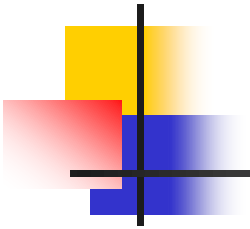

Math 3360: Mathematical Imaging

Lecture 15:
Constrained least square filtering

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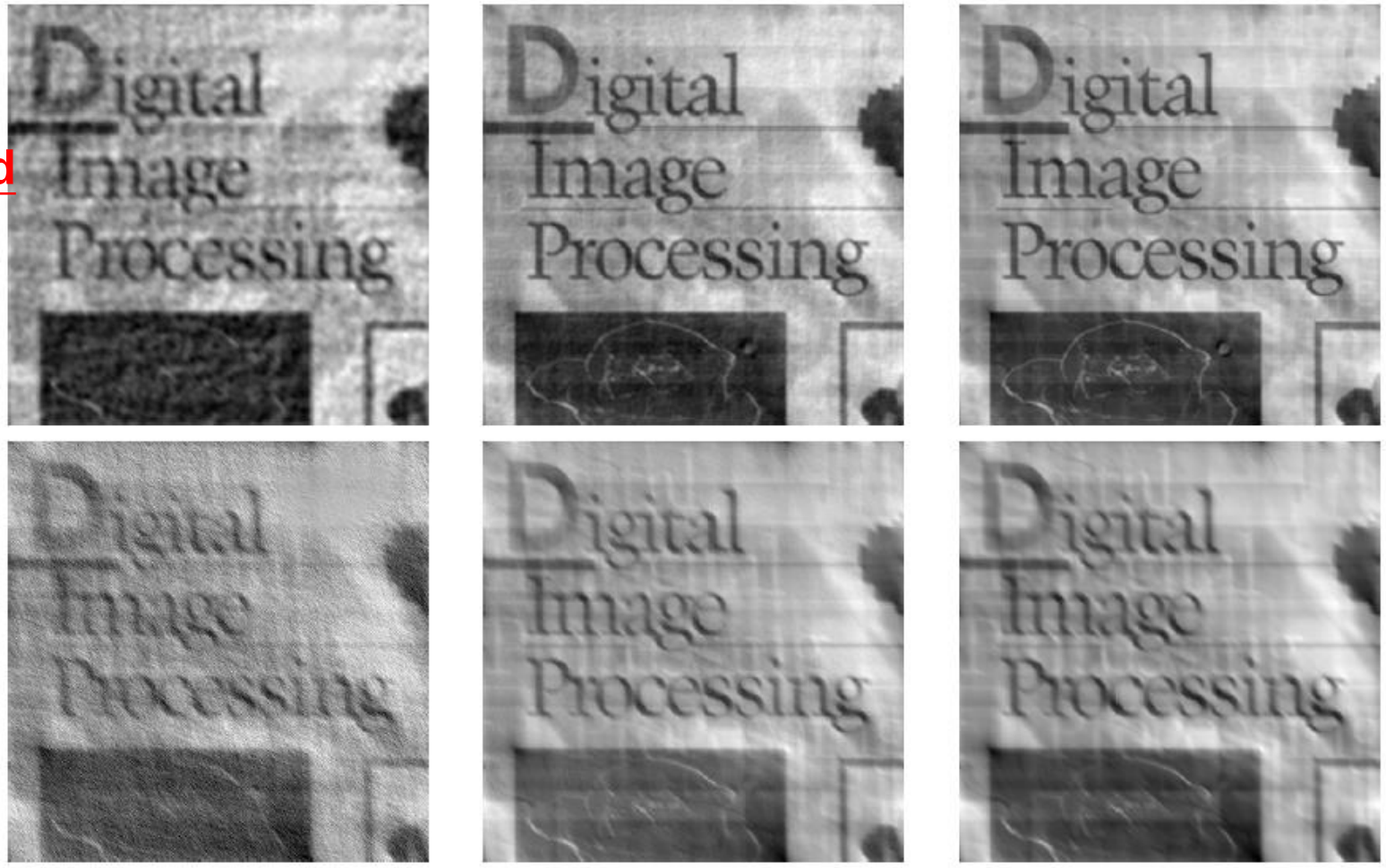
Constrained least square filtering

$$\hat{F}(u,v) = \left[\frac{H^*(u,v)}{|H(u,v)|^2 + \gamma|P(u,v)|^2}\right] G(u,v)$$

Constrained Least Square

Wiener filter



High Noise | Medium Noise | Low Noise

Constrained least square filtering

Blurry image without noise

Constrained least square filtering

Blurry image with noise

Constrained least square filtering

Blurry images

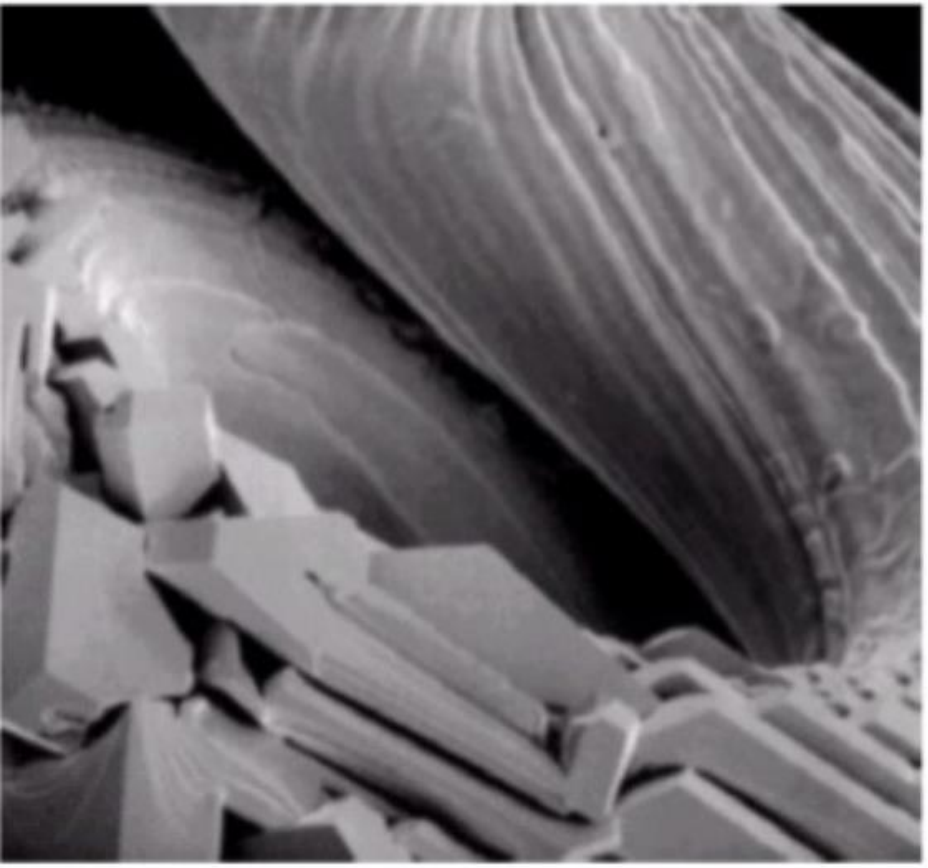
Deblurred images

Laplacian mask

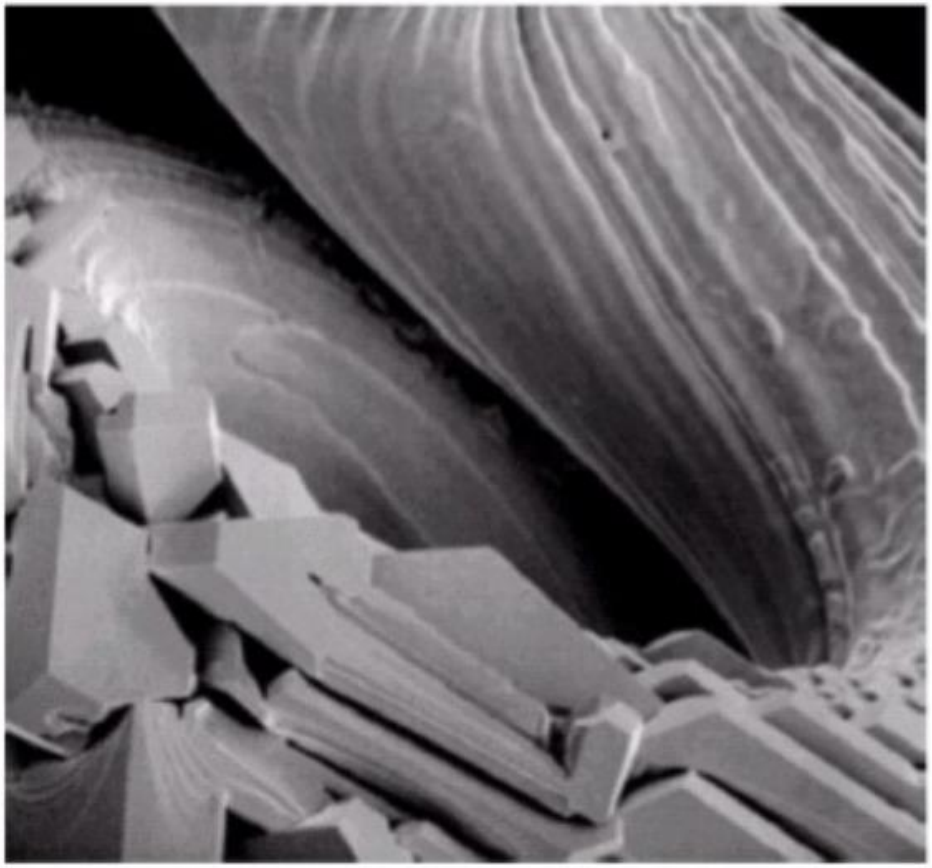
Original image

Laplacian mask

Laplacian mask

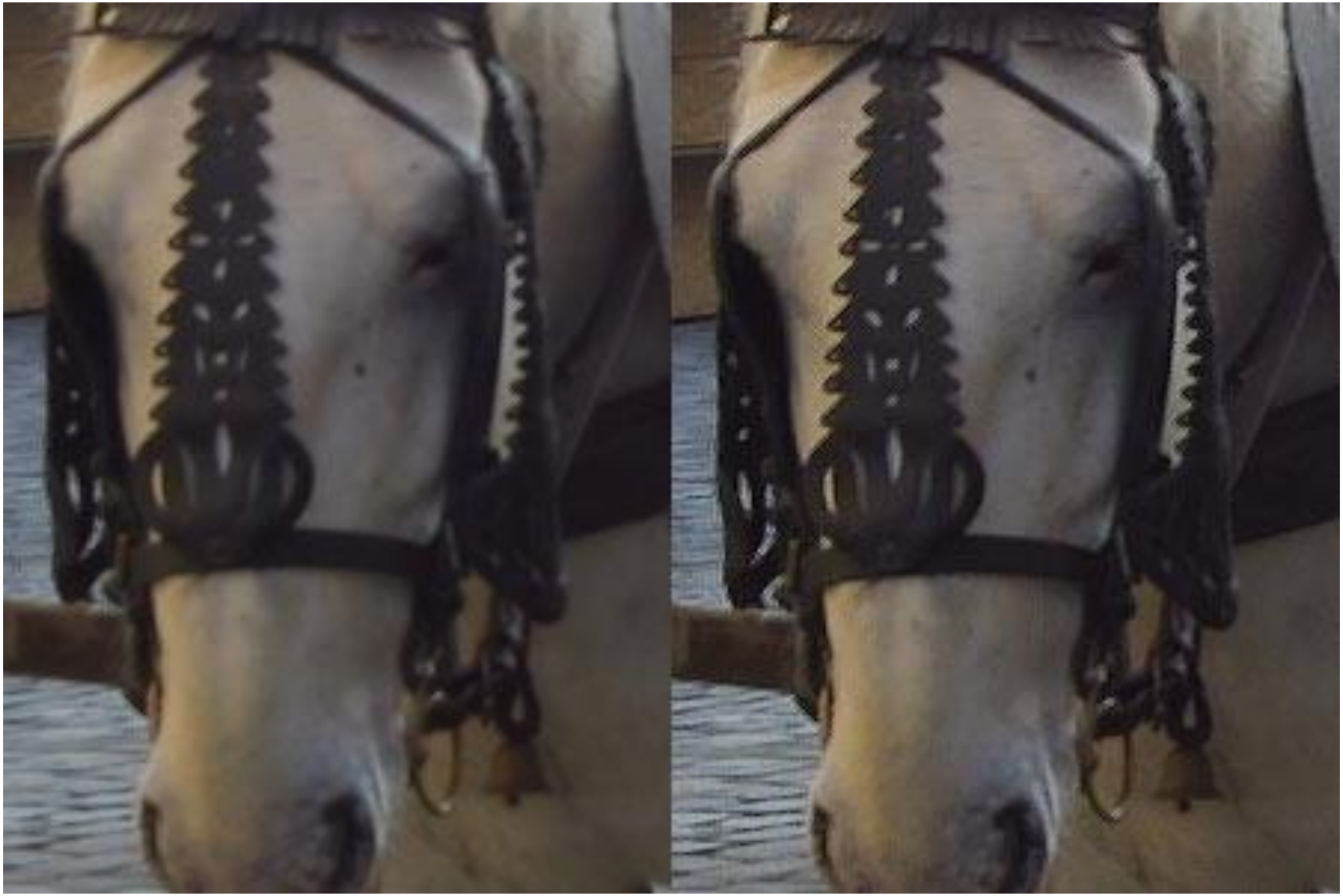


Original image



Laplacian mask

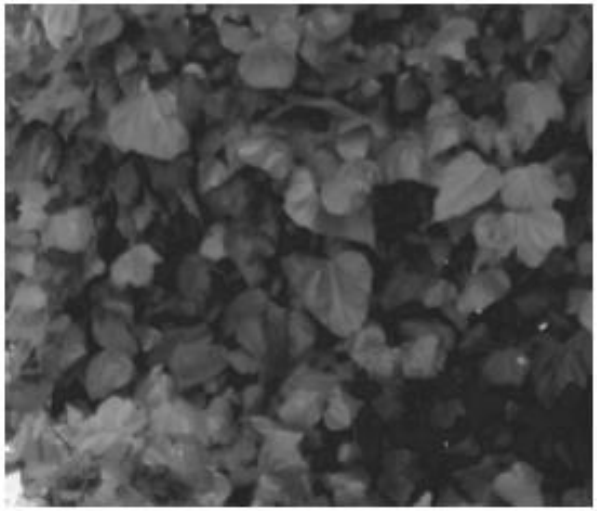
Laplacian mask



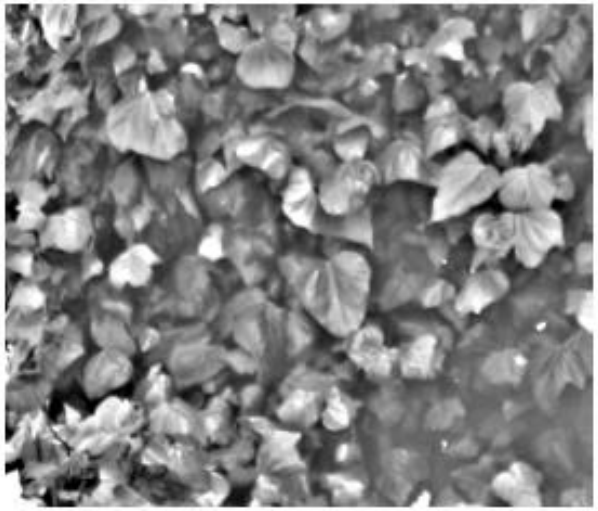
Original image

Laplacian mask

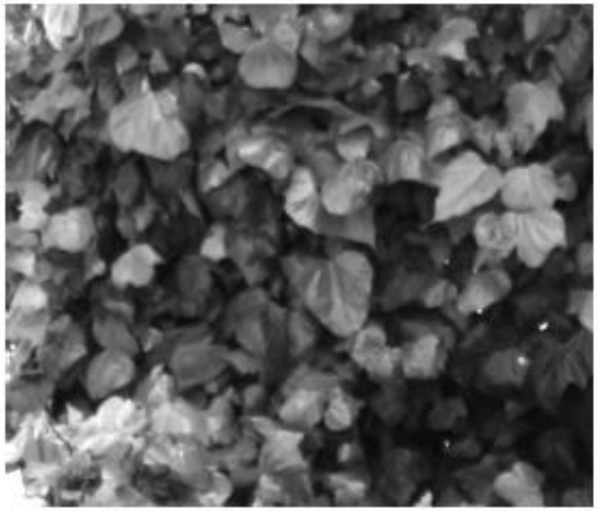
Unsharp masking



(a) Original



(b) Gaussian



(c) Mean

Unsharp masking

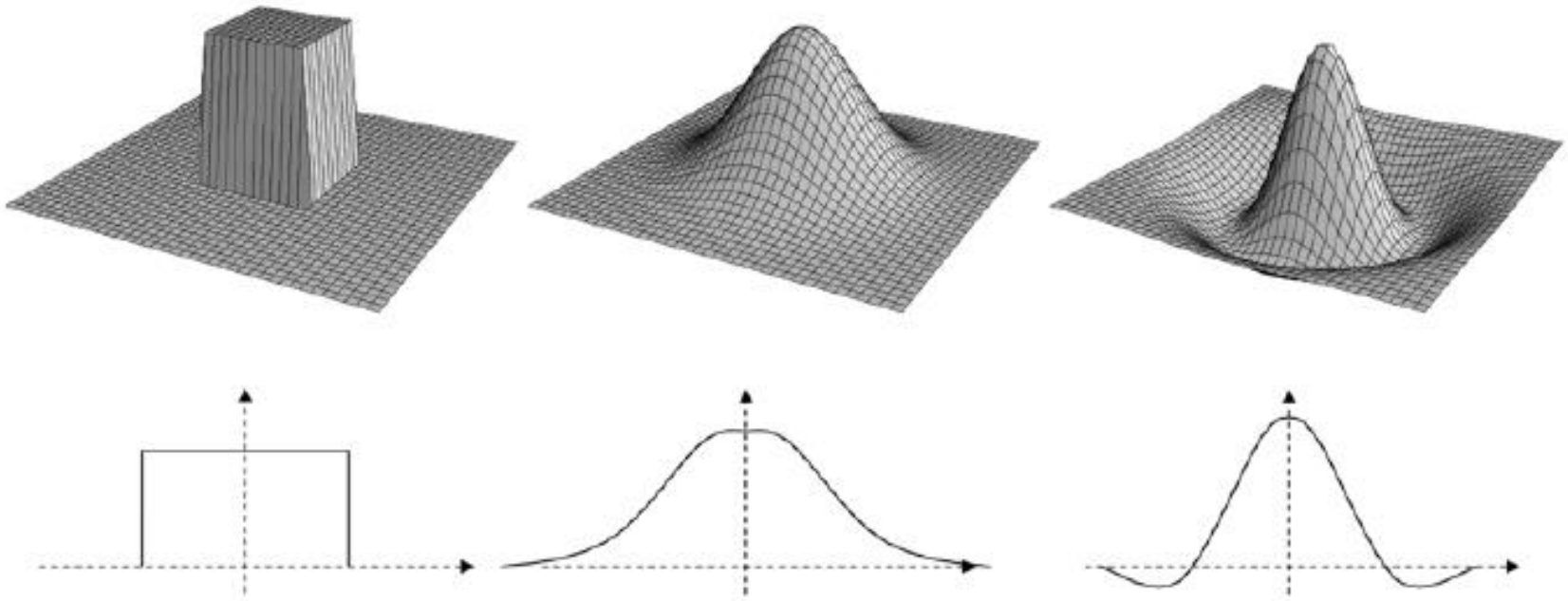
(a) Original

(b) Global, Gaussian 121×121

Linear filter = Convolution

- Linear filtering of a (2M+1)x(2N+1) image I (defined on [-M,M]x[-N,N]) = CONVOLUTION OF I and H
- H is called the filter.
- Different filter can be used:
  - Mean filter
  - Gaussian filter
- Variation of these filters (Non-linear)
  - Median filter
  - Edge preserving mean filter

Type of filter



0	0	0	0	0
0	1	1	1	0
0	1	1	1	0
0	1	1	1	0
0	0	0	0	0

(a)

0	1	2	1	0
1	3	5	3	1
2	5	9	5	2
1	3	5	3	1
0	1	2	1	0

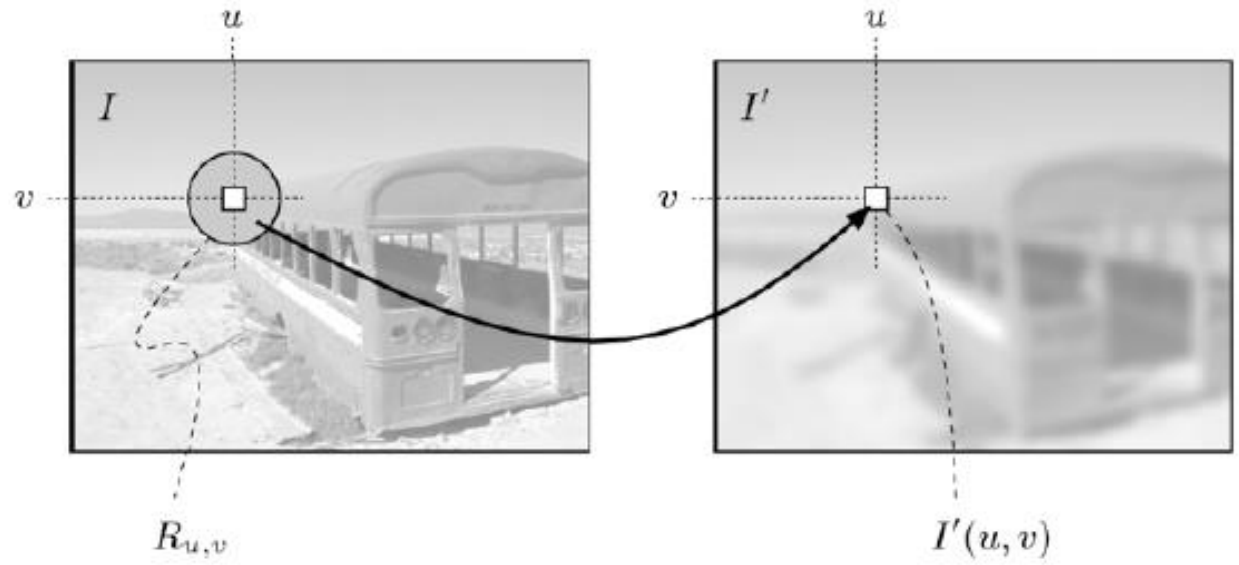
(b)

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

(c)

Mean filter

$$H(i,j) = \begin{bmatrix} 1/9 & 1/9 & 1/9 \\ 1/9 & 1/9 & 1/9 \\ 1/9 & 1/9 & 1/9 \end{bmatrix} = \frac{1}{9}\begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$



$$I'(u,v) \leftarrow \frac{p_0 + p_1 + p_2 + p_3 + p_4 + p_5 + p_6 + p_7 + p_8}{9}$$

Mean filter

Impulse noise

After mean filter

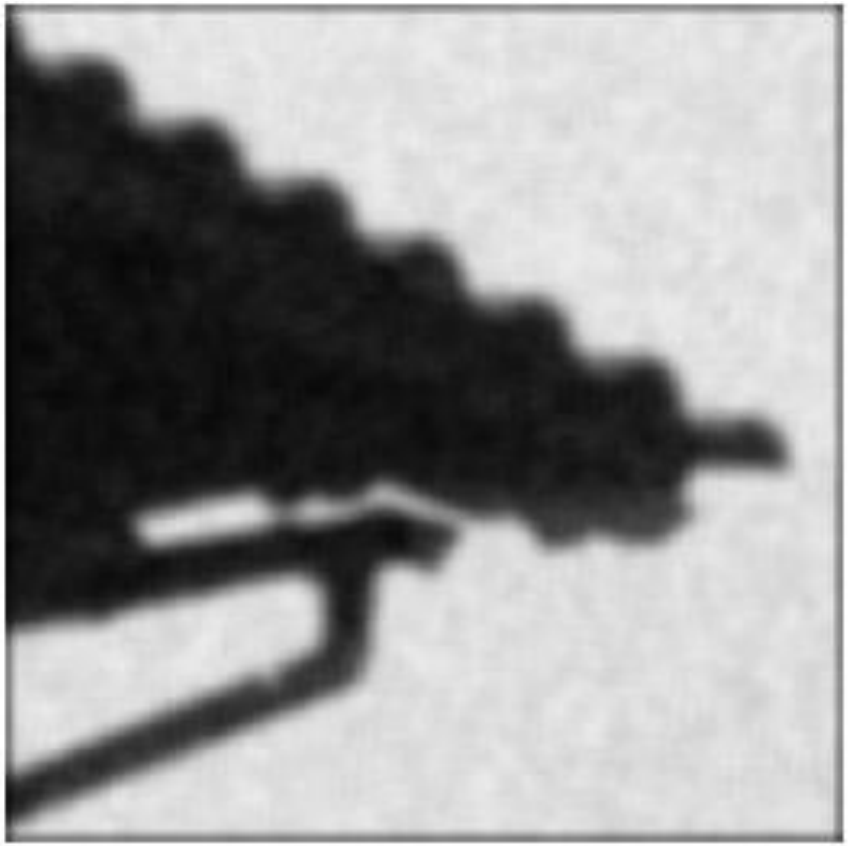
Mean filter

Gaussian noise

After mean filter

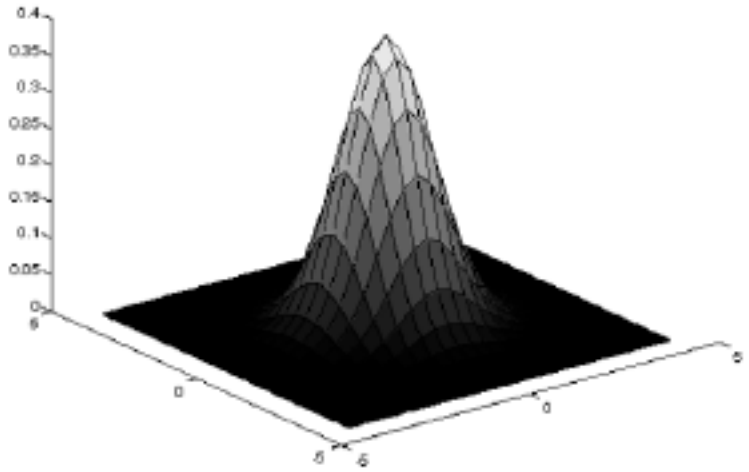
Mean filter

Real image

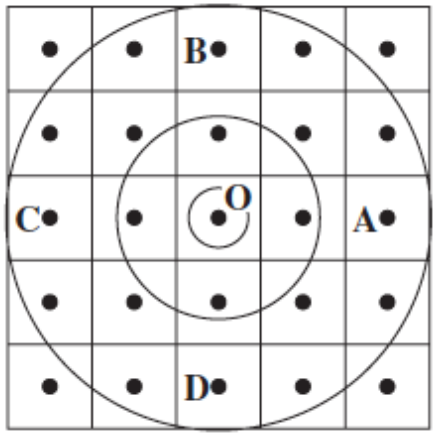


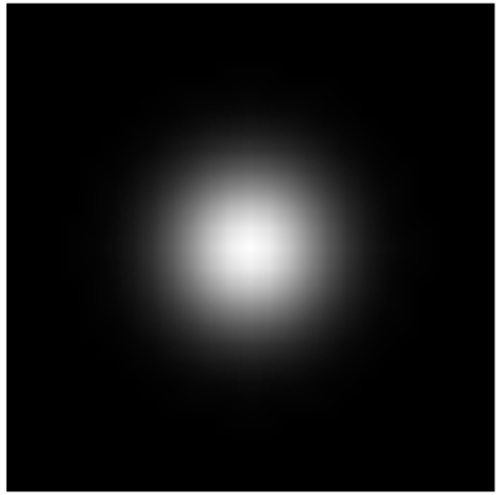
After mean filter

Gaussian filter



Define a function using Gaussian function





Definition of H

Gaussian filter

Real image

After Gaussian filter

Gaussian filter

Real image

After mean filter

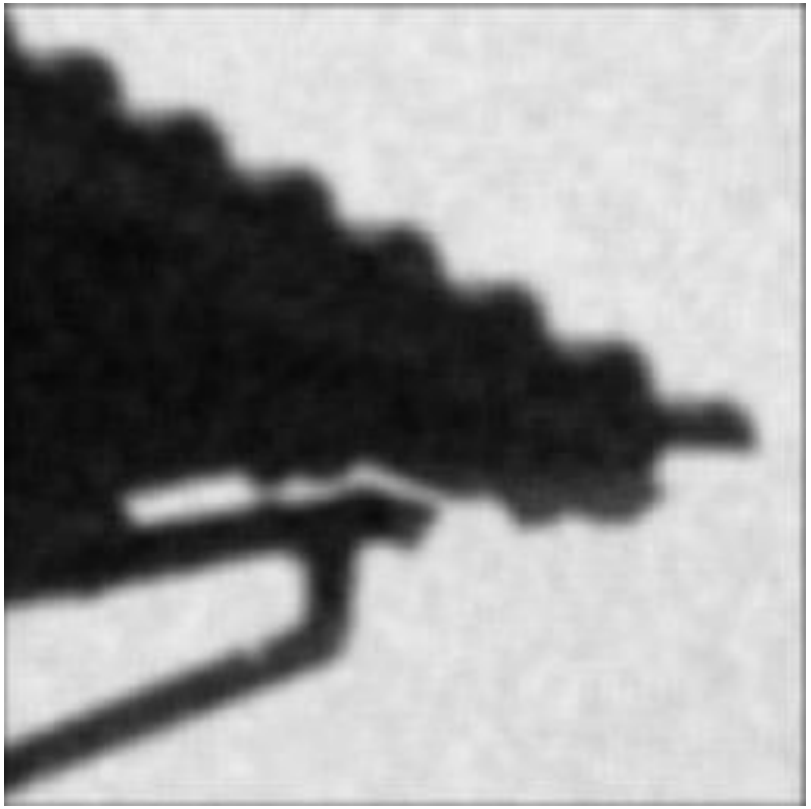
Gaussian filter

Real image

After Gaussian filter

Gaussian filter

Real image



After mean filter

Gaussian filter

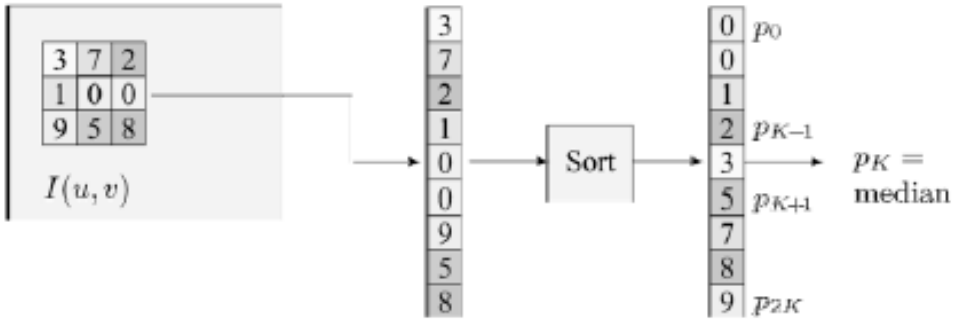
Real image

After Gaussian filter

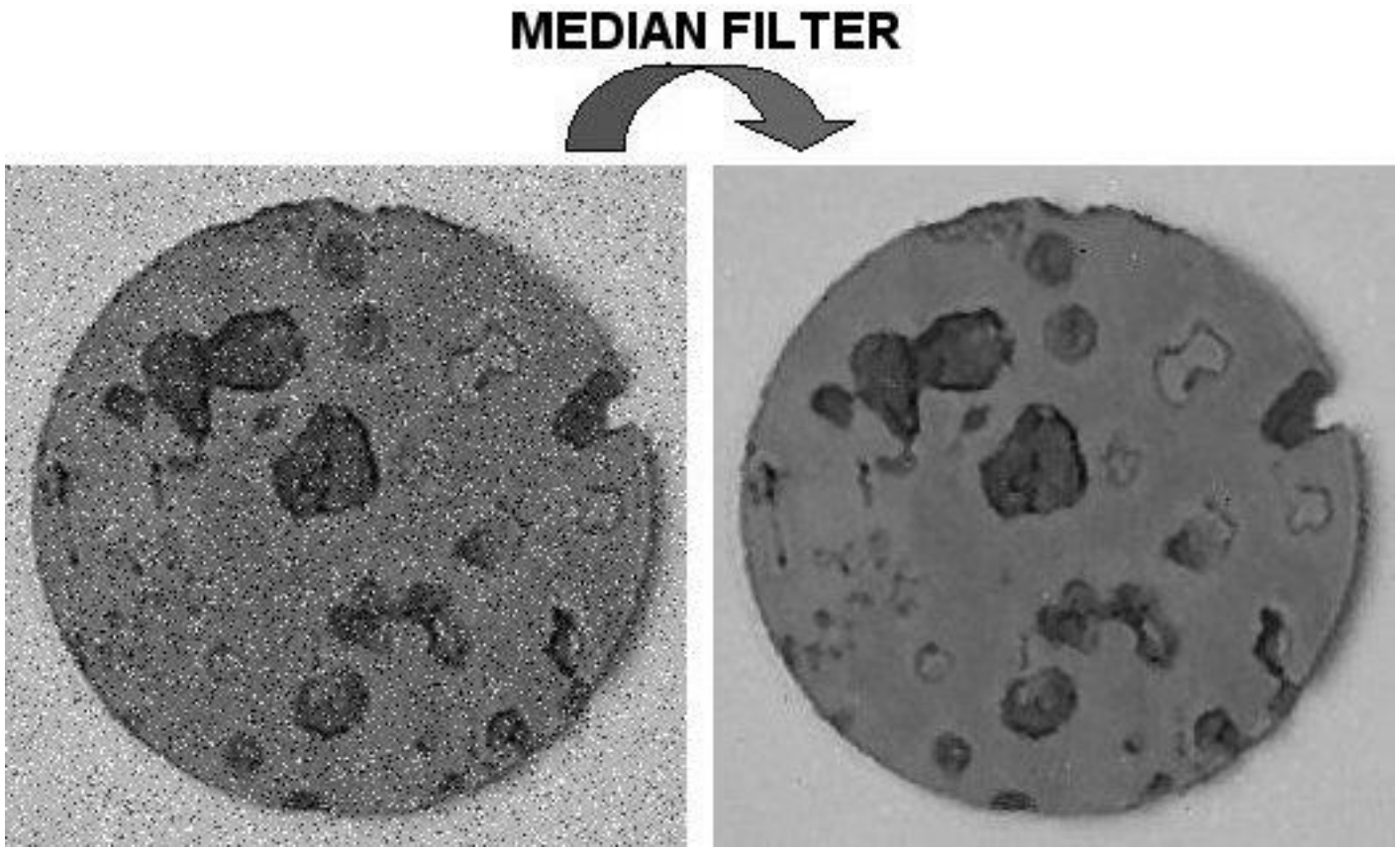
Median filter

- **Median**
  - **Nonlinear filter**
  - **Take median within a local window**

$$I'(u,v) \leftarrow \text{median}\,\{I(u+i, v+j) \mid (i,j) \in R\}$$



Median filter



Real image

After median filter

Median filter

Salt & Pepper

Mean filter

Median filter

Median filter

Noisy image

Median filter

Median filter

Original image

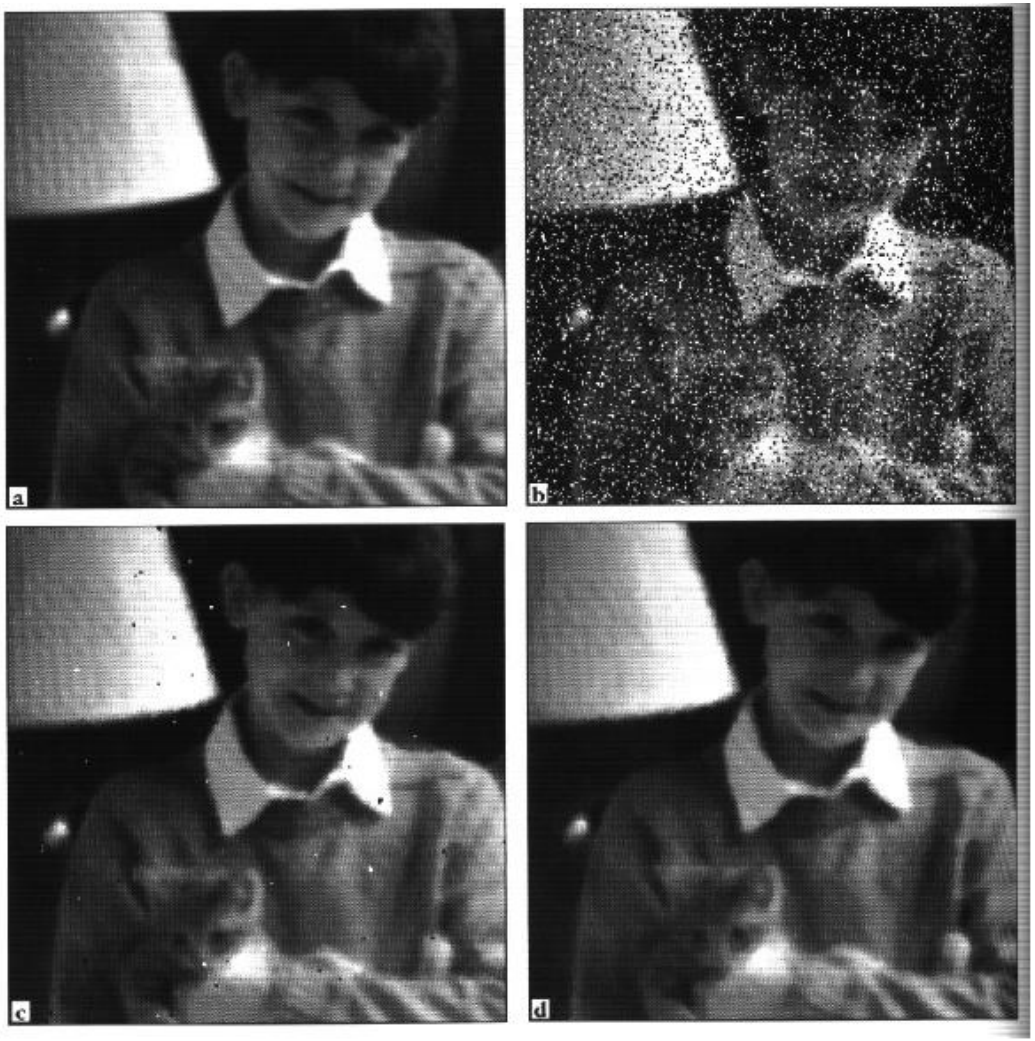


Image a with 10% of the pixels randomly selected and set to black, and another 10% randomly selected and set to white

Application of median filtering to image b using a 3x3 square region

Application of median filtering to image b using a 5x5 square region

Median filter

Noisy image

Median filter

Median filter

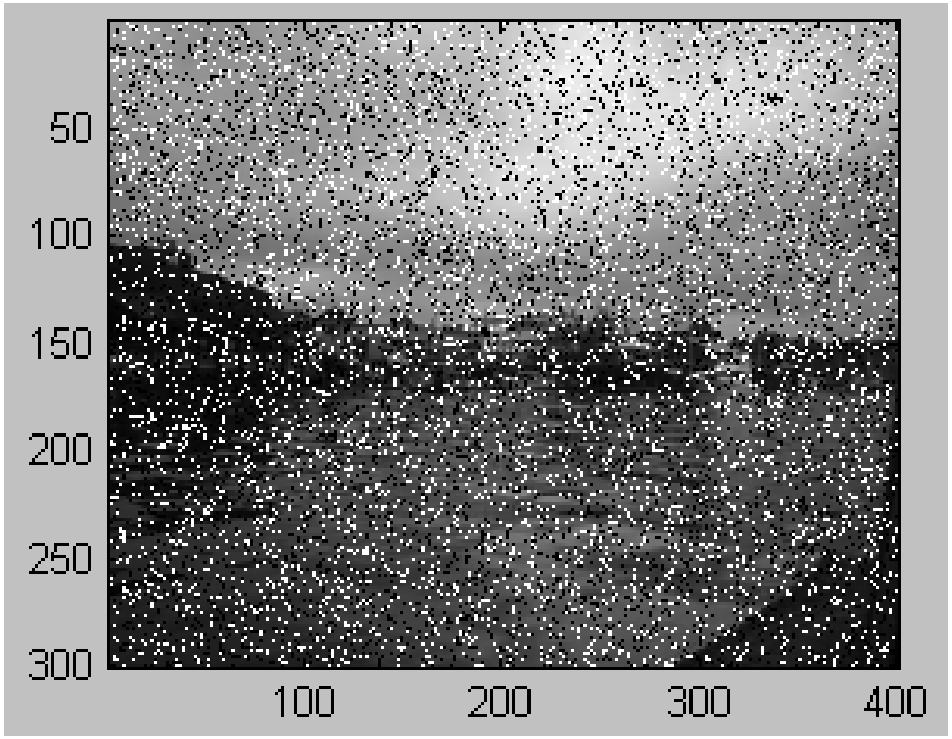
original

added noise

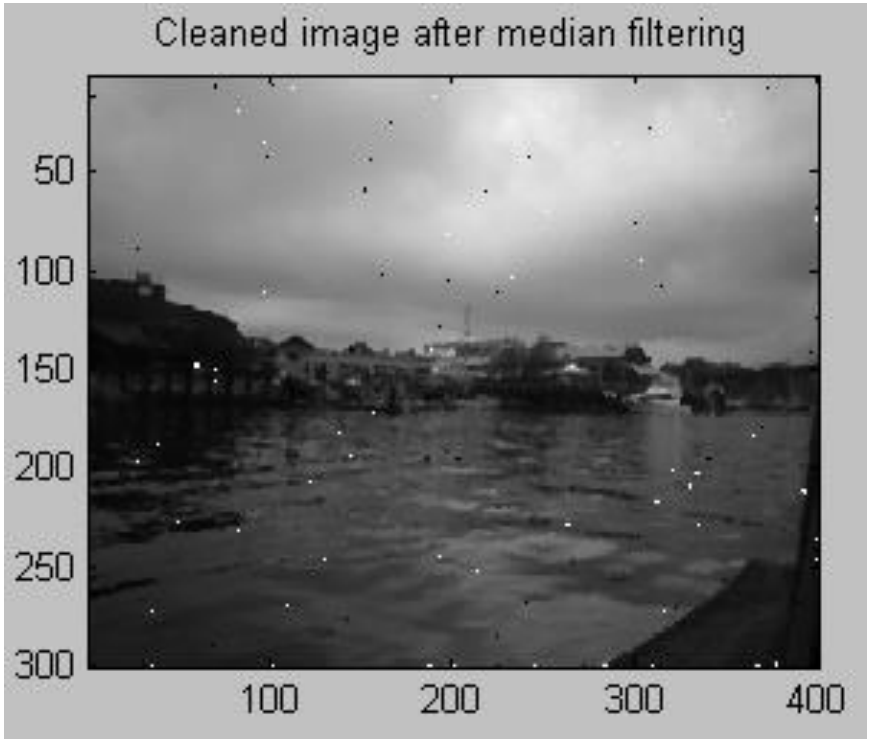
average

median

Median filter

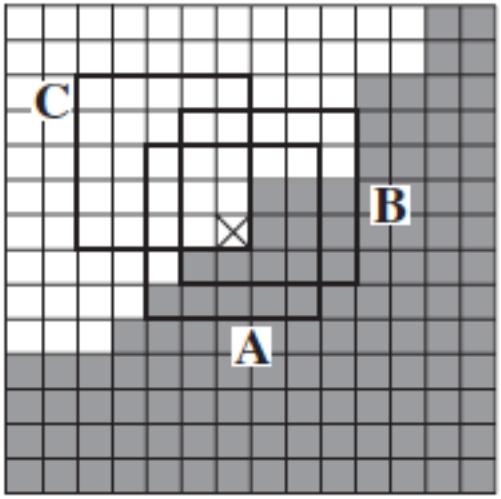


Noisy image
Can you guess what it is?



Median filter

Edge preserving filtering



Step 1: Consider all windows of fixed size around a pixel (not necessarily centered at that pixel)

Step 2: Find a window with the least variance

Step 3: Do a linear filter in that window.

Edge preserving filtering

(a) Original "Greek Flag"

(b) Flat filter

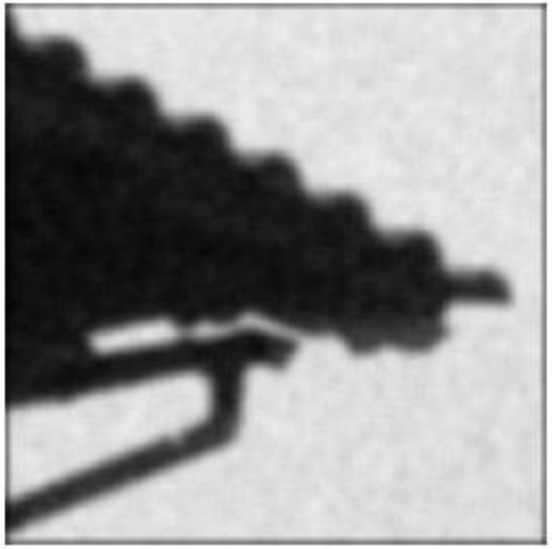
(c) Edge preserving flat

(d) Gaussian filter

(e)Edge preserving Gaussian

Edge preserving filtering

(f) Original "Roof Tiles"



(g) Flat filter

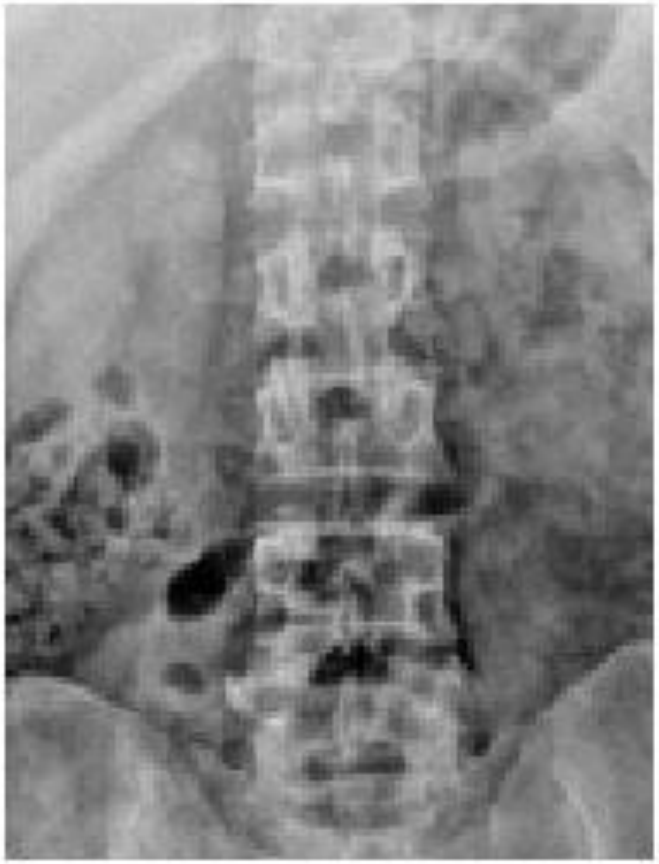
(h) Edge preserving flat

(i) Gaussian filter

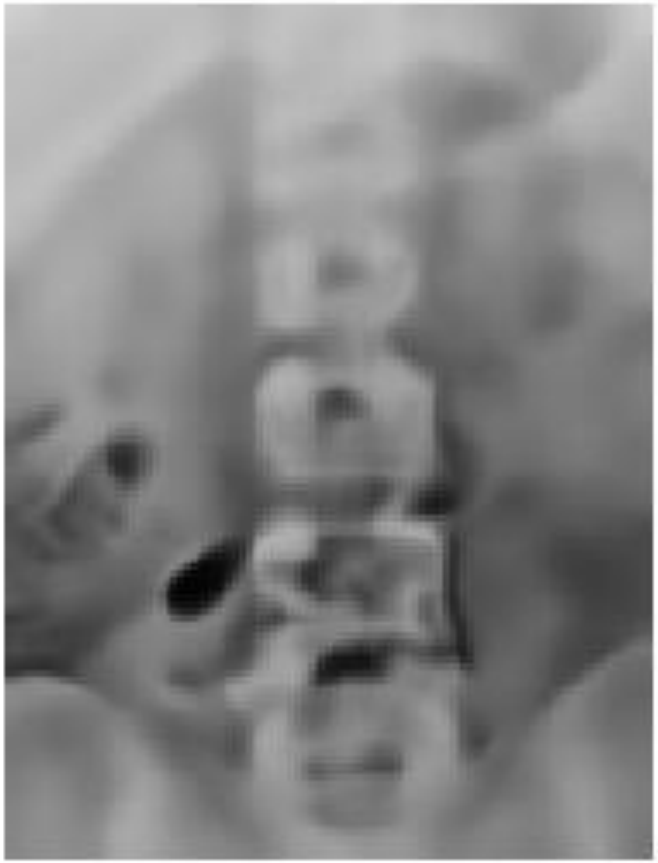
(j)Edge preserving Gaussian

Non-local mean filter

Noisy image



NL-means image



Non-local mean filter

noisy

non-local means

Non-local mean filter

Noisy image

Non-local mean

Non-local mean filter

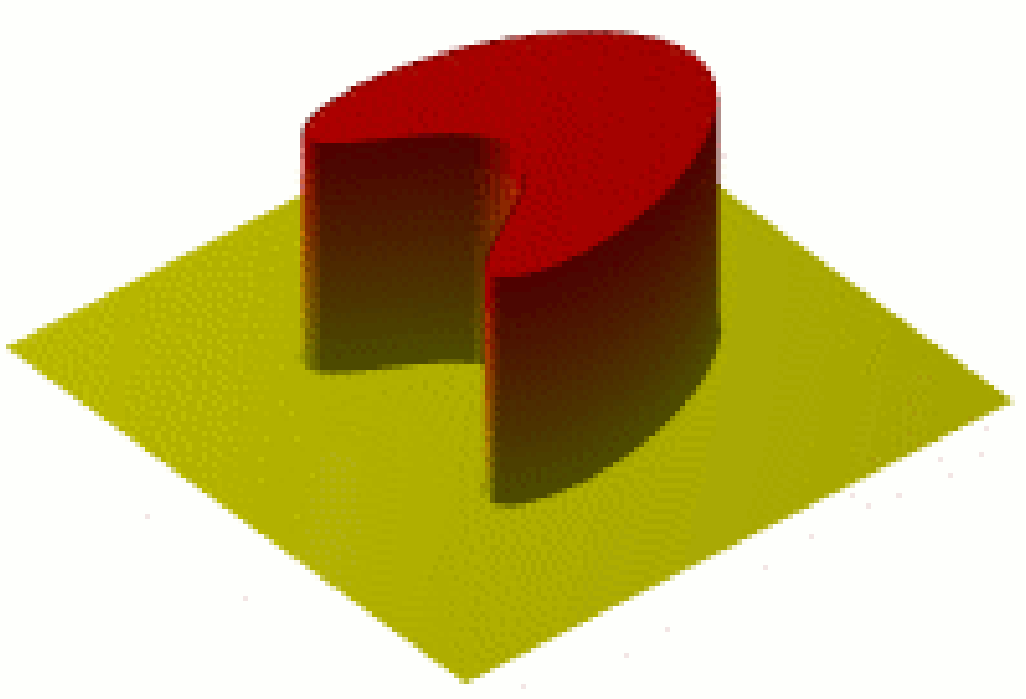
Flat filter

(a)

Non-local mean

(b)

Isotropic diffusion



Isotropic diffusion

Original image

Sigma = 1.98

Sigma = 4.28

Sigma = 8.24

Anisotropic diffusion

Seven iterations

Fourteen iterations

Twenty iterations