

CNN [convolutional neural n/w]

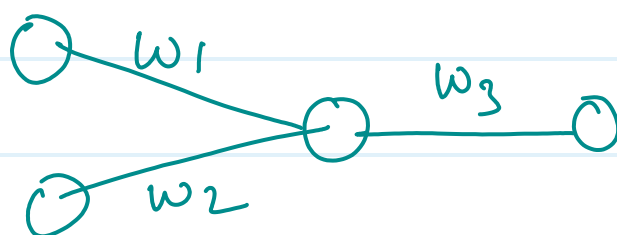
convolution + ANN, Architecture

① weight Initializing Techniques

key point for weight initializing

- ① weight should be small
- ② weight should not be same
- ③ weight should have good variance

w_1	w_2	w_2
0.05	0.06	0.07
0.05	0.1	0.15 ✓



no. of i/p = 2
no. of o/p = 1

① Uniform Distribution -

$$w_{ij} \approx \text{uniform Dist.} \left[\underbrace{\frac{-1}{\sqrt{\frac{\text{no. of}}{1/p}}}}_{\text{lower}}, \underbrace{\frac{1}{\sqrt{\frac{\text{no. of}}{1/p}}}}_{\text{upper}} \right]$$

i = weight
j = Layer

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

② Xavier/Glorot Initialization

① xavier Normal initialization

② xavier uniform initialization

③ $X \sim I$

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

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

⑥ Xavier

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

③ kaiming he initialization

① he Normal

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

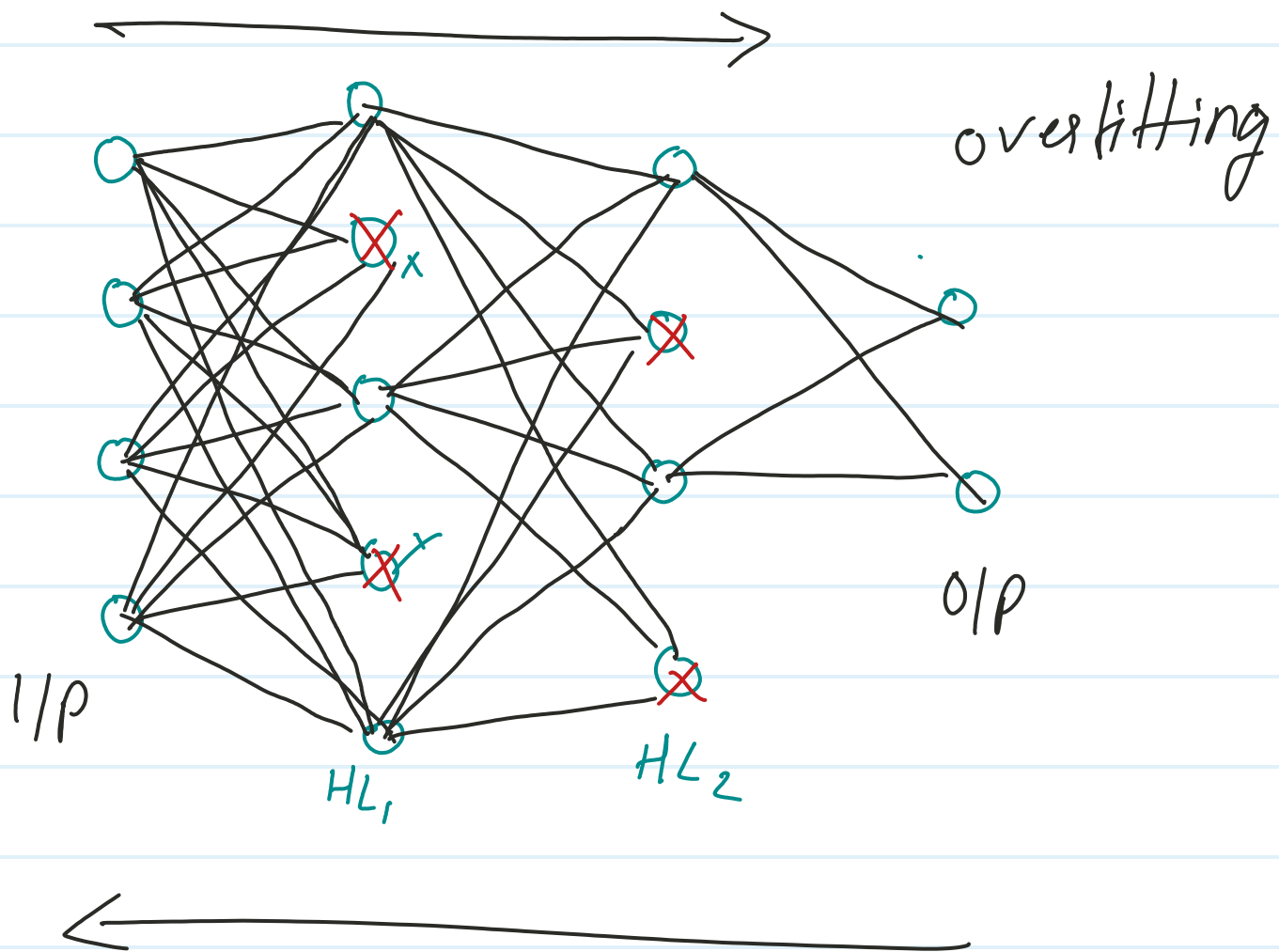
$$\sigma = \sqrt{\frac{2}{\text{no. of } I/p}}$$

② he uniform

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

In the CNN we used most of the time Xavier glot initialization tech.

Dropout Layer - Used to prevent from overfitting.



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

epoch = In every epoch random neuron will be drop.

CNN

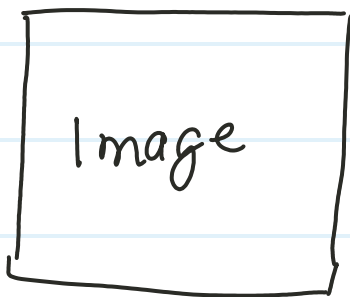
visual cortex -

$(V_1 - V_5)$

V_1 - primary visualization (orientation, edges, lines)

V_2 = Diff. in color, complex recognition

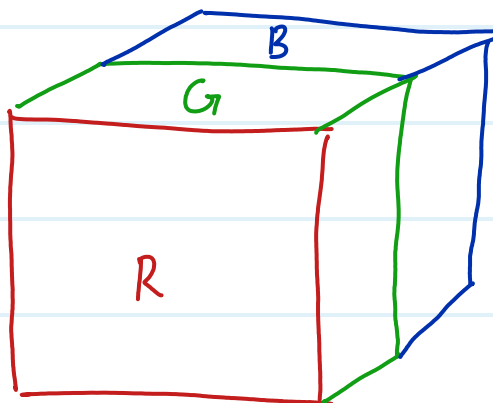
V_3
 V_4
 V_5



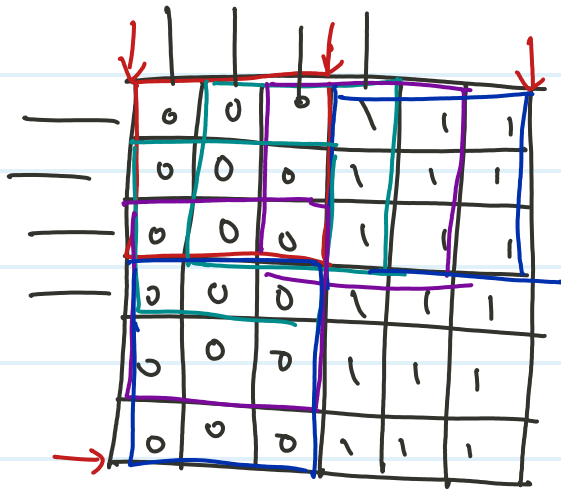
→ RGB

→ Gray scale

→ A - channel for transparency



convolution operation



6 x 6

1/p ↑

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

3 x 3

vertical edge filter

← filter

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

Horizontal edge filter.

stride (shift)

o/p →

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

4 x 4

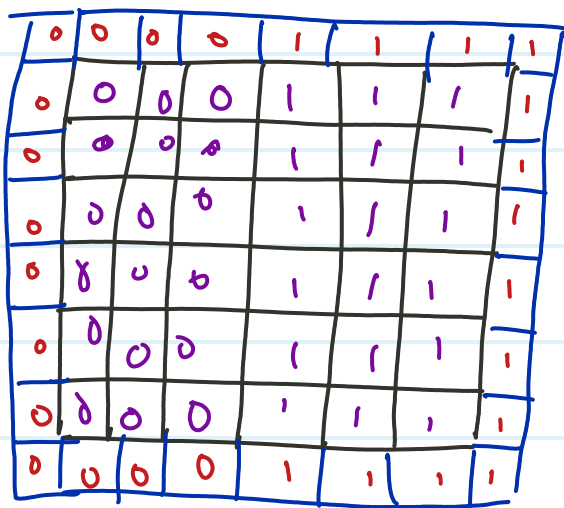
$$(n - f) + 1 =$$

$$6 - 3 + 1 = 4$$

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

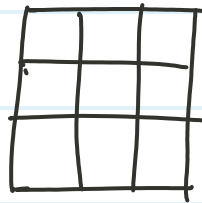
so we are losing some information.

To overcome this we use another technique called padding

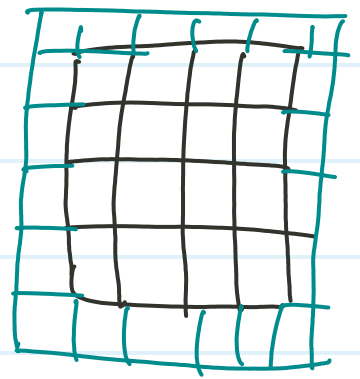


6×6

8×8



3×3



4×4

6×6

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

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

$$2p = 6 - 4$$

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

max pulling

①

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

②

③

④

1	

4	6
6	5

strides jump = 2

min pulling

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

1	1
2	1

Smoothing of image for its use min
and max pooling

mean pulling

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

2.9	2.9
2	2.1