

MUZAFFARPUR INSTITUTE OF TECHNOLOGY

B.Tech 6th Semester Mid-Sem Exam., 2019

DIGITAL IMAGE PROCESSING (ECE)

TIME: 2 Hrs

FULL MARKS: 20

Ques. (1): Explain the following term (any three)

(2 x 3 = 6 marks)

(a) Formula of Fourier Transform in two-dimensional space.

$$F(u, v) = \frac{1}{MN} \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) e^{-j2\pi(ux/M + vy/N)}$$

for $u = 0, 1, 2, \dots, M - 1, v = 0, 1, 2, \dots, N - 1$

(b) Image Negatives

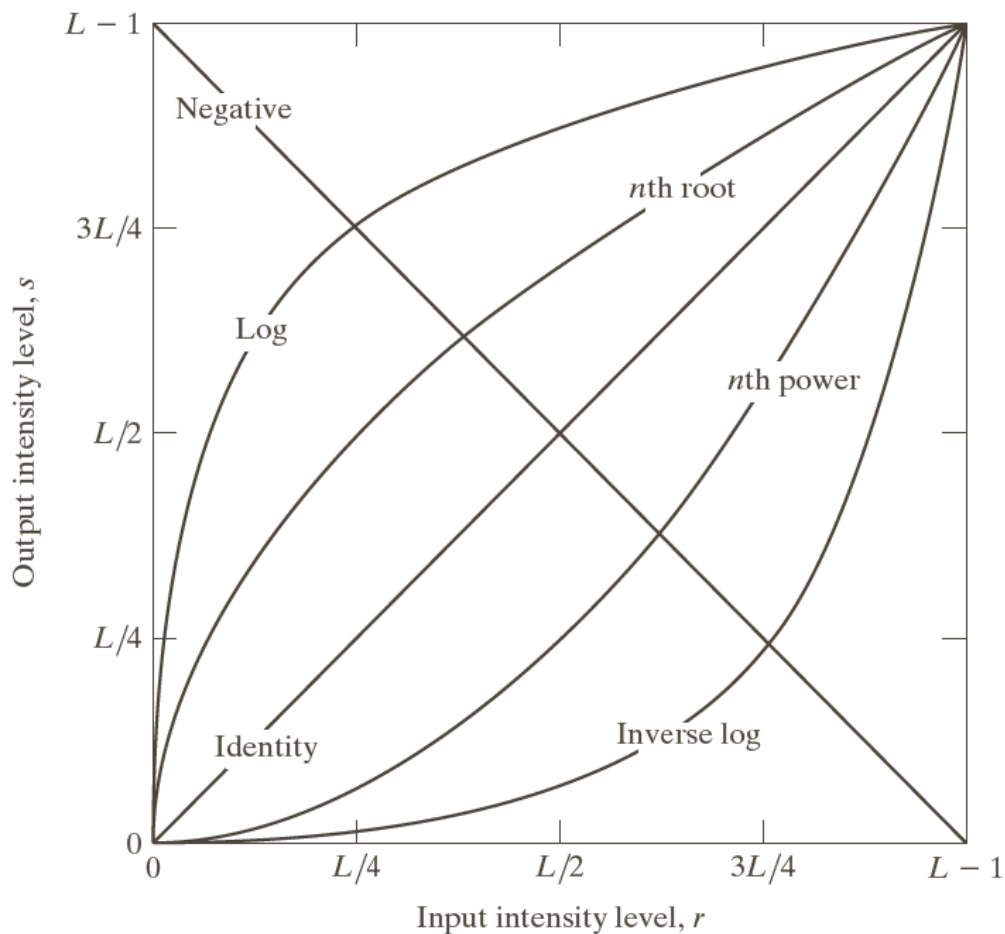


Image negatives

$$s = L - 1 - r$$

(c) Transfer Function of Butterworth High Pass filter

$$|H(u, v)|^2 = \frac{1}{1 + \left[\frac{D_0}{D(u, v)} \right]^{2n}}$$

(d) Smoothing Linear filters


The general implementation for filtering an $M \times N$ image with a weighted averaging filter of size $m \times n$ is given

$$g(x, y) = \frac{\sum_{s=-a}^a \sum_{t=-b}^b w(s, t) f(x+s, y+t)}{\sum_{s=-a}^a \sum_{t=-b}^b w(s, t)}$$


where $m = 2a + 1$, $n = 2b + 1$.

(e) Laplacian in frequency domain

$$\mathfrak{F}\left[\frac{\partial^2 f(x, y)}{\partial x^2} + \frac{\partial^2 f(x, y)}{\partial y^2}\right] = \boxed{-(u^2 + v^2)} F(u, v)$$



$$H_1(u, v) = -(u^2 + v^2)$$

 **Frequency domain**

$$\nabla^2 f = \frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2} \xrightarrow{\text{Spatial domain}} \text{Laplacian operator}$$

Ques.(2): Draw the block diagram of homomorphic filtering approach for image enhancement and explain it.

Homomorphic Filtering

$$\ln : \quad z(x, y) = \ln f(x, y) = \ln i(x, y) + \ln r(x, y)$$

$$\text{DFT} : \quad Z(u, v) = F_i(u, v) + F_r(u, v)$$

$$H(u, v) : \quad S(u, v) = H(u, v)Z(u, v)$$

$$(\text{DFT})^{-1} : \quad s(x, y) = i'(x, y) + r'(x, y)$$

$$\text{exp} : \quad g(x, y) = e^{s(x, y)} = i_0(x, y)r_0(x, y)$$

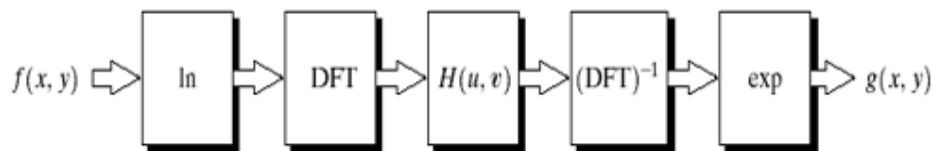
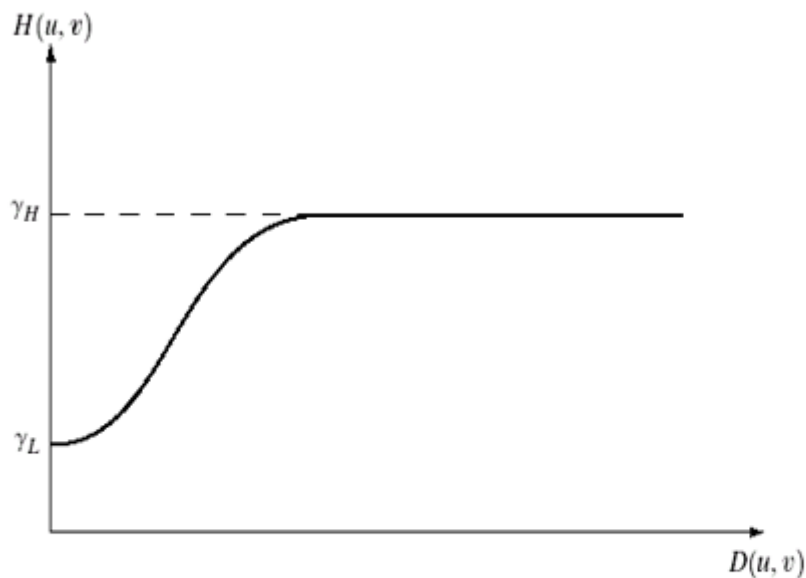


FIGURE 4.31
Homomorphic filtering approach for image enhancement.

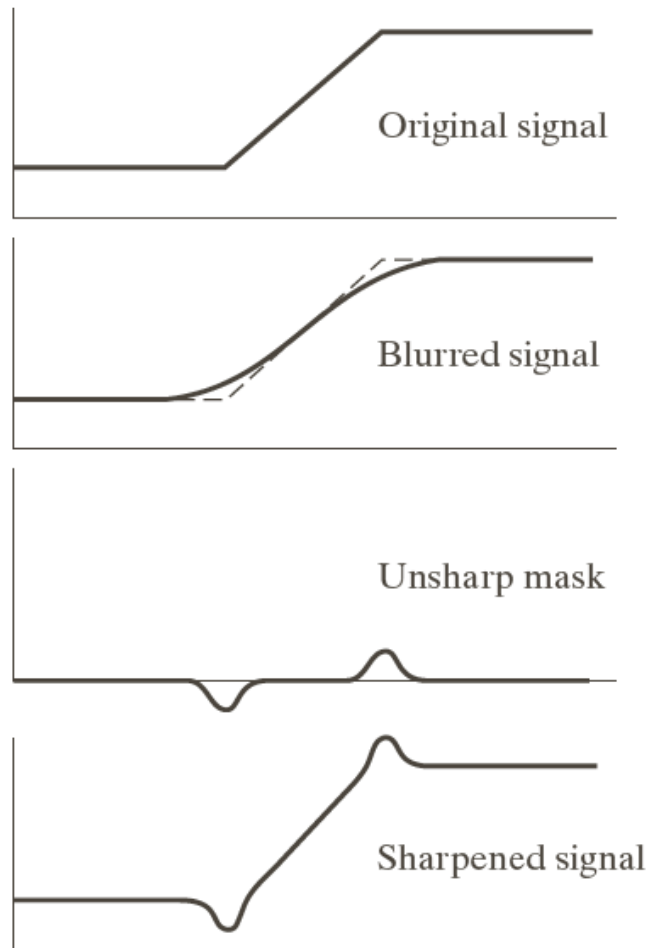


$$H(u, v) = (\gamma_H - \gamma_L) \left[1 - e^{-c \left(\frac{D^2(u, v)}{D_0^2} \right)} \right] + \gamma_L$$

(OR)

Discuss high boost filtering for sharpening and Butterworth low pass filter for smoothing.

(5 marks)



Let $\bar{f}(x, y)$ denote the blurred image, unsharp masking is

$$g_{mask}(x, y) = f(x, y) - \bar{f}(x, y)$$

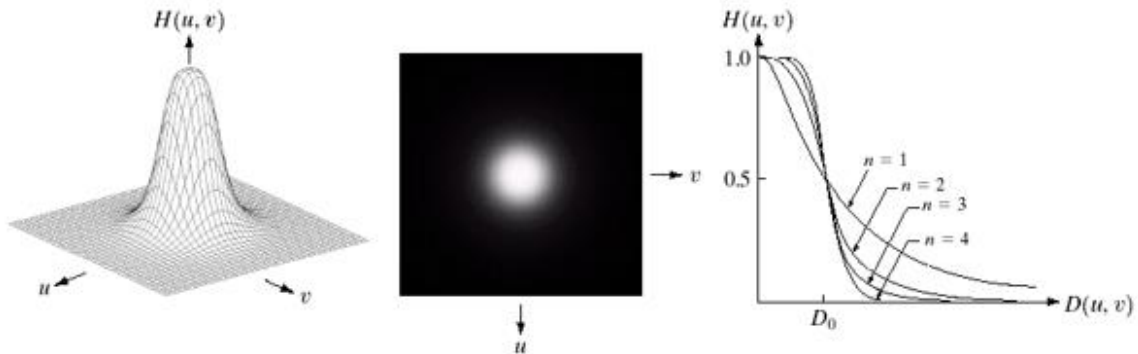
Then add a weighted portion of the mask back to the original

$$g(x, y) = f(x, y) + k * g_{mask}(x, y) \quad k \geq 0$$

when $k > 1$, the process is referred to as highboost filtering.



$$H(u, v) = \frac{1}{1 + [D(u, v) / D_0]^{2n}}$$



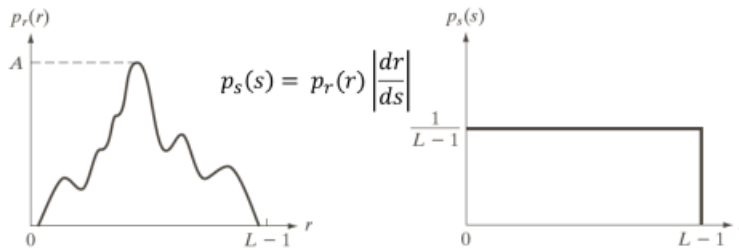
Ques.(3): Explain Histogram Equalization

(5 marks)

Histogram Equalization

The intensity levels in an image may be viewed as random variables in the interval $[0, L-1]$.

Let $p_r(r)$ and $p_s(s)$ denote the probability density function (PDF) of random variables r and s .



a b

FIGURE 3.18 (a) An arbitrary PDF. (b) Result of applying the transformation in Eq. (3.3-4) to all intensity levels, r . The resulting intensities, s , have a uniform PDF, independently of the form of the PDF of the r 's.

Histogram Equalization

$$s = T(r) \quad 0 \leq r \leq L-1$$

- $T(r)$ is a strictly monotonically increasing function in the interval $0 \leq r \leq L-1$;
- $0 \leq T(r) \leq L-1$ for $0 \leq r \leq L-1$.

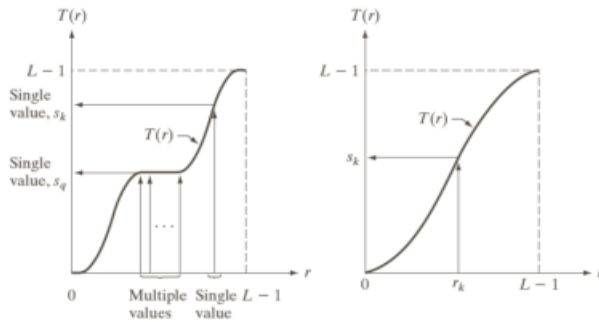


FIGURE 3.17
 (a) Monotonically increasing function, showing how multiple values can map to a single value.
 (b) Strictly monotonically increasing function. This is a one-to-one mapping, both ways.

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Histogram Equalization

$$s = T(r) = (L-1) \int_0^r p_r(w) dw$$

$$\begin{aligned} \frac{ds}{dr} &= \frac{dT(r)}{dr} = (L-1) \frac{d}{dr} \left[\int_0^r p_r(w) dw \right] \\ &= (L-1) p_r(r) \end{aligned}$$

$$p_s(s) = \frac{p_r(r) dr}{ds} = p_r(r) / \left(\frac{ds}{dr} \right) = p_r(r) / ((L-1) p_r(r)) = \frac{1}{L-1}$$

Histogram Equalization

Continuous case:

$$s = T(r) = (L-1) \int_0^r p_r(w) dw$$

Discrete values:

$$\begin{aligned} s_k = T(r_k) &= (L-1) \sum_{j=0}^k p_r(r_j) \\ &= (L-1) \sum_{j=0}^k \frac{n_j}{MN} = \frac{L-1}{MN} \sum_{j=0}^k n_j \quad k=0,1,\dots, L-1 \end{aligned}$$

Ques.(4): Explain image sharpening using Laplace Operator in both spatial and frequency domain

(4 Marks)

Sharpening Spatial Filters: Laplace Operator

The second-order isotropic derivative operator is the Laplacian for a function (image) $f(x,y)$

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

$$\frac{\partial^2 f}{\partial x^2} = f(x+1, y) + f(x-1, y) - 2f(x, y)$$

$$\frac{\partial^2 f}{\partial y^2} = f(x, y+1) + f(x, y-1) - 2f(x, y)$$

$$\begin{aligned} \nabla^2 f &= f(x+1, y) + f(x-1, y) + f(x, y+1) + f(x, y-1) \\ &\quad - 4f(x, y) \end{aligned}$$

Sharpening Spatial Filters: Laplace Operator

0	1	0	1	1	1
1	-4	1	1	-8	1
0	1	0	1	1	1
0	-1	0	-1	-1	-1
-1	4	-1	-1	8	-1
0	-1	0	-1	-1	-1

a b
c d

FIGURE 3.37
 (a) Filter mask used to implement Eq. (3.6-6).
 (b) Mask used to implement an extension of this equation that includes the diagonal terms.
 (c) and (d) Two other implementations of the Laplacian found frequently in practice.

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Sharpening Spatial Filters: Laplace Operator

Image sharpening in the way of using the Laplacian:

$$g(x, y) = f(x, y) + c[\nabla^2 f(x, y)]$$

where,

$f(x, y)$ is input image,

$g(x, y)$ is sharpened images,

$c = -1$ if $\nabla^2 f(x, y)$ corresponding to Fig. 3.37(a) or (b)

and $c = 1$ if either of the other two filters is used.

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Laplacian in Frequency Domain

$$\mathfrak{F}\left[\frac{\partial^2 f(x, y)}{\partial x^2} + \frac{\partial^2 f(x, y)}{\partial y^2}\right] = \boxed{-(u^2 + v^2)} F(u, v)$$

$H_1(u, v) = -(u^2 + v^2)$

↓ Frequency domain
↓ Laplacian operator

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

(OR)

Consider the image segment shown

(a) Let $V = \{0, 1\}$ and compute the lengths of the shortest 4-, 8-, and m-path between p and q. If a particular path does not exist between these two points, explain why.

(b) Repeat for $V = \{1, 2\}$.

3	1	2	1 (q)
2	2	0	2
1	2	1	1
1 (p)	0	1	2