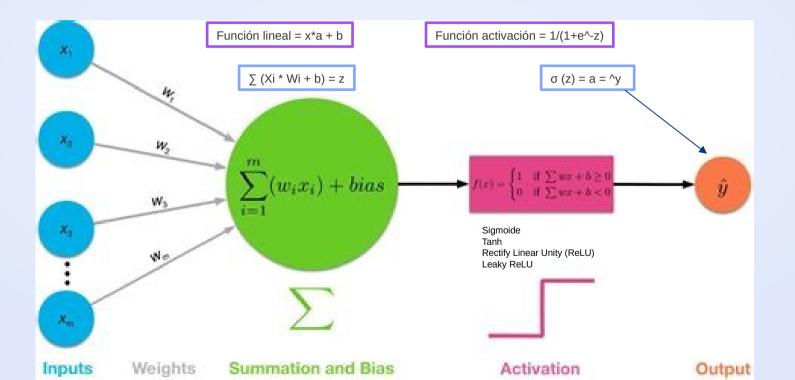




Perceptron



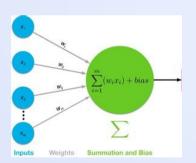




Entropía cruzada

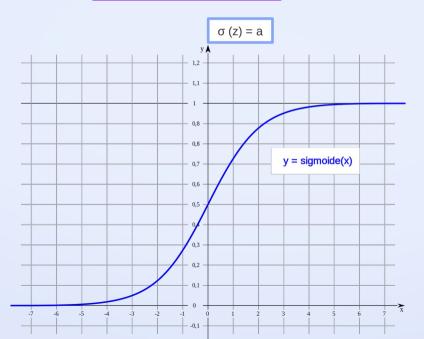
Función lineal = x*m + b

 \sum (Xi * Wi + b) = z



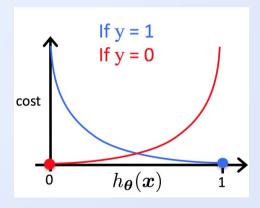
Sigmoide

Función activación = 1/(1+e^-z)



Función perdida = L

Función costo = J

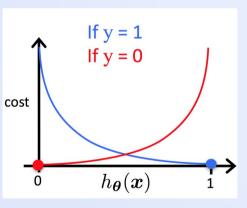






Función costo = J

Función perdida = L

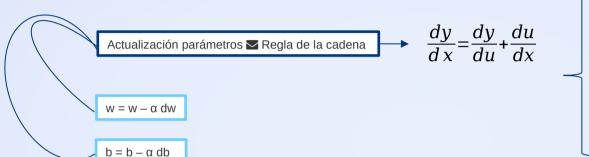


- 1. Caso y = 1: $L(^y, y) = -\log ^y$ Maximizar y para $L(^y, y)$ minimice.
- 2. Caso y = 0: L(^y, y) = log (1-^y)

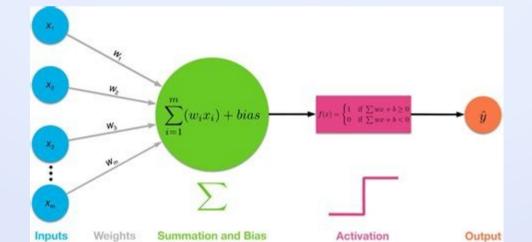
 Maximizar (1-^y) para L (^y,y) minimice, de tal forma que ^y tiene que minimizar.



Redes Neuronales BACKPROPAGATION_



 $dz = \frac{dL(a,y)}{dz} = -\frac{y}{a} + \frac{(1-y)}{(1-a)}$





Clasificador Binario Algoritmo de logistic regression



Hay gato $\searrow y = 1$ No hay gato $\searrow y = 0$





[[0.87467339, 0.11874317, 0.78340092, 0.33238932], [0.97872965, 0.42424313, 0.87790352, 0.86745669], [0.42382851, 0.7793706, 0.95557697, 0.36313912], [0.85210037, 0.85438949, 0.1558208, 0.72611312]]

([[0.03906905, 0.97440963, 0.71961898, 0.32996325], [0.69244103, 0.59412868, 0.71591308, 0.81801062], [0.87850438, 0.20520954, 0.09388833, 0.6499204], [0.41805232, 0.97656932, 0.1016103, 0.49395151]])

([[0.69477392, 0.20554923, 0.17940824, 0.47703951], [0.8679659, 0.01558215, 0.3911708, 0.67345343], [0.34681257, 0.5883467, 0.56270387, 0.01221303], [0.20167319, 0.20900837, 0.25177722, 0.42471632]])

Clasificador Binario Redes Neuronales Convolucionales



Hay gato $\searrow y = 1$ No hay gato $\searrow y = 0$

0	1	1	Ĩ.,	0.	0,	0										
0	0	1	1 _{×0}	1,1	Q	0		••••	>:	Section 1		:1::	4	3	4	1
0	0	0	1,	1 _{×0}	1 _{×1}	0		1	0	1		1	.2	4.	3	3
0	0	0	1	··1	.0	0	*****	0	1	0		1	$\dot{2}$	3	4	1
0	0	1	1	0	0	0		1	0	1		1	3	3	1	1
0	1	1	0	0	0	0	Filtro utilizado						3	1	1	0
1	1	0	0	0	0	0	Filti O utilizado									
	Ima	igen 7	7x7				Kernel 3x3					Imagen filtrada 5x5				

 $\begin{array}{l} \hbox{\tt [[0.87467339,\,0.11874317,\,0.78340092,\,0.33238932],} \\ \hbox{\tt [0.97872965,\,0.42424313,\,0.87790352,\,0.86745669],} \\ \hbox{\tt [0.42382851,\,0.7793706\,,\,0.95557697,\,0.36313912],} \\ \hbox{\tt [0.85210037,\,0.85438949,\,0.1558208\,,\,0.72611312]]} \end{array}$



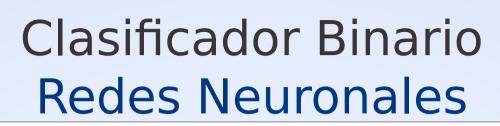
2],],]])

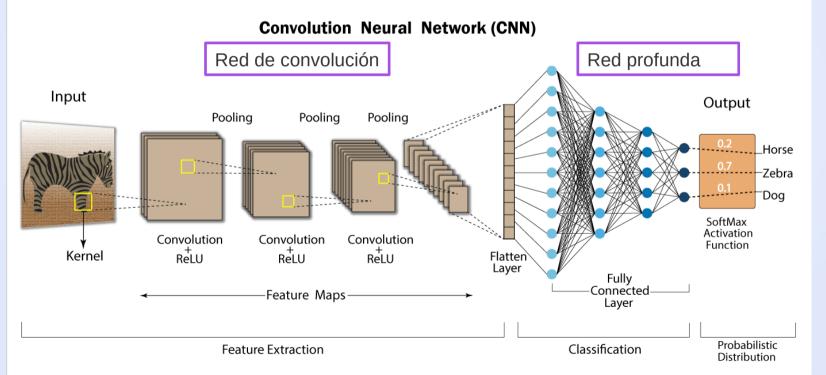
([[0.03906905, 0.97440963, 0.71961898, 0.32996325], [0.69244103, 0.59412868, 0.71591308, 0.81801062], [0.87850438, 0.20520954, 0.09388833, 0.6499204], [0.41805232, 0.97656932, 0.1016103, 0.49395151]])

([[0.69477392, 0.20554923, 0.17940824, 0.47703951], [0.8679659, 0.01558215, 0.3911708, 0.67345343], [0.34681257, 0.5883467, 0.56270387, 0.01221303], [0.20167319, 0.20900837, 0.25177722, 0.42471632]])

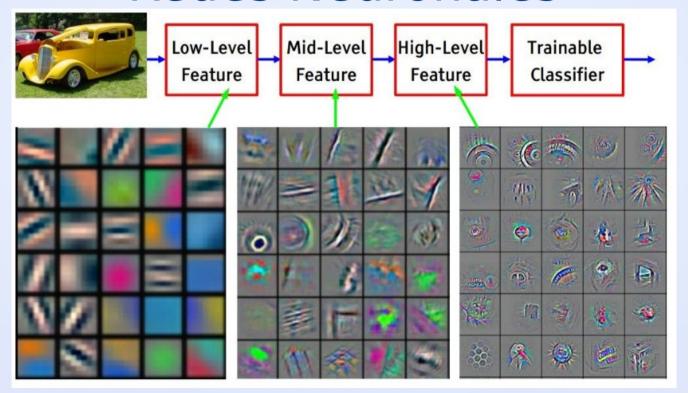
Z =







Clasificador Binario Redes Neuronales





En resumen

CNN

Red encargada de extraer las características de cada imagen y así mejor el entrenamiento



Red neuronal fully connected

Es la encargada de aprender los features de la salida de la cnn.

Funcion de activacion softmax

Dicha función nos entregara un porcentaje de confianza de la clase que detectó.

Fuentes

Cursos

- Especialización en Deep Learning Coursera (cursos 1-4)
- Tensorflow Developer Certificate 2021 Udemy

Libros

- Deep Learning with Python Francois Chollet
- Digital Image Processing Rafael C. Gonzalez
- Hands-On Machine Learning with Scikit Learn and TensorFlow. Concepts, Tools, and Techniques to Build Intelligent Systems - Aurélien Géron

Muchas Gracias

ignacio.bosch@um.edu.ar







CREDITS: This presentation template was created by <u>Slidesgo</u>, including icons by <u>Flaticon</u>, and infographics & images by <u>Freepik</u>

Please keep this slide for attribution

