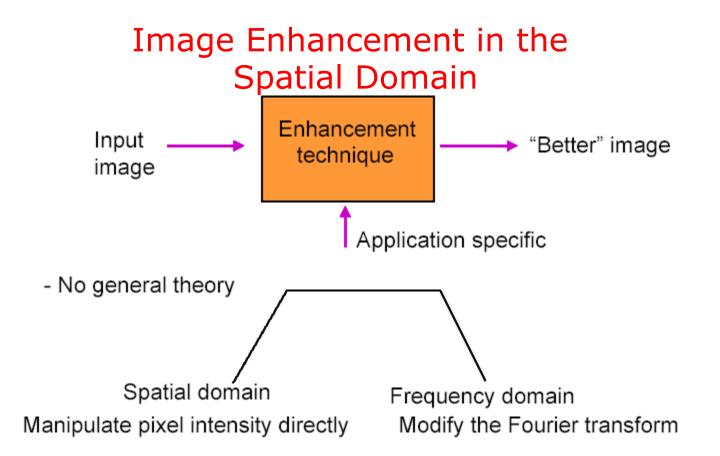
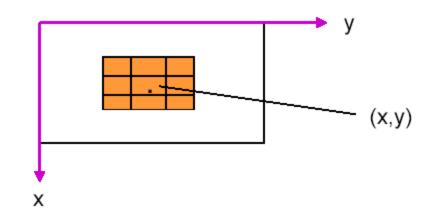
Dr. Pushpa Mamoria Senior Lecturer Dept. of Computer Application UIET, CSJM University, Kanpur http://csjmu.ac.in/all-faculty/dr-pushpa-mamoria/ https://vidwan.inflibnet.ac.in/profile/226676

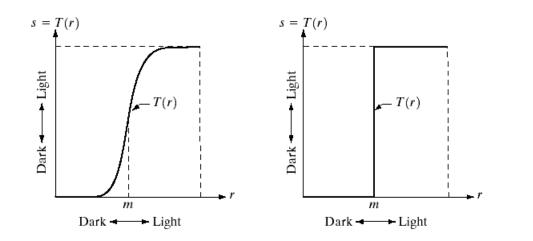


g(x,y) = T[f(x,y)]



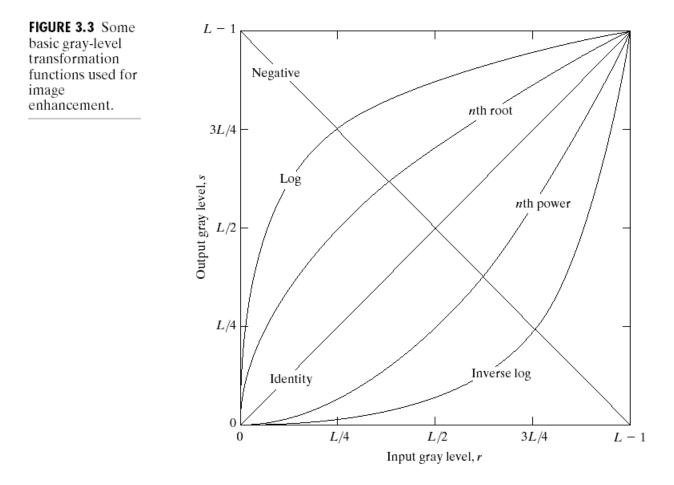
Simplest case: Neighbourhood is (x,y) [g(.) depends only on the value of f at (x,y)]

Gray Level Transformation Functions

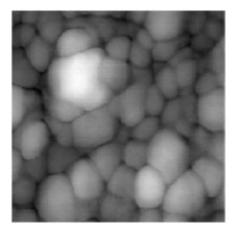


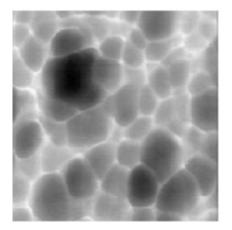


Gray Level Transformation Functions



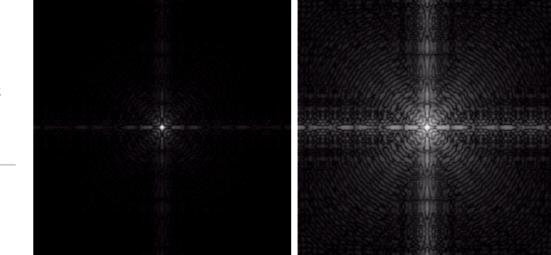
(a) Negative image: Example: g(x,y) = 255 - f(x,y)





(c) Compressing dynamic range s = c log (1 + |r|) c → Scaling factor

Example: Displaying the Fourier Spectrum



a b

FIGURE 3.5 (a) Fourier spectrum. (b) Result of applying the log transformation given in Eq. (3.2-2) with c = 1.

Power Function

$$s = cr^{\gamma}$$

C and \gamma are positive constants.

Often referred to as "gamma correction".

CRT –intensity-to-voltage response follows a power function (typical value of gamma in the range 1.5-2.5.)

Power Function

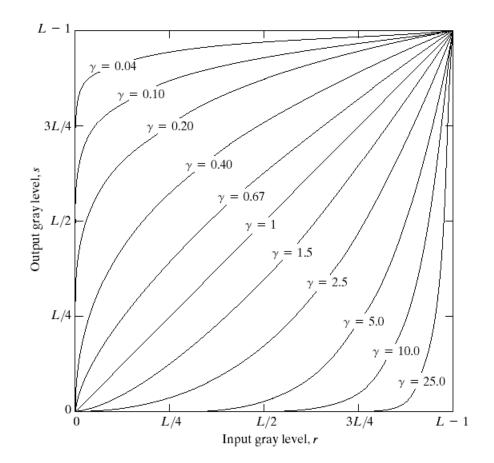
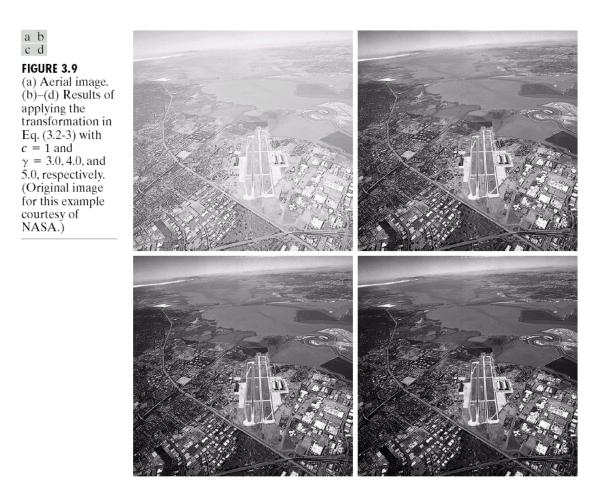
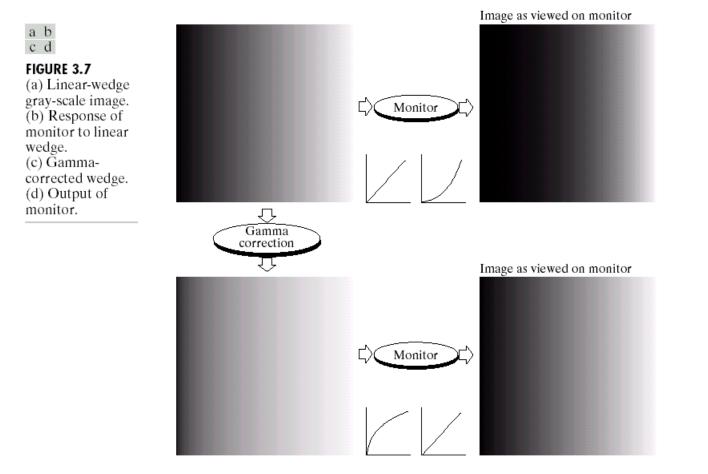


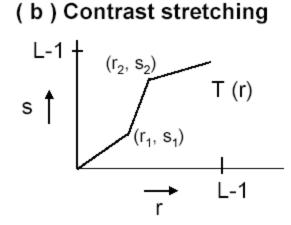
FIGURE 3.6 Plots of the equation $s = cr^{\gamma}$ for various values of γ (c = 1 in all cases).

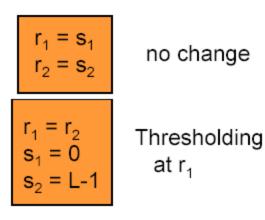
Power Function: Example



Power Function: Gamma Correction







no change

at r₁

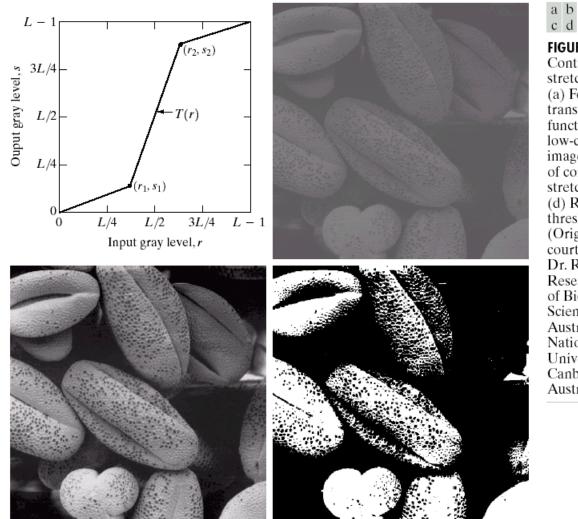
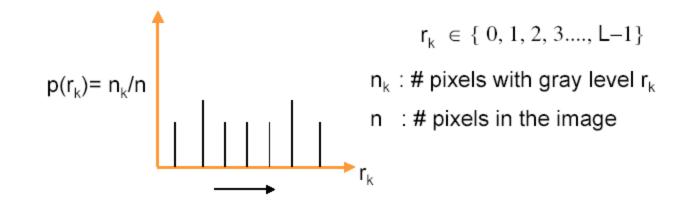
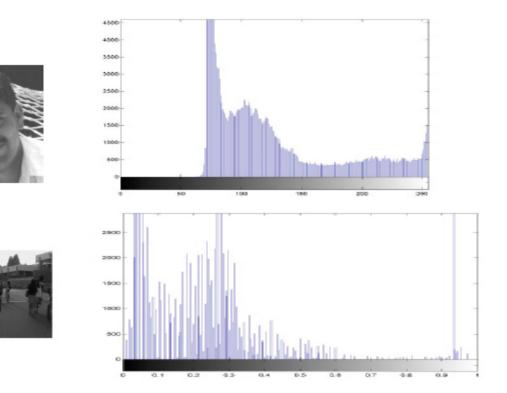


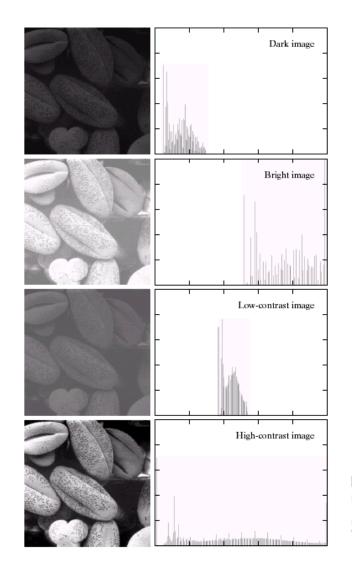
FIGURE 3.10 Contrast stretching. (a) Form of transformation function. (b) A low-contrast image. (c) Result of contrast stretching. (d) Result of thresholding. (Original image courtesy of Dr. Roger Heady, Research School of Biological Sciences, Australian National University, Canberra, Australia.)

Histogram



Histogram: Examples





Histogram: Examples



FIGURE 3.15 Four basic image types: dark, light, low contrast, high contrast, and their corresponding histograms. (Original image courtesy of Dr. Roger Heady, Research School of Biological Sciences, Australian National University, Canberra, Australia.)

Histogram Equalization

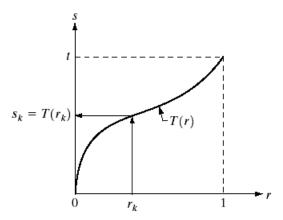


FIGURE 3.16 A

gray-level transformation function that is both single valued and monotonically increasing.

Histogram Equalization

(i) T(r) is single valued valued and monotonically increasing in 0 ≤ r ≤ 1
(ii) 0 ≤ T(r) ≤ 1 for 0 ≤ r ≤ 1
[0, 1] - T → [0, 1]

Inverse transformation : $T^{-1}(s) = r$ $0 \le s \le 1$

 $T^{-1}(s)$ also satisfies (i) and (ii)

The gray levels in the image can be viewed as random variables taking values in the range [0,1].

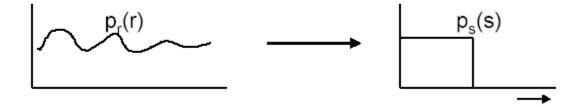
Let $p_r(r)$: p.d.f. of input level r and let $p_s(s)$: p.d.f. of s

Histogram Equalization

- r : Input gray level \in [0, 1]
- s : Transformed gray level \in [0, 1]

s = T (r) T : Transformation function

We are interested in obtaining a transformation function T() which transforms an arbitrary p.d.f. to an uniform distribution



Histogram Equalization

Consider
$$s = T(r) = \int_{0}^{r} p_{r}(w) dw \quad 0 \le r \le 1$$

(Cumulative distribution function of r)

$$p_{s}(s) = p_{r}(r) \frac{dr}{ds}\Big|_{r=T^{-1}(s)};$$

$$\frac{ds}{dr} = \frac{d}{dr} \left[\int_{r}^{0} p_{r}(w) dw \right] = p_{r}(r)$$

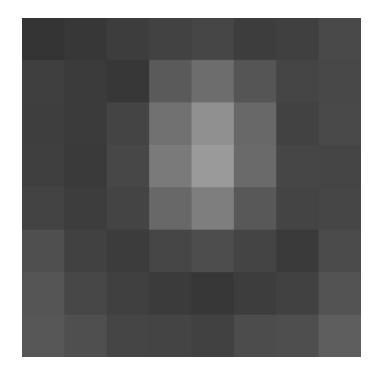
$$\therefore p_{s}(s) = p_{r}(r) \frac{1}{p_{r}(r)} \Big|_{r=T^{-1}(s)} \equiv 1 \qquad 0 \le s \le 1$$

$$p_r(r_k) = \frac{n_k}{n} \qquad 0 \le r_k \le 1 \quad ; \quad k = 0, 1, \dots, L - 1$$

$$L \rightarrow \text{Number of levels}$$

$$s_k = T(r_k) = \sum_{j=0}^k p_r(r_j) = \sum_{j=0}^k \frac{n_j}{n}$$

| 52 | 55 | 61 | 66 | 70 | 61 | 64 | 73 |
|----|----|----|-----|-----|-----|----|----|
| 63 | 59 | 55 | 90 | 109 | 85 | 69 | 72 |
| 62 | 59 | 68 | 113 | 144 | 104 | 66 | 73 |
| 63 | 58 | 71 | 122 | 154 | 106 | 70 | 69 |
| 67 | 61 | 68 | 104 | 126 | 88 | 68 | 70 |
| 79 | 65 | 60 | 70 | 77 | 68 | 58 | 75 |
| 85 | 71 | 64 | 59 | 55 | 61 | 65 | 83 |
| 87 | 79 | 69 | 68 | 65 | 76 | 78 | 94 |

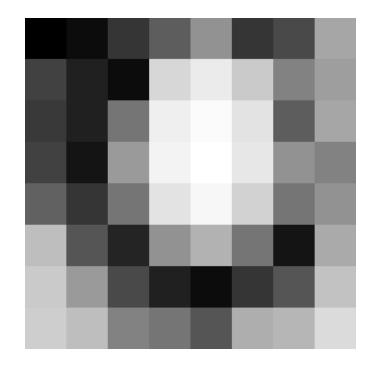


| 52 | 1 | 52 | 1 |
|-----|---|-----|----|
| 55 | 3 | 55 | 4 |
| 58 | 2 | 58 | 6 |
| 59 | 3 | 59 | 9 |
| 60 | 1 | 60 | 10 |
| 61 | 4 | 61 | 14 |
| • | | | |
| • | | | |
| • | | | |
| 144 | 1 | 144 | 63 |
| 154 | 1 | 154 | 64 |

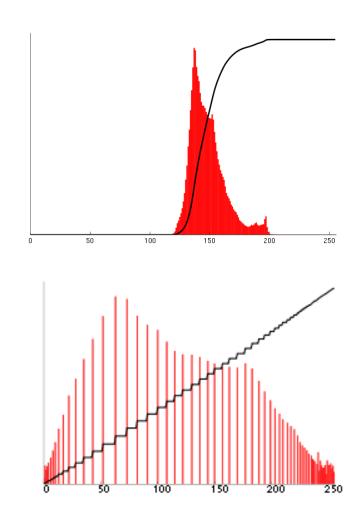
$$h(v) = \operatorname{round}\left(\frac{cdf(v) - cdf_{min}}{(M \times N) - cdf_{min}} \times (L-1)\right)$$

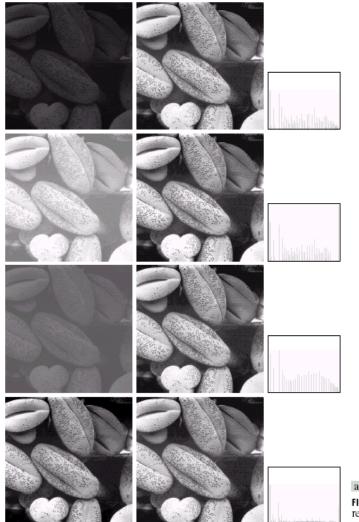
$$h(v) = \operatorname{round}\left(\frac{\operatorname{cdf}(v) - 1}{63} \times 255\right)$$

| 0 | 12 | 53 | 93 | 146 | 53 | 73 | 166 |
|-----|-----|-----|-----|-----|-----|-----|-----|
| 65 | 32 | 12 | 215 | 235 | 202 | 130 | 158 |
| 57 | 32 | 117 | 239 | 251 | 227 | 93 | 166 |
| 65 | 20 | 154 | 243 | 255 | 231 | 146 | 130 |
| 97 | 53 | 117 | 227 | 247 | 210 | 117 | 146 |
| 190 | 85 | 36 | 146 | 178 | 117 | 20 | 170 |
| 202 | 154 | 73 | 32 | 12 | 53 | 85 | 194 |
| 206 | 190 | 130 | 117 | 85 | 174 | 182 | 219 |









Histogram Equalization

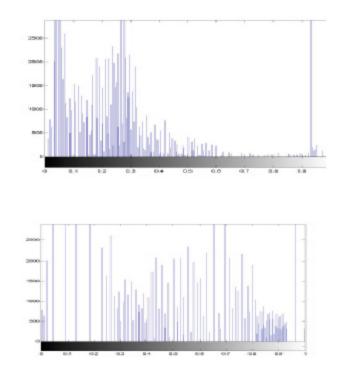
: Examples

a b c FIGURE 3.17 (a) Images from Fig. 3.15. (b) Results of histogram equalization. (c) Corresponding histograms

Histogram Equalization: Examples





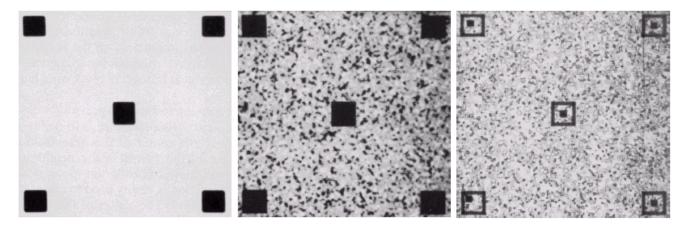


Histogram Specification

Suppose $s = T(r) = \int_{0}^{r} p_{r}(w) dw$ $p_{r}(r) \rightarrow \text{Original histogram} \quad ; \quad p_{z}(z) \rightarrow \text{Desired histogram}$ Let $v = G(z) = \int_{0}^{z} p_{z}(w) dw$ and $z = G^{-1}(v)$ But *s* and *v* are identical p.d.f.

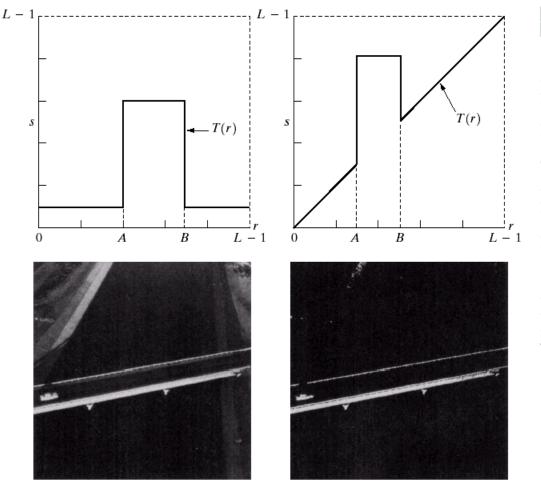
$$\therefore z = G^{-1}(v) = G^{-1}(s) = G^{-1}(T(r))$$

Histogram Equalization: Local Enhancement



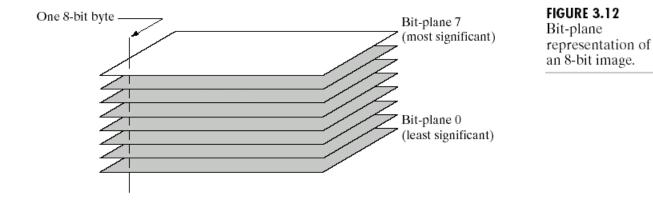
a b c

FIGURE 3.23 (a) Original image. (b) Result of global histogram equalization. (c) Result of local histogram equalization using a 7×7 neighborhood about each pixel.



a b c d FIGURE 3.11 (a) This transformation highlights range [A, B] of gray levels and reduce

[A, B] of gray levels and reduces all others to a constant level. (b) This transformation highlights range [A, B] but preserves all other levels. (c) An image. (d) Result of using the transformation in (a).



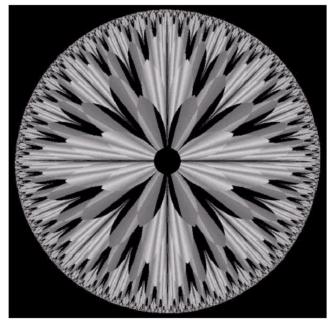


FIGURE 3.13 An 8-bit fractal image. (A fractal is an image generated from mathematical expressions). (Courtesy of Ms. Melissa D. Binde, Swarthmore College, Swarthmore, PA.)

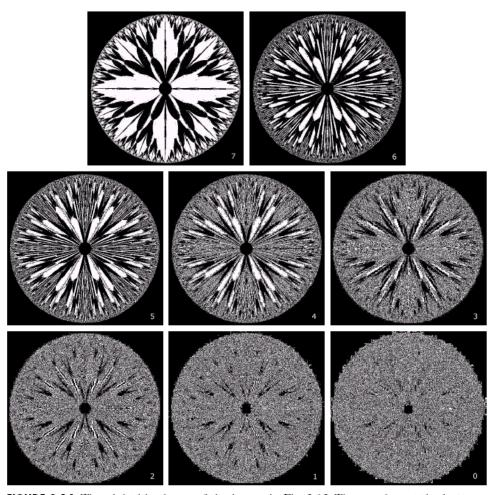
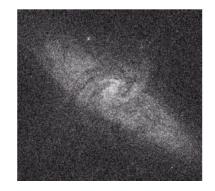


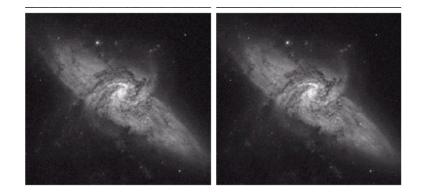
FIGURE 3.14 The eight bit planes of the image in Fig. 3.13. The number at the bottom, right of each image identifies the bit plane.

Image Subtraction

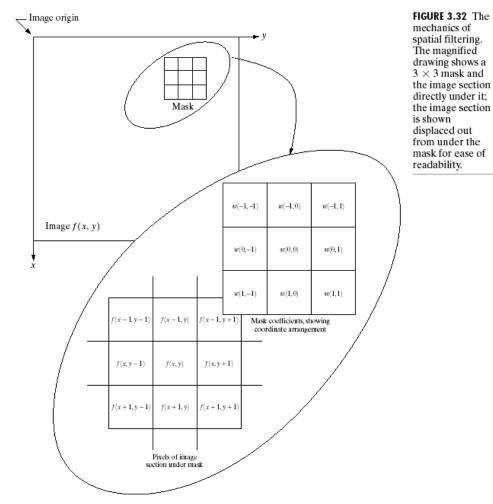




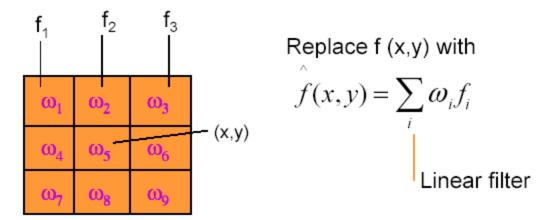




Spatial Filtering

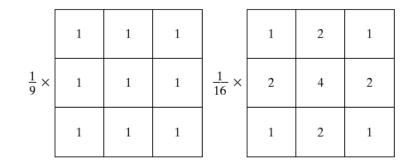


Spatial Filtering



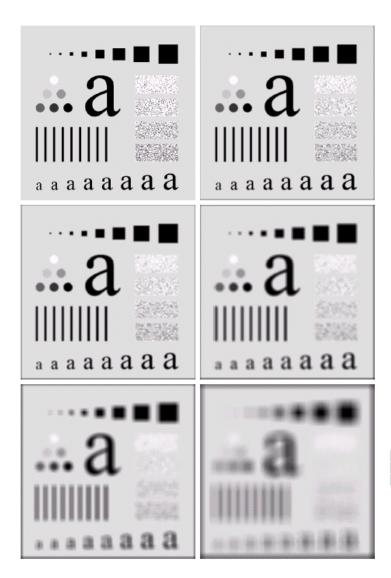
- LPF: reduces additive noise→ blurs the image
 - → sharpness details are lost
 - (Example: Local averaging)

Spatial Filtering: Neighborhood Averaging



a b

FIGURE 3.34 Two 3×3 smoothing (averaging) filter masks. The constant multipli er in front of each mask is equal to the sum of the values of its coefficients, as is required to compute an average.



Spatial Filtering: Neighborhood Averaging

a b FIGURE 3.35 (a) Original image, of size 500×500 pixels. (b)–(f) Results of smoothing with square averaging filter masks of sizes n = 3, 5, 9, 15, and 35, respectively. The black squares at the top are of sizes 3, 5, 9, 15, 25, 35, 45, and 55 pixels, respectively; their borders are 25 pixels apart. The letters at the bottom range in size from 10 to 24 points, in increments of 2 points; the large letter at the top is 60 points. The vertical bars are 5 pixels wide and 100 pixels high; their separation is 20 pixels. The diameter of the circles is 25 pixels, and their borders are 15 pixels apart; their gray levels range from 0% to 100% black in increments of 20%. The background of the image is 10% black. The noisy rectangles are of size 50×120 pixels.

Median Filter

Replace f (x,y) with median [f (x', y')] (x', y') £ neighbourhood

- · Useful in eliminating intensity spikes. (salt & pepper noise)
- · Better at preserving edges.

Example:

| 10 | 20 | 20 |
|----|----|-----|
| 20 | 15 | 20 |
| 25 | 20 | 100 |

→ (10,15,20,20,20,20,20,25,100)

Median=20

So replace (15) with (20)

Median Filter: Example









Sharpening Filter

- · Enhance finer image details (such as edges)
- Detect region /object boundaries.

Example:



Highboost Filter

| 0 | -1 | 0 | -1 | -1 | -1 |
|----|--------------|----|----|-------|----|
| -1 | <i>A</i> + 4 | -1 | -1 | A + 8 | -1 |
| 0 | -1 | 0 | -1 | -1 | -1 |

a b

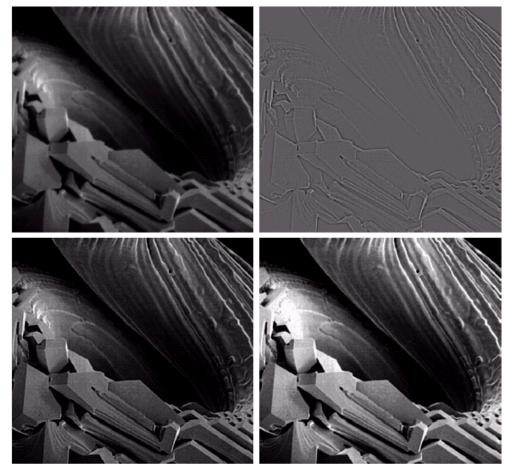
FIGURE 3.42 The high-boost filtering technique can be implemented with either one of these masks, with $A \ge 1$.

Highboost Filter: Example

a b c d

FIGURE 3.43

(a) Same as Fig. 3.41(c), but darker. (a) Laplacian of (a) computed with the mask in Fig. 3.42(b) using A = 0. (c) Laplacian enhanced image using the mask in Fig. 3.42(b) with A = 1. (d) Same as (c), but using A = 1.7.

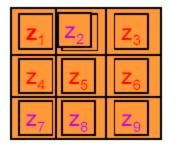


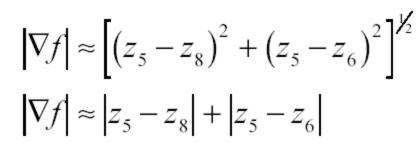
Gradient Filter

Gradient

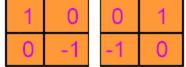
$$\nabla f = \left[\frac{\partial f}{\partial x} \quad \frac{\partial f}{\partial y}\right]^T$$
$$\|\nabla f\| = \left[\left(\frac{\partial f}{\partial x}\right)^2 + \left(\frac{\partial f}{\partial y}\right)^2\right]^{\frac{1}{2}}$$

Gradient Filter



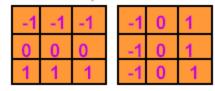


Robert's operator



 $|z_5 - z_9| |z_6 - z_8|$

prewitt



Sobel's

| -1 | -2 | -1 | -1 | 0 | 1 |
|----|----|----|----|---|---|
| 0 | 0 | 0 | -2 | 0 | 2 |
| 1 | 2 | 1 | 5 | 0 | 1 |