Backpropagation

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Gradient Descent

Network parameters
$$\theta = \{w_1, w_2, \dots, b_1, b_2, \dots\}$$

$$\theta^0 \longrightarrow \theta^1 \longrightarrow \theta^2 \longrightarrow \dots$$

Parameters
$$\nabla L(\theta)$$

$$= \begin{bmatrix} \partial L(\theta)/\partial w_1 \\ \partial L(\theta)/\partial w_2 \\ \vdots \\ \partial L(\theta)/\partial b_1 \\ \partial L(\theta)/\partial b_2 \\ \vdots \end{bmatrix}$$
 Compute
$$\nabla L(\theta^0)$$

$$\theta^1 = \theta^0 - \eta \nabla L(\theta^0)$$

$$\theta^2 = \theta^1 - \eta \nabla L(\theta^1)$$

$$\theta^2 = \theta^1 - \eta \nabla L(\theta^1)$$
 Millions of parameters
 To compute the gradients efficiently, we use **backpropagation**.

Compute
$$\nabla L(\theta^0)$$
 $\theta^1 = \theta^0 - \eta \nabla L(\theta^0)$

Compute
$$\nabla L(\theta^1)$$

$$\theta^2 = \theta^1 - \eta \nabla L(\theta^1)$$

Chain Rule

$$y = g(x)$$
 $z = h(y)$

$$\Delta x \to \Delta y \to \Delta z$$

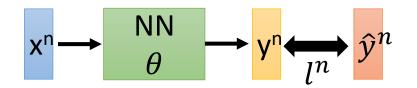
$$\frac{dz}{dx} = \frac{dz}{dy} \frac{dy}{dx}$$

Case 2

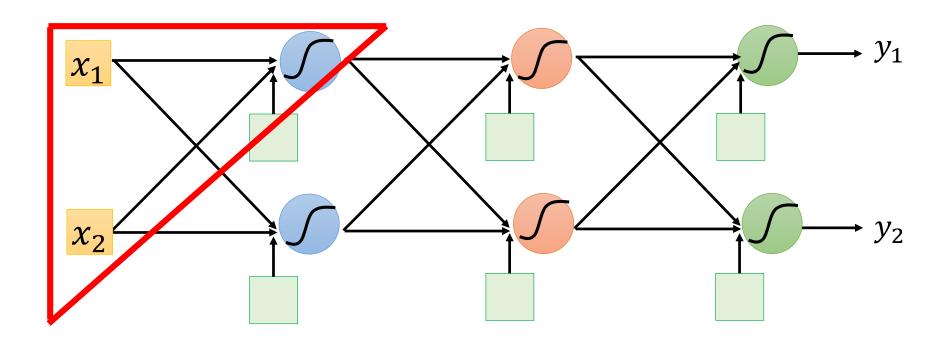
$$x = g(s)$$
 $y = h(s)$ $z = k(x, y)$

$$\Delta s = \frac{\partial z}{\partial x} \frac{\partial x}{\partial s} + \frac{\partial z}{\partial y} \frac{\partial y}{\partial s}$$

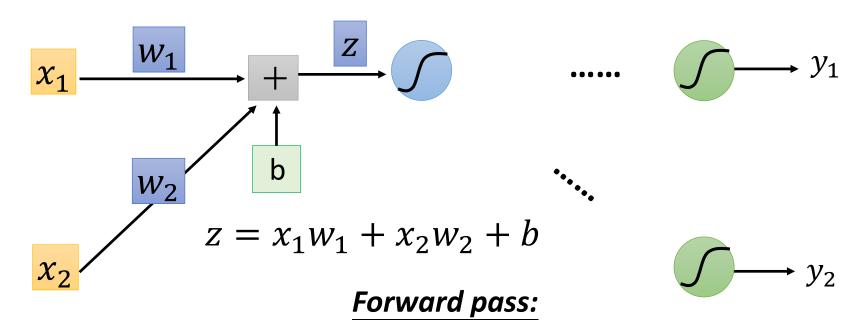
Backpropagation



$$L(\theta) = \sum_{n=1}^{N} l^{n}(\theta) \qquad \qquad \frac{\partial L(\theta)}{\partial w} = \sum_{n=1}^{N} \frac{\partial l^{n}(\theta)}{\partial w}$$



Backpropagation



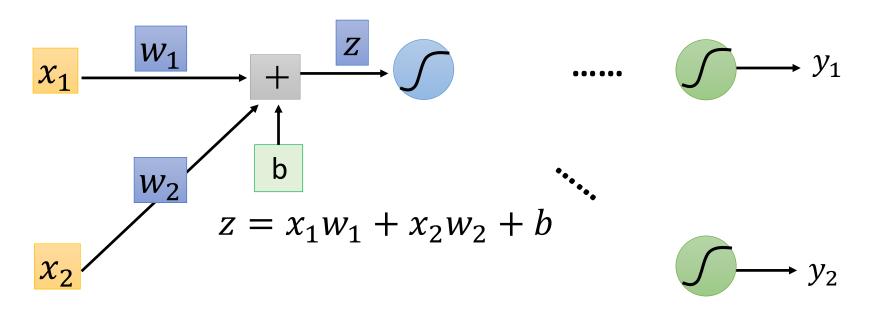
$$\frac{\partial l}{\partial w} = ? \quad \frac{\partial z}{\partial w} \frac{\partial l}{\partial z}$$
(Chain rule)

Compute $\partial z/\partial w$ for all parameters

Backward pass:

Backpropagation – Forward pass

Compute $\partial z/\partial w$ for all parameters



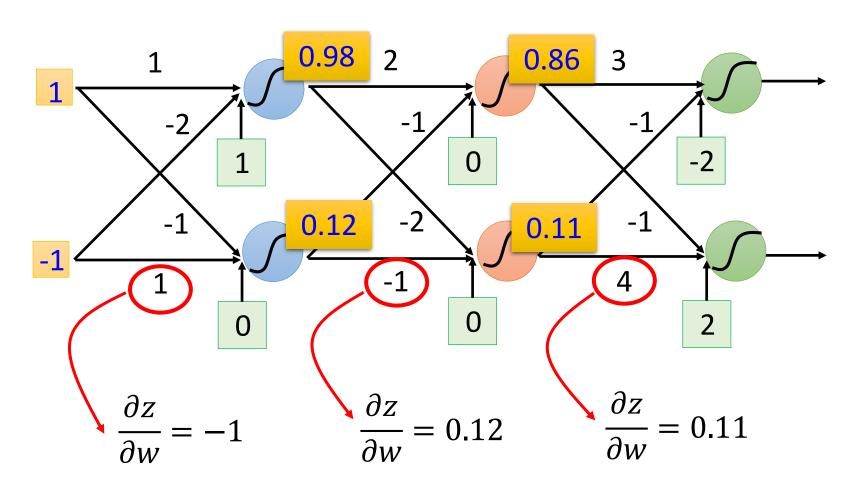
$$\frac{\partial z}{\partial w_1} = ? x_1$$

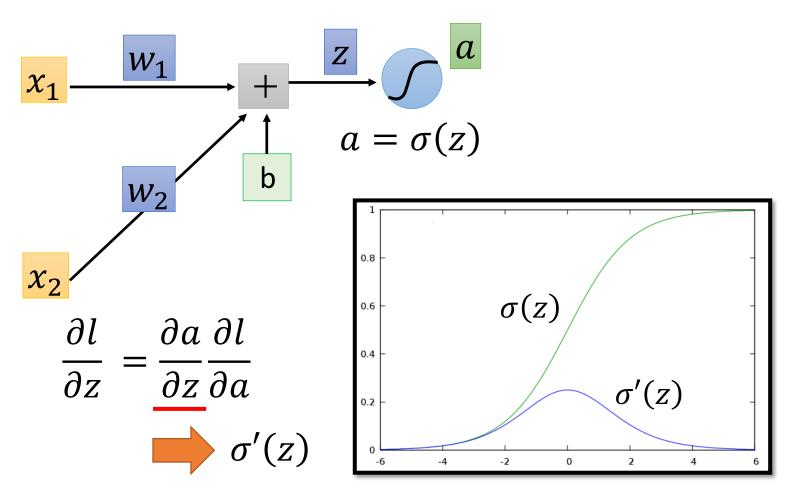
$$\frac{\partial z}{\partial w_2} = ? x_2$$

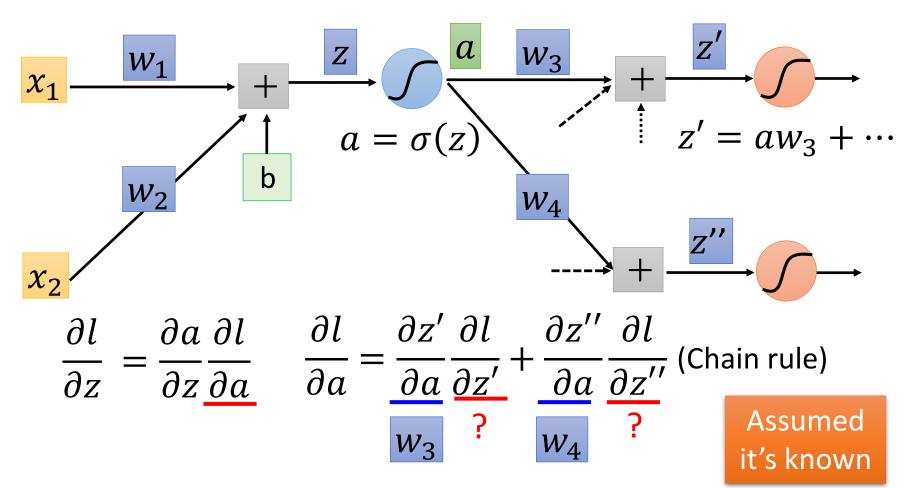
The value of the input connected by the weight

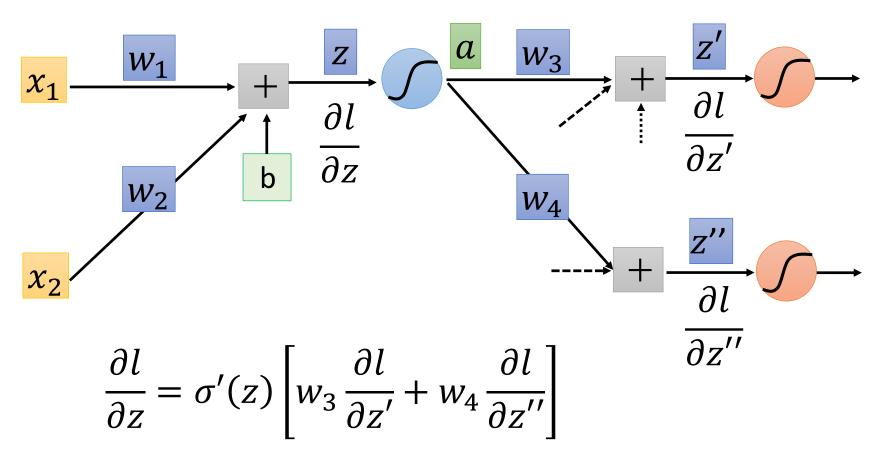
Backpropagation – Forward pass

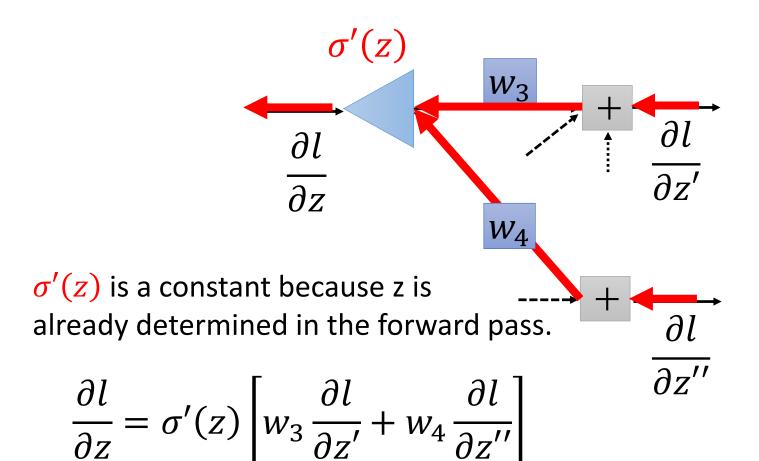
Compute $\partial z/\partial w$ for all parameters



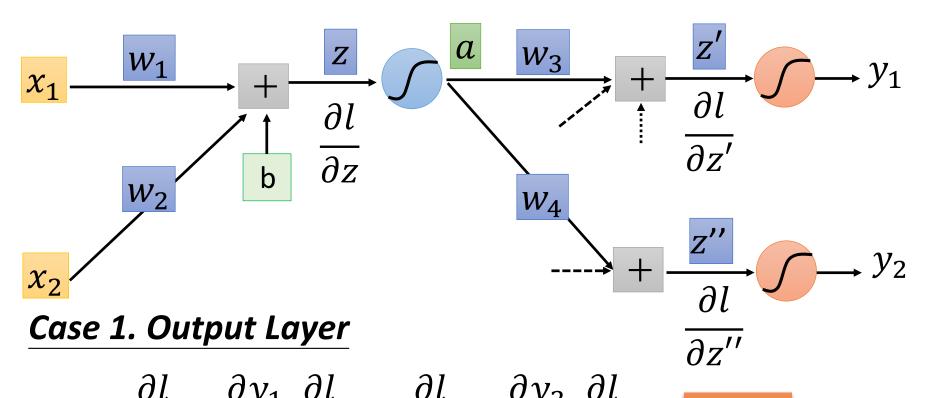








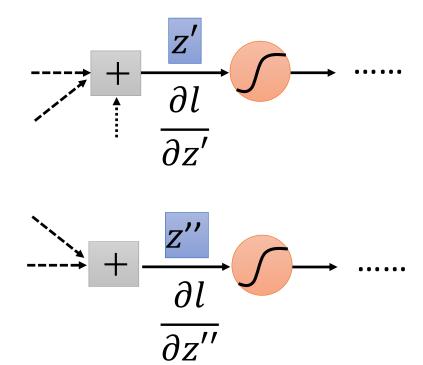
Compute $\partial l/\partial z$ for all activation function inputs z



Done!

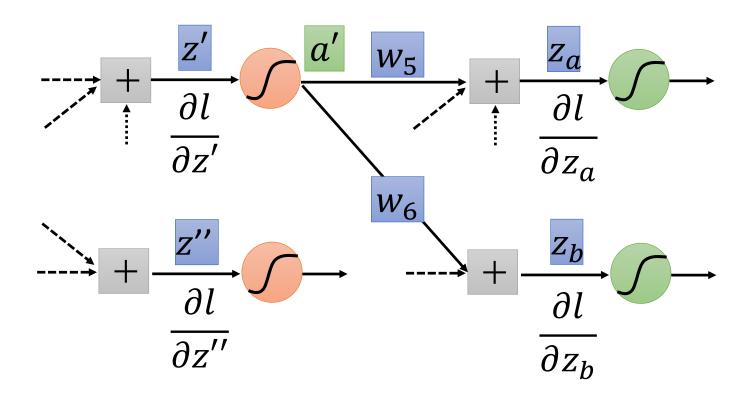
Compute $\partial l/\partial z$ for all activation function inputs z

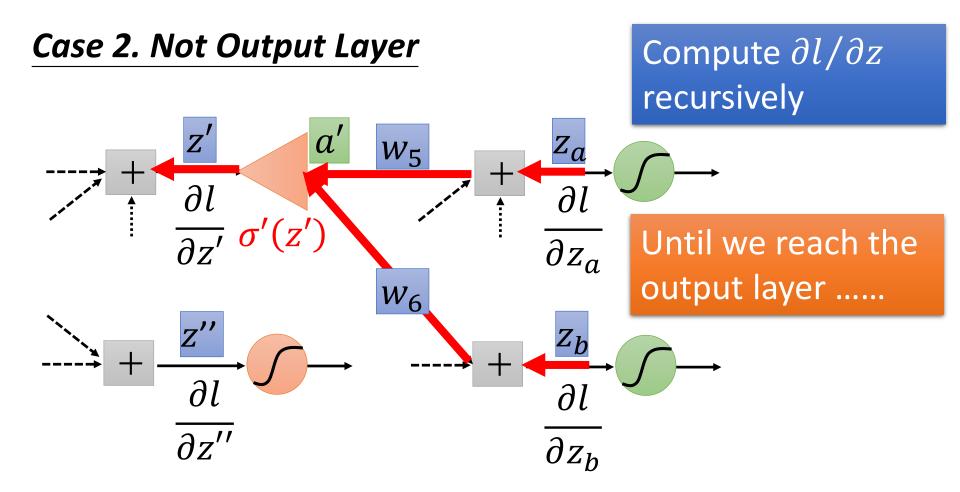
Case 2. Not Output Layer



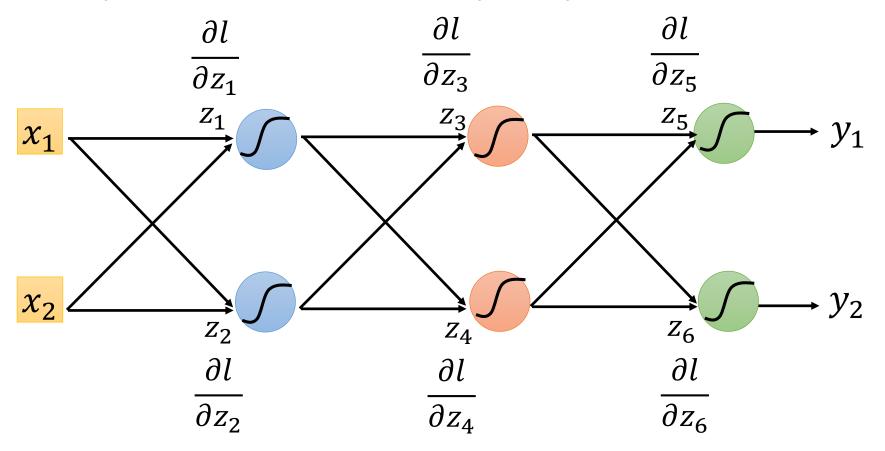
Compute $\partial l/\partial z$ for all activation function inputs z

Case 2. Not Output Layer

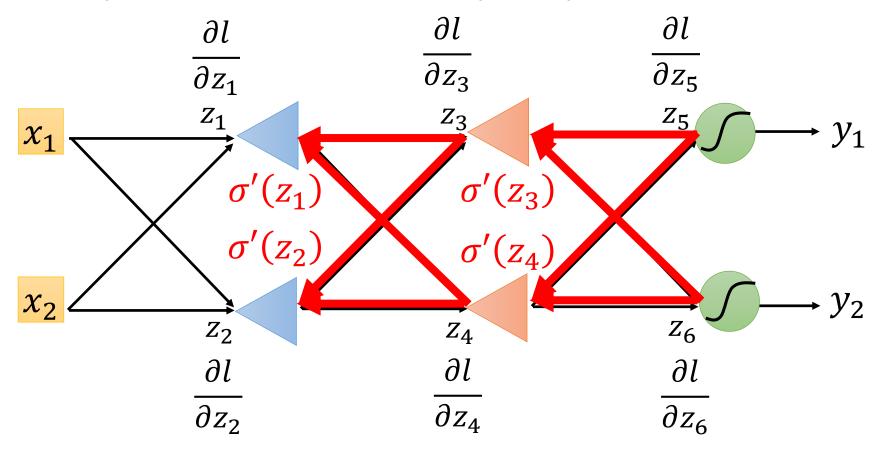




Compute $\partial l/\partial z$ for all activation function inputs z Compute $\partial l/\partial z$ from the output layer



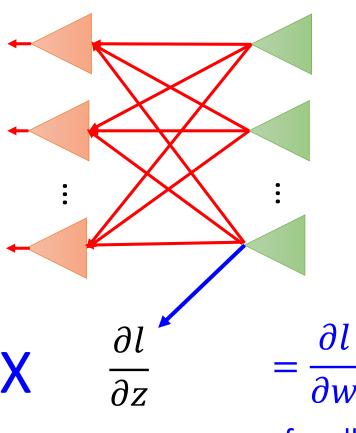
Compute $\partial l/\partial z$ for all activation function inputs z Compute $\partial l/\partial z$ from the output layer



Backpropagation – Summary

Forward Pass

Backward Pass



for all w