

CONVOLUTIONAL NEURAL NETWORK

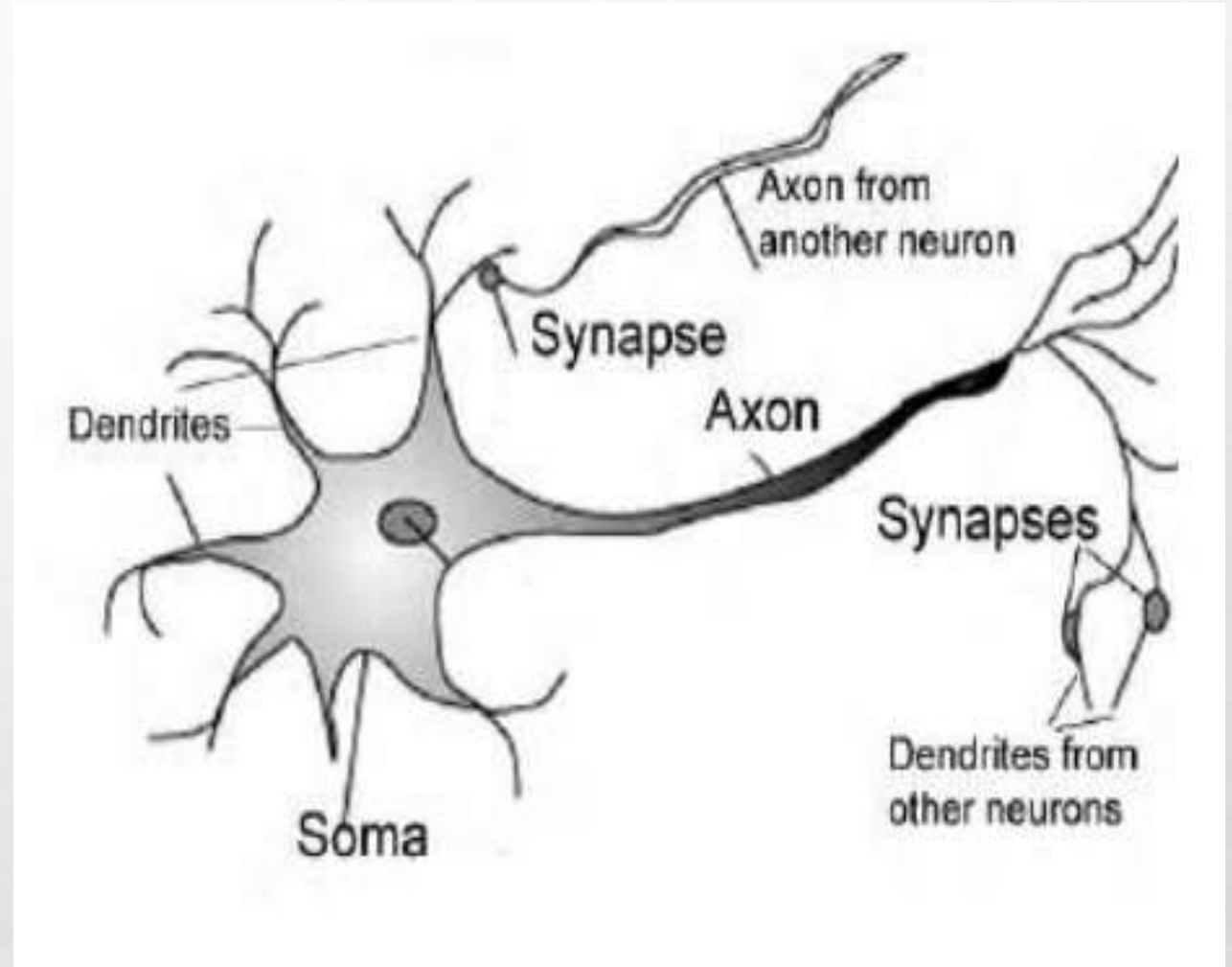


Neural Network & Deep Learning

Algoritma yang mengimitasi kemampuan otak manusia (*biological neuron*) untuk mengenali pola tertentu dan dapat melakukan proses pembelajaran.

Neuron

Unit Cerdas yang
Memiliki
Kemampuan Untuk
Belajar



$$z = \theta_1 a_1 + \theta_2 a_2 + \cdots + \theta_n a_n + b$$

Convolutional Neural Network

1

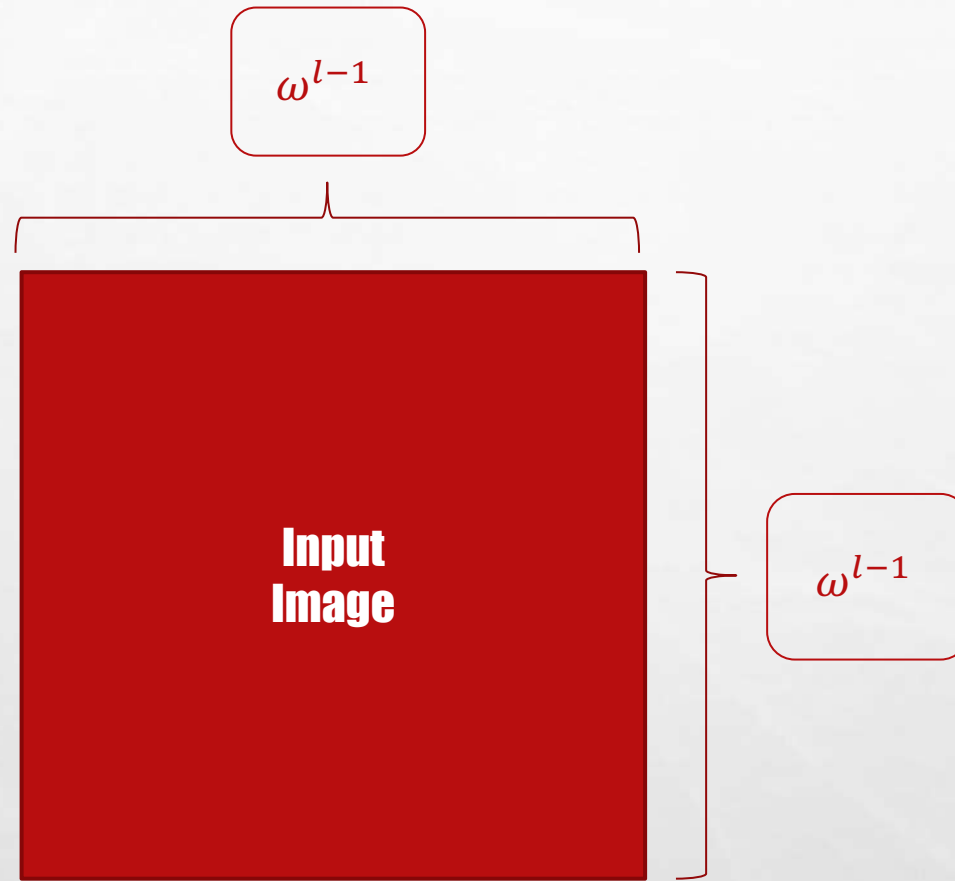
- Local receptive fields

2

- Shared weights

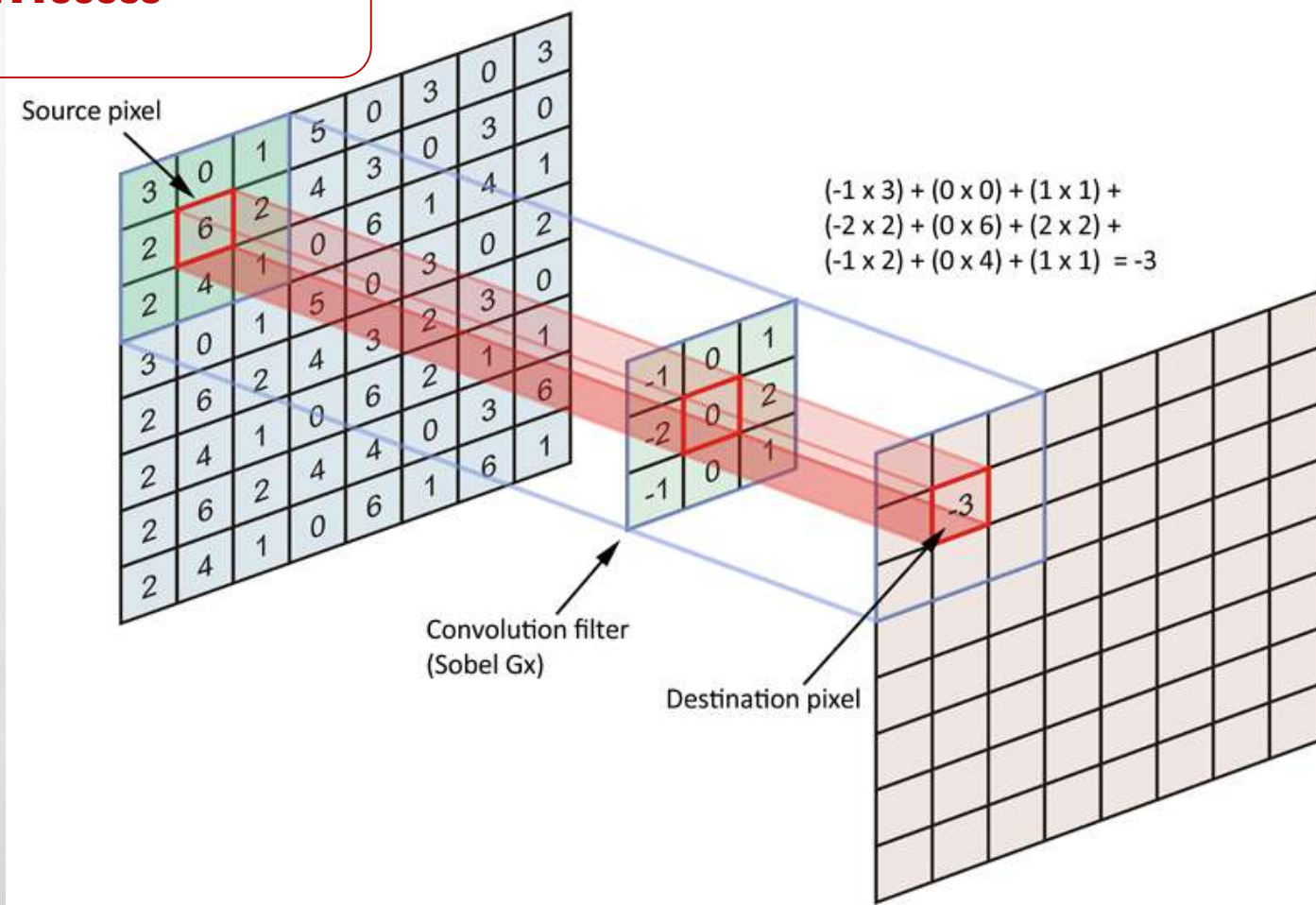
3

- Spatial subsampling

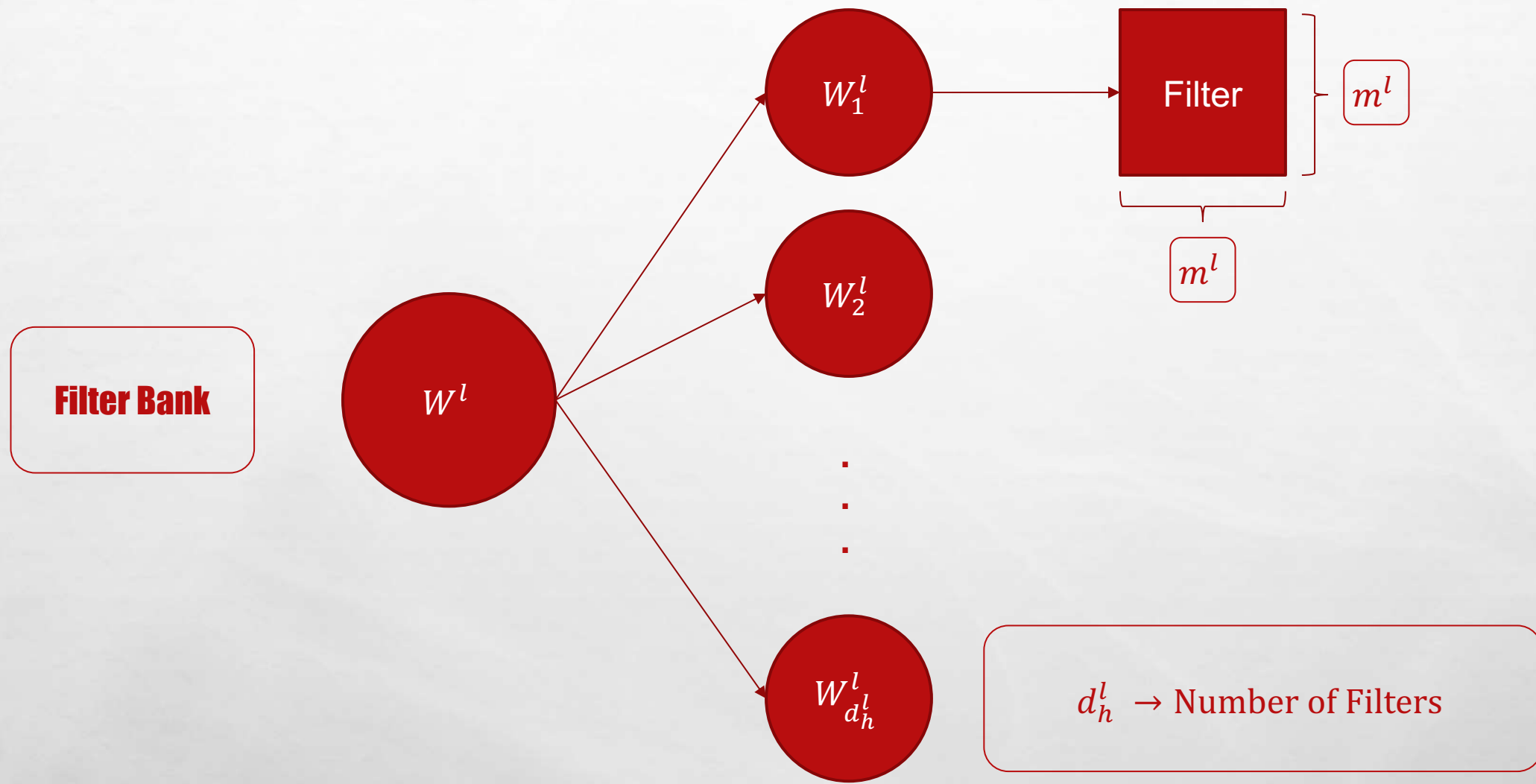


INPUT

Convolution Process



CONVOLUTIONAL LAYER



CONVOLUTIONAL LAYER

Feature Maps

$$g_k^l = I_p^{l-1} * W_k^l$$

Input



$$\omega^{l-1} \times \omega^{l-1}$$

Convolution

Filter

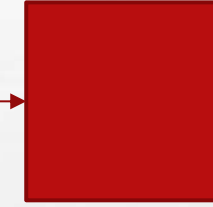


⋮

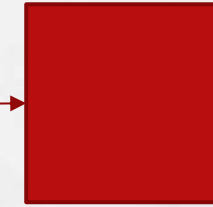


$$m^l \times m^l$$

Feature Maps

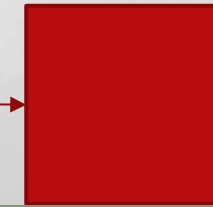


$$k = 1$$



$$k = 2$$

⋮



$$k = d_h^l$$

$$(\omega^{l-1} - m^l + 1) \times (\omega^{l-1} - m^l + 1)$$

CONVOLUTIONAL LAYER

Max Pooling

Reduce
Resolution

Spatial
Invariance

Feature Map



$$\begin{matrix} (\omega^{l-1} - m^l + 1) \\ \times \\ (\omega^{l-1} - m^l + 1) \end{matrix}$$

Window

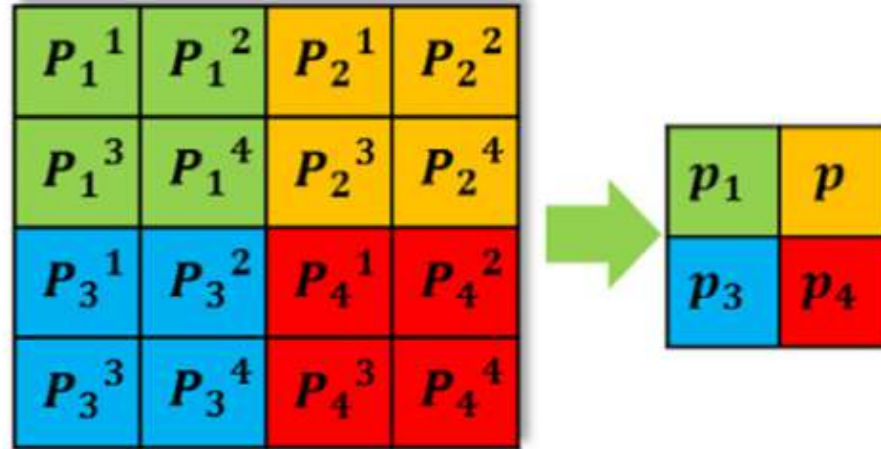


$$s \times s$$

Output



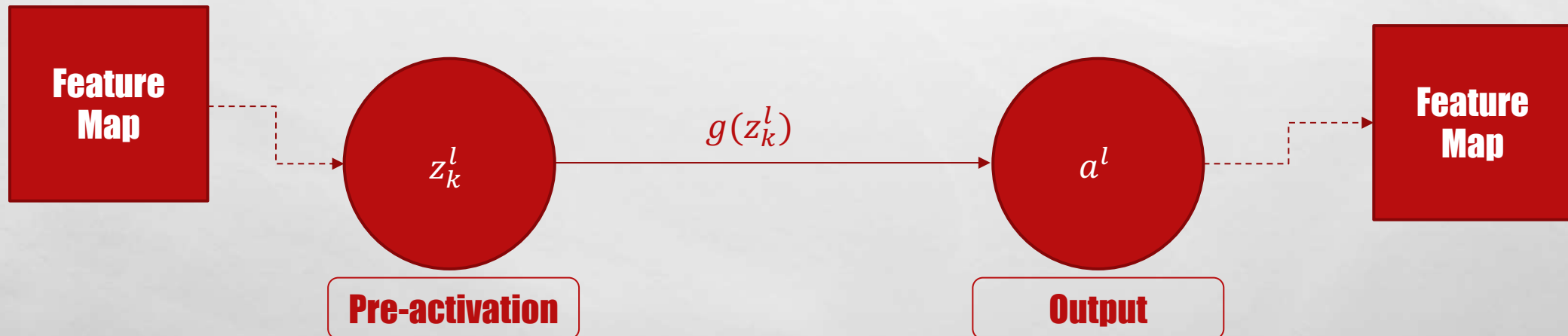
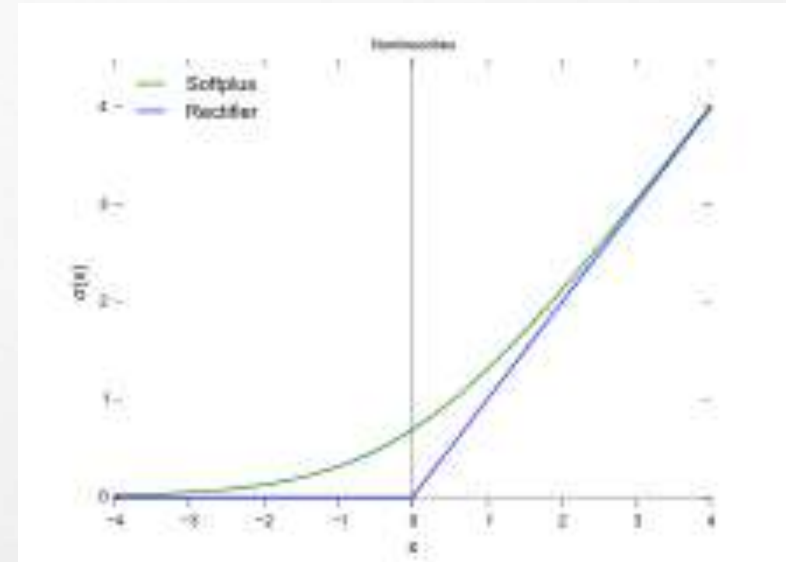
$$\begin{matrix} (\omega^{l-1} - m^l + 1) / s \\ \times \\ (\omega^{l-1} - m^l + 1) / s \end{matrix}$$



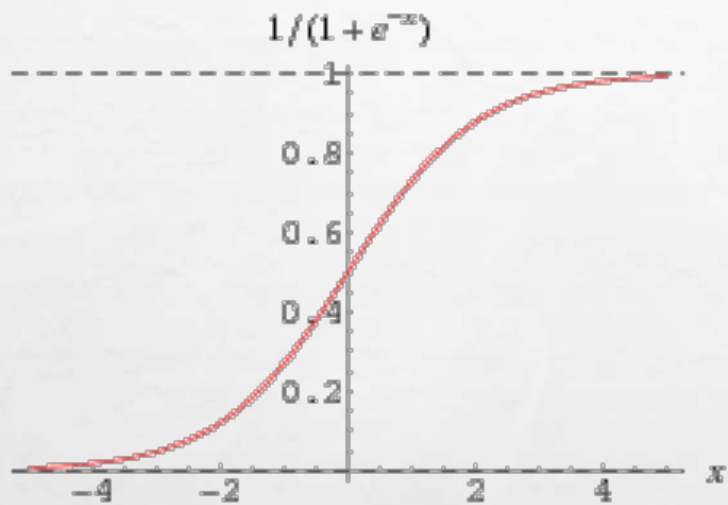
MAX-POOLING LAYER

Activation Function: Relu

$$x^l = f(g_k^l) = \max(0, x)$$

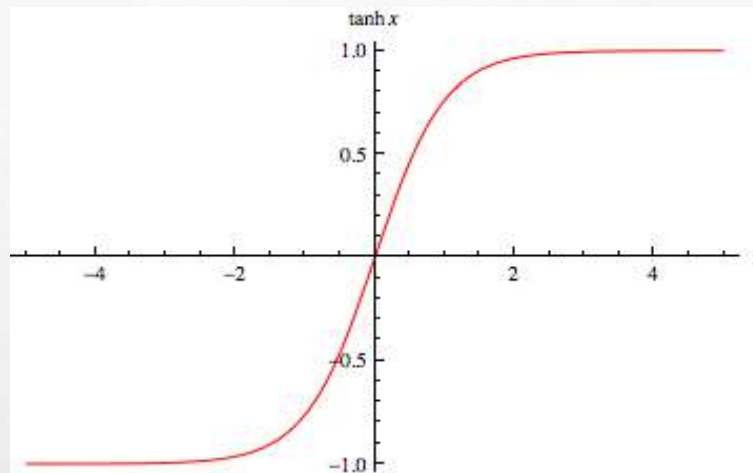


ACTIVATION FUNCTION



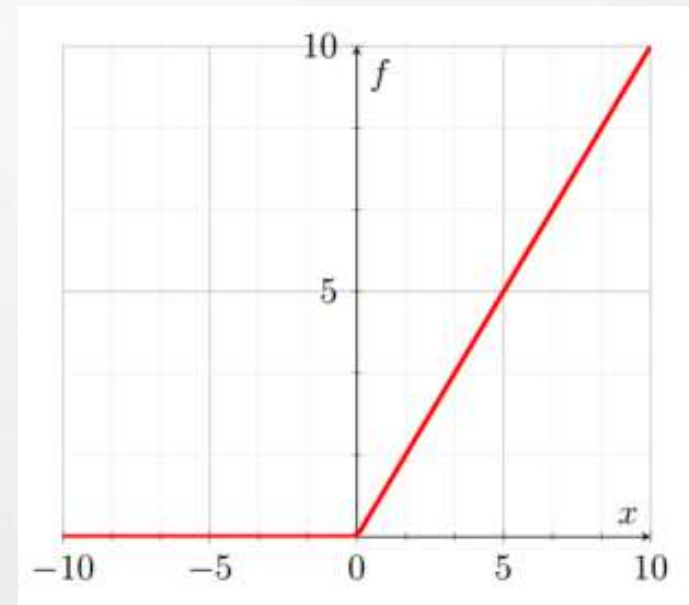
Sigmoid

$$g(z) = \frac{1}{1 + e^{-z}}$$



Tanh

$$g(z) = \frac{e^z - e^{-z}}{e^z + e^{-z}}$$



Relu

$$g(z) = \max(0, z)$$

ACTIVATION FUNCTION

Loss function: Gap antara prediction dan actual

$$J(a^{(L)}, y) = a^{(L)} - y$$

Loss function Regresi

Mean Absolute Error	$MAE(a^{(L)}, y) = \frac{1}{n} \sum_{i=1}^n a_i^L - y $	[2.13]
Mean Square Error	$MSE = \frac{1}{n} \sum_{i=1}^n (a_i^L - y)^2$	[2.14]
Mean Square Logarithmic Error	$MSLE = \frac{1}{n} \sum_{i=1}^n (\log(a_i^L + 1) - \log(y + 1))^2$	[2.15]
L1	$L1 = \sum_{i=1}^n a_i^L - y $	[2.16]
L2	$L2 = \sum_{i=1}^n (a_i^L - y)^2$	[2.17]

Loss function Klasifikasi

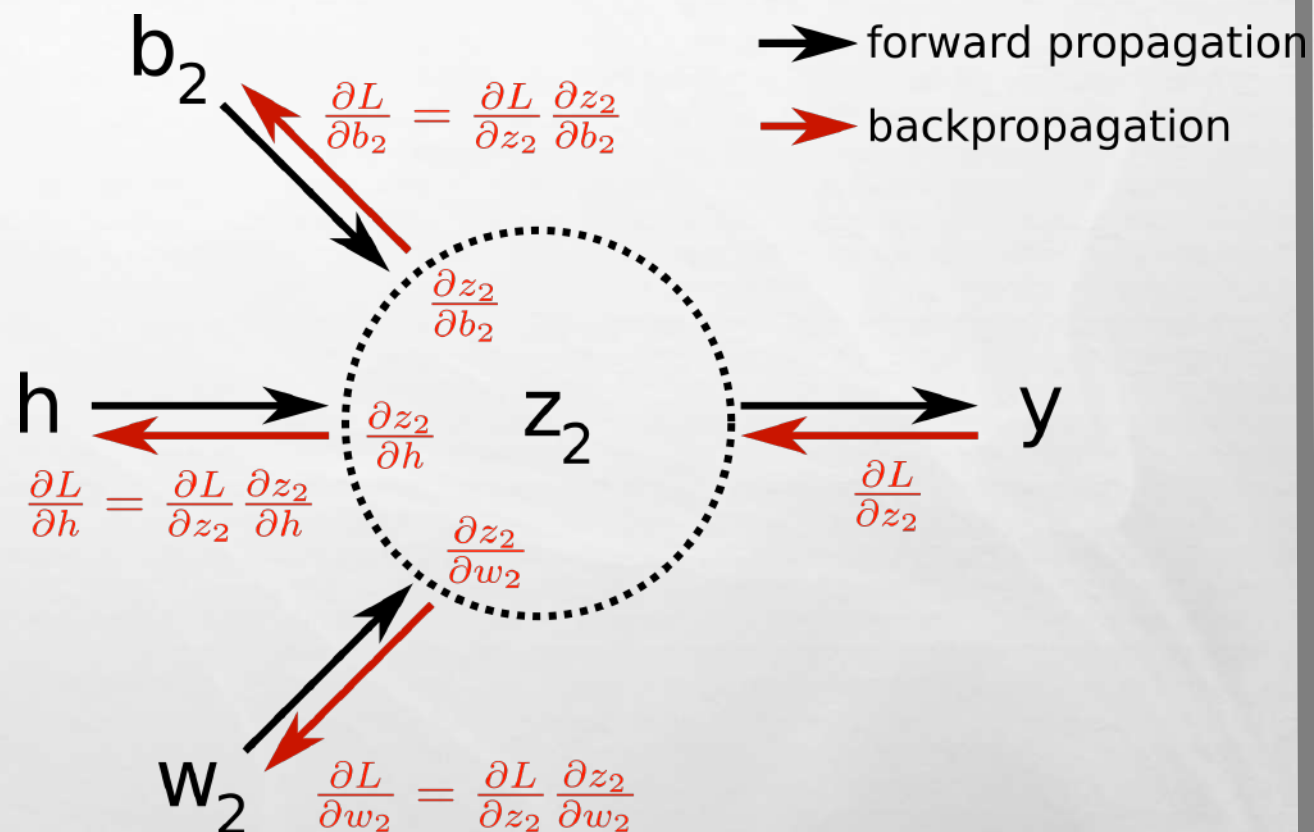
Negative log likelihood	$NLL = -\frac{1}{n} \sum_{i=1}^n \log(a_i^L)$	[2.18]
Binary Cross Entropy	$BCE = -(y_i \log(a_i^L)) + ((1 - y_i) \log(1 - a_i^L))$	[2.19]
Cross Entropy	$CE = -\sum_{i=1}^n y \log(a_i^L)$	[2.20]
Kullback Leibler	$KL = \frac{1}{n} \sum_{i=1}^n (y_i \log(y_i)) - \frac{1}{n} \sum_{i=1}^n y_i \log(a_i^L)$	[2.21]
Hinge	$HINGE = \frac{1}{n} \sum_{i=1}^n \max(0, 1 - y_i \cdot a_i^L)$	[2.22]

LOSS FUNCTION

$$\delta^{(l)} = \sum_{i=1}^n \left(\delta_i^{(l+1)} \cdot \theta_i^{(l+1)} \right) \cdot g'(z^{(l)})$$

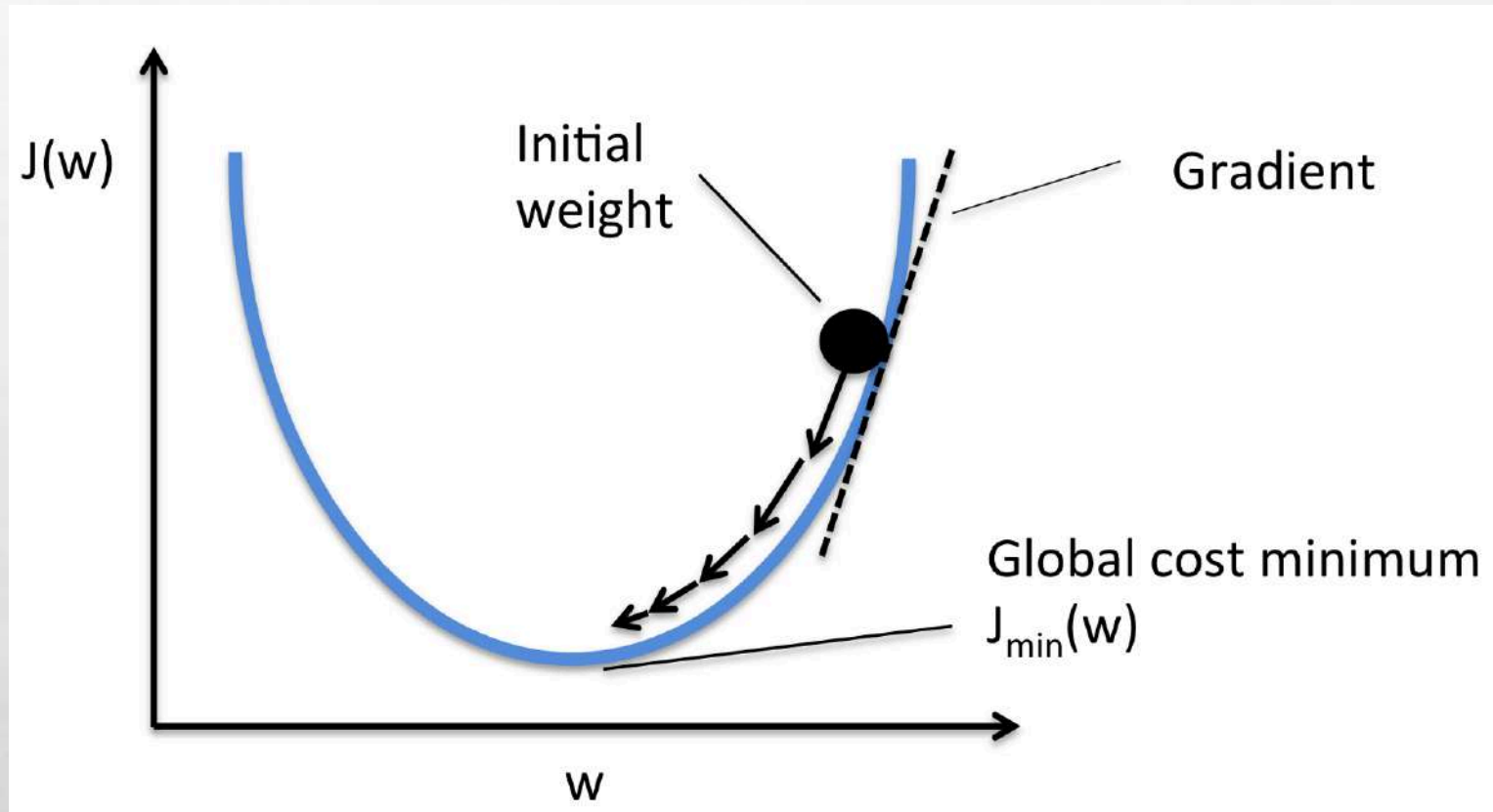
$$\delta_{\theta}^{(l)} = \delta^{(l)} \cdot a^{(l-1)}$$

$$\delta_b^{(l)} = \delta^{(l)}$$



LEARNING ALGORITHM (BACKPROPAGATION)

$$\theta_{new}^l = \theta_{old}^l - \gamma \frac{\partial J(a^{(L)}, y)}{\partial \theta}$$



GRADIENT DESCENT

TERIMA KASIH

