

# UNIVERSITÀ DEGLI STUDI DI PADOVA

Filtering in the frequency domain

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- Filtering in the frequency domain
- Low-pass and high-pass filters
- Band-reject filters
- Ideal, Butterworth, Gaussian filters
- Denoising in frequency



#### Filters in frequency

- Filters can be defined in frequency
- What are the advantages of filtering in frequency?

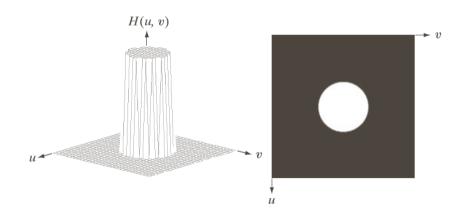


- Filters can be defined in frequency
- What are the advantages of filtering in frequency?
- Remainder of this unit: see many filters and their effect on the images



- Lowpass filters smoothing
  - Ideal
  - Butterworth
  - Gaussian
- Highpass filters sharpening
  - Ideal
  - Butterworth
  - Gaussian
- Selective filters
  - Band-reject
  - Notch

- Expressed in tems of a distance D from the center
- The zero-frequency is in the middle of the figure

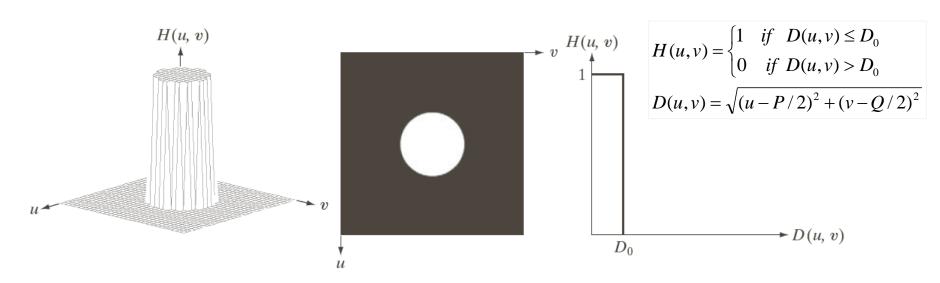


$$H(u,v) = \begin{cases} 1 & \text{if } D(u,v) \le D_0 \\ 0 & \text{if } D(u,v) > D_0 \end{cases}$$
$$D(u,v) = \sqrt{(u-P/2)^2 + (v-Q/2)^2}$$

#### Ideal lowpass filter

IAS-LAB

Ideal lowpass in frequency



**FIGURE 4.40** (a) Perspective plot of an ideal lowpass-filter transfer function. (b) Filter displayed as an image. (c) Filter radial cross section.

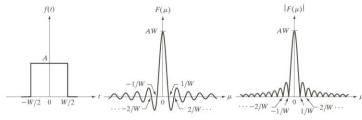


#### Ideal lowpass filter

**IAS-LAB** 

- Ideal lowpass in space
  - Observe the ripples around the main central component
- The filter is bound into a region
  - Corresponds to padding to 0 outside the filter window

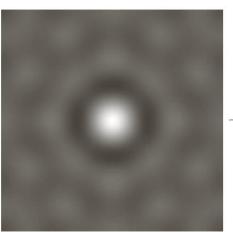
## $F(\mu)$ $|F(\mu)|$



Reminder

abc

FIGURE 4.4 (a) A simple function; (b) its Fourier transform; and (c) the spectrum. All functions extend to infinity in both directions.





#### a b

#### FIGURE 4.43

(a) Representation in the spatial domain of an ILPF of radius 5 and size 1000 × 1000. (b) Intensity profile of a horizontal line passing through the center of the image.

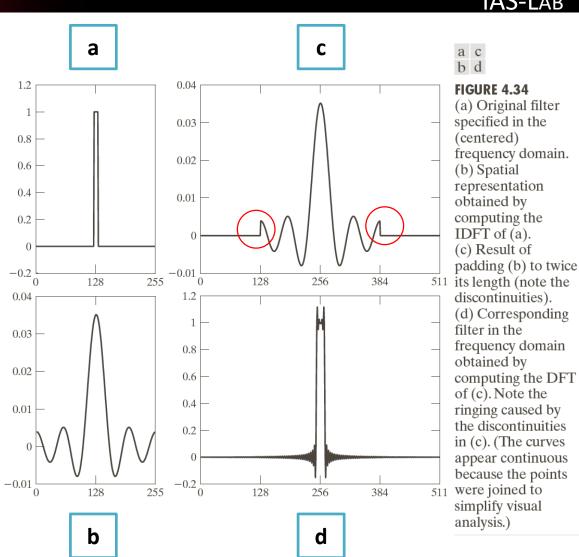


#### Ringing artifacts

#### **IAS-LAB**

- a. Ideal LPF in frequency
- b. Ideal LPF in space
- c. Ideal LPF padded to double length
  - Discontinuities generating the ringing artifacts
- d. DFT of c.

Note the ringing artifacts caused by the padding of a function having a substantial non-zero tails

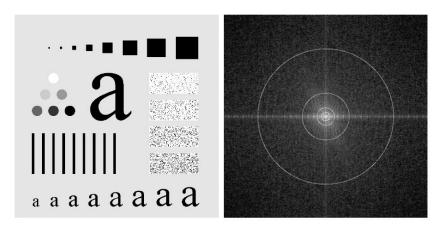




#### ILPF – examples

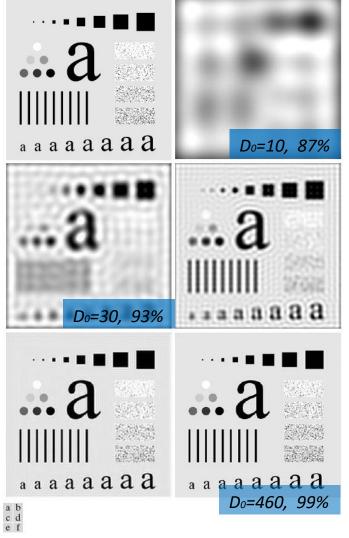
IAS-LAB

## The ringing effect is noticeable



a b

**FIGURE 4.41** (a) Test pattern of size  $688 \times 688$  pixels, and (b) its Fourier spectrum. The spectrum is double the image size due to padding but is shown in half size so that it fits in the page. The superimposed circles have radii equal to 10, 30, 60, 160, and 460 with respect to the full-size spectrum image. These radii enclose 87.0, 93.1, 95.7, 97.8, and 99.2% of the padded image power, respectively.



**FIGURE 4.42** (a) Original image. (b)–(f) Results of filtering using ILPFs with cutoff frequencies set at radii values 10, 30, 60, 160, and 460, as shown in Fig. 4.41(b). The power removed by these filters was 13, 6.9, 4.3, 2.2, and 0.8% of the total, respectively.

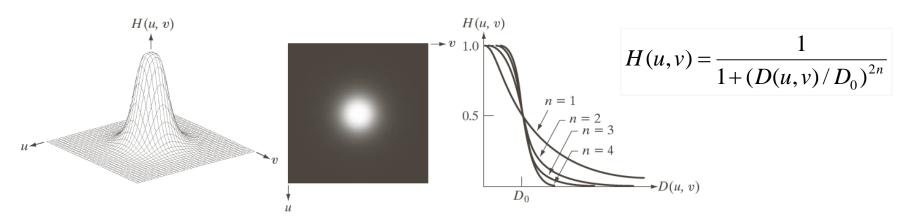
#### Coping with ringing effect

- The ringing effect causes poor image quality
- It is typical of ideal filters
  - Because of their strong transition
- Ringing effects can be reduced or eliminated by using filters showing smoother transitions
  - Butterworth filter
  - Gaussian filter

#### Butterworth lowpass filter

IAS-LAB

- Butterworth in frequency
  - Tuned by means of the parameter n
  - Large values of n cause this filter to be similar to the ILPF

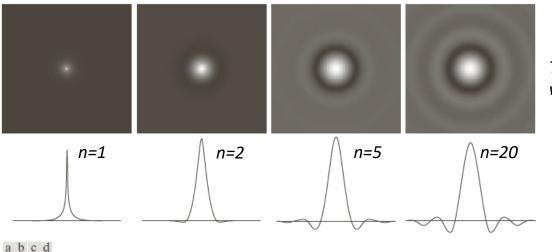


**FIGURE 4.44** (a) Perspective plot of a Butterworth lowpass-filter transfer function. (b) Filter displayed as an image. (c) Filter radial cross sections of orders 1 through 4.

#### Butterworth lowpass filter

IAS-LAB

#### Butterworth in space



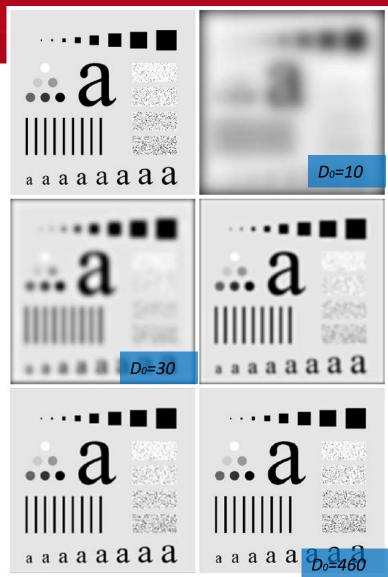
The ringing effect grows with the order of the filter

Why?

**FIGURE 4.46** (a)–(d) Spatial representation of BLPFs of order 1, 2, 5, and 20, and corresponding intensity profiles through the center of the filters (the size in all cases is  $1000 \times 1000$  and the cutoff frequency is 5). Observe how ringing increases as a function of filter order.



## Butterworth – example



a b c d

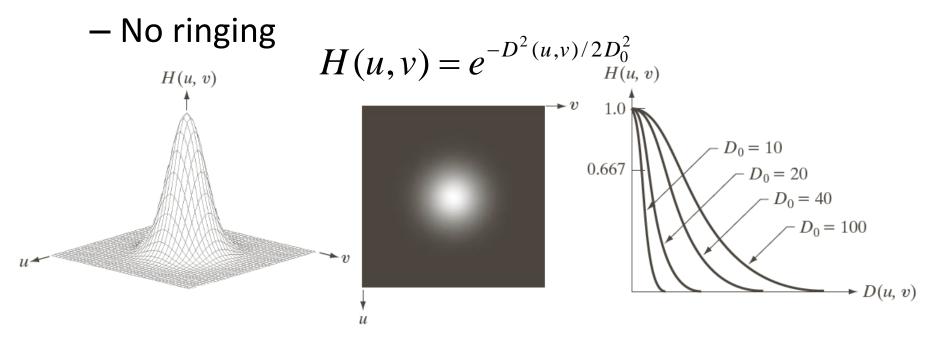
e f

FIGURE 4.45 (a) Original image. (b)-(f) Results of filtering using BLPFs of order 2, with cutoff frequencies at the radii shown in Fig. 4.41. Compare with Fig. 4.42.

#### Gaussian lowpass filter

IAS-LAB

#### Gaussian filter in frequency



**FIGURE 4.47** (a) Perspective plot of a GLPF transfer function. (b) Filter displayed as an image. (c) Filter radial cross sections for various values of  $D_0$ .



#### Gaussian filter – example

#### IAS-LAB

### Filling gaps

Historically, certain computer programs were written using only two digits rather than four to define the applicable year. Accordingly, the company's software may recognize a date using "00" as 1900 rather than the year 2000.

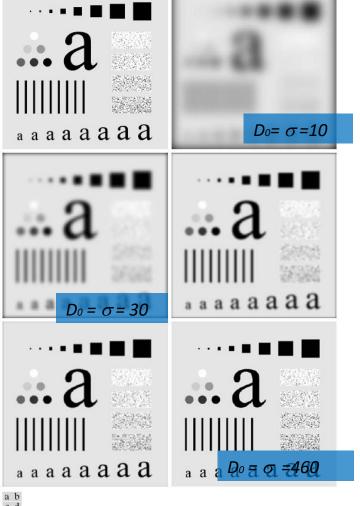
Historically, certain computer programs were written using only two digits rather than four to define the applicable year. Accordingly, the company's software may recognize a date using "00" as 1900 rather than the year 2000.



a b

#### **FIGURE 4.49**

(a) Sample text of low resolution (note broken characters in magnified view). (b) Result of filtering with a GLPF (broken character segments were joined).



**FIGURE 4.48** (a) Original image. (b)–(f) Results of filtering using GLPFs with cutoff frequencies at the radii shown in Fig. 4.41. Compare with Figs. 4.42 and 4.45.



#### Gaussian filter – example

IAS-LAB

#### Beautification

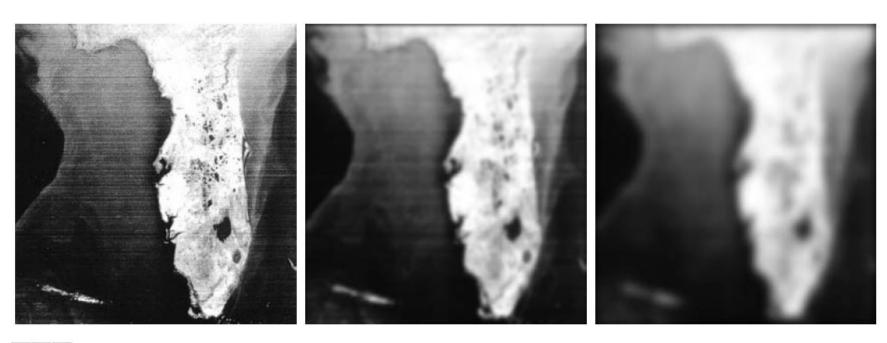


**FIGURE 4.50** (a) Original image (784  $\times$  732 pixels). (b) Result of filtering using a GLPF with  $D_0 = 100$ . (c) Result of filtering using a GLPF with  $D_0 = 80$ . Note the reduction in fine skin lines in the magnified sections in (b) and (c).

#### Gaussian filter – example

IAS-LAB

#### Reducing noise



**FIGURE 4.51** (a) Image showing prominent horizontal scan lines. (b) Result of filtering using a GLPF with  $D_0 = 50$ . (c) Result of using a GLPF with  $D_0 = 20$ . (Original image courtesy of NOAA.)



#### Lowpass filter recap

IAS-LAB

**TABLE 4.4** 

Lowpass filters.  $D_0$  is the cutoff frequency and n is the order of the Butterworth filter.

	Lowpass inters: Dijis the eaton nequency and n is the order of the Batter worth inter.						
Ideal		Butterworth		Gaussian			
$H(u,v) = \begin{cases} 1\\ 0 \end{cases}$	$if D(u, v) \le D_0$ $if D(u, v) > D_0$	$H(u,v) = \frac{1}{1 + [D(u,v)/D_0]^{2n}}$		$H(u,v) = e^{-D^2(u,v)/2D_0^2}$			
a             a a a a a a a		a           a a a a a a a a	-a	a           aaaaaaaa	_a		
a	a	a	a	a	a          a a a a a a a a		

- Lowpass filters smoothing
  - Ideal
  - Butterworth
  - Gaussian
- Highpass filters sharpening
  - Ideal
  - Butterworth
  - Gaussian
- Selective filters
  - Band-reject
  - Notch

- Lowpass filters smoothing
  - Ideal
  - Butterworth
  - Gaussian
- Highpass filters sharpening
  - Ideal
  - Butterworth
  - Gaussian
- Selective filters
  - Band-reject
  - Notch

 A HP filter can be obtained from the corresponding LP:

$$H_{\rm HP}(u,v) = 1 - H_{\rm LP}(u,v)$$

- Features of LP and HP filters (e.g. ringing) are similar for a given type of filter
- Sharpening effect

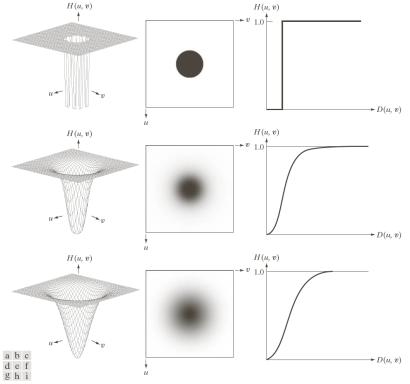


#### Highpass filters

**TABLE 4.5** Highpass filters.  $D_0$  is the cutoff frequency and n is the order of the Butterworth filter.

$$H_{hp}(u,v) = 1 - H_{lp}(u,v)$$

Ideal		Butterworth	Gaussian	
$H(u,v) = \begin{cases} 0 & \text{i} \\ 1 & \text{i} \end{cases}$	$f D(u, v) \le D_0$ $f D(u, v) > D_0$	$H(u, v) = \frac{1}{1 + [D_0/D(u, v)]^{2n}}$	$H(u, v) = 1 - e^{-D^2(u,v)/2D_0^2}$	

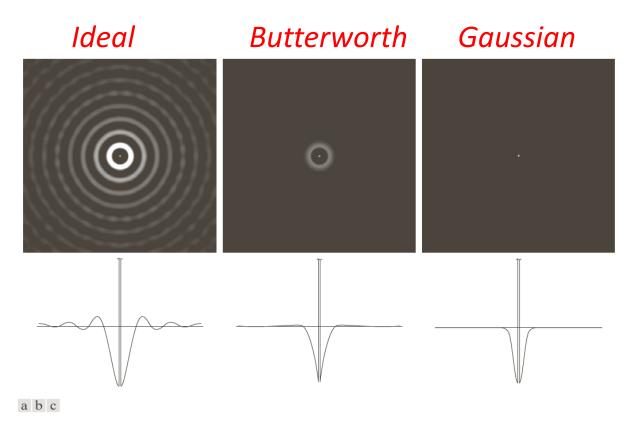


**FIGURE 4.52** Top row: Perspective plot, image representation, and cross section of a typical ideal highpass filter. Middle and bottom rows: The same sequence for typical Butterworth and Gaussian highpass filters.

## Highpass filter in the spatial domain

IAS-LAB

Spatial domain: ringing is present as in lowpass filtering

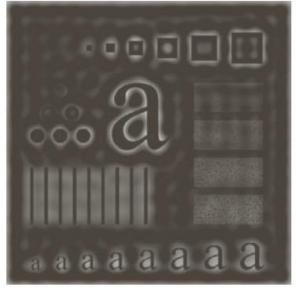


**FIGURE 4.53** Spatial representation of typical (a) ideal, (b) Butterworth, and (c) Gaussian frequency domain highpass filters, and corresponding intensity profiles through their centers.

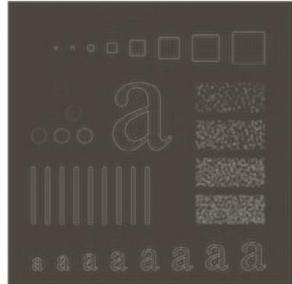


#### Ideal highpass

IAS-LAB







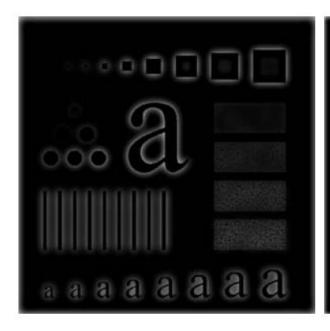
**FIGURE 4.54** Results of highpass filtering the image in Fig. 4.41(a) using an IHPF with  $D_0 = 30, 60, \text{ and } 160.$ 

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



#### Butterworth highpass

IAS-LAB







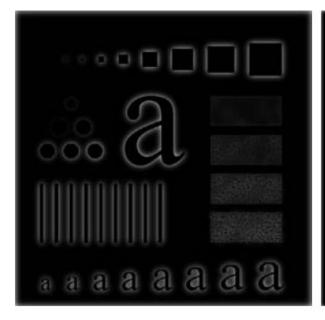
**FIGURE 4.55** Results of highpass filtering the image in Fig. 4.41(a) using a BHPF of order 2 with  $D_0 = 30, 60$ , and 160, corresponding to the circles in Fig. 4.41(b). These results are much smoother than those obtained with an IHPF.

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

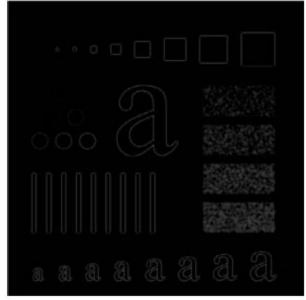


#### Gaussian highpass

IAS-LAB







**FIGURE 4.56** Results of highpass filtering the image in Fig. 4.41(a) using a GHPF with  $D_0 = 30, 60, \text{ and } 160, \text{ corresponding to the circles in Fig. 4.41(b)}$ . Compare with Figs. 4.54 and 4.55.

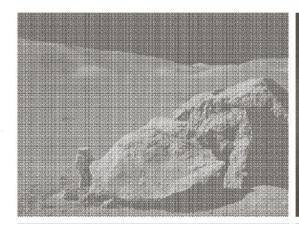
$$H(u, v) = 1 - e^{-\frac{D^2(u, v)}{2D_0^2}}$$

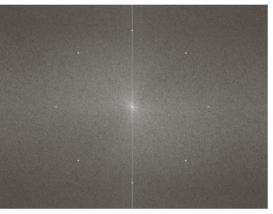
- Lowpass filters smoothing
  - Ideal
  - Butterworth
  - Gaussian
- Highpass filters sharpening
  - Ideal
  - Butterworth
  - Gaussian
- Selective filters
  - Band-reject
  - Notch

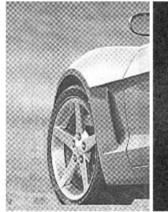
- Lowpass filters smoothing
  - Ideal
  - Butterworth
  - Gaussian
- Highpass filters sharpening
  - Ideal
  - Butterworth
  - Gaussian
- Selective filters
  - Band-reject
  - Notch

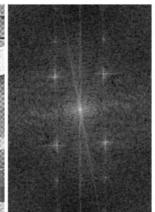
## Noise in frequency domain

- Noise can affect the frequency domain
  - E.g. noise on some frequencies









#### Selective filtering

- Operate on selected frequency bands
  - Band-pass filter
  - Band-reject filter
- Operate on selected frequency rectangles/areas
- Same profiles already discussed
  - Ideal
  - Butterworth
  - Gaussian



#### Band-reject filters

IAS-LAB



**FIGURE 5.15** From left to right, perspective plots of ideal, Butterworth (of order 1), and Gaussian bandreject filters.

#### TABLE 4.6

Bandreject filters. W is the width of the band, D is the distance D(u, v) from the center of the filter,  $D_0$  is the cutoff frequency, and n is the order of the Butterworth filter. We show D instead of D(u, v) to simplify the notation in the table.

	Ideal	Butterworth	Gaussian
$H(u,v) = \begin{cases} 0\\1 \end{cases}$	if $D_0 - \frac{W}{2} \le D \le D_0 + \frac{W}{2}$ otherwise	$H(u, v) = \frac{1}{1 + \left[\frac{DW}{D^2 - D_0^2}\right]^{2n}}$	$H(u, v) = 1 - e^{-\left[\frac{D^2 - D_0^2}{DW}\right]^2}$

$$H_{bp}(u,v) = 1 - H_{br}(u,v)$$



#### Band-reject

IAS-LAB

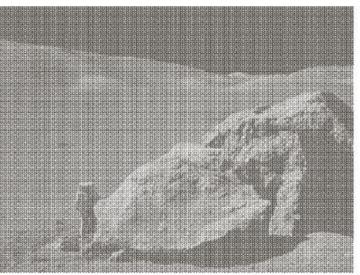
a b c d

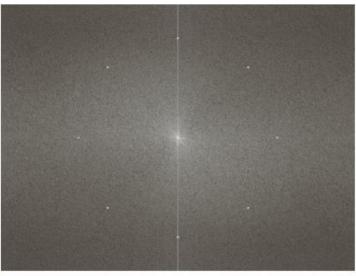
#### FIGURE 5.16

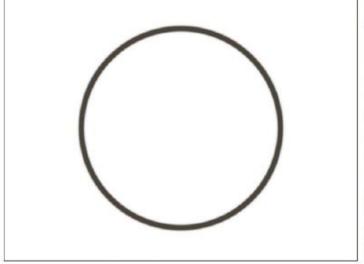
(a) Image corrupted by sinusoidal noise.

(b) Spectrum of (a).

(c) Butterworth bandreject filter (white represents 1). (d) Result of filtering. (Original image courtesy of NASA.)









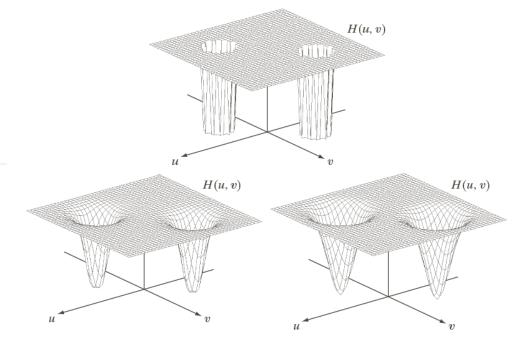
#### Notch filters

IAS-LAB

a b c

#### FIGURE 5.18

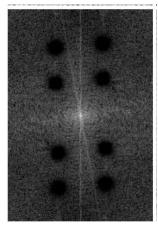
Perspective plots of (a) ideal, (b) Butterworth (of order 2), and (c) Gaussian notch (reject) filters.





## Band-reject and notch – examples

IAS-LAB

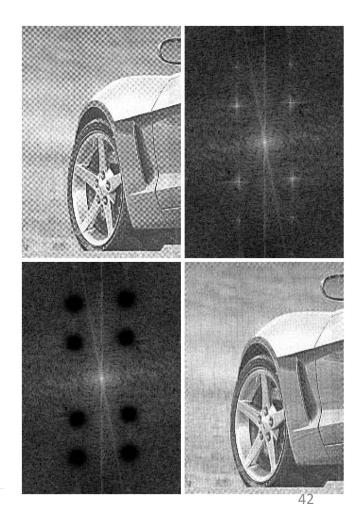


Notch

$$H_{NR}(u,v) = \prod_{k=1}^{4} \left[ \frac{1}{1 + \left[ \frac{D_{0k}}{D_k(u,v)} \right]^{2n}} \right] \left[ \frac{1}{1 + \left[ \frac{D_{0k}}{D_{-k}(u,v)} \right]^{2n}} \right]$$

a b c d

FIGURE 4.64
(a) Sampled newspaper image showing a moiré pattern.
(b) Spectrum.
(c) Butterworth notch reject filter multiplied by the Fourier transform.
(d) Filtered image.



## **General observations**

## Filtering in frequency

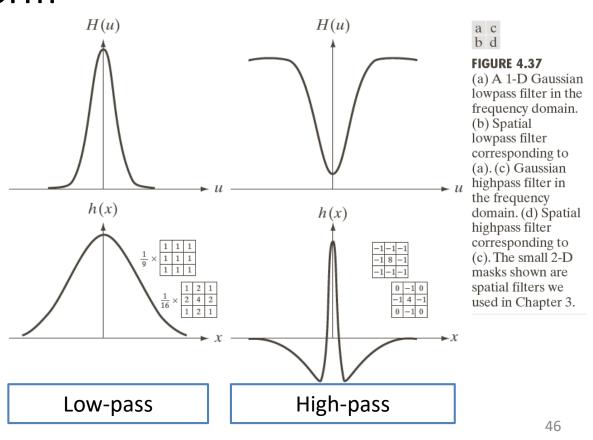
- Several filters previously described perform operations that are similar to the filters in the space domain
  - Smoothing
  - Sharpening / edge highlighting
- What are the differences with respect to the filters in the spatial domain?

## Deriving the spatial mask

- Filtering in frequency lets us operate on frequencies
- The image is provided in the space domain:
  - Need to filter on the space domain: convolution
  - Need to evaluate the filters
- Filtering in frequency means to apply convolution using filters that implement in space operations defined in frequency

#### Deriving the spatial mask

- The spatial mask can be obtained applying the inverse transform
  - Example:Gaussianfilters





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