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BMSAim:-

To implement the Multiple Layer Perceptron to solve

i) XOR problem

ii) iris-dataset

and implement single layer perceptron for iris-dataset.

The Multi-Layer Perceptron Algorithm

* Initialization

- * all weights to small (+ve & -ve) random values.

* Training

- Repeat:

- * for each input vector:-

Forward Phase

- * compute activation of each neuron j in the hidden layer(s) using

$$h_c = \sum_{i=0}^L x_i v_{ic}$$

$$a_c = g(h_c) = \frac{1}{1 + e^{-\beta h_c}}$$

- * work through the network until you get the output layer neurons, which have activations

$$h_k = \sum_j a_j w_{jk} \quad y_k = g(h_k) = \frac{1}{1 + e^{-\beta h_k}}$$

• Backwards Phase

- * compute the error of the output using:-

$$\delta_o(k) = (y_k - t_k) \cdot y_k (1 - y_k)$$

- * compute the error in hidden layers using:-

$$\delta_h(c) = a_c (1 - a_c) \sum_{k=1}^N w_{ck} \delta_o(k)$$

- * update the output layer weights using:-

$$w_{ck} \leftarrow w_{ck} - \eta \delta_o(k) a_c^{\text{hidden}}$$

- * update hidden layers using:-

$$V_i \leftarrow V_i - \eta \delta_h(k) x_i$$

- * (if using sequential updating) randomize the order of input vectors so that you don't train in the same order

— until learning stops.

* Recall

- use the forward phase in the training section above.