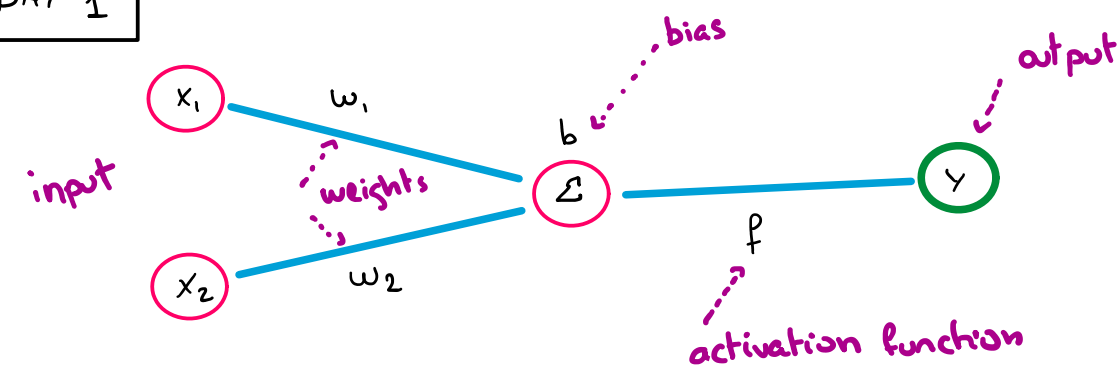


Easy Nets (Day 2)

dimecres, 25 de novembre de 2020 4:16 p. m.

DAY 1



Example:

$f(x) = \frac{1}{1+e^{-x}}$  ,  $\frac{\partial f}{\partial x} = \frac{e^{-x}}{(1+e^{-x})^2} = \frac{1-f(x)}{f(x)}$

then 
$$\text{input } (x_1, x_2) \xrightarrow[\text{bias } b]{\text{weights } w_1, w_2} \underbrace{x_1 \cdot w_1 + x_2 \cdot w_2 + b}_q \xrightarrow[\text{activation } f]{} \frac{1}{1 + e^{-(x_1 \cdot w_1 + x_2 \cdot w_2 + b)}} = f(x_1, w_1, x_2, w_2, b)$$

• to compute partial derivatives, use chain rule:

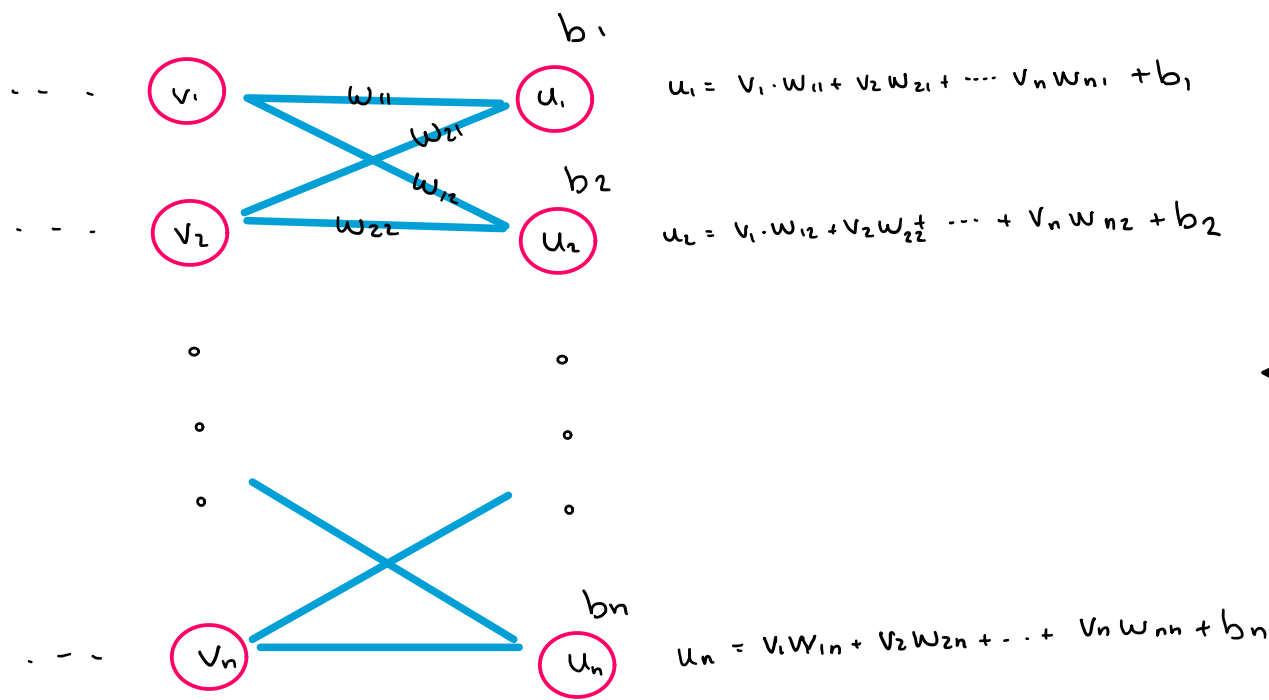
• suppose:  $\begin{cases} x_0 = -1 \\ x_1 = -2 \\ w_0 = 4 \\ w_1 = -4 \\ b = -3 \end{cases}$  , then we have  $q = x_1 w_1 + x_2 w_2 + b$   
 $q = (-2)(-4) + (-1) \cdot 4 - 3 = 1$

$$\frac{\partial f}{\partial x_1} = \frac{\partial f}{\partial q} \cdot \frac{\partial q}{\partial x_1} = \frac{1-f(q)}{f(q)} \cdot w_1 \Big|_{\substack{q=1 \\ w_1=-4}} = \frac{1-f(1)}{f(1)} \cdot (-4) = -4/e \simeq -1.1$$

$$\frac{\partial f}{\partial b} = \frac{\partial f}{\partial q} \cdot \frac{\partial q}{\partial b} = \frac{1-f(q)}{f(q)} \cdot 1 \Big|_{q=1} = \frac{1}{e}$$
 (and so on ...)

$$\frac{\partial f}{\partial w_0} = \frac{\partial f}{\partial q} \cdot \frac{\partial q}{\partial w_0} = \frac{1-f(q)}{f(q)} \cdot x_0 \Big|_{\substack{x_0=-1 \\ q=1}} = -1/e$$

Linear Layer:  $L$



forward

$$\begin{bmatrix} w_{11} & w_{12} & \dots & w_{1n} \\ w_{21} & & & \\ \vdots & & & \\ w_{n1} & \dots & w_{nn} \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix} + \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix} = \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{bmatrix}$$
  
$$\underbrace{\begin{bmatrix} w_{11} & w_{12} & \dots & w_{1n} \\ w_{21} & & & \\ \vdots & & & \\ w_{n1} & \dots & w_{nn} \end{bmatrix}}_W \cdot \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix} + \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix} = \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{bmatrix}$$
  
$$W \cdot x_t + b = x_{t+1}$$

backward

$$\frac{\partial \mathcal{L}}{\partial v_j} = \sum \frac{\partial \mathcal{L}}{\partial u_i} \cdot \frac{\partial u_i}{\partial v_j} = \sum \frac{\partial \mathcal{L}}{\partial u_i} \cdot w_{ji}$$

chain rule

$$\begin{bmatrix} w_{11} & w_{12} & \dots & w_{1n} \\ w_{21} & & & \\ \vdots & & & \\ w_{n1} & \dots & w_{nn} \end{bmatrix} \begin{bmatrix} \partial \mathcal{L} / \partial u_1 \\ \partial \mathcal{L} / \partial u_2 \\ \vdots \\ \partial \mathcal{L} / \partial u_n \end{bmatrix} = \begin{bmatrix} \partial \mathcal{L} / \partial v_1 \\ \partial \mathcal{L} / \partial v_2 \\ \vdots \\ \partial \mathcal{L} / \partial v_n \end{bmatrix}$$
  
$$W^T \delta_{t+1} = \delta_t$$