

# Foundations of PyTorch

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GETTING STARTED WITH PYTORCH FOR MACHINE LEARNING



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[www.loonycorn.com](http://www.loonycorn.com)

# Overview

**Deep learning using neural networks**

**Neurons and activation functions**

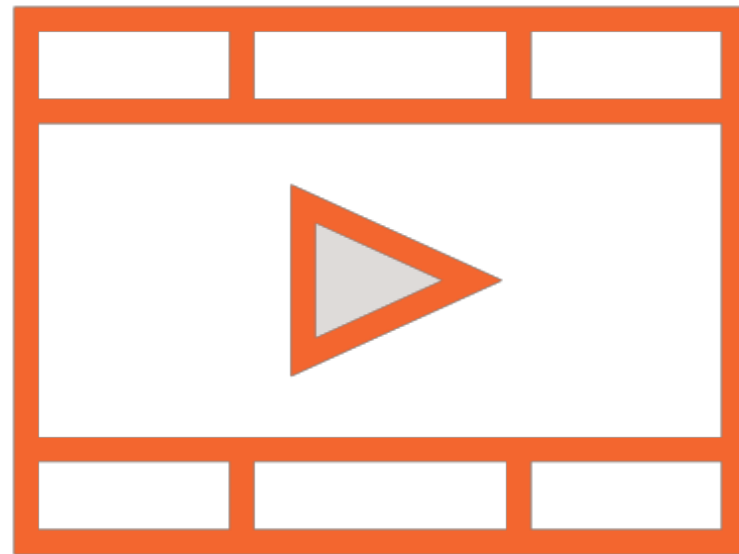
**Introducing PyTorch to build neural networks**

**Understanding the differences between PyTorch and TensorFlow**

# Prerequisites and Course Outline

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# Prerequisites



**Working with Python and Python libraries**

**Basic understanding of machine learning algorithms**

**No prior experience with neural networks necessary**

# Prerequisites



**Understanding Machine Learning by  
David Chappell**

**Understanding Machine Learning with  
Python by Jerry Kurata**

**Building Machine Learning Models in  
Python with scikit-learn by Janani Ravi**

# Course Outline



## **Getting started with PyTorch**

- Introducing neural networks and PyTorch
- Tensor operations and CUDA support

## **Gradients and the Autograd library**

- Gradient descent to train NNs
- Working with gradients in PyTorch

## **Dynamic computation graphs**

- Pros and cons of working with each
- Static graph in TF vs. dynamic graph in PyTorch

# Introducing Neural Networks

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# Reviews: Positive or Negative?





# ML-based Classifier

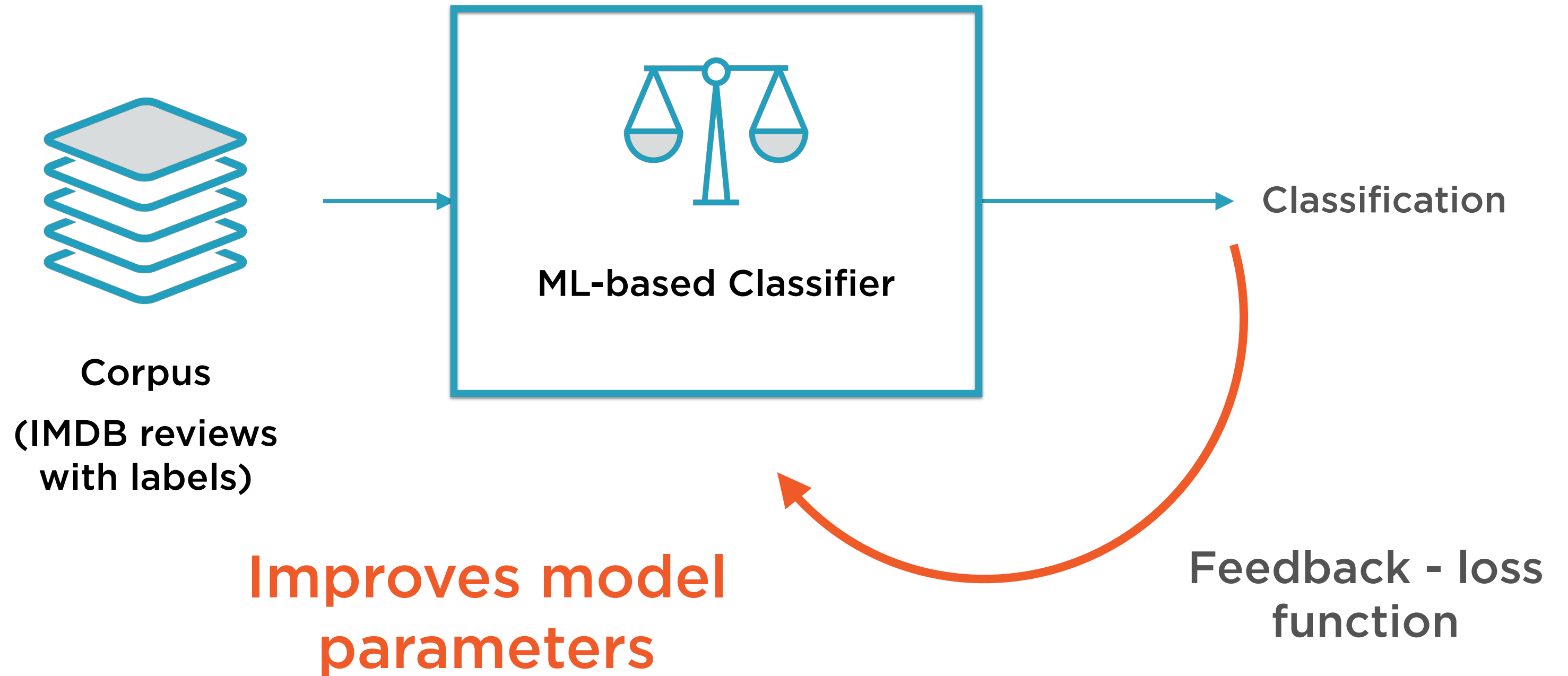
## Training

Feed in a large corpus of data  
classified correctly

## Prediction

Use it to classify new instances  
which it has not seen before

# Training the ML-based Classifier



**“Traditional”** ML-based systems  
rely on experts to decide what  
features to pay attention to

**“Representation”** ML-based  
systems figure out by themselves  
what features to pay attention to

Neural networks are examples  
of such systems

# What is a Neural Network?

## Deep Learning

Algorithms that learn  
what features matter

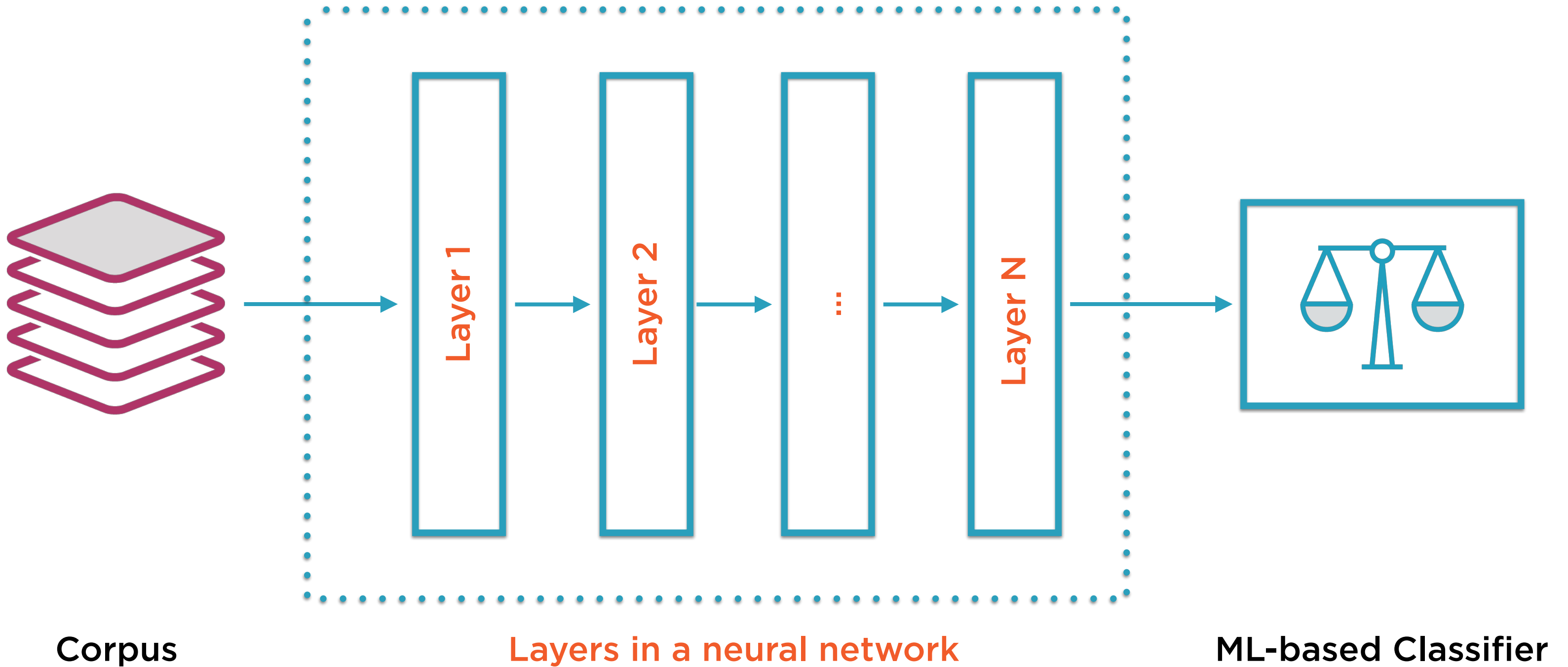
## Neural Networks

The most common class  
of deep learning  
algorithms

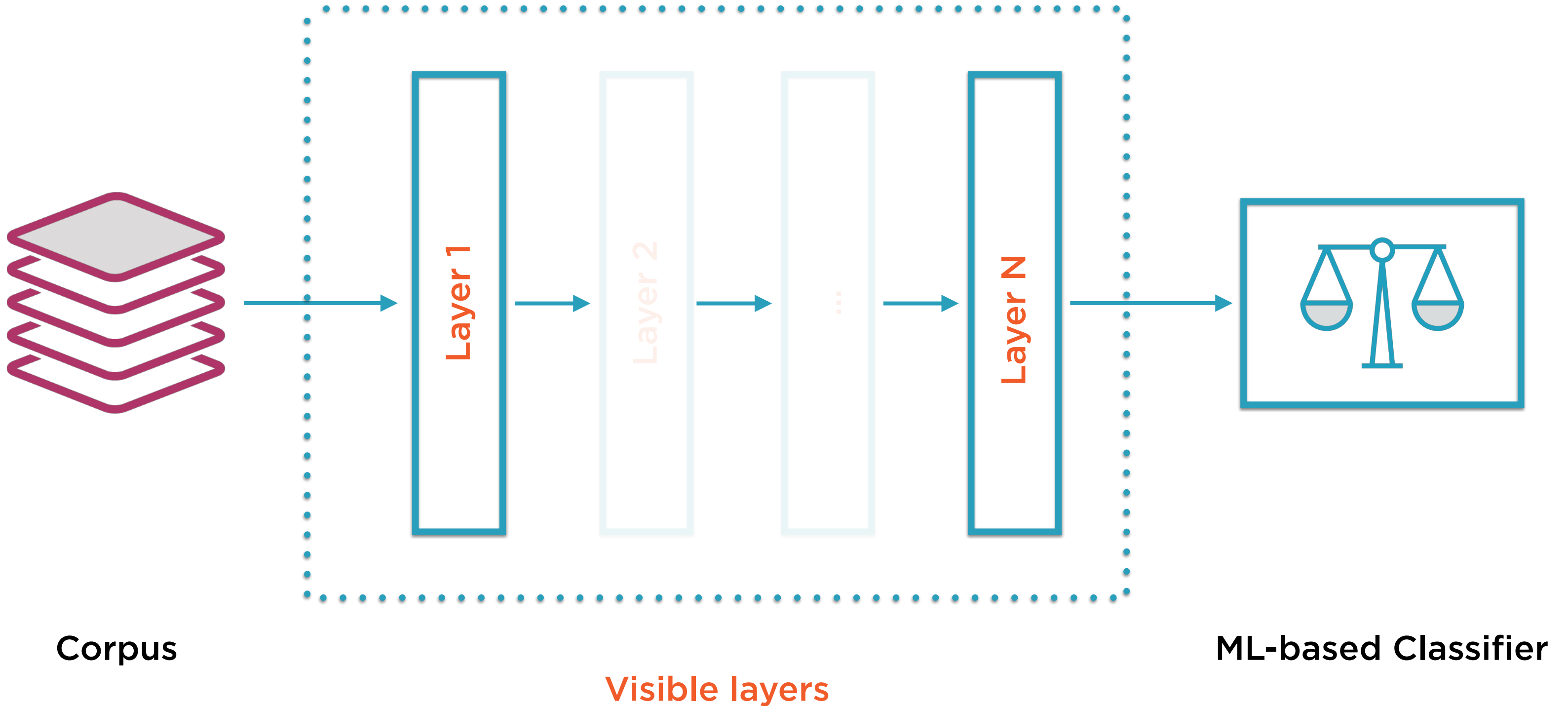
## Neurons

Simple building blocks  
that actually “learn”

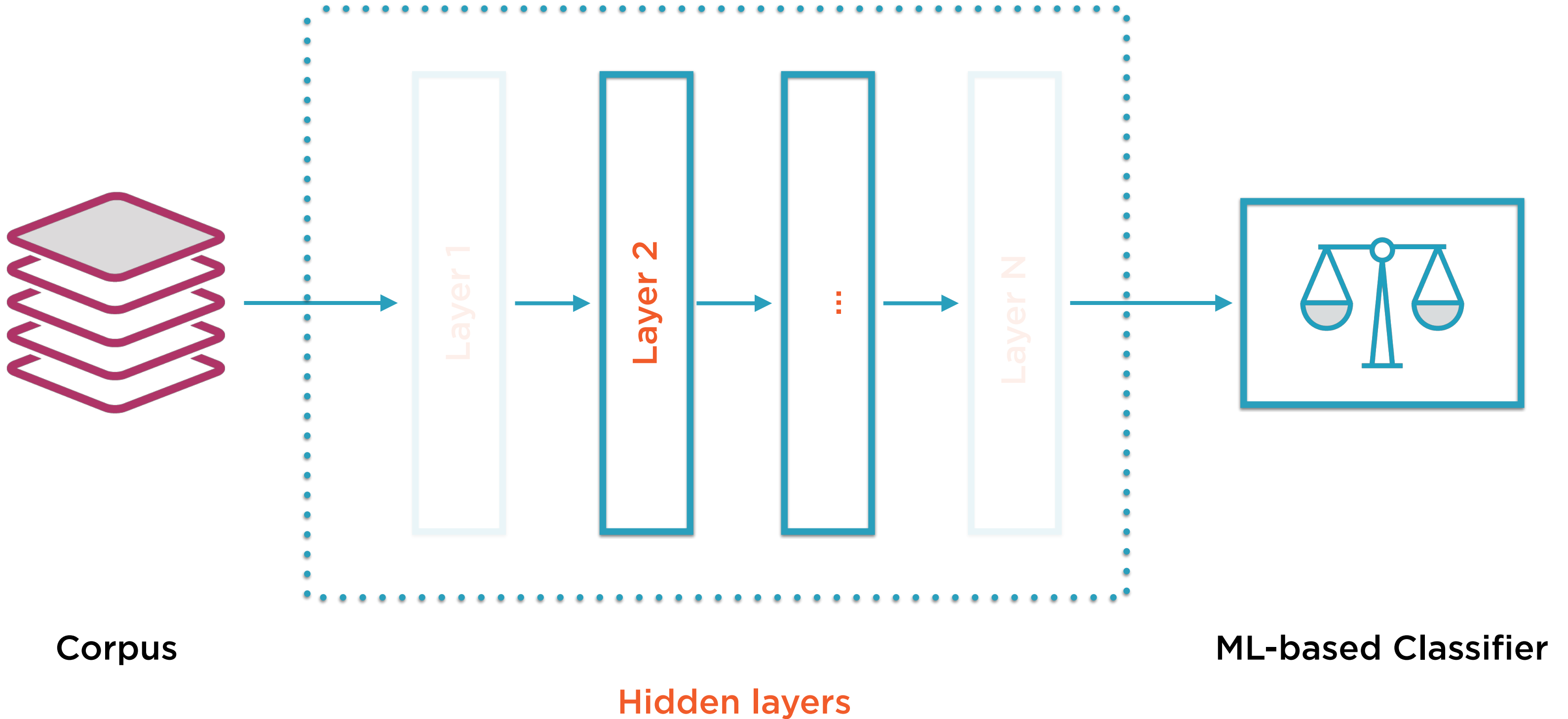
# Neural Networks



# Neural Networks

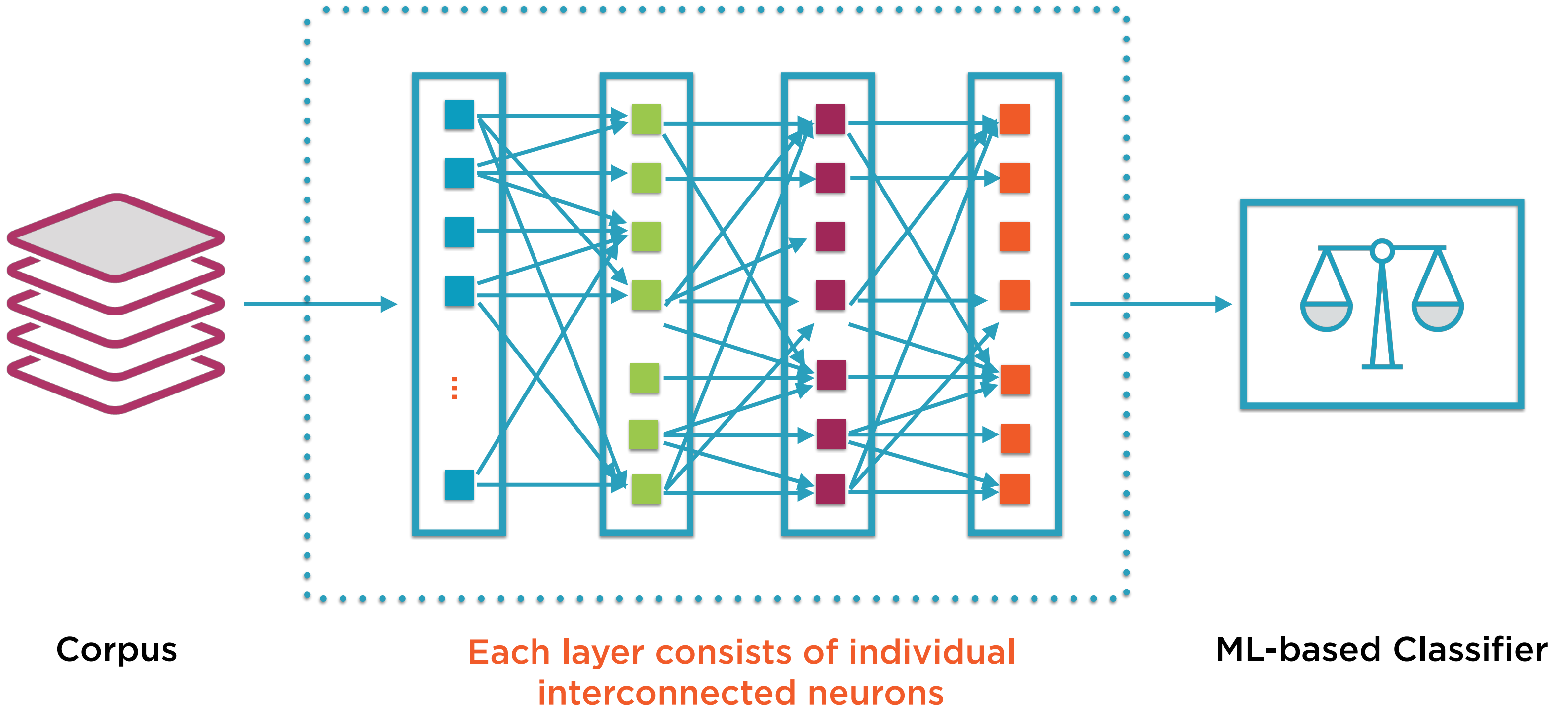


# Neural Networks





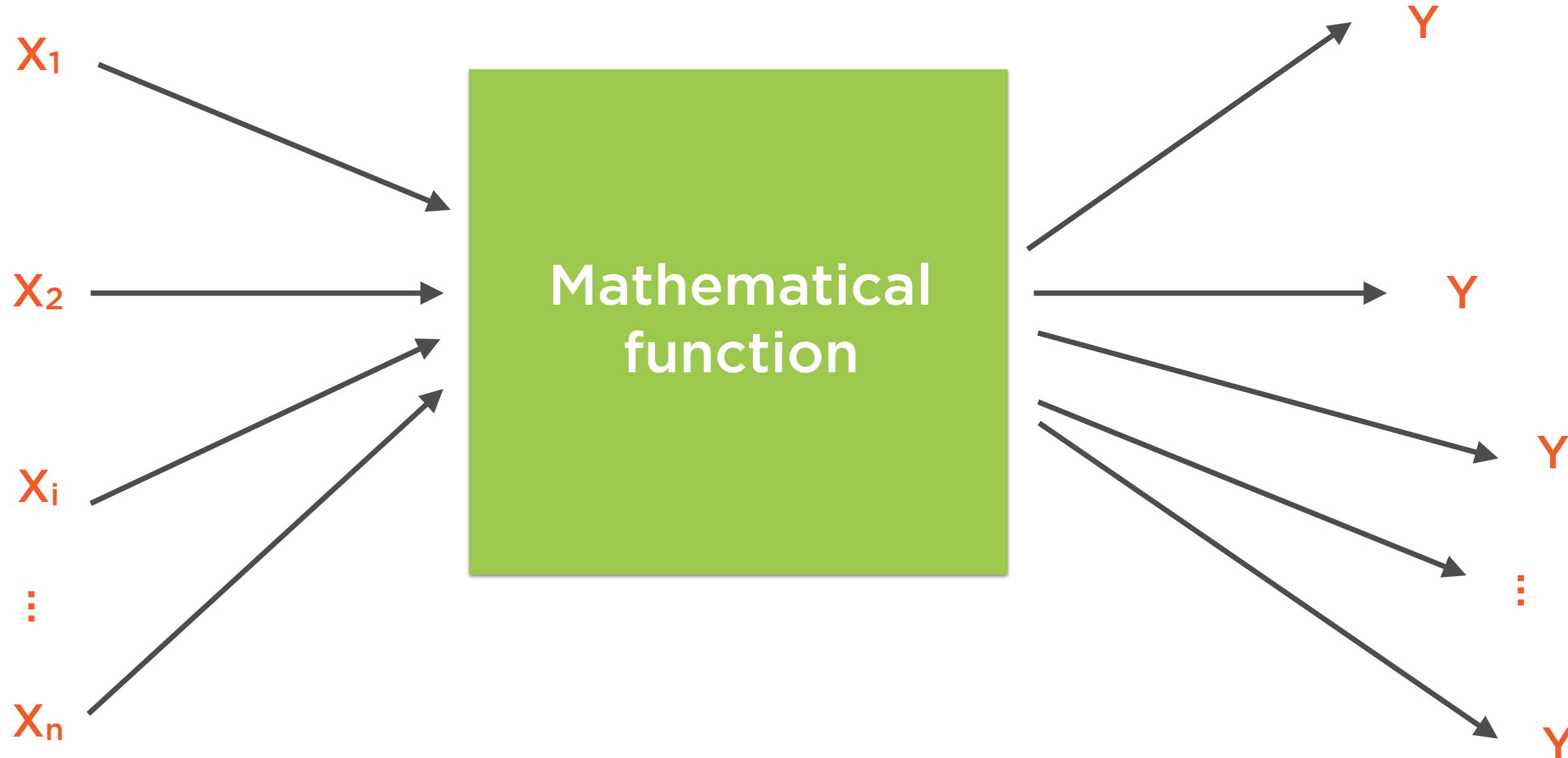
# Neural Networks



# Neurons and Activation Functions

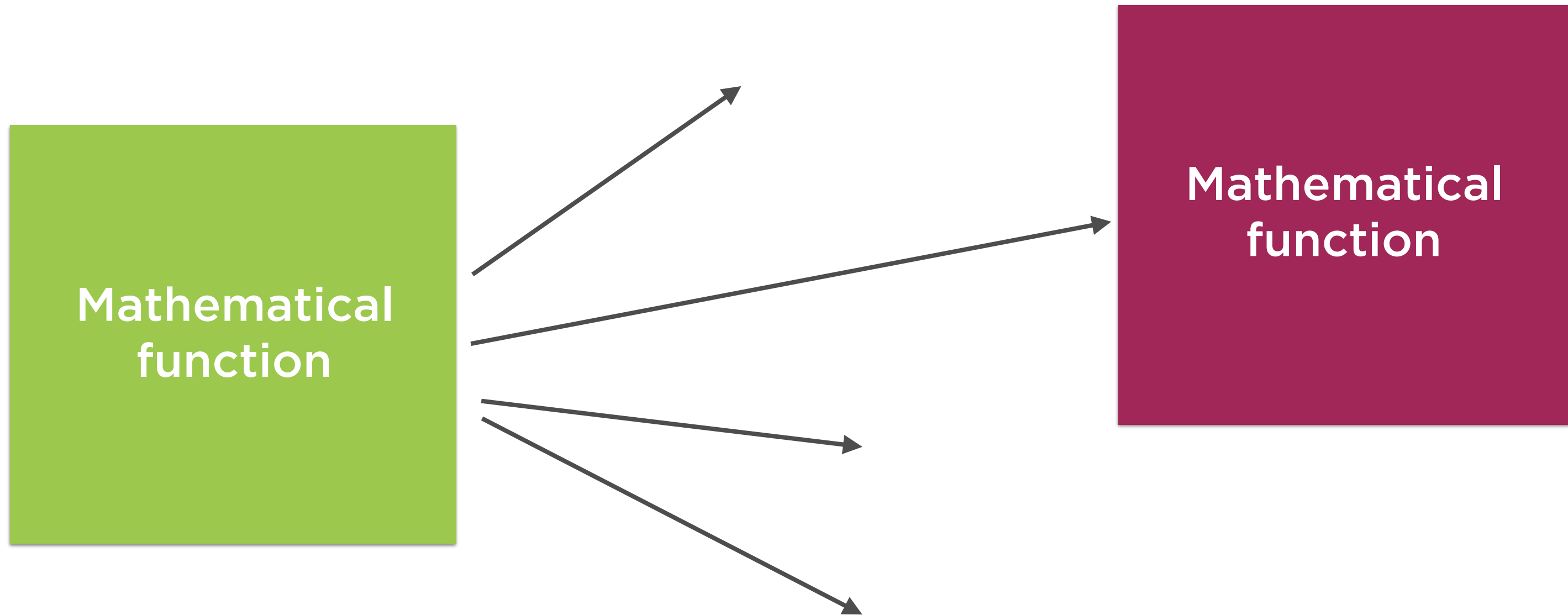
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# Operation of a Single Neuron



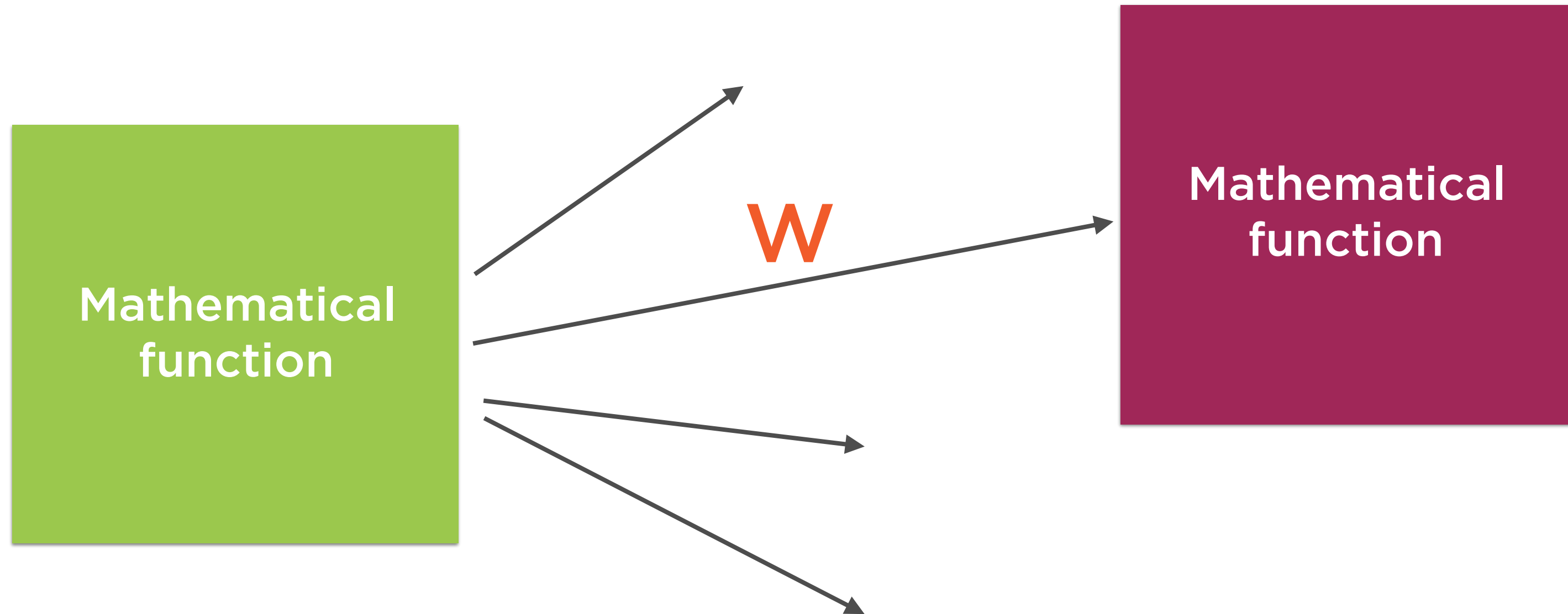
**For an active neuron a change in inputs should trigger a corresponding change in the outputs**

# Operation of a Single Neuron



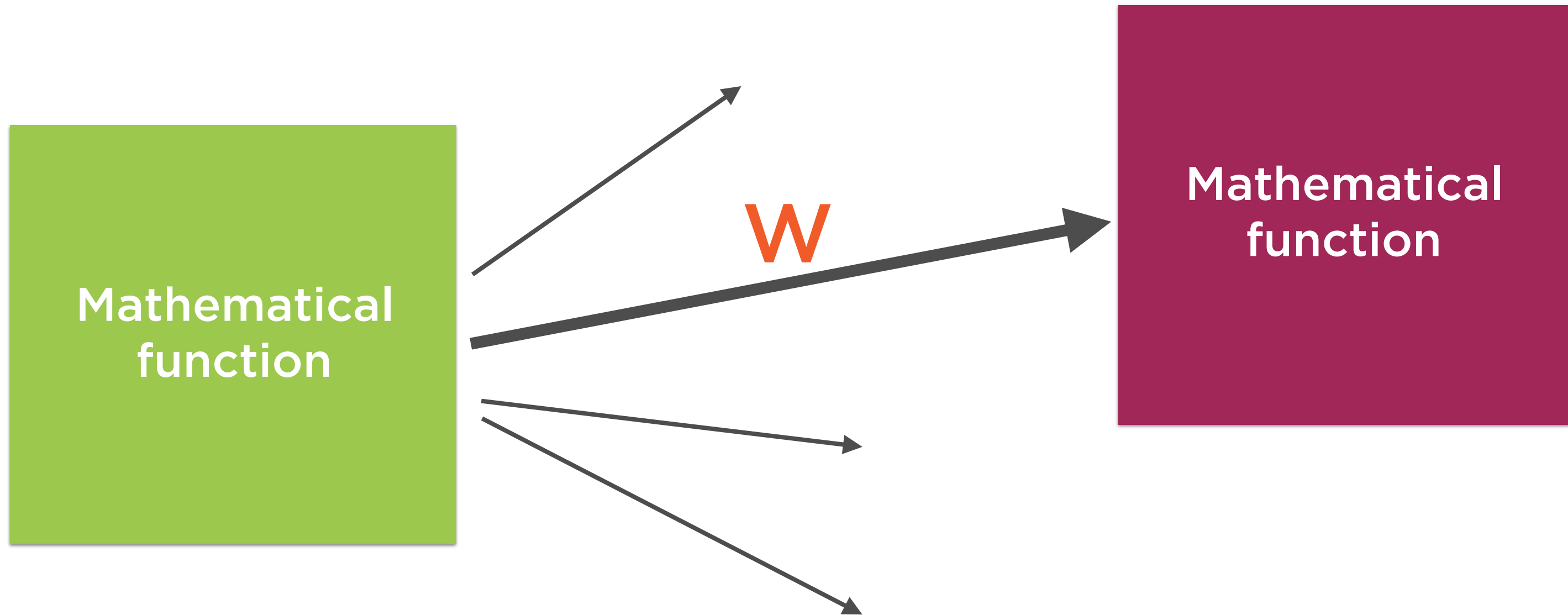
**The outputs of neurons feed into the neurons from the next layer**

# Operation of a Single Neuron



**Each connection is associated with a weight**

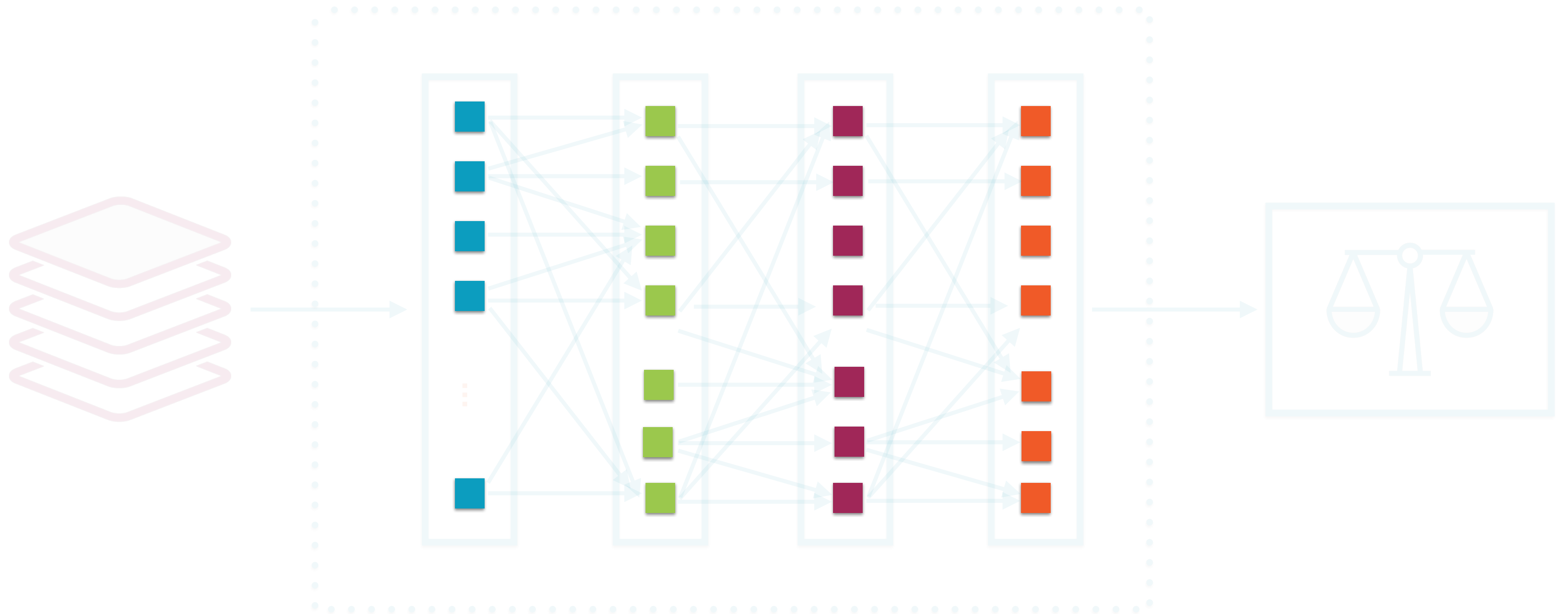
# Operation of a Single Neuron



If the second neuron is sensitive to the output of the first neuron, the **connection between them gets stronger**

**$W$  increases**

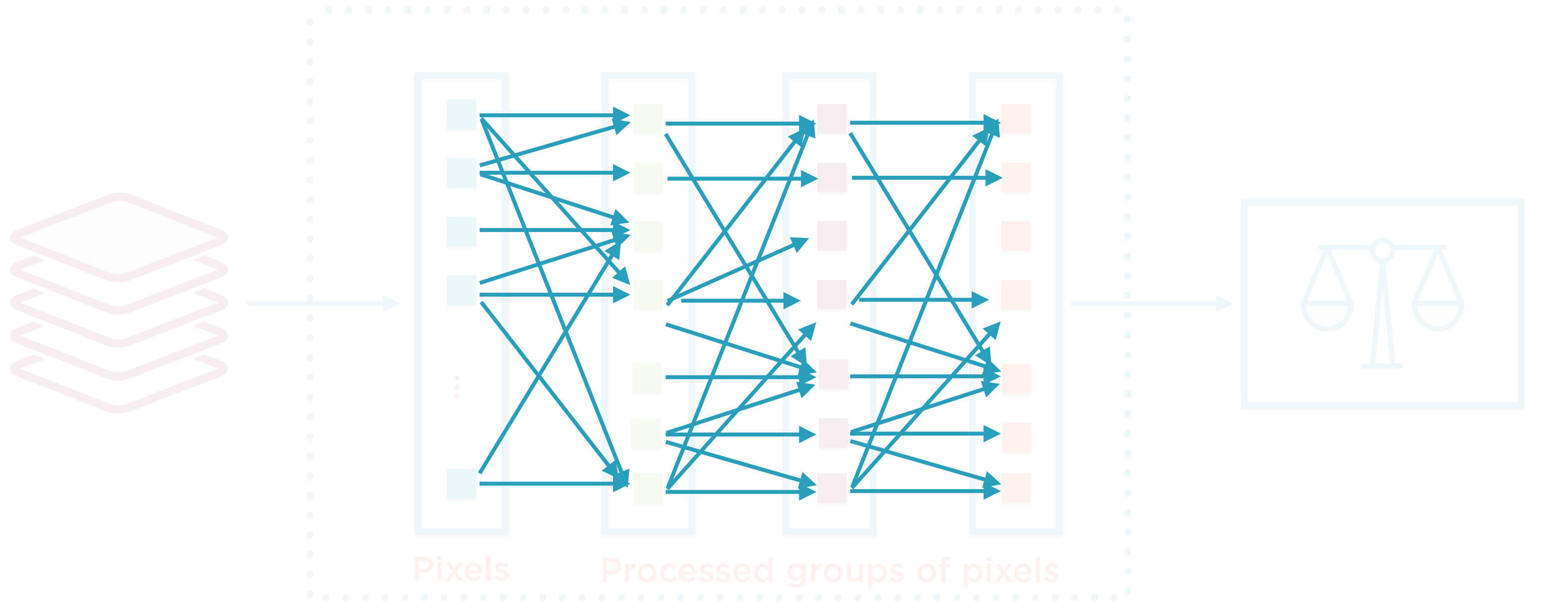
# The Computational Graph



Corpus

**The nodes in the computation graph are** ML-based Classifier  
**neurons (simple building blocks)**

# The Computational Graph



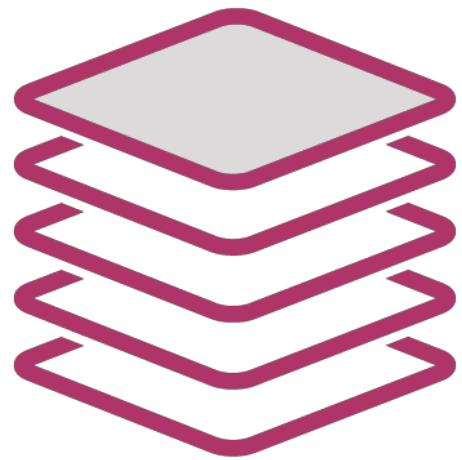
Corpus

**The edges in the computation graph  
are data called tensors**

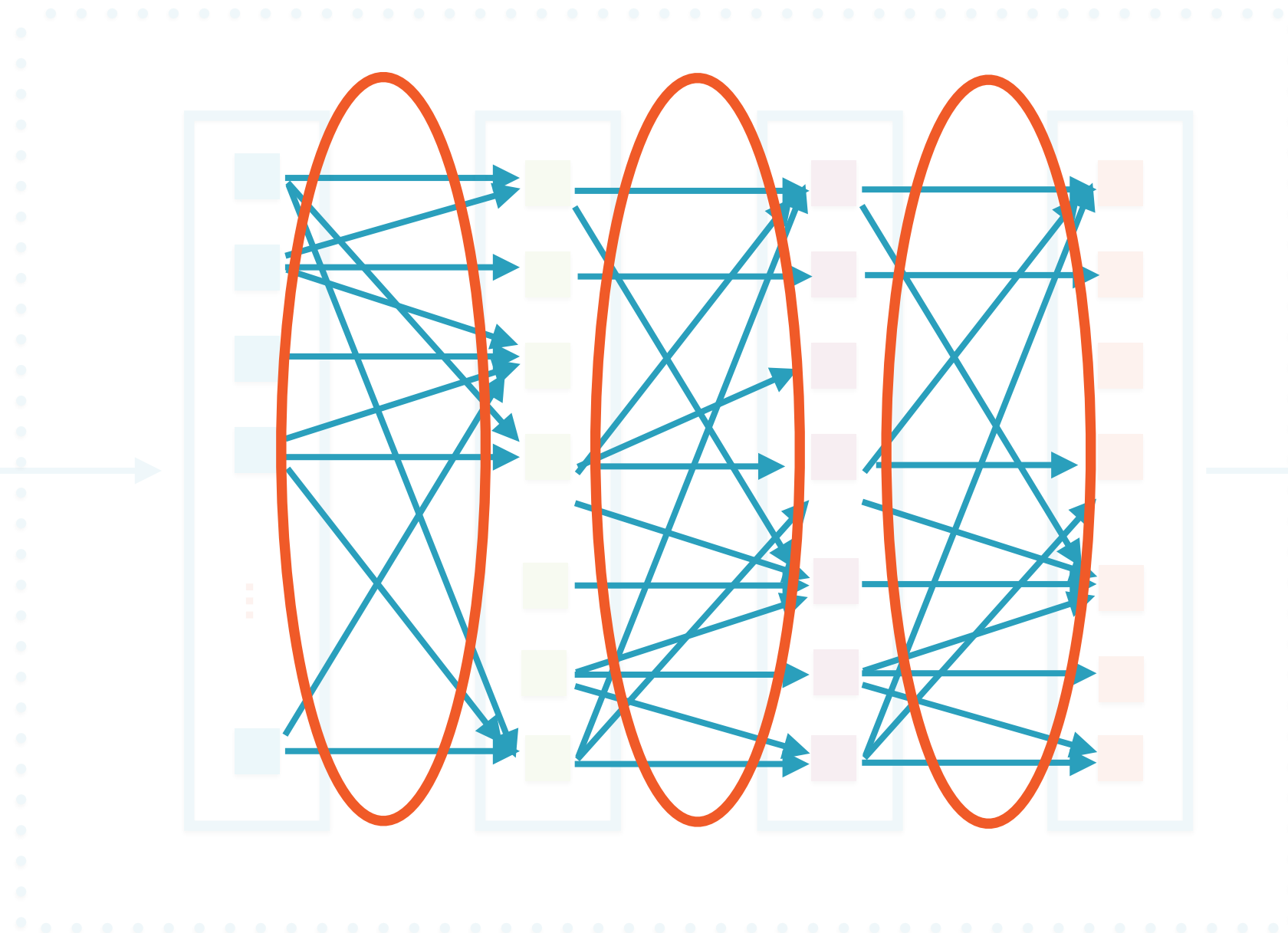
ML-based Classifier



# A Neural Network

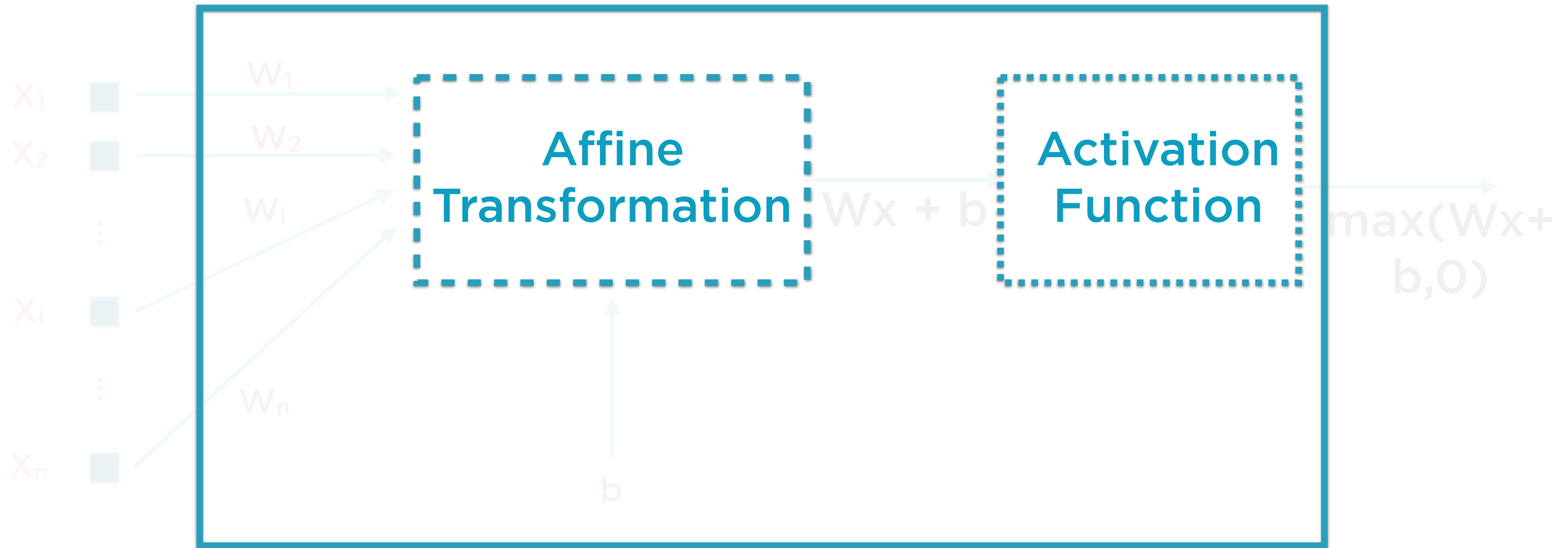


Corpus



Once a neural network is **trained**, all edges have weights which help it make predictions

# Operation of a Single Neuron



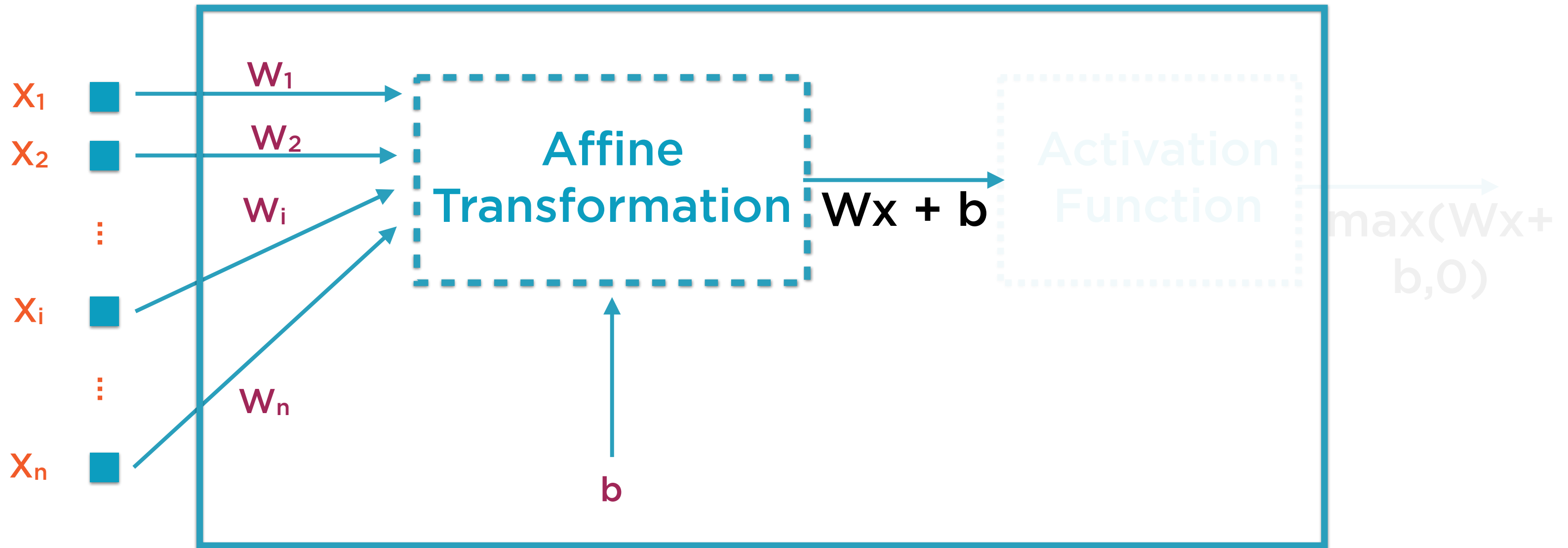
**Each neuron only applies two simple functions to its inputs**

# Affine Transformation



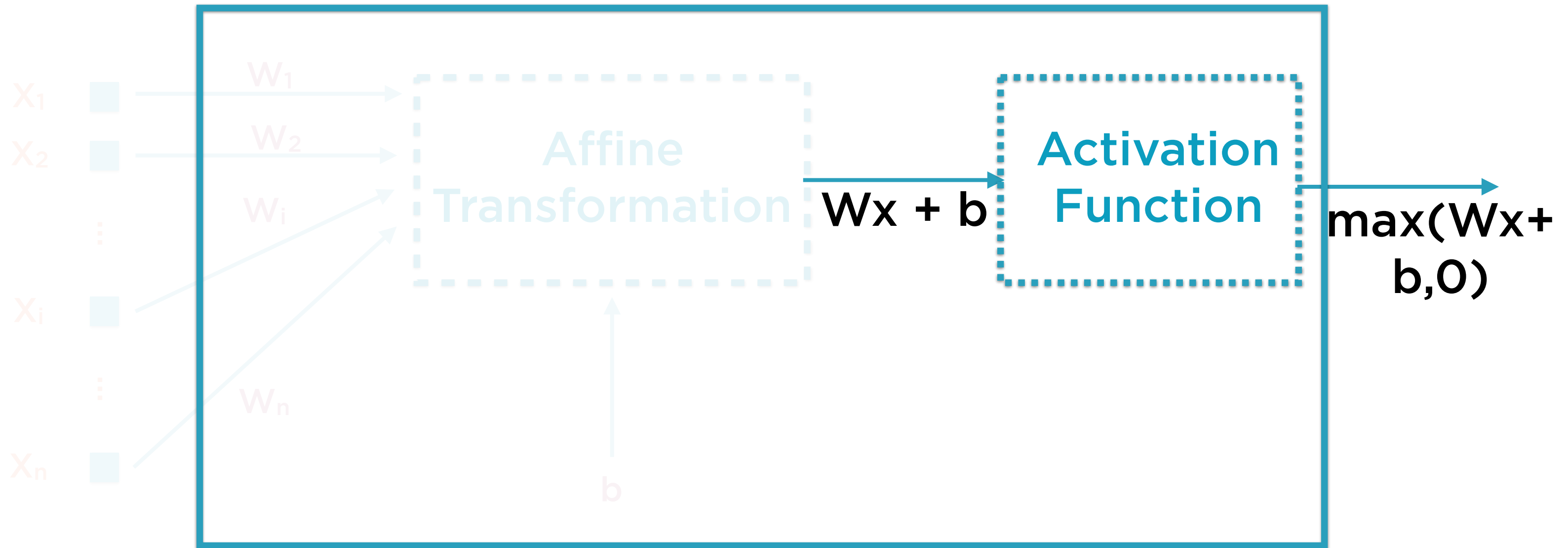
The affine transformation alone can **only** learn **linear** relationships between the inputs and the output

# Affine Transformation



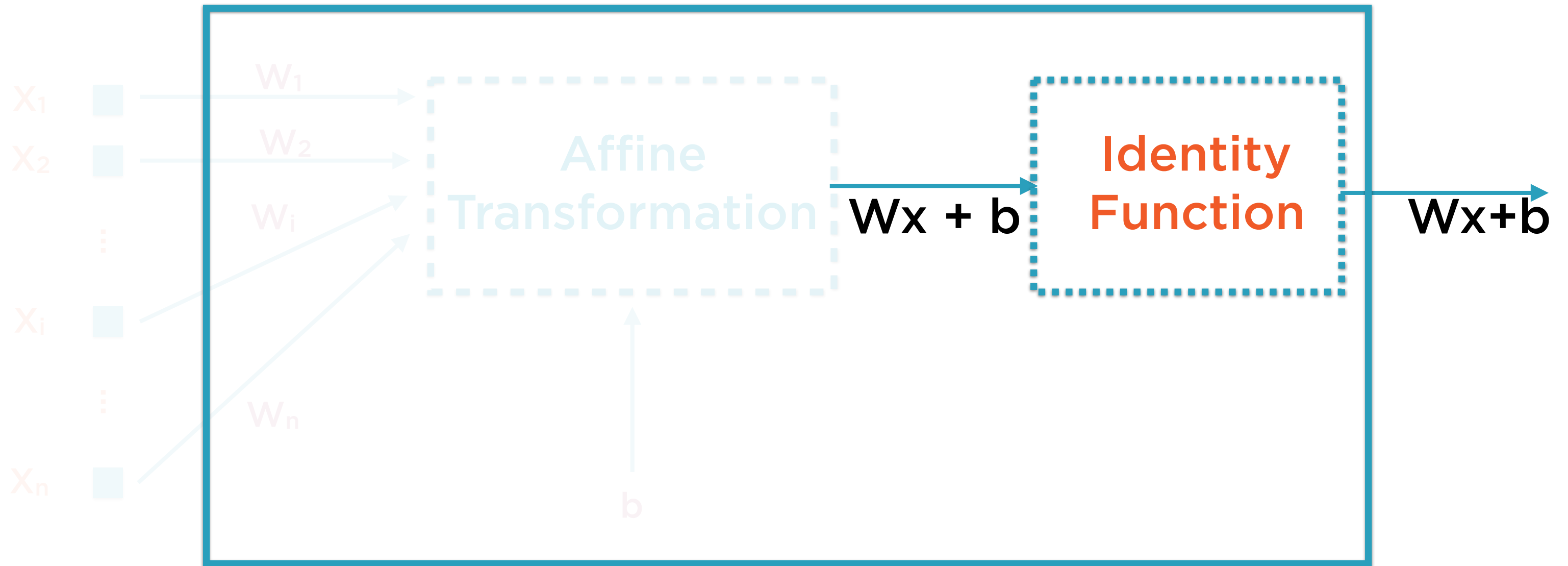
The affine transformation is just a weighted sum with a bias added:  $w_1x_1 + w_2x_2 + \dots + w_nx_n + b$

# Activation Function



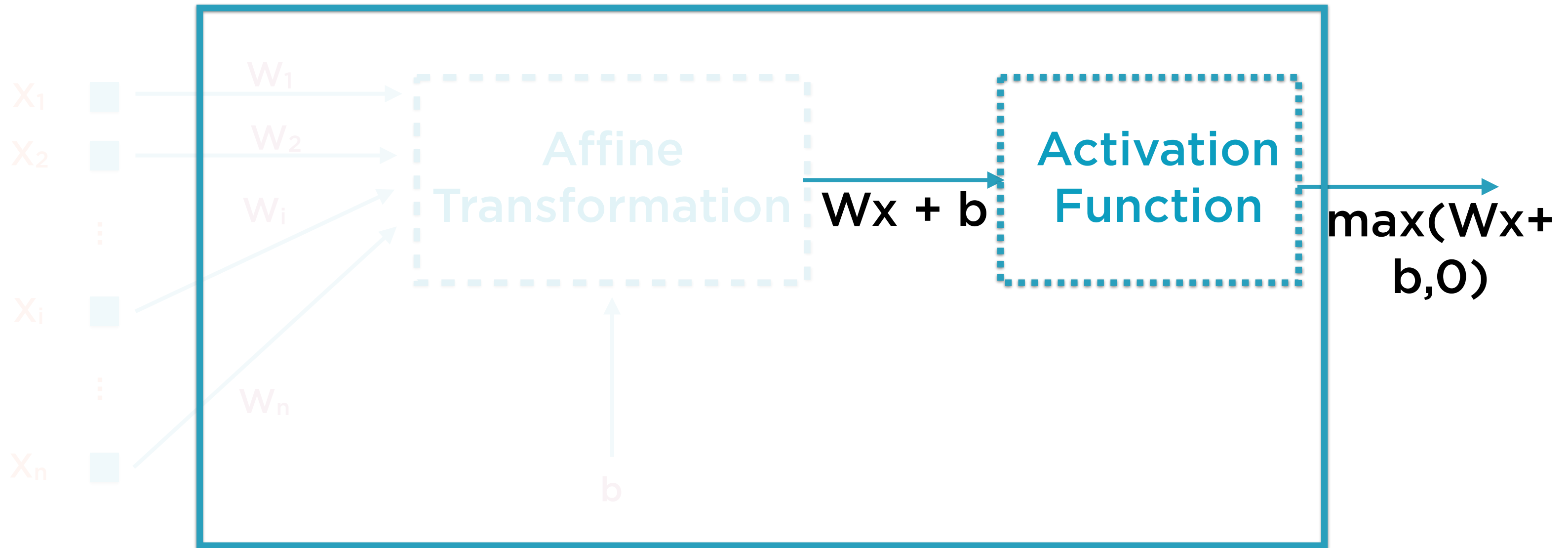
**A function which helps discover non-linear relationships**

# Linear Neuron



**When the activation function is the identity function, the neuron is often referred to as a linear neuron**

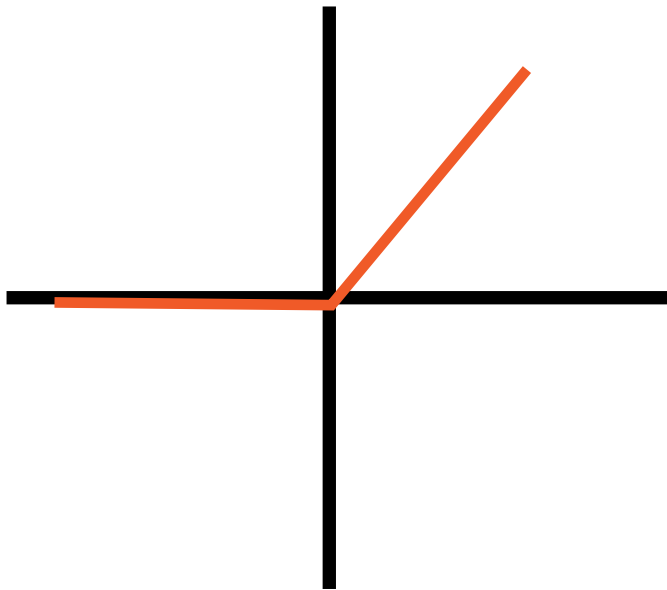
# Activation Function



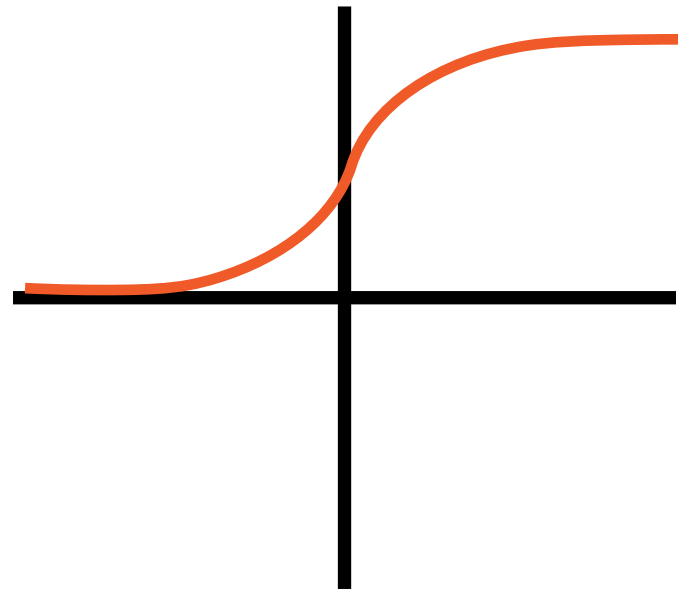
The **combination** of the affine transformation and the activation function can **learn any arbitrary relationship**

# Common Activation Functions

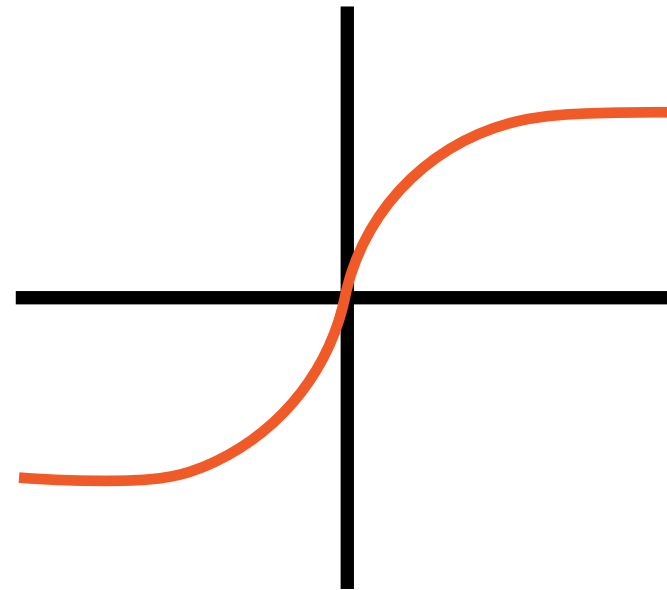
**ReLU**



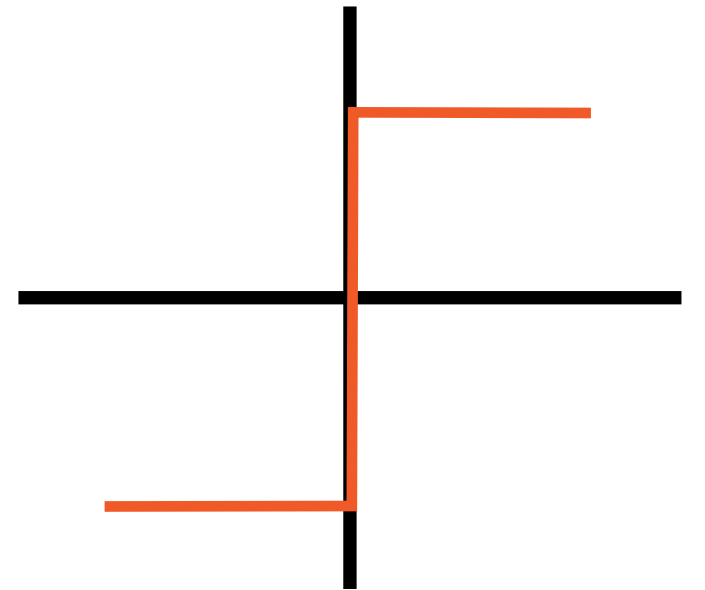
**logit**



**tanh**

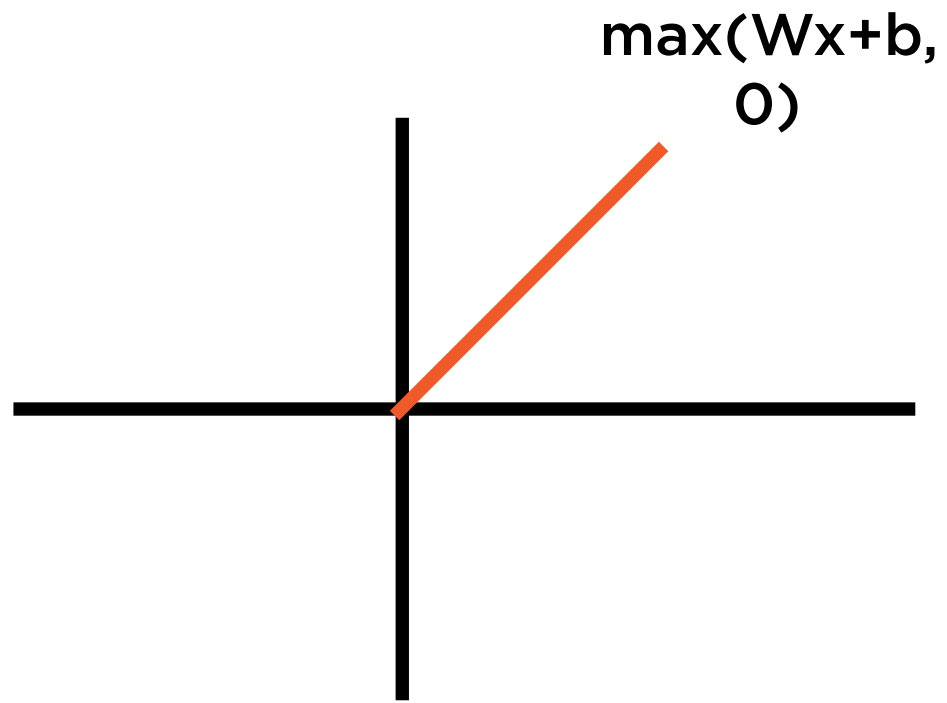


**step**





# ReLU Activation

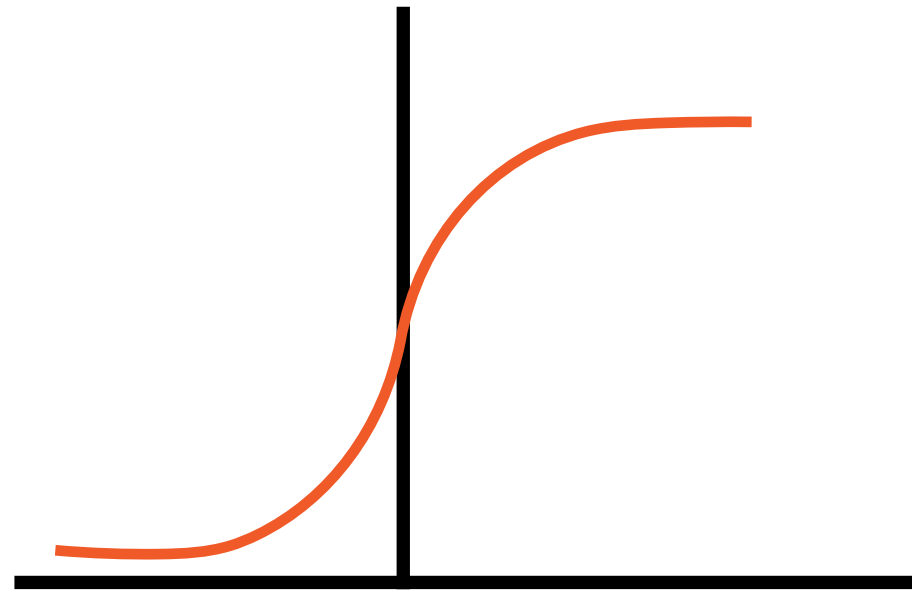


The most common form of the activation function is the ReLU

ReLU : Rectified Linear Unit

$$\text{ReLU}(x) = \max(0, x)$$

# SoftMax Activation



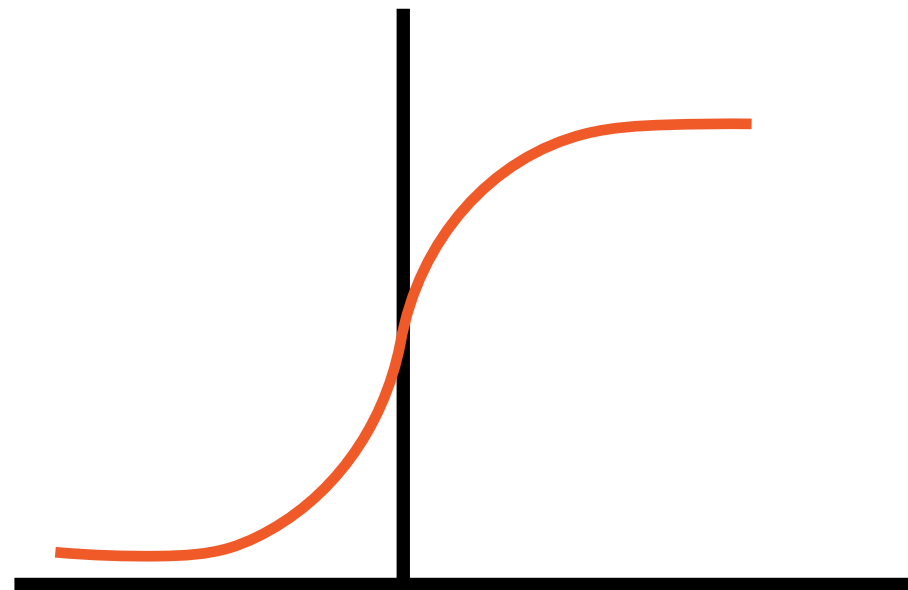
Another very common form of the activation function is the SoftMax

SoftMax(x) outputs a number between 0 and 1

This output can be interpreted as a **probability**

This curve is also called a logit curve

# Importance of Activation

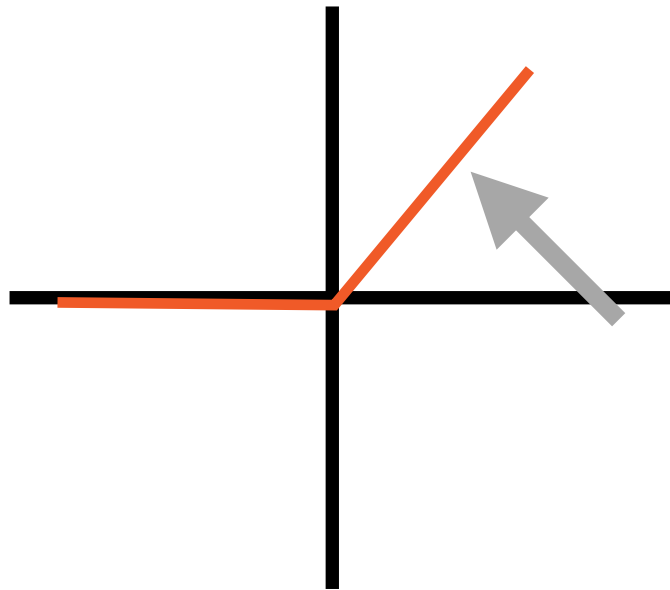


**The choice of activation function is crucial in determining performance**

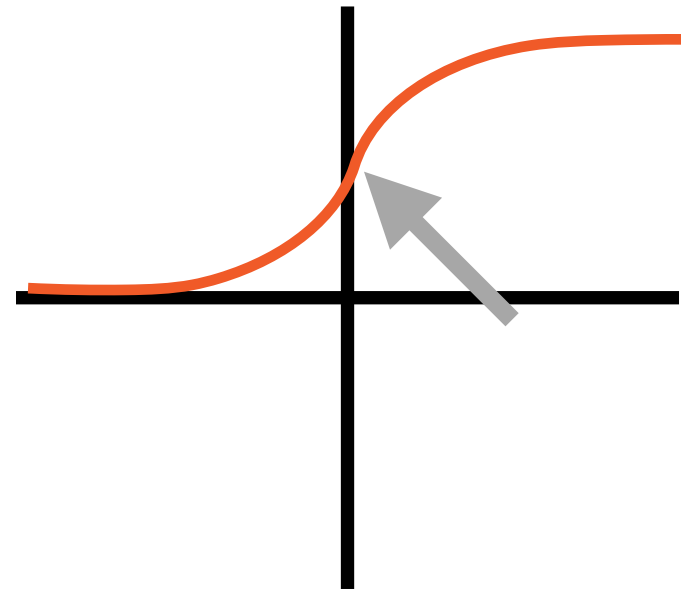
**To see why, we must understand the training process of a Neural Network**

# Active Region

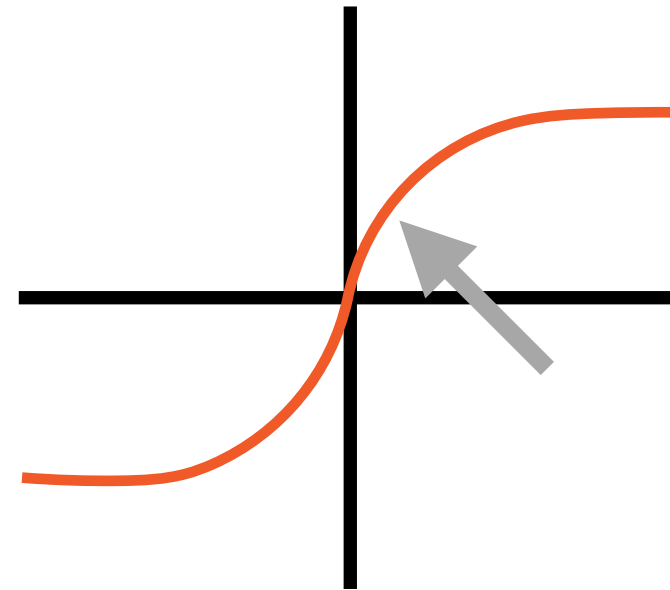
ReLU



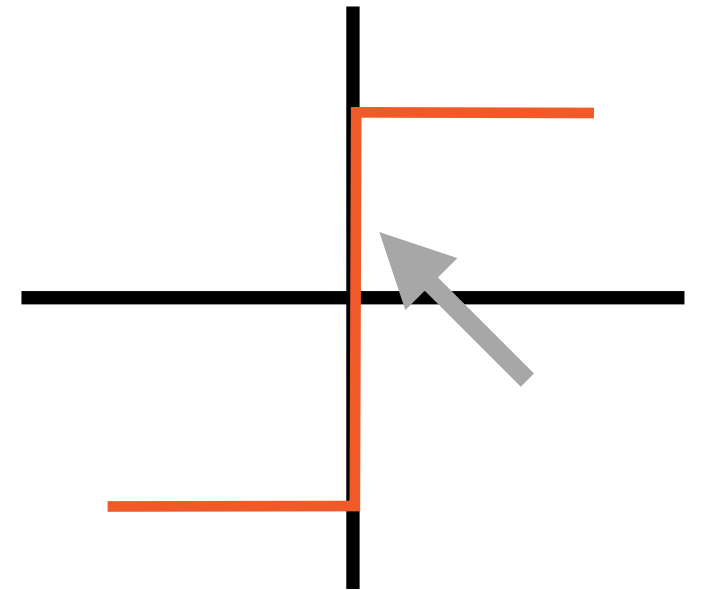
logit



tanh



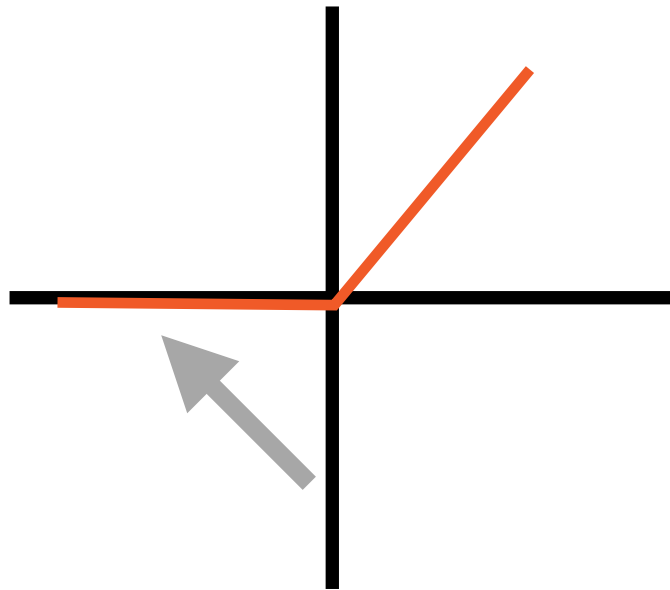
step



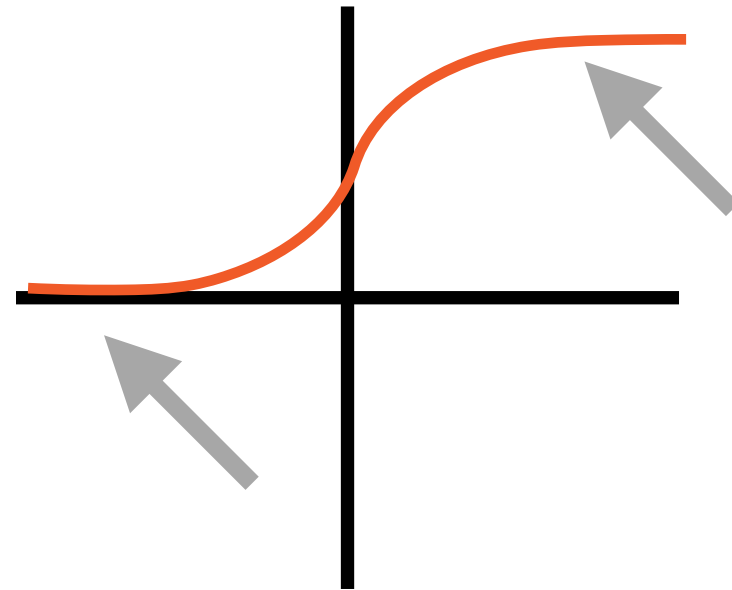
Notice how activation functions have a gradient, this gradient allows them to be sensitive to input changes

# Saturation

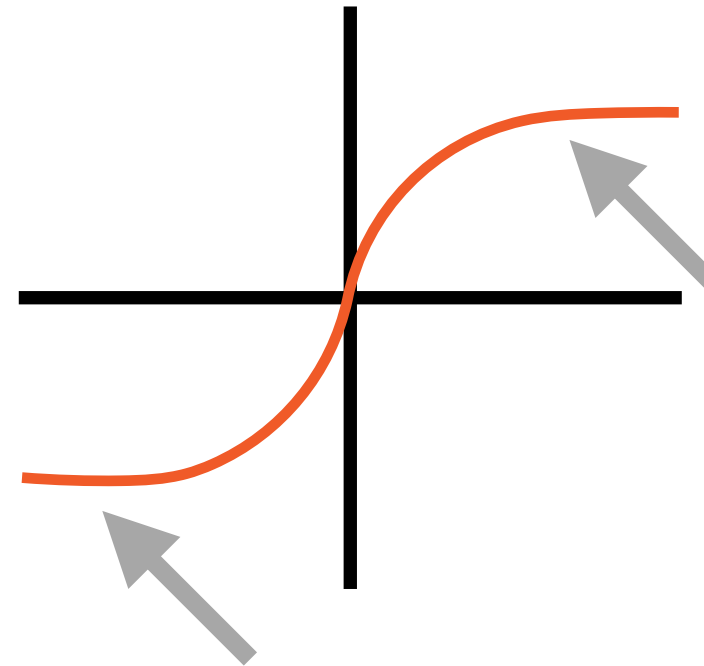
ReLU



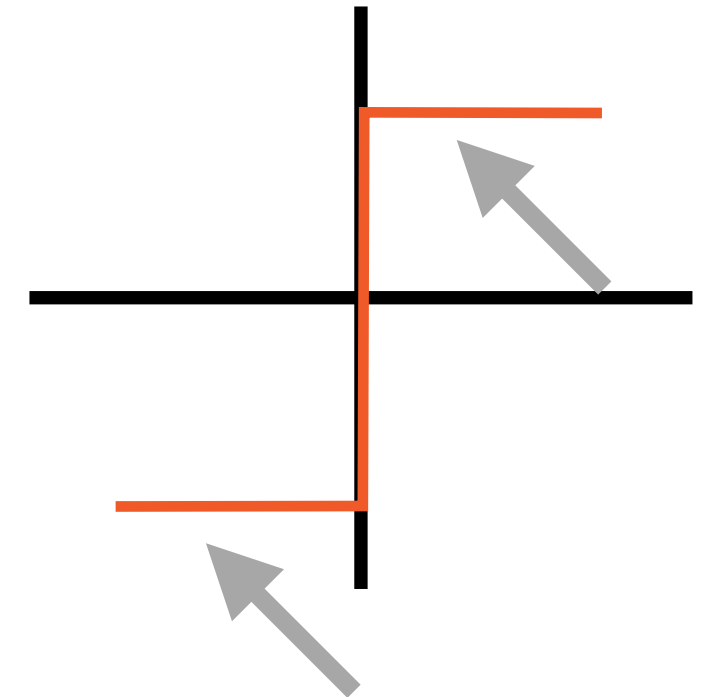
logit



tanh

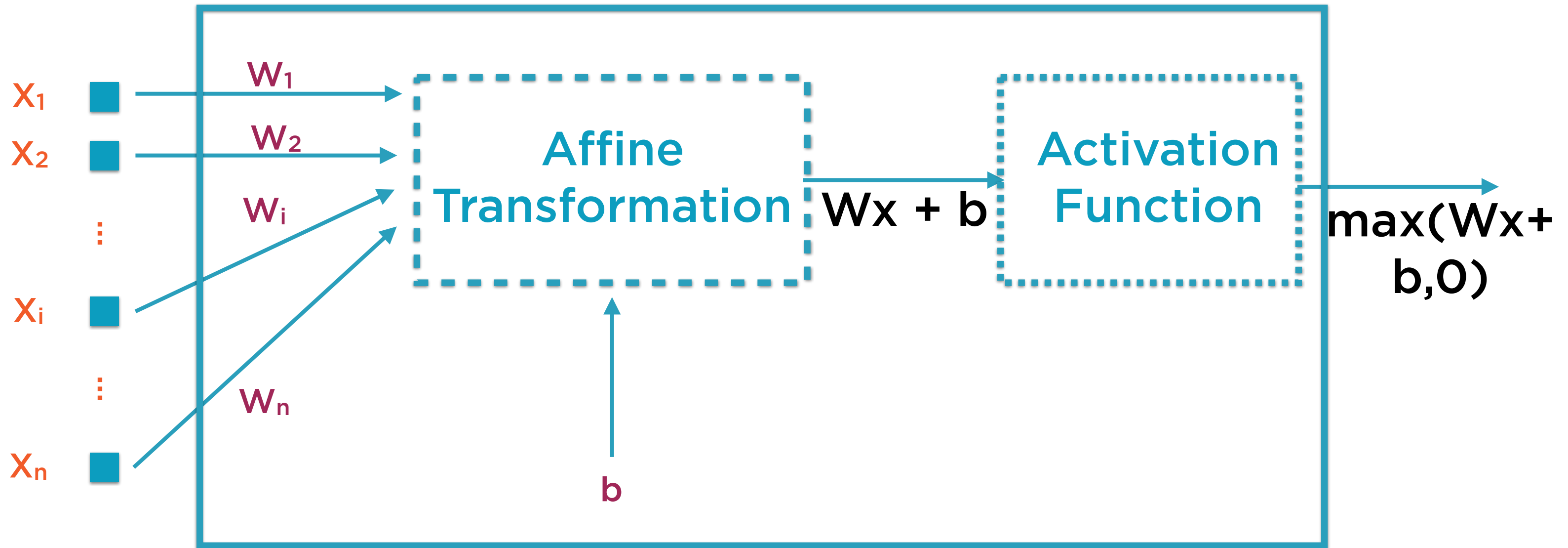


step



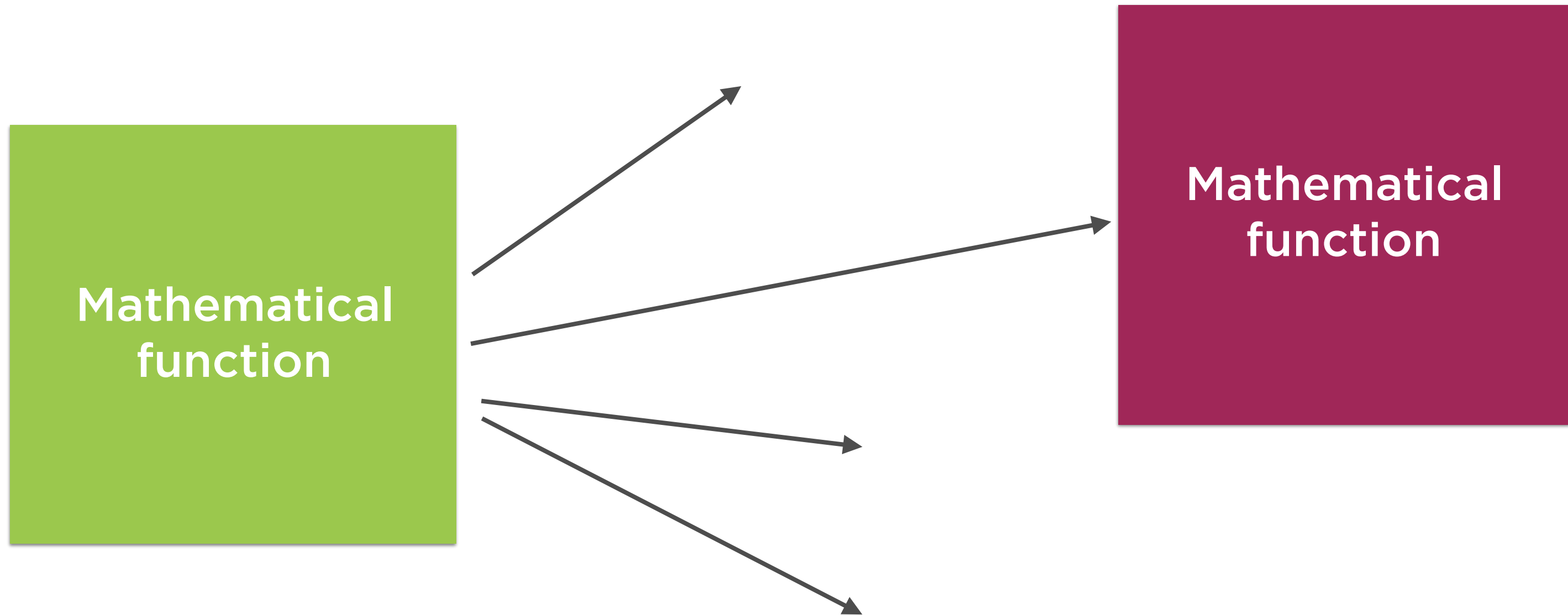
In order to train and adjust the weights of the neural network the activation functions should operate in their active region

# Neuron as a Learning Unit

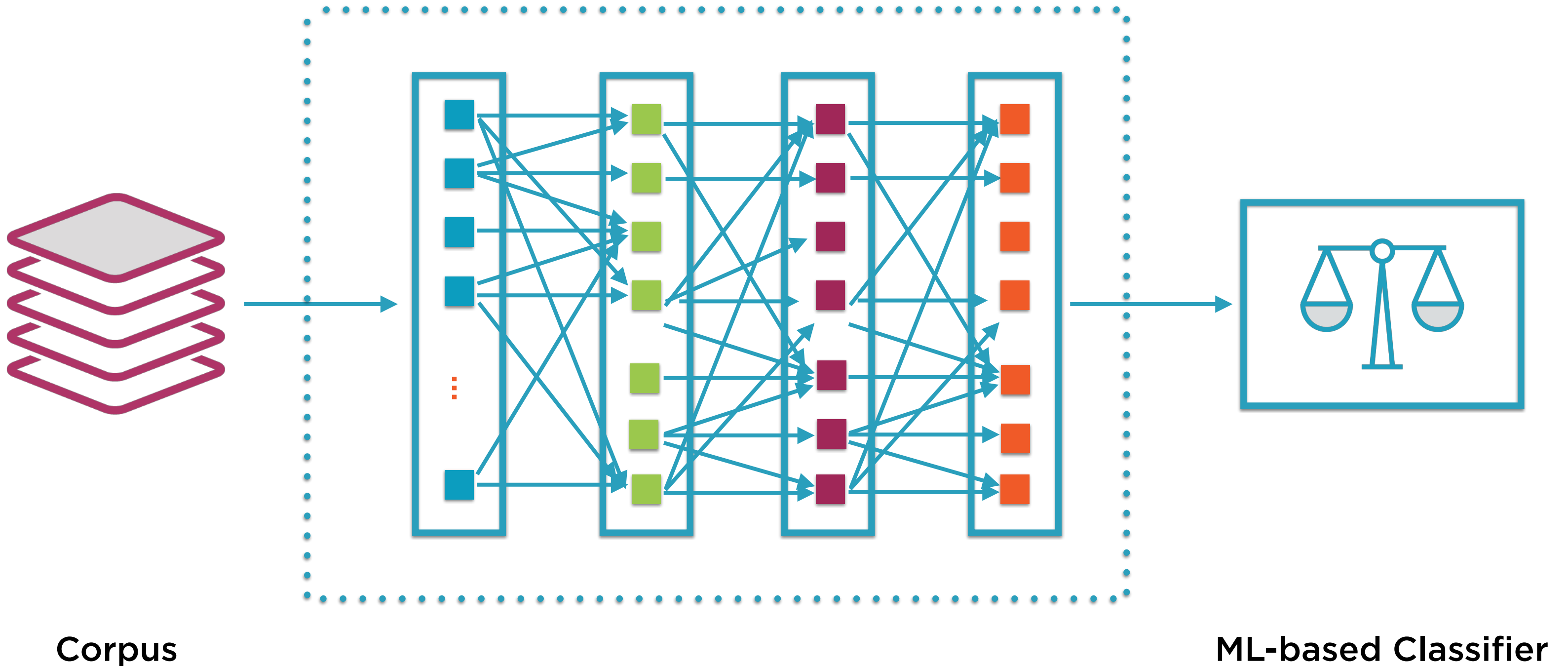


Many of these simple neurons arranged in layers can do magical stuff

# Interconnected Neurons in Layers



# Neural Networks





The **weights** and **biases** of individual neurons are determined during the **training** process

# Introducing PyTorch

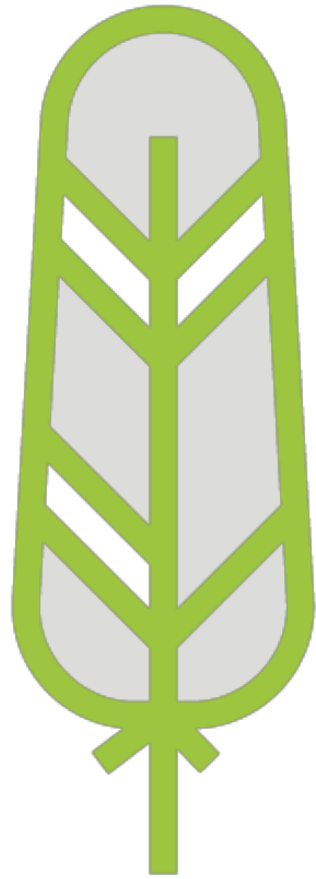
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# PyTorch

A deep learning framework for fast, flexible experimentation.

*<https://pytorch.org/>*

# PyTorch Background



## **Relatively new**

- Initial release October 2016
- Stable release April 2018

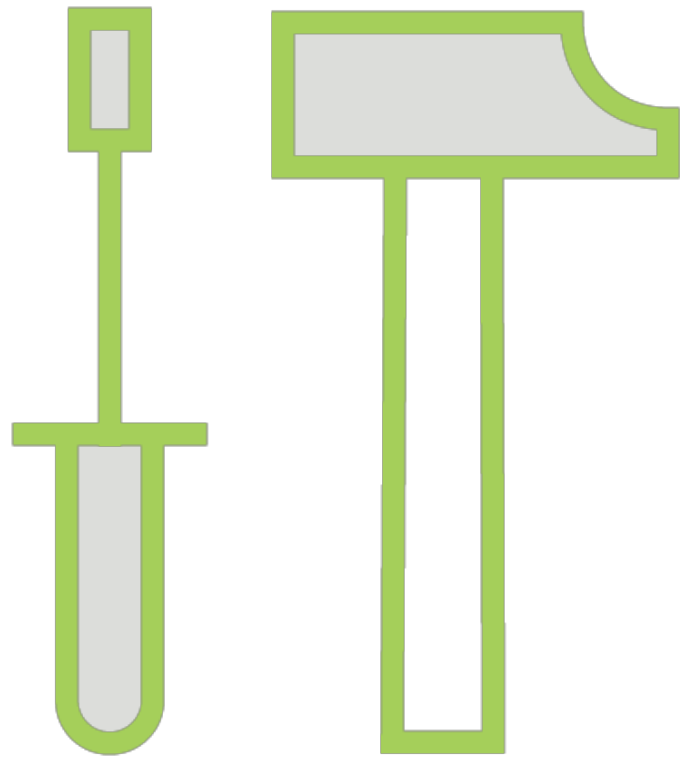
## **Based on Torch**

- Open-source ML library (since 2002!)

## **Facebook connection**

- Developed by Facebook AI researchers

# GPU-ready Tensor Library

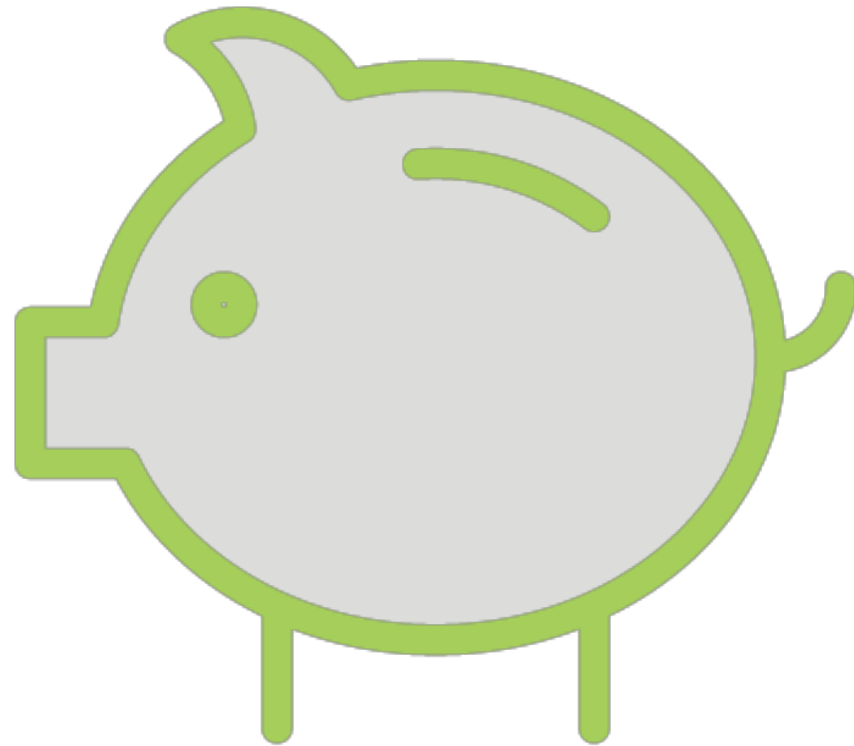


**Tensors for either CPU or GPU**

**Powerful, fast NumPy-like functionality**

- slicing
- indexing
- reductions
- linear algebra

# Tight Python Integration



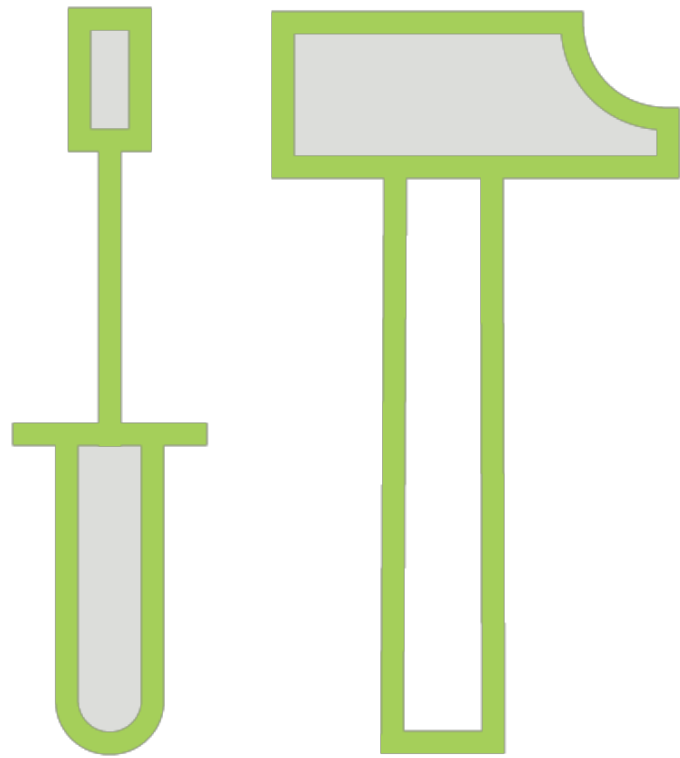
**Deeply tied to Python**

**Approach similar to NumPy/scikit-learn**

**Create neural networks in Python**

**Use existing Python libraries and debuggers**

# Deep Learning Framework

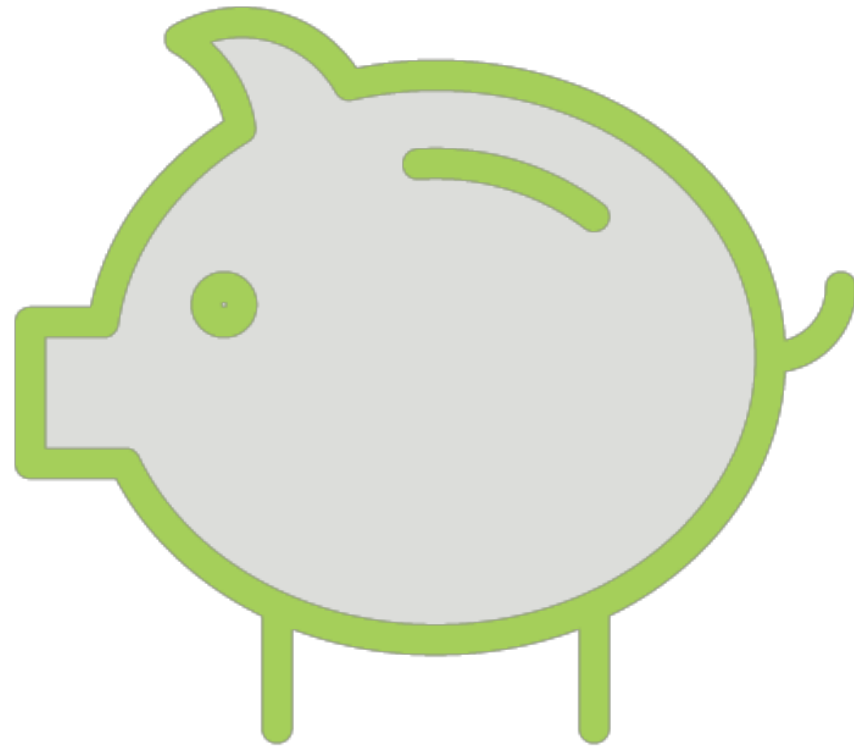


**“Tape-based Autograd”**

**Neural network can be redefined dynamically**

**Different from TensorFlow, CNTK**

# Imperative Execution



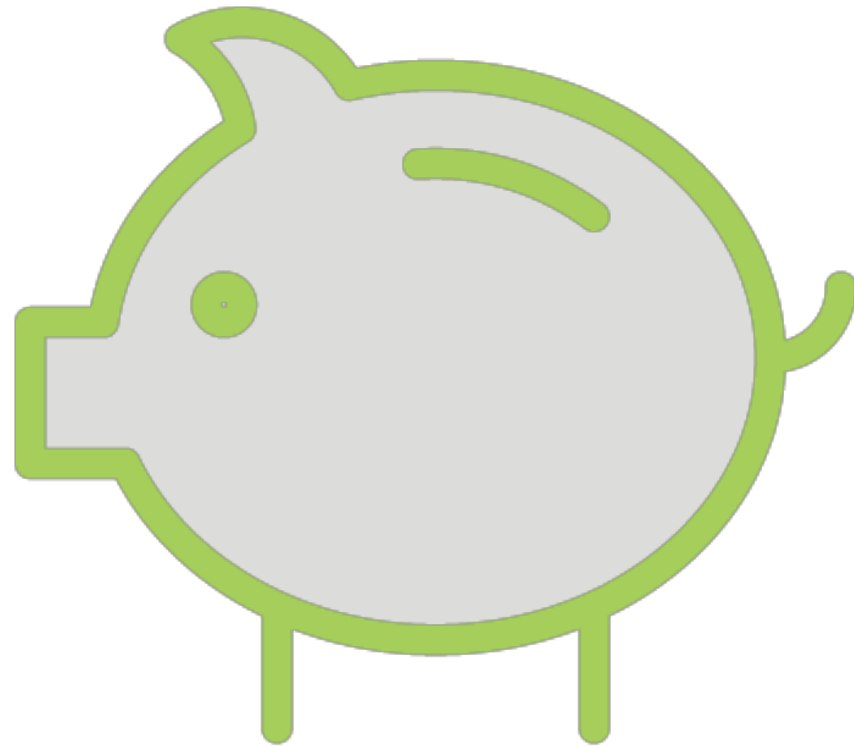
**Write code, run immediately**

**No separate build and run phases**

**Debugging easier**



# Ease of Extensibility



**Easily write new neural network layers**

**Different alternatives**

- Torch API
- Scipy
- C/C++ extension API

# TensorFlow and PyTorch

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# TensorFlow

TensorFlow is an end-to-end open source platform for machine learning. A comprehensive, flexible ecosystem of tools, libraries and community resources to easily build and deploy ML powered applications.

*<https://tensorflow.org/>*

# TensorFlow vs. PyTorch

## TensorFlow

**Originally developed at Google by the Google Brain team**

**First released in November 2015**

**Tensors as fundamental data structures for computation**

**CUDA support for GPUs**

## PyTorch

**Originally developed by AI researchers at Facebook**

**First released in October 2016**

**Tensors as fundamental data structures for computation**

**CUDA support for GPUs**

# TensorFlow vs. PyTorch

## TensorFlow

Computation graph is static

Must be defined before being run

`tf.Session` for separation from  
Python

## PyTorch

Computation graph is dynamic

Can be defined and run as you go

Tightly integrated with Python

# TensorFlow vs. PyTorch

## TensorFlow

Debugging via tfdbg

Visualization using built-in  
TensorBoard

Deployment using TF Serving

`tf.device` and `tf.DeviceSpec` to use  
GPUs (relatively hard)

## PyTorch

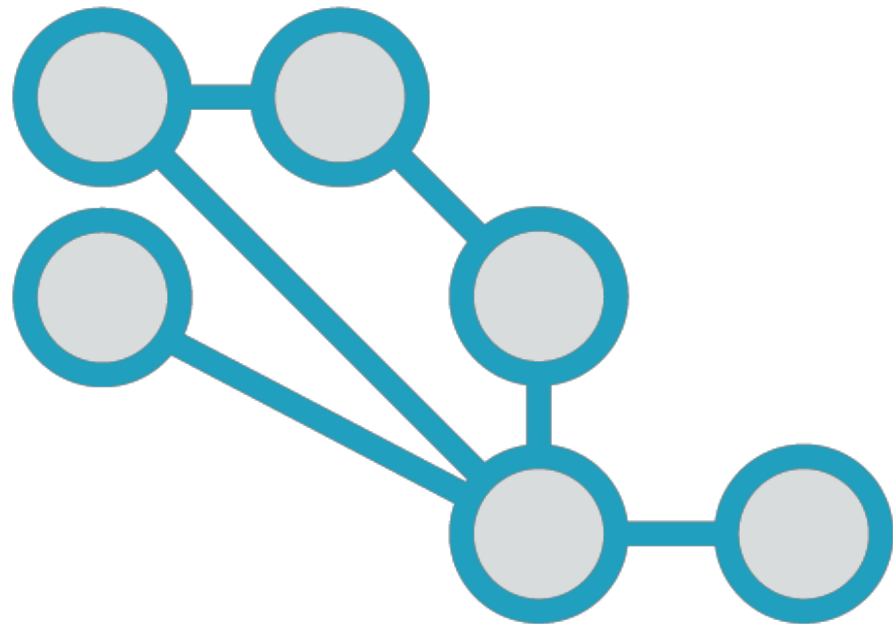
Debugging with PyCharm, pdb

Visualization using matplotlib,  
seaborn

Need to set up REST API e.g. Flask

`torch.nn.DataParallel` to use GPUs  
(relatively easy)

# Learning From PyTorch



**TensorFlow now has eager execution mode for dynamic graph execution**

**Higher level abstraction to build neural network layers using the Keras API**

Demo

**Installation and setup of PyTorch**



# Summary

**Deep learning using neural networks**

**Neurons and activation functions**

**Introducing PyTorch to build neural networks**

**Install and set up PyTorch on your local machine**