



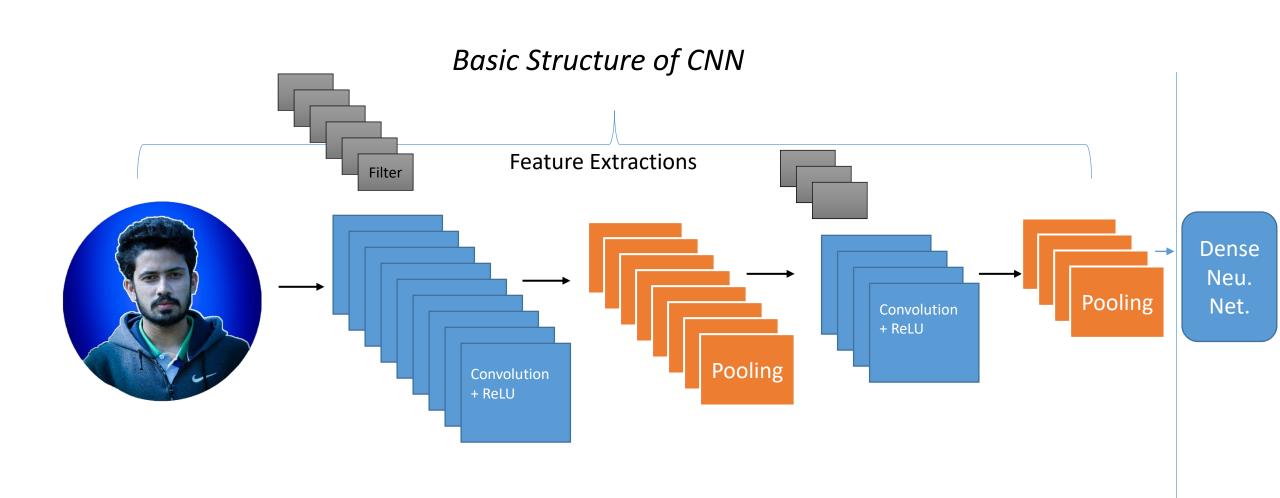
- Images recognition
- Images classifications
- Objects detections
- Decoding facial recognition
- Understanding climate
- Driverless cars
- Human genome mapping
- Predicting earthquakes
- Natural disasters



#### How many layers does CNN have?

There are **three types** of layers in a convolutional neural network: **Convolutional layer**, **Pooling layer**, and **Fully connected layer**. Each of these **layers has** different parameters that can be optimized and performs a different task on the input data.





# Images: Grayscale





Channel: 1

Type: Less Complex

302\* 320 Grayscale image

# Images: RGB



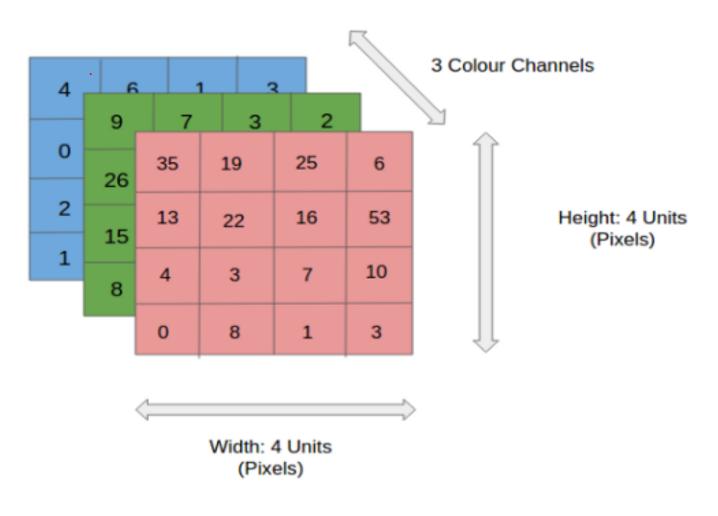


Channel: 3

Type: Complex

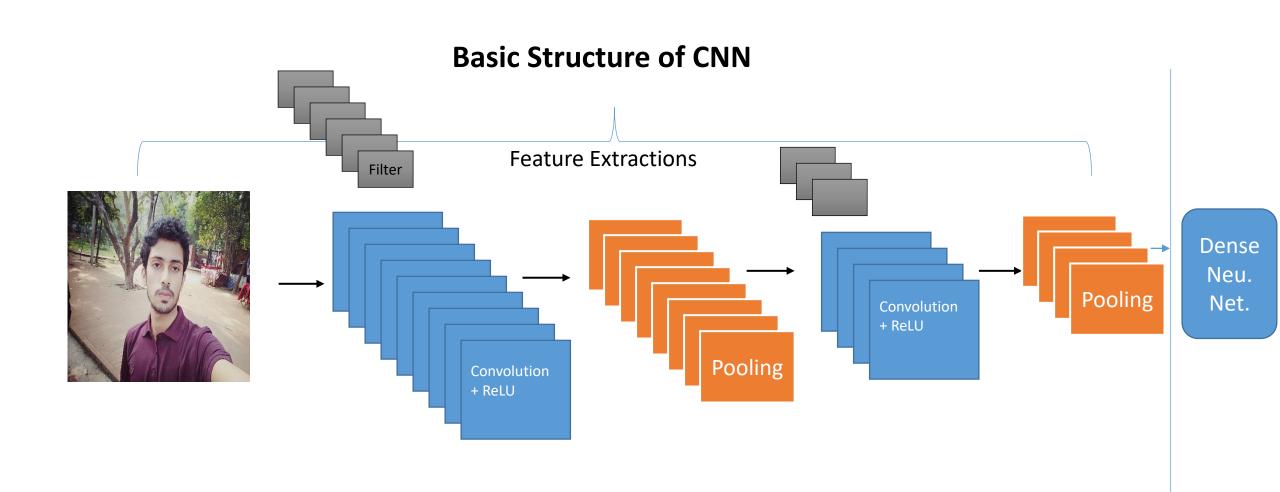
3840\* 2160 RGB image

# **Quest**



4\*4\*3 RGB image







#### **Convolution Layer — The Kernel**

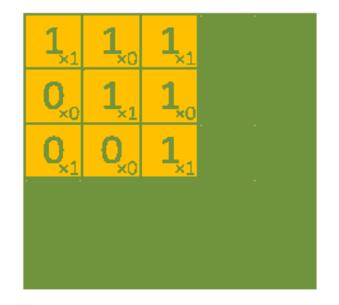
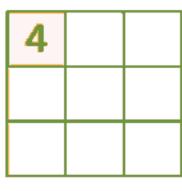
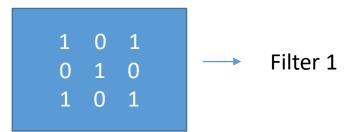


Image 5\*5



Feature Map 3\*3

Convolved Feature

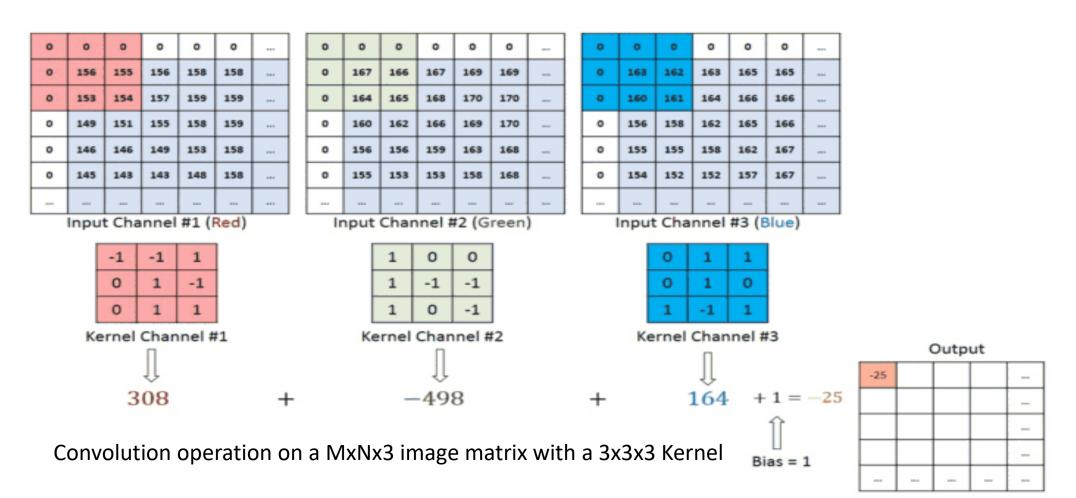


Multiplication: 
$$1*1 + 1*0 + 1*1 + 0*0 + 1*1 + 1*0 + 0*1 + 0*0 + 1*1 = 4$$

Map size = 
$$(N - F + 1) * (N - F + 1)$$
  
=  $(5 - 3 + 1) * (5 - 3 + 1)$   
=  $3*3$  Matrix

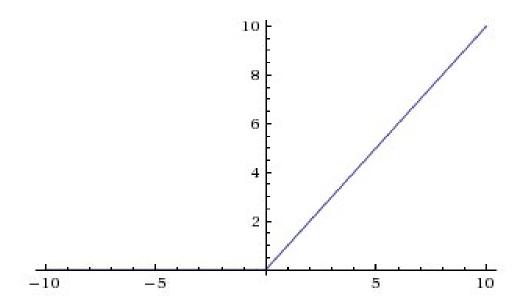


#### **Convolution Layer** — The Kernel





Activation function: Rectified Linear Units (ReLU)

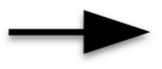




#### Activation function: Rectified Linear Units (ReLU)

#### Filter 1 Feature Map

9	3	5	-8
-6	2	-3	1
1	3	4	1
3	-4	5	1



9	3	5	0
0	2	0	1
1	3	4	1
3	0	5	1



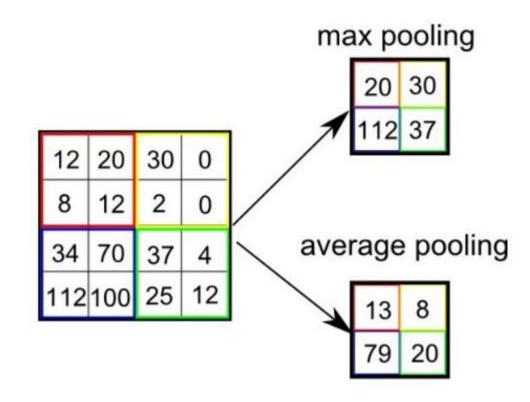
## **Pooling Layer**

3.0	3.0	3.0
3.0	3.0	3.0
3.0	2.0	3.0

3	3	2	1	0
0	0	1	3	1
3	1	2	2	3
2	0	0	2	2
2	0	0	0	1



#### **Pooling Layer**





#### **Padding**

What is padding in CNN?

**Padding** is a term relevant to convolutional neural networks as it refers to the amount of pixels added to an image when it is being processed by the kernel of a **CNN**. For example, if the **padding** in a **CNN** is set to zero, then every pixel value that is added will be of value zero.



#### **Padding**

Why does CNN use padding?

**Padding** is simply a process of adding layers of zeros to our input images so as to avoid the problems mentioned above. This prevents shrinking as, if  $p = number of layers of zeros added to the border of the image, then our <math>(n \times n)$  image becomes  $(n + 2p) \times (n + 2p)$  image after **padding**.

35	19		25	6		0	0	0	0	0	0
35	) 19		25	0	P = 1	0	35	19	25	6	0
10	,		16	F2			0				
13	3 22		16	53		0	13	22	16	53	0
4	3		7	10		0	4	3	7	10	0
9	8		1	3		0	9	8	1	3	0
		- 11	_			0	0	0	0	0	0
4 * 4			4	Input size = $n + 2p$ = $4 + 2*1 - 3 + 1$			E	5 * 6			



#### **Padding**

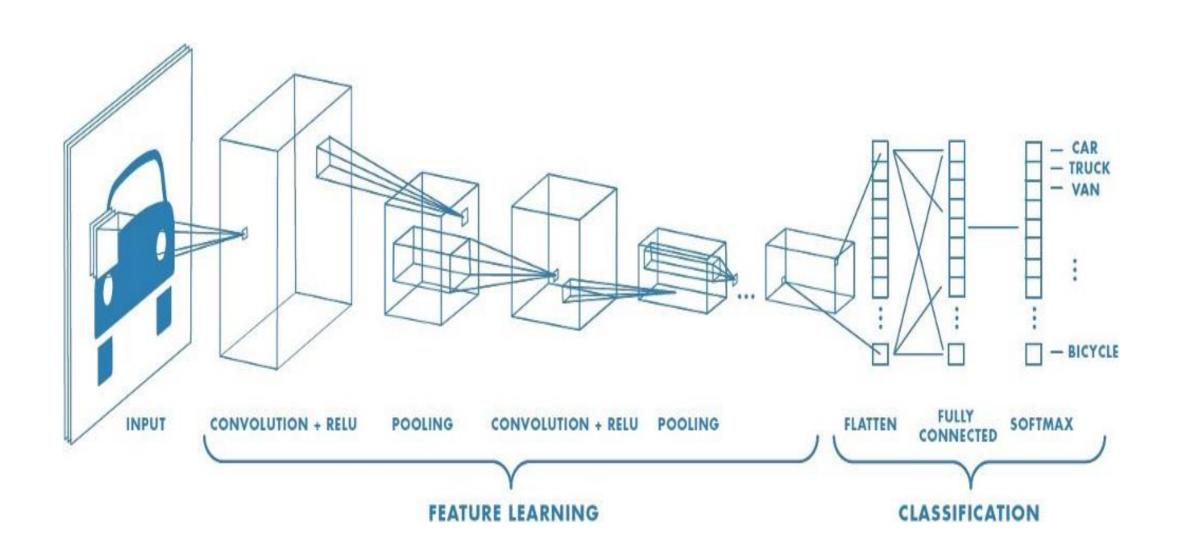
#### **Types of Padding:**

- Valid Padding: It implies no padding at all. The input image is left in its valid/unaltered shape. So,  $[(n \times n) \text{ image}] * [(f \times f) \text{ filter}] \longrightarrow [(n f + 1) \times (n f + 1) \text{ image}]$
- Same Padding: In this case, we add 'p' padding layers such that the output image has the same dimensions as the input image.

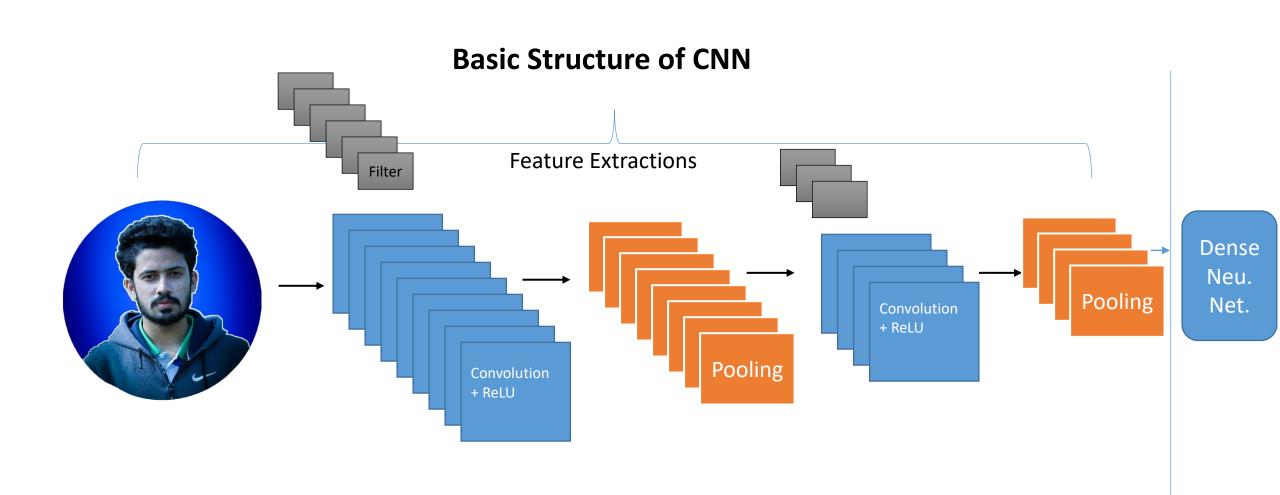
So, 
$$[(n + 2p) \times (n + 2p) \text{ image}] * [(f \times f) \text{ filter}] -> [(n \times n) \text{ image}]$$

Output size = 
$$n + 2p - f + 1$$





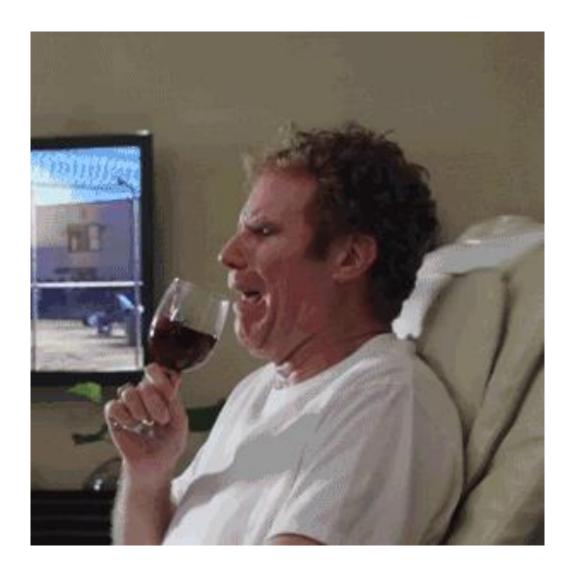




# **Data Augmentation in Deep Learning**



How do I get more data, if I don't have "more data"?





# Data Augmentation in Deep Learning



What is Data Augmentation?

















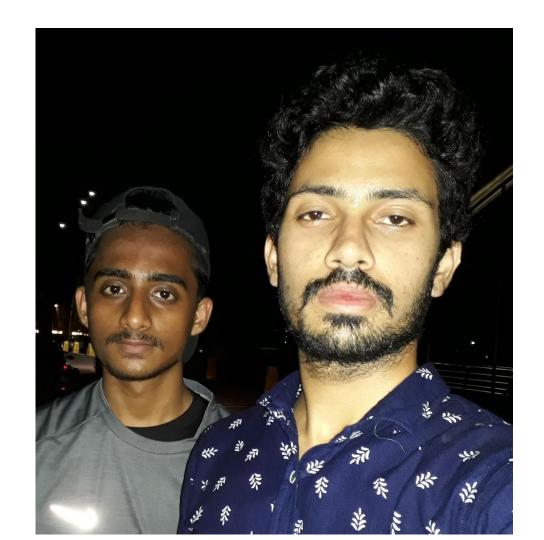






# 1. Zoom in/out

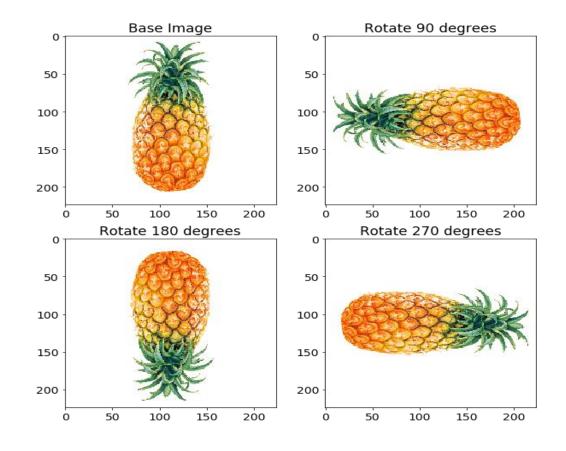








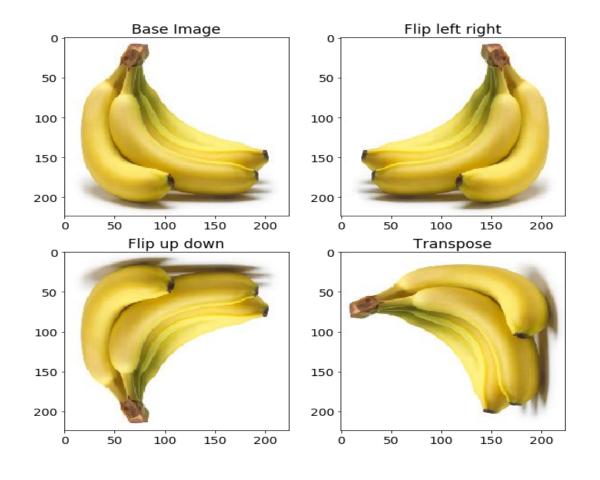
#### 2. Rotation







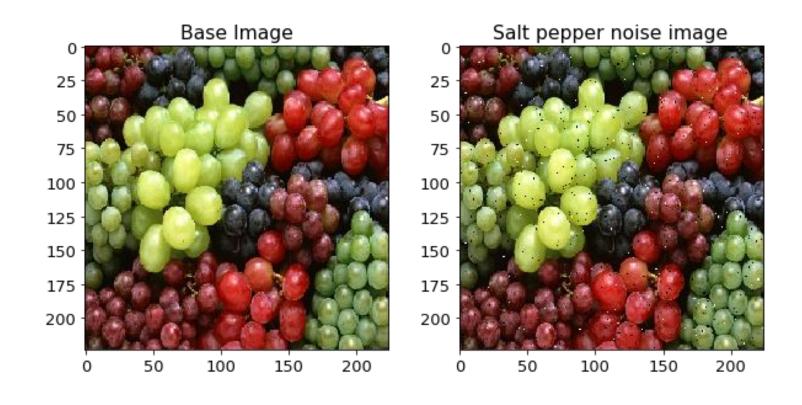
3. Flipping







#### 4. Adding Noise



# Data Augmentation in Deep Learning



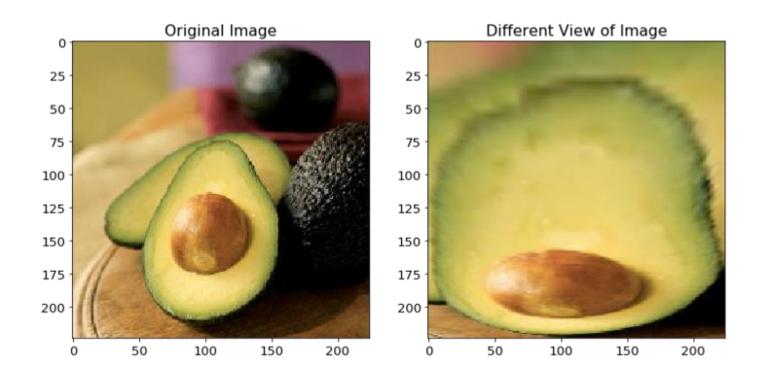
#### **5. Lighting Condition**



# Data Augmentation in Deep Learning



#### 6. Perspective transform



Watch: https://youtu.be/p8e7dGY-Oko



