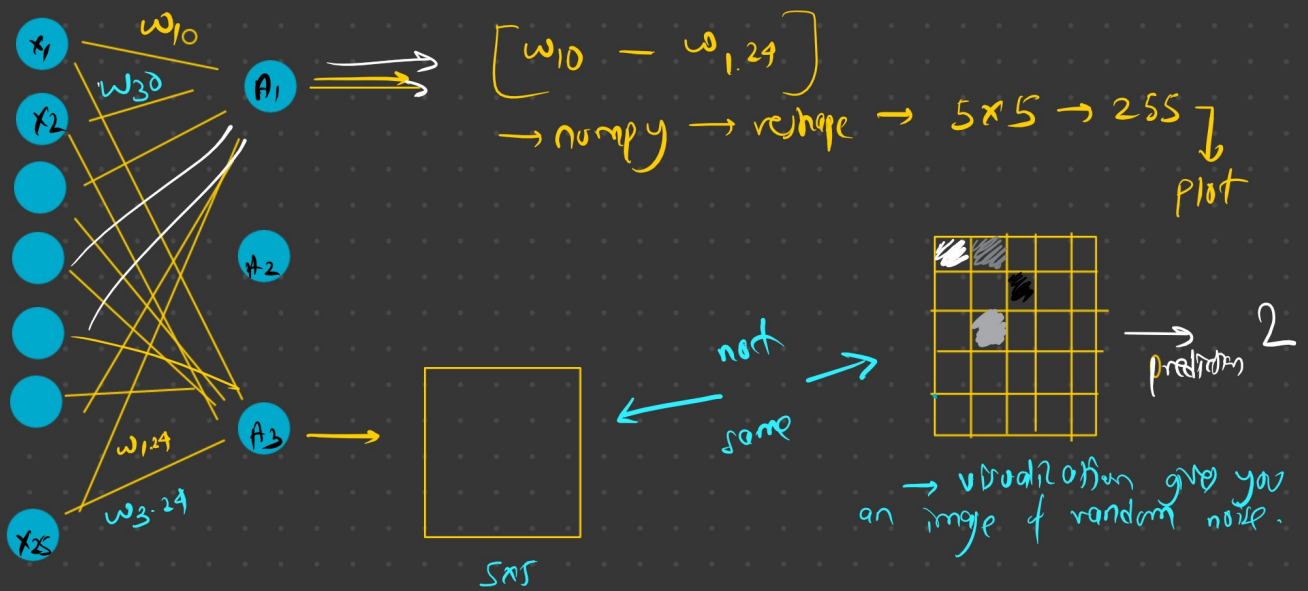


Agenda - 7 Dec 2024

Previous → ANN

Today's → CNN - Convolution Neural Network

 → Flattening → ANN
 $5 \times 5 = 25$



ANN — Disadvantage

→ ANN is not focusing on features of image but finds features which are responsible for prediction.

→ Image — $720 \times 720 \rightarrow \text{RGB}$

— ANN → 1,555,200 → $720 \times 720 \times 3$

→ 15 lakh 55 thousand 200 neuron

→ Higher Computation is required.

→ Not optimized for images.

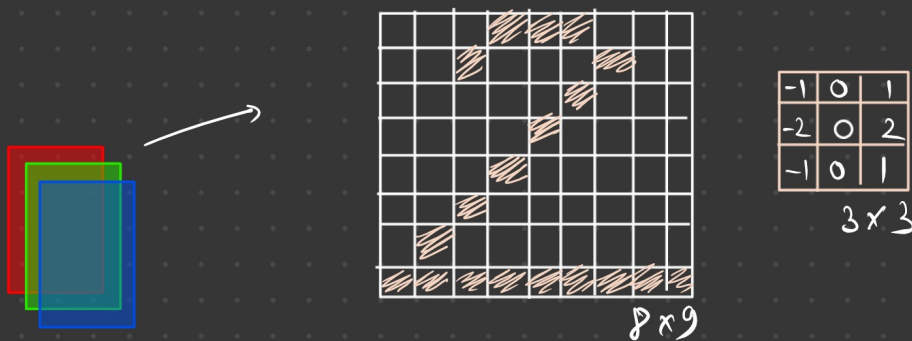
CNN → are a type of deep learning model which is effective for image processing & other tasks that are based on grid-like data.

Components of CNN

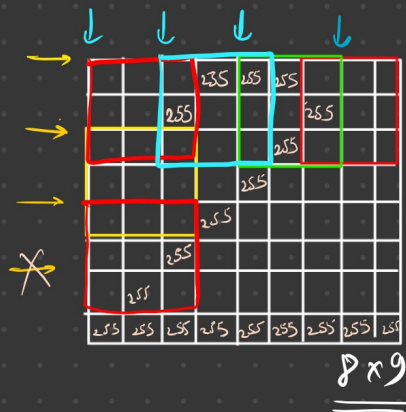
- Convolution layer
- Pooling layer
- Fully connected layer [FC layer]

(1) Convolution layer

— core building block of a CNN. It is responsible for detecting local features in the input data such as edges, shapes & textures.



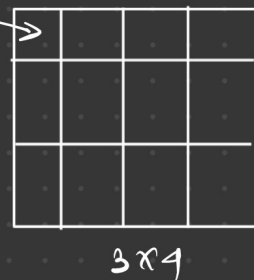
- convolve function
- stride
- kernel → matrix of weights
 - sobel filter
 - Horizontal edges



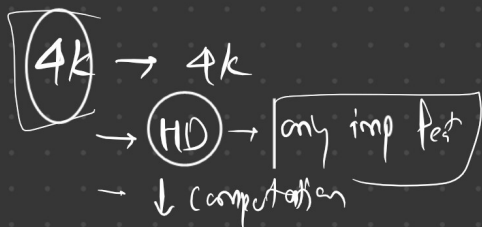
-1	0	1
-2	0	2
-1	0	1

3x3

$$\begin{aligned}
 &-1 \times 0 + 0 + 255 \\
 &-510 + 0 + 0 \\
 &0 + 0 + 0
 \end{aligned}$$



important features based on your kernel

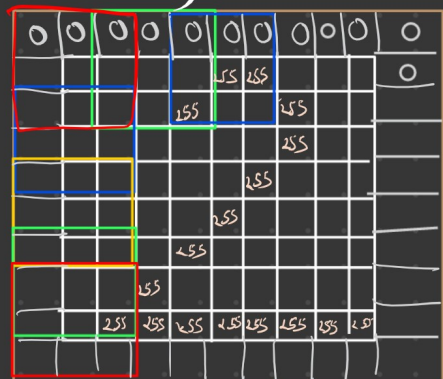


processed \rightarrow 7
Actual \rightarrow 2

reduce the stride \rightarrow 1

\rightarrow Advantage \rightarrow cover more of image
 \rightarrow with stride 2, we will lose information

\rightarrow Padding



\rightarrow we might miss features at the corners of the image

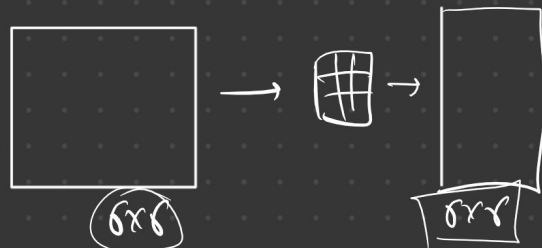
\rightarrow size of kernel

\rightarrow stride

\rightarrow stride - 2

\rightarrow Valid padding (No padding)

\rightarrow same padding



padding = 0 \rightarrow valid

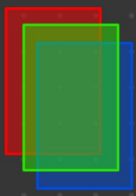
padding = "same" \rightarrow same

= 1 \rightarrow padd 1 pix with 0

= 2 \rightarrow " 2 " " "

= 3 \rightarrow " 3 " " "

Pooling



$\rightarrow R \rightarrow$ Convolution \rightarrow
 \rightarrow stride = 2
 \rightarrow kernel = 3×3

100	70	10	40	60
20	70	20	100	200
250	100	30	170	50
10	11	70	12	145
6	170	0	100	100

5x5



2x2

kernel

stride = 2

max pool

1.5 million \rightarrow 1.2 million \rightarrow pooling
input conv

\rightarrow reduce dimension

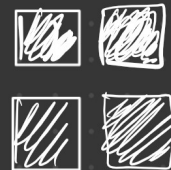
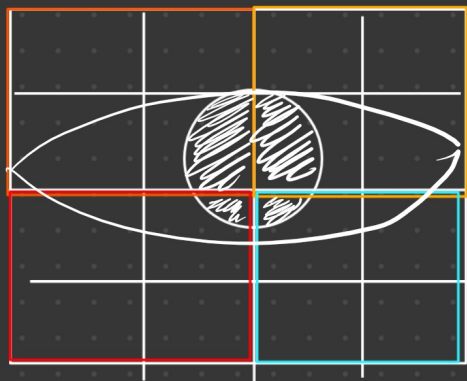
\rightarrow prominent features

100 100 200

val val val

val val val

3x3



Different kinds of pooling

\rightarrow Max pooling

\rightarrow Min pooling

\rightarrow Average pooling

Input image $\rightarrow 60 \times 60$
 — kernel $\rightarrow 3 \times 3$
 — padding $\rightarrow 1$
 — stride $\rightarrow 2$
 \rightarrow pooling $\rightarrow 2$

Image \rightarrow Conv \rightarrow op 1 \rightarrow max pooling \rightarrow op 2

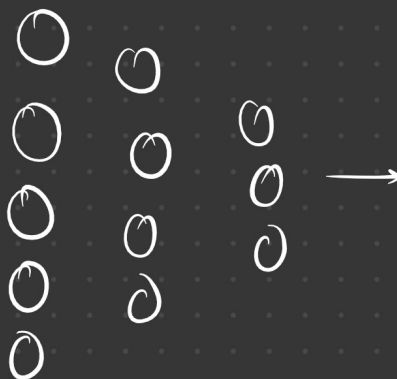
$$\text{output} \rightarrow \frac{\text{Input size} - \text{kernel} / \text{pool size} + 2 \times \text{padding}}{\text{stride}} + 1$$

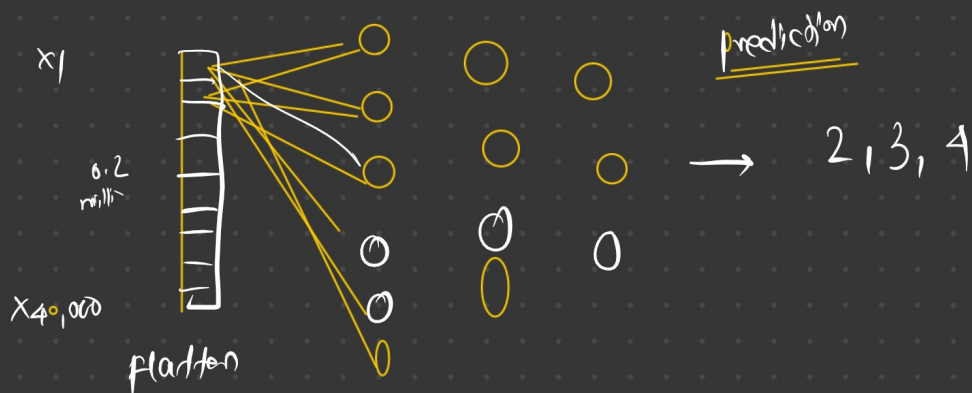
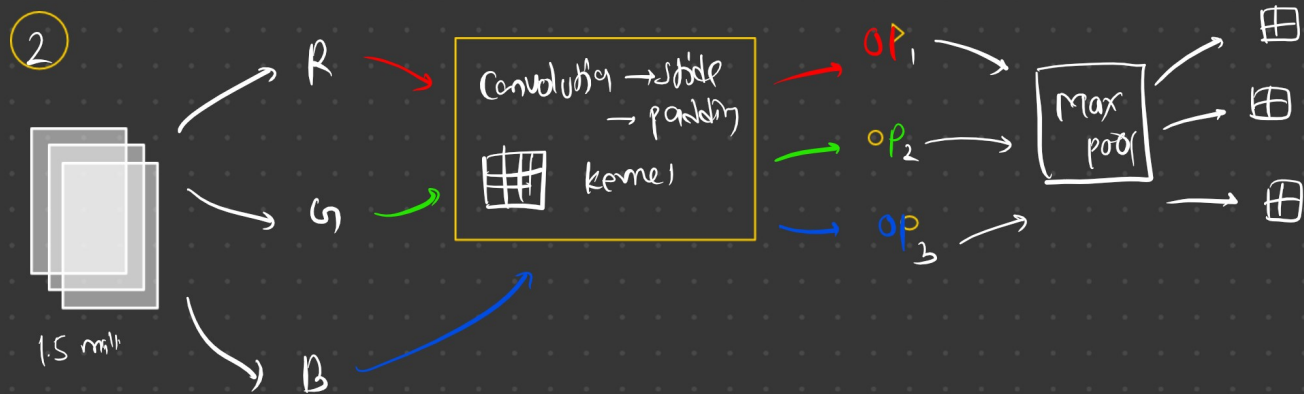
$$\underline{\text{op 1}} \rightarrow \frac{60 - 3 + 2 \times 1}{2} + 1 \rightarrow \boxed{30}$$

$$\underline{\text{op 2}} \rightarrow \frac{30 - 2 + 2 \times 0}{2} + 1 \rightarrow \frac{28}{2} + 1 \rightarrow \boxed{15}$$

FCL

Fully connected layer





- \rightarrow Extract relevant feature or prominent feature
- \rightarrow To reduce dimension of the image

