



# A Succinct Tutorial of Neural Networks in Deep Learning

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## Objectives

- To describe the concept of artificial intelligence, machine learning and deep learning
- To discuss the operation of an artificial neuron
- To discuss the architecture of a neural networks and its elements.
- To explain the operation of a neural network
- To describe the concept of steepest descent utilized in neural networks.
- To explain the training and learning of a neural network

## What is Artificial Intelligence?

Very simple! Artificial Intelligence (AI) is a branch of computer science that is concerned with building computer systems capable of performing tasks the typically require human intelligence. Basically we try to build computer systems that can simulate human intelligence. Artificial Intelligence relies on algorithms to achieve the result.

## What is an algorithm?

An algorithm is a set of instructions that a computer can execute. A complex algorithm can be built on top of other simpler algorithms. Many AI algorithms are capable of learning from data – they can improve themselves.

## What is machine learning?

- Machine learning is a SUBSET of artificial intelligence
- MACHINE LEARNING is an application of AI that can automatically learn and improve from experience
- MACHINE LEARNING is built upon an ALGORITHM that uses computational methods to “learn” information directly from data without relying on a predetermined equation as a model
- MACHINE LEARNING algorithms adaptively improve their performance as the data available for learning increases

## How can one utilize the data to train and enable the algorithms to learn?

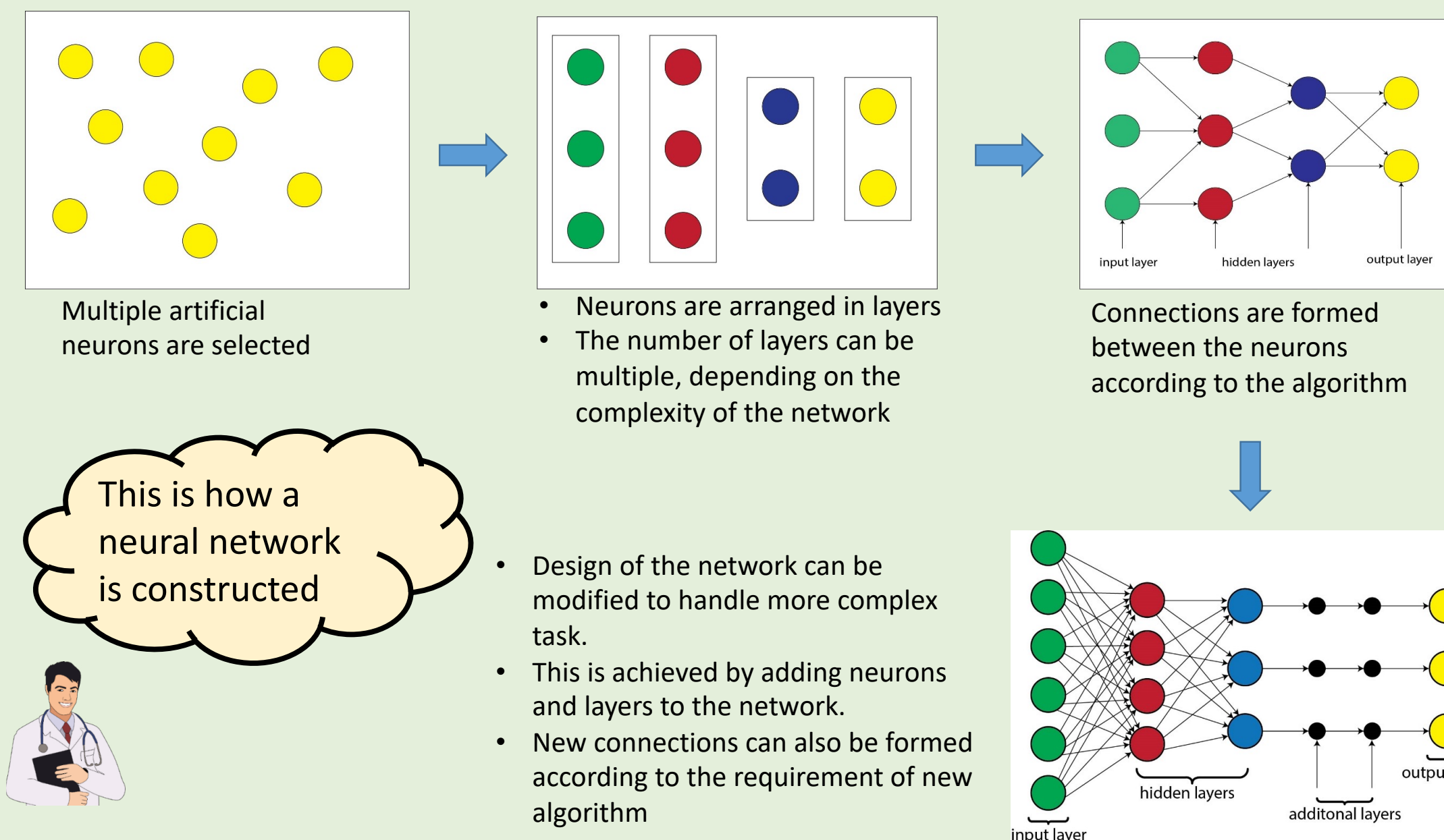
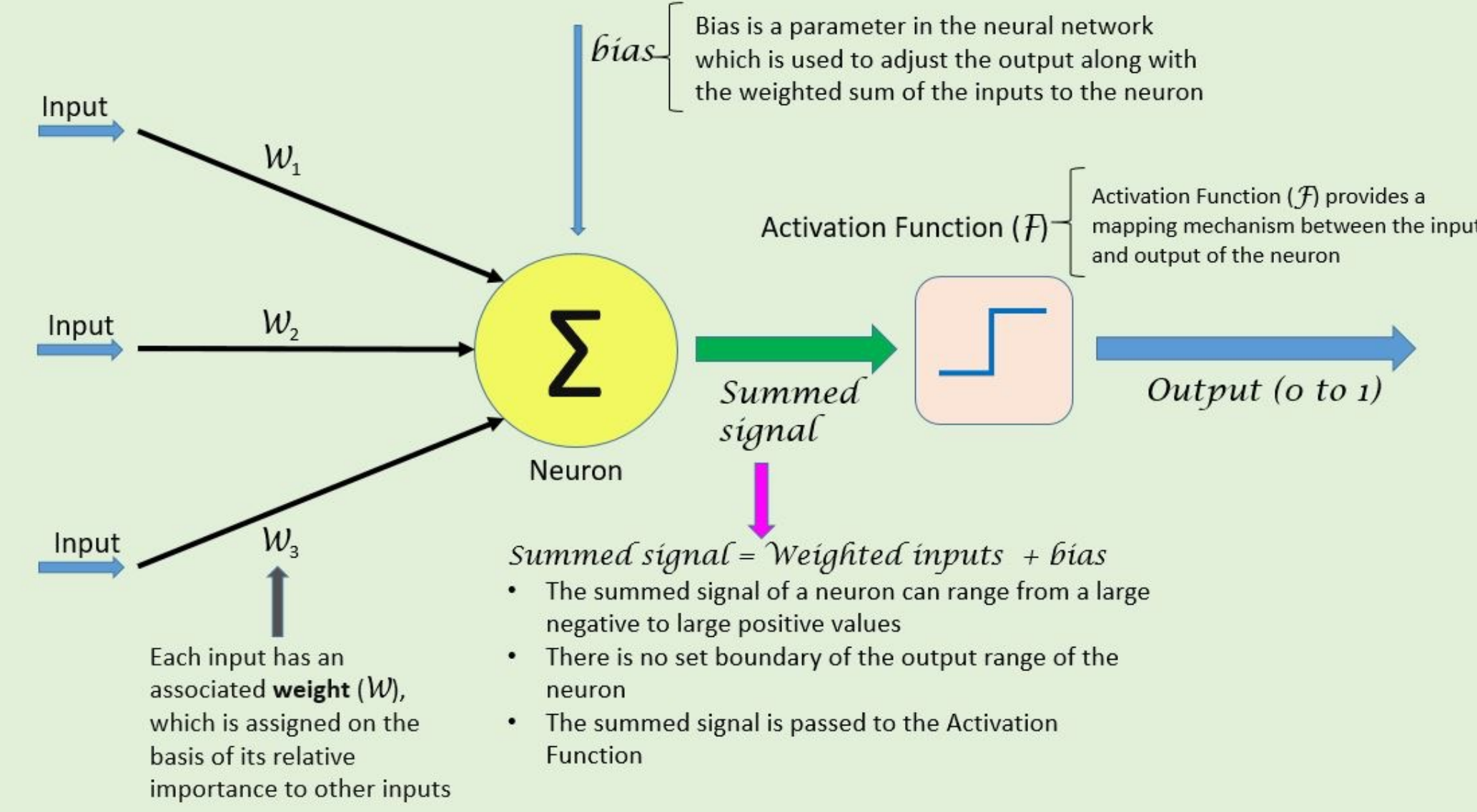
There are four major learning styles of an algorithm

- Supervised learning** – Input data is labelled and specific expected outputs are know. The algorithm generates reasonable predictions for the response to new data
  - Unsupervised learning** – Input data is not labeled and outputs are not known. The algorithm generates results based on the analysis structure of the data
  - Semi-supervised learning** – Input data is a mixture of labelled and unlabeled dataset. The algorithm generates results based on structure of the data
  - Reinforcement learning** – Input data is unlabeled. The machine (algorithm) is trained to make specific decisions. The machine is exposed to an environment where it trains itself continually using trial and error through a feedback loop
- Some practical applications of machine learning
- Image processing and computer vision for face recognition, motion detection, and object detection
  - Computational biology for tumor detection, drug discovery, and DNA sequencing

## What is Deep Learning?

- Deep learning is an artificial intelligence function that mimics the workings of the human thought process
- Deep learning is a subset of machine learning
- Deep learning utilizes a hierarchical level of artificial neural networks built with artificial neuron nodes connected together to form a topological network
- The hierarchical neural network reflects the ALGORITHM of the network in processing/transforming the data
- The word “deep” in “deep learning” refers to the number of layers through which the data is processed
- Deep learning algorithms propagates data through multiple layers of neurons, each of which passes a simplified representation of the data to the next layer
- Deep learning algorithm is capable of learning from the input data to improve its performance

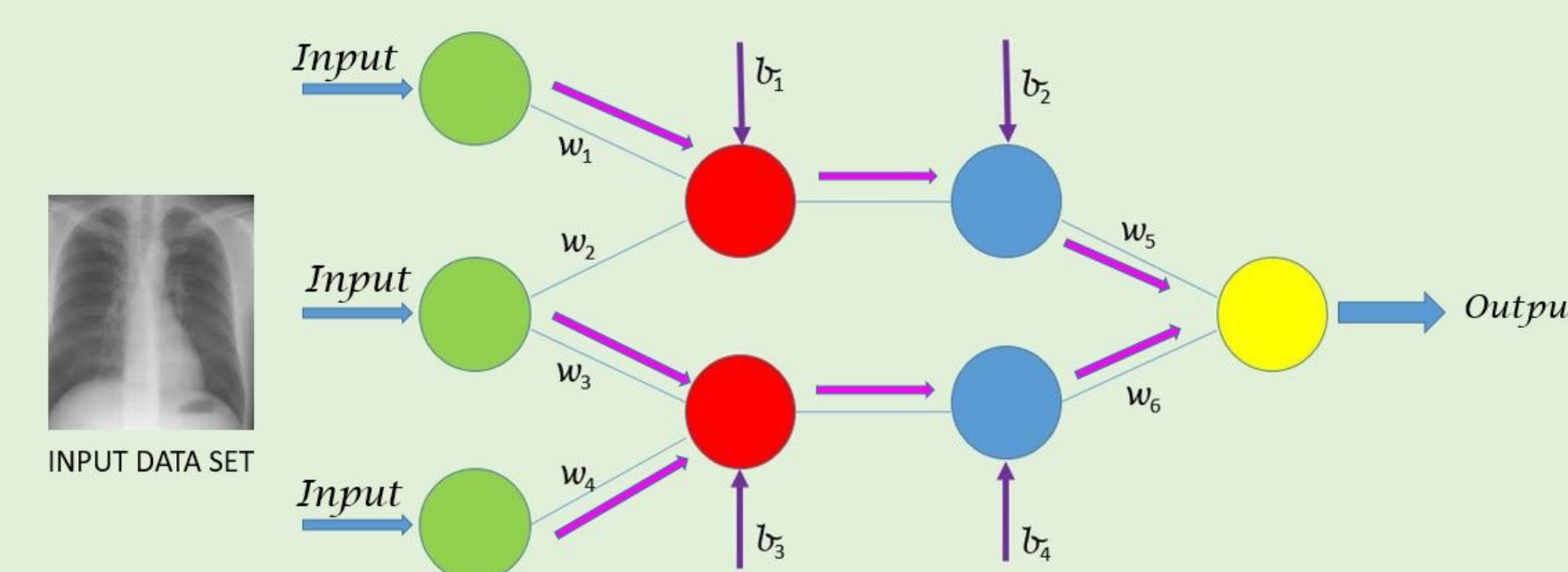
## Structure of an artificial neuron



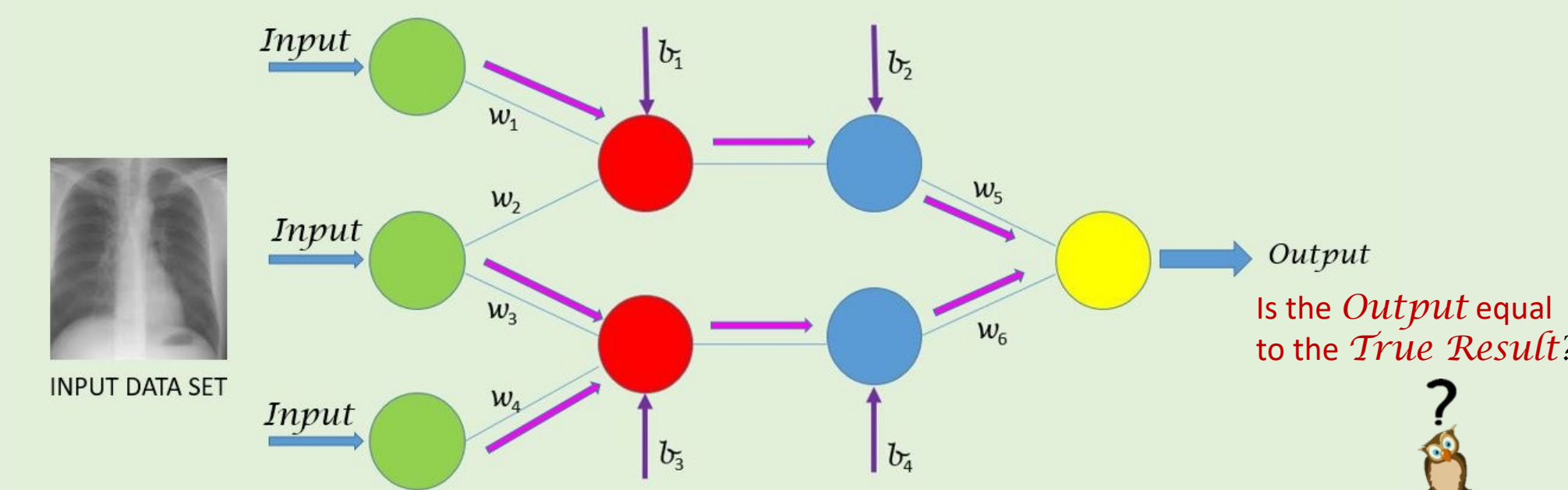
## How can we be assured that the network would perform and achieve the desired results?

- We need to train the network
- Untrained neural network model (newly designed and never been put to use) is like an inexperienced performer – It is essentially ignorant as it must learn to achieve satisfactory performance and yield desired results
- The algorithms of the networks must have actual experience of the intended task through the exposure to a large amount of data
- So by training a neural network on relevant datasets, we seek to improve its performance.
- The way we measure progress is by monitoring the error produced by the network each time it makes a prediction
- Therefore, a very important goal of the network is to reduce the error of the output – This is equivalent to LEARNING to improve itself

## Training of a deep learning neural network

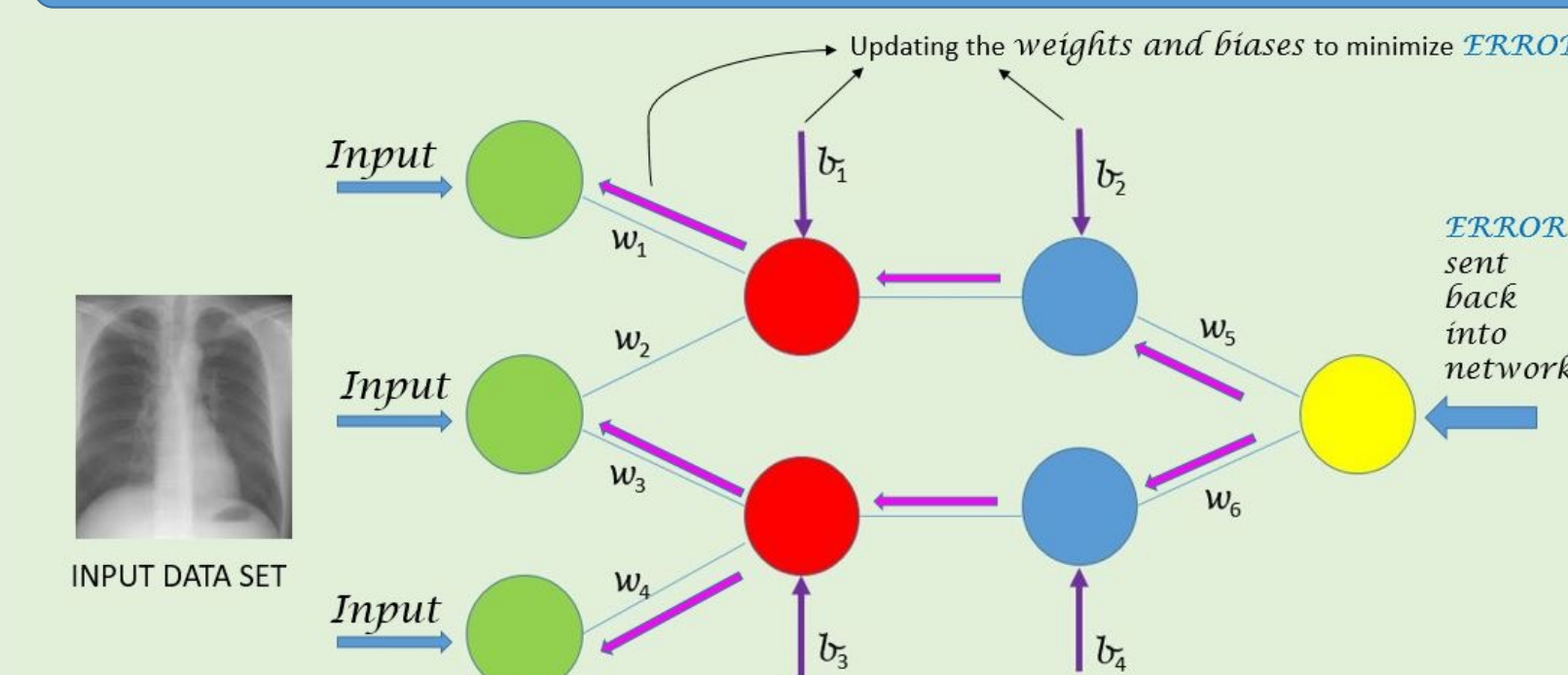


## Training of a deep learning neural network



- No, we do not expect the output to yield the True Result – there would be some expected error
- The error can be calculated  $\rightarrow ERROR = True\ Result - Output$
- We need adjust the network to minimize the ERROR – this is effectively improving the performance of the network

## Training of a deep learning neural network



- We will use a common technique in deep learning network to minimize the ERROR
- We send the ERROR back into the network
- The ERROR propagates backward through the network
- Updating the weights (w) and biases (b) to new values to minimize the ERROR
- This is the technique called Back Propagation

## What is Backpropagation?

- Backpropagation is the central mechanism by which neural networks learn
- Backpropagation is a smart algorithm that looks for the minimum value of the ERROR in changing the weights and biases
- The algorithm gets its name because the weights and biases are updated backwards, from output towards input
- Backpropagation employs a mathematical operation called Gradient Descent
- Gradient Descent calculates the new weights and biases that minimize the ERROR

## Concept of Gradient Descent

- Gradient Descent
  - A mathematical operation that computes the gradient (rate change) of the ERROR
  - It then attempts to find the “direction” to change the weights and biases to reduce the ERROR function most quickly and efficiently

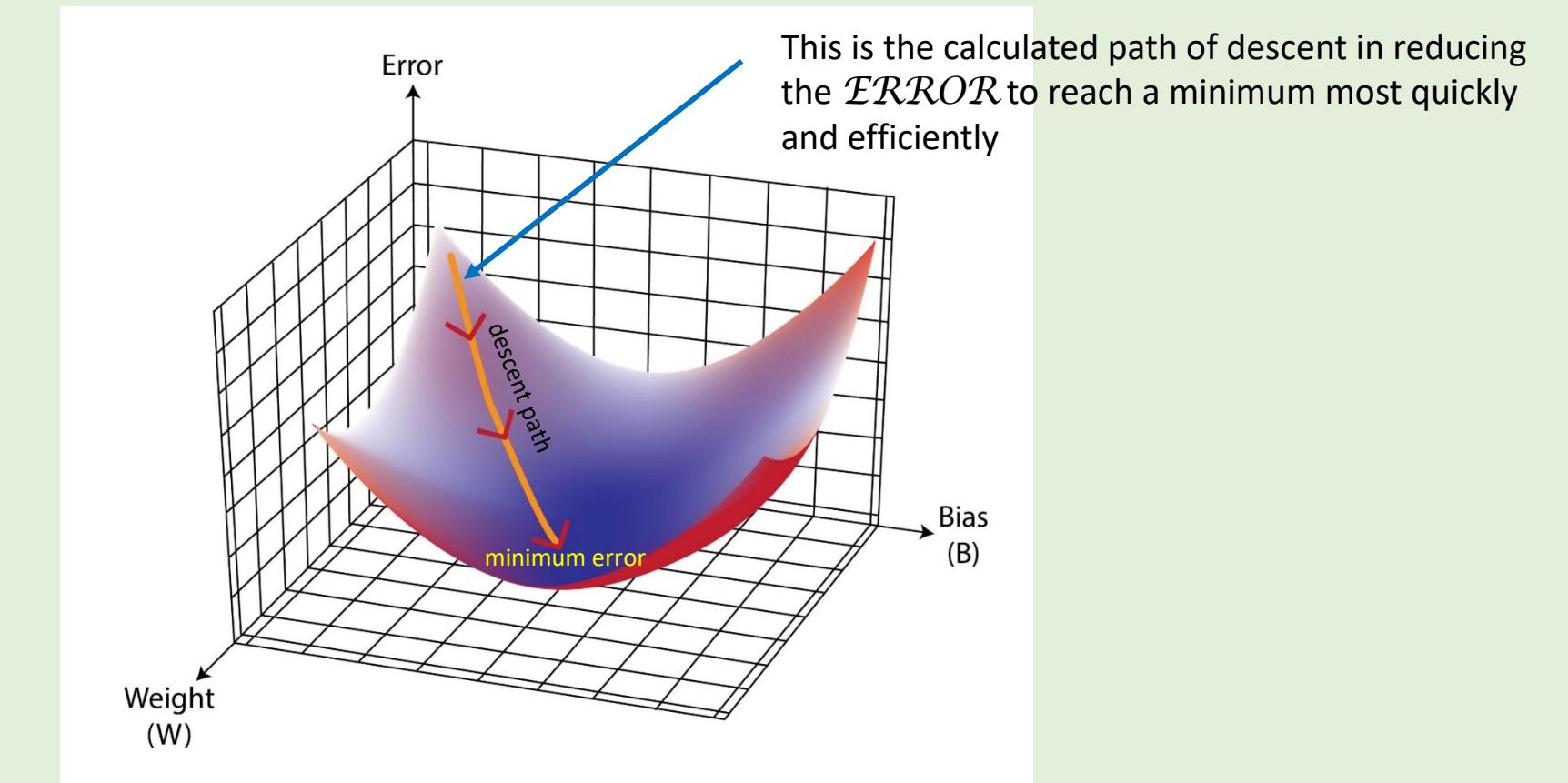


- In plain language, the Gradient Descent optimally relates change in the ERROR and the weights/biases of the network as follows

$$\text{Change in ERROR} = \left[ \frac{\text{Rate of ERROR change}}{\text{Rate of weight change}} \right] \times \text{Change in weight} + \left[ \frac{\text{Rate of ERROR change}}{\text{Rate of bias change}} \right] \times \text{Change in bias}$$

This is essentially a complex process of differential calculus  
But the bottom line to remember  
**CHANGE IN WEIGHTS AND BIASES RESULT IN A CHANGE IN ERROR**

## Concept of Gradient Descent

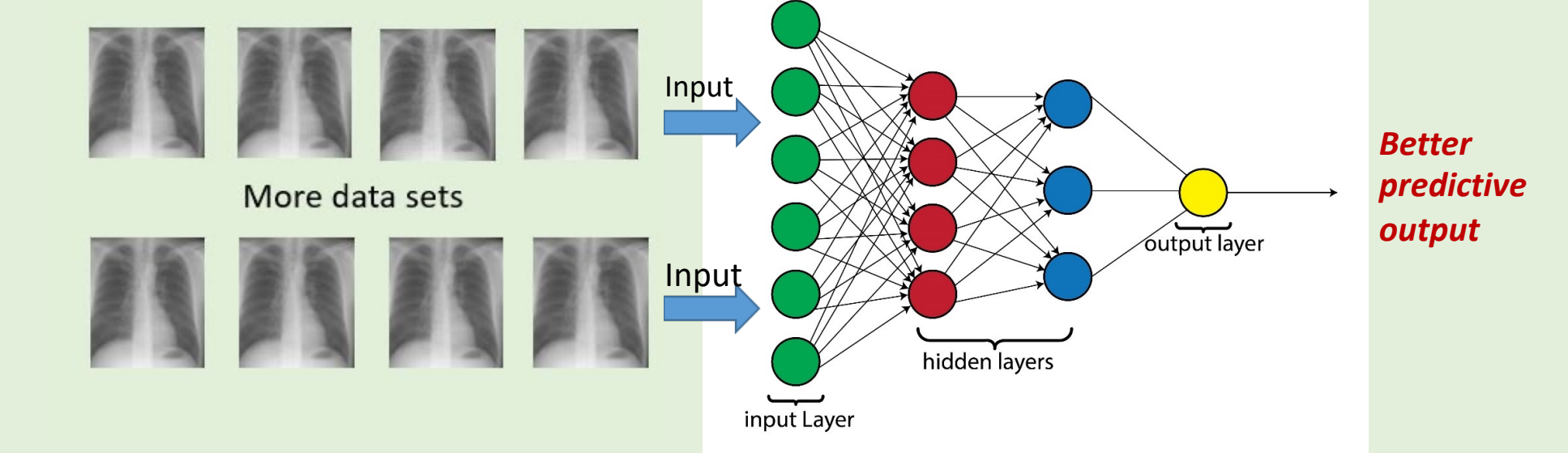


- Graph of the ERROR as a function of the weight and bias, now represented as a surface function of two variables
- Gradient Descent technique finds the direction to change the weights and biases to reduce the ERROR most quickly and efficiently (blue arrow)
- The new weights and biases associated with this descent are then the most optimal for error reduction

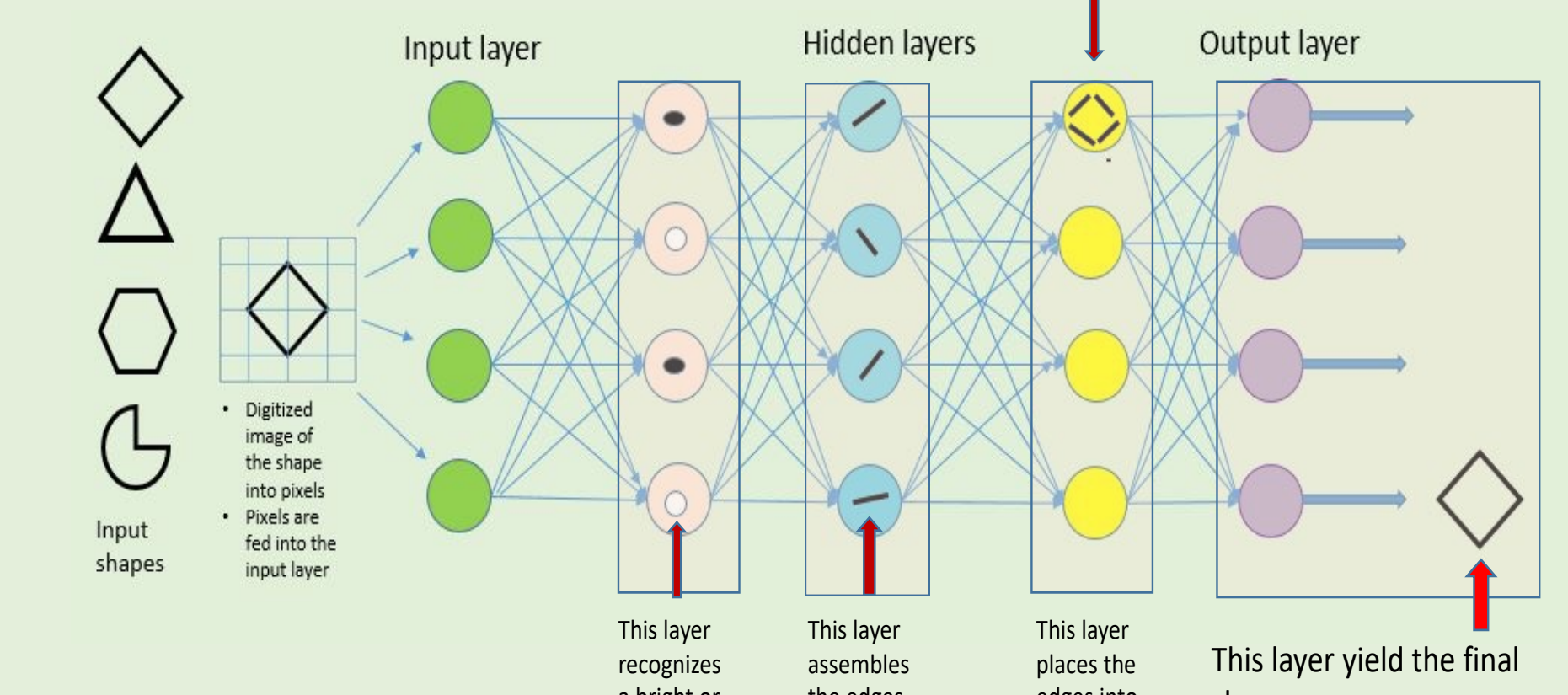
Do we stop now since we have done Backpropagation and found a set of weights biases to reduce the ERROR

No, absolutely not, our work is not done yet

- We will train the network with more data sets
- Each time a new data set is run through the network, the Backpropagation algorithm will improve the performance of the network
- The Gradient Descent force updates of the weights and biases towards a smaller ERROR
- In another word, the network continues to LEARN and improves its predictive output



## Illustration of hierarchical feature of a neural network in deep learning



Let's walk through an example to see how a network can recognize a shape

## FINAL NOTE

- Deep learning utilizes a hierarchical neural network with cascading architecture
- The cascading architecture consists of multiple levels of neural network structure
- The outputs of the hidden neurons in the higher hierarchical level are treated as an equivalent input data to the input neurons at the lower hierarchical level
- Each level of neuron layer learns to transform its input data into a slightly more abstract and composite representation

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