

Deep Learning

Redes Neuronales con TensorFlow

Septiembre 13 – Septiembre 20

Convolucionales

Deep Learning

Redes Neuronales con TensorFlow

Septiembre 13 – Septiembre 20

Convolucionales

- Kernels
- Convolucionales
- Hiperparámetros
- Image Augmentation

import numpy as np a = np.arange(0, 9).reshape(3,3) b = np.arange(0, 9).reshape(3,3)

0	1	2
3	4	5
6	7	8

0	1	2
3	4	5
6	7	8

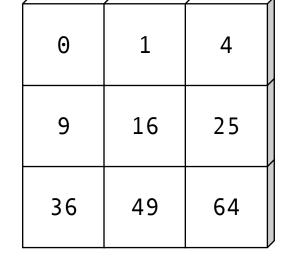


```
>> import numpy as np
>> a = np.arange(0, 9).reshape(3,3)
>> b = np.arange(0, 9).reshape(3,3)
>> a*b
```

Θ	1	2
3	4	5
6	7	8



0	1	2
3	4	5
6	7	8

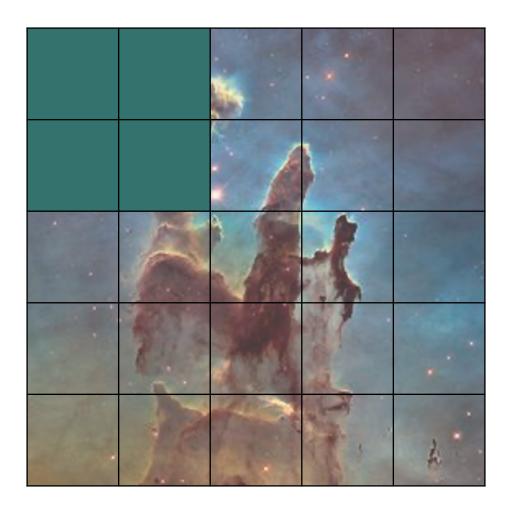




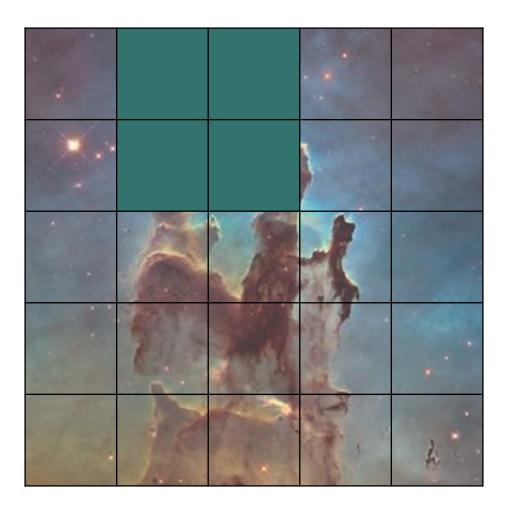
```
>> import numpy as np
>> a = np.arange(0, 9).reshape(3,3)
\Rightarrow b = np.arange(0, 9).reshape(3,3)
>> a*b
>> (a*b).sum()
```

204











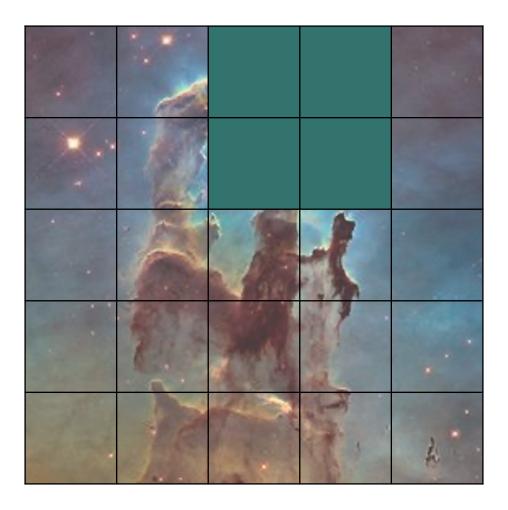






Imagen original



Imagen B/N



Kernels



Imagen original



Imagen B/N



Sharpen

Θ	-1	0
- 1	5	- 1
0	-1	0





Imagen original



Imagen B/N



Laplacian

0	1	0
1	4	1
0	1	0





Imagen original



Imagen B/N



Sobel X

-1	0	1
- 2	0	2
-1	0	1

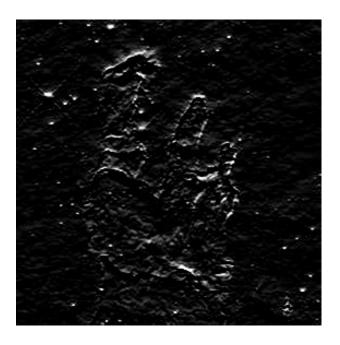




Imagen original



Imagen B/N



Sobel Y

-1	2	-1
0	•	0
1	2	1

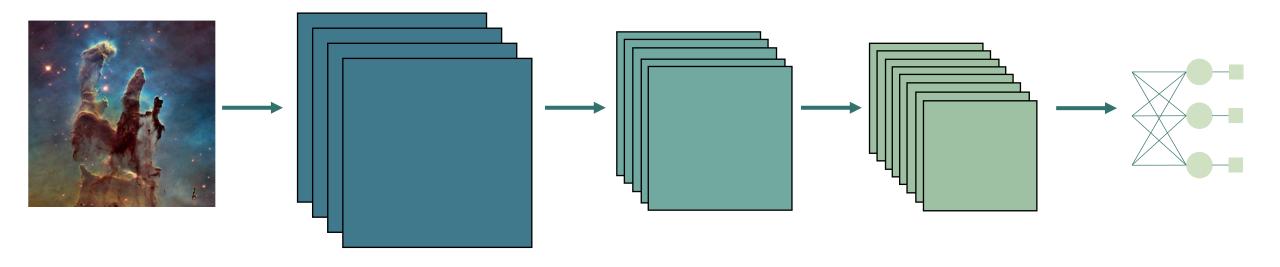




El objetivo de la red neuronal es establecer los valores del kernel

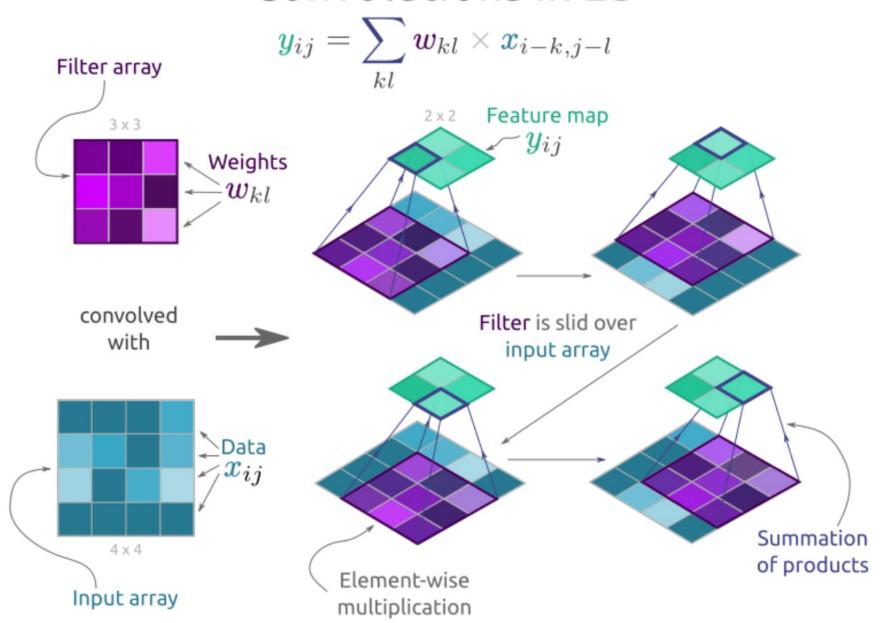


Convolucional





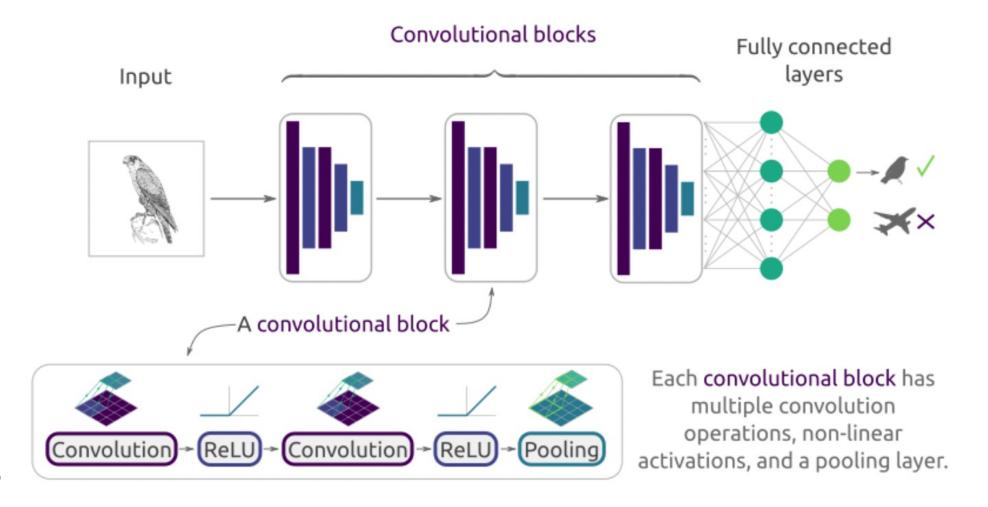
Convolutions in 2D





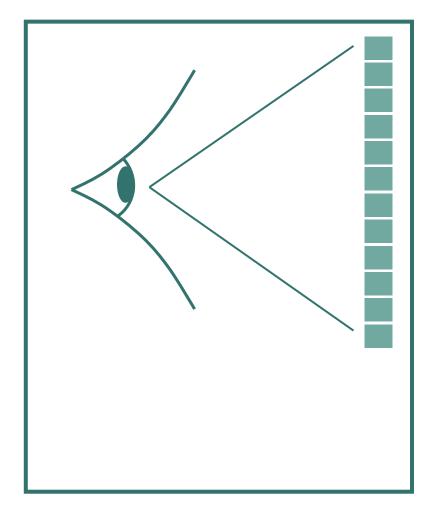
A Basic Convolutional Neural Network

A simplified convolutional neural network made of several convolutional blocks for feature extraction and fully connected layers for image classification.

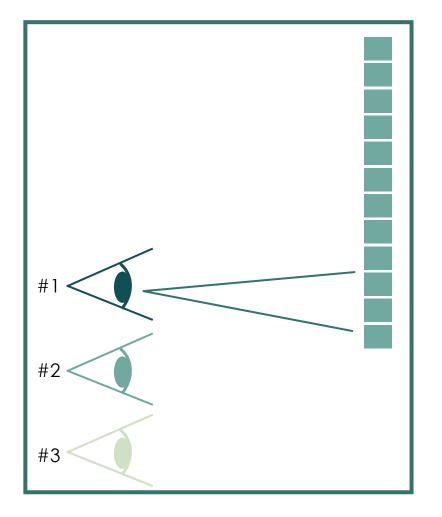




Feedforward Neural Network

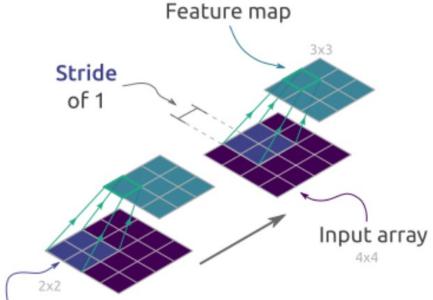


Convolutional Neural Network





Convolution Stride Length

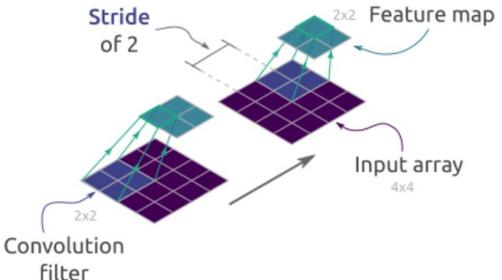


Stride length is how many places the convolution filter is moved between operations and affects the size of the resulting feature map

A larger stride results in a smaller feature map array. Here a stride of 1 results in a 3x3 feature map, while a stride of 2 results in a 2x2 feature map.

Convolution

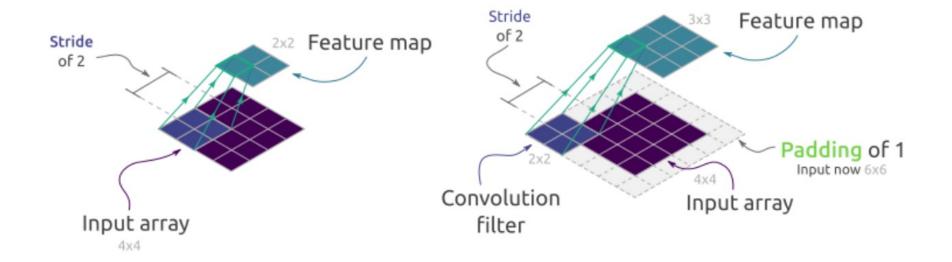
filter





Convolution Padding

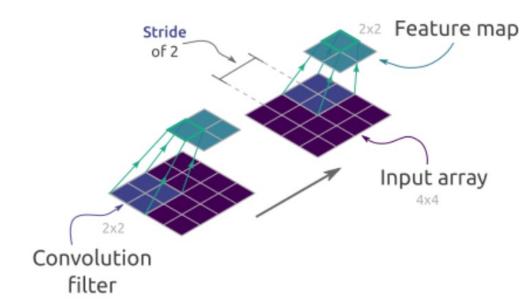
Padding is extra data added to the input array to make the array larger in order to control the size of the resulting feature map



In this example adding a padding of one entry around the input array increases the feature map size from 2x2 to 3x3

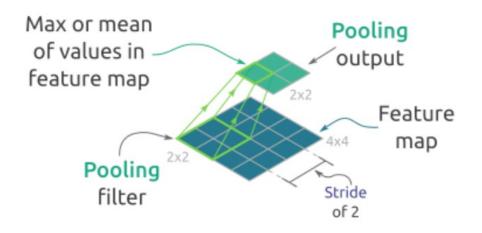


Convolution Pooling

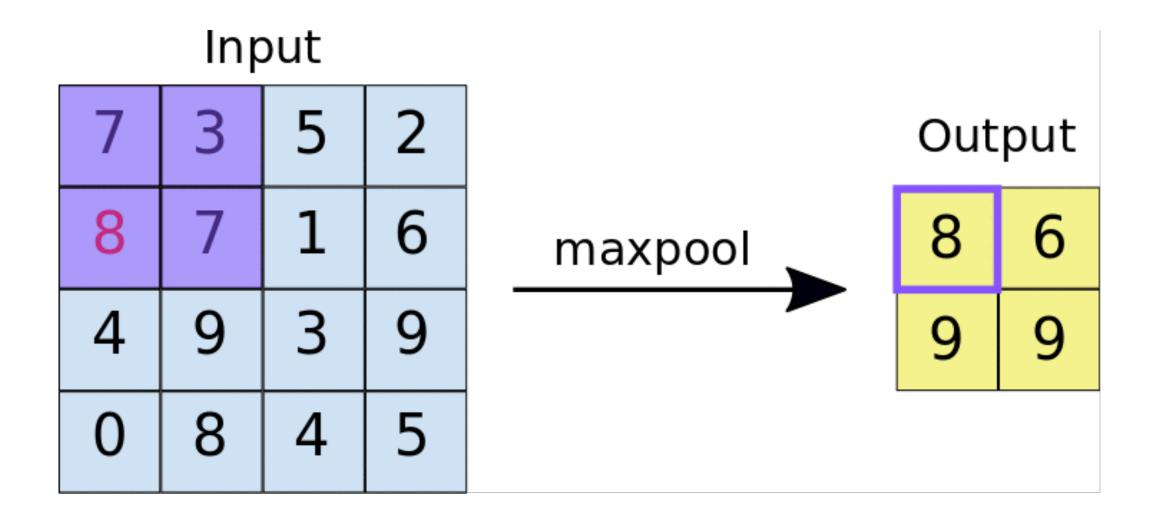


Pooling aggregates the values in the feature map to reduce its dimensions and make the network more robust to postition shift of the object of interest

The most common pooling operations are max pooling, which takes the max value within the pooling filter, and mean pooling, which takes the mean value. 2x2 is a common pooling filter size.

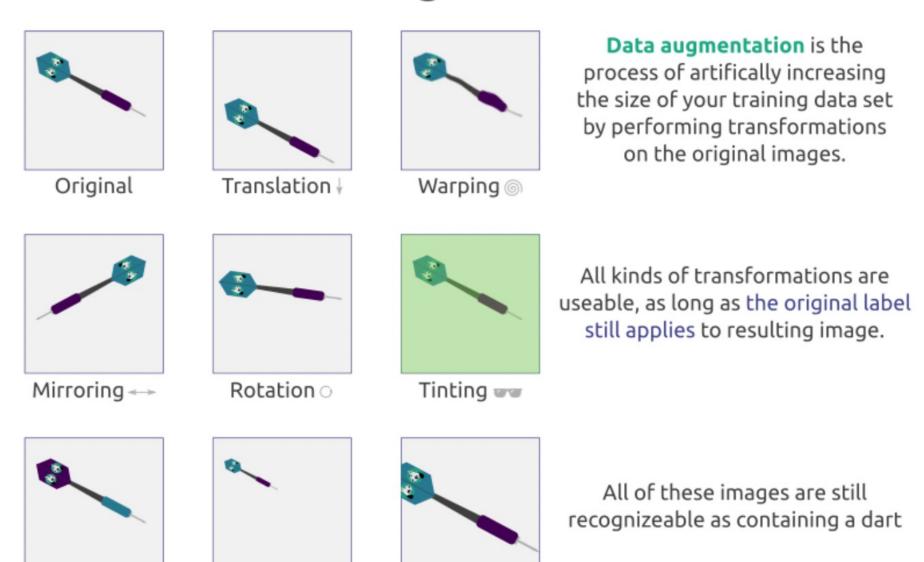








Data Augmentation



Cropping 4



Color change **I**

Scaling 🔚

Preprocesamiento de datos

1. Obtener imágenes



2. Declarar directorios



3. Reescalar / augmentation



4. Flow (crear batches)





Preprocesamiento de datos

1. Obtener imágenes





>> !wget .zip

2. Declarar directorios





>> train_dir = "pizza_sushi/train/"

>> test_dir = "pizza_sushi/test/"

3. Reescalar / augmentation



- >> from tensorflow.keras.preprocessing.image import ImageDataGenerator
- >> train datagen = ImageDataGenerator(rescale=1/255.)
- >> test datagen = ImageDataGenerator(rescale=1/255.)

4. Flow (crear batches)





>> train_data = train_datagen.flow_from_directory(directory=train_dir, target size=(224, 224), class_mode='binary', batch_size=32)

Preprocesamiento de datos

1. Obtener imágenes





>> !wget .zip

2. Declarar directorios





>> train_dir = "pizza_sushi/train/"

>> test_dir = "pizza_sushi/test/"

3. Reescalar / augmentation



- >> from tensorflow.keras.preprocessing.image import ImageDataGenerator
- >> train_datagen_augmented = ImageDataGenerator(rescale = 1/255., rotation range = 20, shear range = 0.2, zoom range = 0.2, width shift range = 0.2, height shift range = 0.2, horizontal flip = True)

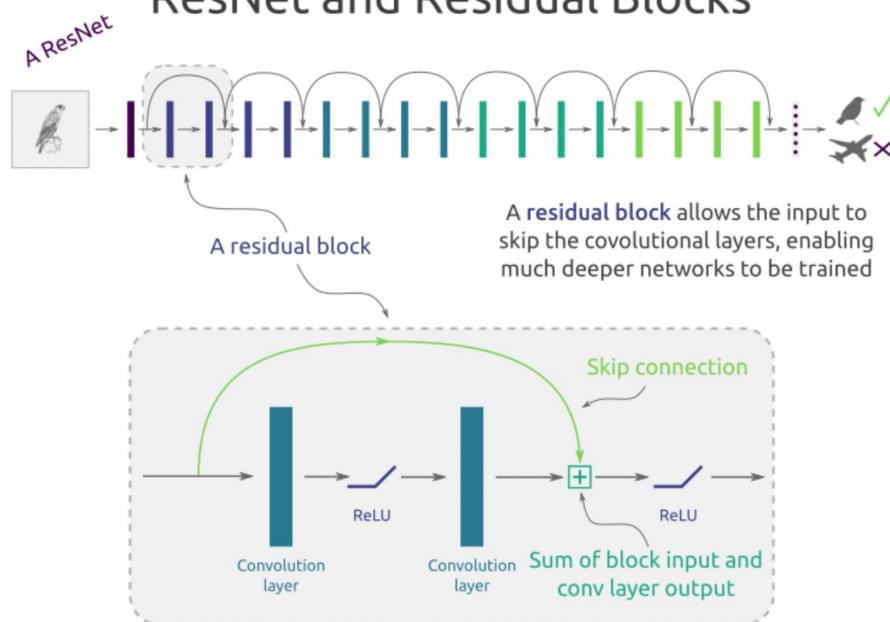
4. Flow (crear batches)





>> train_data = train_datagen.flow_from_directory(directory=train_dir, target size=(224, 224), class_mode='binary', batch_size=32)

ResNet and Residual Blocks





Referencias

- Foto de portada: Joshua Hoehne
- Meor Amer, (2022), "A visual Introduction to Deep Learning"
- Roy Keyes, (2022), "Deep Learning"

GRACIAS

emilio.sandpal@gmail.com

