

Enabling Technologies for Sports (5XSF0) Module 3

Frequency domain processing

Sveta Zinger

(s.zinger@tue.nl)



PdW-SZ / 2017
Fac. EE SPS-VCA

Enabling Technologies for Sports /
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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)



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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)

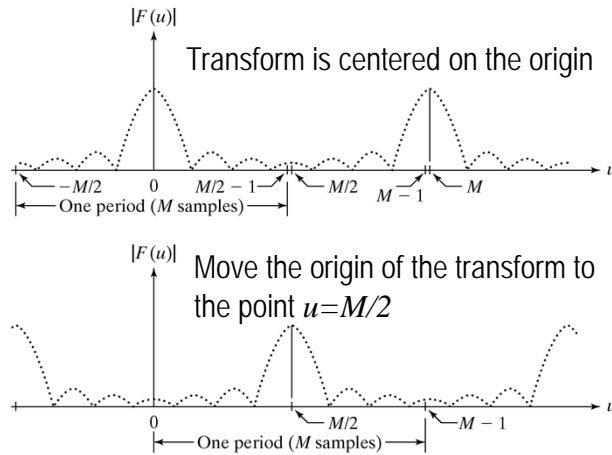


FIGURE 4.1
 (a) Fourier spectrum showing back-to-back half periods in the interval $[0, M - 1]$.
 (b) Centered spectrum in the same interval, obtained by multiplying $f(x)$ by $(-1)^x$ prior to computing the Fourier transform.

Analyzing a transform – (4)

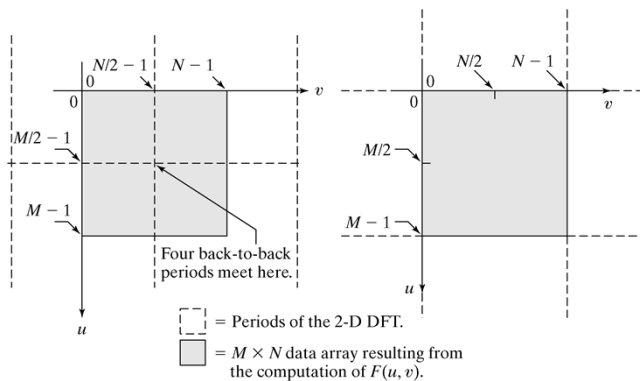


FIGURE 4.2 (a) $M \times N$ Fourier spectrum (shaded), showing four back-to-back quarter periods contained in the spectrum data. (b) Spectrum obtained by multiplying $f(x, y)$ by $(-1)^{x+y}$ prior to computing the Fourier transform. Only one period is shown shaded because this is the data that would be obtained by an implementation of the equation for $F(u, v)$.

Computing and visualizing 2D DFT in Matlab

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* 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:
 $f = \text{real}(\text{ifft2}(F))$

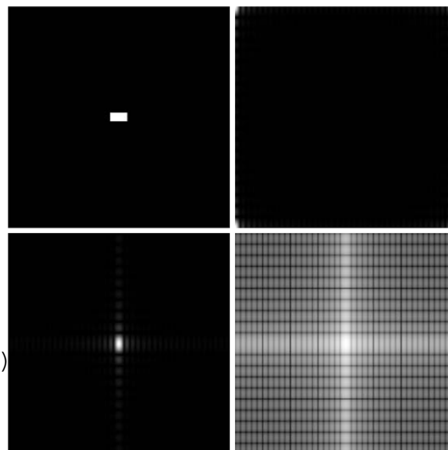
Computing and visualizing 2D DFT in Matlab: example

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a b
c d

FIGURE 4.3
 (a) A simple image.
 (b) Fourier spectrum.
 (c) Centered spectrum.
 (d) Spectrum visually enhanced by a log transformation.

```
Fc=fftshift(F);
figure;
imshow(abs(Fc), [])
```



```
F=fft2(f);
S=abs(F);
figure;
imshow(S, [])
```

```
S2=log(1+
+abs(Fc));
figure;
imshow(S2, [])
```

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

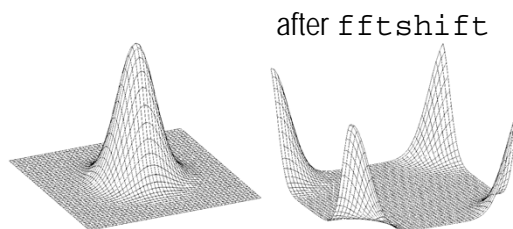
$H(u, v)$ – filter transfer function

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

Transfer function for lowpass filter: example

a b

FIGURE 4.4
Transfer functions of (a) a centered lowpass filter, and (b) the format used for DFT filtering. Note that these are frequency domain filters.



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: $G = H \cdot F$;
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

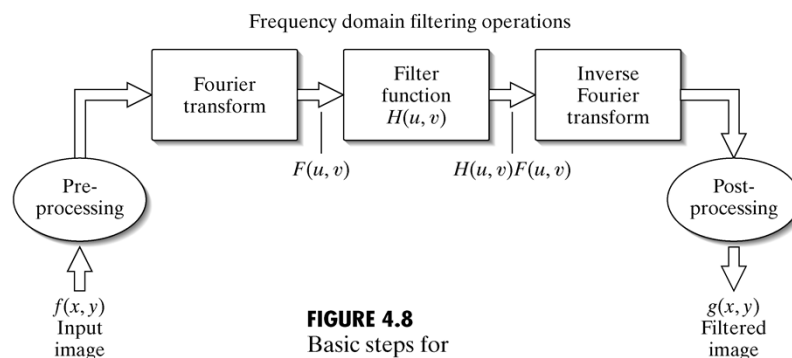


FIGURE 4.8
Basic steps for filtering in the frequency domain.

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:

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

where h is a 2D spatial filter,

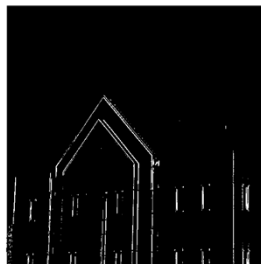
R is the number of rows and

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

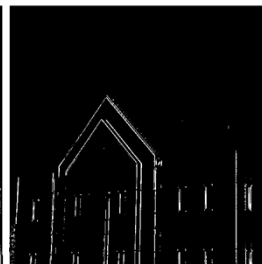
Spatial filtering (Sobel) versus equivalent frequency domain filtering: identical results



initial image



spatial filtering



frequency domain filtering

(thresholded results)

Creating meshgrid arrays for filters in the frequency domain

```
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)

* 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)

- * 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)

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

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

where σ – 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)

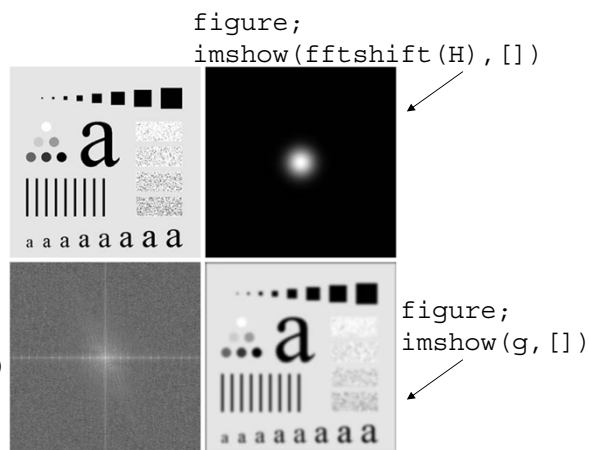
```
PQ=2*size(f);
% Create meshgrid for the transfer function – you need dftuv.m
[U V]=dftuv(PQ(1),PQ(2));
D=sqrt(U.^2+V.^2);
% define the parameter for cut-off frequency D0
D0=0.05*PQ(2); % 5% of the padded image width
% Perform FFT of the original image and obtain the filter transfer function
F=fft2(f,PQ(1),PQ(2));
H=exp(-(D.^2)/(2*(D0^2)));
% Obtain and crop the resulting image
g=real(iff2(H.*F));
g=g(1:size(f,1),1:size(f,2));
```

Gaussian lowpass filter: example – (2)

a b
c d

FIGURE 4.13
Lowpass filtering.
(a) Original image.
(b) Gaussian lowpass filter shown as an image.
(c) Spectrum of (a). (d) Processed image.

```
figure; imshow(log(1+abs(fftshift(F))), [])
```

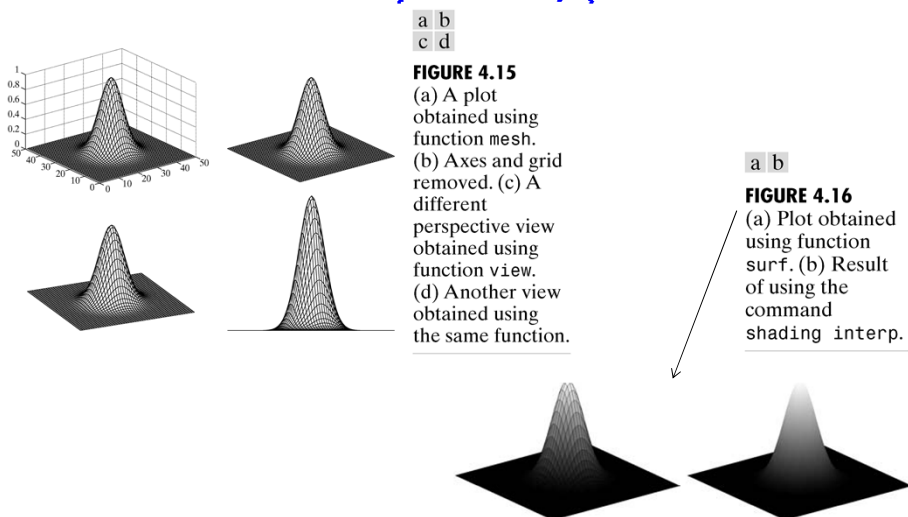


Blurred version of the original

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

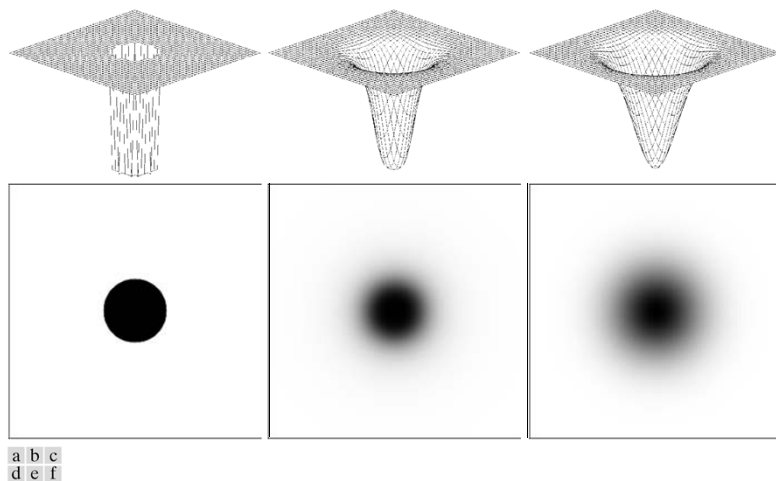


FIGURE 4.17 Top row: Perspective plots of ideal, Butterworth, and Gaussian highpass filters. Bottom row: Corresponding images.

Reference

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