CHN

- Convolutional neural network

COMPUTER VISION

- Computer vision is a field of artificial intelligence
 (AI) that enables computers and systems to derive meaningful information from digital images, videos and other visual inputs and take actions or make recommendations based on that information.
- It mimics human vision to provide vision to machines to capture images and to make prediction based on that.

-> CMN => data it taku in a visval data i.e image or a video'

image => unatrutud data

WHAT IS IMAGE?

- Image is 2-D representation of real world objects made up of Pixels.
- · Images are made from pixels.
- Pixels mathematical range is 0-255.



TYPES OF IMAGES

GREY SCALE IMAGE

Grey scale image is made up of black and white pixels.

Pixel value of black is 0 and white is 255.

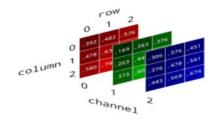




TYPES OF IMAGES

RGB SCALE IMAGE

- RGB stands for red, green and Blue.
- It is made up of three channels i.e red, green and blue channels.
- Overlap of all these three channels is a colorful image.
- Each channel has got the range (0,255).





1. Red =
$$(255, 0, 0)$$
 G: $(0, 255, 0)$
B= $(0, 0, 255)$

CNN

- Convoluted Neural Network is a deep learning algorithm which is widely used for image / object recognition and classification.
- CNN is subset of machine learning highly suitable for computer vision tasks and for applications where object recognition is vital, such as self driving cars and facial recognition.

Different layers in CNN are

Convolution layer

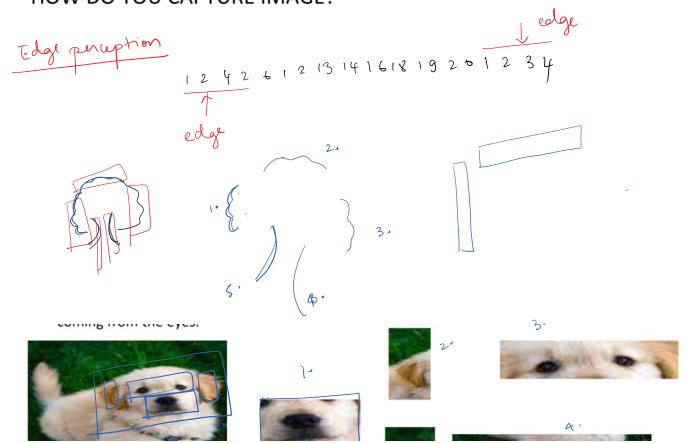
63-7[]pixel

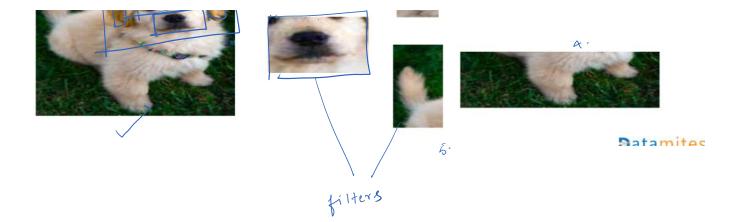
Pooling layer,

] => feature extraction

3. • Fully connected layer

HOW DO YOU CAPTURE IMAGE?





HOW CNN WORKS?

Step1: Take image



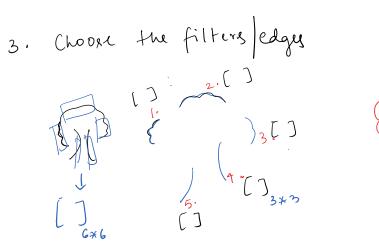
Step2: Convert image into pixels



0	0	0	1	1	1
0	0	0	1	1	1
0	0	0	1	1	1
0	0	0	1	1	1
0	0	0	1	1	1
0	0	0	1	1	1

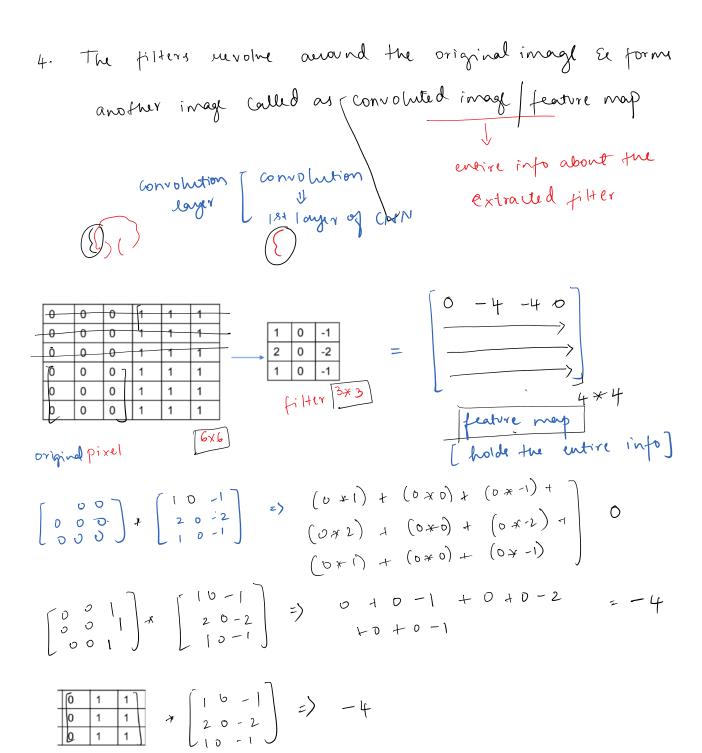
6×6





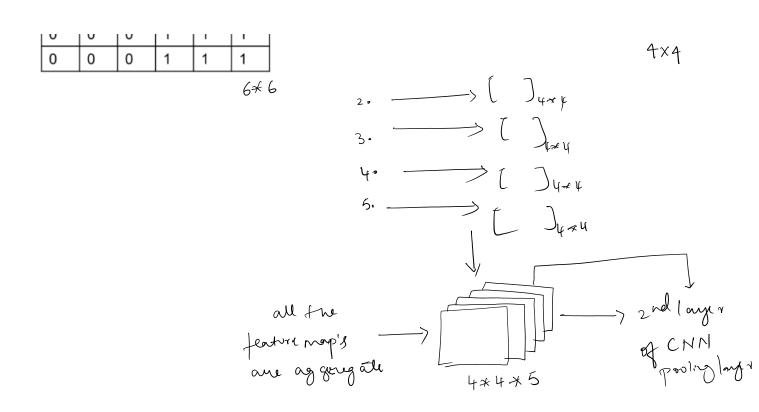


6× 6



							:	1.					
0	0	0	1	1	1	1	n	-1	1	0		_4	Ī n
0	0	0	1	1	1	-		<u> </u>			1 4	4	
0	0	0	1	1	1		0	-2		0	-4	-4	0
0	0	0	1	1	1	1	0	-1		0	-4	-4	0
0	0	0	1	1	1			3*	3	0	-4	-4	0
0	0	0	1	1	1								4

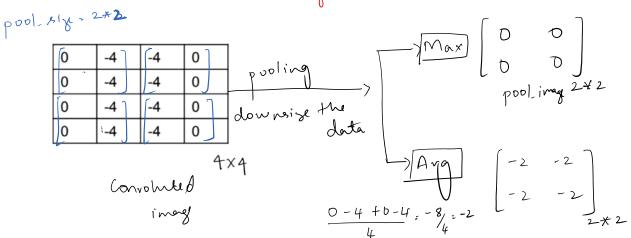
New Section 3 Page 5



2) nd larger of CHN => pooling larger

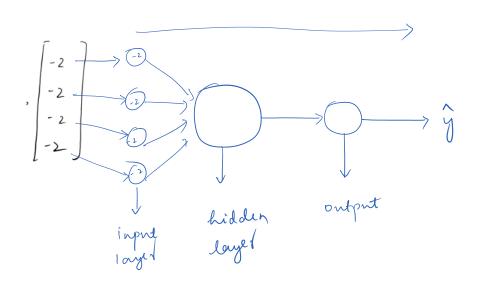
summarize the data and it will endure the

dinersion of feature maps



3) flatter layer
$$\begin{bmatrix}
-2 \\
-2 \\
-2
\end{bmatrix}$$
flattening
$$\begin{bmatrix}
-2 \\
-2 \\
-2
\end{bmatrix}$$
pool-imag

3rd larger og CNN i.e fully connected larger (FC larger)
=> artval learning of the data



$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$						
8 5 2 4 3 1 2 6	2 _/	2	7	1	//	
0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	8	5	2	4		
0 4 9 6 1 0 0 8 5 2 4 0 0 3 1 2 6 0		<u></u>	\sim			
0 4 9 6 1 0 0 8 5 2 4 0 0 3 1 2 6 0	1000		100	1000		
0 3 1 2 6 0		4	9		1	0
	0	3	1	2	6	0

PADDING

- Padding is process of adding new layers to the border of an image.
- The pixel in the corner will only get covers one time but if you
 take the middle pixel it will get covered more than once that
 means we have more information from middle pixels than corner
 pixels.
- Challenges faced are like "Shrinking output" and "loss of information on corners" and to overcome these challenges we have padding concept.
- Padding helps by adding extra rows and columns on the outer dimension of the data.

0	0	0	0	0	0
0	2	2	7	3	0
0	4	9	6	1	0
0	8	5	2	4	0
0	3	1	2	6	0
0	0	0	0	0	0

TYPES OF PADDING

VALID PADDING(NO PADDING)

It says that new layers are not added

ZERO /SAME PADDING

Add zeros at the edges of layers.

CONSTANT PADDING

Adding constant values at the edges of layer

2	2	7	3
4	9	6	1
8	5	2	4
3	1	2	6

0	0	0	0	0	0
0	2	2	7	3	0
0	4	9	6	1	0
0	8	5	2	4	0
0	3	1	2	6	0
0	0	0	0	0	0

1	1	1	1	1	1
1	2	2	7	3	1
1	4	9	6	1	1
1	8	5	2	4	1
1	3	1	2	6	1
1	1	1	1	1	1