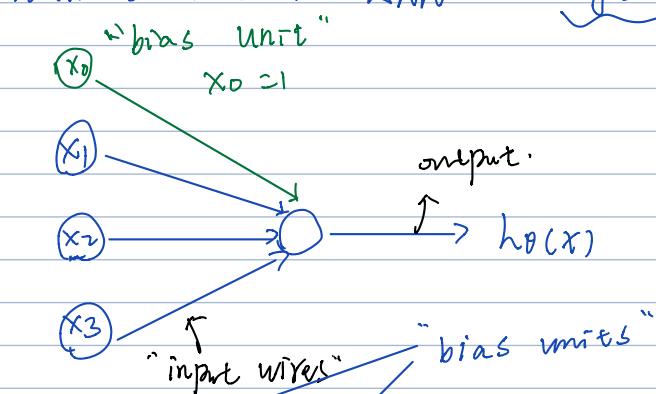


Neuron in your brain:

Input: dendrite (impulse)

Output: Axon → to other Neurons.

Neuron model: ANN ← logistic unit.

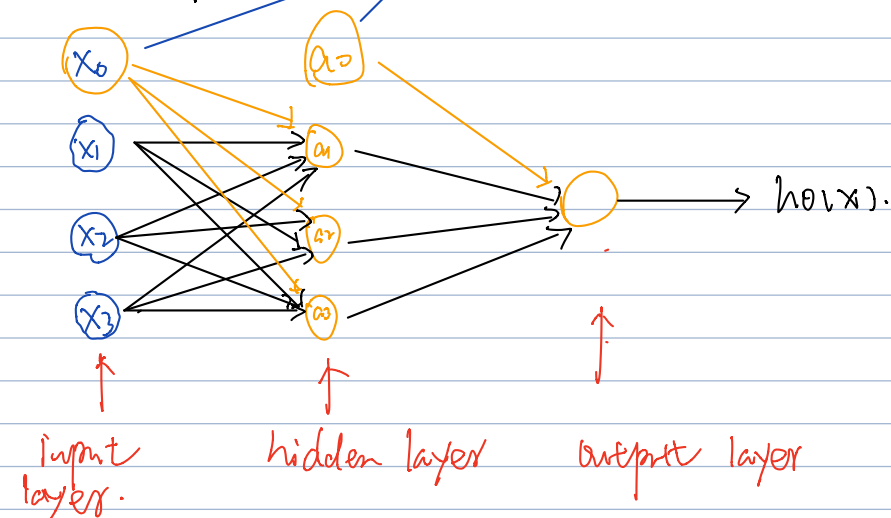


$$X = \begin{bmatrix} x_0 \\ x_1 \\ \vdots \\ x_{n-1} \end{bmatrix}$$

$$\theta = \begin{bmatrix} \theta_0 \\ \vdots \\ \theta_{n-1} \end{bmatrix}$$

↑
"weights"
(the same as parameters)

$$h_{\theta}(x) = \frac{1}{1 + e^{-\theta^T x}}$$



If network has s_j units in layer j , and s_{j+1} units in layer $j+1$, then $\theta^{(j)}$ will be of dimension $s_{j+1} \times (s_j + 1)$

Model representation:

$$z^{(2)} = \theta^{(1)} \times X$$

$$a^{(2)} = g(z^{(2)})$$

x_1, x_2 is binary

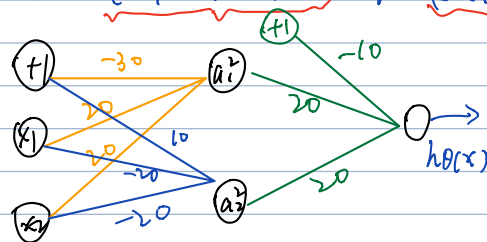
$$h_{\theta}(x) = g(-10 + 20x_1 + 20x_2)$$

x_1	x_2	$h_{\theta}(x)$
0	0	$y \approx 0$
0	1	$y \approx 1$
1	0	$y \approx 1$
1	1	$y \approx 1$

x_1 OR x_2 .

XNOR 网络:

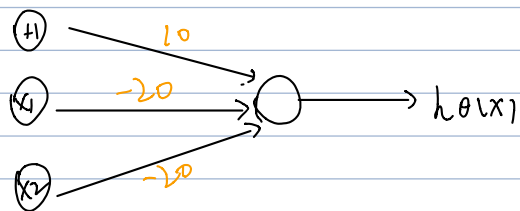
$$XNOR = (x_1 \text{ AND } x_2) \text{ OR } (\text{NOT } x_1 \text{ AND } \text{NOT } x_2)$$



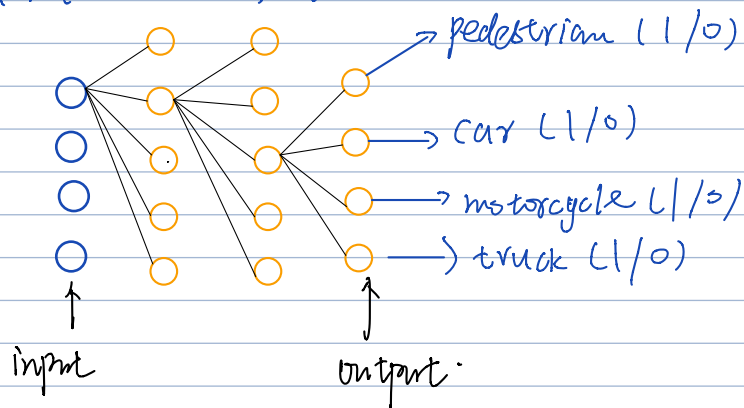
— $x_1 \text{ AND } x_2$
— $\neg x_1 \text{ AND } \neg x_2$
— $x_1 \text{ OR } x_2$

x_1	x_2	$\neg x_1 \wedge \neg x_2$	$h_{\theta}(x)$
0	0	1	1
0	1	0	0
1	0	0	0
1	1	1	1

(NOT x_1 AND NOT x_2)

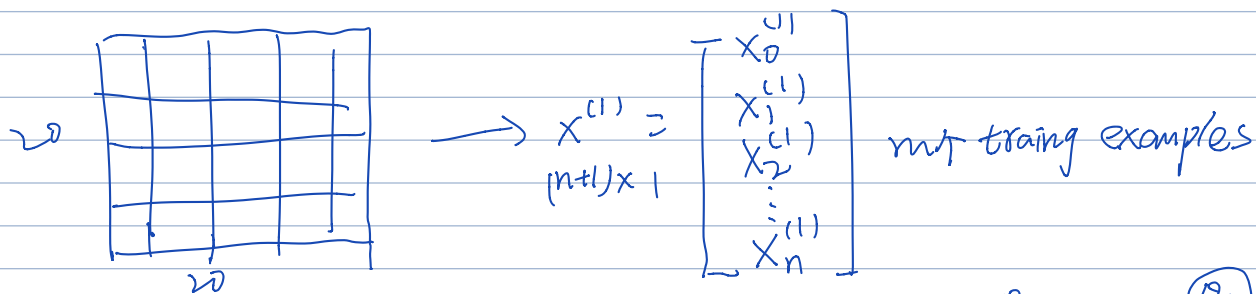


Multiclass classification



输出为四个类别的向量

$$y^{(i)} = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} \text{ or } \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix} \text{ or } \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix} \text{ or } \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}$$



$$X = \begin{bmatrix} -(x^{(1)})^T \\ -(x^{(2)})^T \\ \vdots \\ -(x^{(m)})^T \end{bmatrix}_{m \times (n+1)} = \begin{bmatrix} \theta_0 & \theta_1 & \theta_2 & \dots & \theta_n \\ x_0^{(1)} & x_1^{(1)} & x_2^{(1)} & \dots & x_n^{(1)} \\ x_0^{(2)} & x_1^{(2)} & x_2^{(2)} & \dots & x_n^{(2)} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ x_0^{(m)} & x_1^{(m)} & x_2^{(m)} & \dots & x_n^{(m)} \end{bmatrix}$$

$m = 5000$

$$y = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ \vdots \\ y_m \end{bmatrix}_{m \times 1}$$

$$\theta = \begin{bmatrix} \theta_0 \\ \theta_1 \\ \theta_2 \\ \vdots \\ \theta_n \end{bmatrix}_{(n+1) \times 1}$$

$$h_{\theta}(x) = g(\theta^T X^{(i)}) = g(\underline{X} \cdot \underline{\theta}) \Rightarrow \underline{m \times 1} \quad (n \times 1)(n \times 1)$$

$$J(\theta) = \frac{1}{m} (-y^{(i)} \cdot \log(h_{\theta}(x^{(i)})) - (1 - y^{(i)}) \log(1 - h_{\theta}(x^{(i)}))) + \frac{1}{2m} \sum_{j=1}^n \theta_j^2$$

vectorized

$$J(\theta) = \frac{1}{m} \left[\underbrace{-y^T}_{1 \times m} \cdot \underbrace{g(\underline{X} \cdot \underline{\theta})}_{m \times 1} - \underbrace{(1-y)^T}_{1 \times m} \cdot \underbrace{(1 - g(\underline{X} \cdot \underline{\theta}))}_{m \times 1} \right] + \frac{1}{2m} \cdot \underbrace{\theta [1:n]^T}_{1 \times n} * \underbrace{\theta [1:n]}_{n \times 1}$$

$$\text{grad} = \text{grad}(\text{un-regularized}) + \text{grad}(\text{regularized})$$

$$\frac{\partial}{\partial \theta_j} J(\theta) = \frac{1}{m} (h_{\theta}(x^{(i)}) - y^{(i)}) x_j^{(i)} \leftarrow \text{正则化项.}$$

$$\text{temp} = \theta$$

$$\text{temp}(0) = 0 \leftarrow \theta_0^{(i)} = 0$$

$$\text{grad}(\text{正则化}) = \frac{\lambda}{m} \cdot \theta$$

$$j=0: \theta_0^{(i)} := \theta_0^{(i)} - 2 \left[\frac{1}{m} (h_{\theta}(x^{(i)}) - y^{(i)}) \cdot x_0^{(i)} \right]$$

$$j \neq 0: \theta_j^{(i)} := \theta_j^{(i)} - 2 \left[\frac{1}{m} \sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)}) \cdot x_j^{(i)} + \frac{\lambda}{m} \sum_{j=1}^n \theta_j \right]$$