



# CS231n Lecture 4

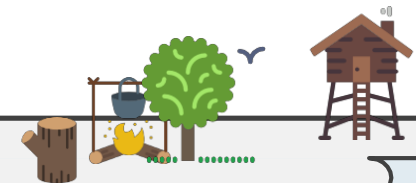
- ☑ BOAZ 10기 박성현
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- ☑ BOAZ 11기 홍지민

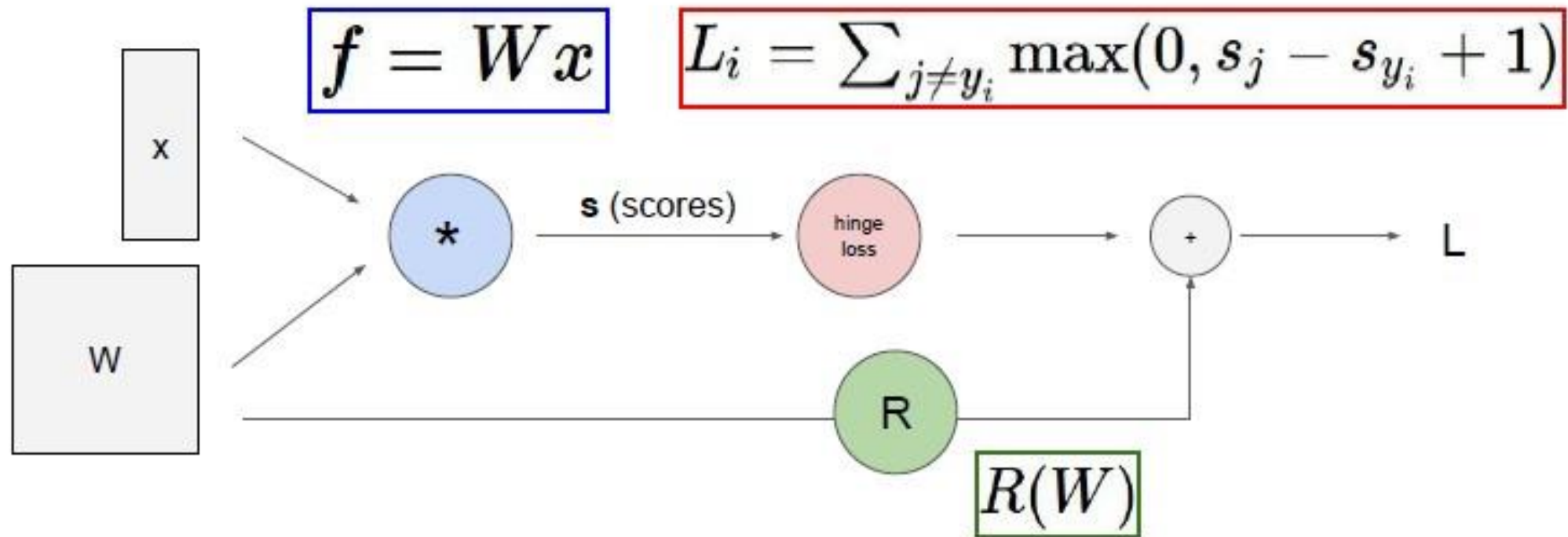
$$\frac{df(x)}{dx} = \lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{h}$$

**Numerical gradient:** slow :(, approximate :(, easy to write :)

**Analytic gradient:** fast :), exact :), error-prone :(

In practice: Derive analytic gradient, check your implementation with numerical gradient





# Backpropagation example

Backpropagation: a simple example

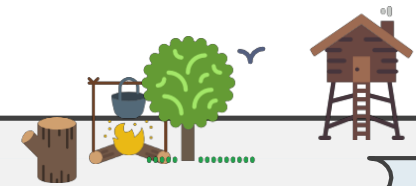
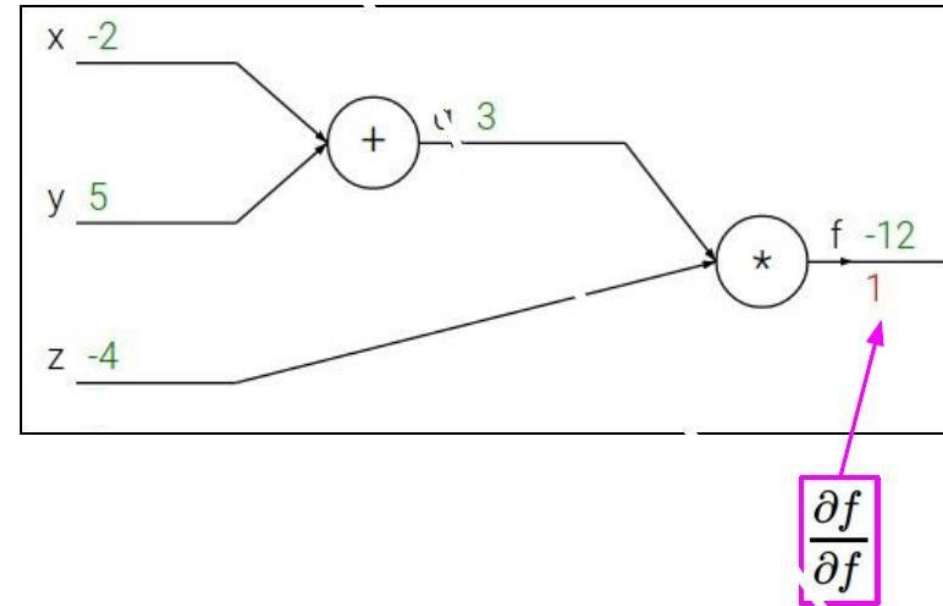
$$f(x, y, z) = (x + y)z$$

e.g.  $x = -2, y = 5, z = -4$

$$q = x + y \quad \frac{\partial q}{\partial x} = 1, \frac{\partial q}{\partial y} = 1$$

$$f = qz \quad \frac{\partial f}{\partial q} = z, \frac{\partial f}{\partial z} = q$$

Want:  $\frac{\partial f}{\partial x}, \frac{\partial f}{\partial y}, \frac{\partial f}{\partial z}$



# Backpropagation example

Backpropagation: a simple example

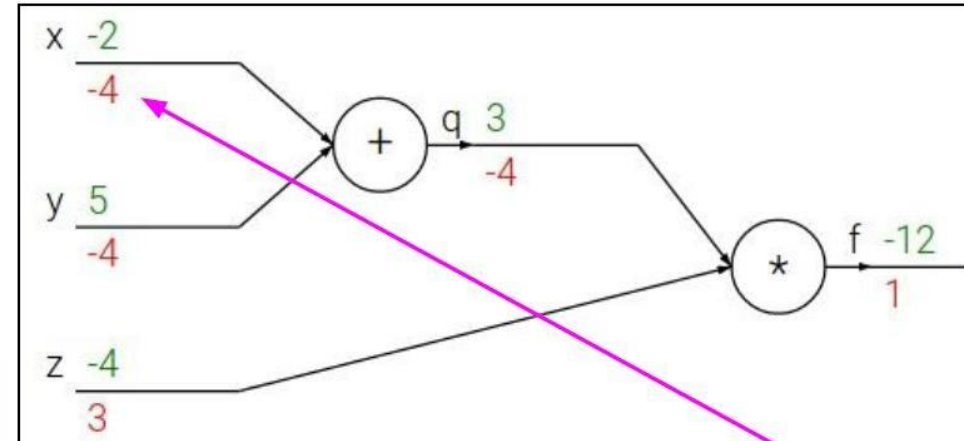
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Want:  $\frac{\partial f}{\partial x}, \frac{\partial f}{\partial y}, \frac{\partial f}{\partial z}$



Chain rule:


$$\frac{\partial f}{\partial x} = \frac{\partial f}{\partial q} \frac{\partial q}{\partial x}$$

$\underbrace{\quad}_z \quad \underbrace{\quad}_1$

$$\frac{\partial f}{\partial x}$$

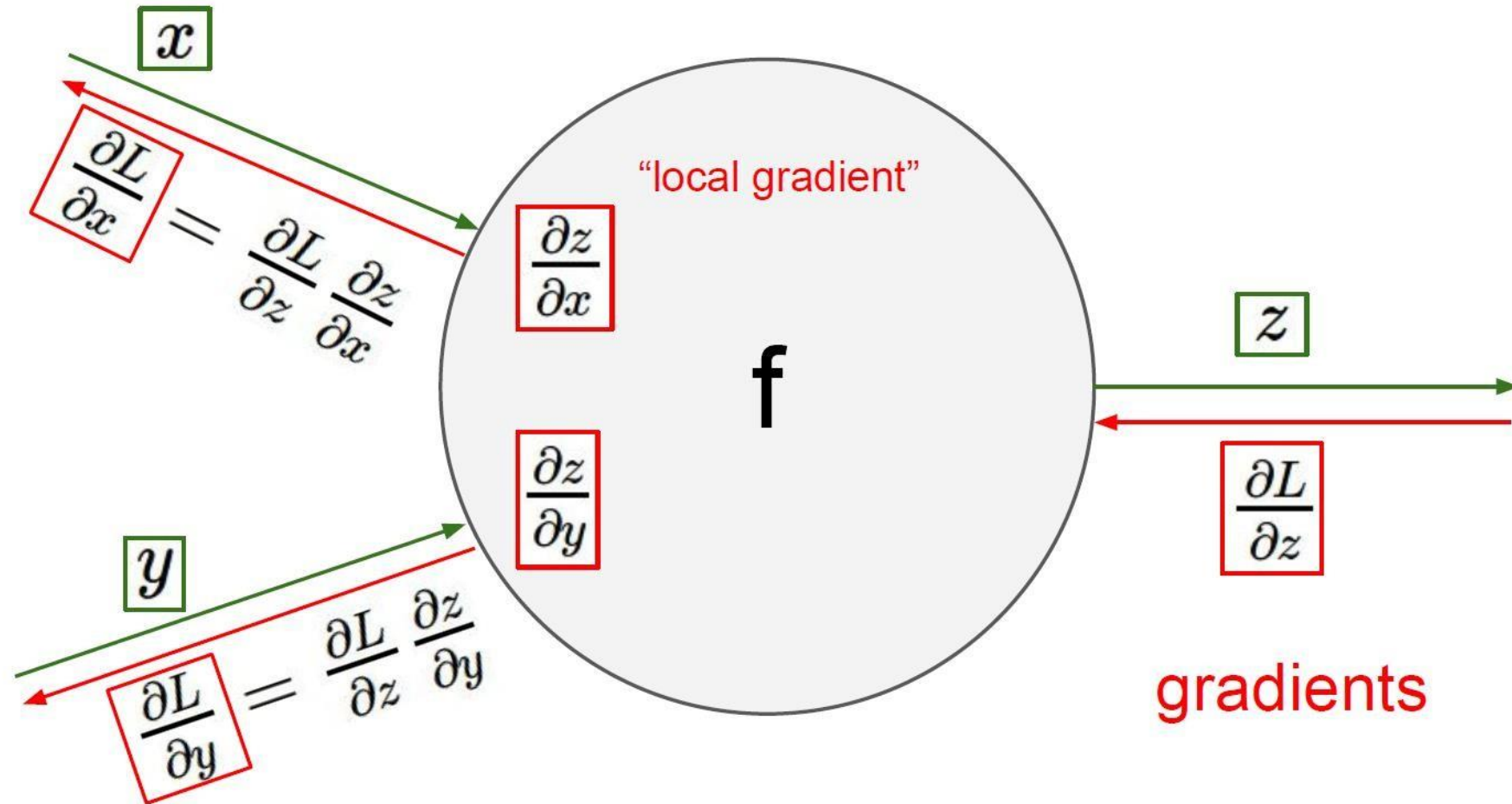
11  
-4



$$f'(x) = (g(h(x)))' = g'(h(x)) h'(x)$$


- keep the inside
  - take derivative of outside
- multiply by derivative of the inside







# What is Jacobian Matrix?

The Jacobian matrix of the function  $\mathbf{F} : \mathbb{R}^3 \rightarrow \mathbb{R}^4$  with components

$$y_1 = x_1$$

$$y_2 = 5x_3$$

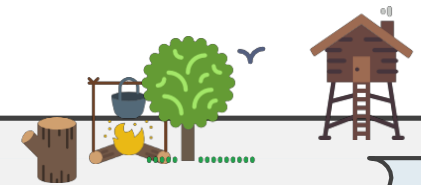
$$y_3 = 4x_2^2 - 2x_3$$

$$y_4 = x_3 \sin x_1$$

is

$$\mathbf{J}_{\mathbf{F}}(x_1, x_2, x_3) = \begin{bmatrix} \frac{\partial y_1}{\partial x_1} & \frac{\partial y_1}{\partial x_2} & \frac{\partial y_1}{\partial x_3} \\ \frac{\partial y_2}{\partial x_1} & \frac{\partial y_2}{\partial x_2} & \frac{\partial y_2}{\partial x_3} \\ \frac{\partial y_3}{\partial x_1} & \frac{\partial y_3}{\partial x_2} & \frac{\partial y_3}{\partial x_3} \\ \frac{\partial y_4}{\partial x_1} & \frac{\partial y_4}{\partial x_2} & \frac{\partial y_4}{\partial x_3} \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 5 \\ 0 & 8x_2 & -2 \\ x_3 \cos x_1 & 0 & \sin x_1 \end{bmatrix}.$$

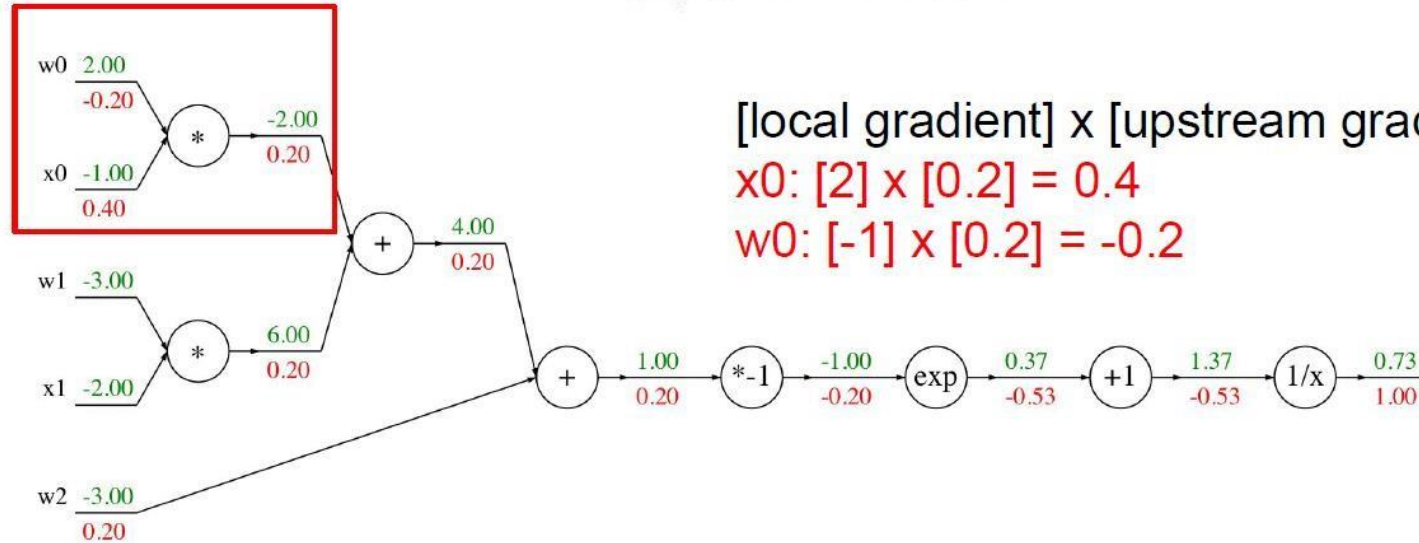
This example shows that the Jacobian need not be a square matrix.





# Another Backpropagation Example

Another example:  $f(w, x) = \frac{1}{1 + e^{-(w_0x_0 + w_1x_1 + w_2)}}$

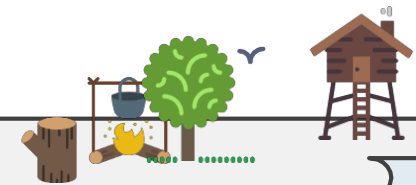


[local gradient] x [upstream gradient]

$x_0: [2] \times [0.2] = 0.4$

$w_0: [-1] \times [0.2] = -0.2$

$f(x) = e^x$	→	$\frac{df}{dx} = e^x$		$f(x) = \frac{1}{x}$	→	$\frac{df}{dx} = -1/x^2$
$f_a(x) = ax$	→	$\frac{df}{dx} = a$		$f_c(x) = c + x$	→	$\frac{df}{dx} = 1$

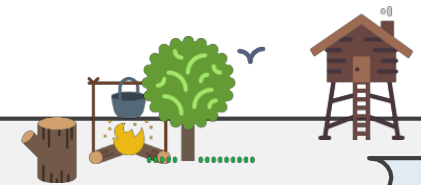
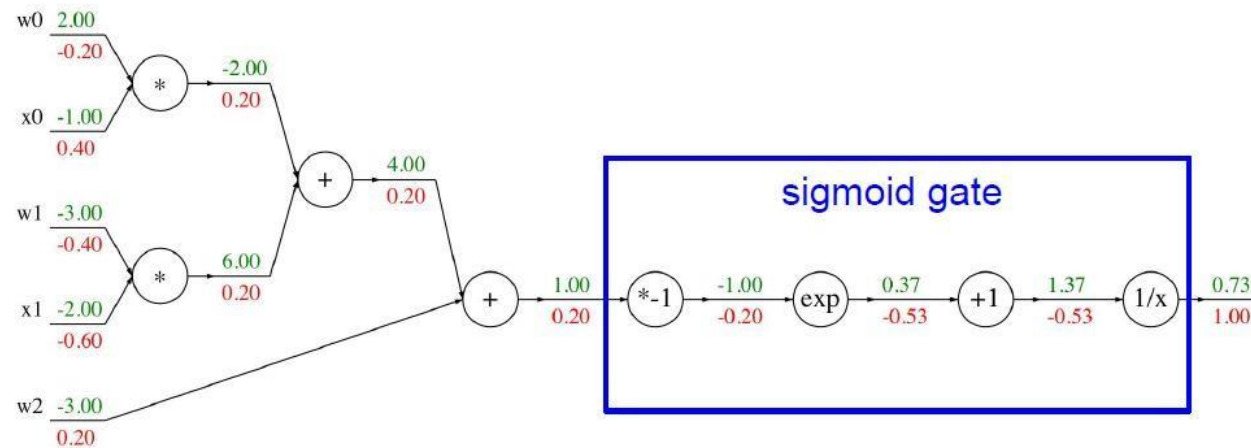


# Derivative of sigmoid function

$$f(w, x) = \frac{1}{1 + e^{-(w_0 x_0 + w_1 x_1 + w_2)}}$$

$$\sigma(x) = \frac{1}{1 + e^{-x}} \quad \text{sigmoid function}$$

$$\frac{d\sigma(x)}{dx} = \frac{e^{-x}}{(1 + e^{-x})^2} = \left( \frac{1 + e^{-x} - 1}{1 + e^{-x}} \right) \left( \frac{1}{1 + e^{-x}} \right) = (1 - \sigma(x)) \sigma(x)$$

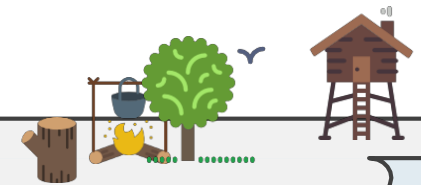
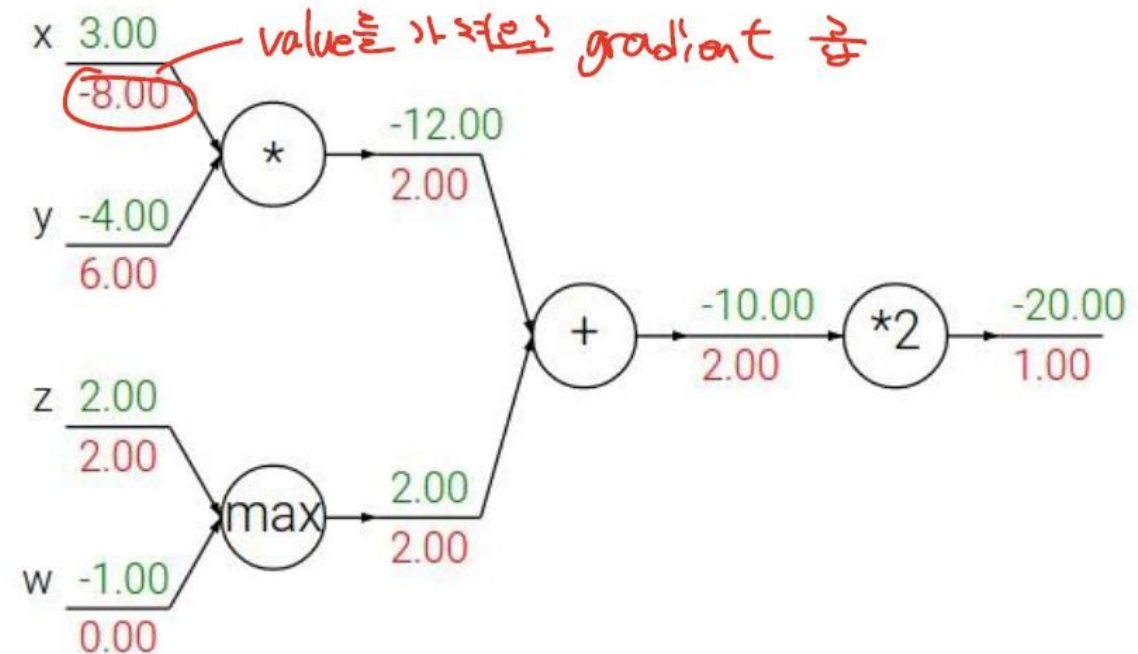


# Patterns in derivative of various gates

**add** gate: gradient distributor

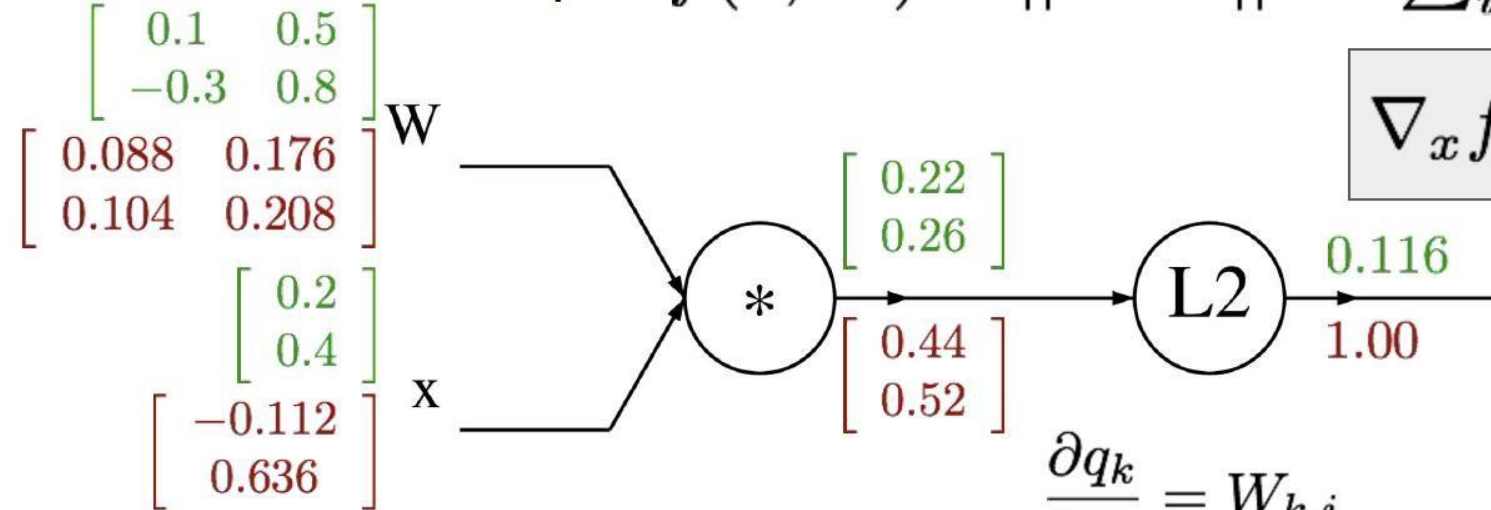
**max** gate: gradient router

**mul** gate: gradient switcher



# Backpropagation with Vector

A vectorized example:  $f(x, W) = ||W \cdot x||^2 = \sum_{i=1}^n (W \cdot x)_i^2$

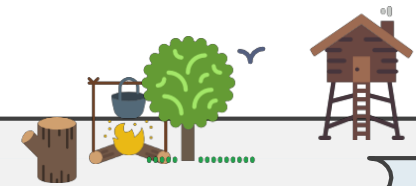


$$\nabla_x f = 2W^T \cdot q$$

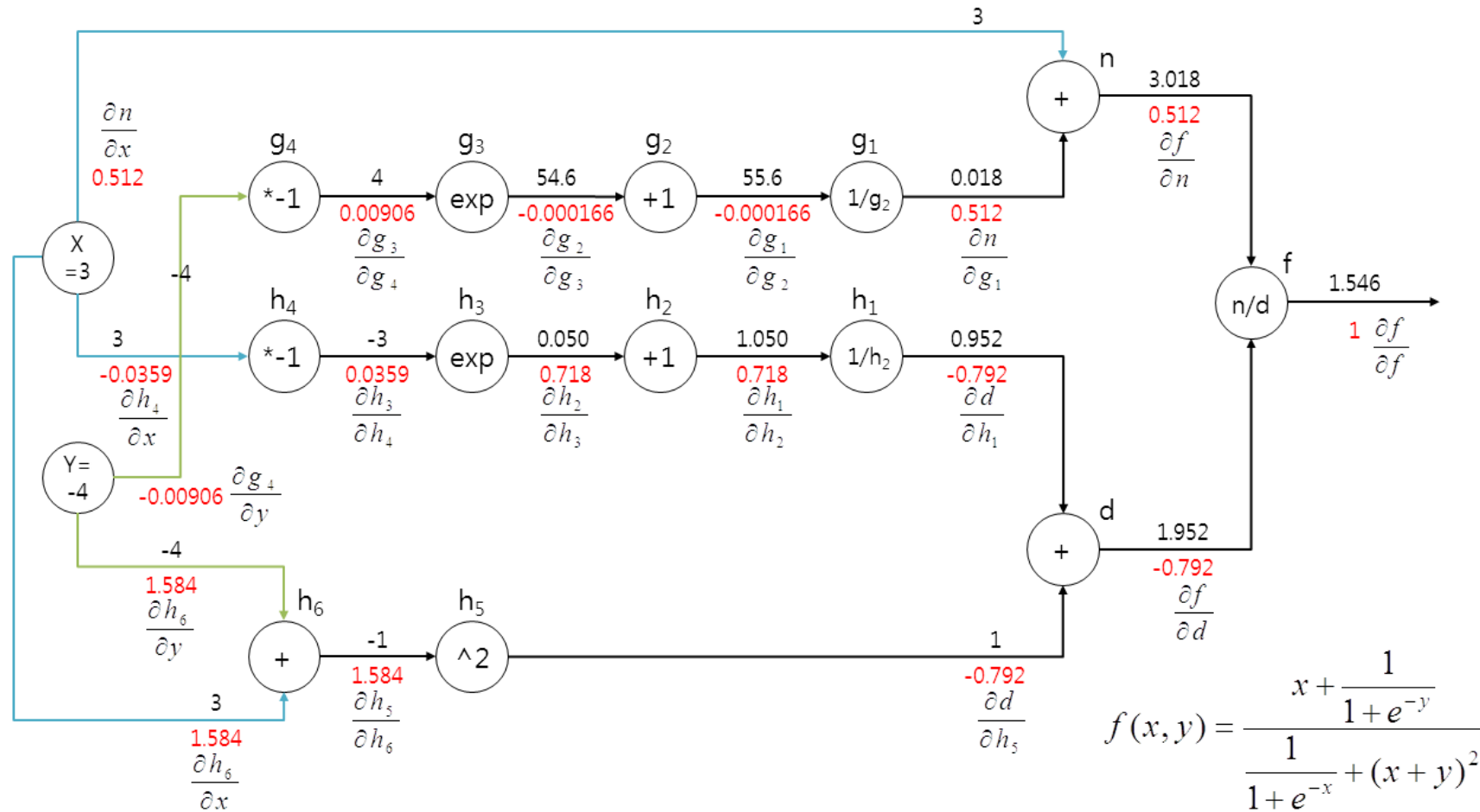
$$q = W \cdot x = \begin{pmatrix} W_{1,1}x_1 + \dots + W_{1,n}x_n \\ \vdots \\ W_{n,1}x_1 + \dots + W_{n,n}x_n \end{pmatrix}$$

$$f(q) = ||q||^2 = q_1^2 + \dots + q_n^2$$

$$\begin{aligned} \frac{\partial q_k}{\partial x_i} &= W_{k,i} \\ \frac{\partial f}{\partial x_i} &= \sum_k \frac{\partial f}{\partial q_k} \frac{\partial q_k}{\partial x_i} \\ &= \sum_k 2q_k W_{k,i} \end{aligned}$$



# Backpropagation with Vector



CS231n : <http://cs231n.stanford.edu/syllabus.html>

