

Averaging

$$g(x, y) = f(x, y) + \eta(x, y)$$

$$\bar{g}(x, y) = \frac{1}{M} \sum_{i=1}^M g_i(x, y)$$

$$E(\bar{g}(x, y)) = f(x, y) \text{ and } \sigma_g^2 = \frac{1}{M} \sigma_\eta^2(x, y)$$

$\eta(x, y) \rightarrow$ Uncorrelated zero mean

$\sigma_\eta^2(x, y) \rightarrow$ Reduces the noise variance

Fig 3.30

Spatial Filtering

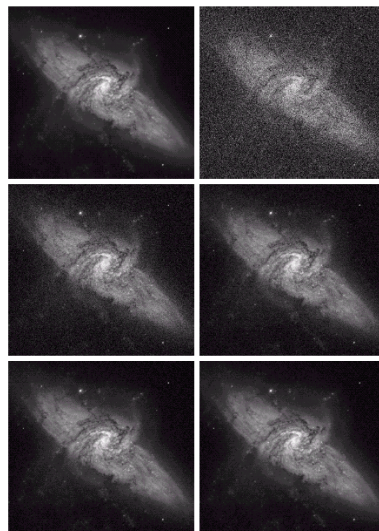
Chapter 3

Required Reading: All sections except 3.8

Image Forensics Project Guidelines

- Process
- Images

Fig 3.30



a b
c d
e f

FIGURE 3.30 (a) Image of Galaxy Pair NGC 3314. (b) Image corrupted by additive Gaussian noise with zero mean and a standard deviation of 64 gray levels. (c)–(f) Results of averaging $K = 8, 16, 64,$ and 128 noisy images. (Original image courtesy of NASA.)

Another example



Images with additive
Gaussian Noise;
Independent
Samples.

```
I=imnoise(J,'Gaussian');
```

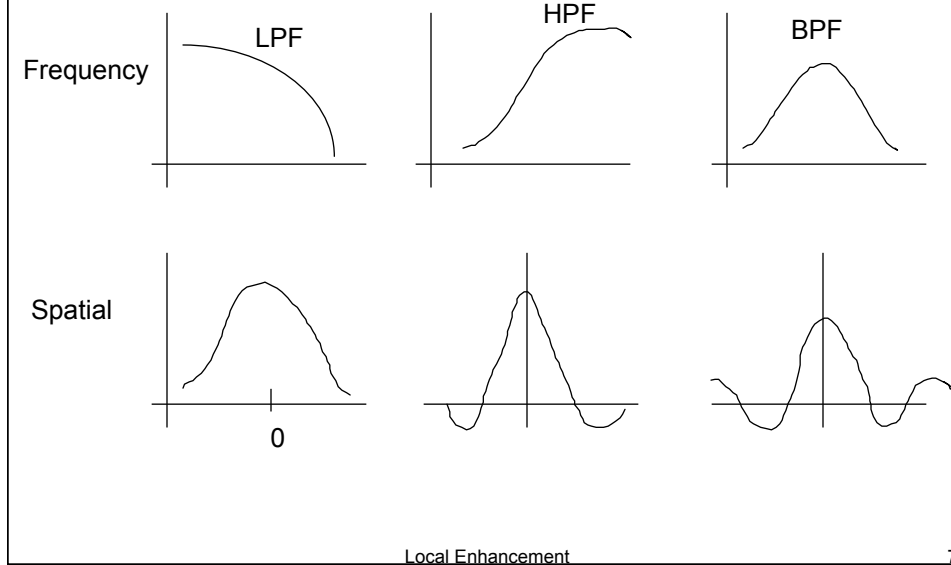


Averaged image

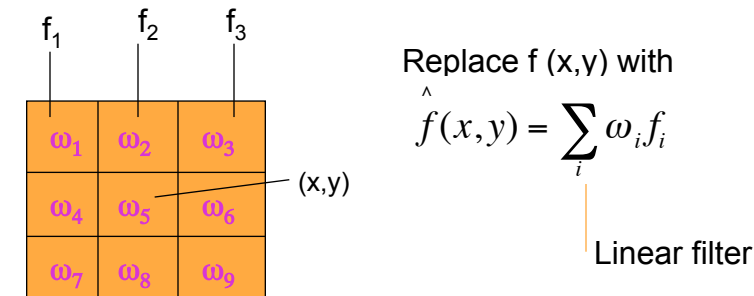


Left: averaged image (10 samples);
Right: original image

Spatial filtering



Smoothing (Low Pass) Filtering



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

Fig 3.35

Fig 3.35: smoothing

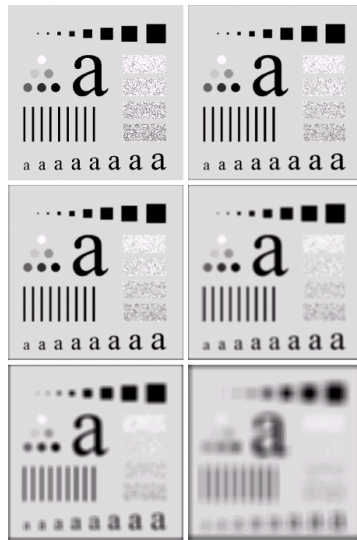
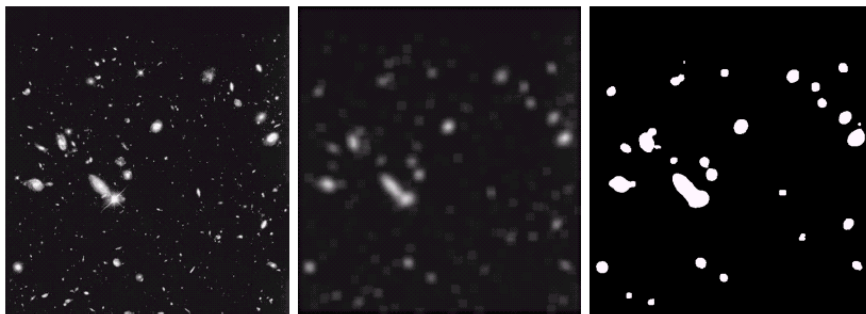


FIGURE 3.35 (a) Original image, of size 500×500 pixels. (b)–(f) Results of smoothing with square averaging filter masks of sizes $n = 3, 5, 9, 15, 25, 35, 45,$ and 55 pixels, respectively. The black squares at the top are of sizes 3, 5, 9, 15, 25, 35, 45, and 55 pixels, respectively; their borders are 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.

Local Enhancement

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Fig 3.36: another example



a b c

FIGURE 3.36 (a) Image from the Hubble Space Telescope. (b) Image processed by a 15×15 averaging mask. (c) Result of thresholding (b). (Original image courtesy of NASA.)

Local Enhancement

10

Image Dithering

- Dithering: to produce visually pleasing signals from heavily quantized data.
 - Halftoning: convert a gray scale image to a binary image by thresholding.
 - Dithering to “add” noise so that the resulting image is smoother than just thresholding (but still it is a binary image)
 - Your homework #4 explores this further with a MATLAB exercise.

Median filtering

Replace $f(x,y)$ with $\text{median}[f(x',y')]$
 $(x',y') \in \mathcal{N}$ 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: Root Signal

Repeated applications of median filter to a signal results in an invariant signal called the “root signal”.

A root signal is invariant to further application of the median filter.

Example: 1-D signal: Median filter length = 3

```
0 0 0 1 2 1 2 1 2 1 0 0 0
0 0 0 1 1 2 1 2 1 1 0 0 0
0 0 0 1 1 1 2 1 1 1 0 0 0
0 0 0 1 1 1 1 1 1 1 0 0 0  root signal
```

Invariant Signals

Invariant signals to a median filter:

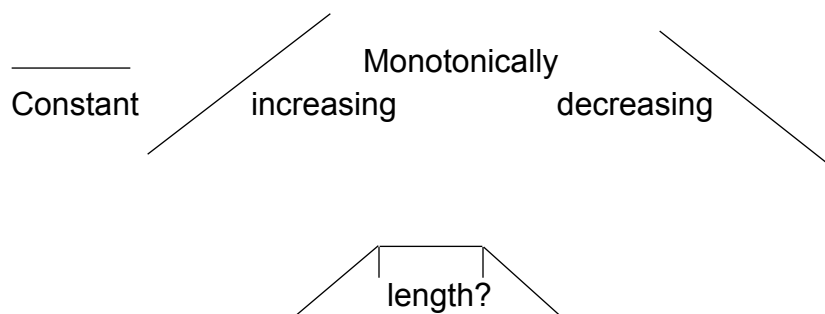
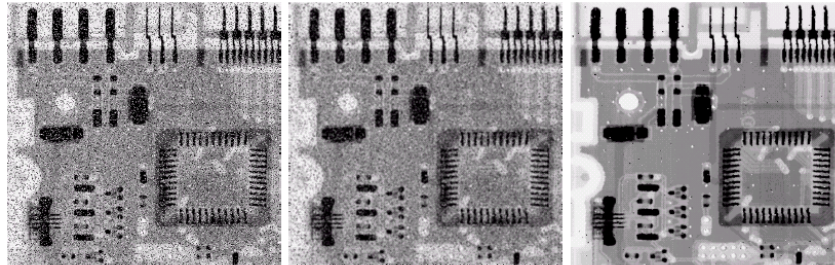


Fig 3.37: Median Filtering example



a b c

FIGURE 3.37 (a) X-ray image of circuit board corrupted by salt-and-pepper noise. (b) Noise reduction with a 3×3 averaging mask. (c) Noise reduction with a 3×3 median filter. (Original image courtesy of Mr. Joseph E. Pascente, Lixi, Inc.)

Media Filter: another example



Original and with salt & pepper noise
`imnoise(image, 'salt & pepper');`

Donoised images



Local averaging
 $K = \text{filter2}(\text{fspecial}(\text{'average'}, 3), \text{image}) / 255.$



Median filtered
 $L = \text{medfil2}(\text{image}, [3 \ 3]);$

Sharpening Filters

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

Example:

-1	-1	-1
-1	8	-1
-1	-1	-1

Unsharp Masking

Subtract Low pass filtered version from the original
emphasizes high frequency information

$I' = A$ (Original) - Low pass

HP = O - LP $A > 1$

$I' = (A - 1) O + HP$

$A = 1 \Rightarrow I' = HP$

$A > 1 \Rightarrow$ LF components added back.

Derivative Filters

1/9

-1	-1	-1
-1	0	-1
-1	-1	-1

Gradient

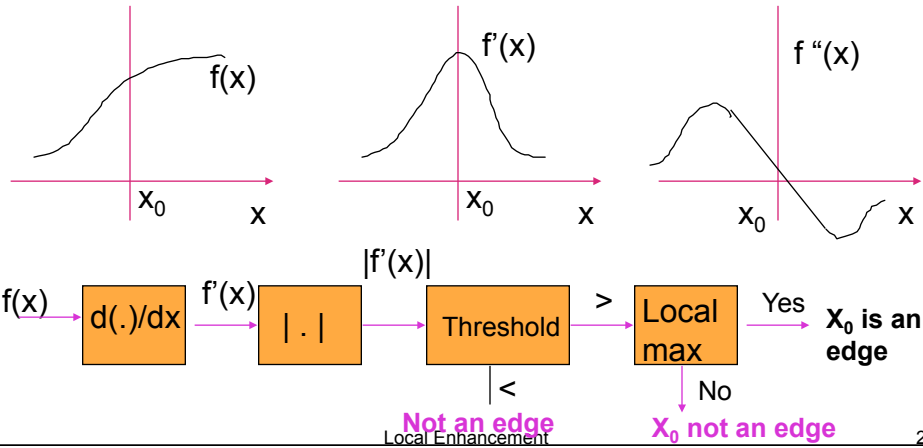
$$\nabla f = \left[\frac{\partial f}{\partial x} \quad \frac{\partial f}{\partial y} \right]^T$$

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

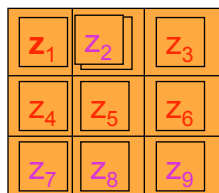
Edge Detection

Gradient based methods

$$\nabla f = \begin{pmatrix} \frac{\partial f}{\partial x} & \frac{\partial f}{\partial y} \end{pmatrix}^T$$



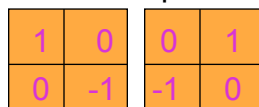
Digital edge detectors



$$|\nabla f| \approx \left[(z_5 - z_8)^2 + (z_5 - z_6)^2 \right]^{1/2}$$

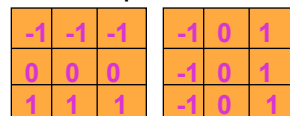
$$|\nabla f| \approx |z_5 - z_8| + |z_5 - z_6|$$

Robert's operator



$$|z_5 - z_9| \quad |z_6 - z_8|$$

prewitt



Sobel's

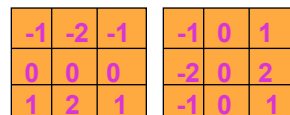
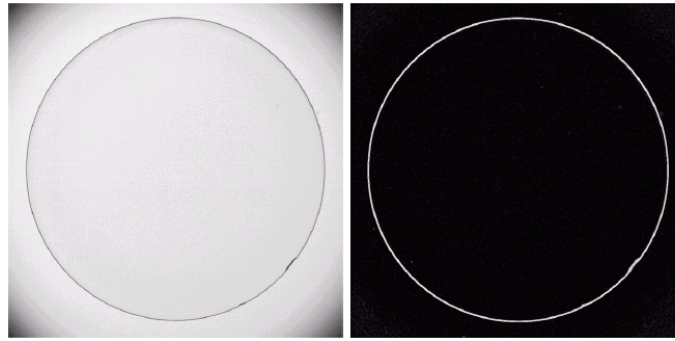


Fig 3.45: Sobel edge detector



a b

FIGURE 3.45
Optical image of contact lens (note defects on the boundary at 4 and 5 o'clock).
(b) Sobel gradient.
(Original image courtesy of Mr. Pete Sites, Perceptics Corporation.)

Laplacian based edge detectors

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

	1	
1	-4	1
	1	

- Rotationally symmetric, linear operator
- Check for the zero crossings to detect edges
- Second derivatives => sensitive to noise.