

Image filtering

In Fourier domain

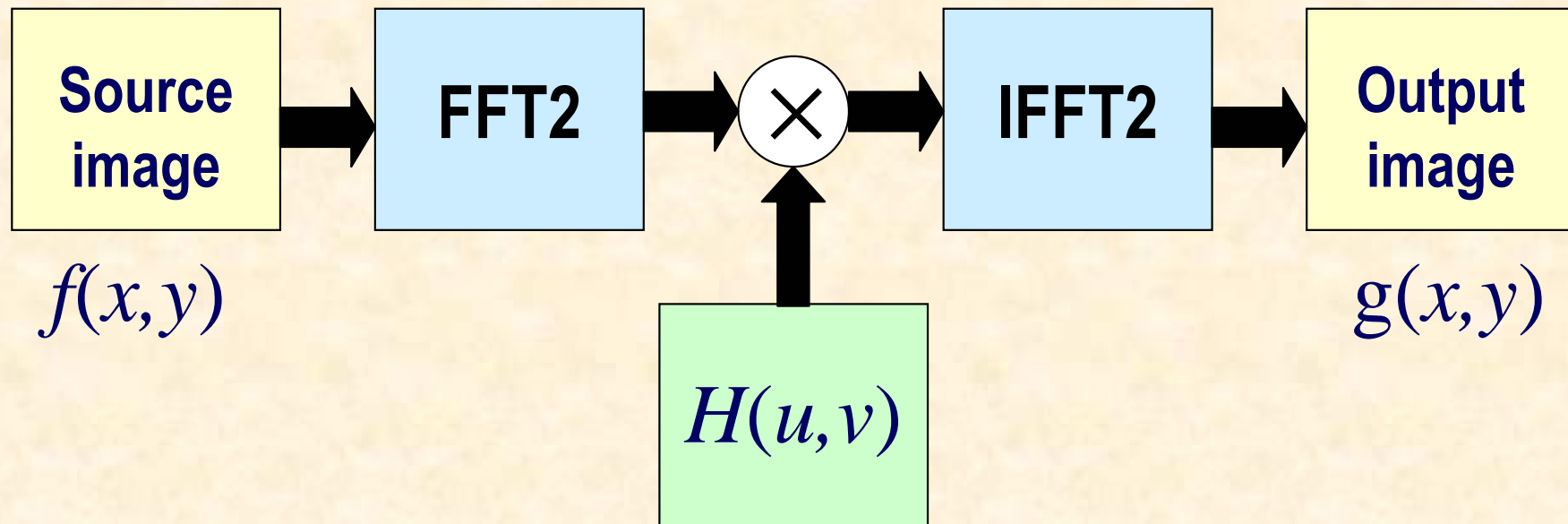
In spatial domain

Linear filters

Non-linear filters



Image filtering in spectrum domain

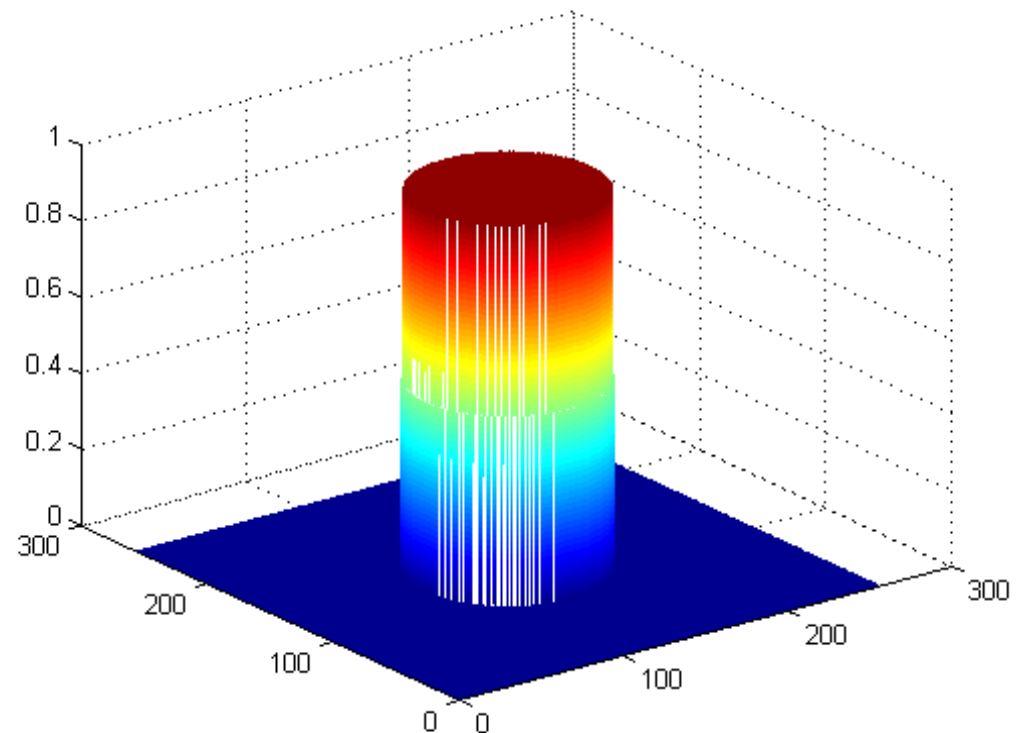
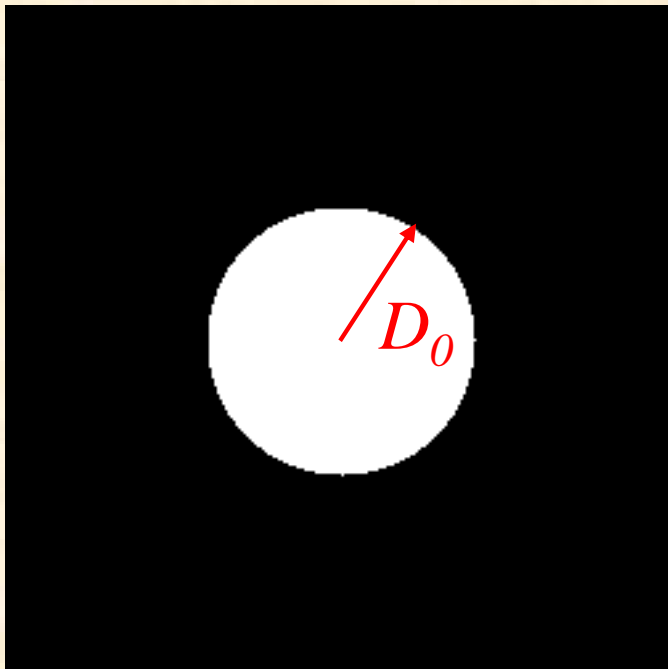


$$g(x,y) = IF\{ H(u,v) F\{f(x,y)\} \}$$

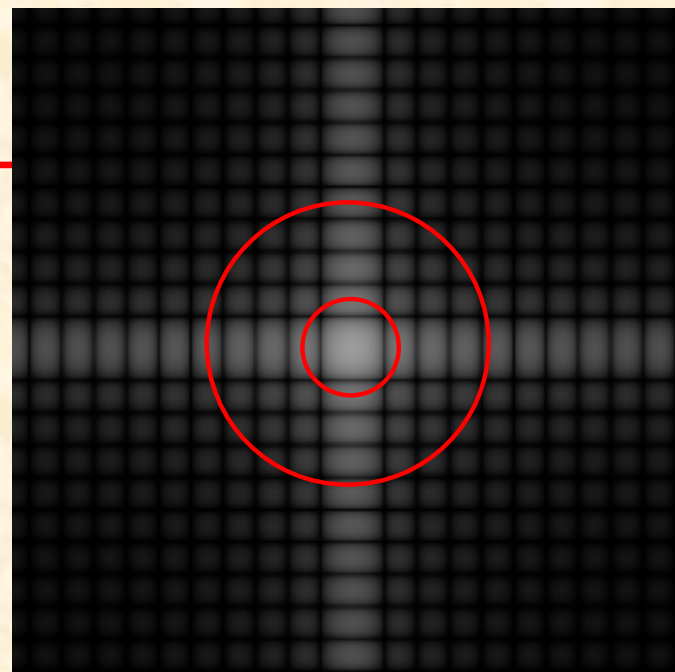
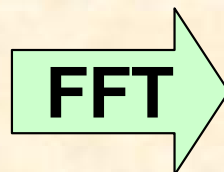
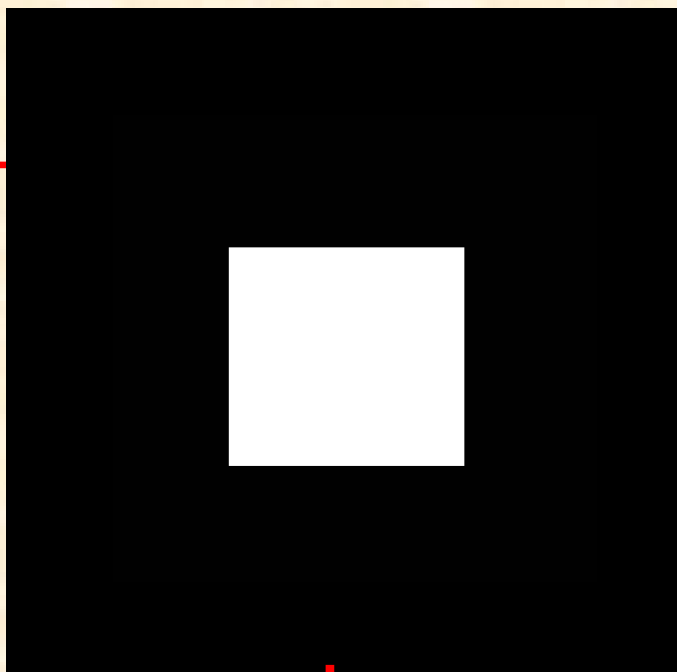
Ideal low-pass filter

$$H(u, v) = \begin{cases} 1 & D(u, v) \leq D_0 \\ 0 & D(u, v) > D_0 \end{cases}$$

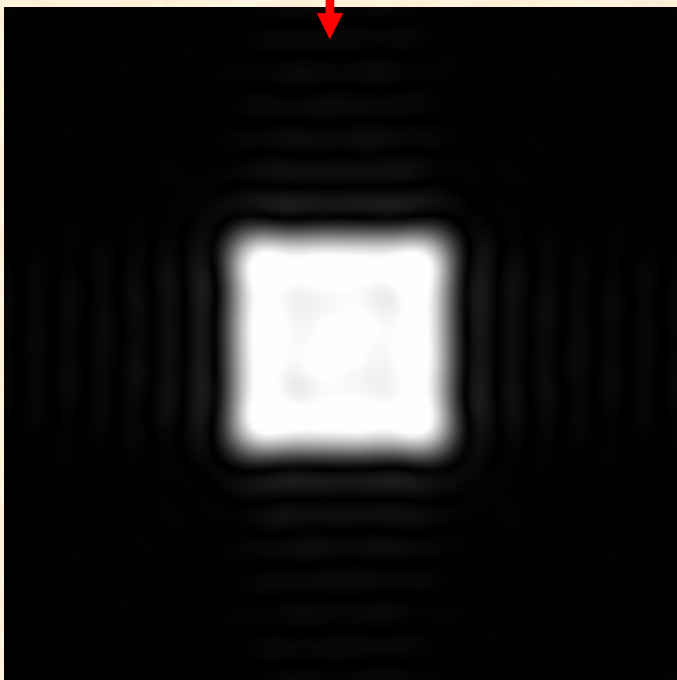
$$D(u, v) = \sqrt{u^2 + v^2}$$



256x256



$D_0=10$



$D_0=70$

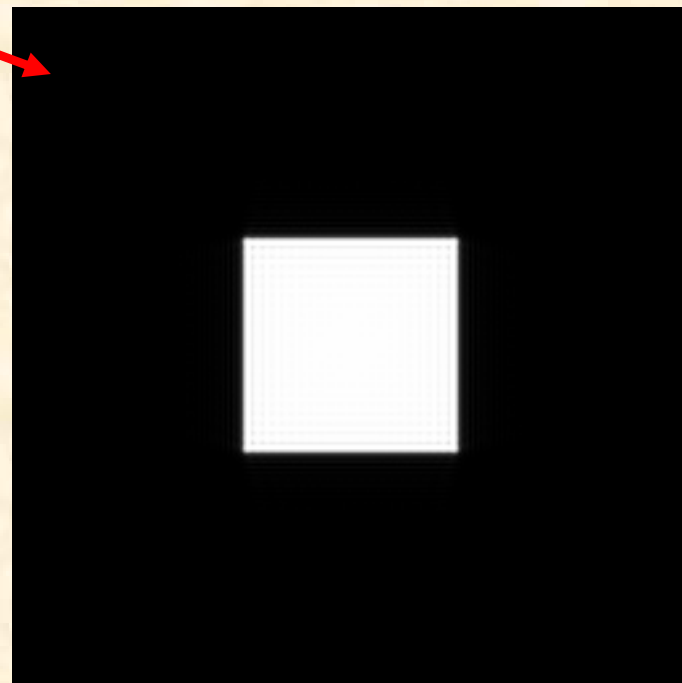


Image after ideal low-pass filtering, $D_0=70$

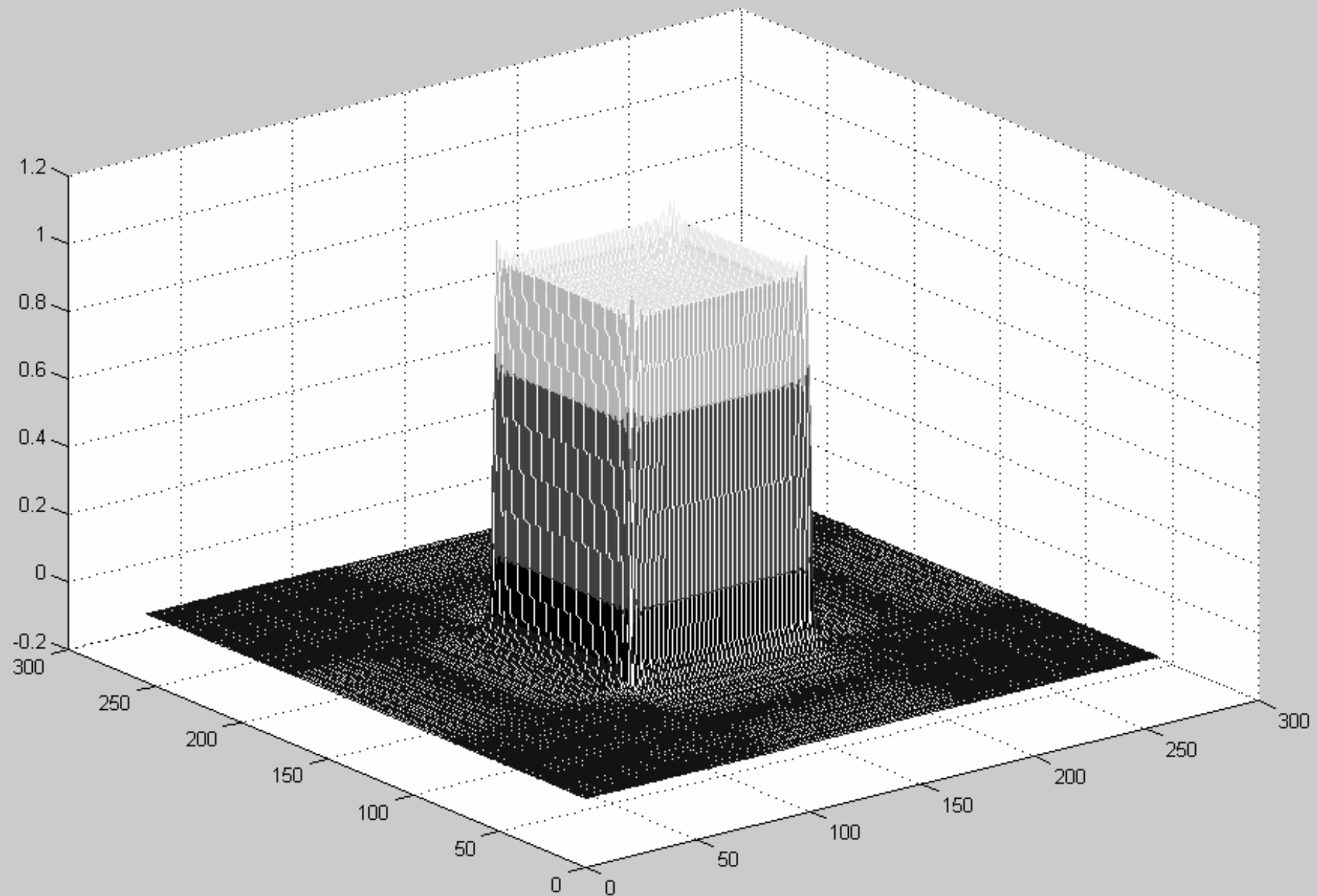
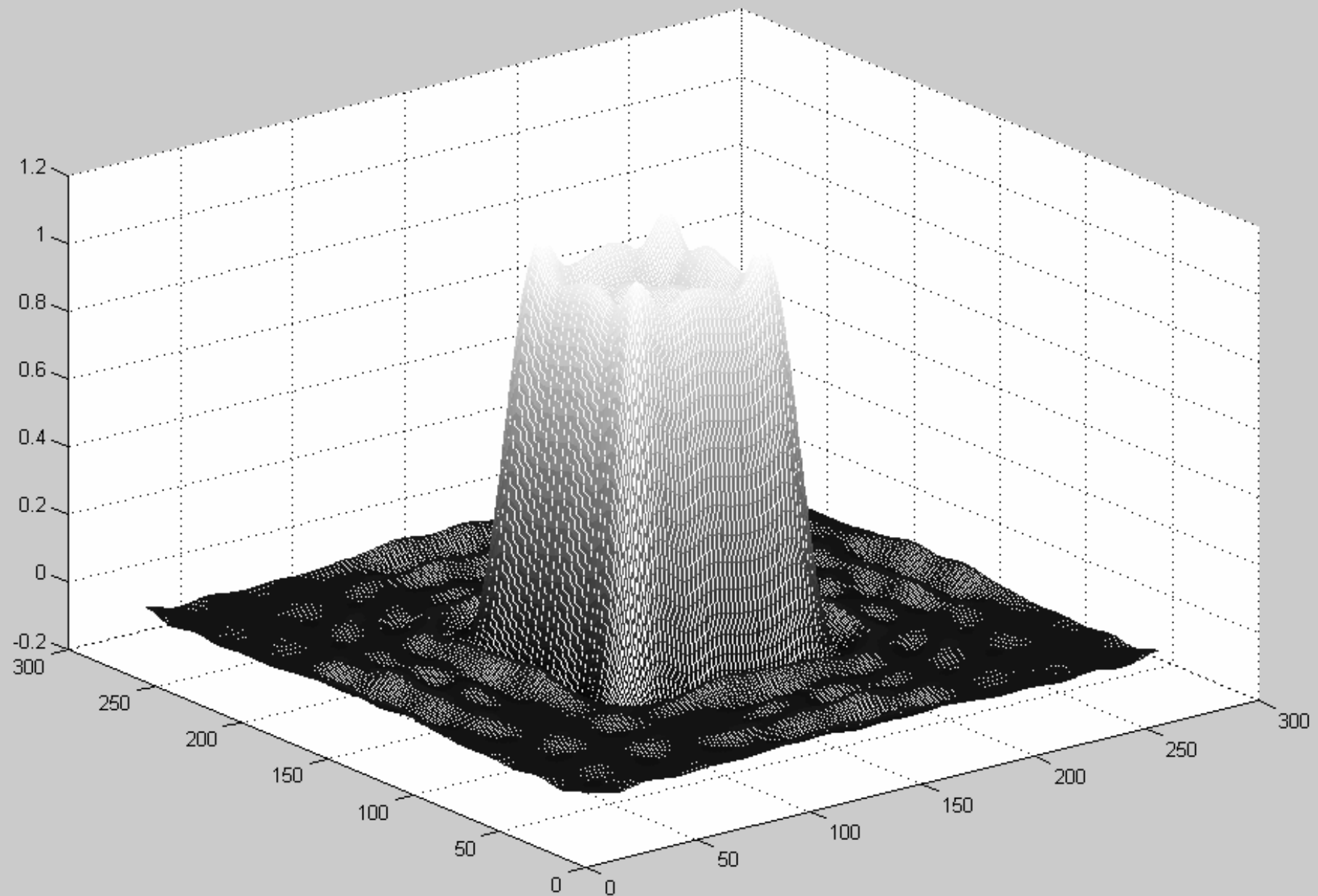


Image after ideal low-pass filtering, $D_0=10$



Ideal low-pass filter - example



$D_0=10$



$D_0=70$

ringing



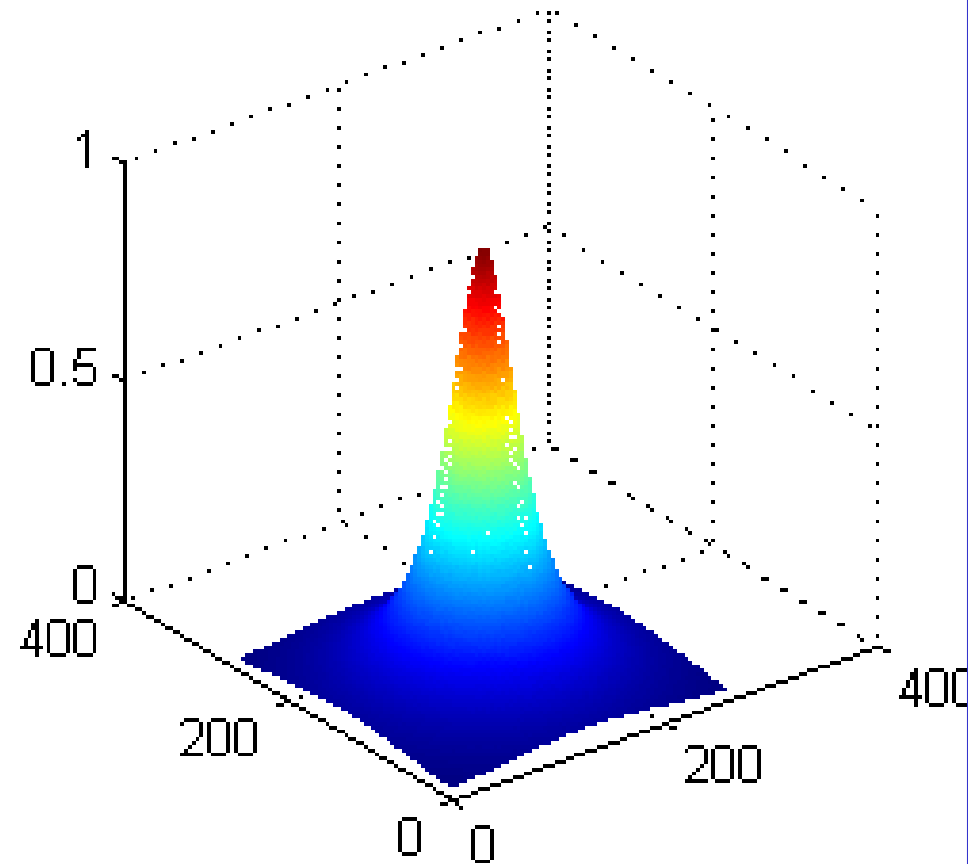
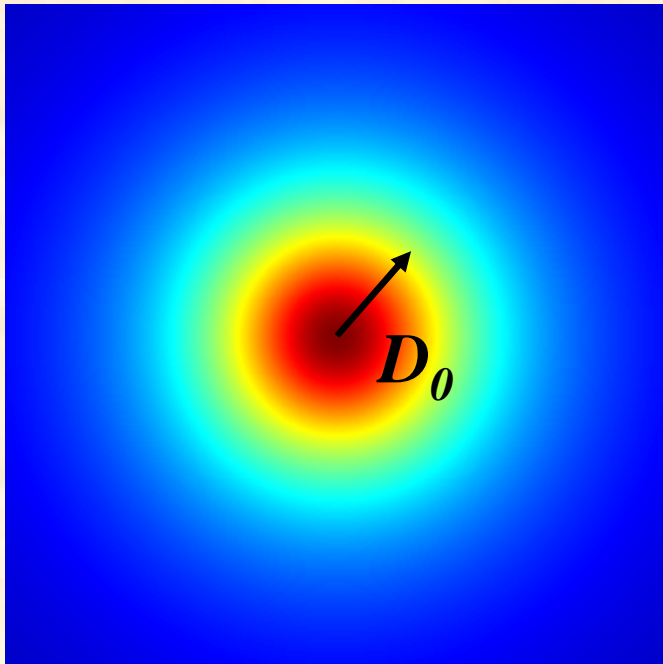
$D_0=30$

Butterworth filter

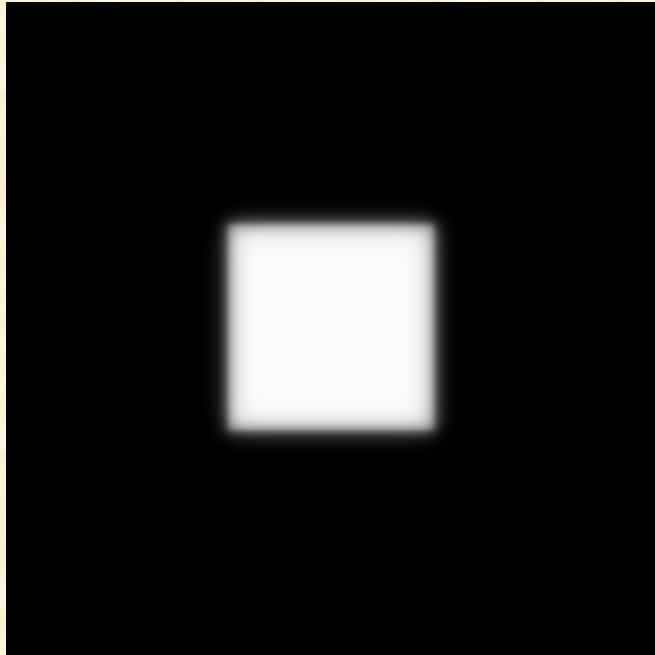
$$H(u, v) = \frac{1}{1 + (\sqrt{2} - 1) [D(u, v) / D_0]^{2n}}$$

$$D(u, v) = \sqrt{u^2 + v^2}, \quad n = 1, 2, \dots$$

n- filter order

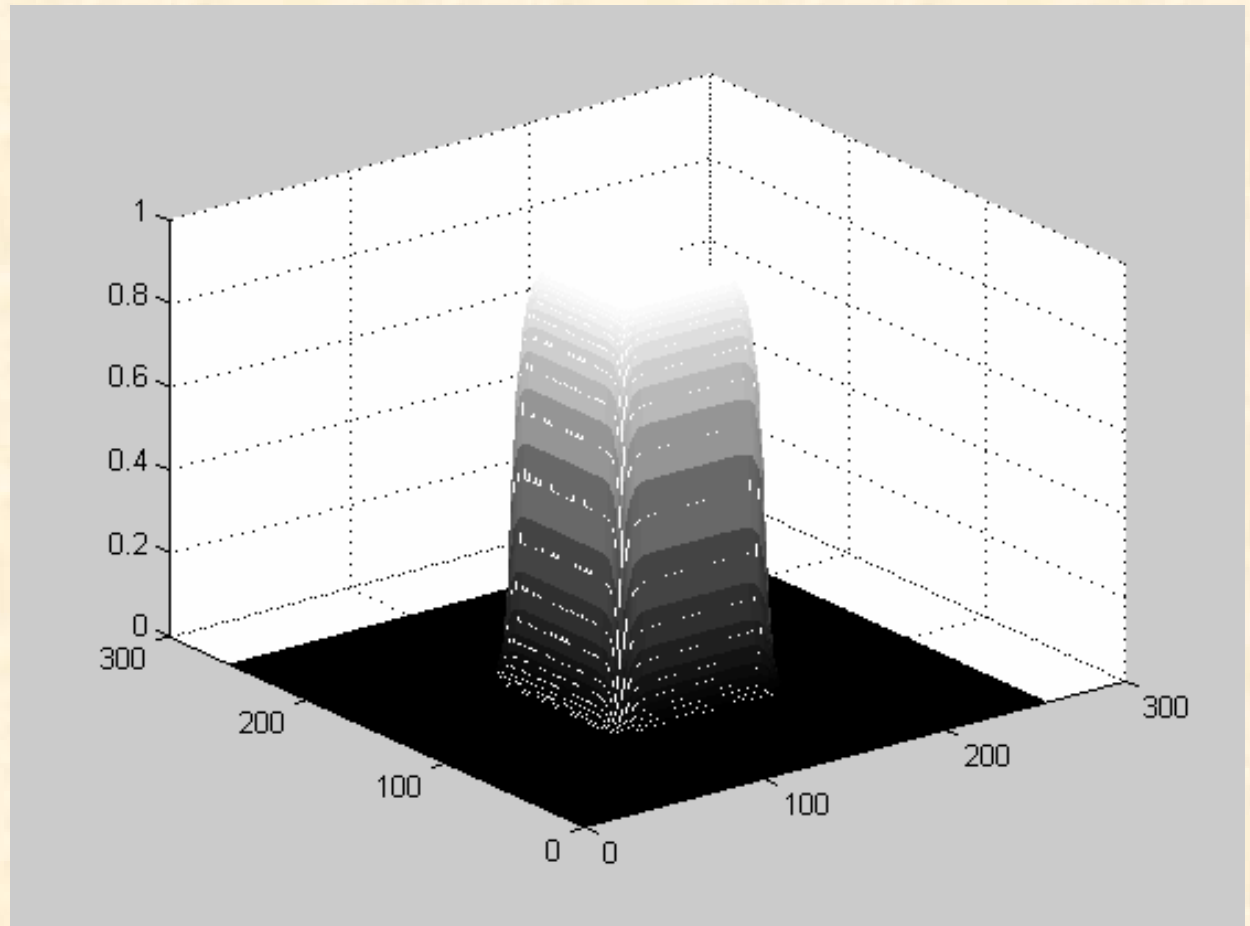


Low-pass filtering



$n = 1$

Butterworth filter



Low-pass Butterworth filter - examples

$n = 1$



$D_0=10$



$D_0=70$

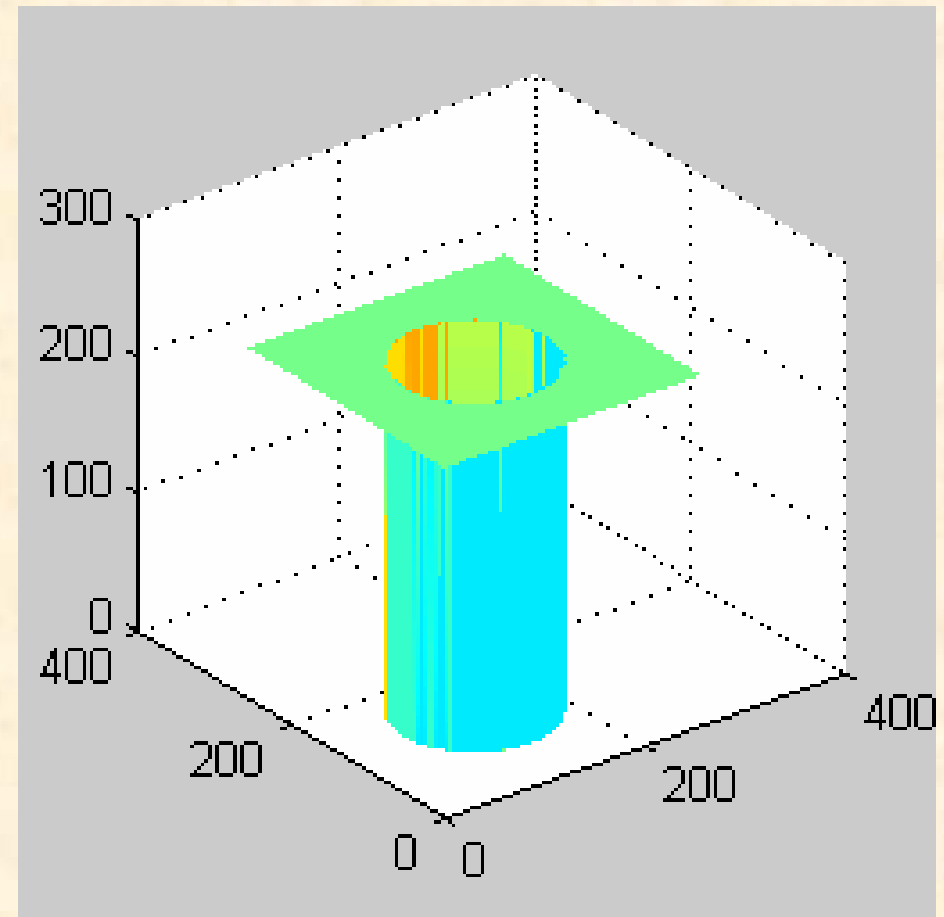
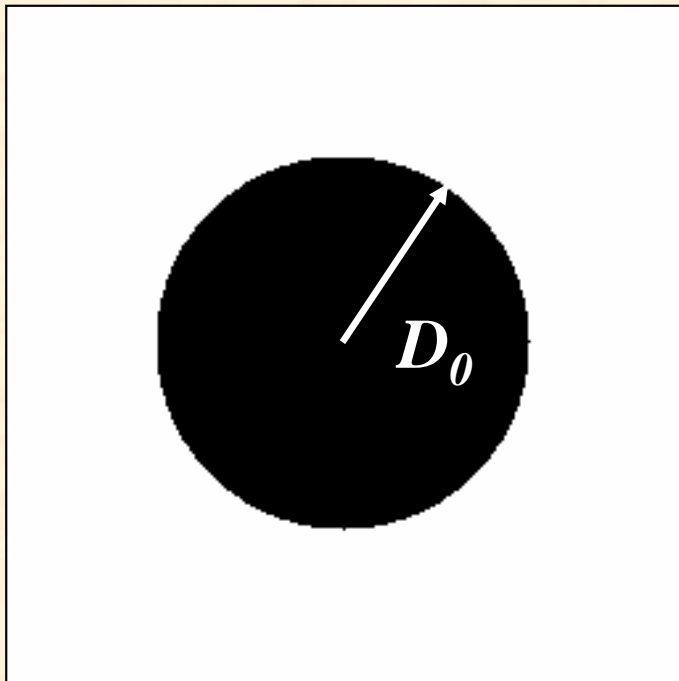


$D_0=30$

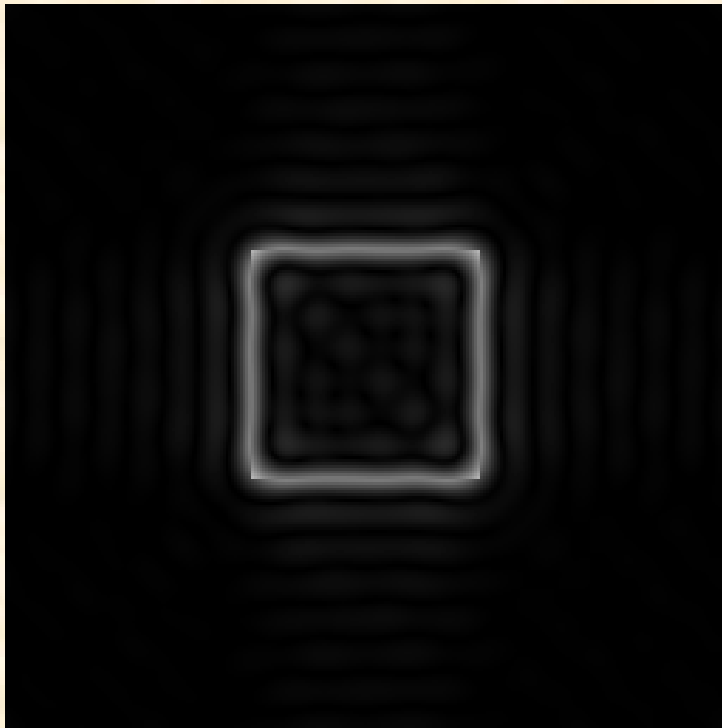
Ideal high-pass filter

$$H(u, v) = \begin{cases} 0 & D(u, v) \leq D_0 \\ 1 & D(u, v) > D_0 \end{cases}$$

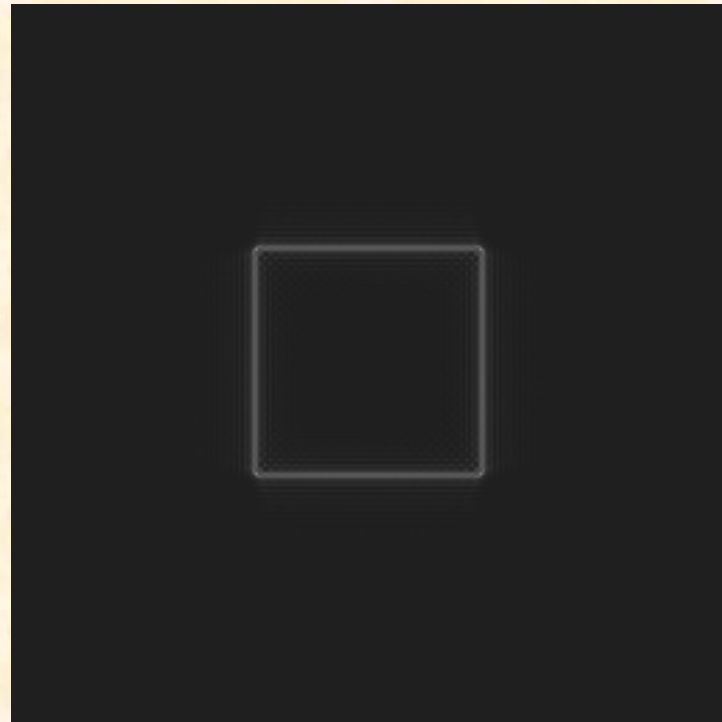
$$D(u, v) = \sqrt{u^2 + v^2}$$



Ideal high-pass filter - example

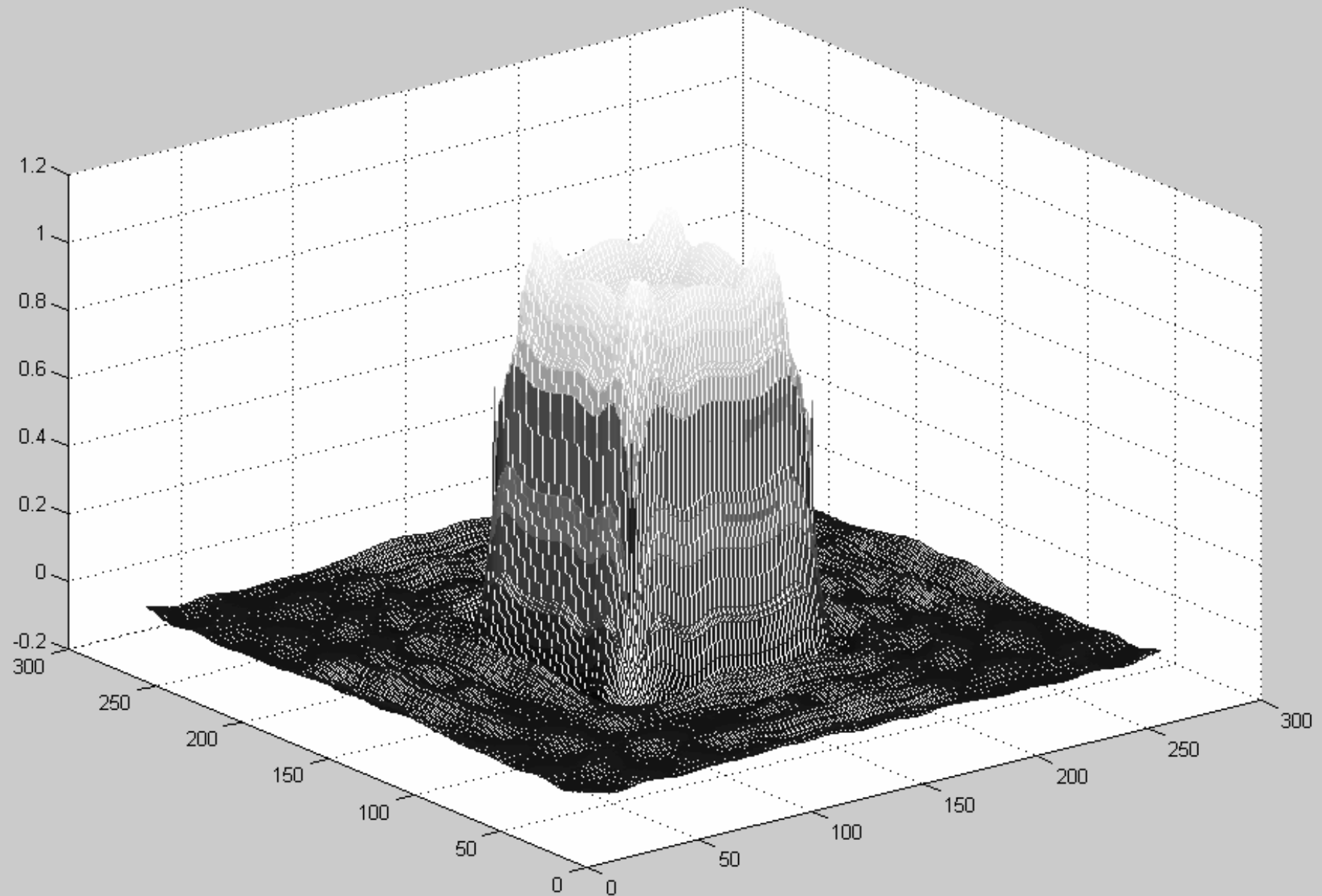


$D_0=10$



$D_0=70$

Image after ideal high-pass filtering, $D_0=10$

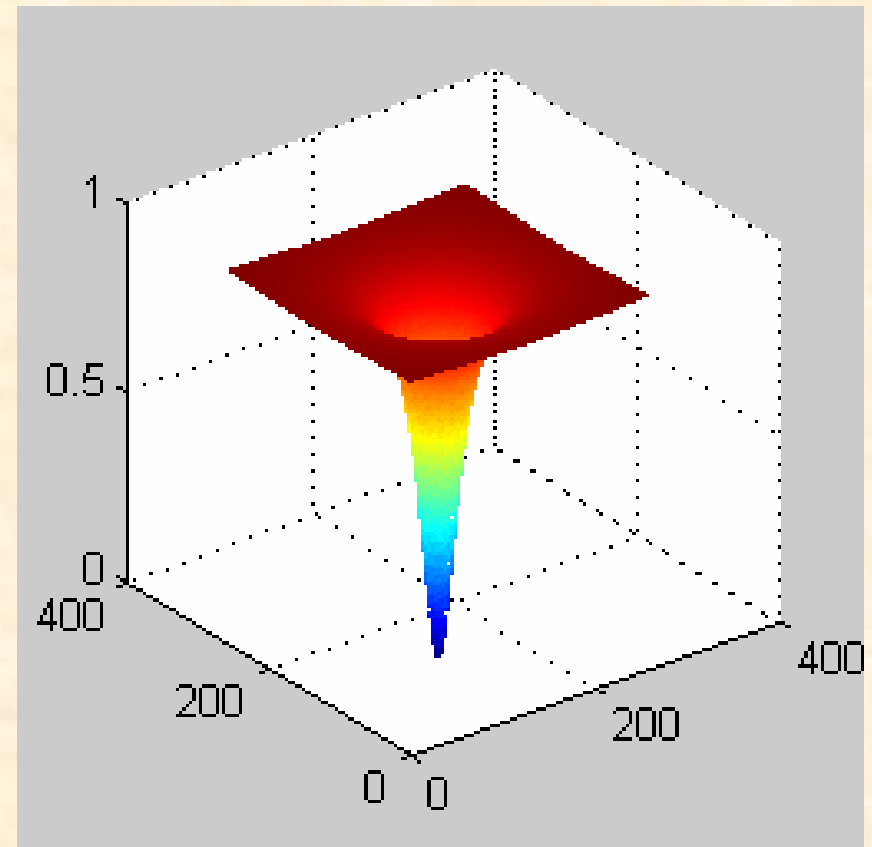
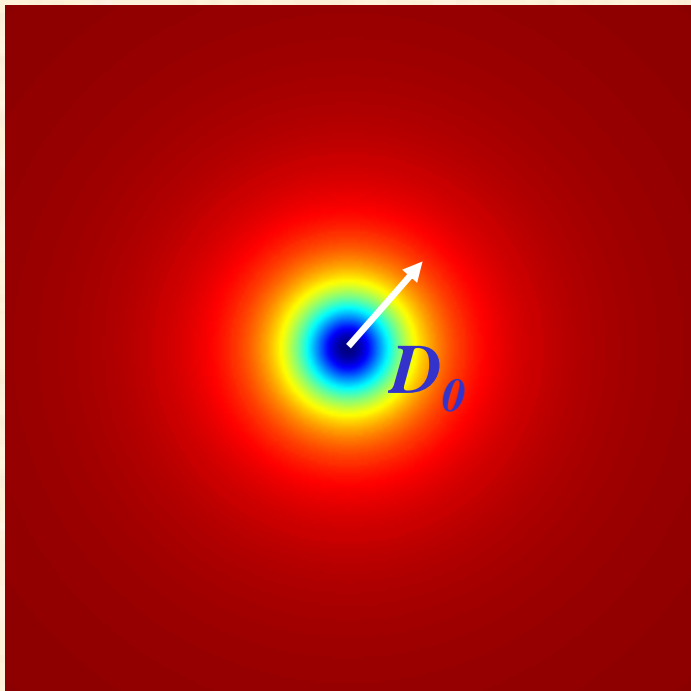


High-pass Butterworth filter 2

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

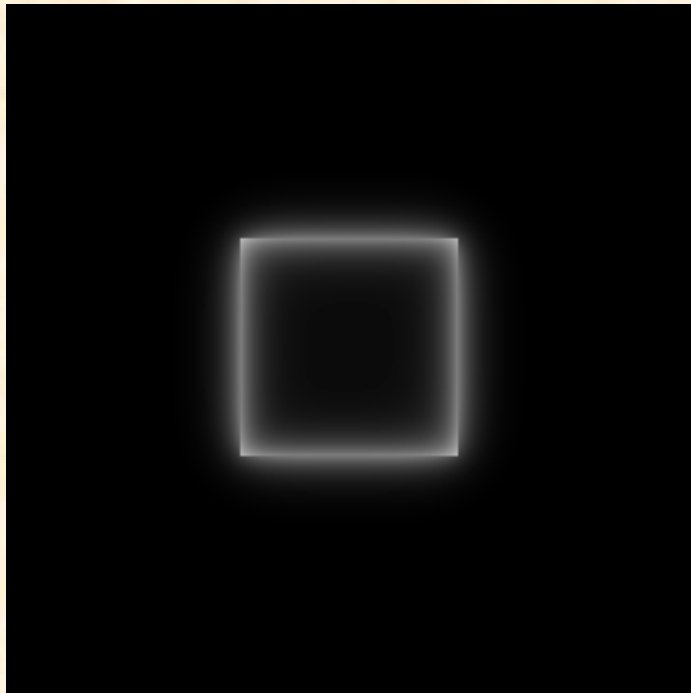
n - filter order

$$D(u, v) = \sqrt{u^2 + v^2} \quad n = 1, 2, \dots$$

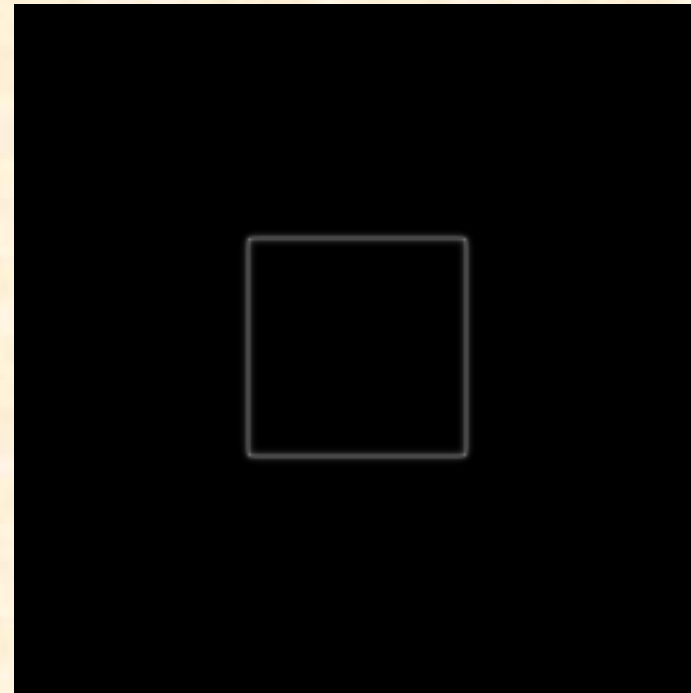


High-pass Butterworth filter - examples

$n = 1$

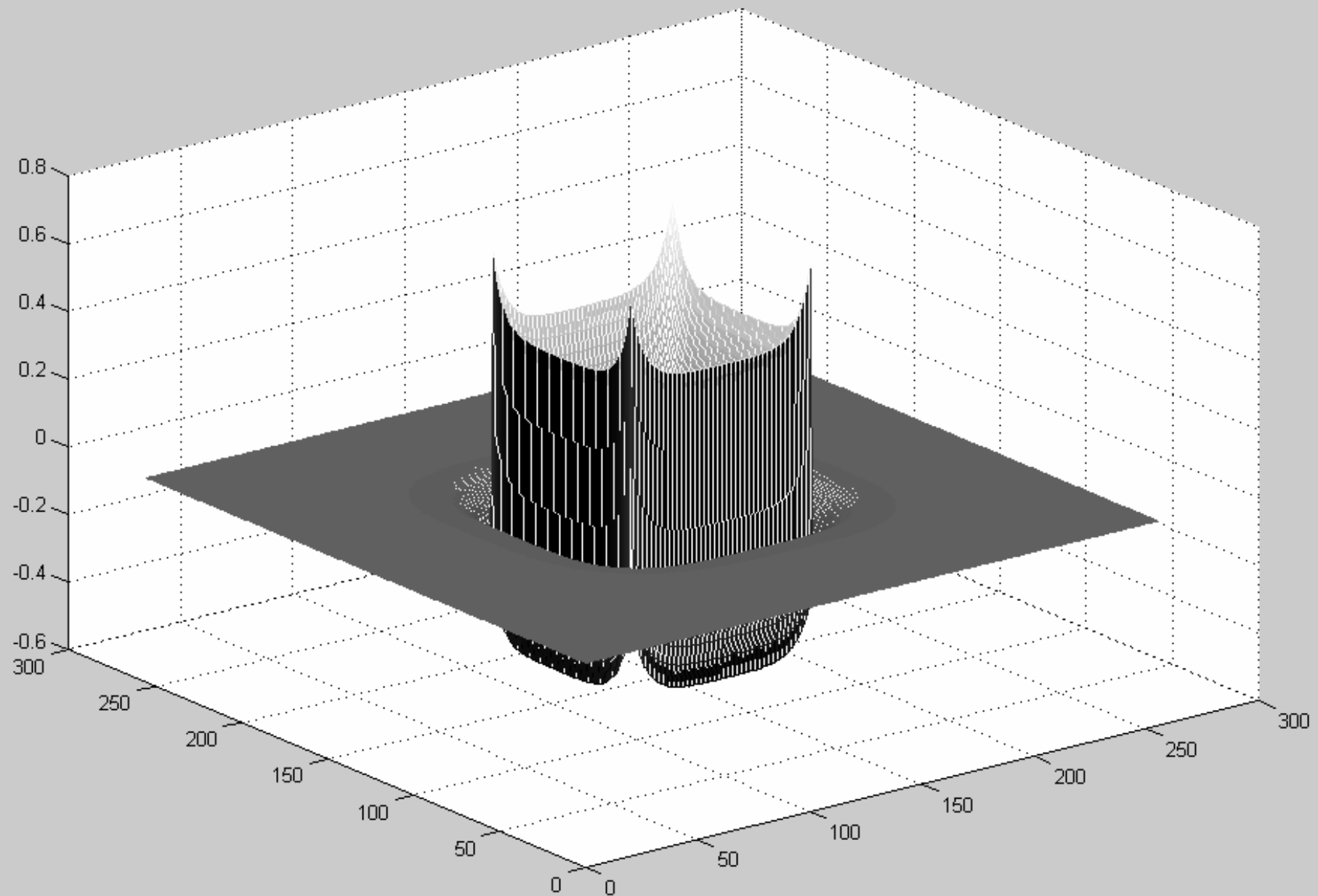


$D_0 = 10$



$D_0 = 70$

Image after Butterworth high-pass filtering



High-pass Butterworth filter - examples

$$n = 1$$

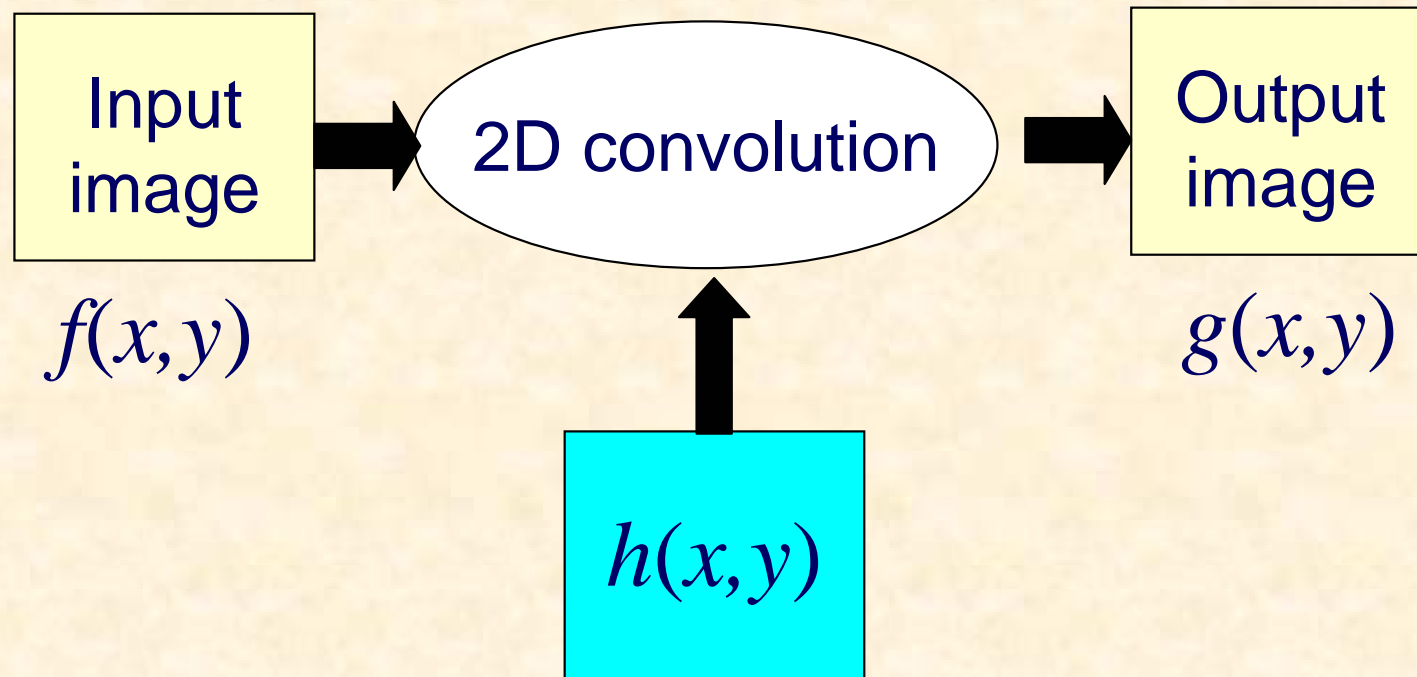


$$D_0=10$$



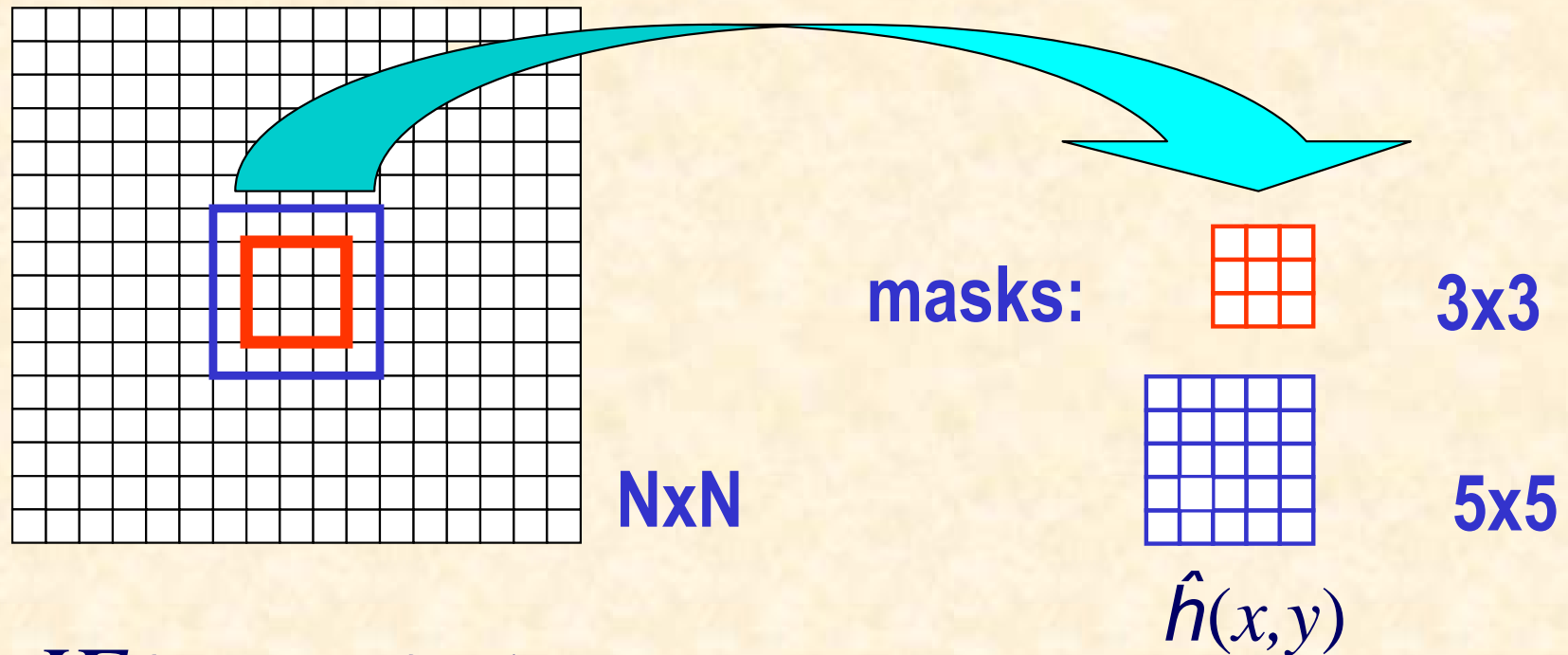
$$D_0=70$$

Image filtering in spatial domain



$$\begin{aligned} g(x,y) &= IF \{ H(u,v) F\{f(x,y)\} \} = \\ &IF \{ H(u,v) \} ** IF \{ F \{f(x,y)\} \} = \\ &h(x,y) ** f(x,y) \end{aligned}$$

Filter definition in spatial domain



$$IF\{H(u,v)\}=h(x,y)$$

\hat{h} is selected so that $F(\hat{h}(x,y)) = \hat{H}(x,y) \approx H(x,y)$

Image and the filter mask convolution

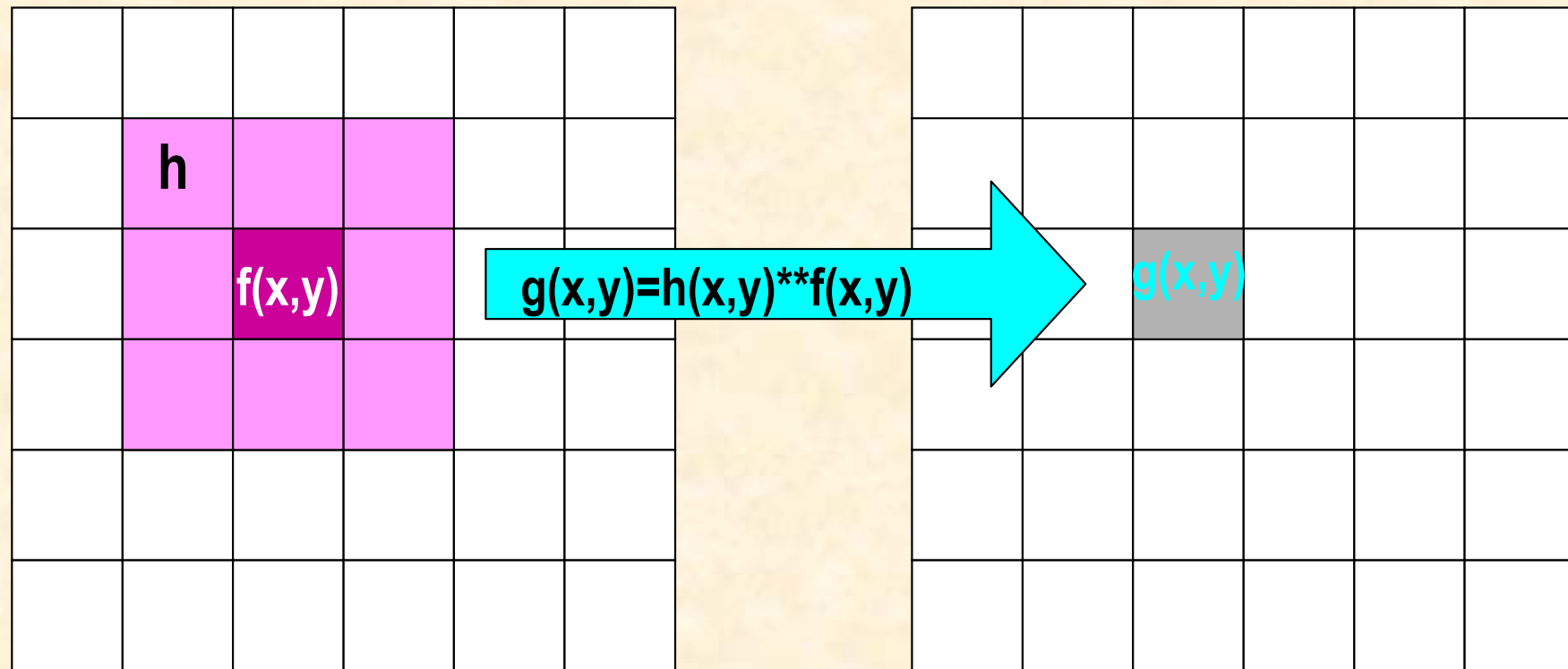
		$h(-1,-1)$	$h(-1,0)$	$h(-1,1)$
		$l(i-1,j-1)$	$l(i-1,j)$	$l(i-1,j+1)$
		$h(0,-1)$	$h(0,0)$	$h(0,1)$
		$l(i,j-1)$	$l(i,j)$	$l(i,j+1)$
		$h(1,-1)$	$h(1,0)$	$h(1,1)$
		$l(i+1,j-1)$	$l(i+1,j)$	$l(i+1,j+1)$

$G(i,j) =$

$$\begin{aligned} & l(i-1,j-1)h(-1,-1) + l(i-1,j)h(-1,0) + l(i-1,j+1)h(-1,1) + \\ & l(i,j-1)h(0,-1) + \mathbf{l(i,j)h(0,0)} + l(i,j+1)h(0,1) + \\ & l(i+1,j-1)h(1,-1) + l(i+1,j)h(1,0) + l(i+1,j+1)h(1,1) \end{aligned}$$

This is true for symmetric masks only !

Computing the filtered image



source image f

output image g

Boundary effects

h						
	f(x,y)					

source image *f*

?					

output image *g*

Boundary effects – 3x3 mask

h					
	f(x,y)				

Boundary columns and rows of (N x N) image are neglected and the filtered image is of size (N-2) x (N-2)

Image filtering – the algorithm

```
f, g : array[0..N-1, 0..N-1] of byte;  
{ size2 – half size of the mask}  
h : array[-size2..size2,-size2..size2] of integer;  
...  
for i:=1 to N-2 do for j:=1 to N-2 do  
    begin  
        g[i,j]:=0;  
        for k:=-size2 to size2 do for l:=-size2 to size2 do  
            g[i,j]:=g[i,j] + f[i+k,j+l] * h[i+k,j+l];  
        end;  
    end;  
end;
```

...

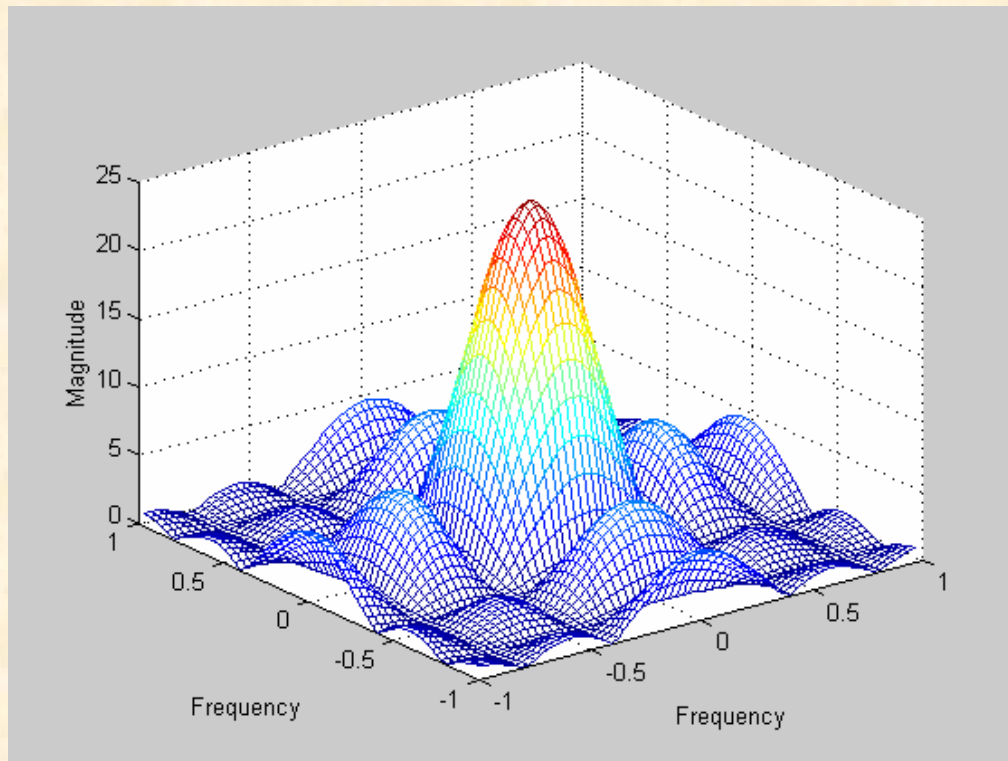
Range check g[i,j] !!!

Low pass filter

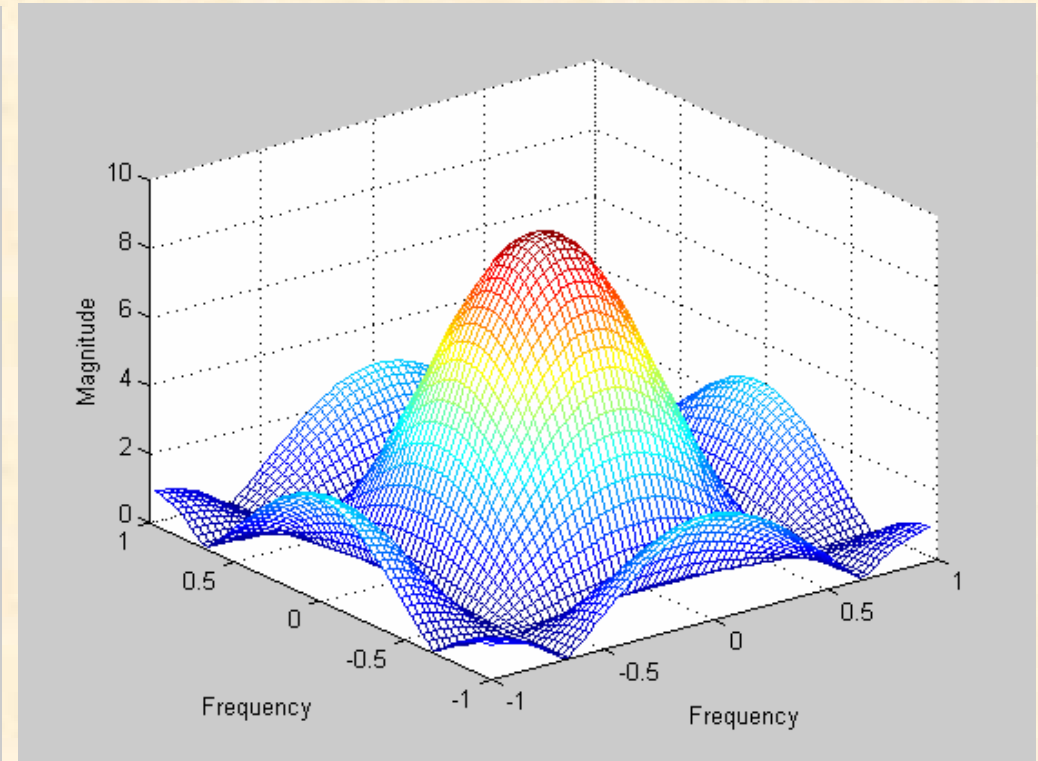
$$h_1 = \frac{1}{9} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \quad h_2 = \frac{1}{25} \begin{bmatrix} 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \end{bmatrix}$$

Can one use mask of even size ?

Frequency characteristics of low pass filters



for 5x5 mask



for 3x3 mask

Low-pass filtering the image



Source image

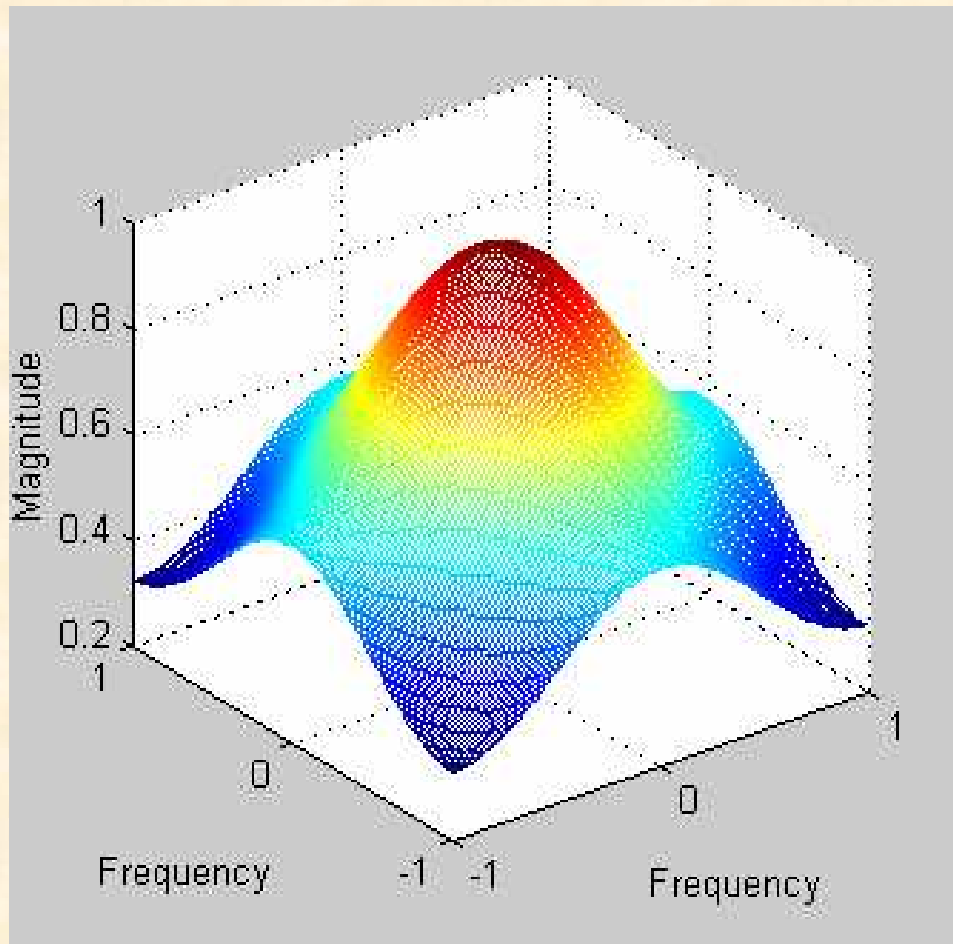


3x3 mask



5x5 mask

Gaussian filter



$$h = \begin{bmatrix} 1 & 2 & 1 \\ 2 & 4 & 2 \\ 1 & 2 & 1 \end{bmatrix}$$

$$h(x, y) = e^{\frac{-\pi(x^2 + y^2)}{d_0^2}}$$

$$H(u, v) = e^{\frac{-\pi d_0^2(u^2 + v^2)}{N}}$$

Image filtering using the Gaussian filter



source image



filtered image

Image low-pass filters - examples



Image distorted by the
Gaussian noise $N(0, 0.01)$

for grayscale
 $<0,1>$



Low pass filter 3x3



Gaussian filter 3x3

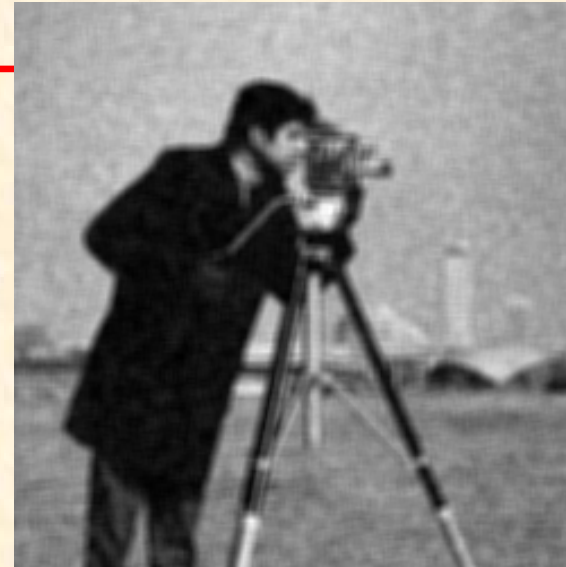


Butterworth filter $D_0=50$

Image low-pass filters - examples



Image distorted by the
Gaussian noise $N(0, 0.01)$



low-pass filter 5x5



Gaussian filter 5x5



Butterworth filter $D_0=30$

Image low-pass filters - examples



Image distorted by the
Gaussian noise $N(0, 0.002)$



Low pass filter 3x3



Gaussian filter 3x3

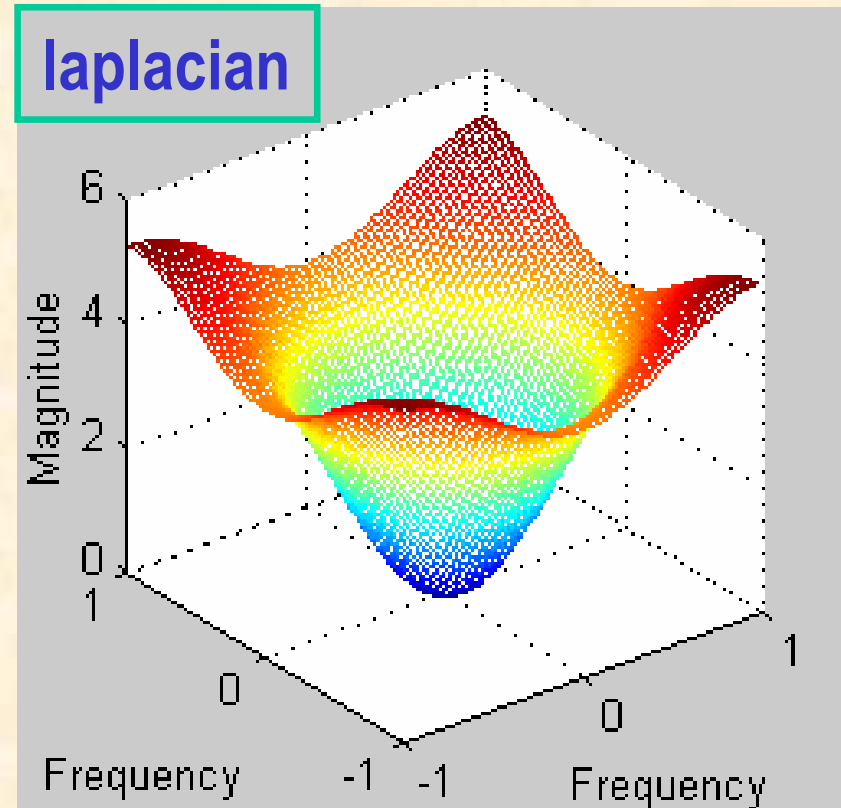
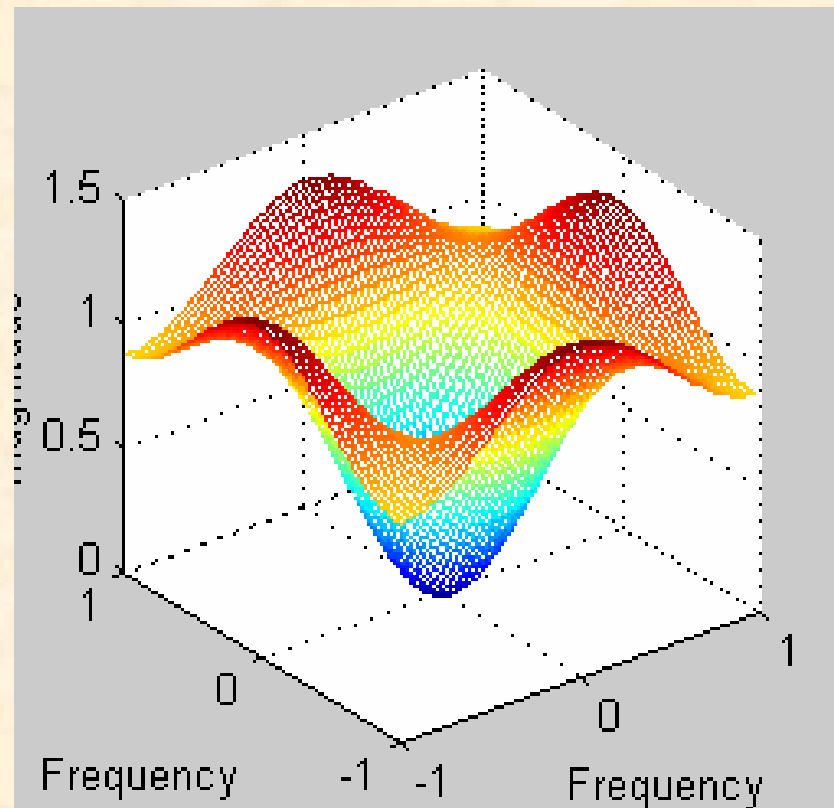


Butterworth filter $D_0=50$

High-pass filters (derivative filters)

$$h_1 = \begin{bmatrix} -1 & -1 & -1 \\ -1 & 8 & -1 \\ -1 & -1 & -1 \end{bmatrix}$$

$$h_2 = \begin{bmatrix} 0.17 & 0.67 & 0.17 \\ 0.67 & -3.33 & 0.67 \\ 0.17 & 0.67 & 0.17 \end{bmatrix}$$



High-pass filtering the image



mask h_1



mask h_2

The „high boost” filter

$$f(x, y) = f_L(x, y) + f_H(x, y)$$

$$\begin{aligned} f_{HB}(x, y) &= Af(x, y) - f_L(x, y) = \\ &= (A - 1)f(x, y) + f(x, y) - f_L(x, y) = \\ &= (A - 1)f(x, y) + f_H(x, y), \quad A \geq 1 \end{aligned}$$

$$A=?$$

$$h_{HB} = \begin{bmatrix} -1 & -1 & -1 \\ -1 & 9A-1 & -1 \\ -1 & -1 & -1 \end{bmatrix}$$

High boost filter - example



Laplace filter



$A=1.1$



$A=1.5$

A modified Laplace filter



$$h_2 = \begin{bmatrix} 0.17 & 0.67 & 0.17 \\ 0.67 & -3.33 & 0.67 \\ 0.17 & 0.67 & 0.17 \end{bmatrix}$$

$$h'_2 = \begin{bmatrix} 0.17 & 0.67 & 0.17 \\ 0.67 & -2.33 & 0.67 \\ 0.17 & 0.67 & 0.17 \end{bmatrix}$$

In order to keep the average value of the image add 1 to the centre element of the Laplace mask

Other high-pass filters

$$h'_3 = \begin{bmatrix} 0 & -1 & 0 \\ -1 & 5 & -1 \\ 0 & -1 & 0 \end{bmatrix}$$

$$h'_4 = \begin{bmatrix} -1 & -1 & -1 \\ -1 & 9 & -1 \\ -1 & -1 & -1 \end{bmatrix}$$



High-pass filters



Blurred image



Sharpened image

```
%MATLAB  
out_image = filter2(filter_mask, in_image);
```

Nonlinear filters

The filtered image is defined by a non-linear function of the source image

Can we compute spectral characteristics for nonlinear filters?

NO

Because transfer characteristics of nonlinear filters depend on image content itself!

Median filter (order statistic filter)

The median ***m*** of a set of values (e.g. image pixels in the filtering mask) is such that half the elements in the set are less than ***m*** and other half are greater than ***m***.

$$x(n)=\{ 1, 5, -7, 101, -25, 3, 0, 11, 7 \}$$

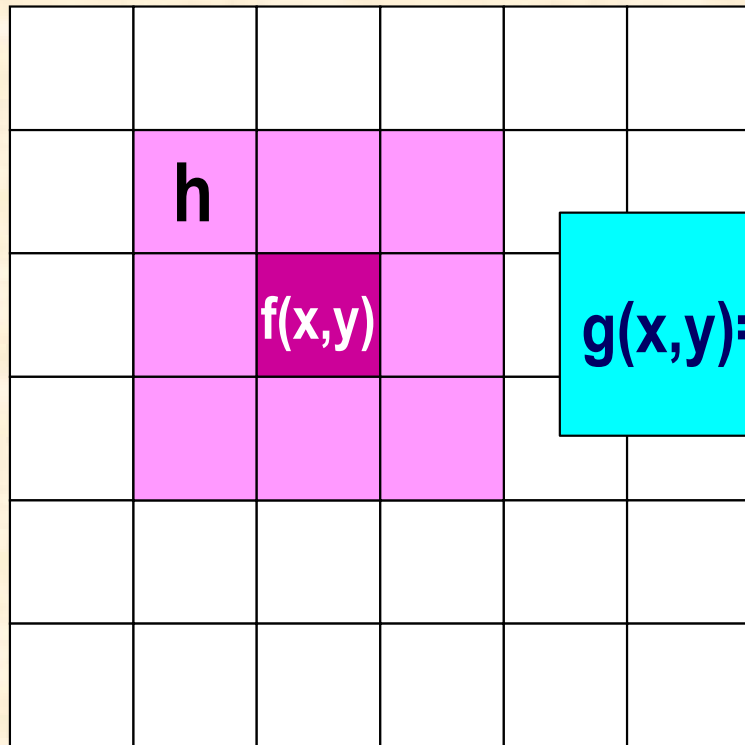
Sorted sequence of elements:

$$x_s(n)=\{-25, -7, 0, 1, \mathbf{3}, 5, 7, 11, 101 \}$$



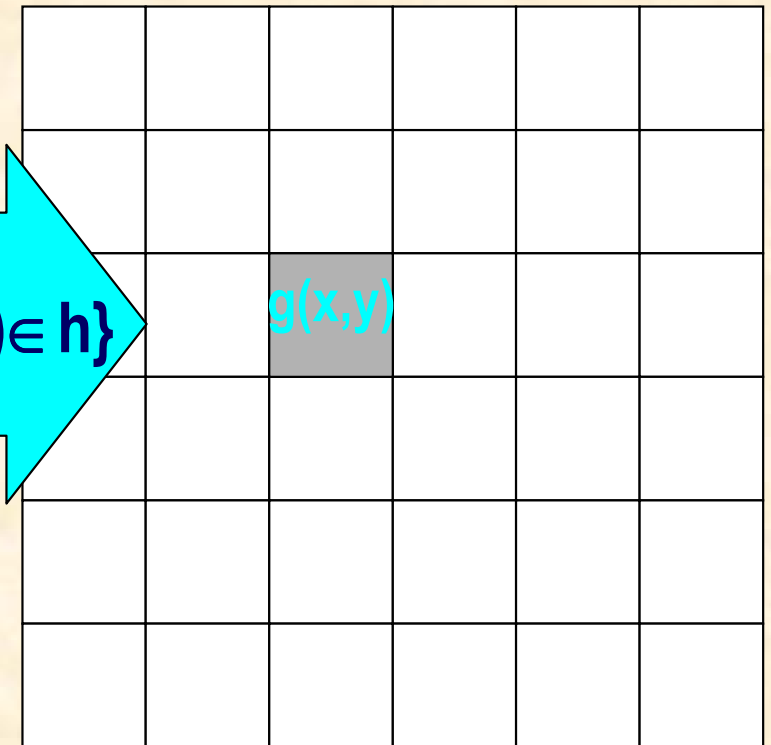
median

Median filtering the image



source image f

$$g(x,y) = \text{median}\{f(x,y); (x,y) \in h\}$$



output image g

Bubble-sort algorithm

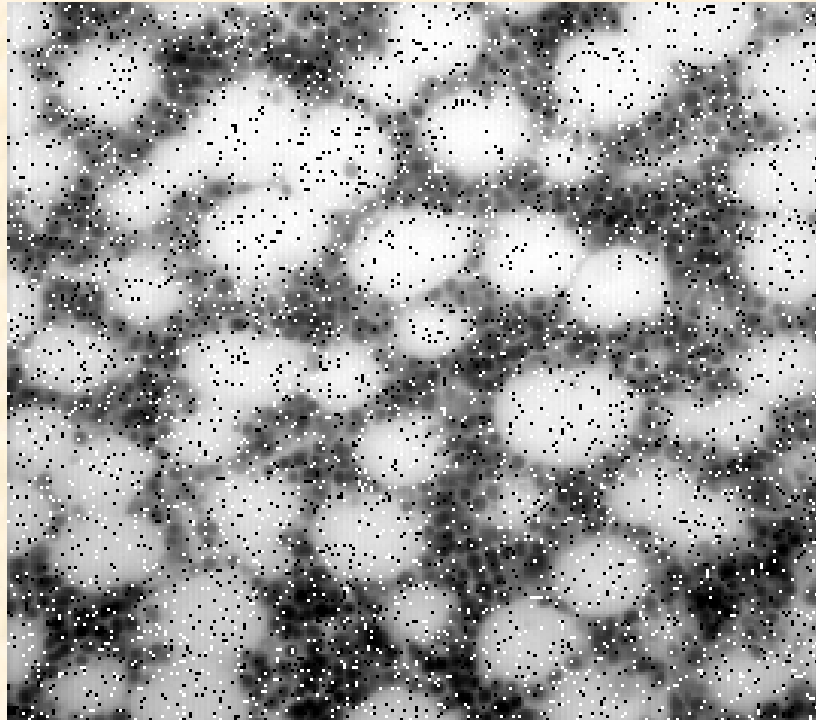
1	2	3	4	5	6	7	8 (iterations)
44	06	06	06	06	06	06	06
55	44	12	12	12	12	12	12
12	55	44	18	18	18	18	18
42	12	55	44	42	42	42	42
94	42	18	55	44	44	44	44
18	94	42	42	55	55	55	55
06	18	94	67	67	67	67	67
67	67	67	94	94	94	94	94

↑
Unsorted sequence

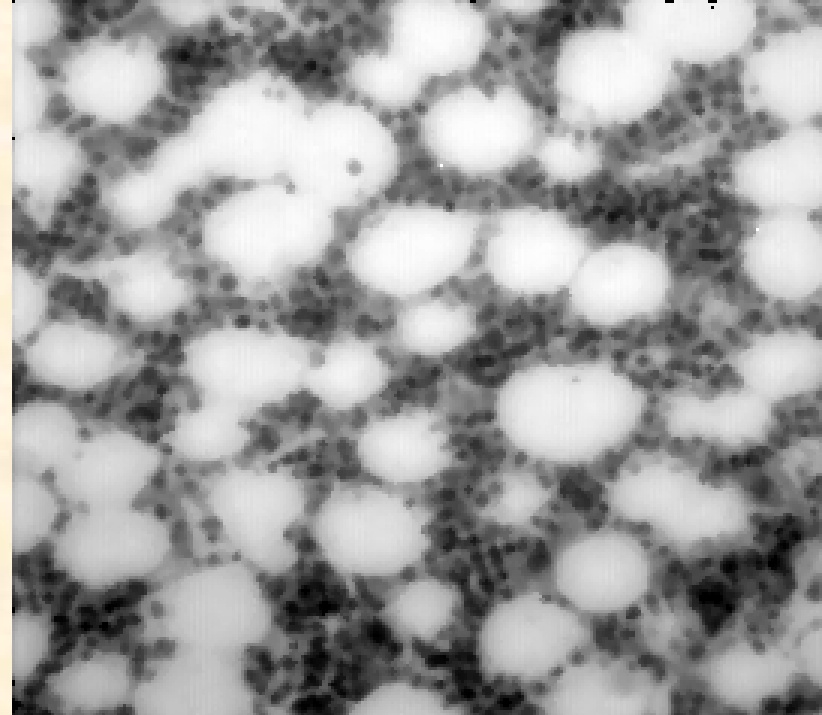
Bubble-sort program

```
a[k], k=1..N – unsorted sequence
for i:=2 to N do
begin
  for j:=N downto i do
    if a[j-1]>a[j] then
      begin
        x=a[j-1]; a[j-1]:=a[j]; a[j]:=x;
      end;
    end;
  end;
end;
```


Demo – median filter



Source image distorted by
„salt and pepper noise”



Enhanced image using
the median filter (3x3)”

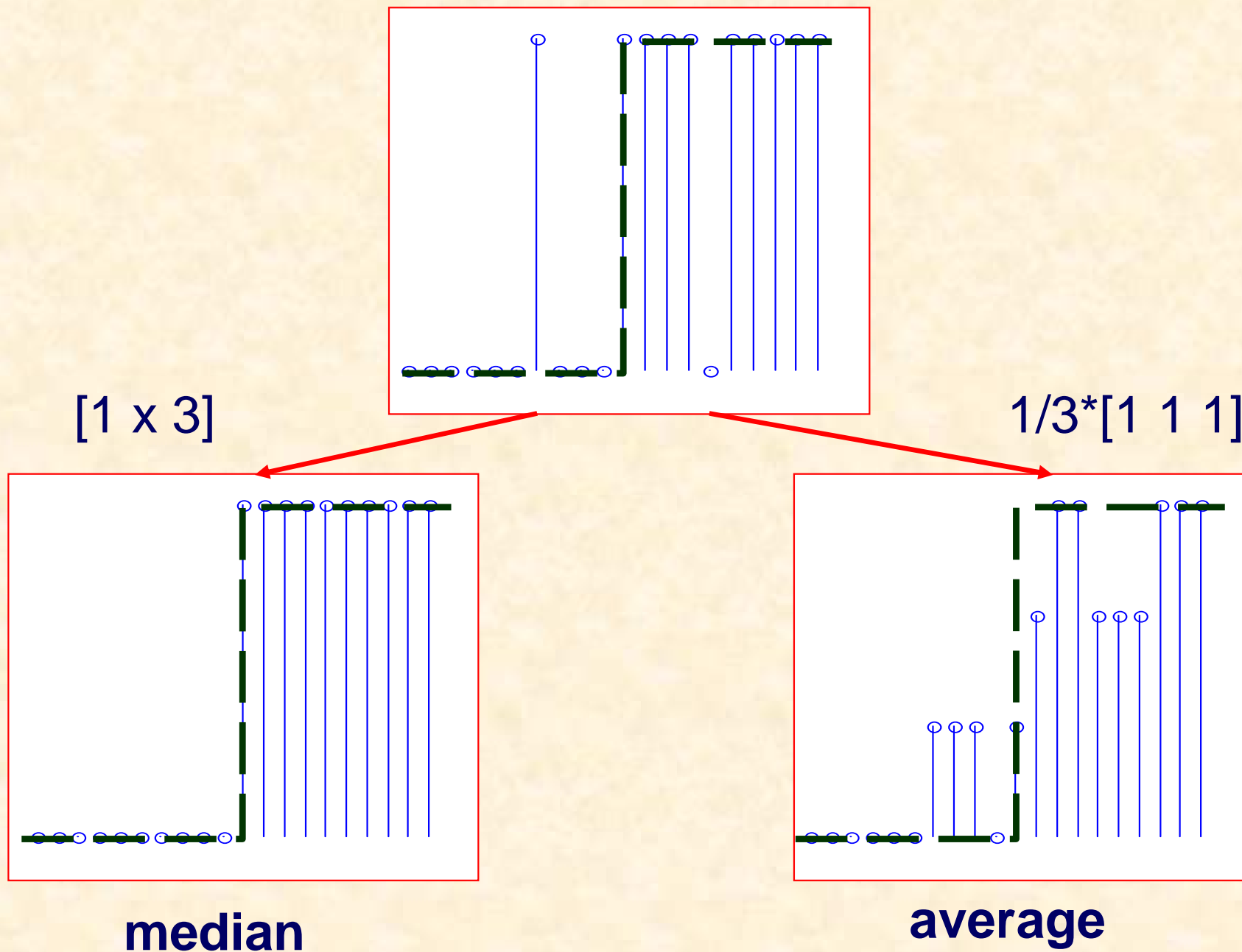
```
%MATLAB  
out_image = medfilt2(in_image, [m n]);
```

Median filter

Median filter:

1. Excellent in reducing impulsive noise (od size smaller than half size of the filtering mask)
2. Keeps sharpness of image edges (as opposed to linear smoothing filters)
3. Values of the output image are equal or smaller than the values of the input image (no rescaling)
4. Large computing cost involved

Median filter



MATLAB Demo – median filter

