

Arquitectura de Redes Neuronales

Universidad Autónoma de Manizales
Semillero de Bioinformática e Inteligencia Artificial



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Fully Connected Neural Networks

Computer Vision

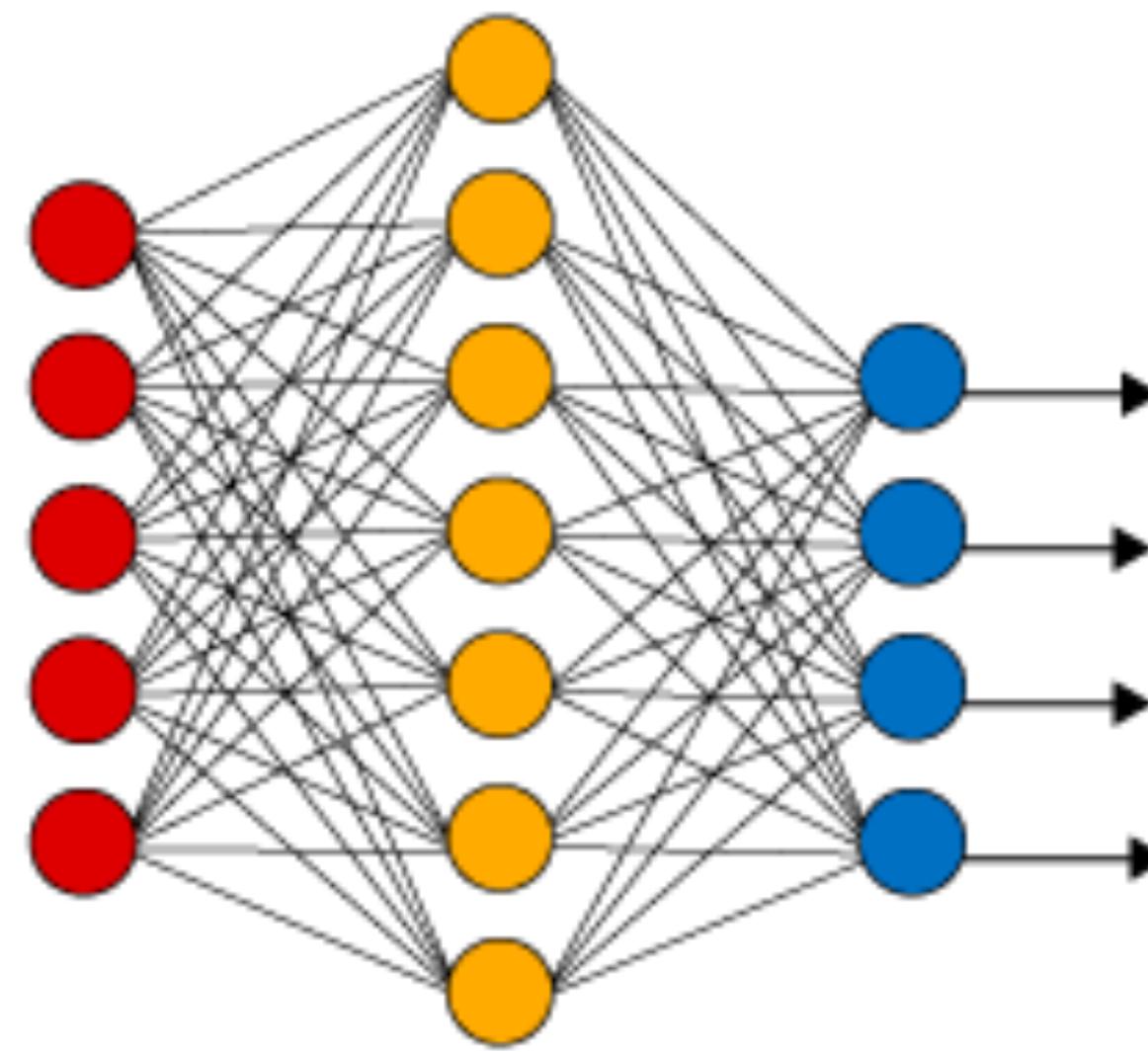
Convolutional Neural Networks

Sequence Modeling

Recurrent Neural Networks

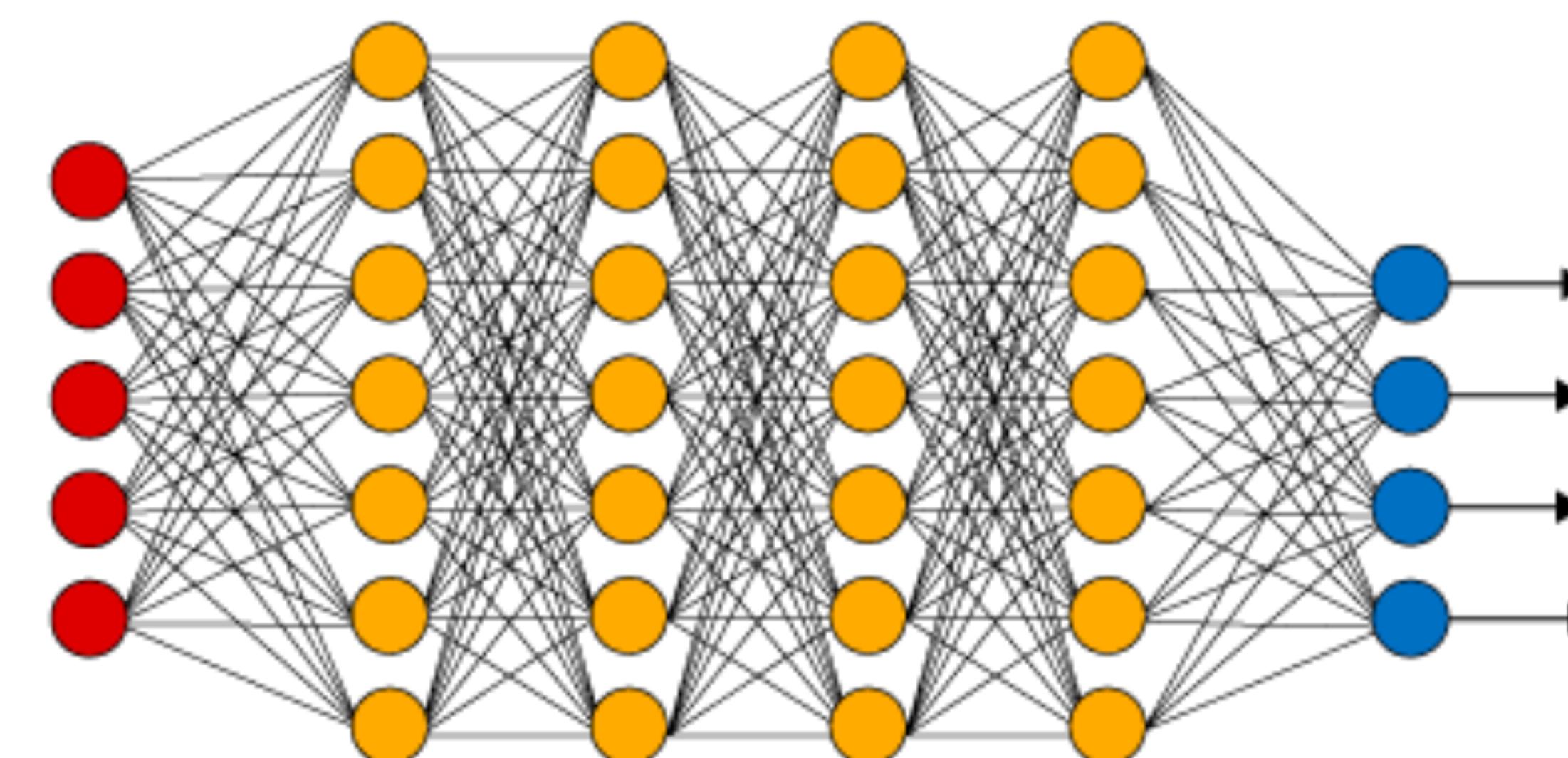
Fully Connected Neural Network

Simple Neural Network



● Input Layer

Deep Learning Neural Network



● Hidden Layer

● Output Layer

<https://miro.medium.com>

Computer Vision

¿Cómo observa un computador?

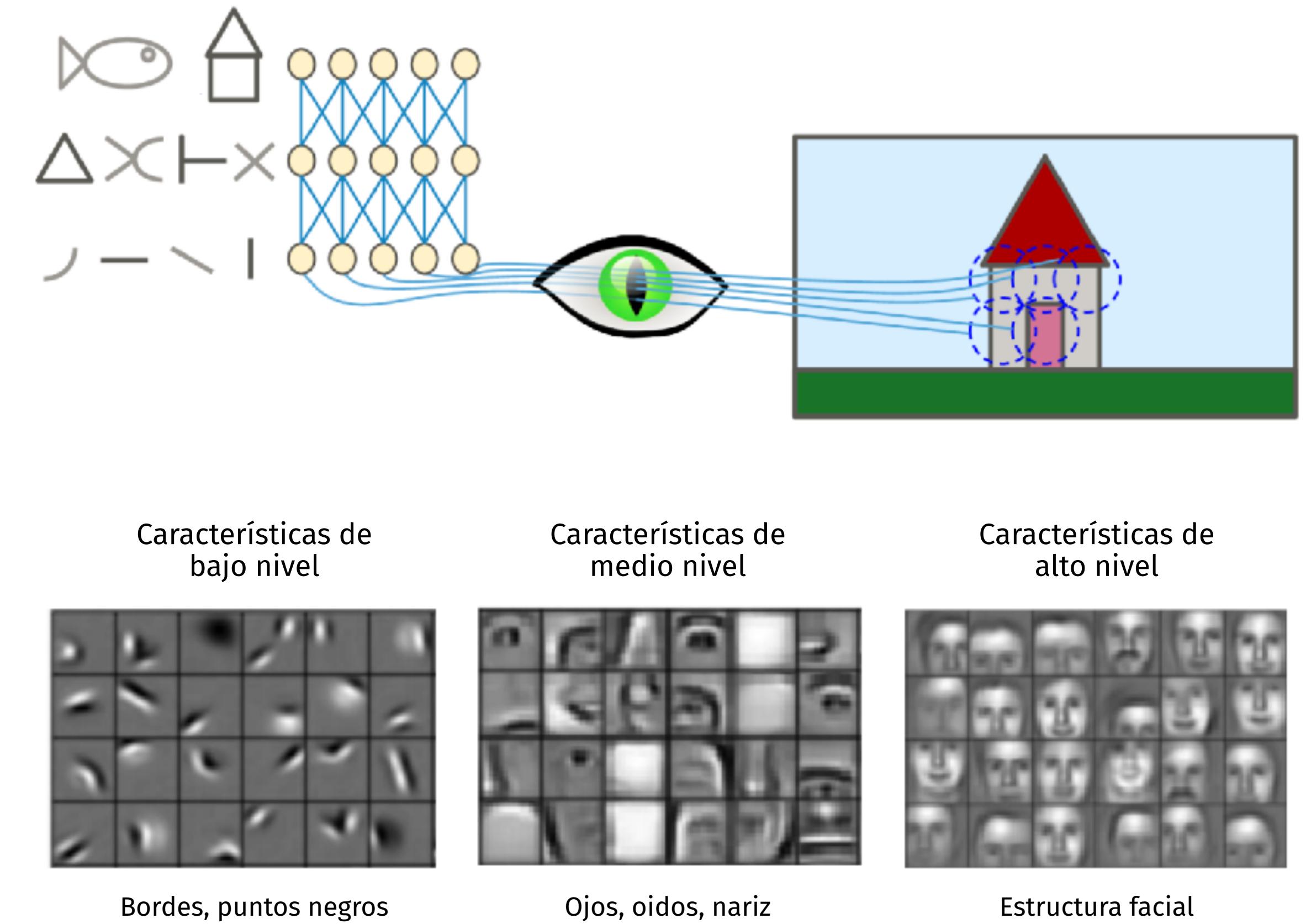


<https://larevueia.fr>

Arquitectura de Redes Neuronales

Computer Vision

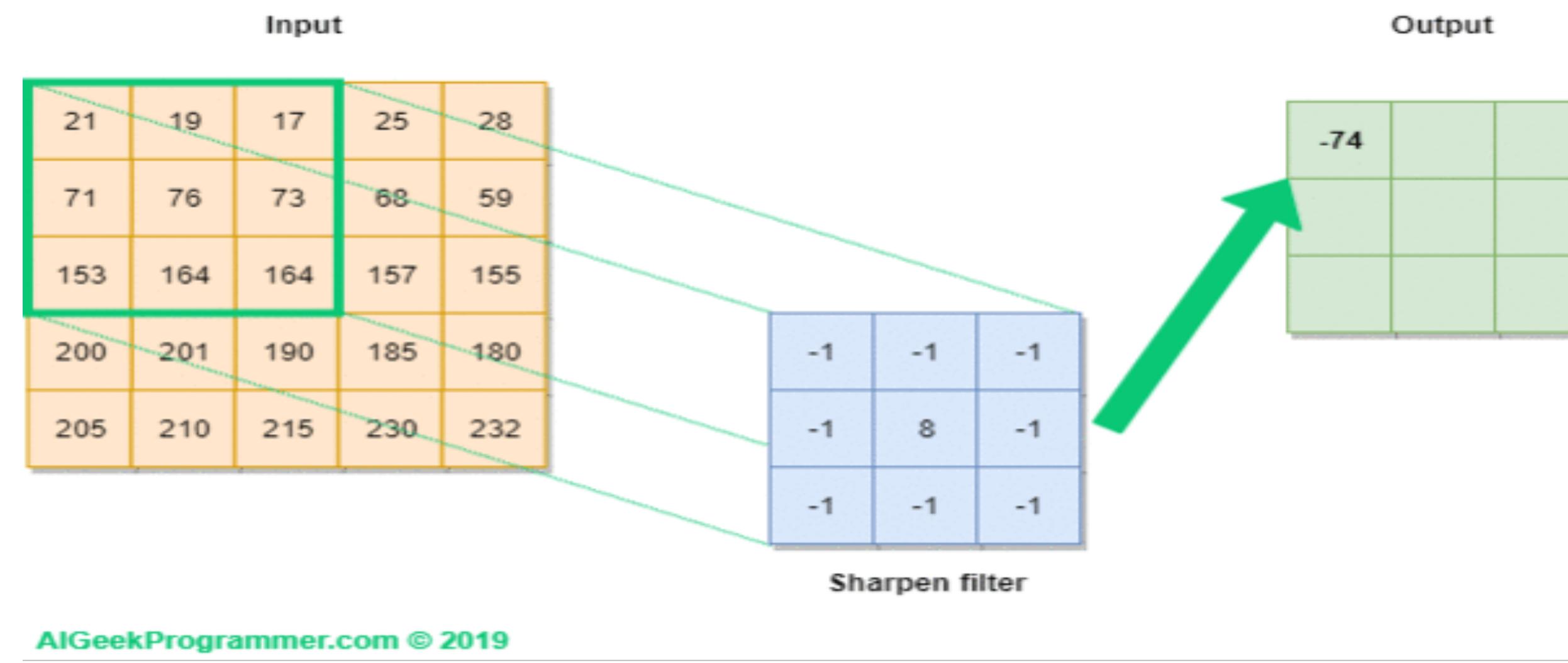
David H. Hubel y Torsten Wiesel (1959): Muchas neuronas tienen un pequeño **campo receptivo local**, reaccionan sólo a los estímulos visuales situados en una región limitada del campo visual



Convolutional Neural Networks

Convolución

Mapa de activación
Mapa de características



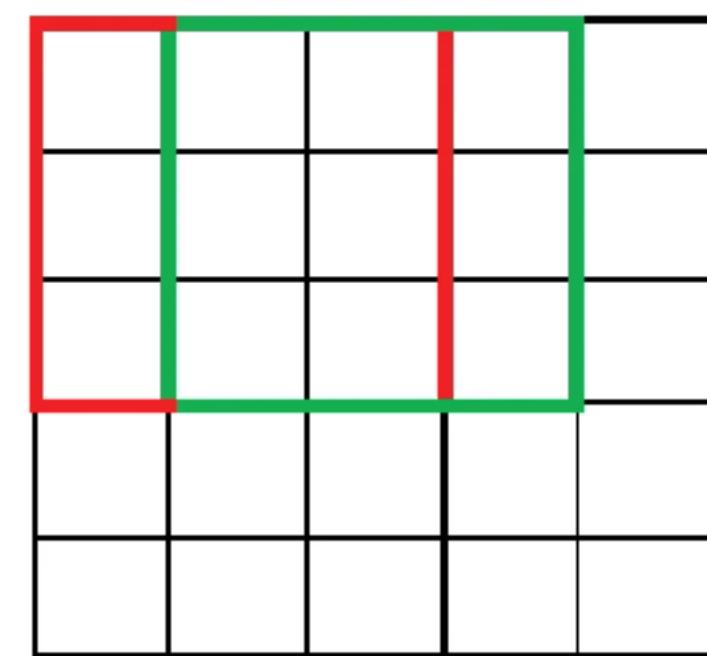
<https://aigeekprogrammer.com>

Arquitectura de Redes Neuronales

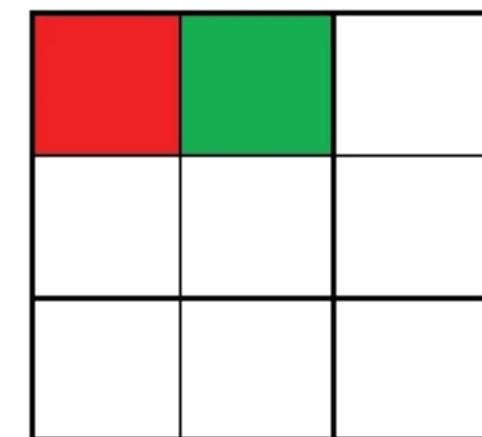
Convolutional Neural Networks

Stride y Padding

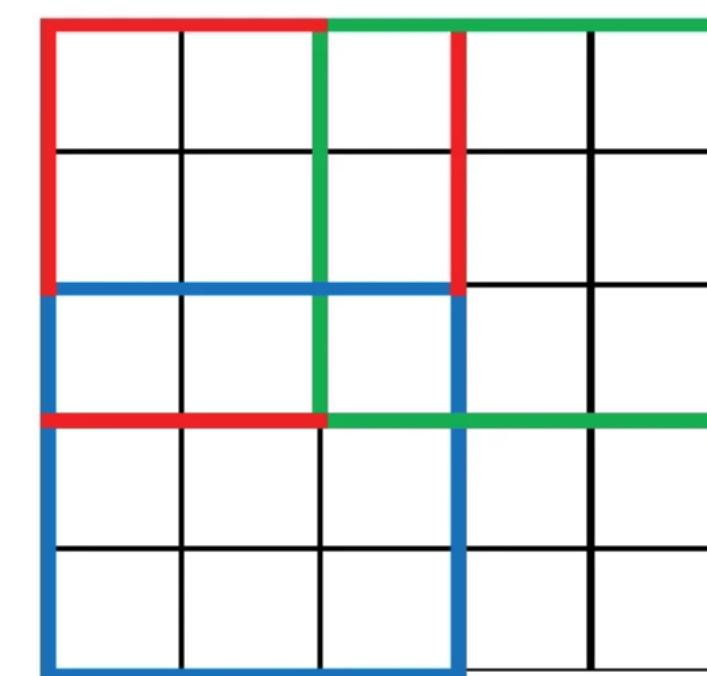
Stride = 1



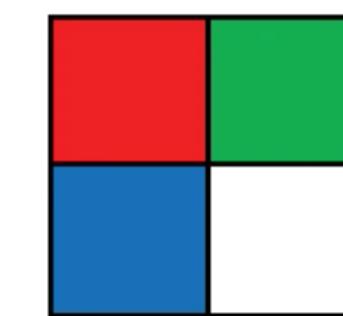
Salida



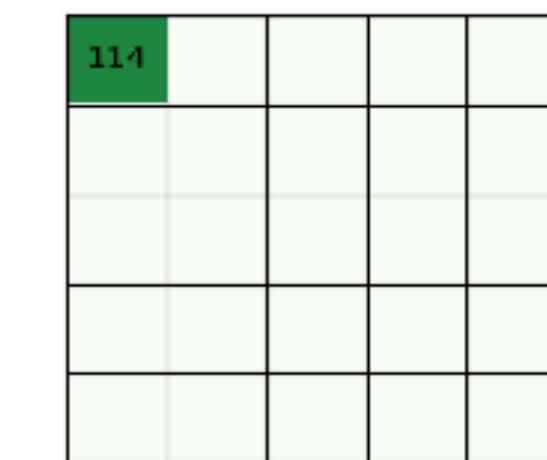
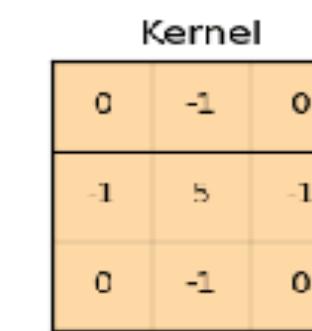
Stride = 2



Salida



0	0	0	0	0	0	0	0
0	60	113	56	139	85	0	0
0	73	121	54	84	128	0	0
0	131	99	70	129	127	0	0
0	80	57	115	69	134	0	0
0	104	126	123	95	130	0	0
0	0	0	0	0	0	0	0



$$n_{out} = \left(\frac{n_{in} + 2p - k}{s} \right) + 1$$

n_{in} = Tamaño de entrada

p = Tamaño del padding

k = Tamaño del filtro de la convolución

s = Tamaño del stride

n_{out} = Tamaño del mapa de características

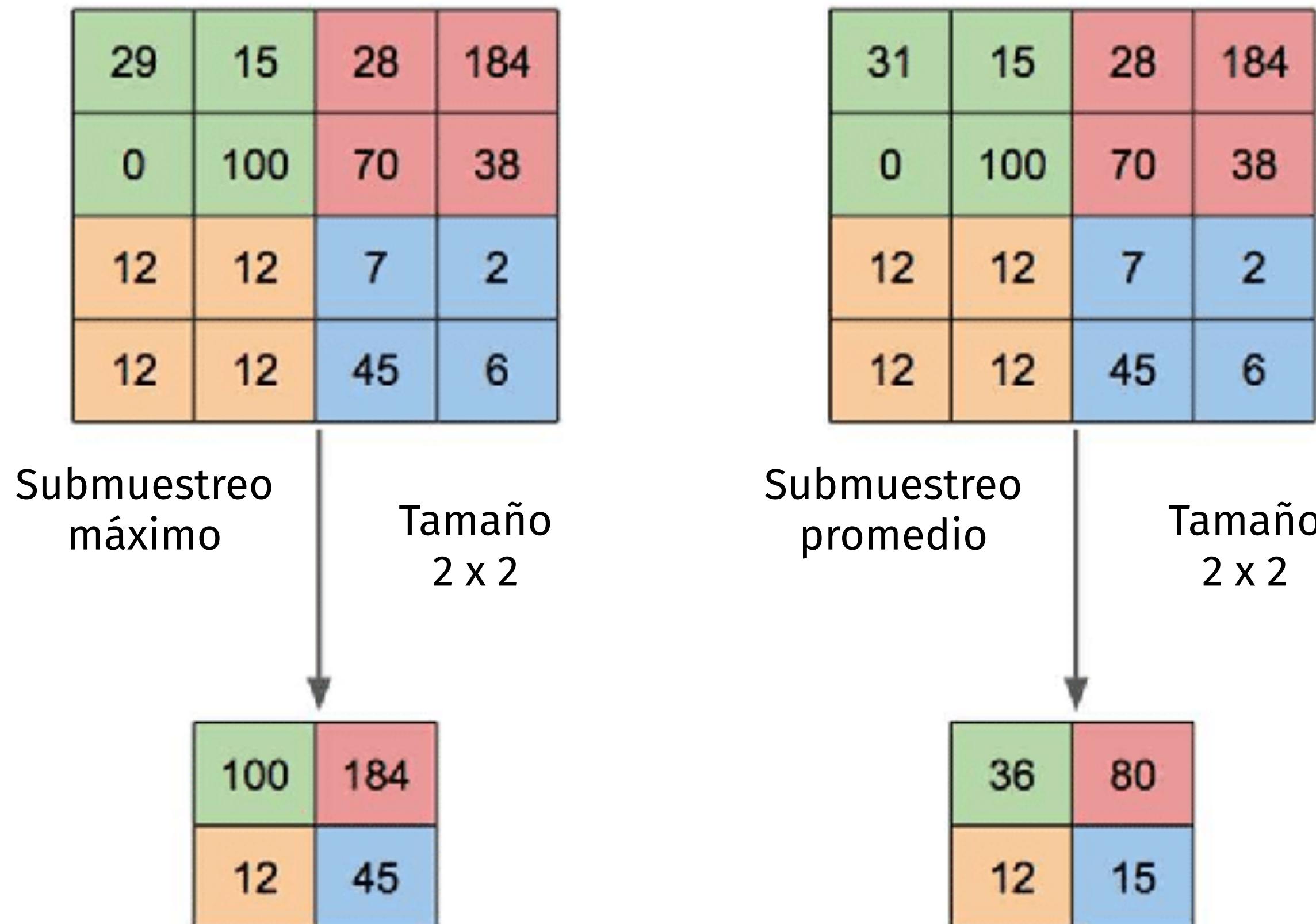
<https://www.analyticsvidhya.com/blog/2022/03/basics-of-cnn-in-deep-learning/>

<https://setosa.io/ev/image-kernels/>

<https://medium.com/@draj0718/zero-padding-in-convolutional-neural-networks-bf1410438e99>

Convolutional Neural Networks

Pooling



$$M_p = \frac{M}{p}$$

M = Tamaño del mapa original

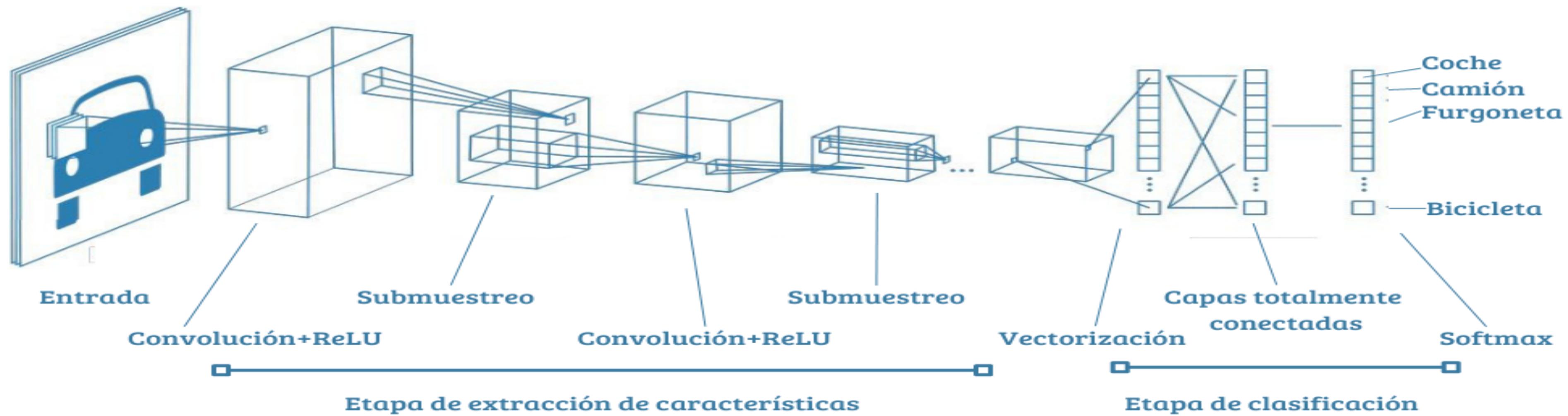
p = Tamaño del submuestreo

M_p = Tamaño del mapa después de aplicar el submuestreo

<https://www.linkedin.com/pulse/max-pooling-combining-channels-using-11-convolutions-field-n-bhatt/>
<http://dx.doi.org/10.1088/1742-6596/1201/1/012052>

Convolutional Neural Networks

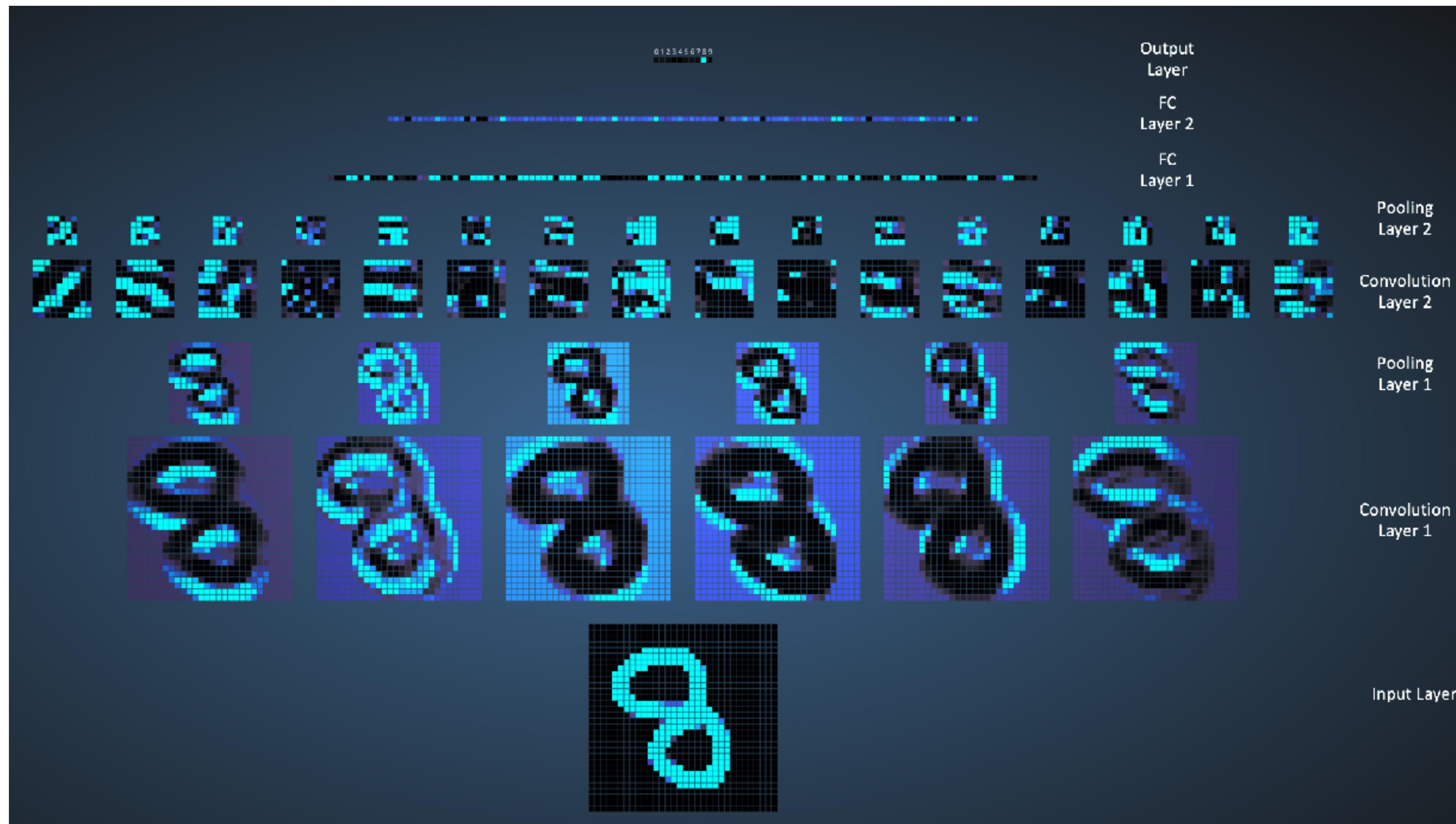
Simulación de una CNN



<https://www.mathworks.com/discovery/convolutional-neural-network-matlab.html>

Arquitectura de Redes Neuronales

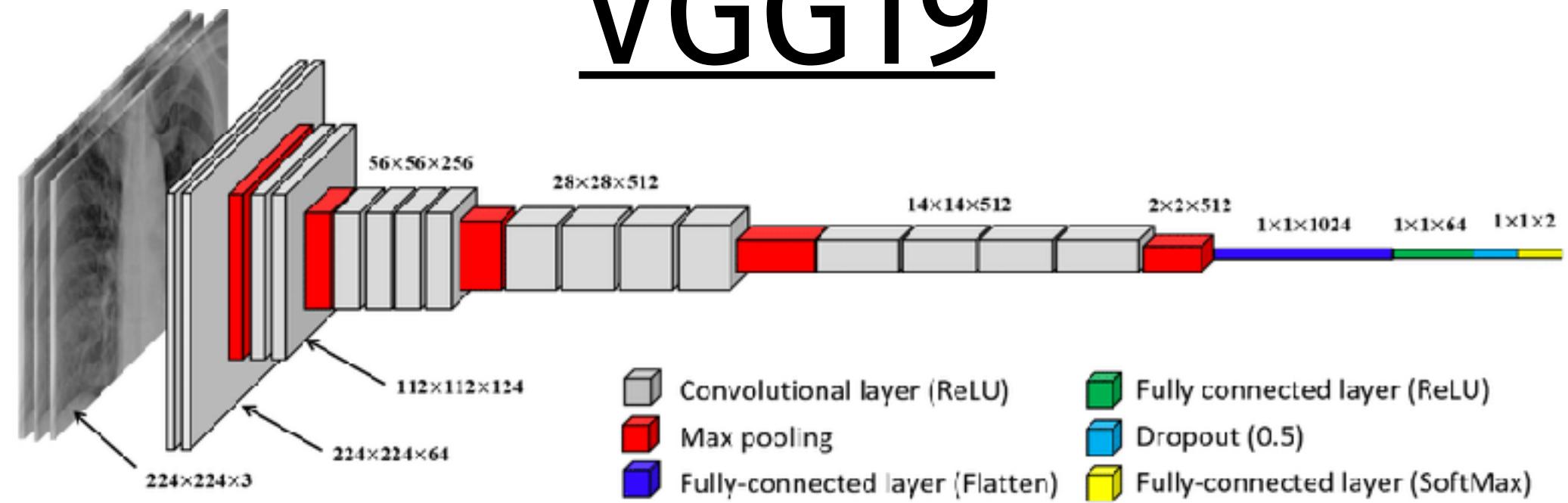
Convolutional Neural Networks



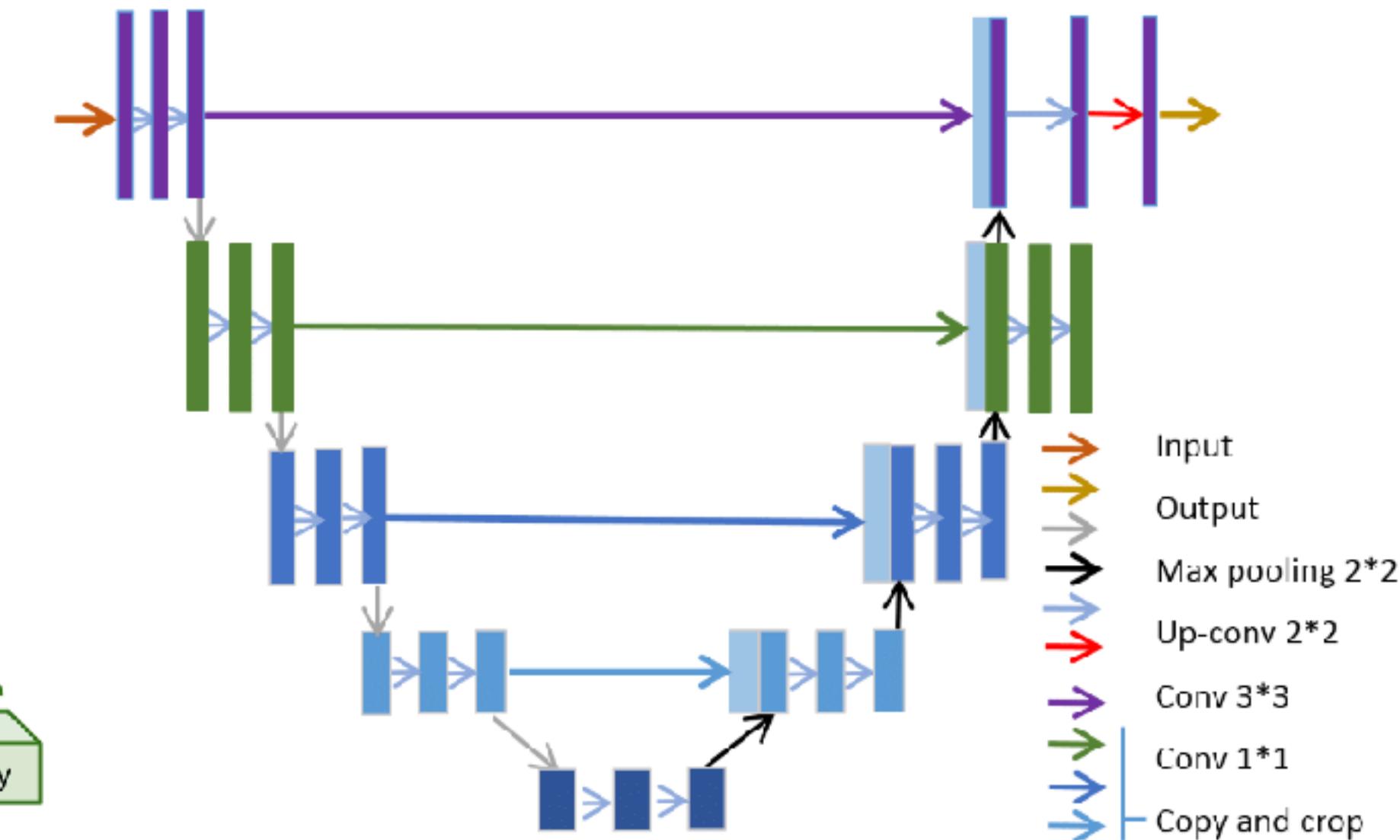
<https://www.kdnuggets.com/2016/11/intuitive-explanation-convolutional-neural-networks.html/3>

Convolutional Neural Networks

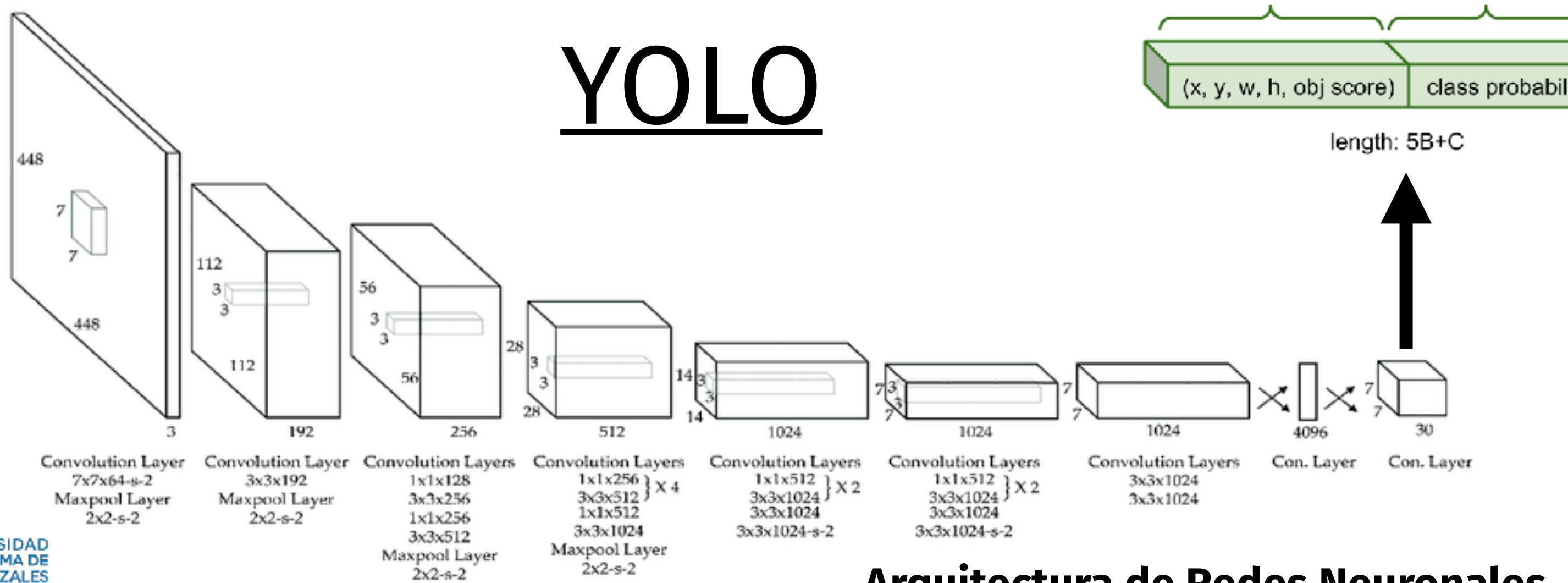
VGG19



U-Net



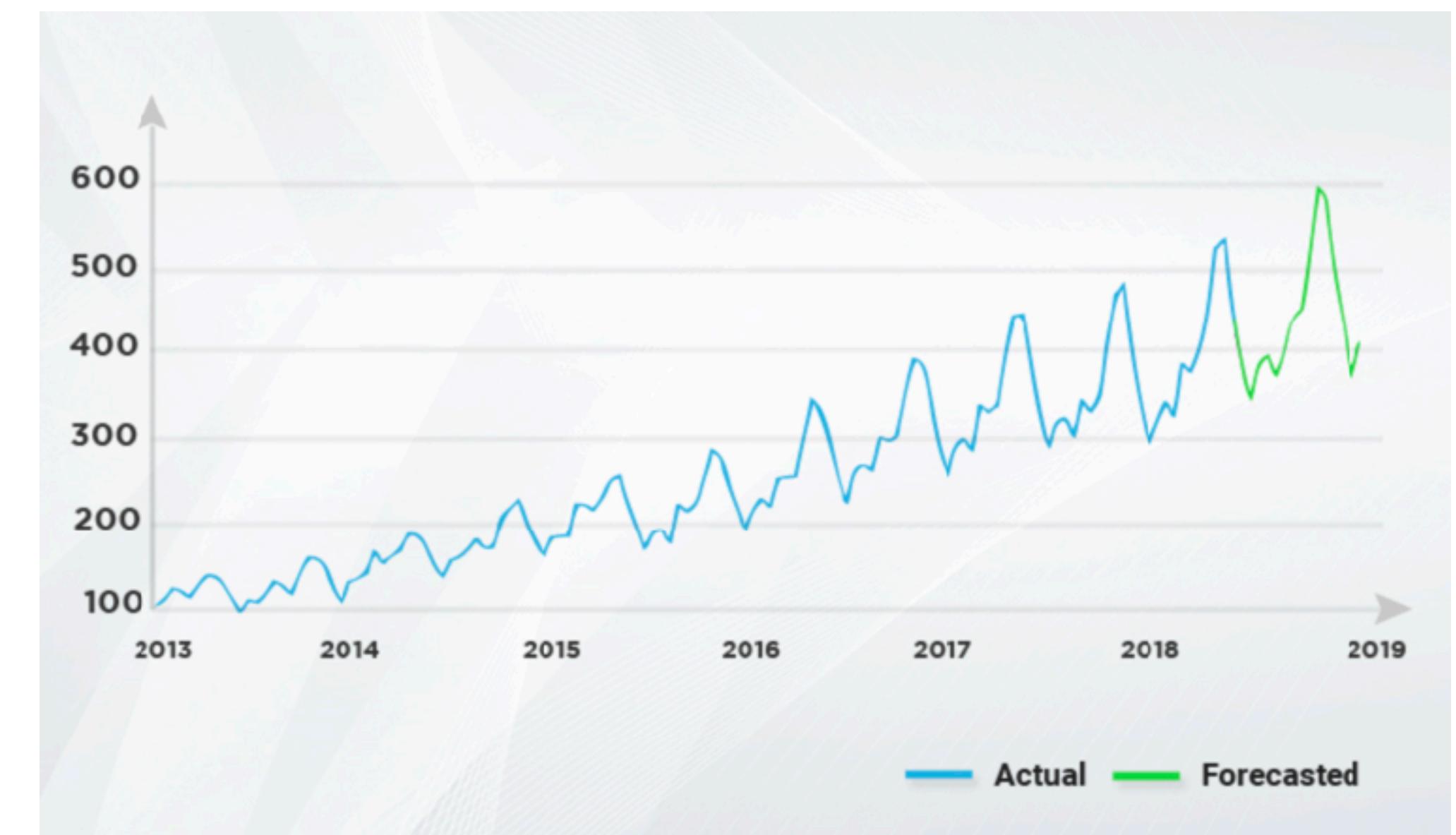
YOLO



<https://www.analyticsvidhya.com>

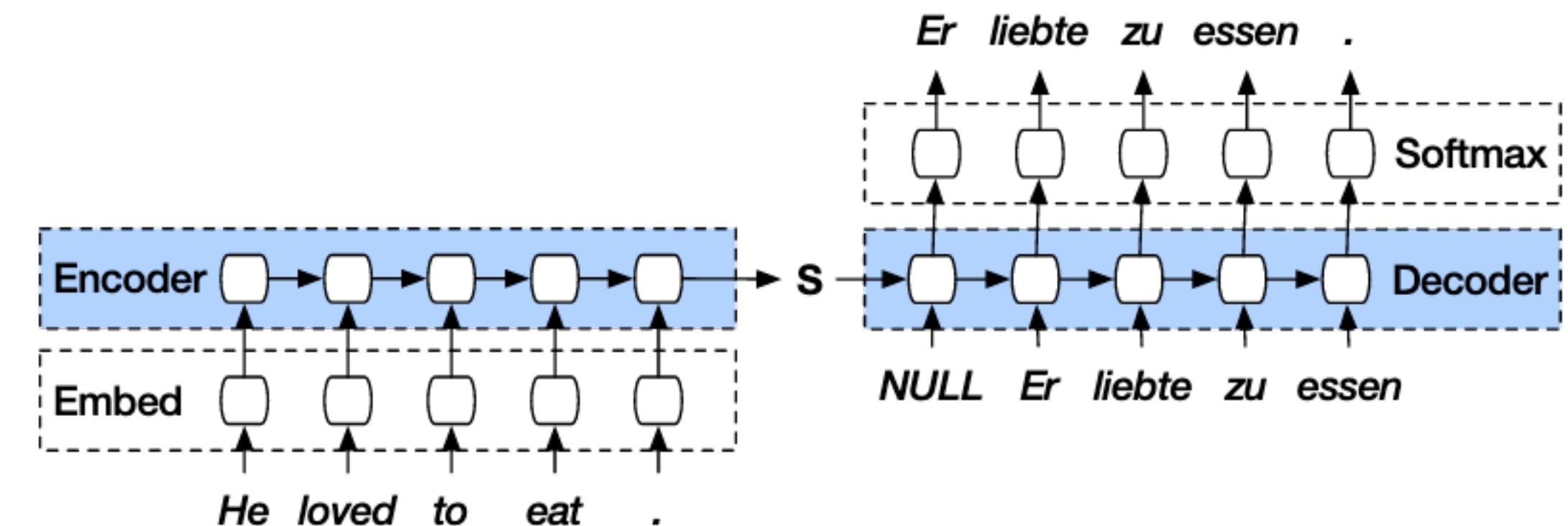
Sequence Modeling

Se refiere al procesamiento de datos secuenciales como las series de tiempo u oraciones y textos.



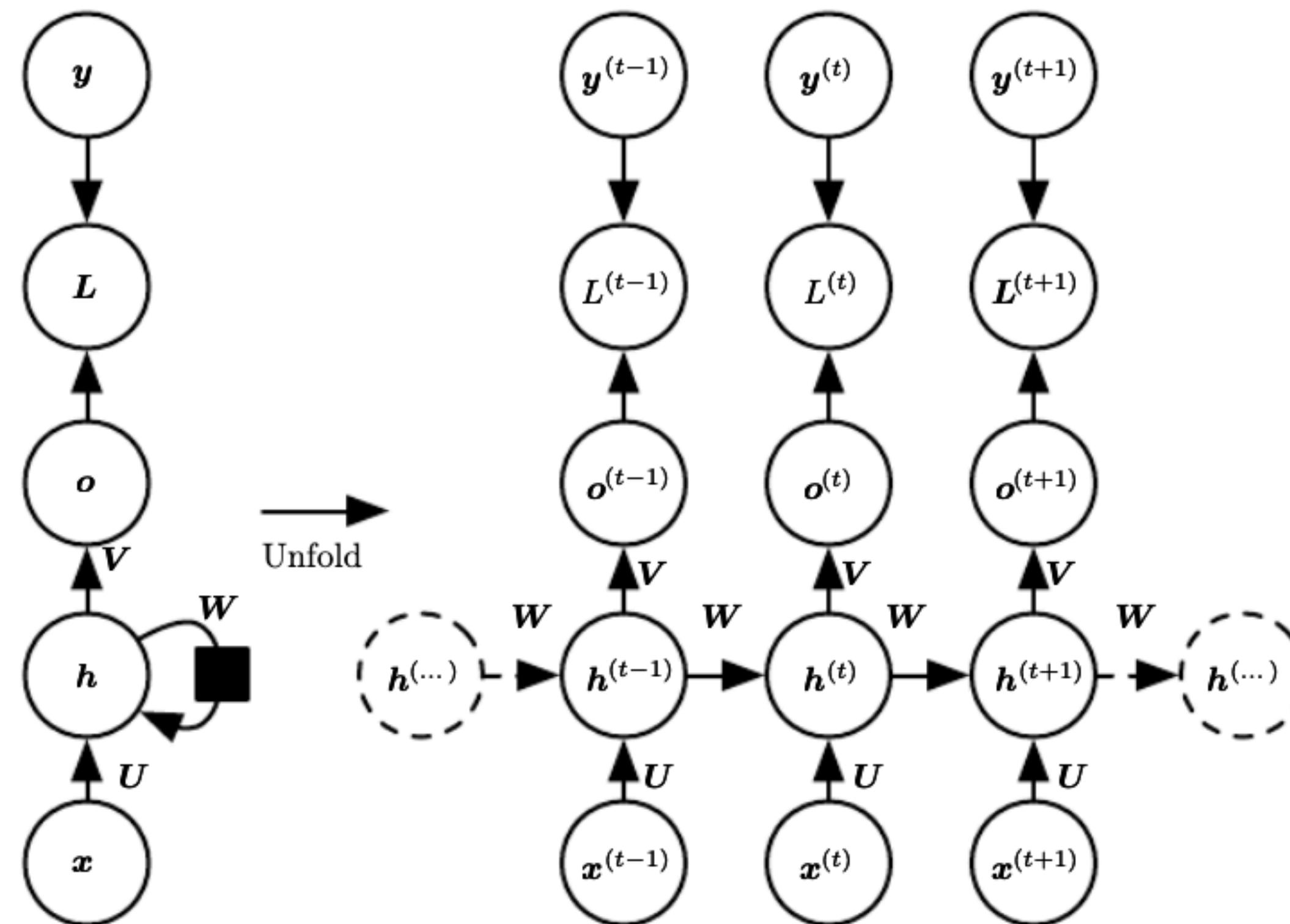
<https://www.springboard.com>

La característica principal de los datos secuenciales es que las muestras poseen algún tipo de dependencia entre ellas.



<https://smerity.com>

Recurrent Neural Networks

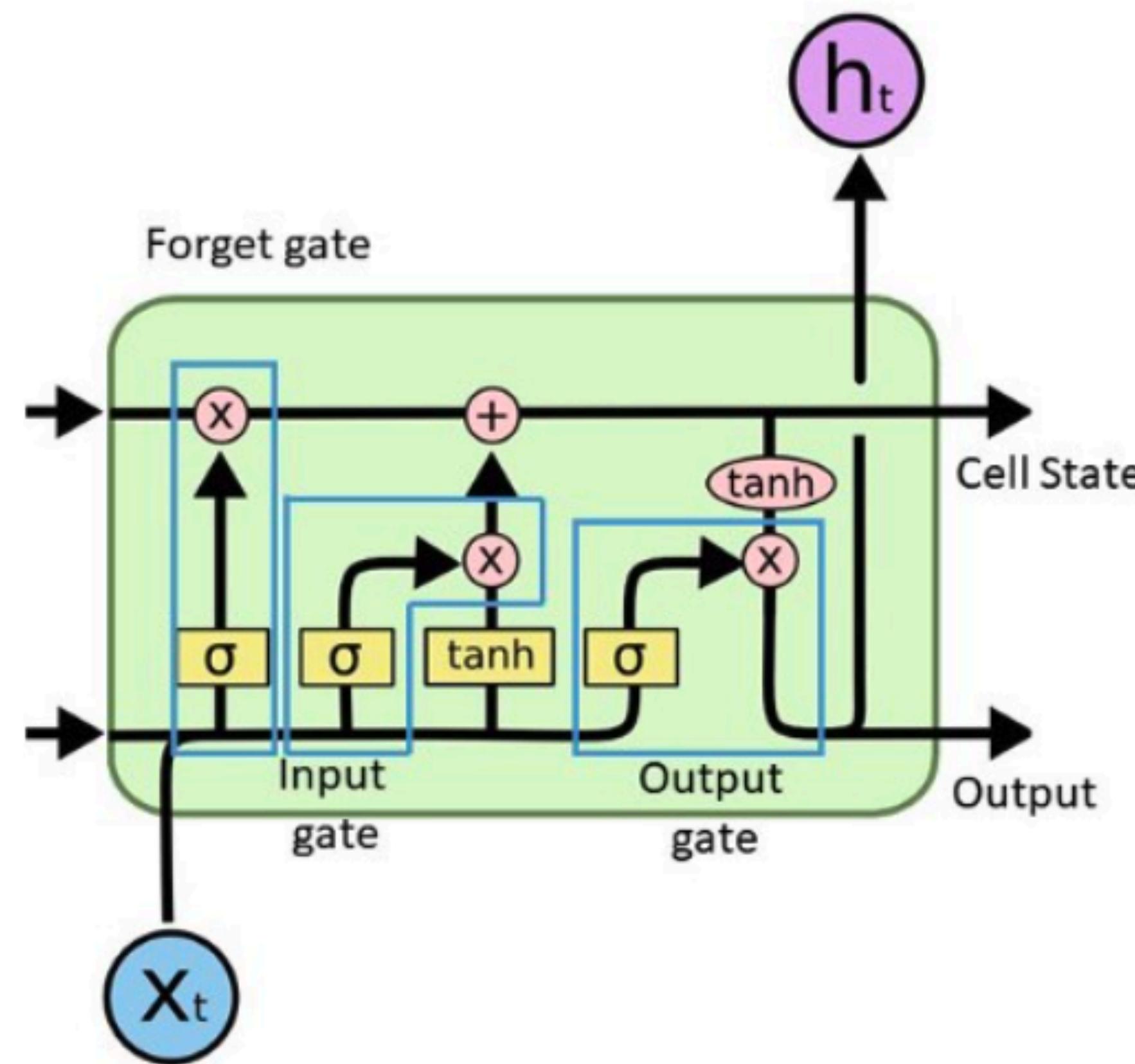


[Deep Learning by Ian Goodfellow,
Yoshua Bengio and Aaron Courville](#)

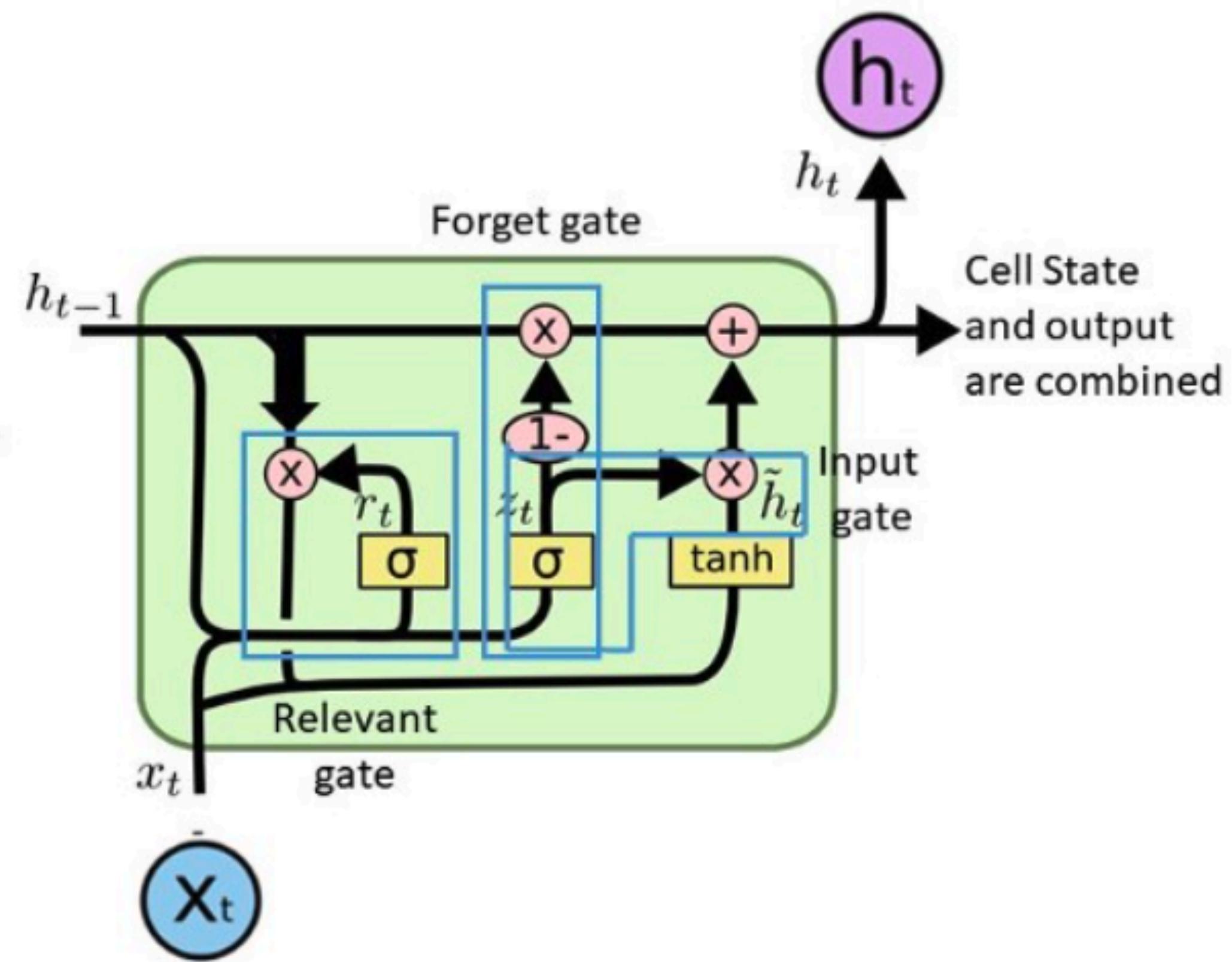
$$\begin{aligned} \mathbf{a}^{(t)} &= \mathbf{b} + \mathbf{W}\mathbf{h}^{(t-1)} + \mathbf{U}\mathbf{x}^{(t)}, \\ \mathbf{h}^{(t)} &= \tanh(\mathbf{a}^{(t)}), \\ \mathbf{o}^{(t)} &= \mathbf{c} + \mathbf{V}\mathbf{h}^{(t)}, \\ \hat{\mathbf{y}}^{(t)} &= \text{softmax}(\mathbf{o}^{(t)}), \end{aligned}$$

Recurrent Neural Networks

LSTM



GRU



<https://deeplearningmath.org>

Recurrent Neural Networks

Backpropagation through time

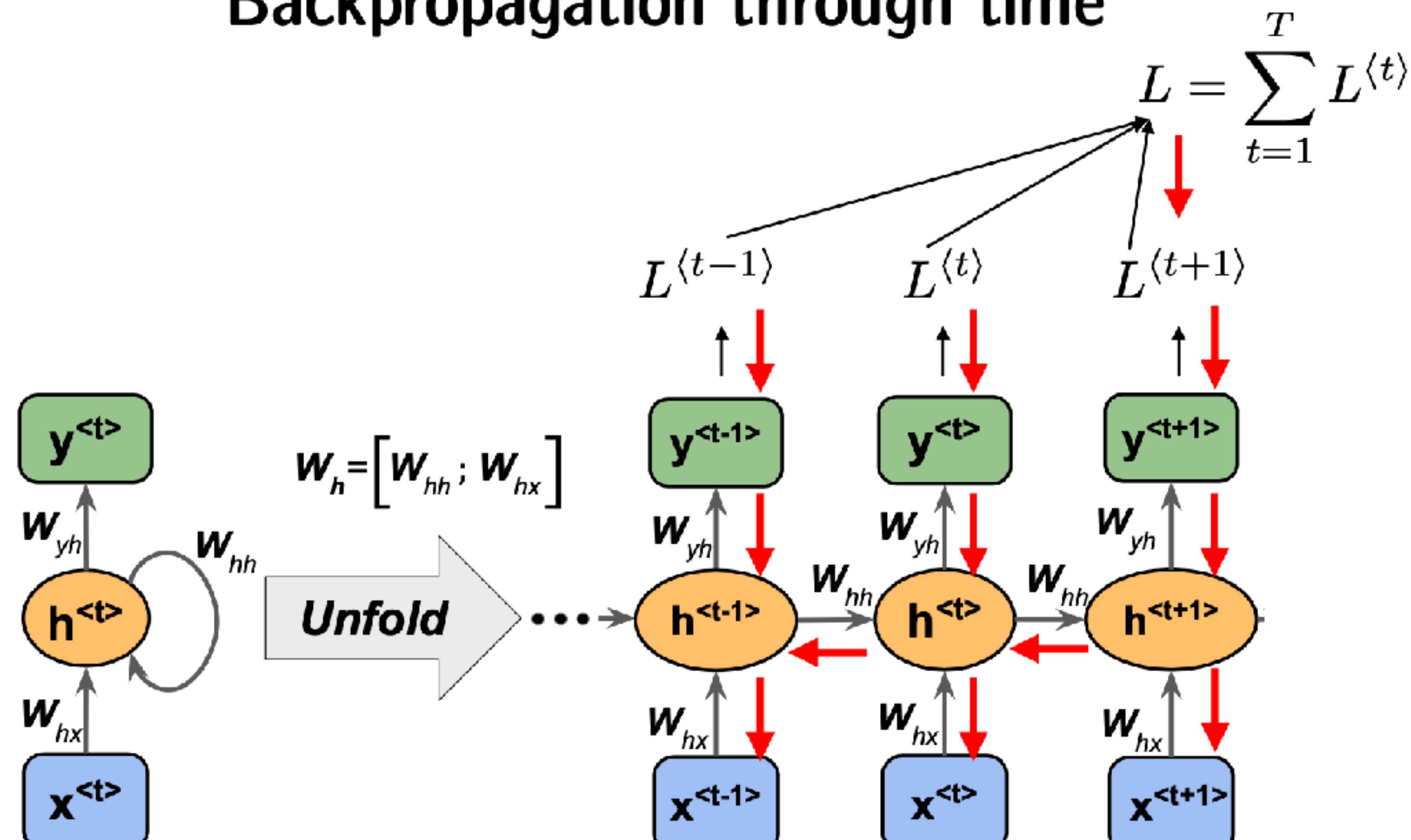


Figure: Sebastian Raschka, Vahid Mirjalili. Python
Machine Learning. 3rd Edition. Birmingham, UK: Packt
Publishing. 2019

<https://deeplearningmath.org>

Recurrent Neural Networks

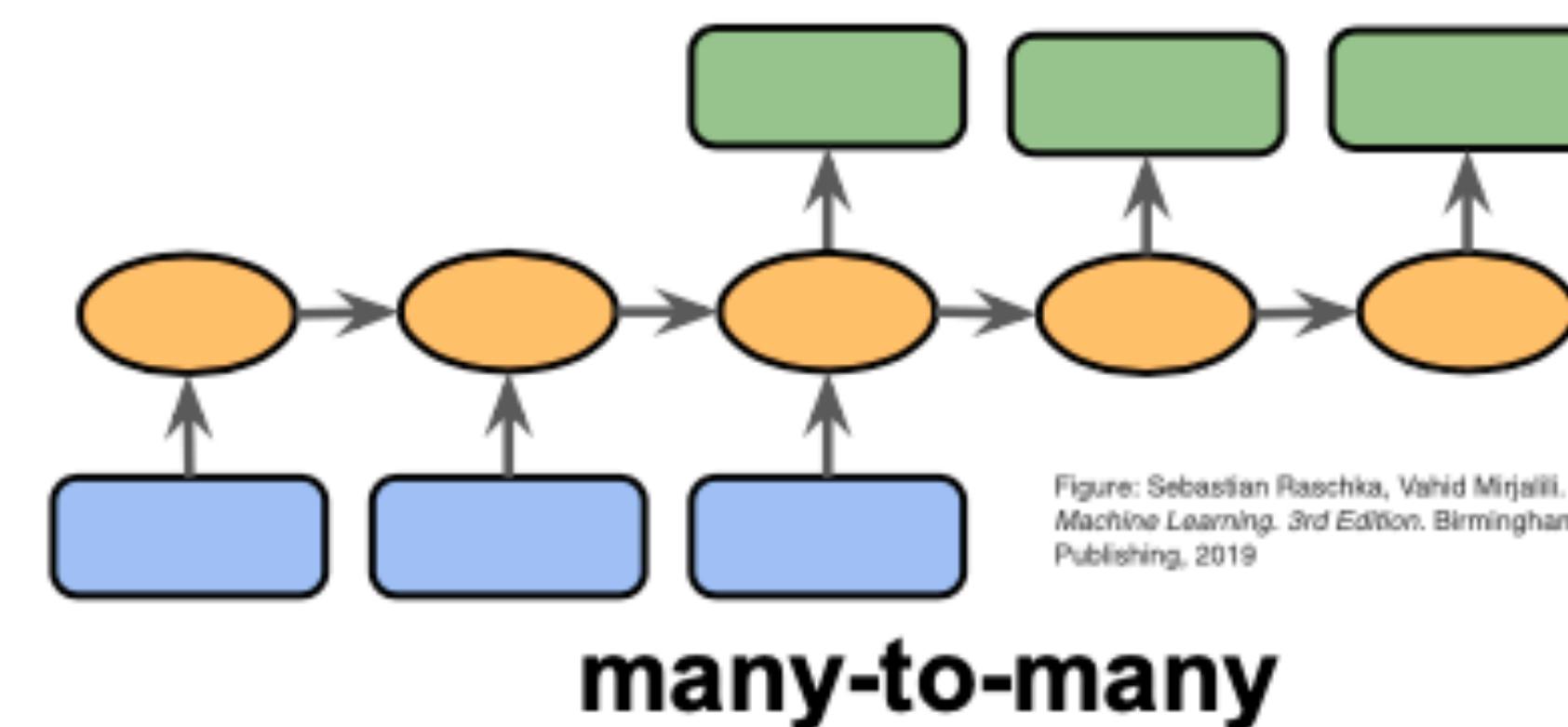
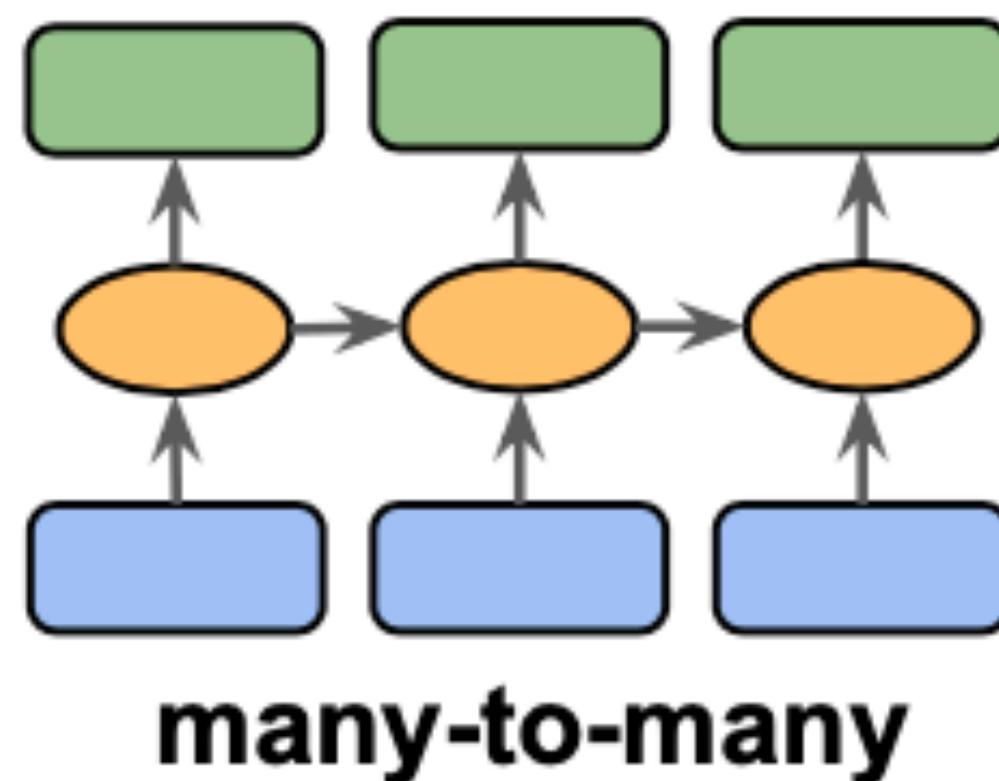
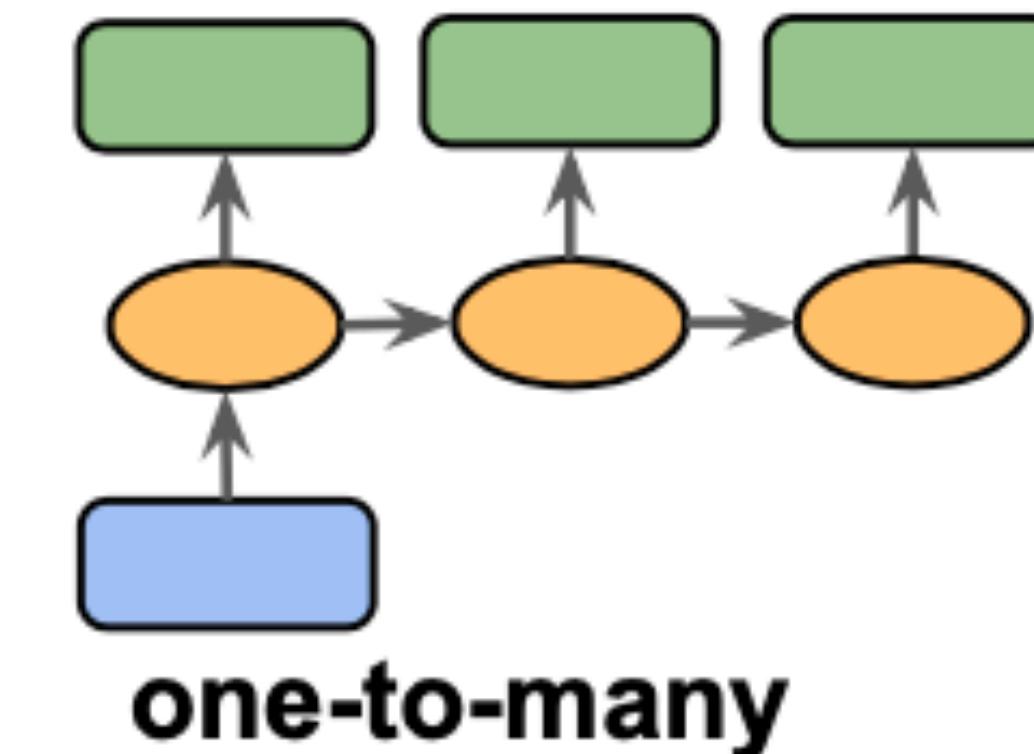
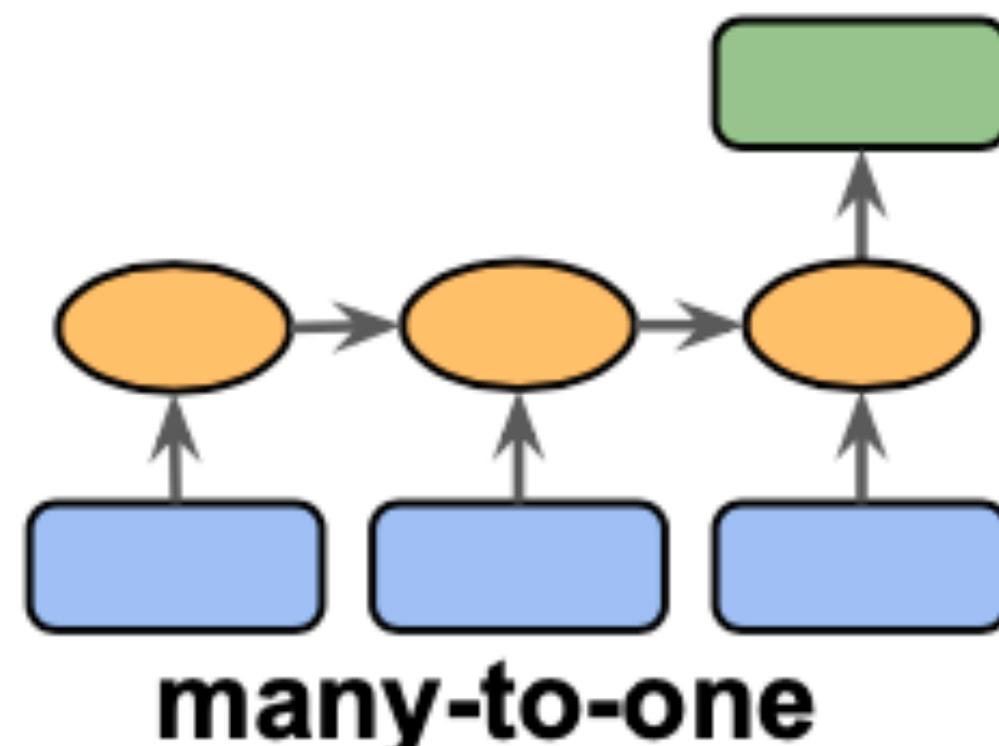


Figure: Sebastian Raschka, Vahid Mirjalili. Python Machine Learning. 3rd Edition. Birmingham, UK: Packt Publishing, 2019

<https://deeplearningmath.org>