

# Digital Image Processing (DIP)

## Lec-(4-5)

By

Dr. Akram Alsubari

e-mail:- [akram.alsubari87@gmail.com](mailto:akram.alsubari87@gmail.com)

# Pervious Lecture

- Pattern Recognition

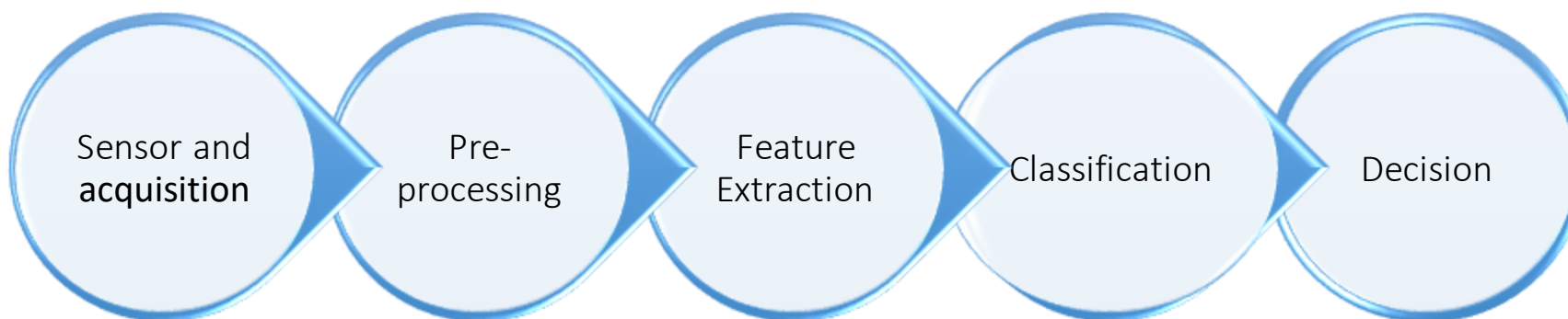
- Pattern

- **Spatial patterns** (patterns are located in space)
- **Temporal patterns** (Distributed in time)
- **Abstract patterns** (patterns are distributed neither in space nor time)

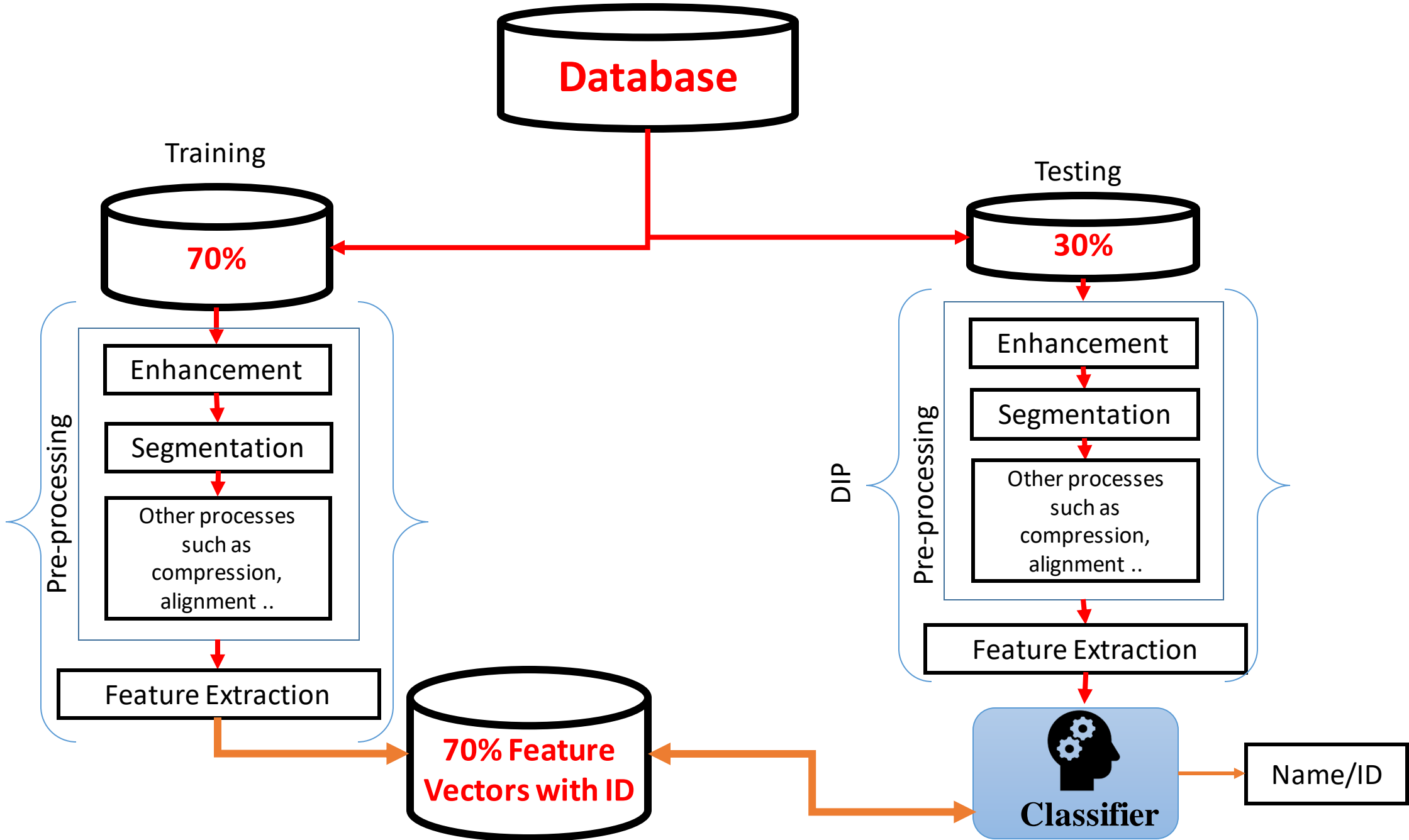
Pattern : A Visible Entity

Re**ognition** = Re + **Cognition**

↓                      ↓  
Labelling ← Learning



نفرض انها 100 صوره لعشره اشخاص وكل شخص له عشره صور



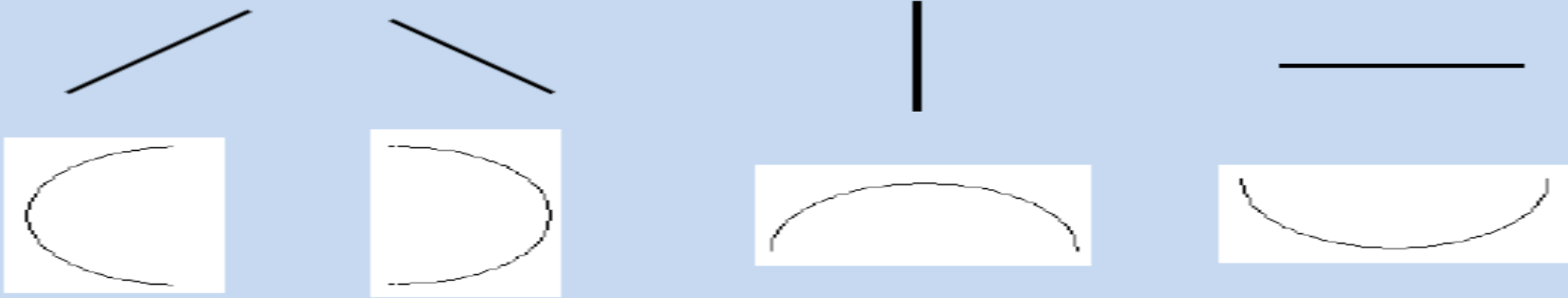
# Example : Feature Extraction

Objects :

A B C D E F

Feature?...

Line and Curve Segments



| Object | 0 | 45 | 90 | 145 | Top semi circle | Bottom Semi circle | Left Semi circle | Right Semi circle |
|--------|---|----|----|-----|-----------------|--------------------|------------------|-------------------|
| A      | 1 | 1  | 0  | 1   | 0               | 0                  | 0                | 0                 |
| B      | 0 | 0  | 1  | 0   | 0               | 0                  | 0                | 2                 |
| C      | 0 | 0  | 0  | 0   | 0               | 0                  | 1                | 0                 |
| D      | 0 | 0  | 1  | 0   | 0               | 0                  | 0                | 1                 |
| E      | 3 | 0  | 1  | 0   | 0               | 0                  | 0                | 0                 |
| F      | 2 | 0  | 1  | 0   | 0               | 0                  | 0                | 0                 |

Label/class

Feature Vector

# Face Recognition in MATLAB

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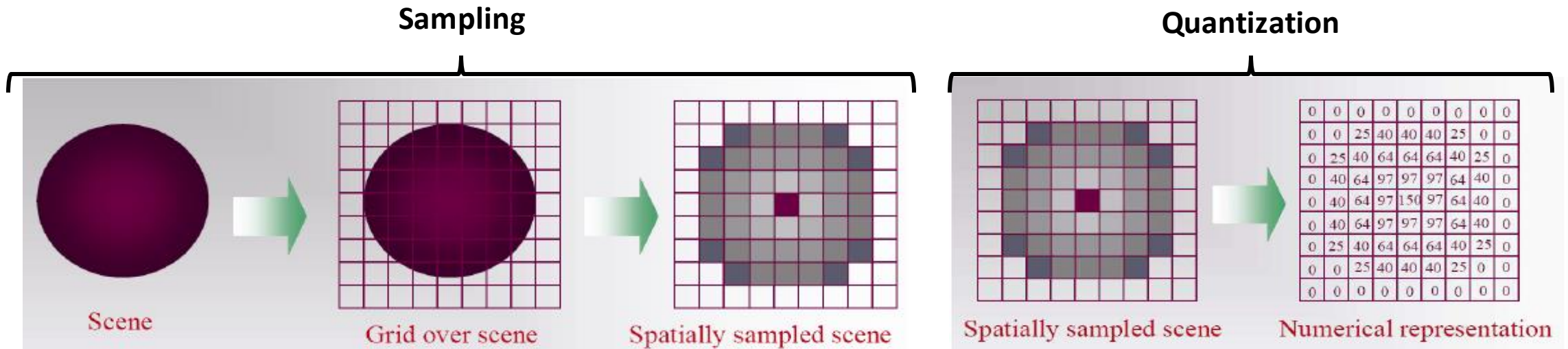
```
clear
imgcoll=imageSet('24_ORL','recursive');
[training, test]=partition(imgcoll,[0.5 0.5]);% split the database into training sets and testing sets
K=1;
for i=1:size(training,2)
    for j=1:training(i).Count
        trainingFeatures(K,:)= extractHOGFeatures(read(training(i),j));
        traininglabel{K}=training(i).Description;
        K=K+1;
    end
end
faceclfssifier= fitcknn(trainingFeatures,traininglabel);
cn=0;
for i=1:size(test,2)
    for j=1:test(i).Count
        InserFace=read(test(i),j);
        FaceFeatures=extractHOGFeatures(InserFace);
        personlabel=predict(faceclfssifier,FaceFeatures);
        if strcmp(personlabel,test(i).Description)
            cn=cn+1;
        end
    end
end
Recognition_rate=(cn/(i*j))*100
```



# Image Digitization

# Image Sensing and Acquisition

- An image captured by **sensor** is expressed as continuous function  $f(x, y)$  of two co-ordinates in the plane.
- **Image Digitization**
  - **Image sampling** is to convert/sample continuous function  $f(x, y)$  into matrix with M rows and N columns
  - **Image quantization** is to assign to each continuous sample an integer value





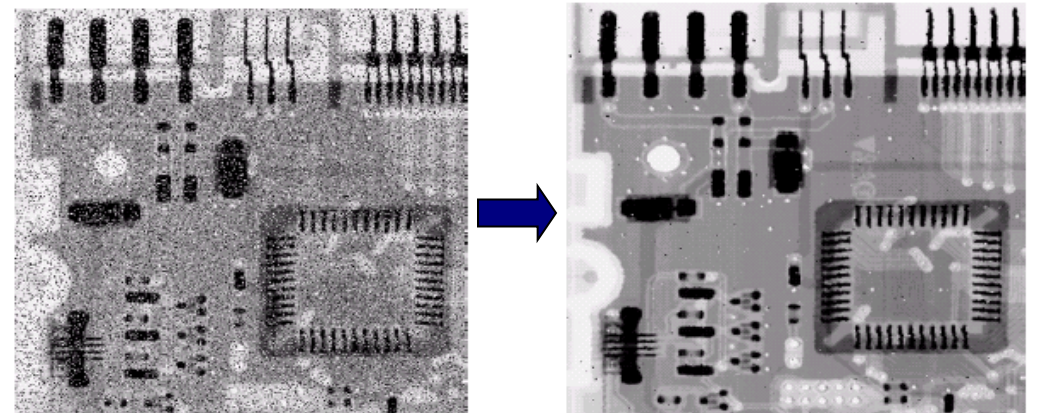
# Image Enhancement



# Enhancement

Image enhancement is the process of making images more useful

- Highlighting interesting detail in images
- Removing noise from images
- Making images more visually appealing



# Enhancement

There are two broad categories of image enhancement techniques

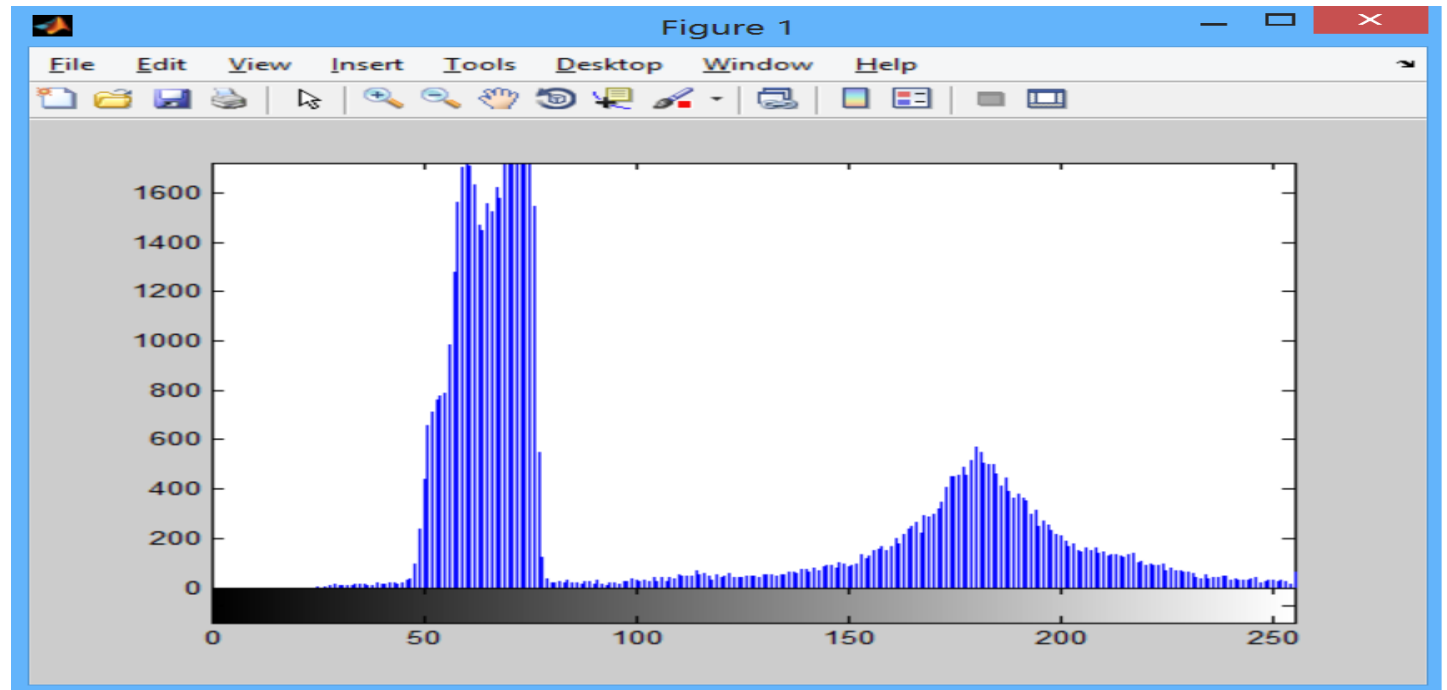
- **Spatial domain** techniques
  - Direct manipulation of image pixels
- **Frequency domain** techniques
  - Manipulation of Fourier transform or wavelet transform of an image

# Image Histogram

- The histogram of an image shows us the distribution of grey levels in the image

```
f = imread('coins.png');
```

```
figure, imhist(f);
```

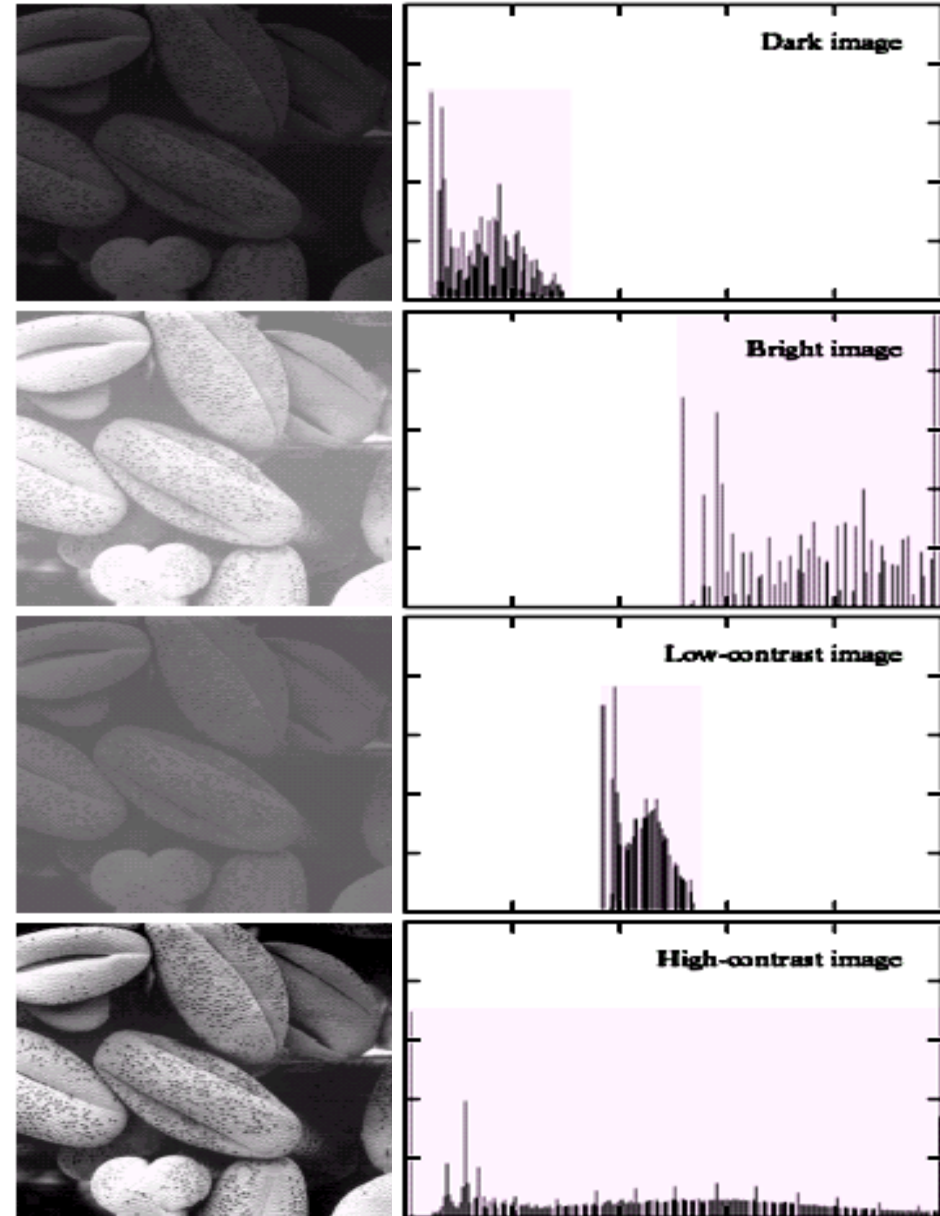


# Histogram Examples

A selection of images and their histograms

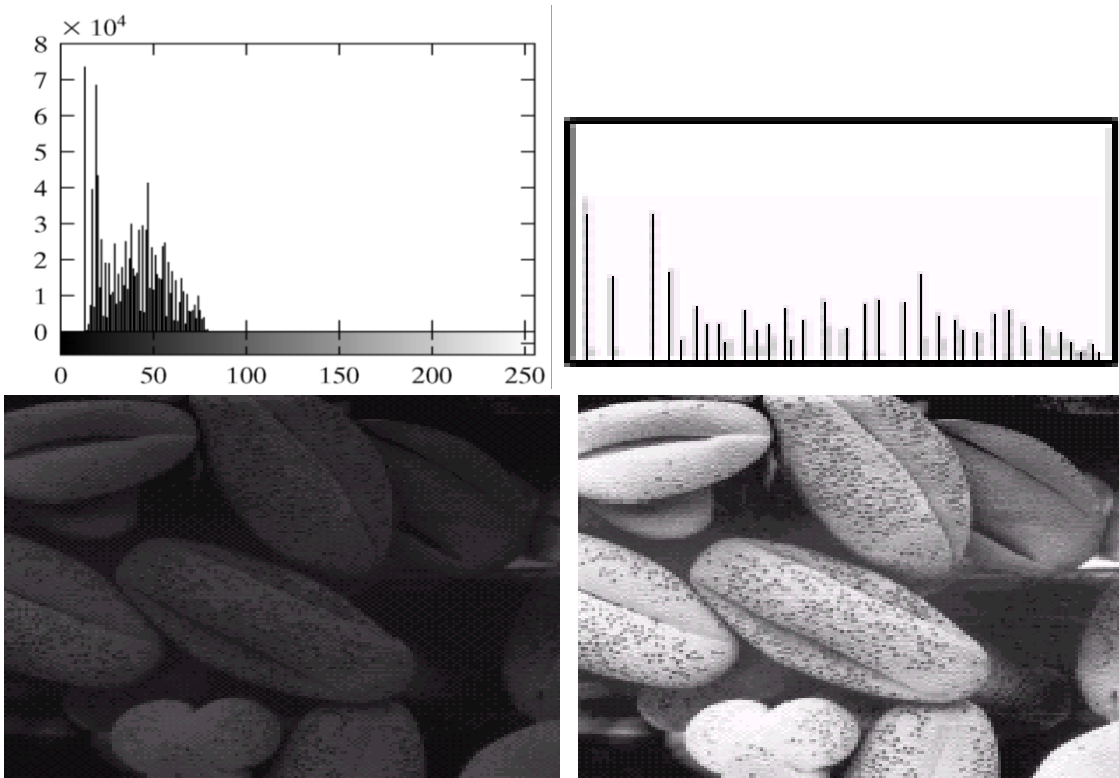
Notice the relationships between the images and their histograms

Note that the high contrast image has the most evenly spaced histogram



# Histogram Equalisation

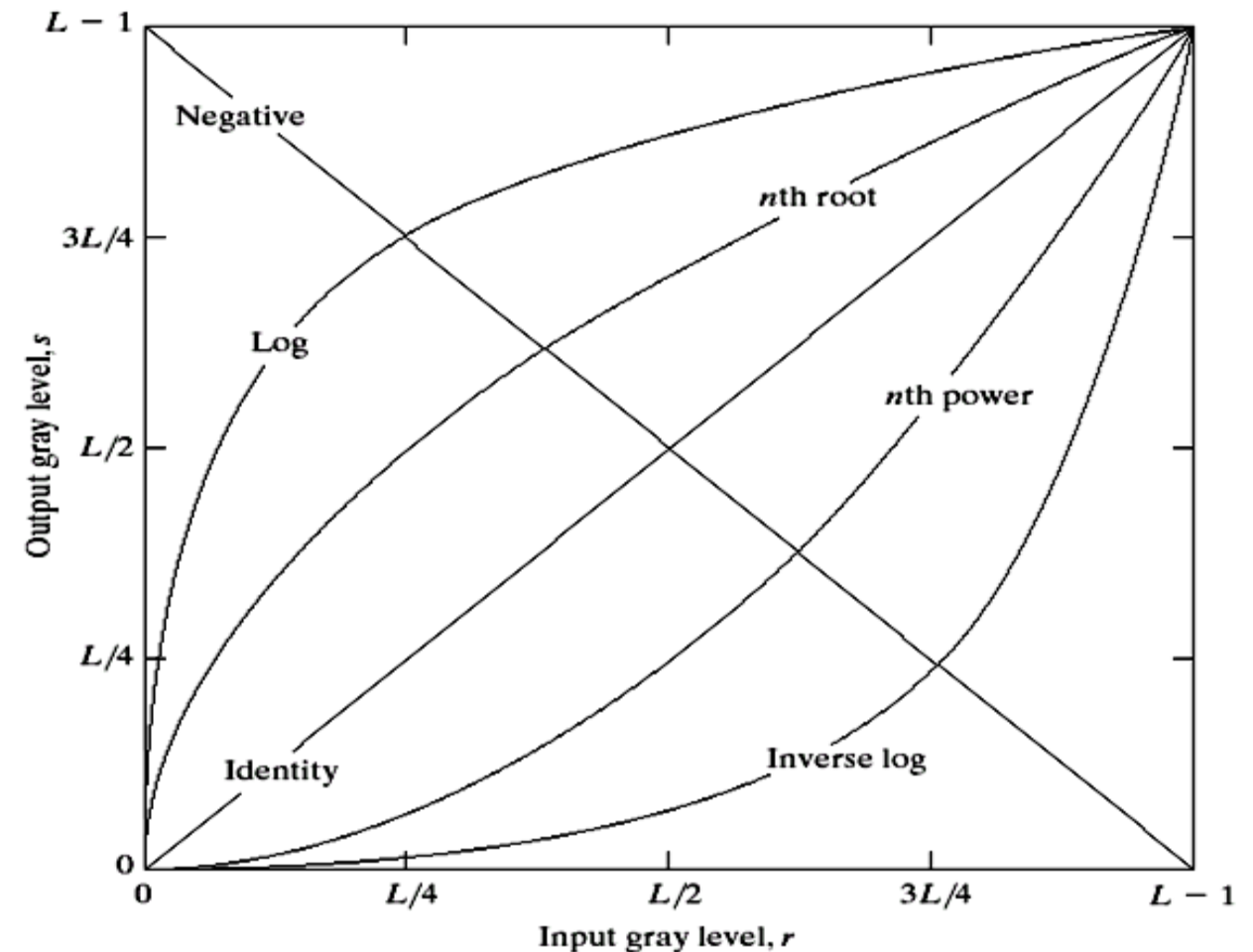
- Spreading out the frequencies in an image (or equalising the image) is a simple way to improve dark or washed out images



# Basic Grey Level Transformation

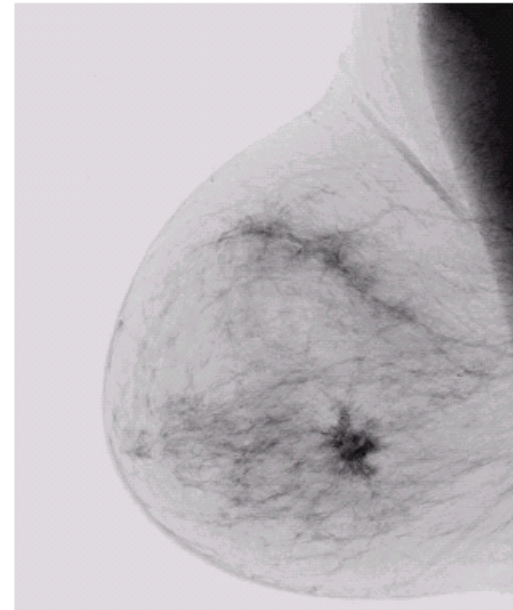
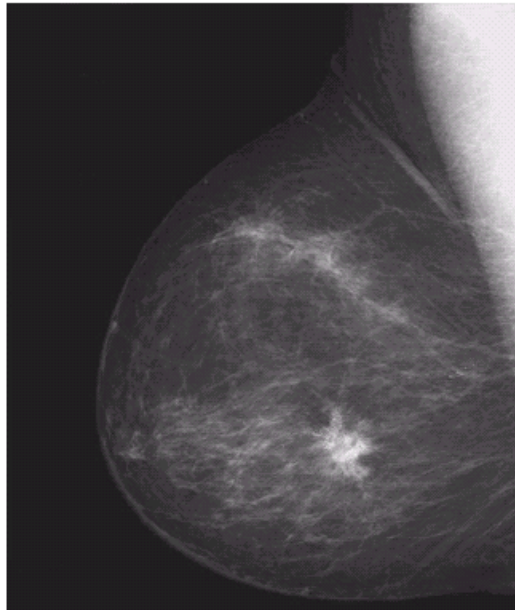
- Basic Grey Level Transformation

- 1) Linear (negative and identity)
- 2) Logarithmic
- 3) Power – law



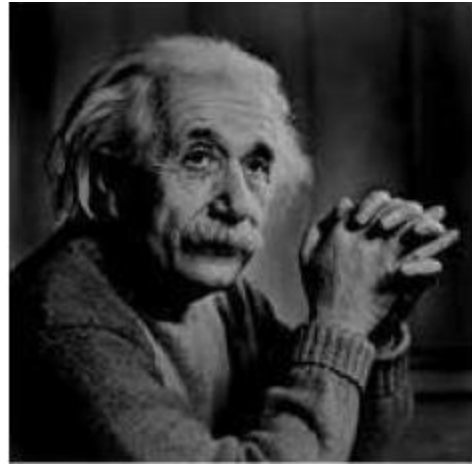
# Negative Transformation

- $s = \textit{intensity}_{max} - r$



# Logarithmic Transformation

$$s = c \log(r + 1)$$



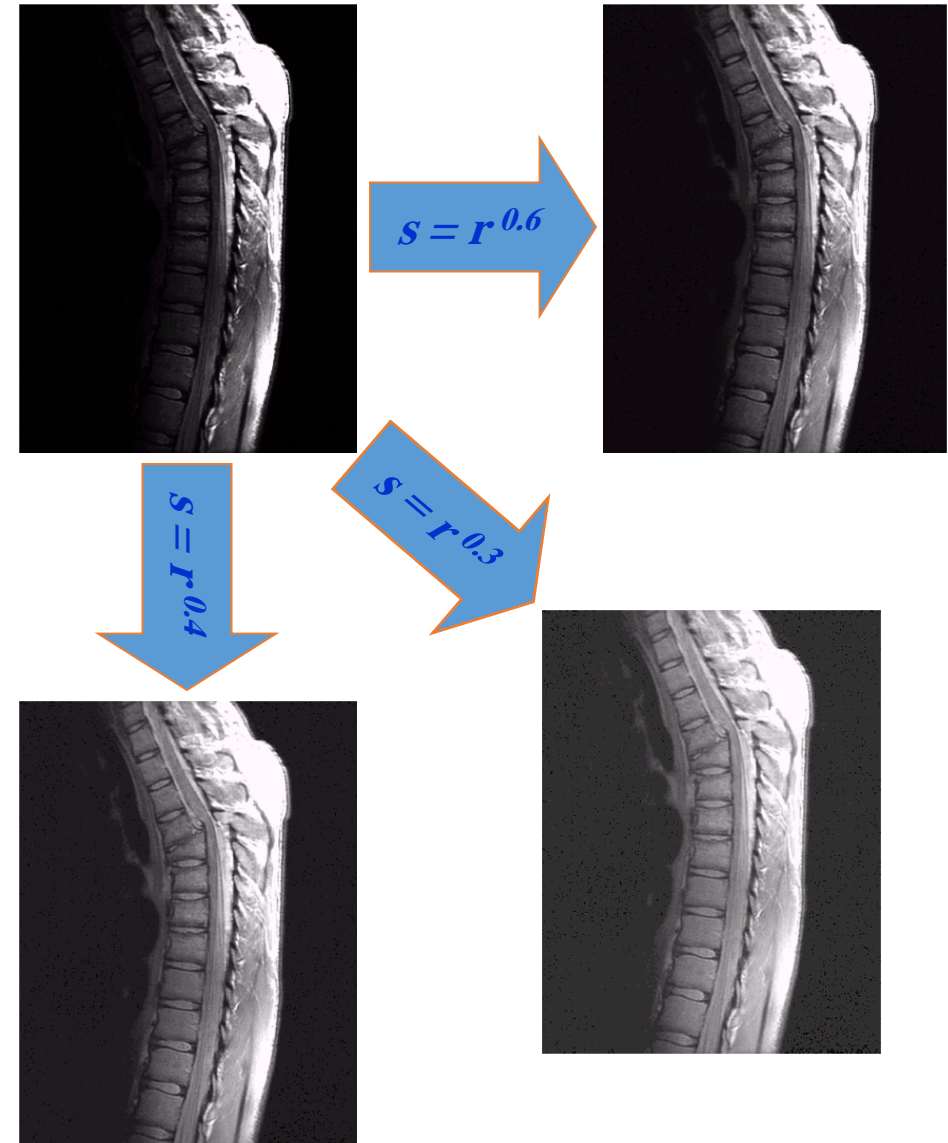
Log functions are particularly **useful when** the input grey level values **may have an extremely large range** of values



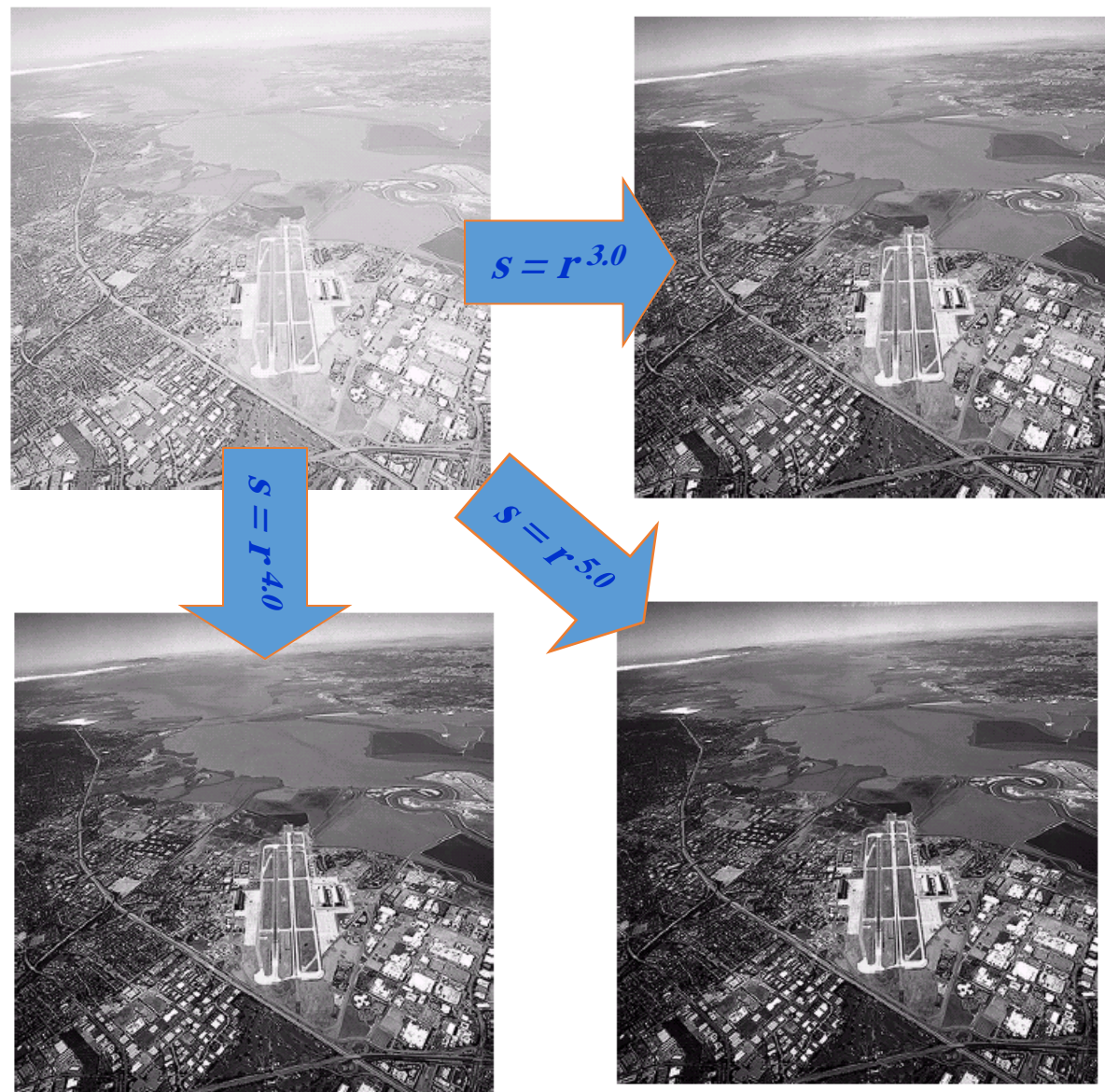
# Power – law Transformation

$$s = cr^\gamma$$

Map a narrow range  
of dark input values  
into a wider range  
of output values or vice  
versa



Cont..



# Brightness and Contrast

- **Brightness:**

intensity of light emit by a particular light source.

- **Contrast:**

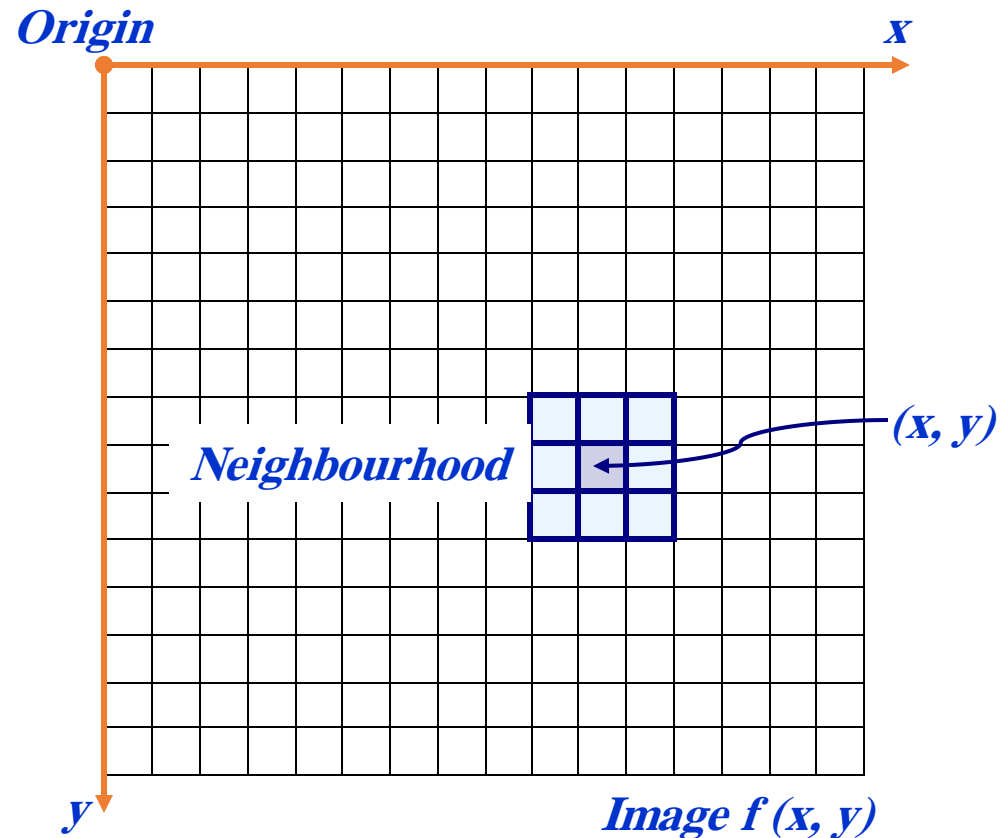
the difference between maximum and minimum pixel intensity in an image.

# CODES

```
%% Read an 8 bit image and then apply different image enhancement techniques
clear; clc;
close all
Img= imread('pout.tif');
BI= imadjust(Img);      % Adjust image intensity values
BR= Img./2;            % Brightness Reduction by dividing each pixel by 2
BW= im2bw(Img);         % Convert Gray Image into Binary Image Based on Thresholding
BW2= Img<110;           % another way to convert gray image into binary image, where the Thresholding is 110
NI = imcomplement(Img); % Negative of the Image
LT= log(double(Img));   % Log Transformation
DImg=double(Img);
a=1;gamma=2;
PLT= a*(DImg).^gamma; % Power Law Transform
subplot(3,3,1);
imshow(Img); title ('Original Image')
subplot (3,3,2);
imshow(BI); title ('image Improvement')
subplot(3,3,3);
imshow(BR); title ('Brightness Reduction')
subplot(3,3,4);
imshow(BW); title ('Binary Image')
subplot(3,3,5);
imshow(BW2); title ('Thresholding=110')
subplot(3,3,6);
imshow(NI); title ('Negative of the Image')
subplot(3,3,7);
imshow (LT, []); title ('Log Transformation ')
subplot(3,3,8);
imshow(PLT, []); title ('Power Low Transform')
```

# Neighbourhood Operations

- Neighbourhood operations simply operate on a larger neighbourhood of pixels than point operations
- Neighbourhoods are mostly a rectangle around a central pixel
- Any size rectangle and any shape filter are possible

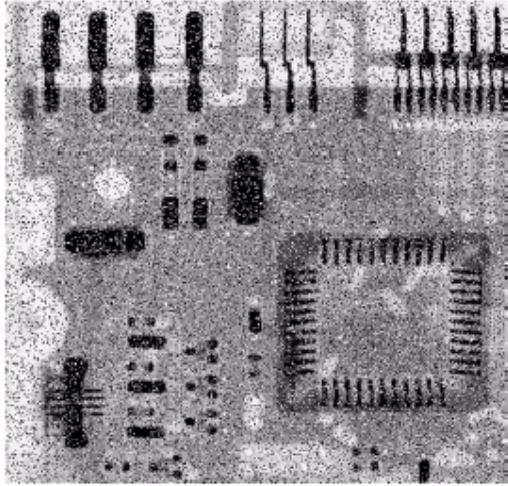


# Simple Neighbourhood Operations

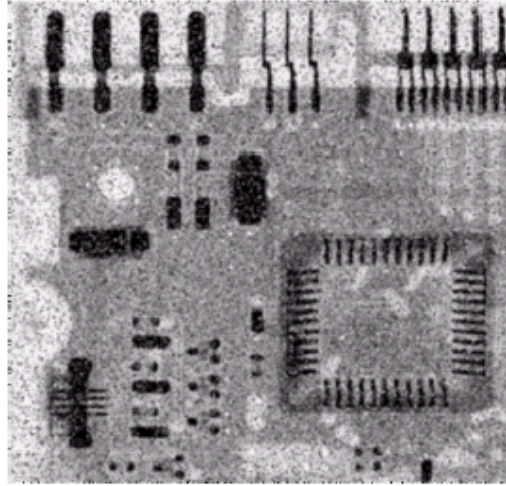
- Some simple neighbourhood operations include:
  - **Min:** Set the pixel value to the minimum in the neighbourhood
  - **Max:** Set the pixel value to the maximum in the neighbourhood
  - **Median:** The median value of a set of numbers is the midpoint value in that set (e.g. from the set [1, 7, 15, 18, 24] 15 is the median). Sometimes the median works better than the average



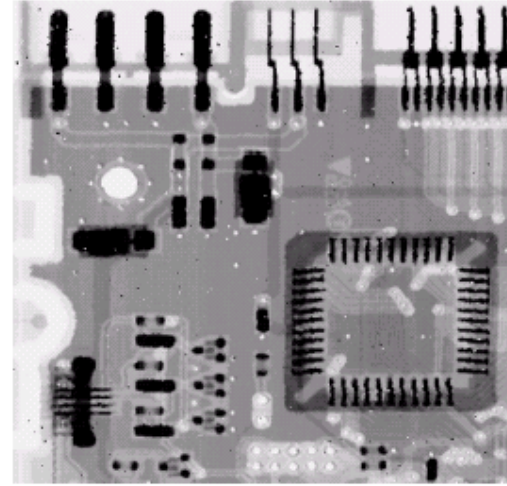
# Averaging Filter Vs. Median Filter Example



**Original Image  
With Noise**



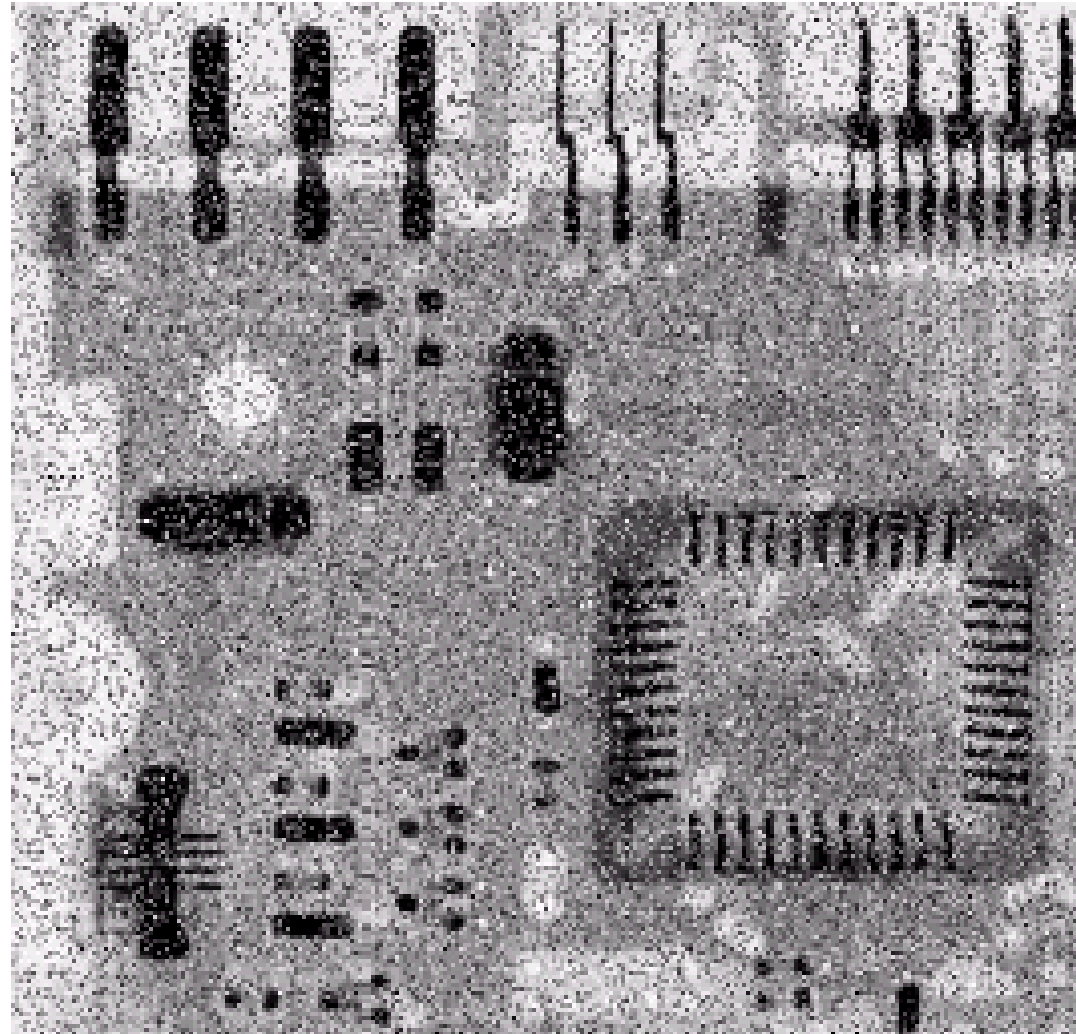
**Image After  
Averaging Filter**



**Image After  
Median Filter**

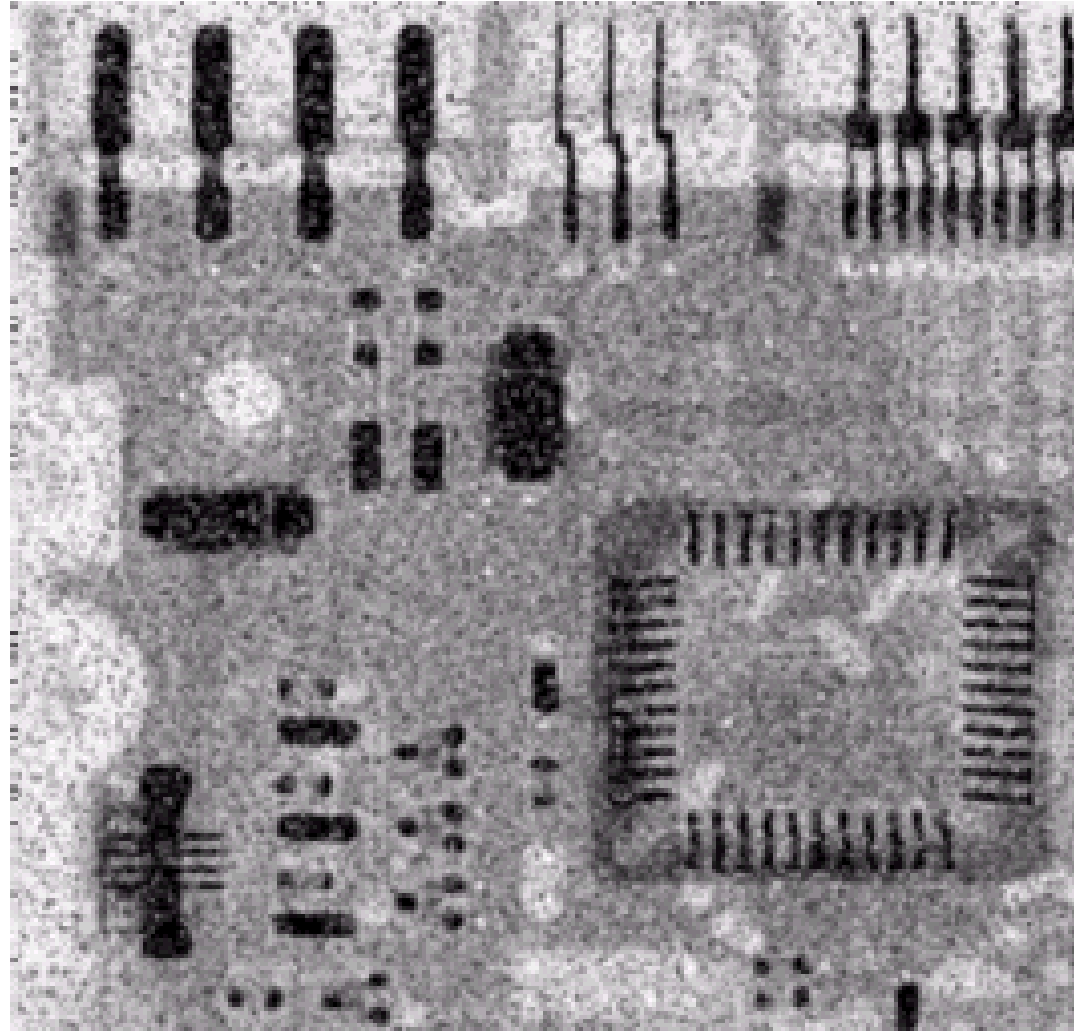
- Filtering is often used to remove noise from images
- Sometimes a median filter works better than an averaging filter

# Averaging Filter Vs. Median Filter Example

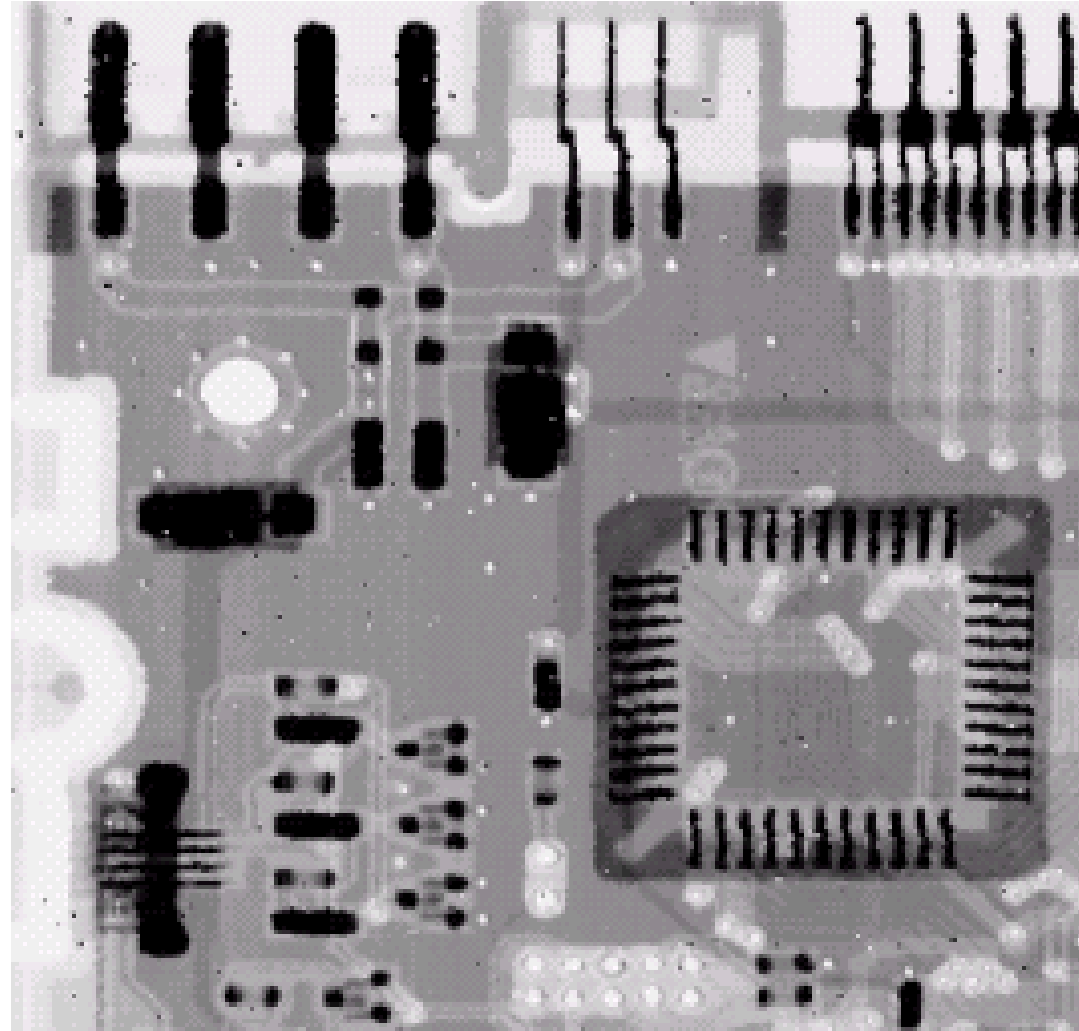




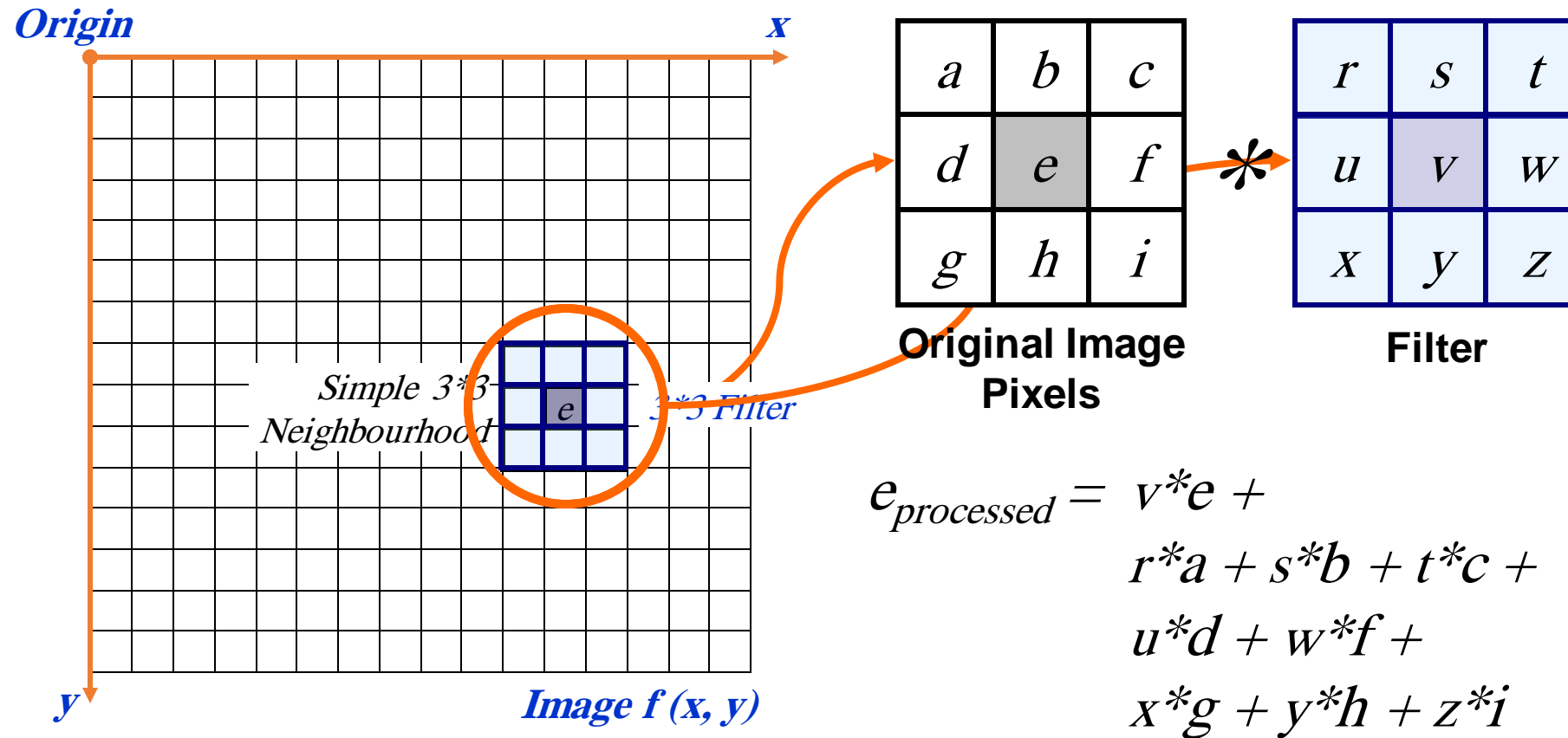
# Averaging Filter Vs. Median Filter Example



# Averaging Filter Vs. Median Filter Example

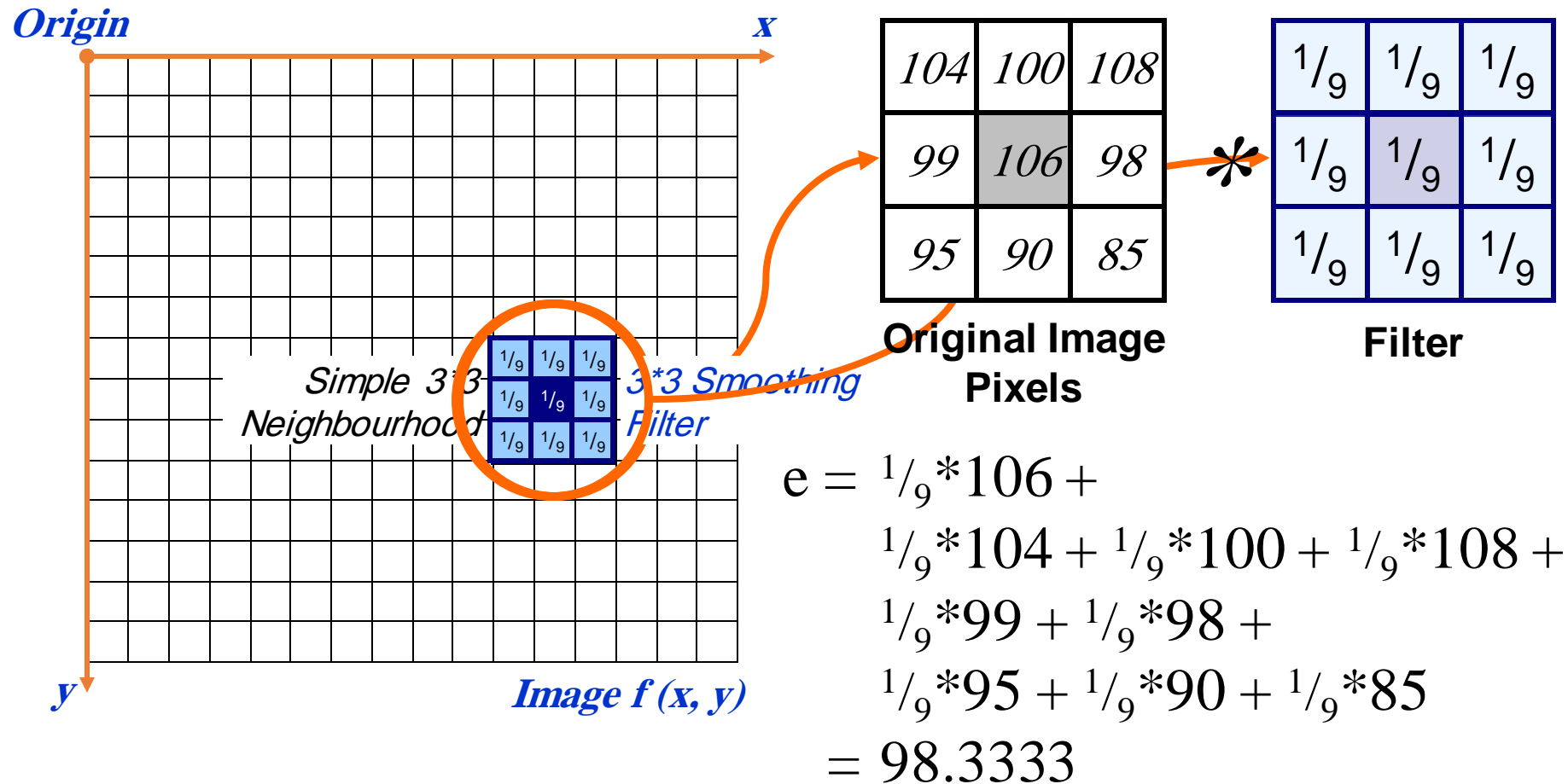


# The Spatial Filtering Process



The above is repeated for every pixel in the original image to generate the filtered image

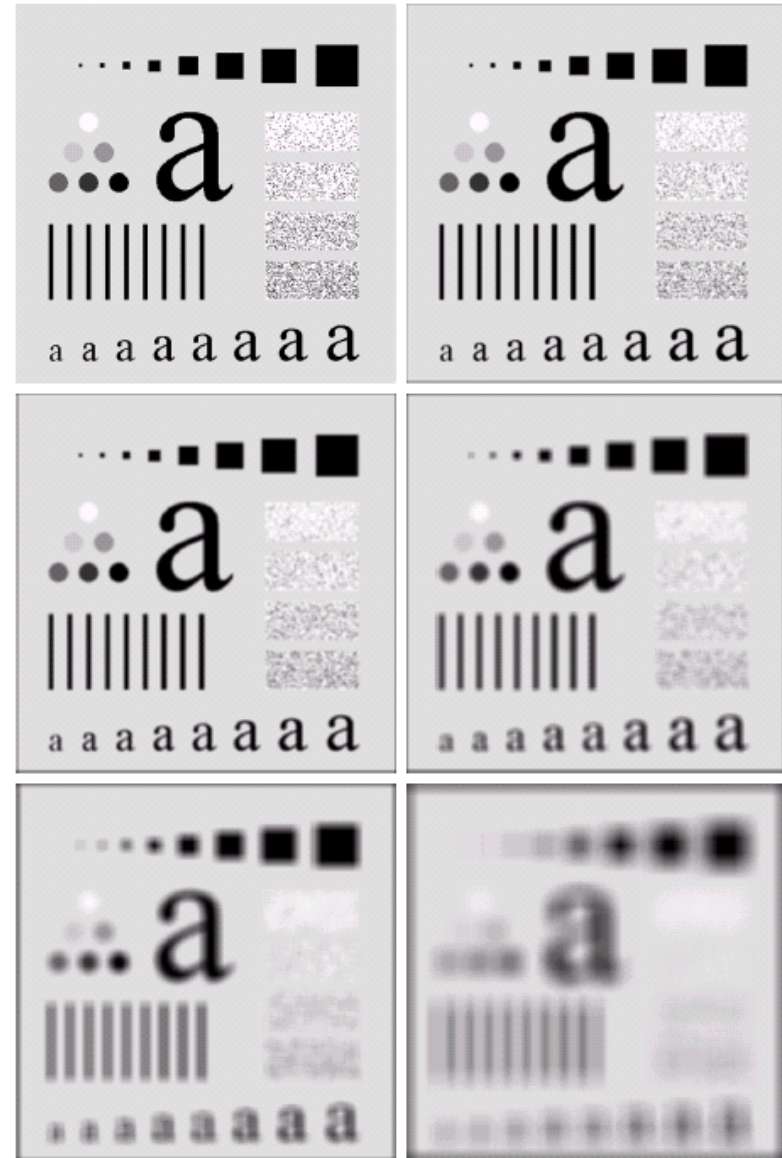
# Smoothing Spatial Filtering



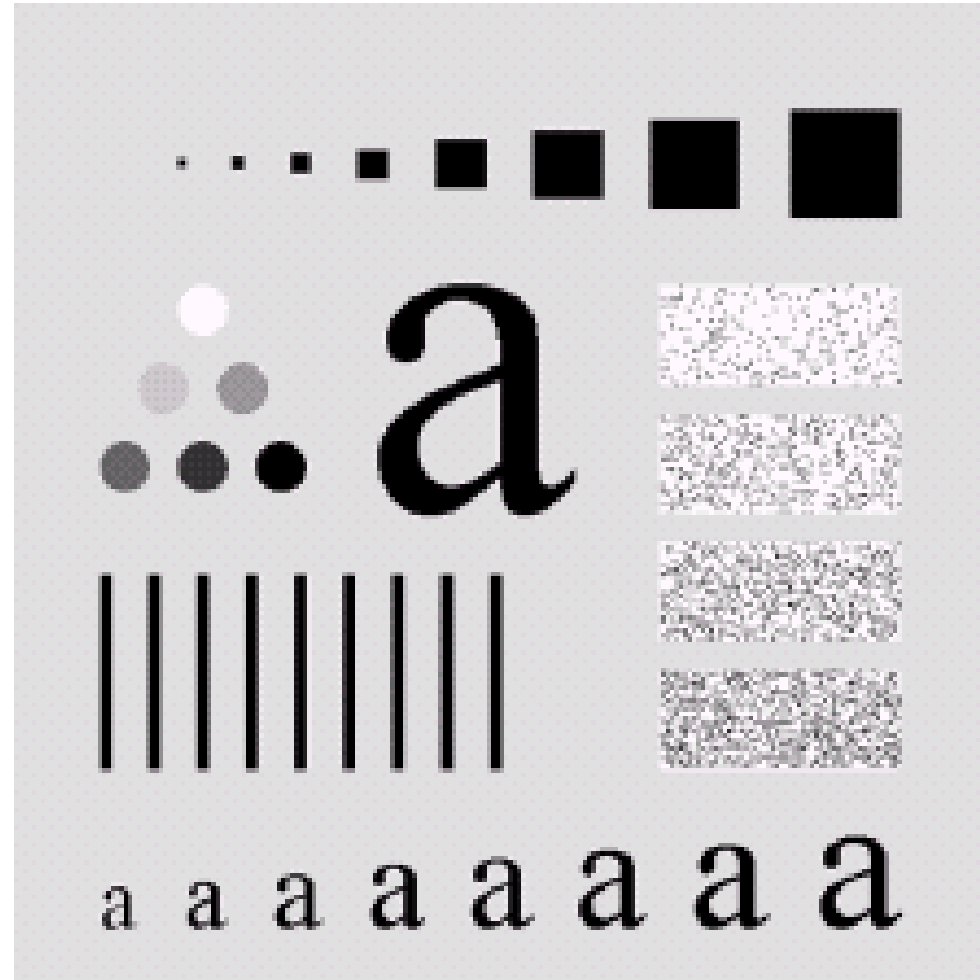
The above is repeated for every pixel in the original image to generate the smoothed image

# Image Smoothing Example

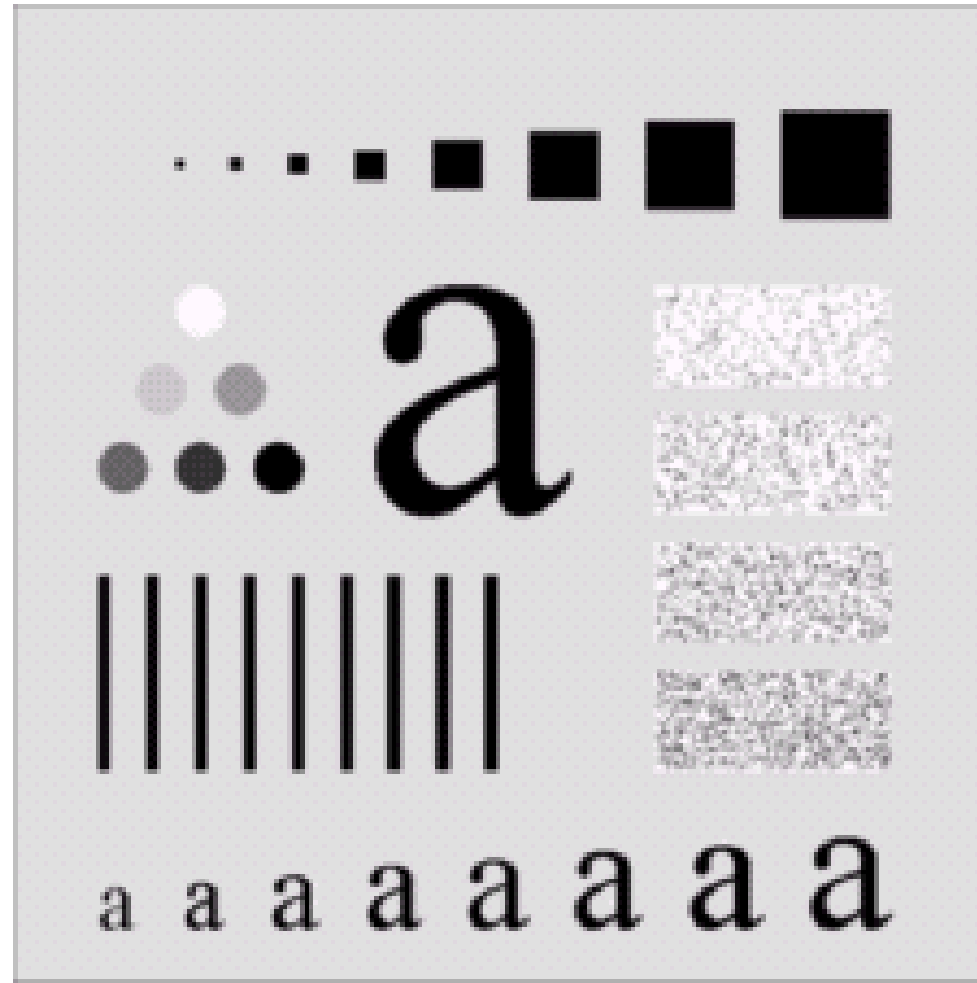
- The image at the top left is an original image of size 500\*500 pixels
- The subsequent images show the image after filtering with an averaging filter of increasing sizes
  - 3, 5, 9, 15 and 35
- Notice how detail begins to disappear



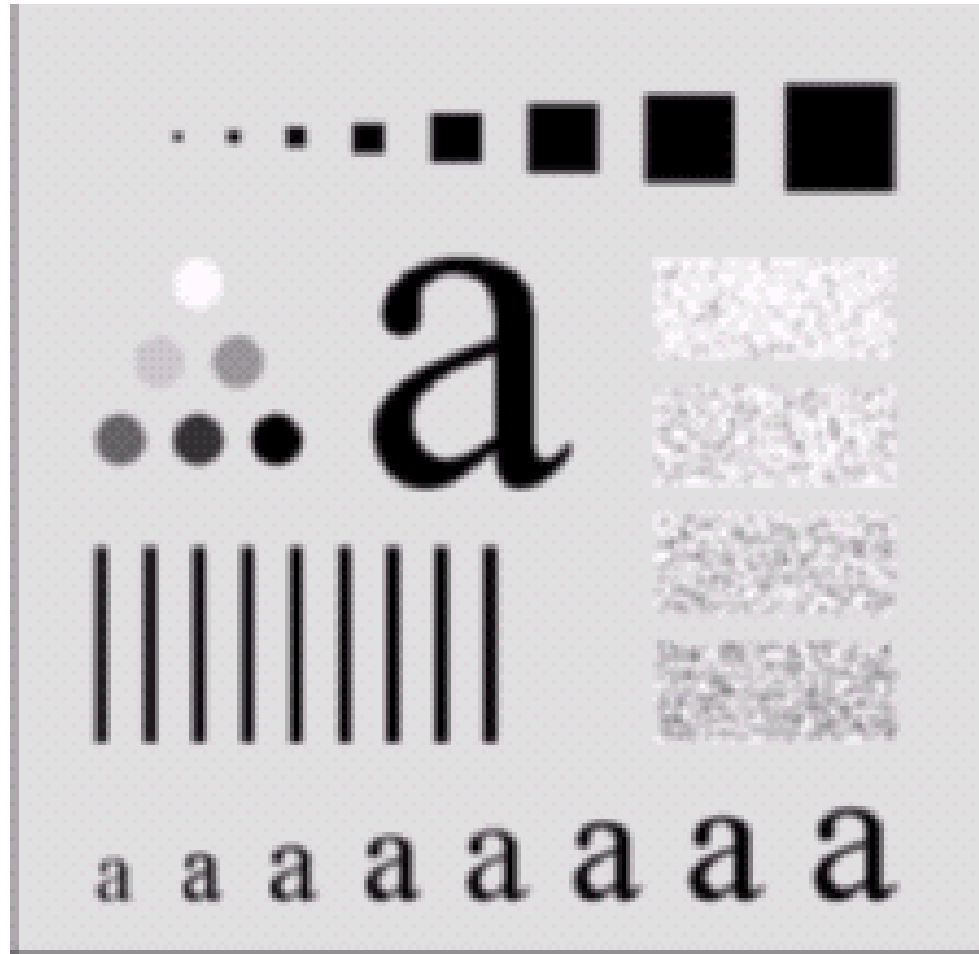
# Image Smoothing Example



# Image Smoothing Example

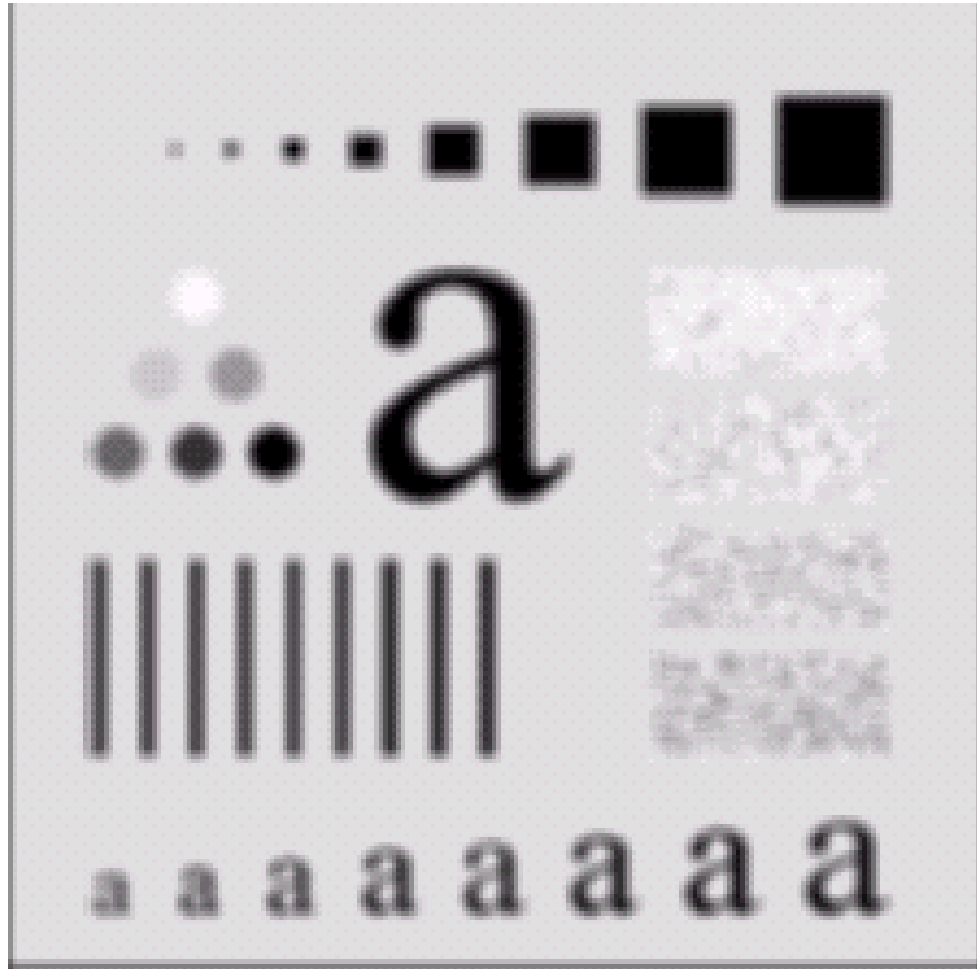


# Image Smoothing Example

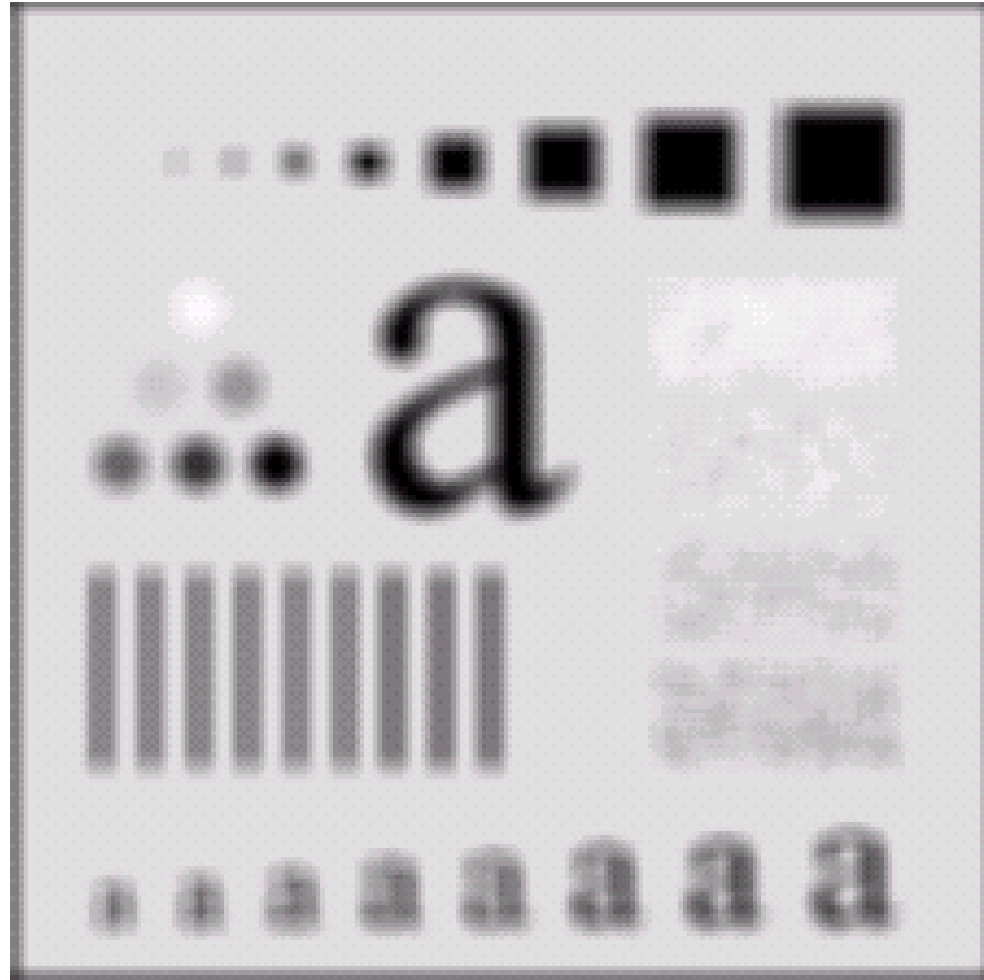




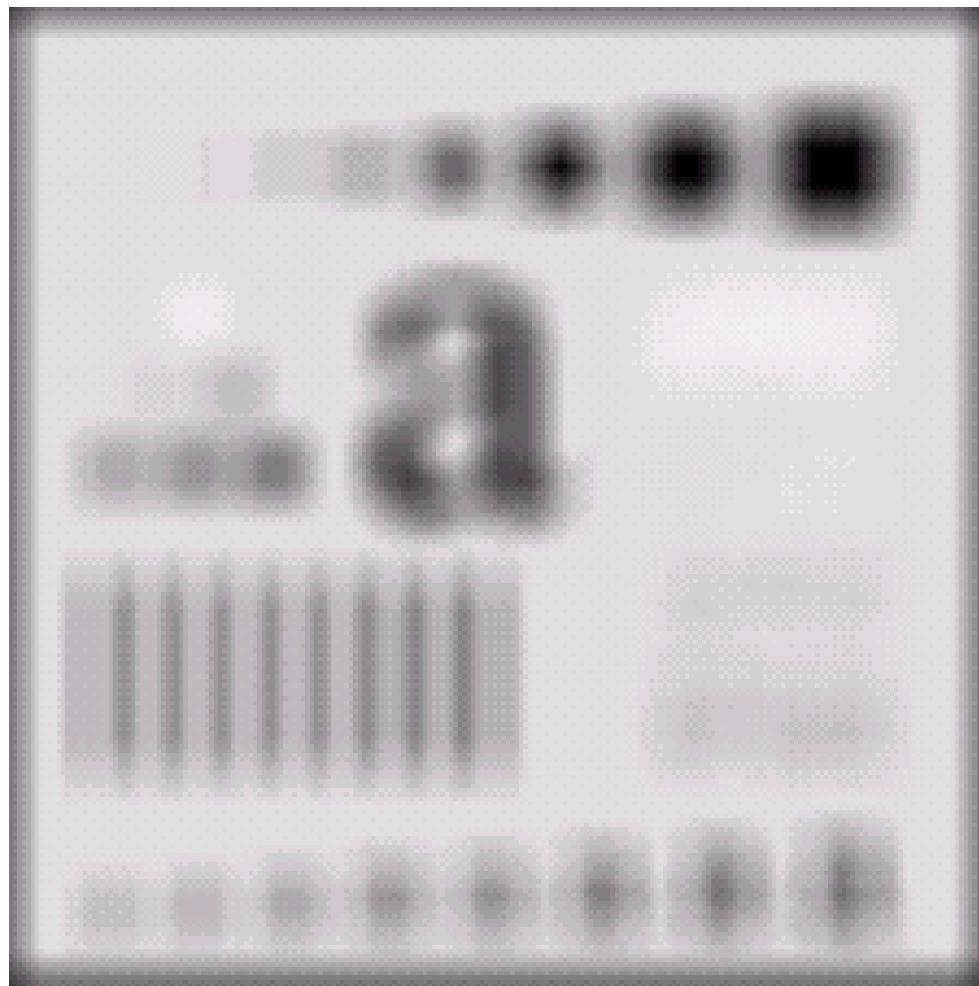
# Image Smoothing Example



# Image Smoothing Example



# Image Smoothing Example



# Weighted Smoothing Filters

- More effective smoothing filters can be generated by allowing different pixels in the neighbourhood different weights in the averaging function

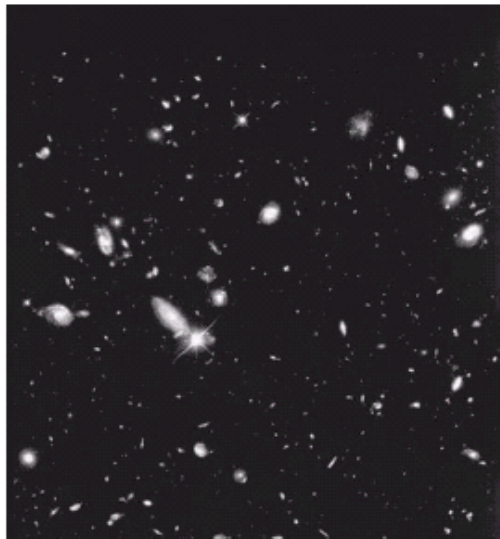
- Pixels closer to the central pixel are more important
- Often referred to as a *weighted averaging*

|                |                |                |
|----------------|----------------|----------------|
| $\frac{1}{16}$ | $\frac{2}{16}$ | $\frac{1}{16}$ |
| $\frac{2}{16}$ | $\frac{4}{16}$ | $\frac{2}{16}$ |
| $\frac{1}{16}$ | $\frac{2}{16}$ | $\frac{1}{16}$ |

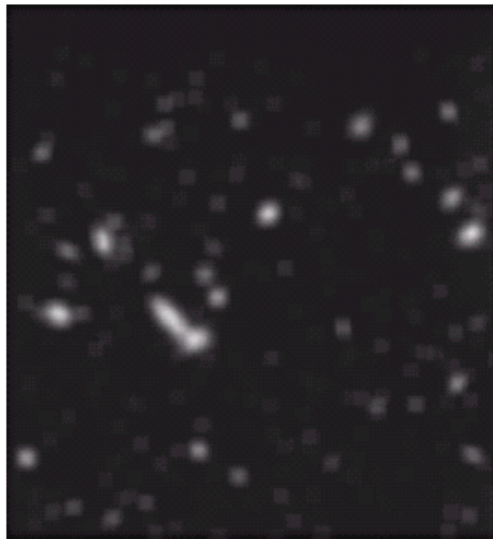
Weighted  
averaging filter

# Another Smoothing Example

- By smoothing the original image we get rid of lots of the finer detail which leaves only the gross features for thresholding



**Original Image**



**Smoothed Image**



**Thresholded Image**

# Sharpening Spatial Filters

Previously we have looked at smoothing filters which remove fine detail

*Sharpening spatial filters* seek to highlight fine detail

- Remove blurring from images
- Highlight edges

Sharpening filters are based on *spatial differentiation*

```
clc
clear all
close all
a=imread('horse.jpg');
%Addition of salt and pepper noise
b=imnoise(a,'salt & pepper',0.1);
%Defining the box and median filters
h1=1/9*ones(3,3);
h2=1/25*ones(5,5);
c1=conv2(b,h1,'same');
c2=conv2(b,h2,'same');
c3=medfilt2(b,[3 3]);
c4=medfilt2(b,[5 5]);
subplot(3,2,1),imshow(a),title('Original image')
subplot(3,2,2),imshow(b),title('Salt & pepper noise')
subplot(3,2,3),imshow(uint8(c1)),title('3 x 3 smoothing')
subplot(3,2,4),imshow(uint8(c2)),title('5 x 5 smoothing')
subplot(3,2,5),imshow(uint8(c3)),title('3x 3 Median filter')
subplot(3,2,6),imshow(uint8(c4)),title('5 x 5 Median filter')
```

# Challenge (1)



**Thank You**