

# AVD862-Digital Image Processing

## Assignment -1

27-09-2025

SAURABH KUMAR  
SC22B146

①  $L=8, MN=4096$

$r_k$	$n_k$	$C_k$	$S_k = (L-1) \frac{C_k}{MN}$
0	790	790	1.35 $\rightarrow$ 1
1	<del>850</del> 1023	1813	3.09 $\rightarrow$ 3
2	850	2663	4.55 $\rightarrow$ 5
3	656	3319	5.67 $\rightarrow$ 6
4	329	3648	6.23 $\rightarrow$ 6
5	245	3893	6.65 $\rightarrow$ 7
6	122	4015	6.86 $\rightarrow$ 7
7	81	4096	7.00 $\rightarrow$ 7

Resulting Histogram:

0	:	0
1	:	790 (from $n_0$ )
2	:	0
3	:	1023 (from $n_1$ )
4	:	0
5	:	850 (from $n_2$ )
6	:	985 (from $n_3, n_4 : 656 + 329$ )
7	:	448 (from $n_5, n_6, n_7 : 245 + 122 + 81$ )

Total = 4096

Mapping:

0	$\rightarrow$ 1
1	$\rightarrow$ 3
2	$\rightarrow$ 5
3	$\rightarrow$ 6
4	$\rightarrow$ 6
5	$\rightarrow$ 7
6	$\rightarrow$ 7
7	$\rightarrow$ 7

②

$r_k$	$n_k$	Target count ( $\times 4096$ )	$P_r(k)$	$C_k$	$P_z(k) = \frac{C_k}{MN}$
0	0 $\rightarrow$ 0		0	790	0.193
1	0 $\rightarrow$ 0		0	1813	0.443
2	0 $\rightarrow$ 0		0	2663	0.650
3	0.15 $\rightarrow$ 614.4		0.15	3319	0.810
4	0.20 $\rightarrow$ 819.2		0.35	3648	0.891
5	0.30 $\rightarrow$ 1228.8		0.65	3893	0.951
6	0.20 $\rightarrow$ 819.2		0.85	4015	0.981
7	0.15 $\rightarrow$ 614.4		1.00	4096	1.000

For each  $\mathcal{M}_k$ , find smaller  $\mathbb{Z}$  such that  $P_{\mathbb{Z}}(k) \geq P_{\mathbb{Z}_C}(k)$  :  
(Mapping)

$$\mathcal{M}_0 \rightarrow \mathbb{Z}_4$$

$$\mathcal{M}_1 \rightarrow \mathbb{Z}_5$$

$$\mathcal{M}_2 \rightarrow \mathbb{Z}_6$$

$$\mathcal{M}_3 \rightarrow \mathbb{Z}_6$$

$$\mathcal{M}_4 \rightarrow \mathbb{Z}_7$$

$$\mathcal{M}_5 \rightarrow \mathbb{Z}_7$$

$$\mathcal{M}_6 \rightarrow \mathbb{Z}_7$$

$$\mathcal{M}_7 \rightarrow \mathbb{Z}_7$$

③ An ideal low pass filter is a filter that passes all frequencies components,  $|w|$ , less than a cutoff,  $w_c$ , with gain 1, and completely rejects (gain 0) frequencies above  $w_c$ .

$$\text{In ID, } H(w) = \begin{cases} 1, & |w| \leq w_c \\ 0, & |w| > w_c. \end{cases}$$

No, in general, ideal LPF is theoretical with perfect cutoff. Its impulse response in the spatial domain is an infinite sinc function (non-causal).

Impulse response:

$$h(n) = \frac{w_c}{\pi}, \quad n=0$$

$$= \frac{\sin(w_c n)}{\pi n}, \quad n \neq 0.$$

④ Properties of convolution:

Ⓐ Linearity:  $a(f * g) + b(f * h) = f * (ag + bh)$

Ⓑ Commutative:  $f * g = g * f$ .

Ⓒ Associative:  $f * (g * h) = (f * g) * h$

Ⓓ Distributive:  $f * (g + h) = (f * g) + (f * h)$

Ⓔ Shift invariance: If  $f'_x = f(x - x_0)$ , then

Ⓕ Convolution theorem:  $f'_x * g = (f * g)(x - x_0)$

theorem:

Convolution in spatial domain is multiplication in freq domain, and vice-versa.



$$\begin{bmatrix} 0 & \frac{1}{6} & 0 \\ \frac{1}{6} & \frac{1}{3} & \frac{1}{6} \\ 0 & \frac{1}{6} & 0 \end{bmatrix} * \begin{bmatrix} 5 & 15 & 20 & 45 & 50 \\ 10 & 35 & 40 & 60 & 70 \\ 20 & 75 & 100 & 85 & 75 \\ 30 & 40 & 85 & 135 & 80 \\ 40 & 10 & 45 & 50 & 65 \end{bmatrix}$$

Applying the mask on shaded / dotted pixels:

$$\begin{bmatrix} 5 & 15 & 20 & 45 & 50 \\ 10 & 35.83 & 50.83 & 60.83 & 70 \\ 20 & 57.5 & 81.67 & 90 & 75 \\ 30 & 46.67 & 81.67 & 95 & 80 \\ 40 & 10 & 45 & 50 & 65 \end{bmatrix} \approx \begin{bmatrix} 5 & 15 & 20 & 45 & 50 \\ 10 & 36 & 51 & 61 & 70 \\ 20 & 58 & 82 & 90 & 75 \\ 30 & 47 & 82 & 95 & 80 \\ 40 & 10 & 45 & 50 & 65 \end{bmatrix}$$

⑤

$$\begin{bmatrix} 5 & 15 & 20 & 45 & 50 \\ 10 & 35 & 45 & 60 & 70 \\ 20 & 75 & 100 & 85 & 75 \\ 30 & 40 & 85 & 135 & 80 \\ 40 & 10 & 45 & 50 & 65 \end{bmatrix}$$

↓  
sort and  
take median

$$\begin{bmatrix} 5 & 15 & 20 & 45 & 50 \\ 10 & 20 & 45 & 60 & 70 \\ 20 & 40 & 75 & 80 & 75 \\ 30 & 40 & 75 & 80 & 80 \\ 40 & 10 & 45 & 50 & 65 \end{bmatrix}$$

- ⑤ Laplacian is a second-order derivative operator that measures the sum of second-order partial derivatives.

$$\nabla^2 f = \frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2}$$

In image processing, it highlights regions of rapid intensity change, useful for edge detection.

Types of Laplacian operator:

- (i) 4-Neighbour Laplacian:

$$\begin{bmatrix} 0 & 1 & 0 \\ 1 & -4 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

- (ii) 8-Neighbour Laplacian:

$$\begin{bmatrix} 1 & 1 & 1 \\ 1 & -8 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

- (iii) Laplacian of Gaussian (LoG): Gaussian Smoothing followed by Laplacian to detect edges.

Eg. Mexican Hat kernel:

$$\begin{bmatrix} 0 & 0 & -1 & 0 & 0 \\ 0 & -1 & -2 & -1 & 0 \\ -1 & -2 & 16 & -2 & -1 \\ 0 & -1 & -2 & -1 & 0 \\ 0 & 0 & -1 & 0 & 0 \end{bmatrix}$$