

## Tutorial-5

### Understanding CNN and Implementation

Day-5 : ~~Deep~~ Deep Learning

ANN  $\rightarrow$  Theoretical, Practical

CNN  $\rightarrow$  Convolutional neural network as Images, video frames  
 $\downarrow$   
 Image classification, object detection

Agenda:

① CNN vs Human Brain

② Convolution Operations

- $\rightarrow$  Convolution
- $\rightarrow$  padding (prevent to lose img size. If my img size is decreasing we are losing some info. In order to prevent loss we use padding)
- $\rightarrow$  Strides (moves one cell to other cell is called strides)  $\rightarrow$  stride=1  
 jump 1 cell, stride=2 jump 2 cell.
- $\rightarrow$  Filters (kernels) (helps to extract information from image)

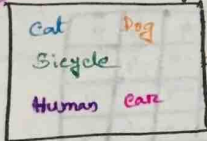
③ Max Pooling

④ Flattening (As like as ANN Dense)

⑤ Practical Implementation (CNN)

① CNN vs Human Brain

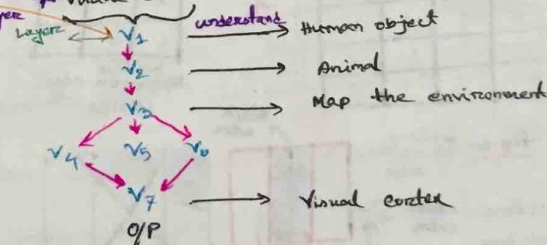
Human Brain:



Image

Cerebral Cortex

As soon as passes input Visual Cortex (identifies specific image)

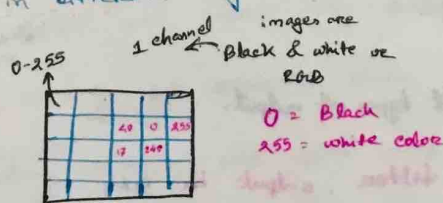


CNN:

we can divide in different stage.

① Convolution

Images  $\rightarrow$

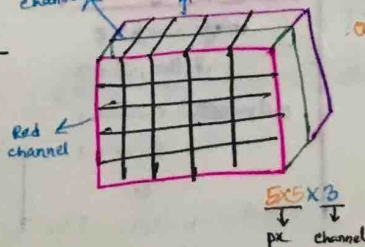


Green channel

Blue channel 5x5 px

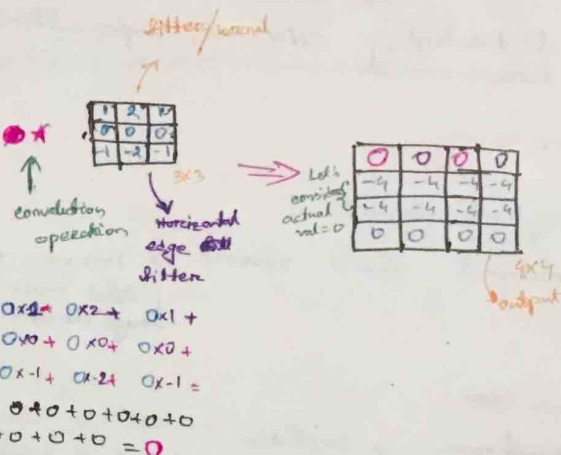
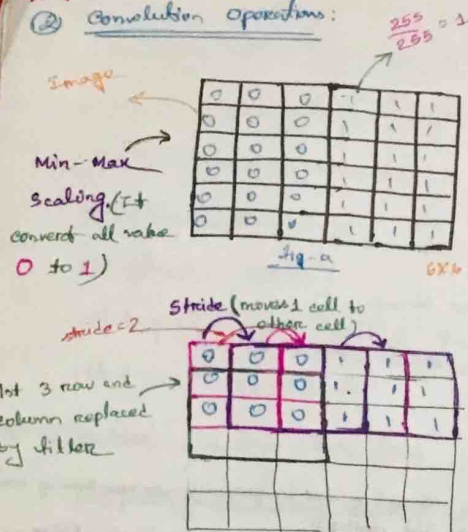
RGB

total channel=3



Step-1:

## ② Convolution operation:



stage-2:

$$0 + 0 + 1 + 0 + 0 + 0 + 0 + 0 - 1 = 0$$

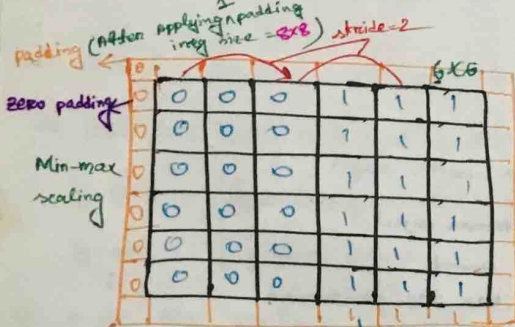
stage-3:

$$0 + 2 + 1 + 0 + 0 + 0 + 0 + 0 - 2 - 1 = 0$$

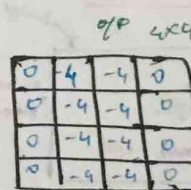
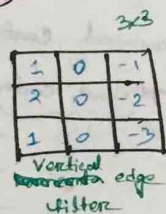
stage-4:

$$1 + 2 + 1 + 0 + 0 + 0 + 0 - 2 - 2 - 1 = 0$$

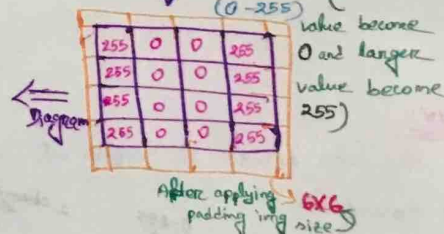
Never hard coded the filter value. We should update it during b/w propagation. Here we use activation function  $\text{ReLU}(\max(0, x))$  or other activation func.



0 = black colour  
255 = white colour



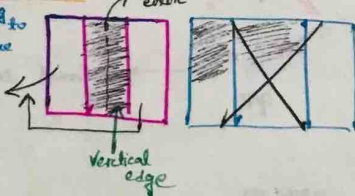
revert back (smaller (0-255))



Multiple filters are -

- ① Vertical
- ② Horizontal
- ③ Round object

helps to stretch into from img.



For different filter, we will get different types of output.

When passes 6x6 image in a 3x3 filter, output is 4x4

Mathematical formula:

$$\begin{aligned} \text{output} &= n - f + 1 \\ \text{image size, } n &= 6, P = \text{padding} = 1 \\ &= 6 - 3 + 1 \\ \text{filter, } f &= 3, s = \text{stride} = 1 \\ &= 3 + 1 \end{aligned}$$

$$\text{output} = n - f + 1$$

$$\text{output} = \frac{n + 2P - f + 1}{s}$$

$$\text{output} = \frac{n + 2P - f + 1}{s} \rightarrow \text{updating formula}$$

Padding:

If my img size is decreasing, we loose some info to prevent loss we use padding. It means protecting the image adding another layer top of it. What types of value comes inside padding? For this diff types of padding. One is 0 padding, nearest the value put there.

After applying padding

image size,  $n = 8$

filter,  $f = 3$

$$\text{output} = n - f + 1$$

$$= 8 - 3 + 1$$

$$= 6$$

$$\text{output} = 8 + 2 \times 1 - 3 + 1$$

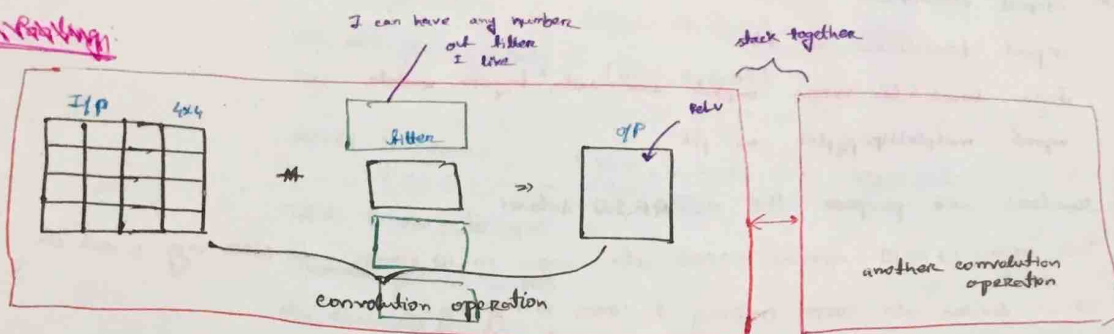
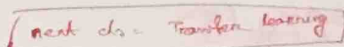
$$= 6$$

$$n = 8$$

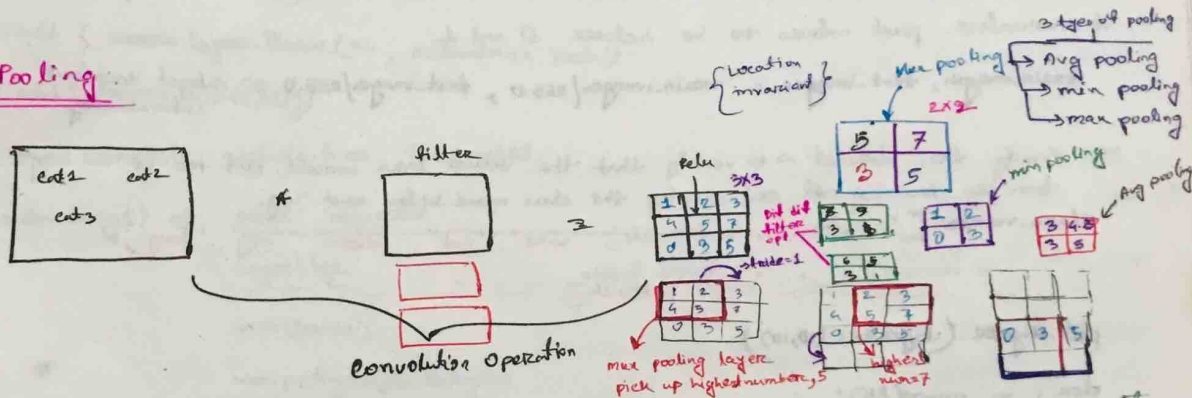
$$f = 3$$

$$P = 1$$





③ Max Pooling



The concept is called location invariant. Location invariant says that objects present anywhere but as <sup>we</sup> pass through many convolution layers I should be able to extract the info in a better and clear way. And for that we specially use something called as Max pooling. It extracts clear info.

3 types of pooling:

- Max (pick highest value from the filter)
- Min (u min u u u u)
- Avg (summation all value and provides avg value)

④ Flattening Layer: ( ~~pooling~~ <sup>helps</sup> <sup>pooling</sup> converts into elongate input )  
As Like as ANN dense.

