

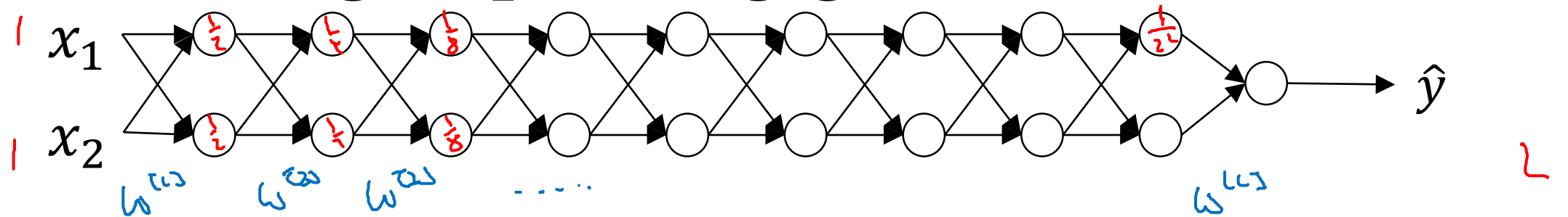


deeplearning.ai

Setting up your
optimization problem

Vanishing/exploding
gradients

Vanishing/exploding gradients



$g(z) = z$ $b^{(L)} = 0$

$\hat{y} = w^{(L)} \left(w^{(L-1)} w^{(L-2)} \dots w^{(2)} w^{(1)} x \right)$

$a^{(L)} = z^{(L)} = w^{(L)} x$

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

$a^{(L-1)} = g(z^{(L-1)}) = g(w^{(L-1)} a^{(L-2)})$

$w^{(1)} > I$

$w^{(2)} < I \quad \begin{bmatrix} 0.9 & \\ & 0.9 \end{bmatrix}$

$w^{(2)} = \begin{bmatrix} 1.5 & 0 \\ 0 & 1.5 \end{bmatrix}$

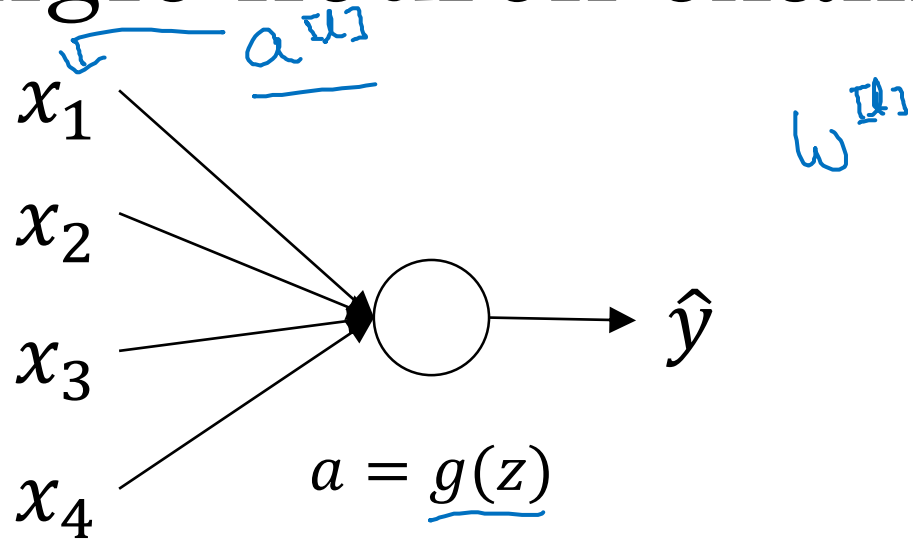
0.5 (above 1.5)
 6.5 (below 1.5)

$\hat{y} = w^{(L)} \begin{bmatrix} 1.5 & 0 \\ 0 & 1.5 \end{bmatrix}^{L-1} x$

0.5 (above 1.5)
 6.5 (below 1.5)

$1.5^{L-1} \times$
 $6.5^{L-1} \times$

Single neuron example



$$z = w_1 x_1 + w_2 x_2 + \dots + w_n x_n$$

large $n \rightarrow$ Smaller w_i

$$\text{Var}(w_i) = \frac{1}{n} \frac{2}{n}$$

$$\underline{W^{[1]}} = \text{np.random.randn}(\text{shape}) * \text{np.sqrt}\left(\frac{2}{n^{[1-1]}}\right)$$

ReLU $g^{[2]}(z) = \text{ReLU}(z)$

Other variants:

tanh

$$\frac{1}{n^{[l-1]}}$$

Xavier initialization

$$\frac{2}{n^{[l-1]} + n^{[1]}}$$