Neural Networks

Layer 1 Input Layer	Layer 2 Hidden Layer	Layer 3 Hidden Layer	Layer L Output Layer
$x_1 = a_1^{(1)}$	$\begin{pmatrix} a_1^{(2)} = g(z_1^{(2)}) \\ z_1^{(2)} = (\Theta^{(1)}a^{(1)})_1 \end{pmatrix}$	$\begin{bmatrix} a_1^{(3)} = g(z_1^{(3)}) \\ z_1^{(3)} = (\Theta^{(2)}a^{(2)})_1 \end{bmatrix} \dots$	$\begin{bmatrix} a_1^{(L)} = g(z_1^{(L)}) = \hat{y}_1 \\ z_1^{(L)} = (\Theta^{(L-1)}a^{(L-1)})_1 \end{bmatrix}$
$ a_0^{(1)} = 1 $ $ a_0^{(1)} = 1 $ $ \Theta^{(1)} = \begin{bmatrix} \Theta_{10}^{(1)} & \Theta_{11}^{(1)} & \cdots \\ \Theta_{20}^{(1)} & \Theta_{21}^{(1)} & \cdots \\ \vdots & \vdots & \ddots \\ \Theta_{s_20}^{(1)} & \Theta_{s_{21}}^{(1)} & \cdots \end{bmatrix} $	$ \begin{array}{c} \theta_{1s_{1}}^{(1)} \\ \theta_{1s_{1}}^{(1)} \\ \theta_{2s_{1}}^{(1)} \\ \vdots \\ \theta_{s_{2}s_{1}}^{(1)} \end{array} $ $ \begin{vmatrix} a_{2}^{(2)} = g(z_{2}^{(2)}) \\ z_{2}^{(2)} = (\Theta^{(1)}a^{(1)})_{2} \end{vmatrix} $ $ \vdots $ $ \theta^{(2)} = \begin{bmatrix} \theta_{10}^{(2)} & \theta_{10}^{(2)} \\ \theta_{20}^{(2)} & \theta_{20}^{(2)} \\ \vdots & \vdots \\ \theta_{s_{3}0}^{(2)} & \theta_{s_{2}}^{(2)} \end{vmatrix} $	$\begin{bmatrix} z_1^2 & \cdots & \Theta_{1s_2}^{(2)} \\ z_1^2 & \cdots & \Theta_{2s_2}^{(2)} \\ \vdots & \ddots & \vdots \end{bmatrix} \qquad \begin{bmatrix} a_2^{(3)} = g(z_2^{(3)}) \\ z_2^{(3)} = (\Theta^{(2)}a^{(2)})_2 \end{bmatrix} \qquad \cdots$	$\begin{bmatrix} a_2^{(L)} = g(z_2^{(L)}) = \hat{y}_2 \\ z_2^{(L)} = (\Theta^{(L-1)}a^{(L-1)})_2 \end{bmatrix}$ \vdots
$x_n = a_{s_1}^{(1)}$	$\begin{pmatrix} a_{s_2}^{(2)} = g(z_{s_2}^{(2)}) \\ z_{s_2}^{(2)} = (\Theta^{(1)}a^{(1)})_{s_2} \end{pmatrix}$	$ \begin{pmatrix} a_{s_3}^{(3)} = g(z_{s_3}^{(3)}) \\ z_{s_3}^{(3)} = (\Theta^{(2)}a^{(2)})_{s_3} \end{pmatrix} \dots $	$\begin{bmatrix} a_{s_L}^{(L)} = g(z_{s_L}^{(L)}) = \hat{y}_k \\ z_{s_L}^{(L)} = (\Theta^{(L-1)}a^{(L-1)})_{s_L} \end{bmatrix}$

Backpropagation Algorithm

$$\begin{bmatrix} a_1^{(1)} & a_1^{(2)} & a_1^{(3)} & \cdots & a_1^{(L-1)} & a_1^{(L)} \\ a_2^{(1)} & a_2^{(2)} & a_2^{(2)} & a_2^{(3)} & \cdots & a_2^{(L-1)} \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ a_{s_1}^{(1)} & a_{s_2}^{(2)} & a_{s_3}^{(2)} & a_{s_3}^{(3)} & \cdots & a_{s_{L-1}}^{(L-1)} & a_{s_{L-1}}^{(L-1)} \\ a_{s_1}^{(2)} & a_{s_2}^{(2)} & a_{s_3}^{(3)} & \cdots & a_{s_{L-1}}^{(L-1)} & a_{s_{L-1}}^{(L-1)} \\ a_{s_1}^{(1)} & a_{s_2}^{(2)} & a_{s_3}^{(2)} & a_{s_3}^{(2)} & a_{s_1}^{(L-1)} & a_{s_1}^{(L-1)} \\ a_{s_1}^{(1)} & a_{s_2}^{(2)} & a_{s_2}^{(2)} & a_{s_2}^{(2)} & a_{s_1}^{(2)} & a_{s_2}^{(2)} \\ a_{s_1}^{(2)} & a_{s_2}^{(2)} & a_{s_2}^{(2)} & a_{s_2}^{(2)} & a_{s_2}^{(2)} \\ a_{s_1}^{(2)} & a_{s_2}^{(2)} & a_{s_2}^{(2)} & a_{s_2}^{(2)} & a_{s_2}^{(2)} \\ a_{s_1}^{(2)} & a_{s_2}^{(2)} & a_{s_2}^{(2)} & a_{s_2}^{(2)} & a_{s_2}^{(2)} \\ a_{s_2}^{(2)} & a_{s_2}^{(2)} & a_{s_2}^{(2)} & a_{s_2}^{(2)} & a_{s_2}^{(2)} \\ a_{s_2}^{$$

$$\begin{split} & \Delta_{ij}^{(l)} = 0 \\ & For \ i \ from \ 1 \ to \ m \ do \\ & a^{(1)} = x^{(i)} \\ & Perform \ forward \ propagation \ to \ compute \ a^{(l)} \ for \ l = 2, \dots, L \\ & \delta^{(L)} = a^{(L)} - y^{(i)} \\ & Perform \ back \ propagation \ to \ compute \ \delta^{(l)} \ for \ l = L - 1, \dots, 2 \\ & \delta^{(l)} = \left[\Theta^{(l)}\right]^T \delta^{(l+1)}.\times \ g'\left(z^{(l)}\right) = \left[\Theta^{(l)}\right]^T \delta^{(l+1)}.\times \ a^{(l)}\left(1 - a^{(l)}\right) \\ & \Delta^{(l)}_{ij} \coloneqq \Delta^{(l)}_{ij} + a^{(l)}_{j} \delta^{(l+1)}_{i} \ or \ \Delta^{(l)} \coloneqq \Delta^{(l)} + \delta^{(l+1)}\left[a^{(l)}\right]^T \\ & D^{(l)}_{ij} \coloneqq \begin{cases} \frac{1}{m} \Delta^{(l)}_{ij} + \frac{\lambda}{m} \Theta^{(l)}_{ij} \ if \ j = 0 \\ & \frac{1}{m} \Delta^{(l)}_{ij} \ if \ j = 0 \end{cases} \\ & \frac{\partial}{\partial \Theta^{(l)}_{ij}} J(\Theta) = D^{(l)}_{ij} \end{split}$$

Examples

x_1	x_2	$x_1 \& x_2$
0	0	0
0	1	0
1	0	0
1	1	1

x_1	x_2	$x_1 \vee x_2$
0	0	0
0	1	1
1	0	1
1	1	1

x_1	x_2	$x_1 \oplus x_2$
0	0	0
0	1	1
1	0	1
1	1	0











