



deeplearning.ai

Multi-class  
classification

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Trying a softmax  
classifier

# Understanding softmax

(4,1)

$$\mathbf{z}^{[L]} = \begin{bmatrix} 5 \\ 2 \\ -1 \\ 3 \end{bmatrix}$$

$$\mathbf{t} = \begin{bmatrix} e^5 \\ e^2 \\ e^{-1} \\ e^3 \end{bmatrix}$$

$$C=4 \quad \mathbf{g}^{[L]}(.)$$

"soft max"

$$\mathbf{a}^{[L]} = \mathbf{g}^{[L]}(\mathbf{z}^{[L]}) = \begin{bmatrix} e^5 / (e^5 + e^2 + e^{-1} + e^3) \\ e^2 / (e^5 + e^2 + e^{-1} + e^3) \\ e^{-1} / (e^5 + e^2 + e^{-1} + e^3) \\ e^3 / (e^5 + e^2 + e^{-1} + e^3) \end{bmatrix} = \begin{bmatrix} 0.842 \\ 0.042 \\ 0.002 \\ 0.114 \end{bmatrix}$$

"hard max"

$$\begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

Softmax regression generalizes logistic regression to  $C$  classes.

If  $C=2$ , softmax reduces to logistic regression.

# Loss function

$$y^{(1)} = \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix} \quad \hat{y}^{(1)} = \begin{bmatrix} 0.3 \\ 0.2 \\ 0.1 \\ 0.4 \end{bmatrix} \quad C=4$$

$$L(\hat{y}, y) = - \sum_{i=1}^4 y_i \log \hat{y}_i$$

$y_2 \log \hat{y}_2 = -\log \hat{y}_2$

make  $\hat{y}_2$  bigger,  $L \downarrow$

$$J(W^{[1]}, b^{[1]}, \dots) = \frac{1}{m} \sum_{i=1}^m L(\hat{y}^{(i)}, y^{(i)})$$

$$Y = [y^{(1)}, y^{(2)}, \dots, y^{(m)}]$$

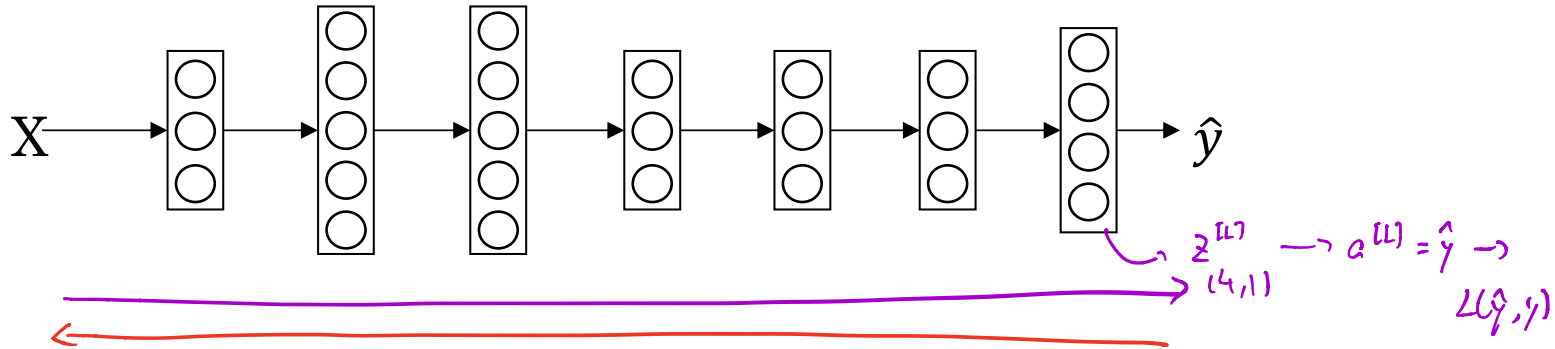
$(4, m)$

$$= \begin{bmatrix} 0 & & \\ 1 & \dots & \\ 0 & & \end{bmatrix}$$

$$\hat{Y} = [\hat{y}^{(1)}, \hat{y}^{(2)}, \dots, \hat{y}^{(m)}]$$

$$= \begin{bmatrix} 0.3 & & \\ 0.2 & \dots & \\ 0.1 & & \\ 0.4 & & \end{bmatrix}$$

# Summary of softmax classifier



Backprop :

$$\frac{\partial \mathcal{L}}{\partial z^{[L]}} = \hat{y} - y$$