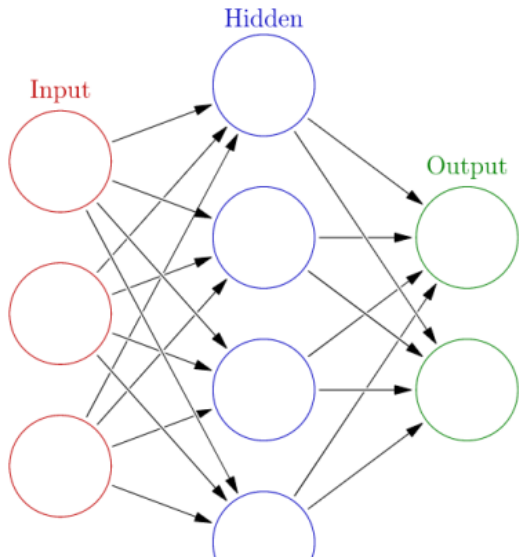


# NEURAL NETWORKS

Jan-Philipp Kolb

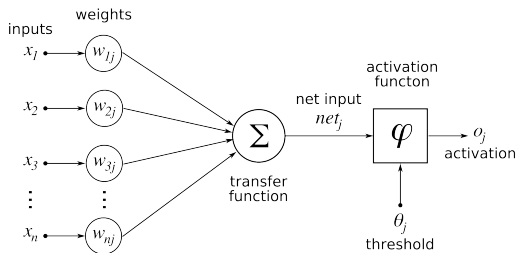
04 Juni, 2019

# EXAMPLES OF MULTI-NEURON



# ARTIFICIAL NEURON

- ▶ Inputs correspond to raw data values
- ▶ The transfer function sums all the inputs together (cumulative inputs).
- ▶ If the summed input values reach a specified threshold, the activation function generates an output signal (all or nothing).
- ▶ The output signal then moves to a raw output or other neurons.
- ▶ This basic artificial neuron is combined with multiple other artificial neurons to create an ANN.



# ACTIVATION FUNCTIONS

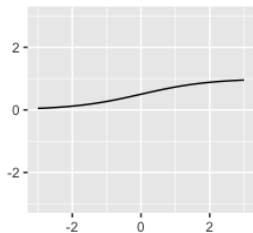
- ▶ The capability of ANNs to learn any function, (given sufficient training data examples) are dependent on the appropriate selection of the **activation function(s)** present in the network.
- ▶ They enable the ANN to learn non-linear properties present in the data.
- ▶ The input into the activation function is the weighted sum of the input features from the preceding layer.
- ▶ Let  $o_j$  be the output from the  $j$ th neuron in a given layer for a network for  $k$  input vector features.

$$o_j = \Phi\left(b_j + \sum_{i=1}^p w_i x_i\right)$$

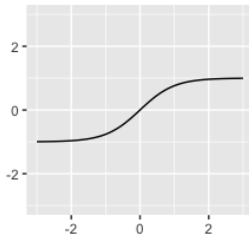
# COMMON ANN ACTIVATION FUNCTIONS

Common ANN Activation Functions

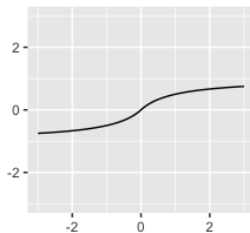
logistic



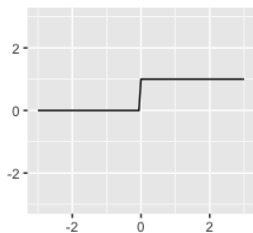
tanh



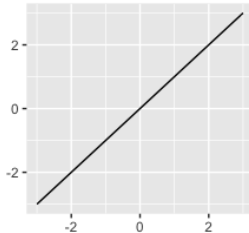
softsign



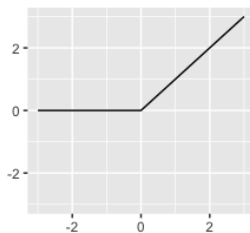
binary step



linear



rectified linear unit (ReLU)



## THE OUTPUT ( $o_j$ )...

- ▶ ... can feed into the output layer of a neural network, or in deeper architectures may feed into additional hidden layers.
- ▶ The activation function determines if the sum of the weighted inputs plus a bias term is sufficiently large to trigger the firing of the neuron.
- ▶ No universal best choice for the activation function, researchers have provided information regarding what activation functions work well for ANN solutions to many common problems.
- ▶ The choice of the activation function governs the required data scaling necessary for ANN analysis.

# HOW ANNs LEARN

- ▶ We have some features ( $X$ ) describing an output ( $y$ )
- ▶ To begin training our notional single-layer one-neuron neural network we initially randomly assign weights.
- ▶ We then run the neural network with the random weights and record the outputs generated.
- ▶ This is called a forward pass. Output values, in our case called  $y$ , are a function of the input values ( $X$ ), the random initial weights ( $w$ ) and our choice of the threshold function ( $T$ ).

$$\hat{y} = f(X, w, T)$$

# CHOICE OF THE PERFORMANCE FUNCTION

- ▶ Once we have our ANN output values ( $\hat{y}$ ) we can compare them to the data set output values ( $y$ ).
- ▶ To do this we use a performance function  $P$ .
- ▶ The choice of the performance function is a choice of the analyst, we choose Sum of Squared Errors (SSE).