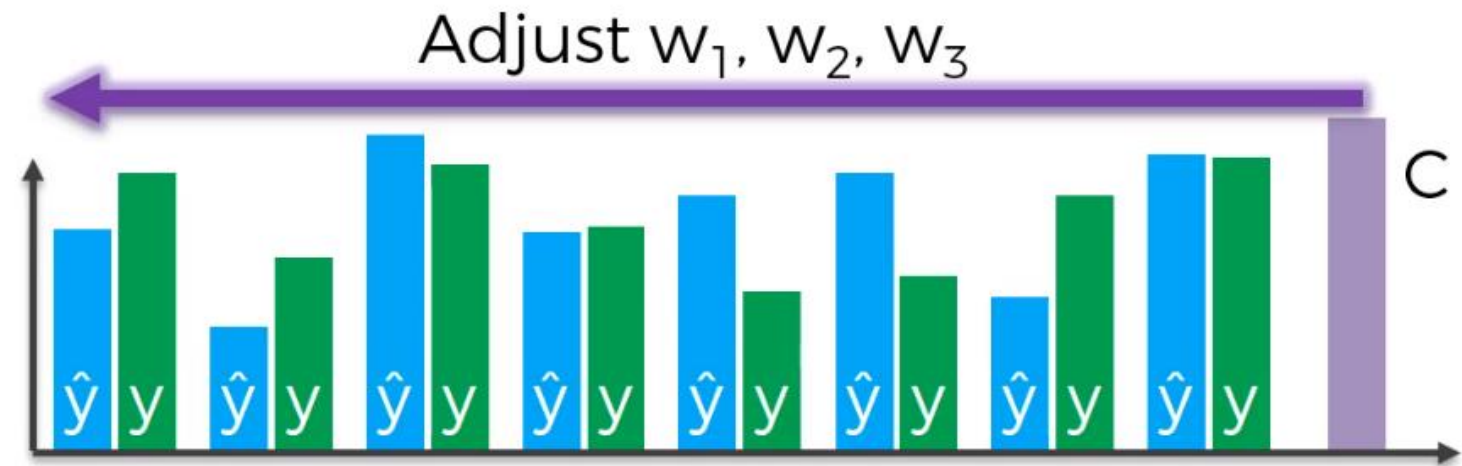


# How do Neural Networks learn?



Row ID	Study Hrs	Sleep Hrs	Quiz	Exam
1	12	6	78%	93%
2	22	6.5	24%	68%
3	115	4	100%	95%
4	31	9	67%	75%
5	0	10	58%	51%
6	5	8	78%	60%
7	92	6	82%	89%
8	57	8	91%	97%

$$C = \sum \frac{1}{2}(\hat{y} - y)^2$$

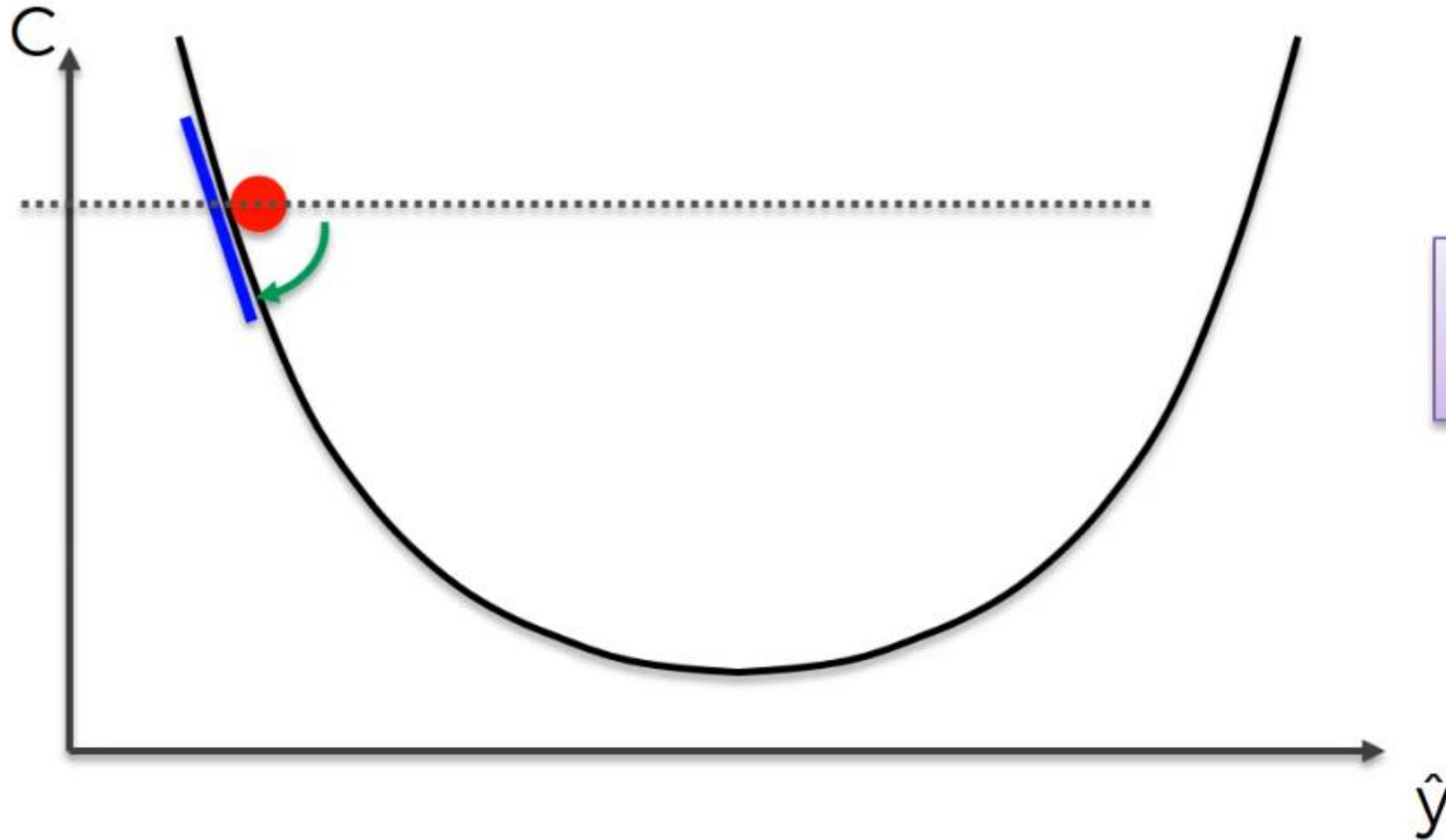




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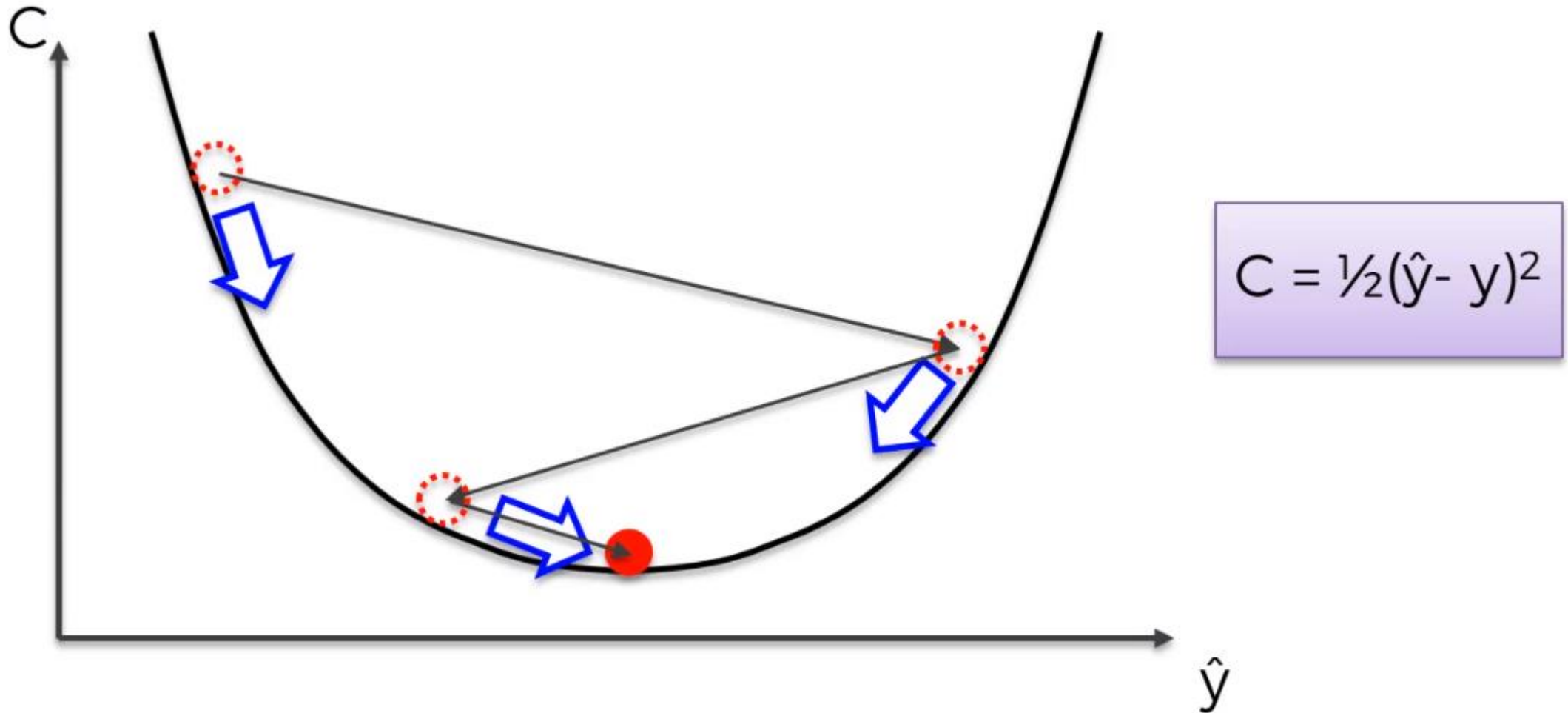
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# Gradient Descent

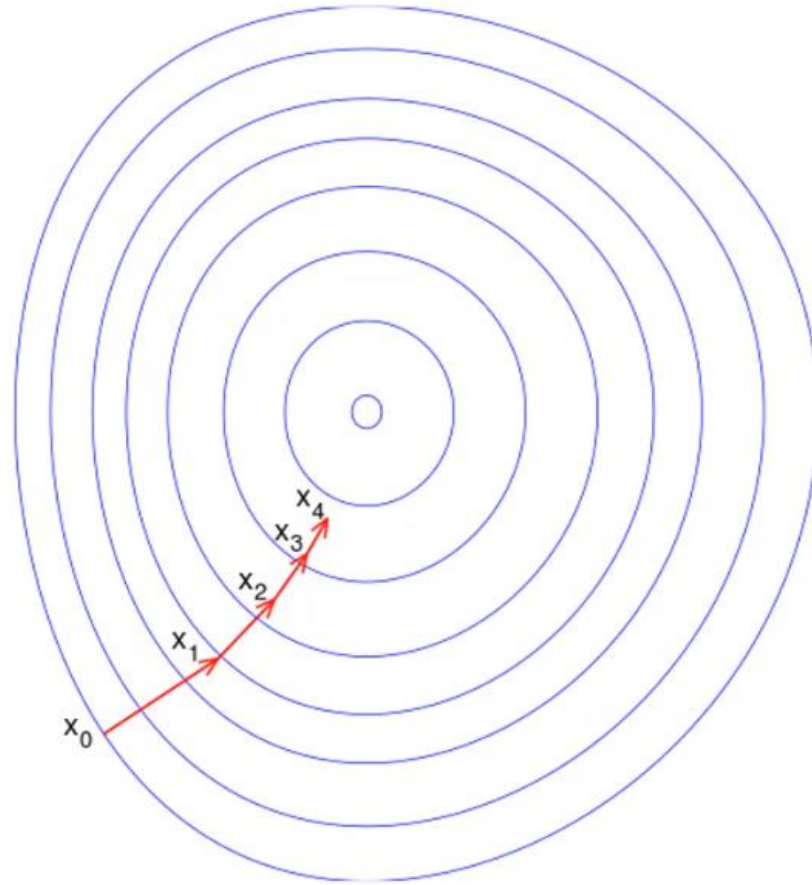


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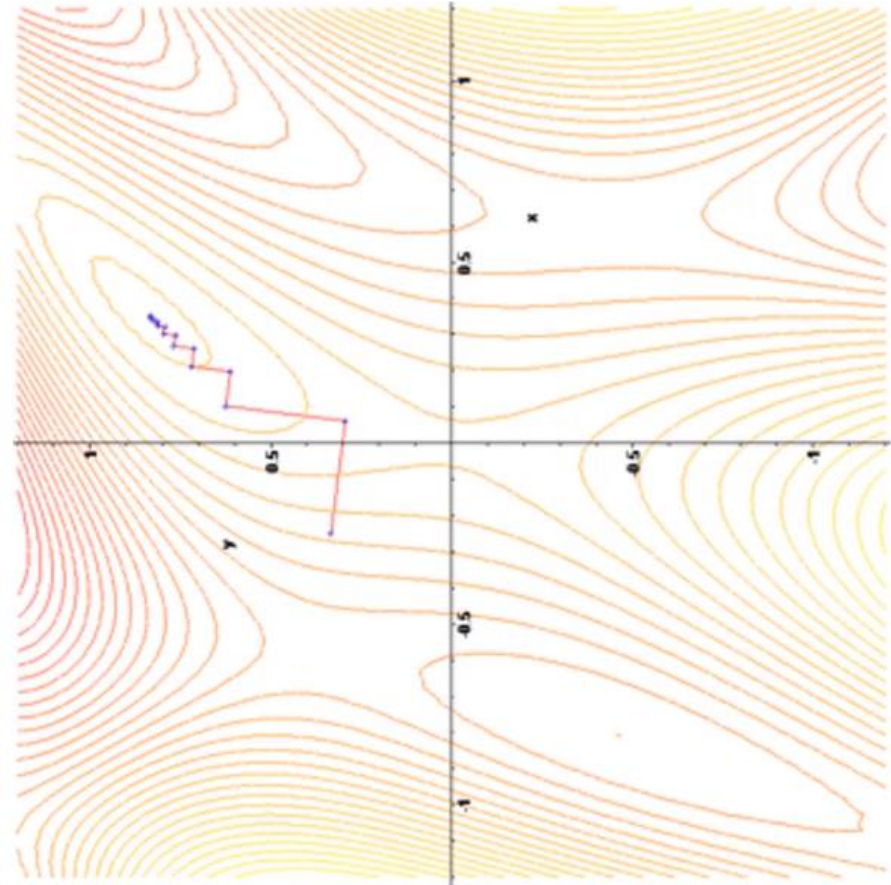
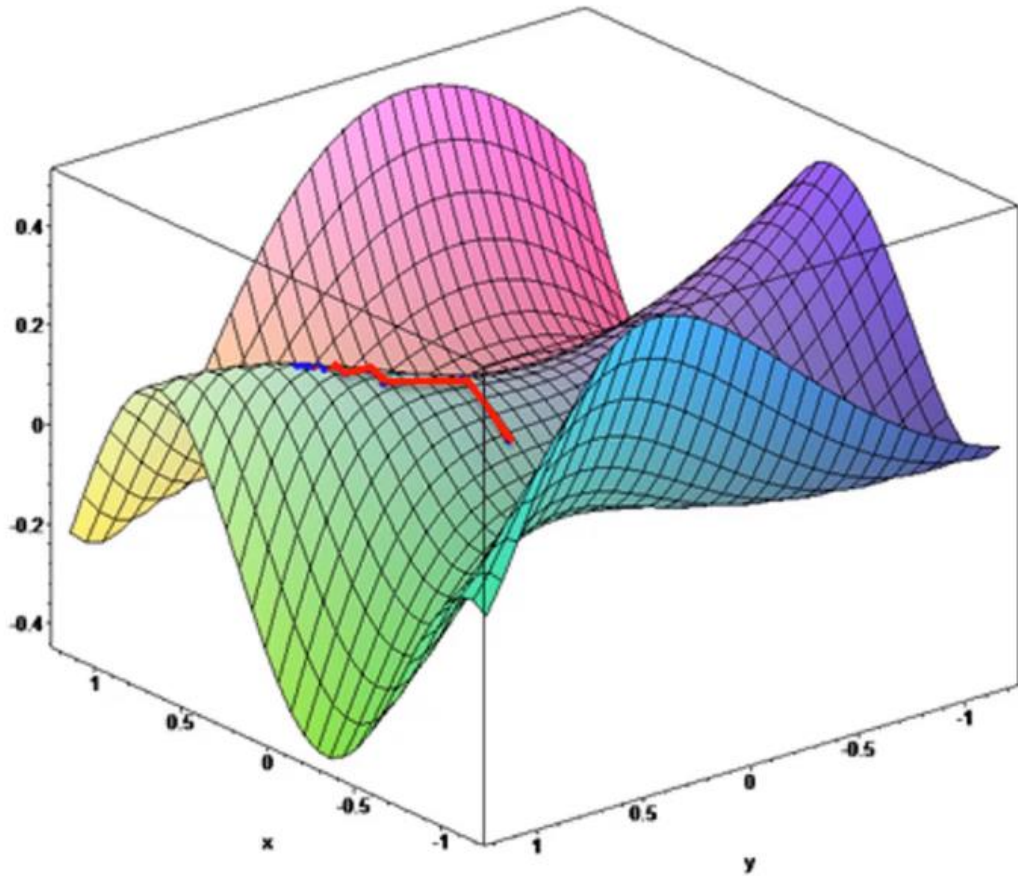
# Gradient Descent



# Gradient Descent



# Gradient Descent



# Gradient Descent

Step 1: Take the derivative of the **Cost Function** for each parameter in it. In fancy Machine Learning lingo, take the **Gradient** of the **Loss Function**.

Step 2: Pick random values for the parameters.

Step 3: Plug the parameter values into the derivatives (**Gradient**).

Step 4: Calculate the Step Sizes:  $\text{Step Size} = \text{Slope} * \text{Learning Rate}$

Step 5: Calculate the New Parameters:

$$\text{New Parameter} = \text{Old Parameter} - \text{Step Size}$$

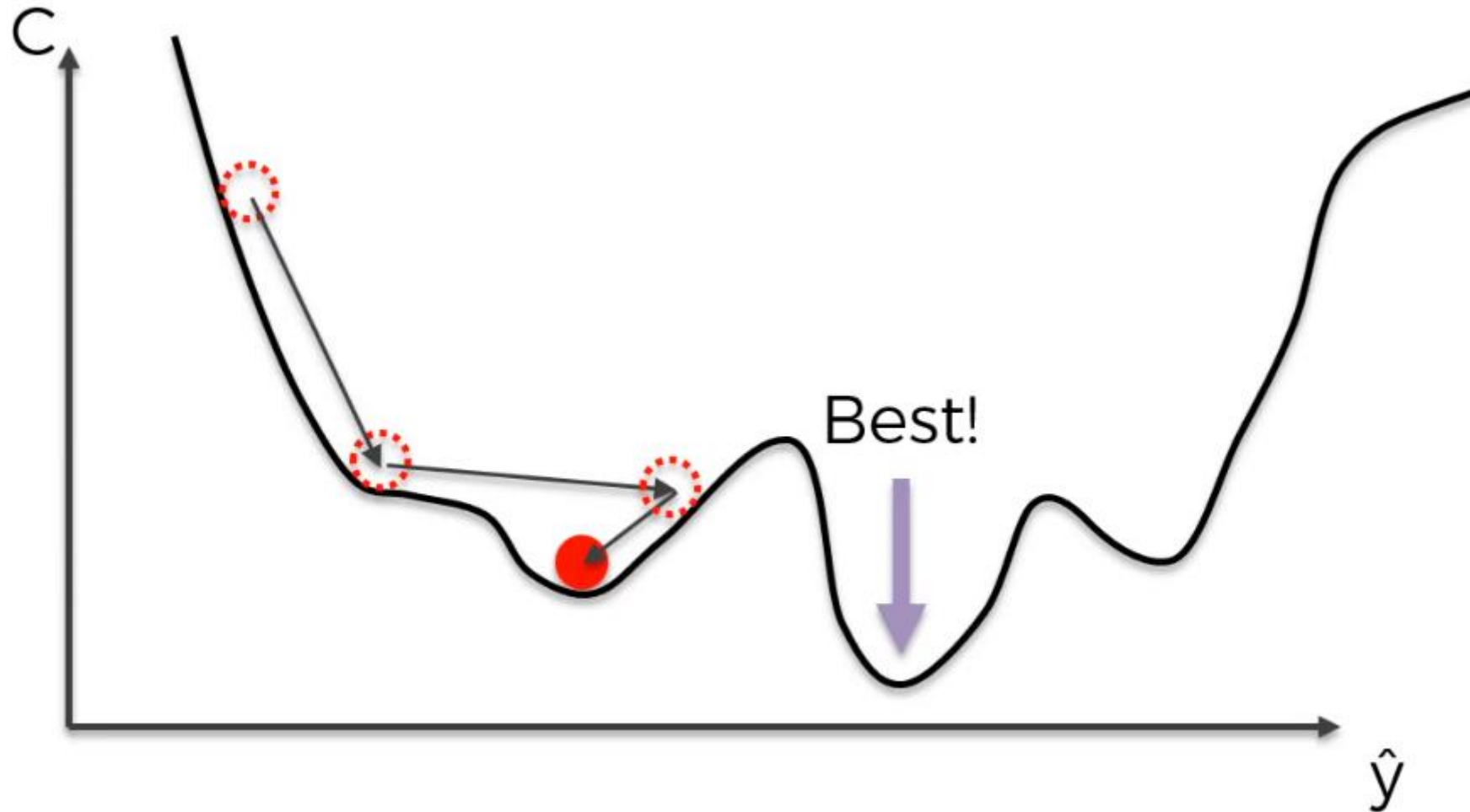
Step 6: Go back to **Step 3** and repeat until **Step Size** is very small, or you reach the **Maximum Number of Steps**

# Summary

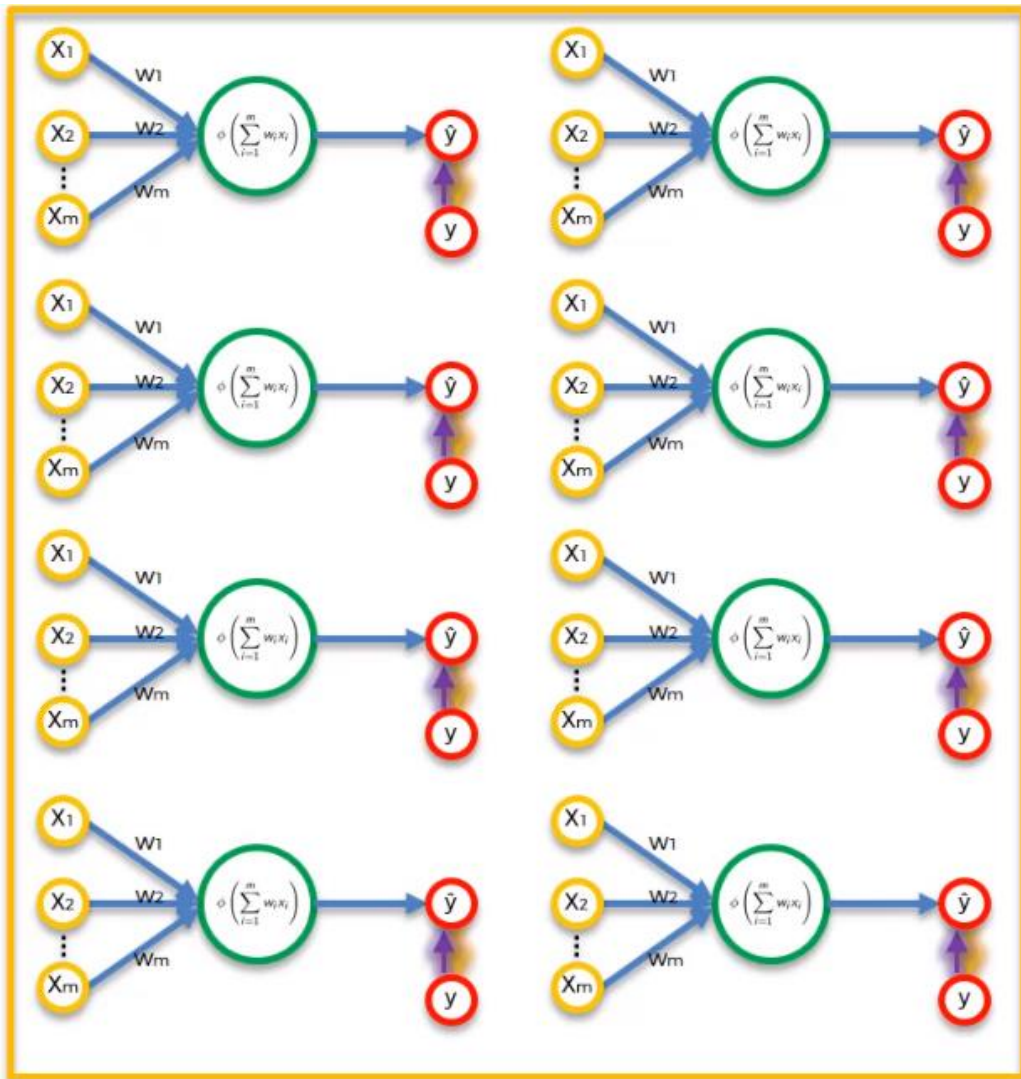
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# Stochastic Gradient Descent



# Stochastic Gradient Descent

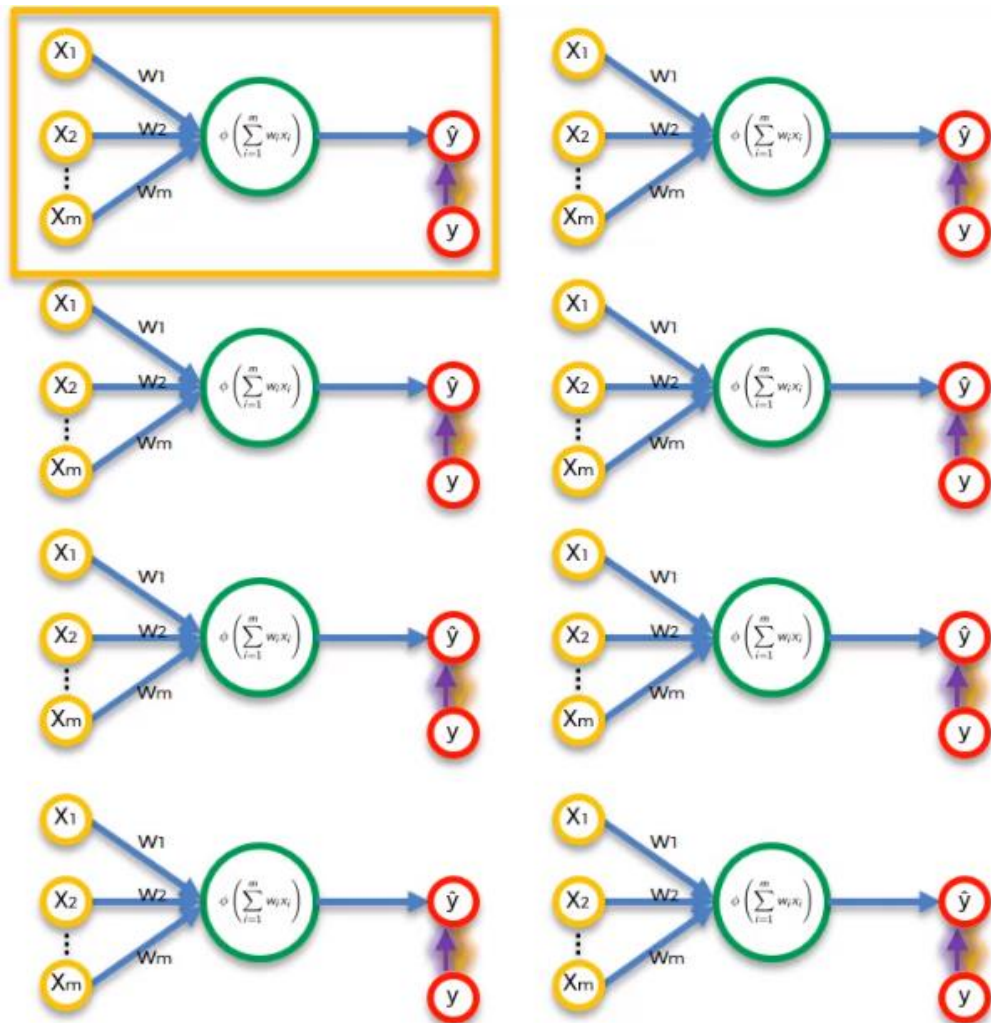


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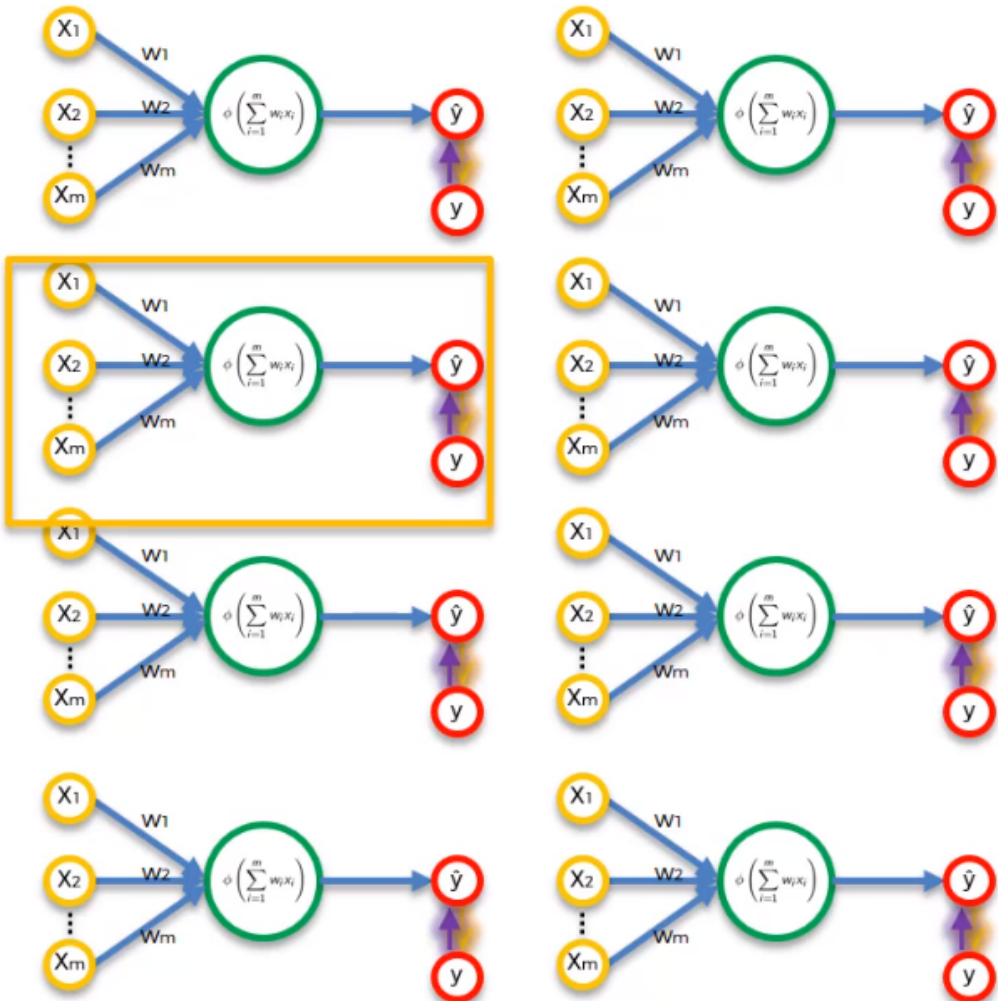


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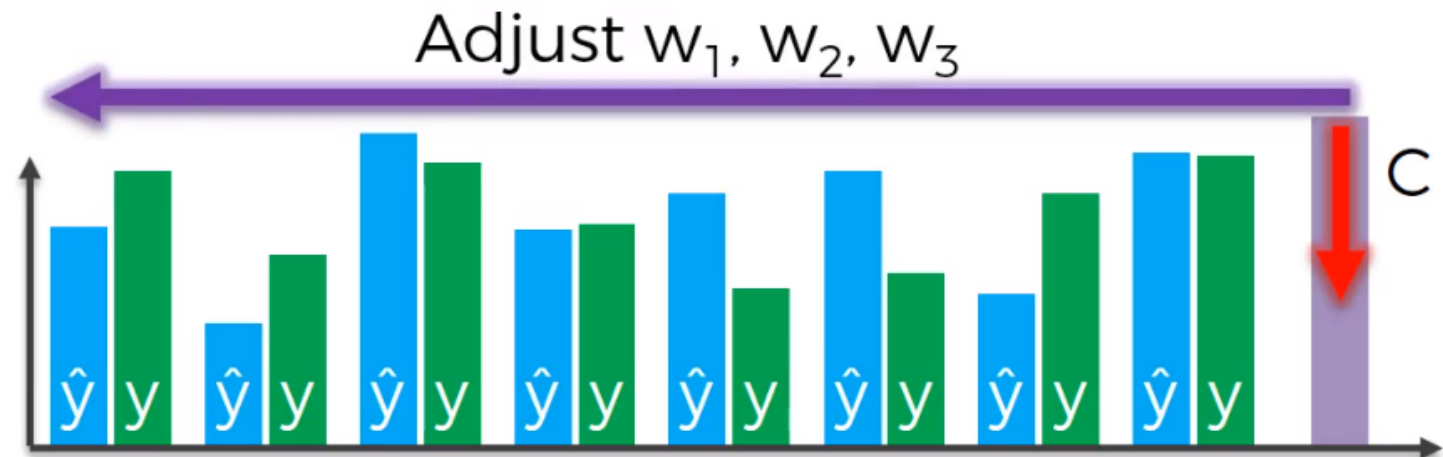


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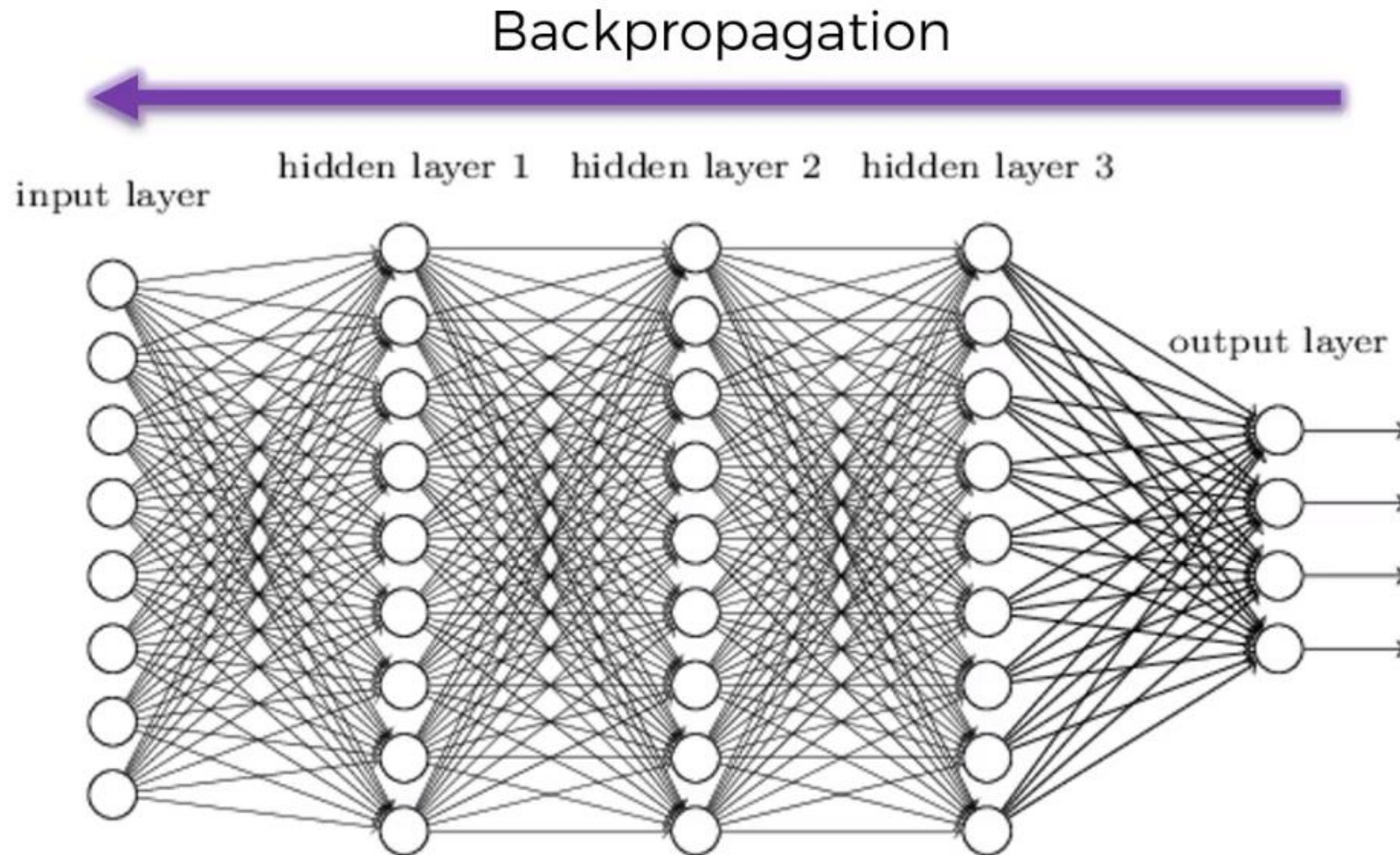
Batch  
Gradient  
Descent

Stochastic  
Gradient  
Descent

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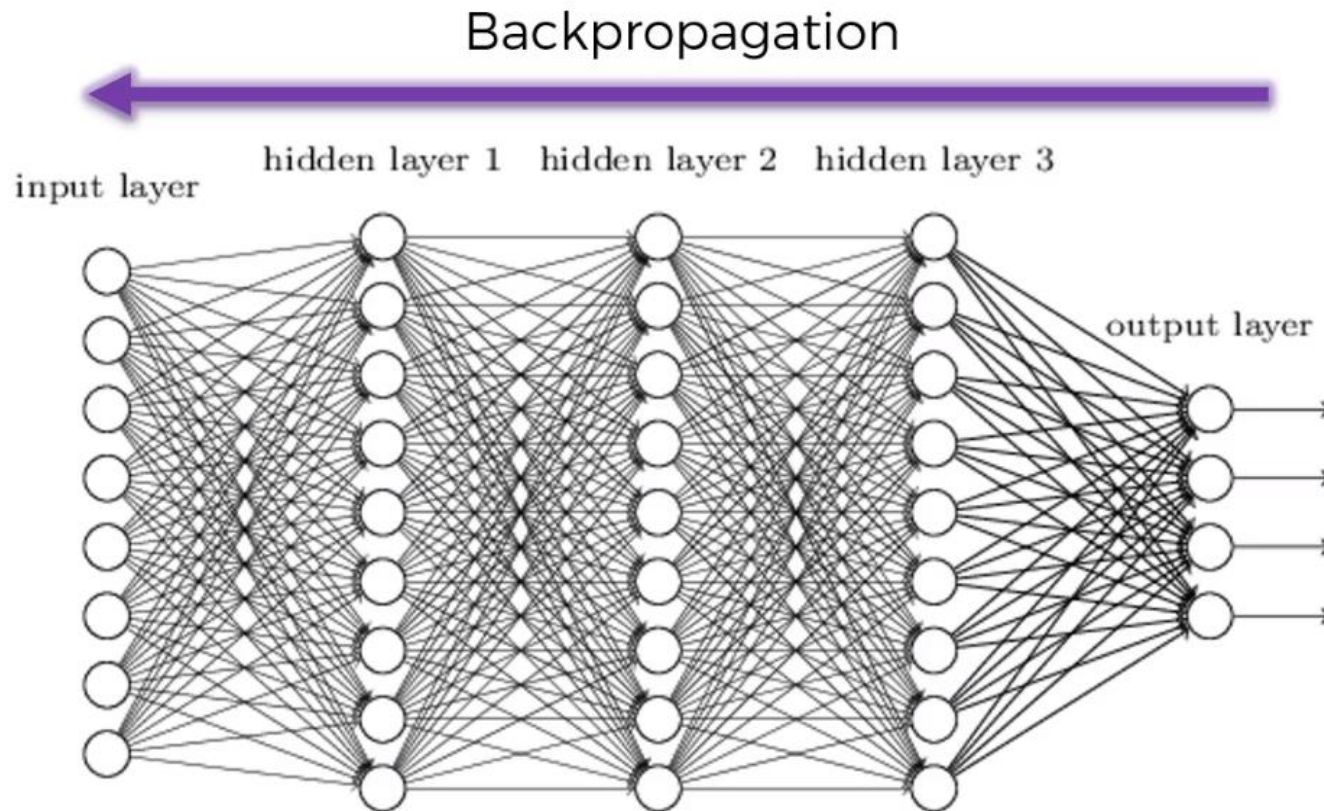
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# Backpropagation





# Backpropagation



## Gradient of the cost function

**Input:** 270 weights/biases

**Output:** 270 weights/biases nudges

**Parameters:** 1 number (the cost)

## Cost function

**Input:** 270 weights/biases

**Output:** 1 number (the cost)

**Parameters:** Many, many, many training example

## Neural Network function

**Input:** 8 numbers (features)

**Output:** 4 numbers

**Parameters:** 270 weights/biases

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# Algorithm summary

**STEP 1:** Randomly initialise the weights to small numbers close to 0 (but not 0).



**STEP 2:** Input the first observation of your dataset in the input layer, each feature in one input node.



**STEP 3:** Forward-Propagation: from left to right, the neurons are activated in a way that the impact of each neuron's activation is limited by the weights. Propagate the activations until getting the predicted result  $y$ .



**STEP 4:** Compare the predicted result to the actual result. Measure the generated error.



**STEP 5:** Back-Propagation: from right to left, the error is back-propagated. Update the weights according to how much they are responsible for the error. The learning rate decides by how much we update the weights.



**STEP 6:** Repeat Steps 1 to 5 and update the weights after each observation (Reinforcement Learning). Or:  
Repeat Steps 1 to 5 but update the weights only after a batch of observations (Batch Learning).



**STEP 7:** When the whole training set passed through the ANN, that makes an epoch. Redo more epochs.

# Algorithms used

- Neural Network architecture/topology: **Deep Feed Forward (DFF)**
- Activation function: **Rectifier** and **sigmoid**
- Training method/Optimizer:
  - **Backpropagation** to calculate the derivatives and
  - **Gradient Descent** which descends through the gradient, i.e. adjust the parameters
- Cost/loss function:

$$C = \sum \frac{1}{2}(\hat{y} - y)^2$$