







# Deep Learning: Keras and Tensorflow

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**UNIBA:** *http://www.uniba.it* 

**DIB:** *http://www.di.uniba.it* 

**KDDE:** *http://kdde.di.uniba.it* 

## **Deep Learning frameworks**





## Tensorflow



- **Open-source library** for Deep Neural Network
- Developed by Google Brain team
- First version developed in 2015; version 2.0 in 2018
- Can be used in a wide variety of programming languages:
  - Python, Javascript, C++, and Java
- Supports running computations on a variety of types of devices:
  - including CPU and GPU
- Basic type:
  - Tensor: multi-dimensional arrays with a uniform type, (kind of) like *np.arrays*



https://www.tensorflow.org/

# Keras

- Open-source library that provide an interface for DL architectures
- High-level API written in Python:
  - Running on top of **TensorFlow** platform
- Provides modular abstractions and building blocks for developing artificial neural networks:
  - contains implementations of commonly neural-network building blocks (e.g., Convolutional layer, Dense layer), activation functions, loss function.

https://keras.io/





## **Keras the functional API**



- The Keras *functional API* is a way to create models
  - more flexible than the <u>tf.keras.Sequential</u> API.
  - The functional API can handle models with non-linear topology, shared layers, and even multiple inputs or outputs.
- The main idea is that a deep learning model is usually a directed acyclic graph (DAG) of layers.
  - the functional API builds graphs of layers.

```
model = Sequential()
model.add(Dense(4,activation='relu'))
model.add(Dense(4,activation='relu'))
model.add(Dense(4,activation='relu'))
input_layer = Input(shape=(3,))
Layer_1 = Dense(4, activation="relu")(input_layer)
Layer_2 = Dense(4, activation="relu")(Layer_1)
```

### https://colab.research.google.com/

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### Colab is essentially the Google version of a Jupyter Notebook. Advantages:

**Google Colab** 

- - zero configuration Ο
  - free access to GPUs & CPUs Ο
  - sharing of code.



Free platform from Google that allows users to code in Python.





## Google Colab:GPU



- Edit
- Notebook settings

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https://colab.research.google.com/

## **Exercise 1: CNN**

Classification of MNIST dataset:

- 1 CNN
  - 3 Convolutional layers with padding = 0 and stride =1
  - 16,32,64 neurons
  - A MaxPooling layer
  - A Dropout layer
  - $\circ$  A Dense layer with 128 neurons
- Creation of adversarial examples using Adversarial Robustness Toolbox library
  - <u>https://adversarial-robustness-toolbox.readthedocs.io/en/latest/modules/att</u> <u>acks/evasion.html</u>



### **Deep Learning**

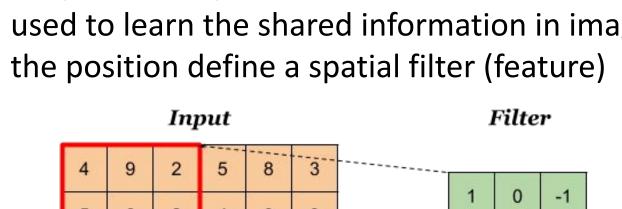
### https://indoml.com

Result

2 = 4\*1 + 9\*0 + 2\*(-1) +

5\*1 + 6\*0 + 2\*(-1) +

2\*1 + 4\*0 + 5\*(-1)



### \* -1 -1

### • A transformation filter:

**CNNs: filters** 

 $n_H x n_W = 6 x 6$ 

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• used to learn the shared information in image pixels

Parameters:

Padding: p = 0

Size:

Stride:

f = 3

S = 1

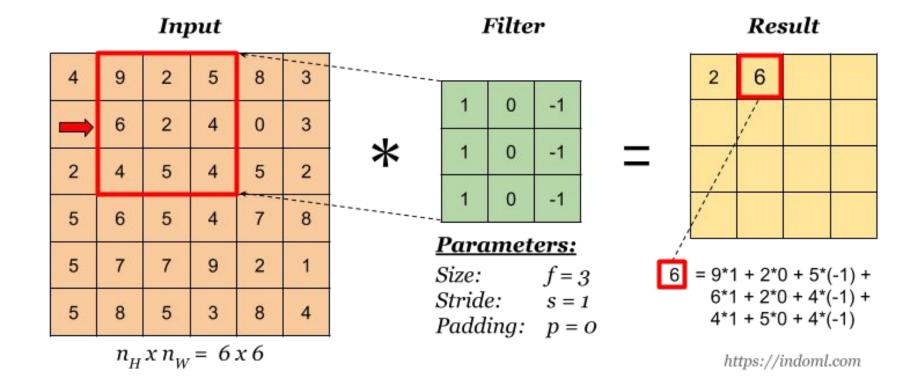


### Deep Learning

## **CNNs: stride**

### • Stride:

 $\circ$  how many cells filter is moved next in one step.

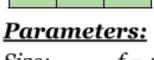




### **Deep Learning**

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### 1 = 2\*1 + 5\*0 + 3\*(-1) + Size: f=32\*1 + 4\*0 + 3\*(-1) + Stride: s = 25\*1 + 4\*0 + 2\*(-1) Padding: p = ohttps://indoml.com



Filter

0

0

0

1

1

1

-1

-1

-1

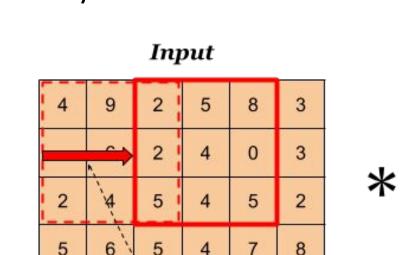
2



## **CNNs: stride**

### Stride:

how many cells filter is moved next in one step. Ο



4

9

3

7.

5

Dimension: 6 x 6

7

8

5

5

8

4

2

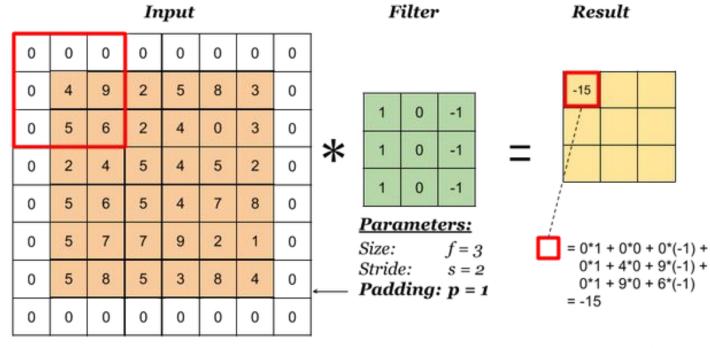
8



## **CNNs: padding**



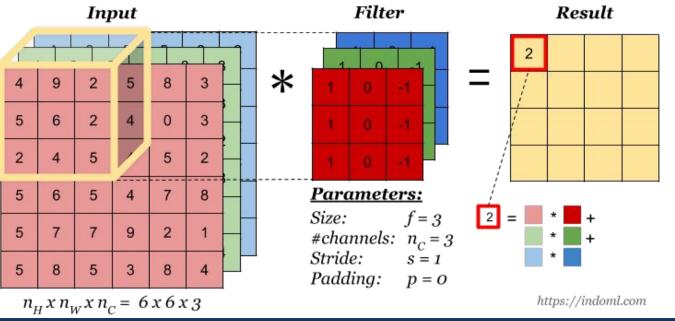
- Add the pixels around the border of feature map in order to maintain the spatial footprint
- The value of the padded features is 0





## **CNNs: convolution on 3 channels**

- Channels
  - RGB input: 3 channels red, green and blue
  - A convolution layer:
    - receives the image (h×w×c) as input
    - generates as output an activation map of dimensions h'×w'×c'.
  - The number of channels is the **depth** of the matrices involved in the convolutions



## **Exercise 2**



- Goal: create an Adversarial Autoencoder
  - Using MINST Clothes dataset

## **Exercise 3**



- Goal: create an unsupervised GAN
  - Using MINST Clothes dataset