

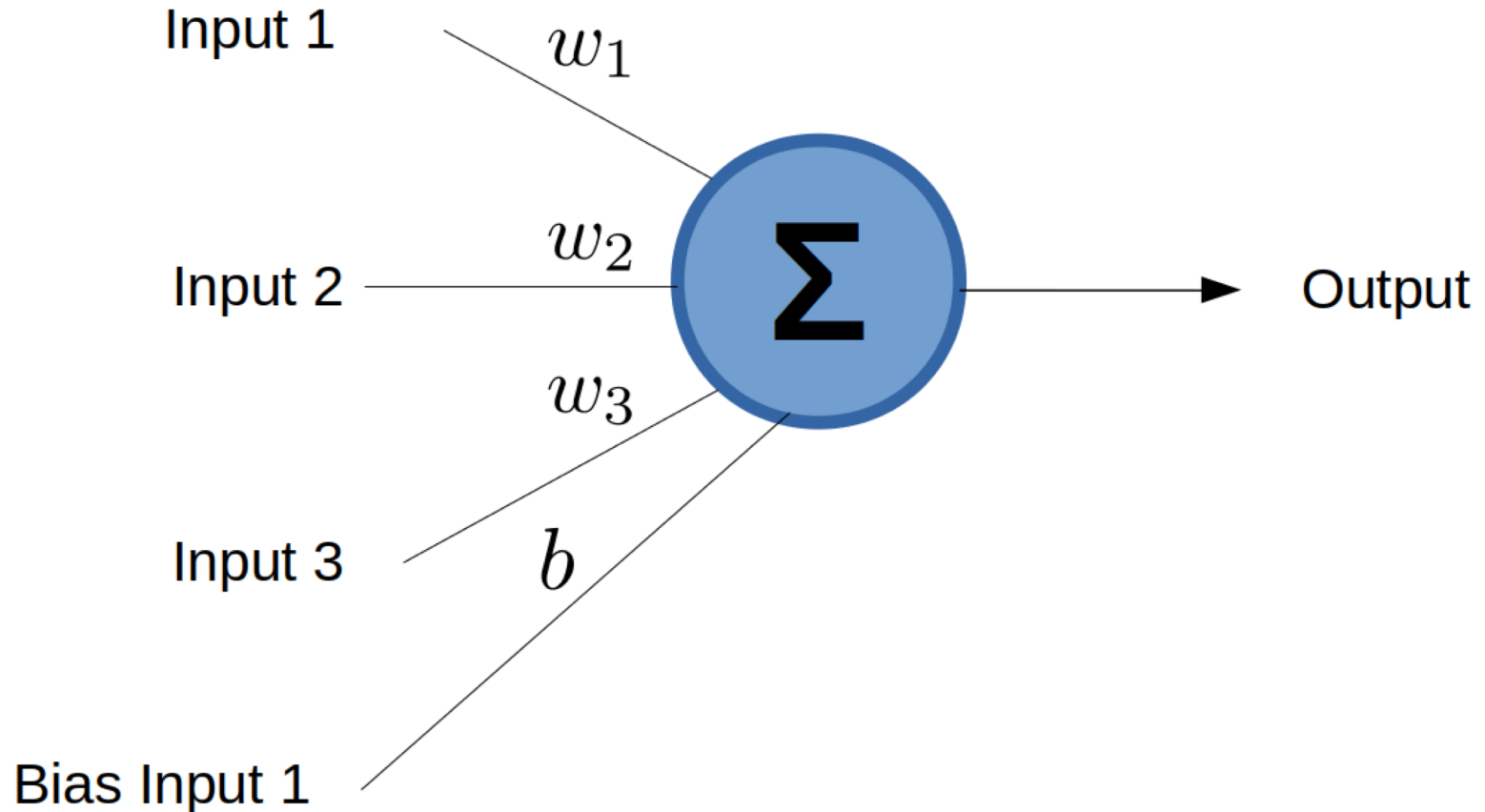


Multilayer Perceptron

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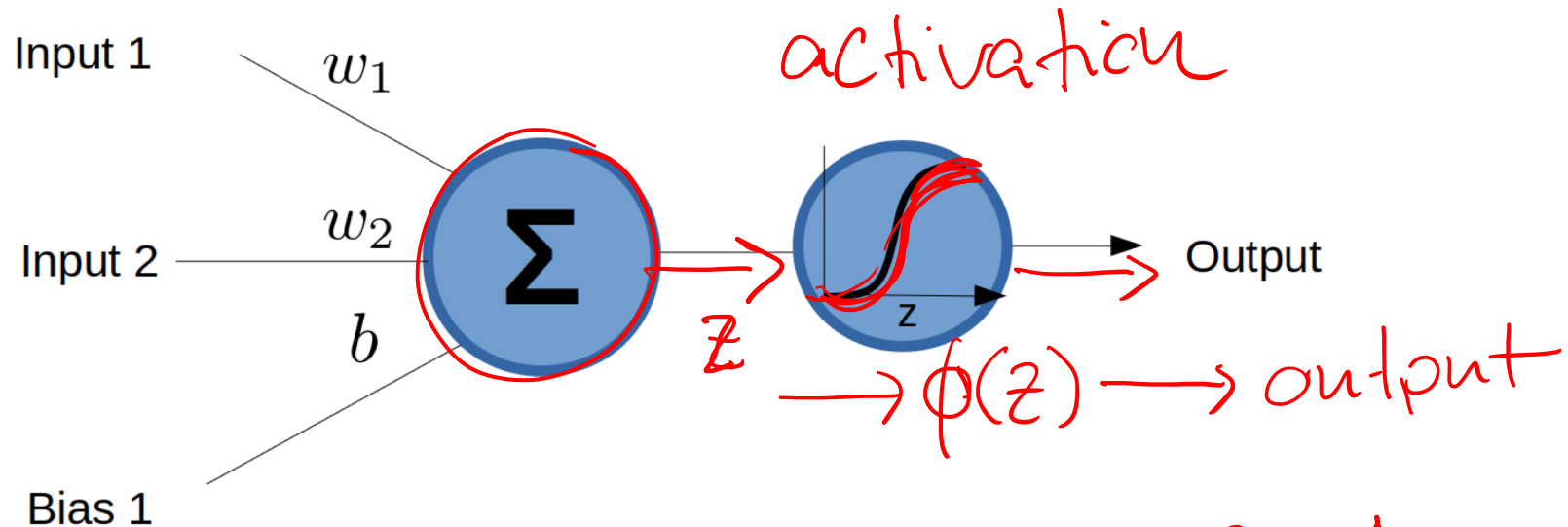
Linear Perceptron with Bias

- The input to the bias term is always 1



Nonlinear Perceptron

Add nonlinear activation function



Output is calculated via: $a = \phi(\mathbf{w}^T \mathbf{x} + b)$

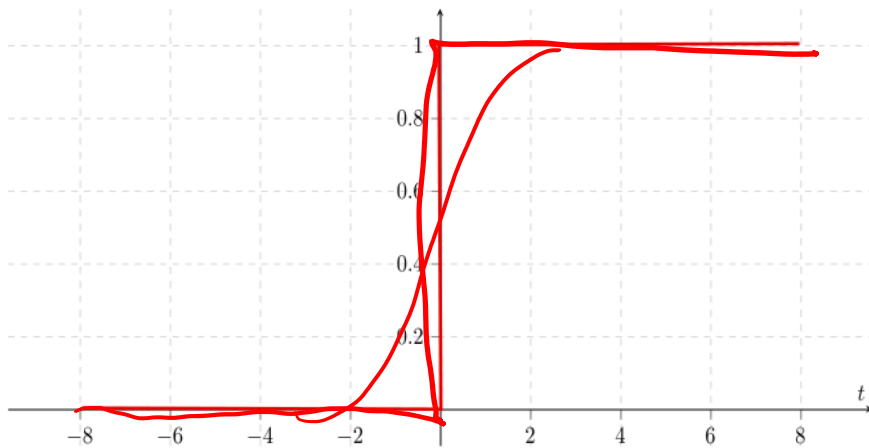
Possible activation function:

Sigmoid

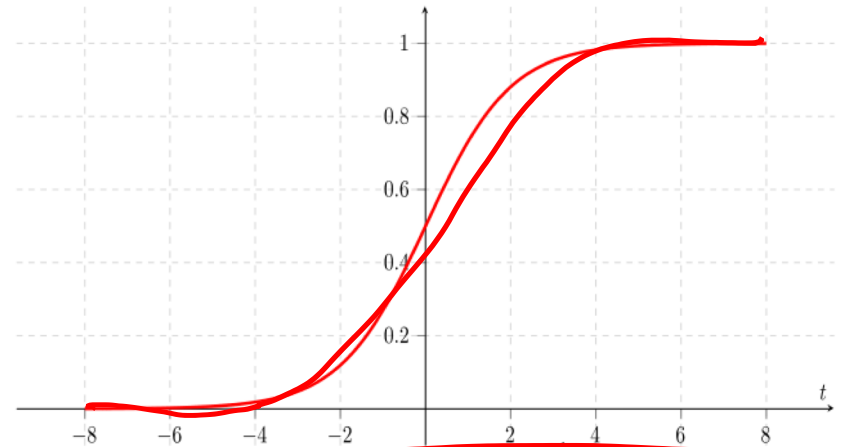
$$\phi(z) = \frac{1}{1 + \exp(-z)}$$

Sigmoid Activation Function

A **soft** version of a threshold unit



$$f(z) \begin{cases} 1 & \text{if } z > 0 \\ 0 & \text{if } z \leq 0 \end{cases}$$



$$\phi(z) = \frac{1}{1 + \exp(-z)}$$

Nice property

$$\frac{\partial \phi(z)}{\partial z} = \phi(z)(1 - \phi(z))$$

Multi-Layer Perceptron

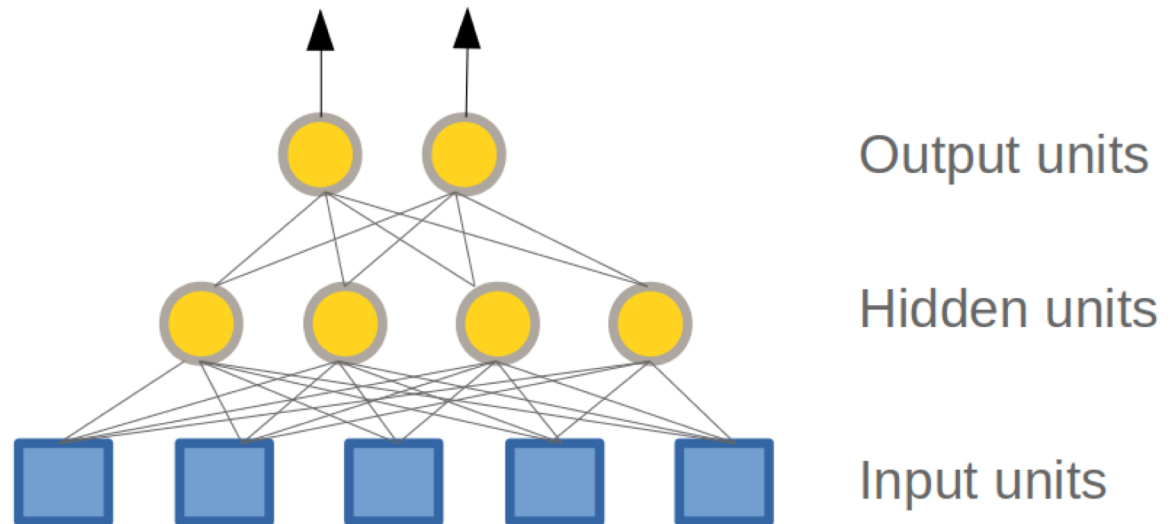
- Artificial Neural Network (ANN)

- Hierarchy of neurons

- Input layer, hidden layers, output layer

- 2 Layers = all continuous functions

- 3+ Layers = all functions



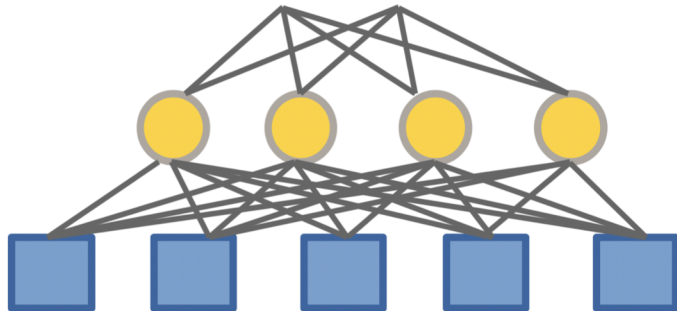
Matrix Representation for Layer

- Each layer can have multiple nodes
- Our output equation only calculated one node

$$a = \phi(\mathbf{w}^T \mathbf{x} + b)$$

- Change equation to matrix notation

$$A = \phi(WX + \mathbf{b})$$



A = activations

W = weights

X = inputs

b = vector of biases

Summary

- | **Nonlinear activation functions**
- | **ANN: multiple layers of neurons**
- | **Popular activation function: sigmoid**
- | **Sigmoid has a simple derivative**
- | **We can represent each layer in matrix notation**