

CNN [convolutional neural n/w]

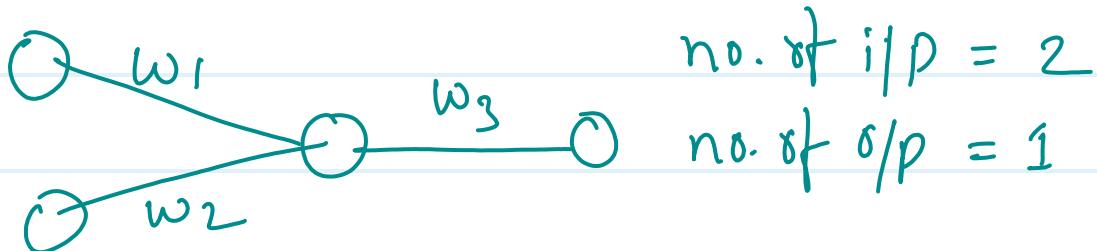
convolution + ANN, Architecture

① weight Initializing Techniques

key point for weight initializing

- Ⓐ weight should be small
- Ⓑ weight should not be same
- Ⓒ weight should have good variance

w_1	w_2	w_3
0.05	0.06	0.07
0.05	0.1	0.15 ✓



① Uniform Distribution -

$w_{ij} \approx \text{uniform dist. } \left[\frac{-1}{\sqrt{\text{no. of } I/P}}, \frac{1}{\sqrt{\text{no. of } I/P}} \right]$

$\overset{\text{lower}}{\swarrow}$ $\overset{\text{upper}}{\searrow}$

i = weight

j = Layer

$$\Rightarrow \left[-\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}} \right]$$

② Xavier/Glorot Initialization

ⓐ Xavier Normal Initialization

ⓑ Xavier uniform initialization

ⓐ X N I

$$w_{ij} \approx N(0, \sigma)$$

$$\sigma = \sqrt{\frac{2}{\text{no. of I/P} + \text{no. of O/P}}}$$

⑤ X UI

$$w_{ij} \approx \text{uniform dist. } \left[\frac{-\sqrt{6}}{\sqrt{I/P + O/P}}, \frac{\sqrt{6}}{\sqrt{I/P + O/P}} \right]$$

③ kaiming he Initialization

④ he Normal

$$w_{ij} \approx N(0, \sigma)$$

$$\sigma = \sqrt{\frac{2}{\text{no. of I/P}}}$$

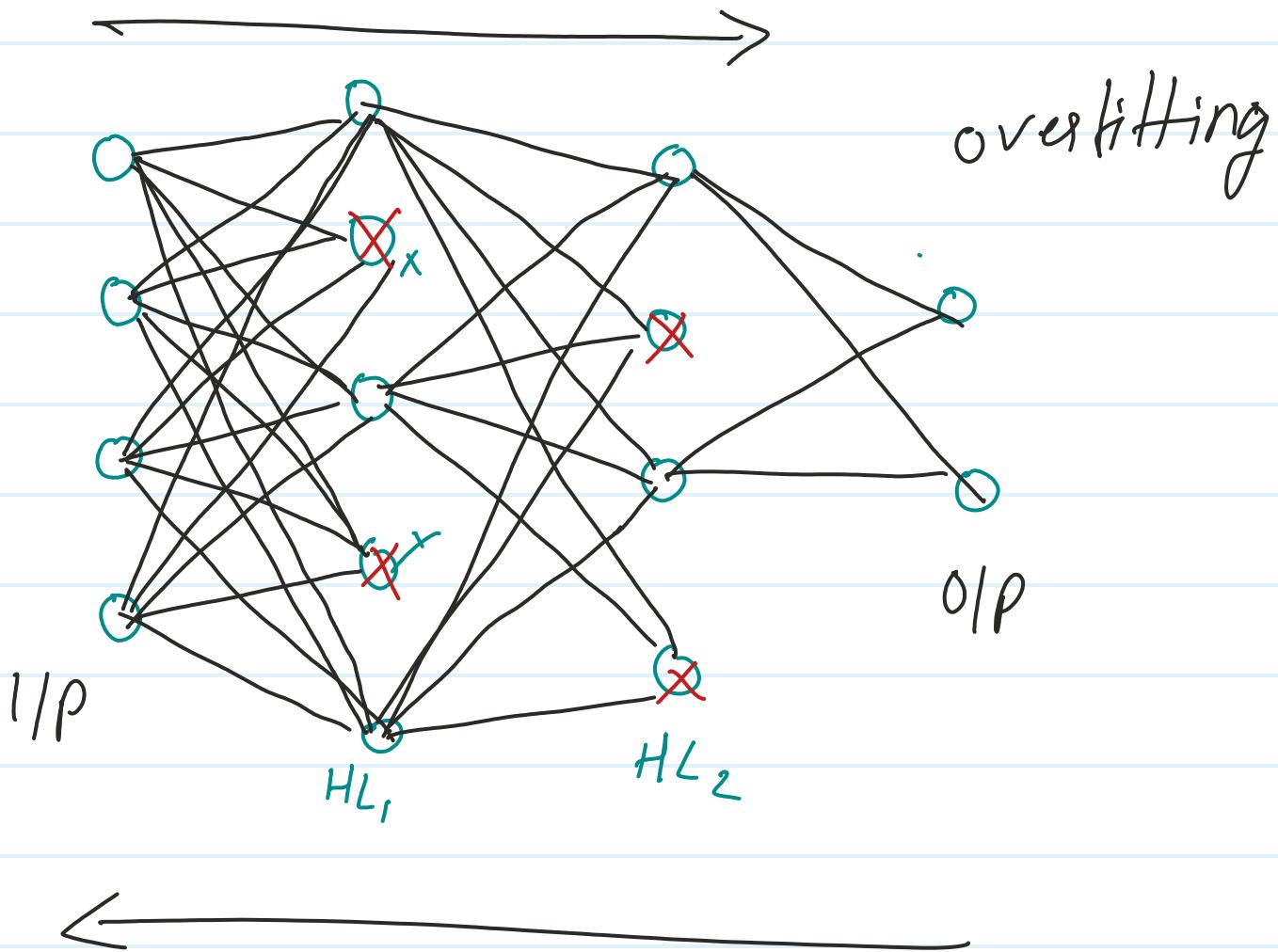
⑥ he uniform

$$w_{ij} \approx \text{uni. } \left[\frac{-\sqrt{6}}{\sqrt{I/P}}, \frac{\sqrt{6}}{\sqrt{I/P}} \right]$$

In the CNN we used most of the time xavier glorot initialization tech.

Dropout Layer -

Used to prevent from over fitting.



$$P_{HL_1} = 0.5, \quad P_{HL_2} = 0.35, \quad P_{HL_3} = 0.25$$

epoch = In every epoch random neuron will be drop.

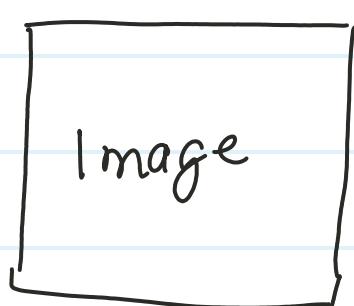
CNN

visual cortex -

$$(V_1 - V_5)$$

V_1 - primary visualization (orientation, edges, lines)

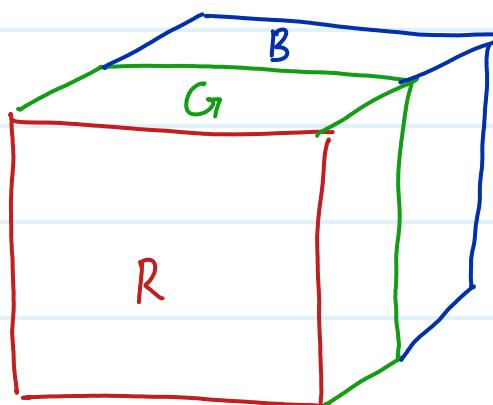
V_2 = Diff. in color, complex recognition



→ RGB

→ Gray Scale

→ A-channel for transparency



Convolution operation

0	1	-1		
0	0	0	1	0
0	0	0	1	0
0	0	1	1	
0	0	1	1	
1	0	1	1	
1	0	1	1	

6×6

$1 / P \uparrow$

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

3×3

vertical edge filter

\leftarrow filter

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

Horizontal edge filter.

stride (shift)

$O/P \rightarrow$

0	-4	-4	6
0	-4	-4	0
0	-4	-4	0
0	-4	-4	0

4×4

$$(n - f) + 1 =$$

$$6 - 3 + 1 = 4$$

we have 6×6 metrics with filter image by 3×3 metrics but getting final image by 4×4 ,

so we are losing some information.

To overcome this we use another technique called padding

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

6×6

.	.	.
3 \times 3	.	.

.	.	.
4 \times 4	.	.

$\rightarrow 8 \times 8$

6×6

Padding

$$n - f + 2p + 1 = 6$$

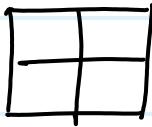
$$6 - 3 + 2p + 1 = 6$$

$$2p = 6 - 4$$

$$p = \frac{2}{2} = \boxed{1}$$

max pooling

2	1	4	2
3	0	0	1
1	2	3	1
2	1	4	0

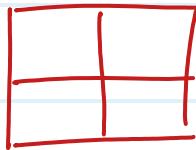


3	4
2	4

strides jump = 2

min pooling

1	2	1	2
3	4	4	5
4	5	5	9
6	2	3	1

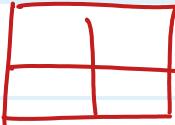


1	1
2	1

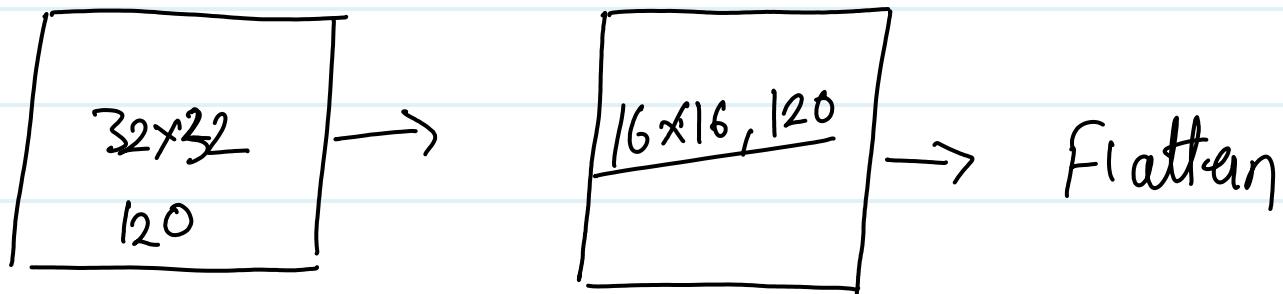
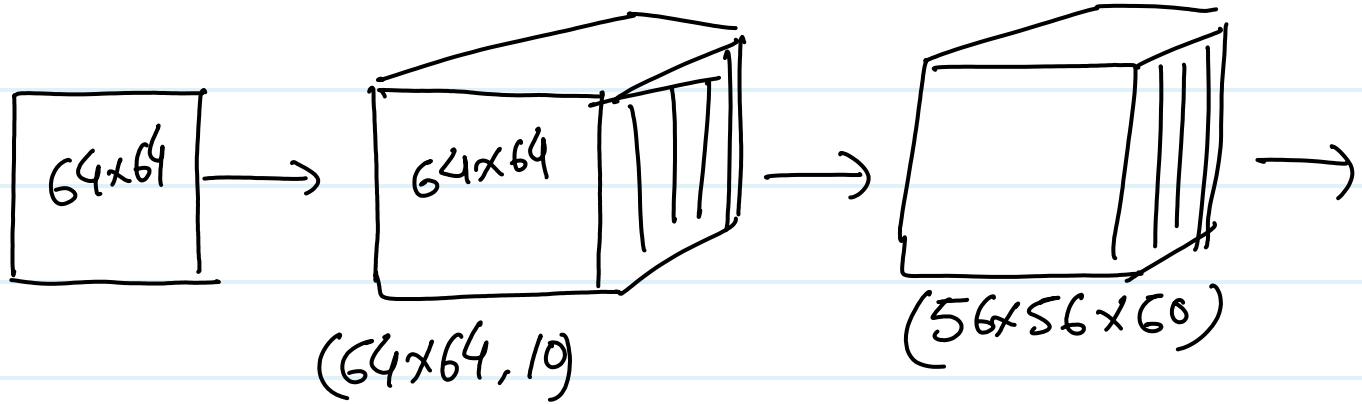
Smoothing of image for its use min
and max pooling

mean pooling (Avg pooling)

1	3	2	1
3	4	3	9
2	1	2	4
3	2	4	1



2.9	2.9
2	2.1



$16 \times 16 \times 120$

