

# Neighbourhood Processing

## Neighbourhood

$(x-1, y+1)$   $(x, y+1)$   $(x+1, y+1)$

$(x-1, y)$   $(x, y)$   $(x+1, y)$

$(x-1, y-1)$   $(x, y-1)$   $(x+1, y-1)$

✳️ 4-neighbours of  $P$ ,  $N_4(P)$

✳️ Diagonal neighbours of  $P$ ,  $N_D(P)$

} All these together called:  
8-neighbors of  $P$ ,  $N_8(P)$

## Distance Measures

$P(x, y)$ ,  $Q(s, t)$

• Euclidean Distance:  $D_e(P, Q) = \sqrt{(x-s)^2 + (y-t)^2}$

•  $D_4$  distance (city block distance):  $D_4(P, Q) = |x-s| + |y-t|$   
↳ Pixels with  $D_4 = 1$

•  $D_8$  distance (chessboard distance):  $D_8(P, Q) = \max(|x-s|, |y-t|)$   
↳ Pixels with  $D_8 = 1$

# Linear vs Non-Linear operations

Conditions of Linear operations:

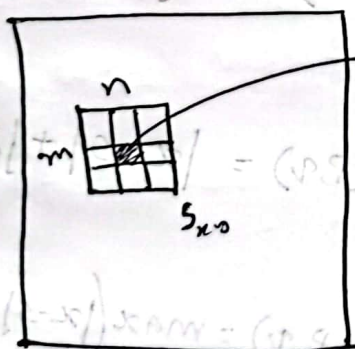
① Additivity property:

$$H[f(x_1, y_1) + f(x_2, y_2)] = H[f(x_1, y_1)] + H[f(x_2, y_2)]$$

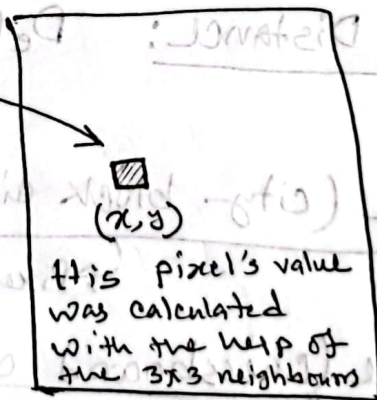
② Homogeneity property:

$$H[a f(x, y)] = a H[f(x, y)]$$

## Neighbourhood operations



Input



This pixel's value was calculated with the help of the 3x3 neighbours

Output



## Neighbourhood Operation : Mechanism

~~\*\*\*~~ Neighbourhood processing is also called Special filtering.

~~\*\*\*~~ Special filtering consists of:

① A neighbourhood (defined by the filter kernel or mask)

② A predefined Operation

→ Linear Operation

→ Non-Linear Operation

~~\*\*\*~~ A processed image is generated as the centers of the filter kernel visits each pixel in the input image.

## Linear Special Filtering

### \*\* Formula:

- Kernel shape :  $m \times n$
- Numbers of pixels padded :  $p$
- Stride :  $s$
- Input image shape :  $M \times N$
- Output image shape :  $H \times W$

$$H = \text{floor} \left( \frac{M + 2p - m}{s} + 1 \right)$$

$$W = \text{floor} \left( \frac{N + 2p - n}{s} + 1 \right)$$

if we want input and output to be same shape:

$$\begin{aligned} \text{No. of padding pixels, } p &= \frac{m-1}{2} \\ &= \frac{n-1}{2} \end{aligned}$$



## Math Demo:

input (6x6)

7	7	6			
7	7	6	5		
6	6	4	3		
	5	3	2		

Kernel (3x3)

0	-1	0
-1	5	-1
0	-1	0

- Kernel shape : 3x3
- $S = 1$
- Padding = Nearest ;  $P = 1$
- input image =  $M \times N$   
6x6
- Out image =  $H \times W$

7.0

7.-1-

6.0

7.-1

7.5

6.-1

6.0

6.-1

4.0

9

$$H = \left\lfloor \frac{M + 2P - m}{S} + 1 \right\rfloor$$

$$= \left\lfloor \frac{6 + 2 \cdot 1 - 3}{1} + 1 \right\rfloor$$

$$= 6$$

$$W = \left\lfloor \frac{N + 2P - n}{S} + 1 \right\rfloor$$

$$= \left\lfloor \frac{6 + 2 \cdot 1 - 3}{1} + 1 \right\rfloor$$

$$= 6$$

So, input (x,y) = 7, output

(x,y) = 9

7	8	7
6	7	6
5	6	5

# Padding

Kernel stride  $\rightarrow$  border pixel problem.

0	1	0
1	2	1
0	1	0

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

Solution:

Solution-①: Don't visit border pixels.  
- image shrinks.

Solution-②: Do Padding

Padding Type:

Linear Padding:

Nearest  $\rightarrow$  then  $\rightarrow$   $\rightarrow$

$\rightarrow 0$

① Zero Padding: extend pixel value = 0

② Constant Padding: image extend pixel value = some constant

④ Mirror Padding: Mirror the edge pixels

③ Clamp / Edge / Nearest: Repeat edge pixels.

Mirror Padding  $\rightarrow$

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

Nearest padding  $\rightarrow$

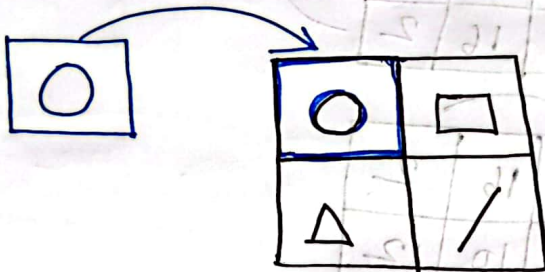
		7	8	4
3	3	3	4	5
2	2	2	9	6
2	2	2	9	
2	2	2	9	



# Correlation

\*\*\* Sliding dot product.

\*\*\* It is the measure of similarity of the two signals as a function of the displacement of one function relative to other.



\*\*\* একটি <sup>filter kernel</sup> weight matrixe থাকে। যেকোনো slide করে এর original image এর উপর। যেখানে যেখানে similarity বরা বরা হয়।

\*\*\* Linear Spatial filtering is basically a 2D signal correlation operation.

# Convolution

**\*\*** If we rotate the original filter  $w(x,y)$   $180^\circ$  and then find the correlation, it's called convolution.

original filter :

4	5	6
9	10	12
13	16	2

$90^\circ$  rotate :

13	16	2
9	10	12
4	5	6

$180^\circ$  rotate :

2	16	13
12	10	9
6	5	4

**\*\*** for symmetric filters convolution and correlation gives same result.



# Smoothing Spatial Filters

- Linear
  - Box / average filtering
  - Gaussian filtering
- Non-Linear
  - Median filtering

## Linear Smoothing Filters

### Average Filter:

→ as pixel average

$$w(x, y) = \frac{1}{mn}$$

$$\frac{1}{9} \times \begin{array}{|c|c|c|} \hline 1 & 1 & 1 \\ \hline 1 & 1 & 1 \\ \hline 1 & 1 & 1 \\ \hline \end{array}$$

### Gaussian Filter:

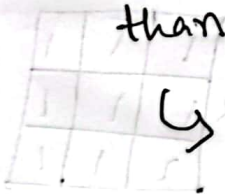
→ weighted

$$w(x, y) = e^{-\frac{x^2 + y^2}{2\sigma^2}}$$

## Applications

- ① Blurring
- ② Sharpening the edges
- ③ Denoising
- ④ Low pass Filtering (getting rid of high variations)

**Ex** For the same kernel size the gaussian blur preserves more information than average filtering.



↳ edge better preserve

$$\frac{y+y}{2}$$

8

7 between

$$= (b \times w)$$



## Non-Linear smoothing filtering / order statistic filtering

✱✱ Kernel ମାତ୍ର ନୁହେଁ

✱✱ neighbourhood ଏବଂ value ଥିବା ସମସ୍ତ  
ଏହା Statistical operation ଚାଲିଥାଏ value  
ପ୍ରତି ଏହା କରାଯାଏ ।

① Median filtering: ଅନ୍ତରାଳର value  
(after sorting)

② Max and Min filtering.

③ Midpoint filtering

④ Alpha-trimmed mean filtering.

# Median filtering

\*\* Gaussian / Box filtering is Smoothing  
but, Median filtering - is also  
smoothing but Median filtering  
better.

Box > Gaussian > Median filtering (1)  
Linear Non-Linear (11)

(111)

(1111)



## Unsharp Masking & High Boost Filtering

① Blur the original image:

$$f'(x,y) = \text{Blur}[f(x,y)]$$

②  $g_{\text{mask}}(x,y) = f(x,y) - f'(x,y)$

③  $g(x,y) = f(x,y) + g_{\text{mask}}(x,y)$

