

Assignment -8

Title: program for creating a back propagation feed-forward neural network.

Aim: Write a python program for creating a back propagation feed-forward neural network

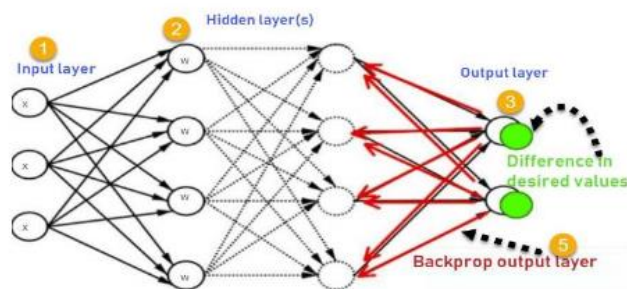
Theory:

Back propagation Neural Network:

Back propagation is a widely used algorithm for training feed forward neural networks. It computes the gradient of the loss function with respect to the network weights. It is very efficient, rather than naively directly computing the gradient concerning each weight. This efficiency makes it possible to use gradient methods to train multi-layer networks and update weights to minimize loss; variants such as gradient descent or stochastic gradient descent are often used.

The back propagation algorithm works by computing the gradient of the loss function with respect to each weight via the chain rule, computing the gradient layer by layer, and iterating backward from the last layer to avoid redundant computation of intermediate terms in the chain rule.

Backpropagation Algorithm:



Step 1: Inputs X, arrive through the preconnected path.

Step 2: The input is modeled using true weights W. Weights are usually chosen randomly.

Step 3: Calculate the output of each neuron from the input layer to the hidden layer to the output layer.

Step 4: Calculate the error in the outputs

Back propagation Error= Actual Output - Desired Output

Step 5: From the output layer, go back to the hidden layer to adjust the weights to reduce the error.

Step 6: Repeat the process until the desired output is achieved

Parameters :

x = inputs training vector $x=(x_1, x_2, \dots, x_n)$.

t = target vector $t=(t_1, t_2, \dots, t_n)$.

δ_k = error at output unit.

δ_j = error at hidden layer.

α = learning rate.

V_{0j} = bias of hidden unit j .

Training Algorithm :

Step 1: Initialize weight to small random values.

Step 2: While the stepsstopping condition is to be false do step 3 to 10.

Step 3: For each training pair do step 4 to 9 (Feed-Forward).

Step 4: Each input unit receives the signal unit and transmits the signal x_i signal to all the units.

Step 5 : Each hidden unit Z_j ($j=1$ to a) sums its weighted input signal to calculate its net input

$$z_{in j} = v_{0j} + \sum_i x_i v_{ij} \quad (i=1 \text{ to } n)$$

Applying activation function $z_j = f(z_{in j})$ and sends this signal to all units in the layer about i.e output units

For each output l =unit $y_k = (k=1$ to $m)$ sums its weighted input signals.

$$y_{in k} = w_{0k} + \sum_j z_j w_{jk} \quad (j=1 \text{ to } a)$$

and applies its activation function to calculate the output signals.

$$y_k = f(y_{in k})$$

Backpropagation Error :

Step 6: Each output unit y_k ($k=1$ to n) receives a target pattern corresponding to an input pattern then error is calculated as:

$$\delta_k = (t_k - y_k) + y_{in k}$$

Step 7: Each hidden unit Z_j ($j=1$ to a) sums its input from all units in the layer above

$$\delta_{in j} = \sum_k \delta_k w_{jk}$$

The error information term is calculated as :

$$\delta_j = \delta_{in j} + z_{in j}$$

Updation of weight and bias :

Step 8: Each output unit y_k ($k=1$ to m) updates its bias and weight ($j=1$ to a). The weight correction term is given by :

$$\Delta w_{jk} = \alpha \delta_k z_j$$

and the bias correction term is given by $\Delta w_k = \alpha \delta_k$.

therefore $w_{jk}(\text{new}) = w_{jk}(\text{old}) + \Delta w_{jk}$

$$w_{0k}(\text{new}) = w_{0k}(\text{old}) + \Delta w_k$$

for each hidden unit z_j ($j=1$ to a) update its bias and weights ($i=0$ to n) the weight connection term

$$\Delta v_{ij} = \alpha \delta_j x_i$$

and the bias connection on term

$$\Delta v_{0j} = \alpha \delta_j$$

Therefore $v_{ij}(\text{new}) = v_{ij}(\text{old}) + \Delta v_{ij}$

$$v_{0j}(\text{new}) = v_{0j}(\text{old}) + \Delta v_{0j}$$

Step 9: Test the stopping condition. The stopping condition can be the minimization of error, number of epochs.

Need for Backpropagation:

Backpropagation is “backpropagation of errors” and is very useful for training neural networks. It’s fast, easy to implement, and simple. Backpropagation does not require any parameters to be set, except the number of inputs. Backpropagation is a flexible method because no prior knowledge of the network is required.

Conclusion:

We have successfully implemented back propagation feed-forward neural network.

Questions:

1. Explain Translation System for Face-To-Face Dialog & Intelligent Help Systems Using ANN.
2. Explain Building Blocks of CNNs.
3. Explain Architecture of CNN.
4. Explain Convolution/ Pooling layers of CNN.