

# Introduction to Medical Imaging (5XSA0)

## Frequency domain processing

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## What do we consider in frequency domain processing?

- \* **Filtering in the frequency domain via the Fourier transform**
  - can be used for image enhancement, restoration, compression
- \* **How to perform frequency domain processing in Matlab**

(The slides are based on "Digital Image Processing Using Matlab", R. C. Gonzalez, R. E. Woods, S. L. Eddins)

## 2D Discrete Fourier Transform

- \* For image  $f(x,y)$  ( $x=0,1,2,\dots,M-1$  and  $y=0,1,2,\dots,N-1$ ), discrete Fourier transform (DFT) is

$$F(u,v) = \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x,y) e^{-j2\pi(ux/M + vy/N)}$$

where  $u=0,1,2,\dots,M-1$  and  $v=0,1,2,\dots,N-1$

- \* **Frequency domain is the coordinate system spanned by  $F(u,v)$  with  $u$  and  $v$  as frequency variables**

## Inverse Discrete Fourier Transform

- \* **Inverse DFT is given by**

$$f(x,y) = \frac{1}{MN} \sum_{u=0}^{M-1} \sum_{v=0}^{N-1} F(u,v) e^{j2\pi(ux/M + vy/N)}$$

where  $x=0,1,2,\dots,M-1$  and  $y=0,1,2,\dots,N-1$ ,  
the values  $F(u,v)$  are called Fourier coefficients

- \*  **$F(0,0)$  is the DC component (Direct current – from electrical engineering) of the Fourier transform**

## Analyzing a transform – (1)

- \* Even if  $f(x,y)$  is real, the transform in general is complex
- \* **Spectrum – the magnitude of  $F(u,v)$  – is the principal method of visually analyzing a transform**
- \* **Fourier spectrum is defined as**

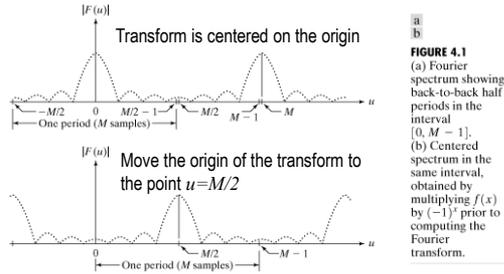
$$|F(u,v)| = [R^2(u,v) + I^2(u,v)]^{1/2}$$

where  $R(u,v)$  and  $I(u,v)$  represent the real and imaginary components of  $F(u,v)$

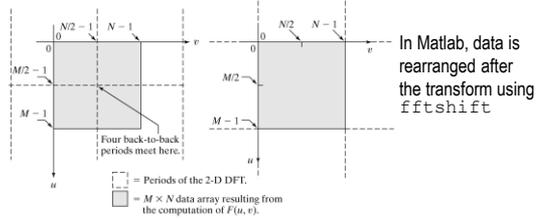
## Analyzing a transform – (2)

- \* **Fourier spectrum is symmetric about the origin**
$$|F(u,v)| = |F(-u,-v)|$$
- \* **DFT is infinitely periodic in both  $u$  and  $v$  directions, the periodicity is determined by  $M$  and  $N$**
- \* **Image obtained by taking the inverse DFT is also infinitely periodic; DFT implementations compute only one period –  $M \times N$**

## Analyzing a transform – (3)



## Analyzing a transform – (4)

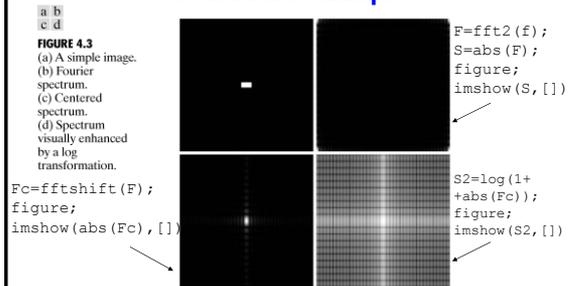


## Computing and visualizing 2D DFT in Matlab

\* DFT and its inverse are obtained in practice using a Fast Fourier Transform (FFT):  $F = \text{fft2}(f)$

- it is necessary to pad the input image with zeros:  $F = \text{fft2}(f, P, Q)$  – pads the input so that the resulting function is of size  $P \times Q$
- Fourier spectrum:  $S = \text{abs}(F)$
- Inverse Fourier transform:  $f = \text{ifft2}(F)$ ;
  - obtain an image containing only real values:  $f = \text{real}(\text{ifft2}(F))$

## Computing and visualizing 2D DFT in Matlab: example



## Filtering in the frequency domain

\* Convolution theorem

$$f(x, y) * h(h, y) \Leftrightarrow H(u, v) F(u, v)$$

and

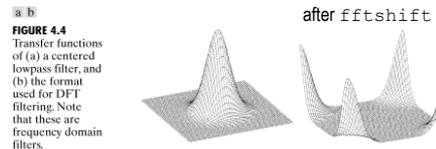
$$f(x, y) h(h, y) \Leftrightarrow H(u, v) * F(u, v)$$

symbol "\*" indicates convolution

$H(u, v)$  – filter transfer function

\* Frequency domain filtering: select a filter transfer function that modifies  $F(u, v)$  in a specified manner

## Transfer function for lowpass filter: example



Lowpass filter attenuates the high-frequency components of  $F(u, v)$ , while leaving the low frequencies relatively unchanged.

Result of lowpass filtering is image blurring (smoothing).

## Padding functions with zeros

For functions  $f(x,y)$  and  $h(x,y)$  of size  $A \times B$  and  $C \times D$  respectively, form two extended (padded) functions, both of size  $P \times Q$ , by appending zeros to  $f$  and  $g$ .

Wraparound error is avoided by choosing  $P \geq A + C - 1$  and  $Q \geq B + D - 1$

If the functions are of the same size,  $M \times N$ , then the padding values are  $P \geq 2M - 1$  and  $Q \geq 2N - 1$

## Basic steps (1-2) in DFT filtering

Step-by-step procedure involving Matlab functions, where  $f$  – image to be filtered,

$g$  – result,

$H(u,v)$  – filter function of the same size as the padded image.

1. Obtain the padding parameters:  $PQ = 2 * \text{size}(f)$ .
2. Obtain the Fourier transform with padding:  
 $F = \text{fft2}(f, PQ(1), PQ(2))$ .

## Basic steps (3-6) in DFT filtering

3. Generate a filter function  $H$  of size  $PQ1 \times PQ2$ . If the filter function is centered, let  $H = \text{fftshift}(H)$  before using the filter.
4. Multiply the transform by the filter:  $G = H .* F$ ;
5. Obtain the real part of the inverse FFT of  $G$ :  
 $g = \text{real}(\text{ifft2}(G))$ ;
6. Crop the top left rectangle of the original size:  
 $g = g(1:\text{size}(f,1), 1:\text{size}(f,2))$ ;

## Filtering procedure summarized

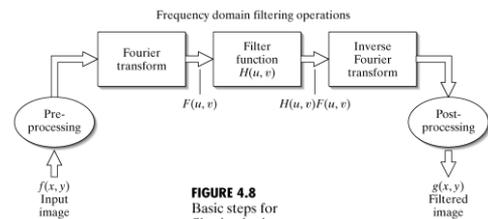


FIGURE 4.8  
Basic steps for filtering in the frequency domain.

## Converting spatial filters into equivalent frequency domain filters

- \* Choice between filtering in spatial or frequency domain may depend on the computational efficiency

- \* Filter in the frequency domain:

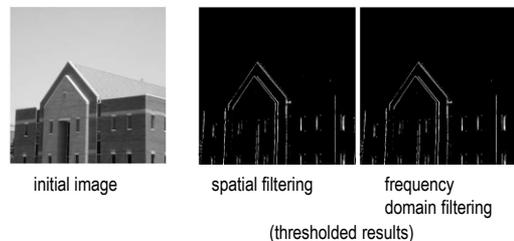
$$H = \text{freqz2}(h, R, C),$$

where  $h$  is a 2D spatial filter,

$R$  is the number of rows and

$C$  is the number of columns that we wish filter  $H$  to have

## Spatial filtering (Sobel) versus equivalent frequency domain filtering: identical results



## Creating meshgrid arrays for filters in the frequency domain

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```
function [U V]=dftuv(M,N)
% DFTUV Computes meshgrid frequency matrices. U and V are both M-by-N
% Set up range of variables
u=0:(M-1); v=0:(N-1);
% Compute the indices for use in meshgrid
idx=find(u>M/2); u(idx)=u(idx)-M;
idy=find(v>N/2); v(idy)=v(idy)-N;
% Compute the meshgrid arrays
[U V]=meshgrid(v,u);
```

## Lowpass frequency domain filters – (1)

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\* **Ideal lowpass filter (ILPF)** has the transfer function

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

where  $D_0$  – specified nonnegative number,

$D(u,v)$  – distance from point  $(u,v)$  to the center of the filter

– Ideal filter “cuts off” (multiplies by 0) all components of  $F$  outside the circle and leaves unchanged (multiplies by 1) all components on, or inside, the circle

## Lowpass frequency domain filters – (2)

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\* **Butterworth lowpass filter (BLPF)** of order  $n$ , with a cutoff frequency at a distance  $D_0$  from the origin, has the transfer function

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

– BLPF transfer function does not have a sharp discontinuity at  $D_0$

## Lowpass frequency domain filters – (3)

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\* **Transfer function of a Gaussian lowpass filter (GLPF)**

$$H(u,v) = e^{-D^2(u,v)/2\sigma^2}$$

where  $\sigma$  – standard deviation

– By letting  $\sigma = D_0$ , we obtain the expression for GLPF in terms of the cutoff parameter:

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

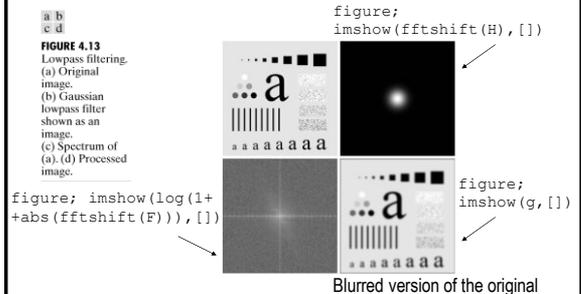
## Gaussian lowpass filter: example – (1)

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```
PQ=2*size(f);
% Create meshgrid for the transfer function – you need dftuv.m
[U V]=dftuv(PQ(1),PQ(2));
D=sqrt(U.^2+V.^2);
% define the parameter for cut-off frequency D0
D0=0.05*PQ(2); % 5% of the padded image width
% Perform FFT of the original image and obtain the filter transfer function
F=fft2(f,PQ(1),PQ(2));
H=exp(-(D.^2)/(2*(D0.^2)));
% Obtain and crop the resulting image
g=real(ifft2(H.*F));
g=g(1:size(f,1),1:size(f,2));
```

## Gaussian lowpass filter: example – (2)

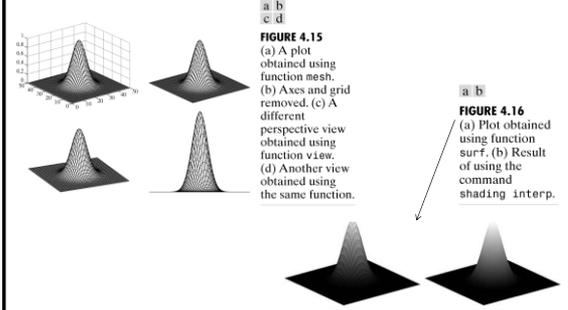
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### 3D plots – (1)

- \* Draw a plot: **mesh**, **surf**
- \* Set or switch off axis: **axis**
- \* Switch on or off the grid: **grid**
- \* Change the viewing point:
  - view
  - Click on the “Rotate 3D” button in the figure window’s toolbar

### 3D plots – (2)



### Sharpening frequency domain filters: highpass filtering

- \* Highpass filtering sharpens the image by attenuating the low frequencies and leaving the high frequencies of the Fourier transform relatively unchanged
- \* Transfer function of a highpass filter:

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

where  $H_{lp}(u, v)$  – transfer function of the corresponding lowpass filter

### Highpas filters: examples

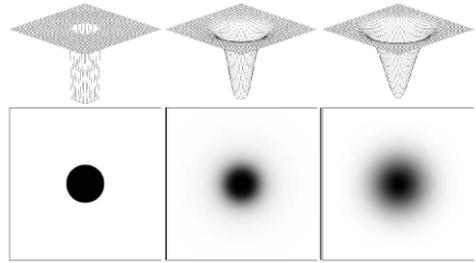


FIGURE 4.17 Top row: Perspective plots of ideal, Butterworth, and Gaussian highpass filters. Bottom row: Corresponding images.

### Reference

- Rafael C. Gonzalez, Richard E. Woods, Steven L. Eddins, “Digital Image Processing Using Matlab”, Pearson Education, 2004
- Chapter 4