





NPTEL ONLINE CERTIFICATION COURSES

Course Name: Deep Learning

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Topic

Lecture 22: Multilayer Perceptron -II

CONCEPTS COVERED

Concepts Covered:

☐ Neural Network

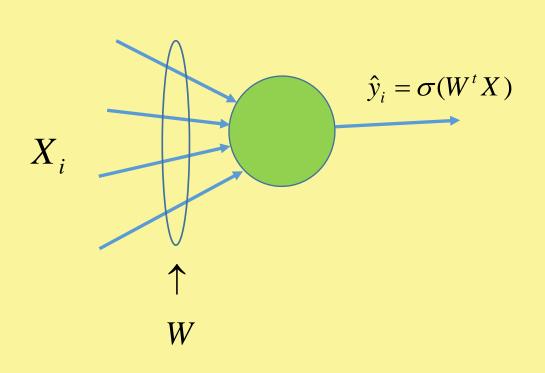
☐ Feed Forward NN

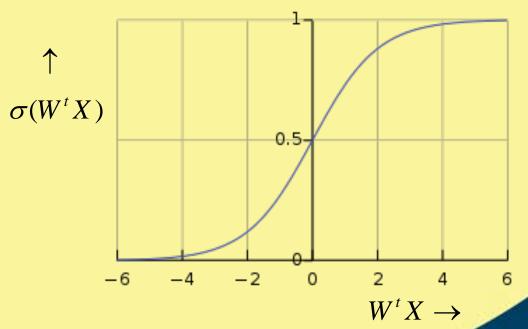
■ Back Propagation Learning





Single Layer Network- Single Output with nonlinearity







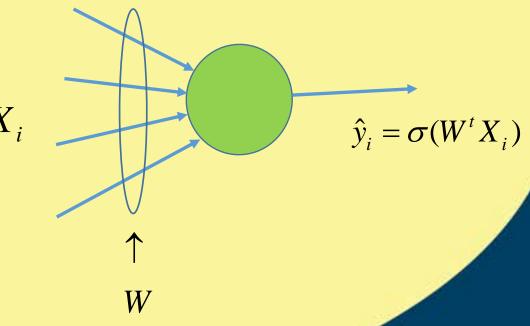
Single Layer Network- Single Output with nonlinearity

$$E = \frac{1}{2}(\hat{y}_i - y_i)^2 = \frac{1}{2}(\sigma(W^t X_i) - y_i)^2$$

$$\nabla_{\mathbf{W}} E = \hat{\mathbf{y}}_i (1 - \hat{\mathbf{y}}_i) (\hat{\mathbf{y}}_i - \mathbf{y}_i) X_i$$

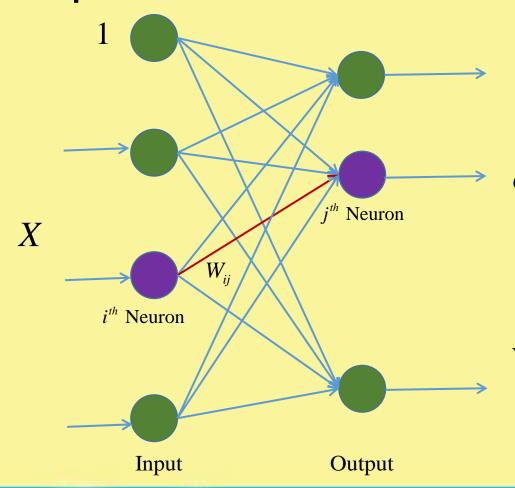
Weight updation rule ⇒

$$W \leftarrow W - \eta \hat{y}_i (1 - \hat{y}_i) (\hat{y}_i - y_i) X_i$$





Back Propagation Learning: Single Layer Multiple Output



$$o_j = \frac{1}{1 + e^{-\theta_j}} \qquad \theta_j = \sum_{i=1}^D W_{ij} x_i$$

$$E = \frac{1}{2} \sum_{j=1}^{M} (o_j - t_j)^2$$

$$\frac{\partial E}{\partial W_{ij}} = \frac{\partial E}{\partial o_j} \cdot \frac{\partial o_j}{\partial \theta_j} \cdot \frac{\partial \theta_j}{\partial W_{ij}}$$
$$= (o_j - t_j) o_j (1 - o_j) x_i$$

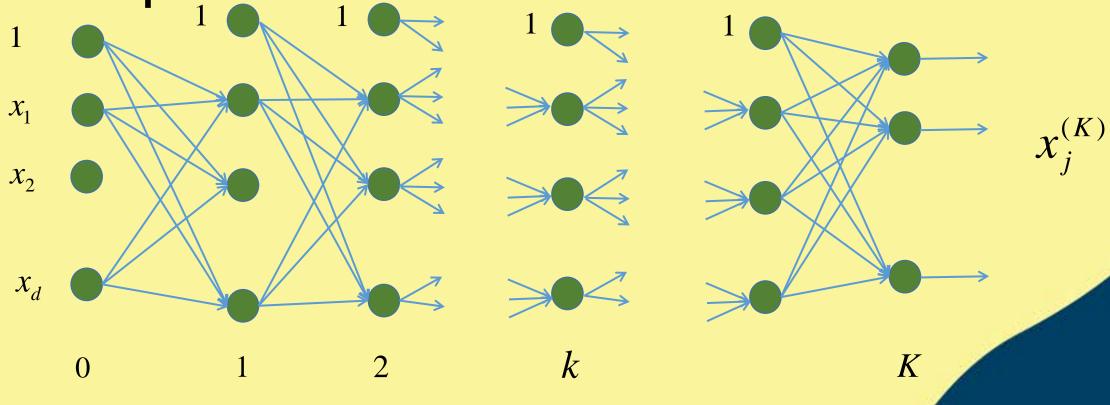
Weight updation rule ⇒

$$W_{ij} \leftarrow W_{ij} - \eta(o_j - t_j)o_j(1 - o_j)x_i$$



Multilayer

Perceptron



 $M_k \to \text{No. of nodes in } k^{th} \text{ layer}$









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Thank you