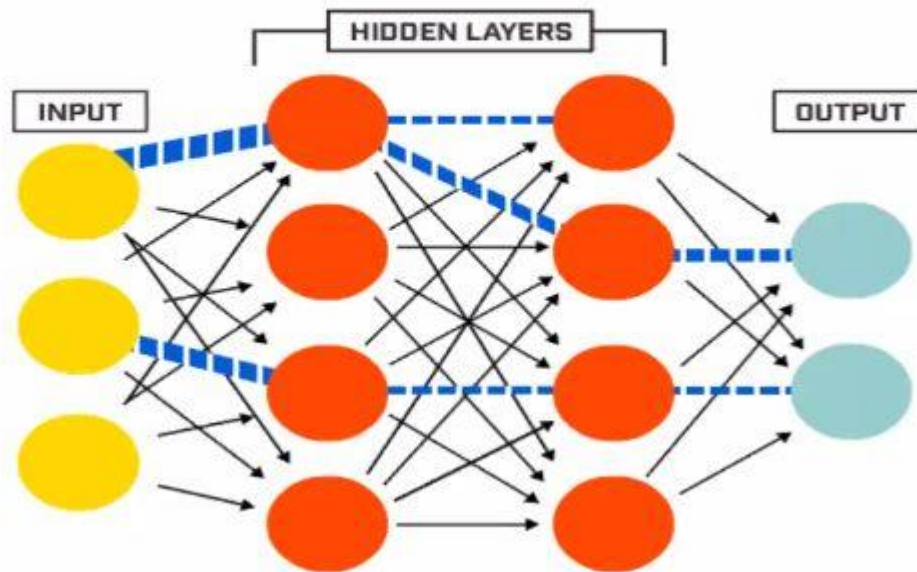
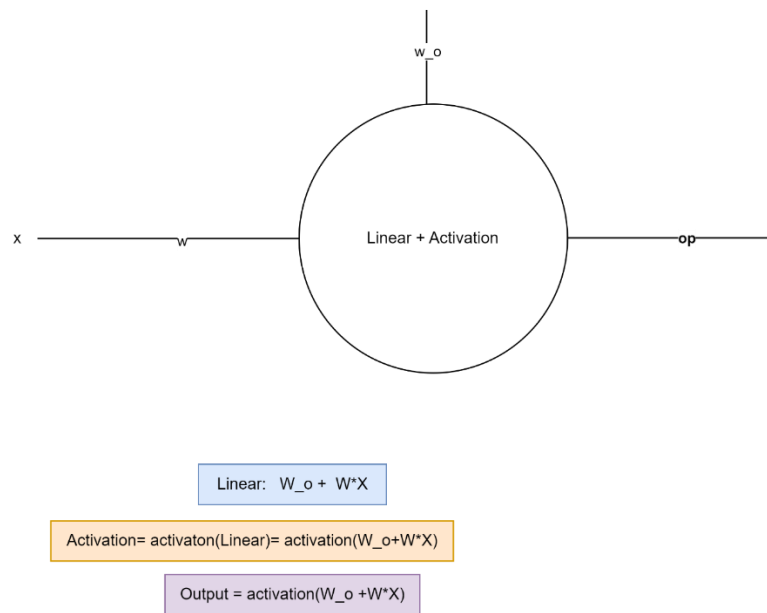


1. Multilayer Perceptron/ANN



1. **The Number of Hidden Layer**
2. **The number of neurons,**
3. **Weight Initialization**
4. **Activation function**
5. **Loss Function**
6. **Metrics**
7. **Optimizer,**
8. **Learning rate,**
9. **Batch size**
10. **Epochs**
11. **Iteration**
12. **Dropout rate etc**

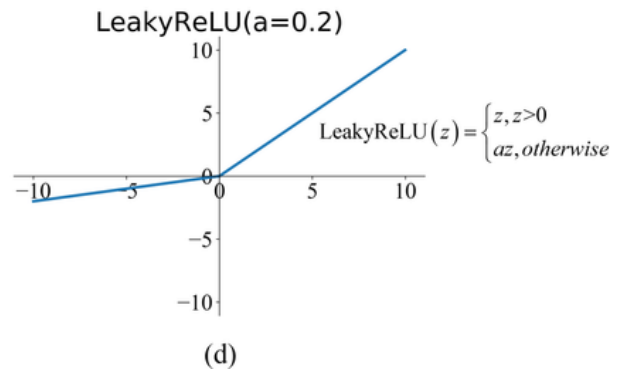
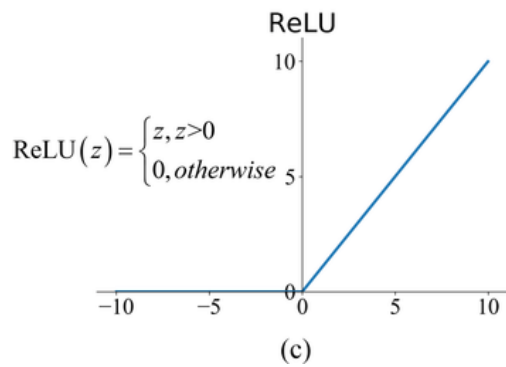
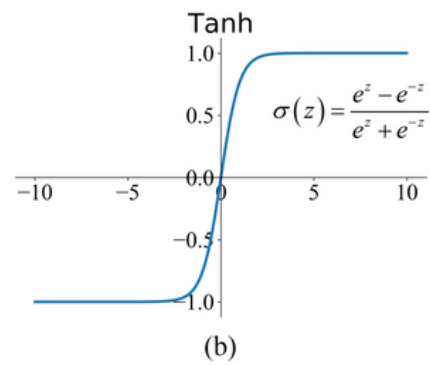
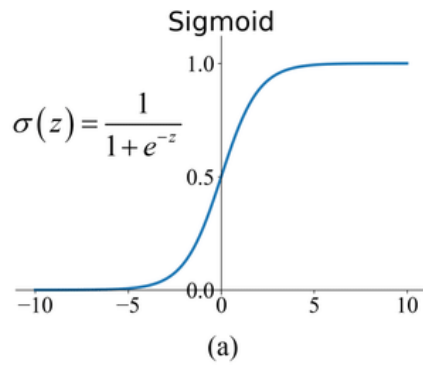
what is Neuron - How neuron works



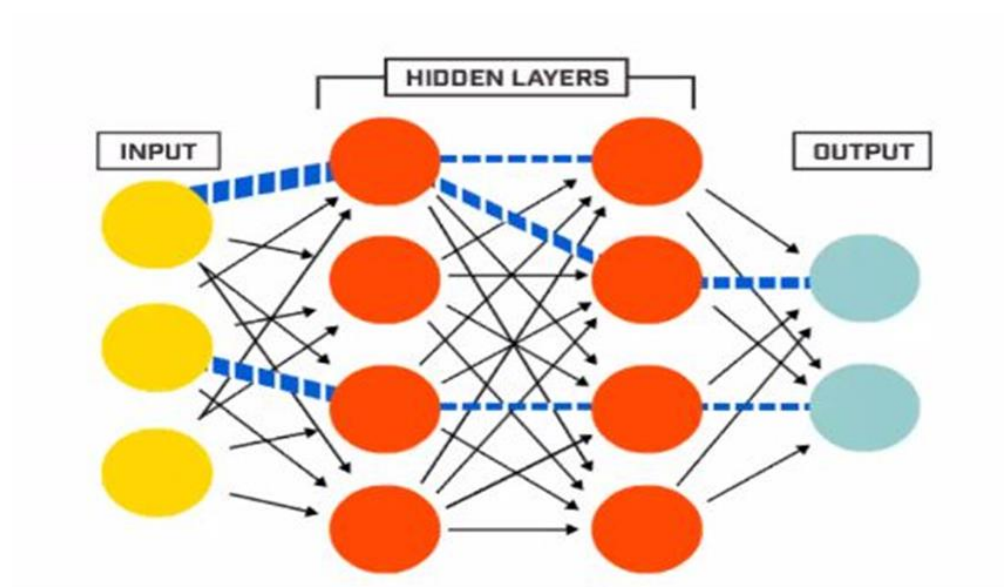
1. In a single neuron two operations are performed
2. Neuron is Linear combinations and Activations
3. If there no activations it is simply a Linear network
4. It will not understand and patterns, the deeper thinking is not possible
5. In order to get the more patterns, we need to **apply Non linearity**
6. This non linearity is given by activation functions
7. The perfect example is sigmoid activation function
8. It is look like s curve, which means it is Nonlinear curve
9. Now it is the time, understand human brain neuron

There are many activation functions available

- Sigmoid
- SoftMax
- Tanh
- ReLU Rectified Linear unit
- Leaky ReLU



Let's calculate the number of parameters for the given architecture:



1. **Input layer:** There are no parameters, only inputs.
2. **First hidden layer:**
 - a. Total parameters in this layer: $(3 \times 4) + 4 = 16$ parameters.
3. **Second hidden layer:**
 - a. Total parameters in this layer: $(4 \times 4) + 4 = 20$ parameters.
4. **Output layer:**
 - a. Total parameters in this layer: $(4 \times 2) + 2 = 10$ parameters.

Total parameters = 36

Let's calculate the number of parameters for the given architecture:

- Input layer: 10 neurons
- First hidden layer: 10 neurons
- Second hidden layer: 5 neurons
- Output layer: 2 neurons

For each layer, the number of parameters can be calculated as follows:

1. **Input layer:** There are no parameters, only inputs.
2. **First hidden layer:**
 - Each neuron has 10 weights (one for each input) and 1 bias term.
 - Total parameters in this layer: $(10 \times 10) + 10 = 110$ parameters.
3. **Second hidden layer:**
 - Each neuron has 10 weights (one for each neuron in the previous layer) and 1 bias term.
 - Total parameters in this layer: $(10 \times 5) + 5 = 55$ parameters.
4. **Output layer:**
 - Each neuron has 5 weights (one for each neuron in the previous layer) and 1 bias term.
 - Total parameters in this layer: $(5 \times 2) + 2 = 12$ parameters.

Now, summing up the parameters from all layers:

110 parameters in the first hidden layer

55 parameters in the second hidden layer

12 parameters in the output layer

=177 parameters in