

Image Processing and Visual Communications

Two-Dimensional Discrete Fourier Transform (2D-DFT)

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Summary of FT, FS, DTFT/DSFT, DFS, DFT and FFT

Fourier Transform (FT):

$$\begin{array}{ccc}
 x(t) & \xleftrightarrow{X(\Omega) = \int_{-\infty}^{\infty} x(t)e^{-j\Omega t} dt} & X(\Omega) \\
 \text{(continuous)} & & \text{(continuous)} \\
 & \xleftarrow{x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(\Omega)e^{j\Omega t} d\Omega} &
 \end{array}$$

Fourier Series (FS):

$$\begin{array}{ccc}
 x(t) & \xleftrightarrow{X_m = \frac{1}{T} \int_{-T/2}^{T/2} x(t)e^{-jm\Omega_0 t} dt} & X_m \\
 \text{(continuous, periodic)} & & \text{(discrete)} \\
 & \xleftarrow{x(t) = \sum_{m=-\infty}^{\infty} X_m e^{jm\Omega_0 t}} &
 \end{array}$$

$\Omega_0 = \frac{2\pi}{T}$

Discrete Time/Space Fourier Transform (DTFT/DSFT):

$$\begin{array}{ccc}
 x(n) & \xleftrightarrow{X(\omega) = \sum_{n=-\infty}^{\infty} x(n)e^{-j\omega n}} & X(\omega) \\
 \text{(discrete)} & & \text{(continuous, periodic)} \\
 & \xleftarrow{x(n) = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(\omega)e^{j\omega n} d\omega} &
 \end{array}$$

Summary of FT, FT, DTFT/DSFT, DFS, DFT and FFT

Discrete
Fourier
Series
(DFS):

$$\begin{array}{ccc}
 \tilde{x}(n) & \begin{array}{c} \xrightarrow{\tilde{X}_m = \sum_{n=0}^{N-1} \tilde{x}(n) e^{-j2\pi mn/N}} \\ \xleftarrow{\tilde{x}(n) = \frac{1}{N} \sum_{m=0}^{N-1} \tilde{X}_m e^{j2\pi mn/N}} \end{array} & \tilde{X}_m \\
 \text{(discrete, periodic)} & & \text{(discrete, periodic)}
 \end{array}$$

Discrete
Fourier
Transform
(DFT):

$$\begin{array}{ccc}
 x(n) & \begin{array}{c} \xrightarrow{X(m) = \sum_{n=0}^{N-1} x(n) e^{-j2\pi mn/N}} \\ \xleftarrow{x(n) = \frac{1}{N} \sum_{m=0}^{N-1} X(m) e^{j2\pi mn/N}} \end{array} & X(m) \\
 \text{(discrete, finite)} & & \text{(discrete, finite)}
 \end{array}$$

Fast Fourier Transform (FFT): Fast algorithm for computing DFT

2D DFT and Inverse DFT (IDFT)

$$F(u, v) = \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) \exp \left[-2\pi i \left(\frac{xu}{M} + \frac{yv}{N} \right) \right]$$

$$f(x, y) \xleftrightarrow{\hspace{10em}} F(u, v)$$

$$f(x, y) = \frac{1}{MN} \sum_{u=0}^{M-1} \sum_{v=0}^{N-1} F(u, v) \exp \left[2\pi i \left(\frac{xu}{M} + \frac{yv}{N} \right) \right]$$

M, N : image size

x, y : image pixel position

u, v : spatial frequency

often used
short notation:

$$W_N = e^{-j2\pi/N}$$

The Meaning of DFT and Spatial Frequencies

- Important Concept

Any signal can be represented as a linear combination of a set of **basic components**

$$f(x, y) = \frac{1}{MN} \sum_{u=0}^{M-1} \sum_{v=0}^{N-1} F(u, v) \exp \left[2\pi i \left(\frac{xu}{M} + \frac{yv}{N} \right) \right]$$

- **Fourier components**: sinusoidal patterns
 - **Fourier coefficients**: weighting factors assigned to the Fourier components
- **Spatial frequency**: The frequency of Fourier component
- **Not to be confused with electromagnetic frequencies** (e.g., the frequencies associated with the colors of light)

Real Part, Imaginary Part, Magnitude, Phase, Spectrum

Real part:

$$R = \text{Real}(F)$$

Imaginary part:

$$I = \text{Imag}(F)$$

Magnitude-phase
representation:

$$F(u, v) = |F(u, v)|e^{-j\phi(u, v)}$$

Magnitude
(spectrum):

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

Phase
(spectrum):

$$\phi(u, v) = \tan^{-1} \left[\frac{I(u, v)}{R(u, v)} \right] \quad \text{for } R > 0$$

Power
Spectrum:

$$P(u, v) = |F(u, v)|^2$$

2D DFT Properties

Mean of image/
DC component:

$$\bar{f}(x, y) = F(0, 0) = \frac{1}{MN} \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y)$$

Highest frequency
component:

$$f(x, y)(-1)^{x+y} \Leftrightarrow F(u - M/2, v - N/2)$$

“Half-shifted”
Image:

$$f(x - M/2, y - N/2) \Leftrightarrow F(u, v)(-1)^{u+v}$$

Conjugate
Symmetry:

$$F(u, v) = F^*(-u, -v)$$

Magnitude
Symmetry:

$$|F(u, v)| = |F(-u, -v)|$$

2D DFT Properties

Spatial domain
differentiation:

$$\frac{\partial^n f(x, y)}{\partial x^n} \Leftrightarrow (ju)^n F(u, v)$$

Frequency domain
differentiation:

$$(-jx)^n f(x, y) \Leftrightarrow \frac{\partial^n F(u, v)}{\partial u^n}$$

Distribution law:

$$\Im[f_1(x, y) + f_2(x, y)] = \Im[f_1(x, y)] + \Im[f_2(x, y)]$$

Laplacian:

$$\nabla^2 f(x, y) \Leftrightarrow -(u^2 + v^2)F(u, v)$$

Spatial domain

Periodicity:

$$f(x, y) = f(x + M, y) = f(x, y + N) = f(x + M, y + N)$$

Frequency domain

periodicity:

$$F(u, v) = F(u + M, v) = F(u, v + N) = F(u + M, v + N)$$

Computation of 2D-DFT

Fourier transform matrices:

remember $W_N = e^{-j2\pi/N}$

$$\mathbf{F}_N = \begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & W_N & W_N^2 & \dots & W_N^{N-1} \\ \vdots & \vdots & \vdots & & \vdots \\ \vdots & \vdots & \vdots & & \vdots \\ 1 & W_N^{N-1} & W_N^{2(N-1)} & \dots & W_N^{(N-1)^2} \end{bmatrix}$$

$$\mathbf{F}_N^* = \begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & W_N^{-1} & W_N^{-2} & \dots & W_N^{1-N} \\ \vdots & \vdots & \vdots & & \vdots \\ \vdots & \vdots & \vdots & & \vdots \\ 1 & W_N^{1-N} & W_N^{2(1-N)} & \dots & W_N^{-(N-1)^2} \end{bmatrix}$$

relationship: $\mathbf{F}_N^{-1} = \frac{1}{N} \mathbf{F}_N^*$

In particular, for $N = 4$:

$$\mathbf{F}_4 = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{bmatrix}$$

$$\mathbf{F}_4^* = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & j & -1 & -j \\ 1 & -1 & 1 & -1 \\ 1 & -j & -1 & j \end{bmatrix}$$

Computation of 2D-DFT

- To compute the 1D-DFT of a 1D signal \mathbf{x} (as a vector):

$$\tilde{\mathbf{x}} = \mathbf{F}_N \mathbf{x}$$

To compute the inverse 1D-DFT:

$$\mathbf{x} = \frac{1}{N} \mathbf{F}_N^* \tilde{\mathbf{x}}$$

- To compute the 2D-DFT of an image \mathbf{X} (as a matrix):

$$\tilde{\mathbf{X}} = \mathbf{F}_N \mathbf{X} \mathbf{F}_N$$

To compute the inverse 2D-DFT:

$$\mathbf{X} = \frac{1}{N^2} \mathbf{F}_N^* \tilde{\mathbf{X}} \mathbf{F}_N^*$$

Computation of 2D-DFT: Example

- A 4x4 image
- Compute its 2D-DFT:

$$\mathbf{X} = \begin{bmatrix} 1 & 3 & 6 & 8 \\ 9 & 8 & 8 & 2 \\ 5 & 4 & 2 & 3 \\ 6 & 6 & 3 & 3 \end{bmatrix}$$

$$\tilde{\mathbf{X}} = \mathbf{F}_4 \mathbf{X} \mathbf{F}_4 =$$

MATLAB function: *fft2*

lowest frequency
component

highest frequency
component

$$= \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{bmatrix} \begin{bmatrix} 1 & 3 & 6 & 8 \\ 9 & 8 & 8 & 2 \\ 5 & 4 & 2 & 3 \\ 6 & 6 & 3 & 3 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{bmatrix}$$

$$= \begin{bmatrix} 21 & 21 & 19 & 16 \\ -4-3j & -1-2j & 4-5j & 5+j \\ -9 & -7 & -3 & 6 \\ -4+3j & -1+2j & 4+5j & 5-j \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{bmatrix}$$

$$= \begin{bmatrix} 77 & 2-5j & 3 & 2+5j \\ 4-9j & -11+8j & -4-7j & -5-4j \\ -13 & -6+13j & -11 & -6-13j \\ 4+9j & -5+4j & -4+7j & -11-8j \end{bmatrix}$$

Computation of 2D-DFT: Example

$$\tilde{\mathbf{X}} = \begin{bmatrix} 77 & 2-5j & 3 & 2+5j \\ 4-9j & -11+8j & -4-7j & -5-4j \\ -13 & -6+13j & -11 & -6-13j \\ 4+9j & -5+4j & -4+7j & -11-8j \end{bmatrix}$$

Real part:

$$\tilde{\mathbf{X}}_{real} = \begin{bmatrix} 77 & 2 & 3 & 2 \\ 4 & -11 & -4 & -5 \\ -13 & -6 & -11 & -6 \\ 4 & -5 & -4 & -11 \end{bmatrix}$$

Imaginary part:

$$\tilde{\mathbf{X}}_{imag} = \begin{bmatrix} 0 & -5 & 0 & 5 \\ -9 & 8 & -7 & -4 \\ 0 & 13 & 0 & -13 \\ 9 & 4 & 7 & -8 \end{bmatrix}$$

Magnitude:

$$\tilde{\mathbf{X}}_{magnitude} = \begin{bmatrix} 77 & 5.39 & 3 & 5.39 \\ 9.85 & 13.60 & 8.06 & 6.4 \\ 13 & 14.32 & 11 & 14.32 \\ 9.85 & 6.40 & 8.06 & 13.60 \end{bmatrix}$$

Phase:

$$\tilde{\mathbf{X}}_{phase} = \begin{bmatrix} 0 & -1.19 & 0 & 1.19 \\ -1.15 & 2.51 & -2.09 & -2.47 \\ 3.14 & 2.00 & 3.14 & -2.00 \\ 1.15 & 2.47 & 2.09 & -2.51 \end{bmatrix}$$

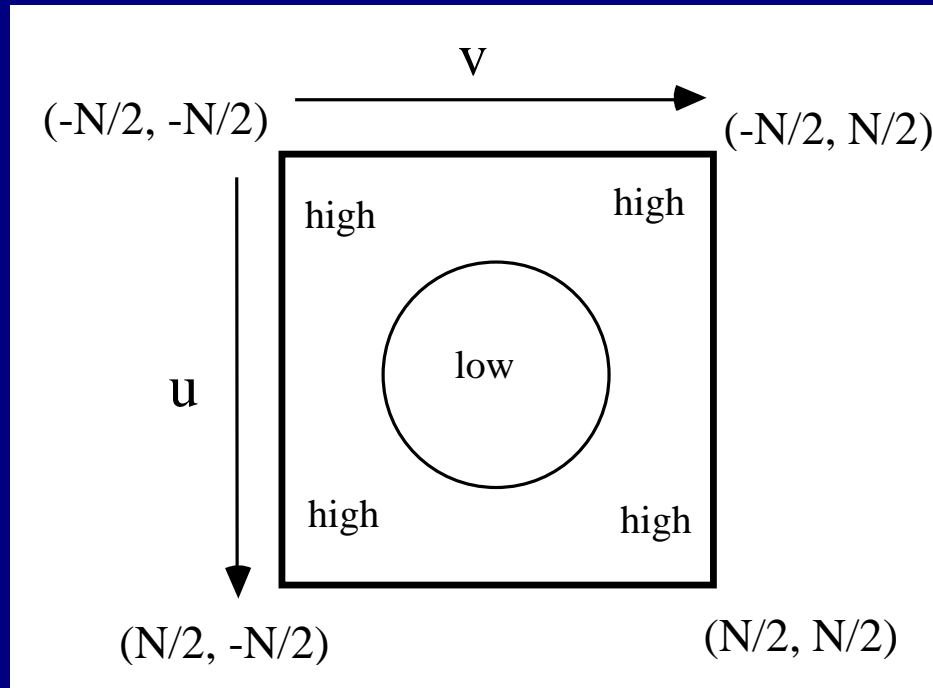
Computation of 2D-DFT: Example

- Compute the inverse 2D-DFT:

$$\begin{aligned}
 \mathbf{F}_4^* \tilde{\mathbf{X}} \mathbf{F}_4^* &= \frac{1}{4^2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & j & -1 & -j \\ 1 & -1 & 1 & -1 \\ 1 & -j & -1 & j \end{bmatrix} \begin{bmatrix} 77 & 2-5j & 3 & 2+5j \\ 4-9j & -11+8j & -4-7j & -5-4j \\ -13 & -6+13j & -11 & -6-13j \\ 4+9j & -5+4j & -4+7j & -11-8j \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & j & -1 & -j \\ 1 & -1 & 1 & -1 \\ 1 & -j & -1 & j \end{bmatrix} \\
 &= \frac{1}{4} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & j & -1 & -j \\ 1 & -1 & 1 & -1 \\ 1 & -j & -1 & j \end{bmatrix} \begin{bmatrix} 21 & 21 & 19 & 16 \\ -4-3j & -1-2j & 4-5j & 5+j \\ -9 & -7 & -3 & 6 \\ -4+3j & -1+2j & 4+5j & 5-j \end{bmatrix} \\
 &= \begin{bmatrix} 1 & 3 & 6 & 8 \\ 9 & 8 & 8 & 2 \\ 5 & 4 & 2 & 3 \\ 6 & 6 & 3 & 3 \end{bmatrix} = \mathbf{X}
 \end{aligned}$$

MATLAB function: *ifft2*

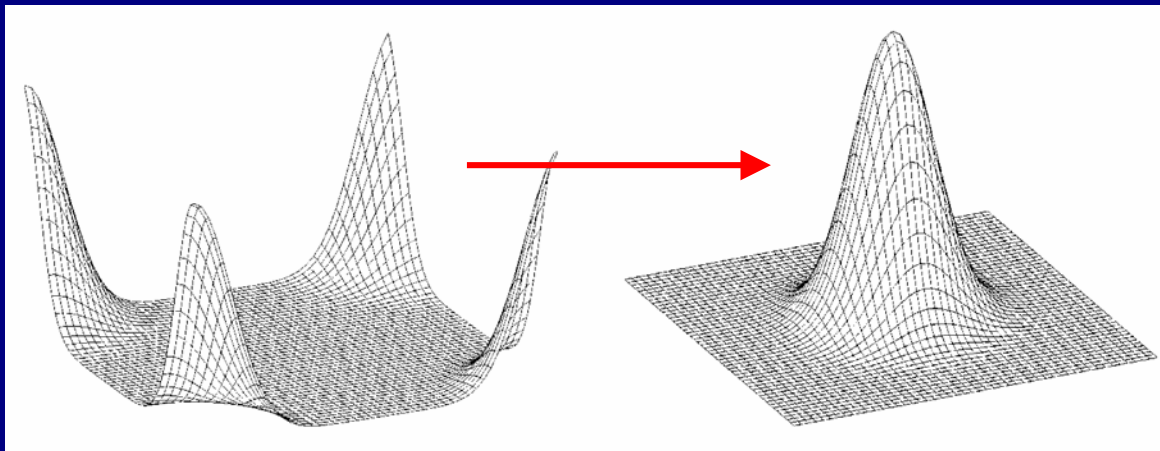
Centered Representation



MATLAB
function: *fftshift*

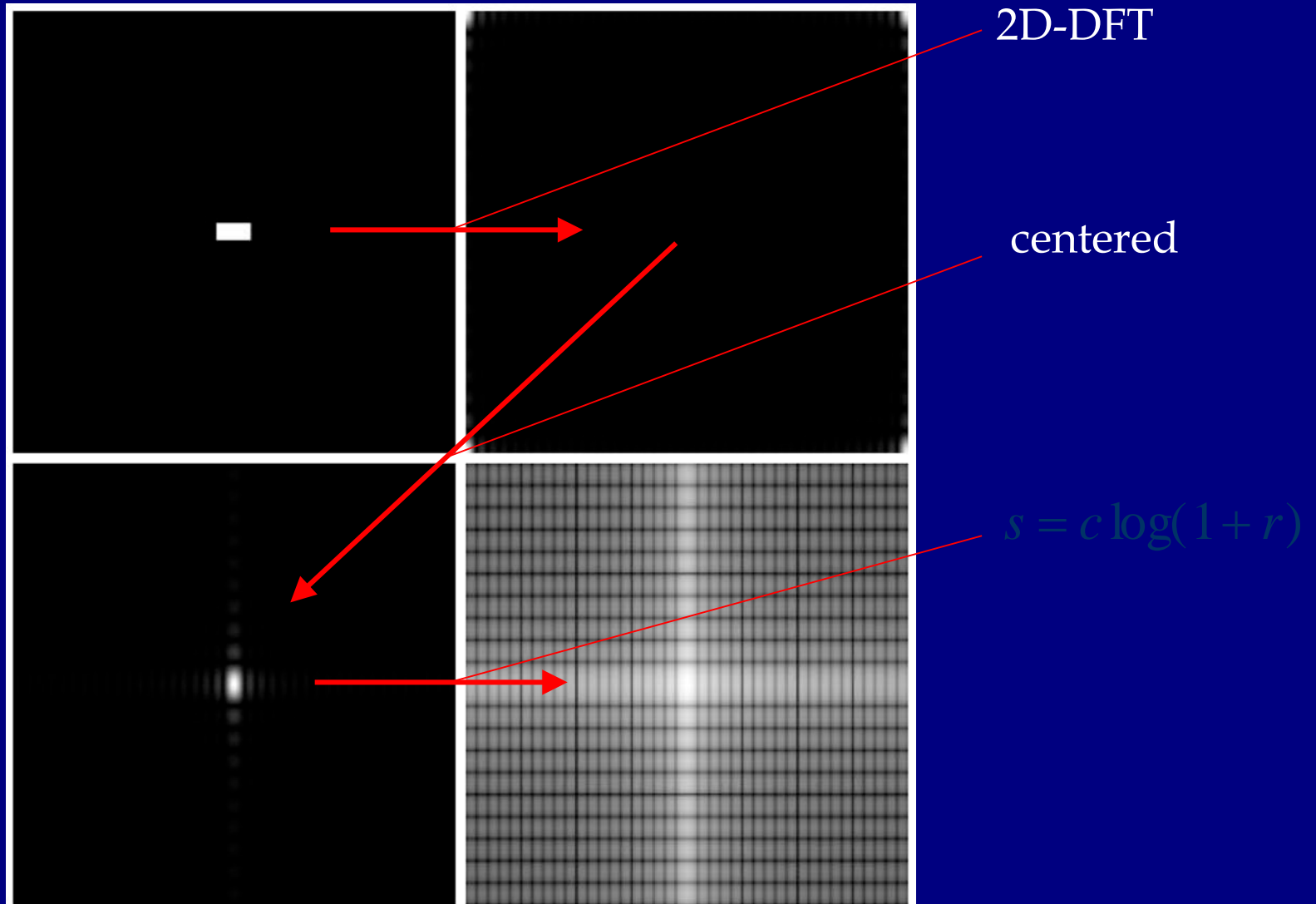
From Prof. Al Bovik

Example:



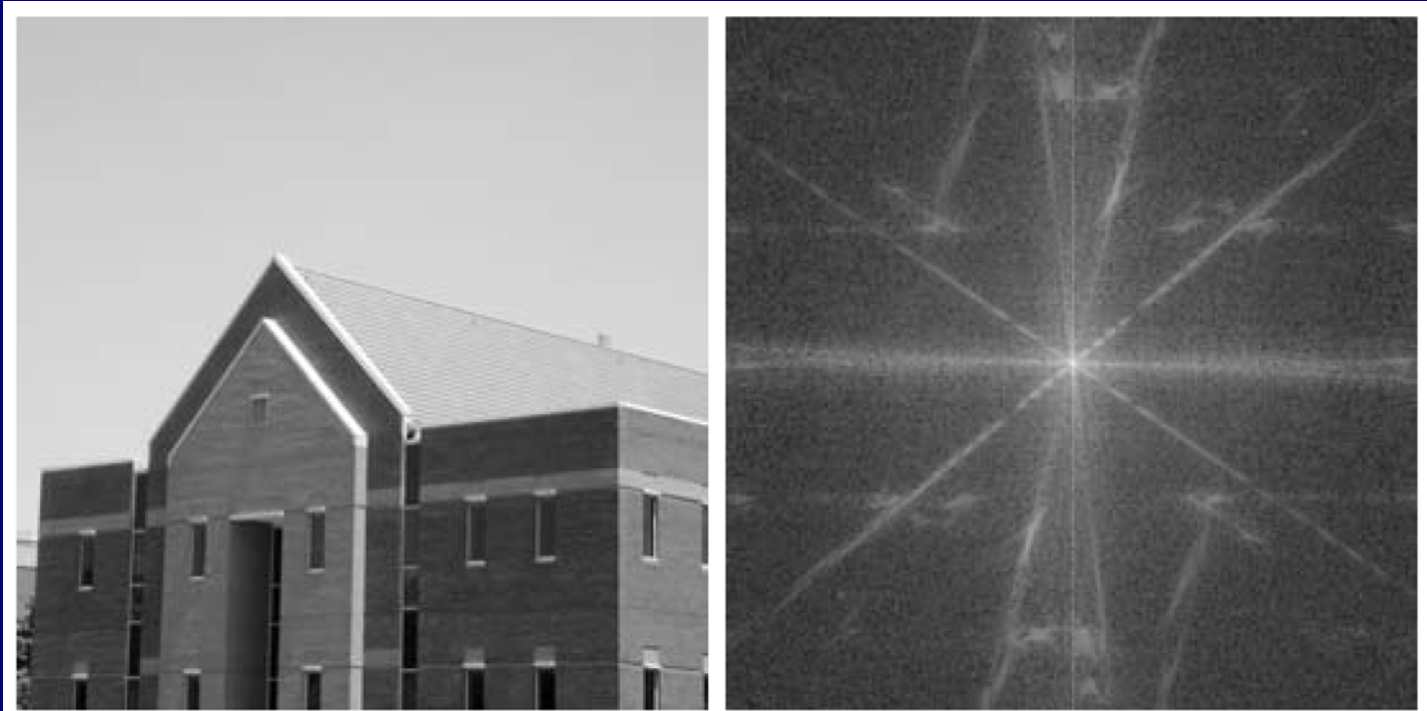
From [Gonzalez
& Woods]

Log-Magnitude Visualization



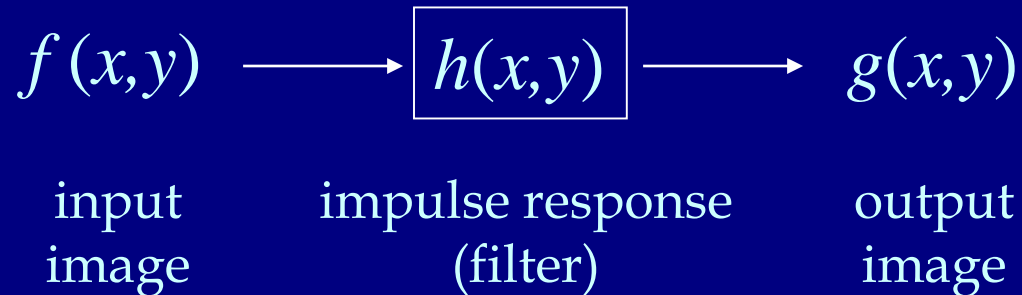
From [Gonzalez & Woods]

Apply to Images

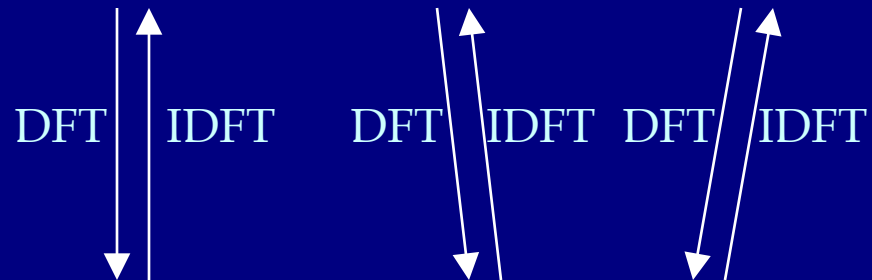


2D-DFT \rightarrow centered \rightarrow log intensity transformation

Convolution Theorem

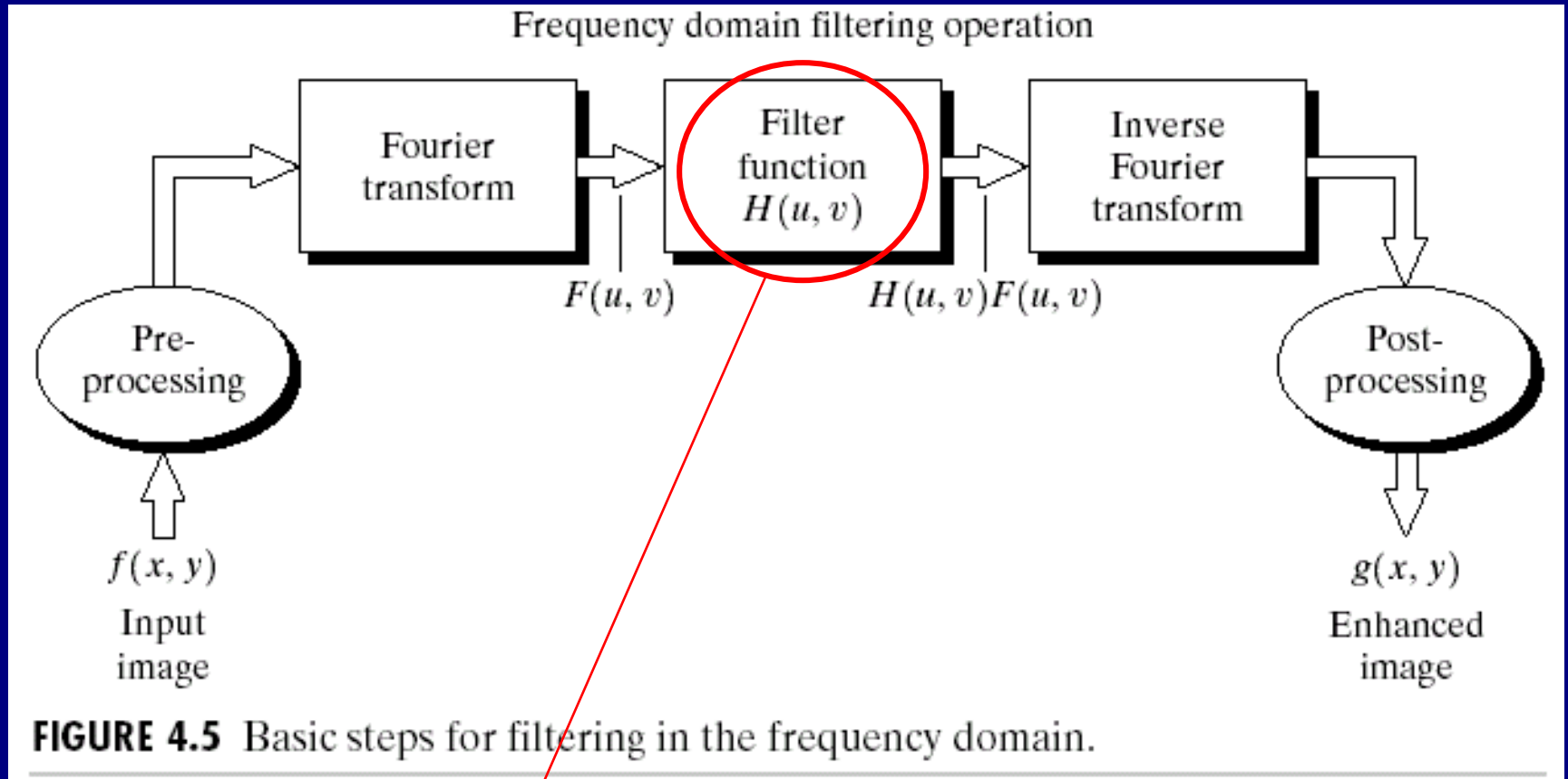


$$g(x, y) = f(x, y) \otimes h(x, y)$$



$$G(u, v) = F(u, v) H(u, v)$$

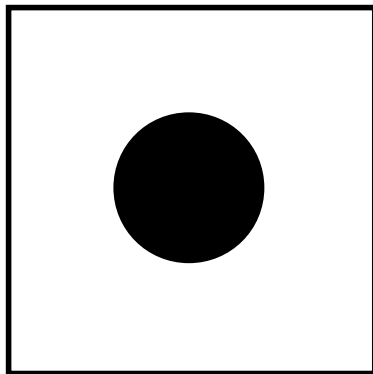
Frequency Domain Filtering



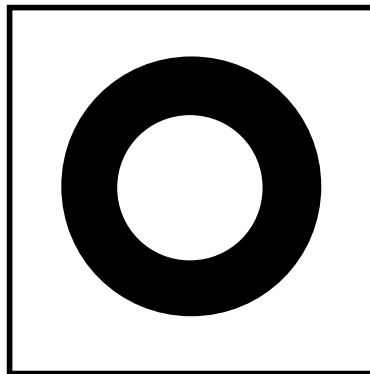
Filter design: design $H(u, v)$

2D-DFT Domain Filter Design

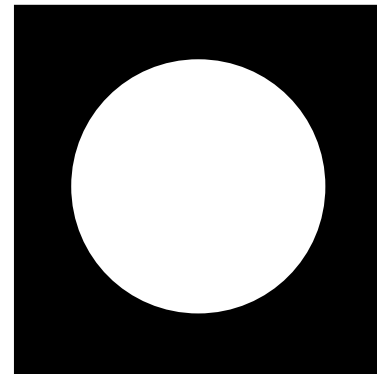
- Ideal lowpass, bandpass and highpass



low-frequency
mask



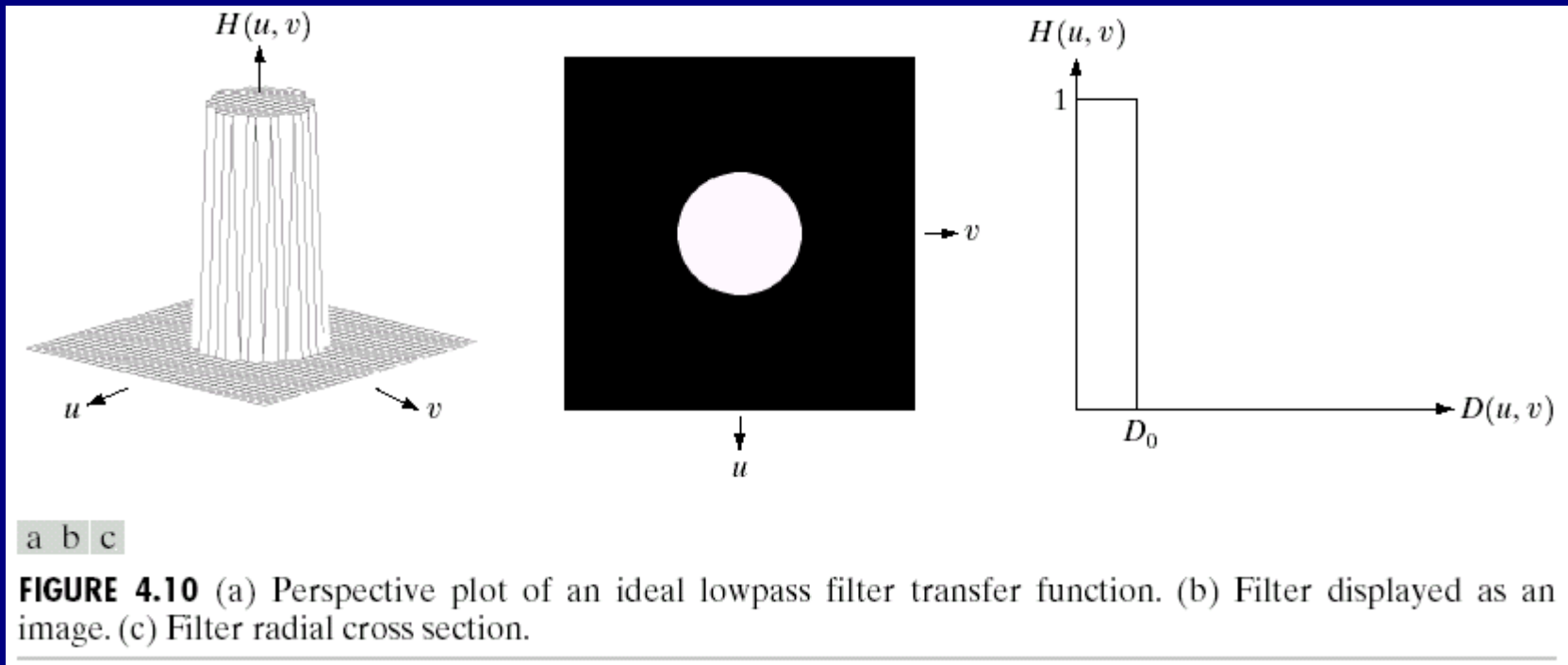
mid-frequency
mask



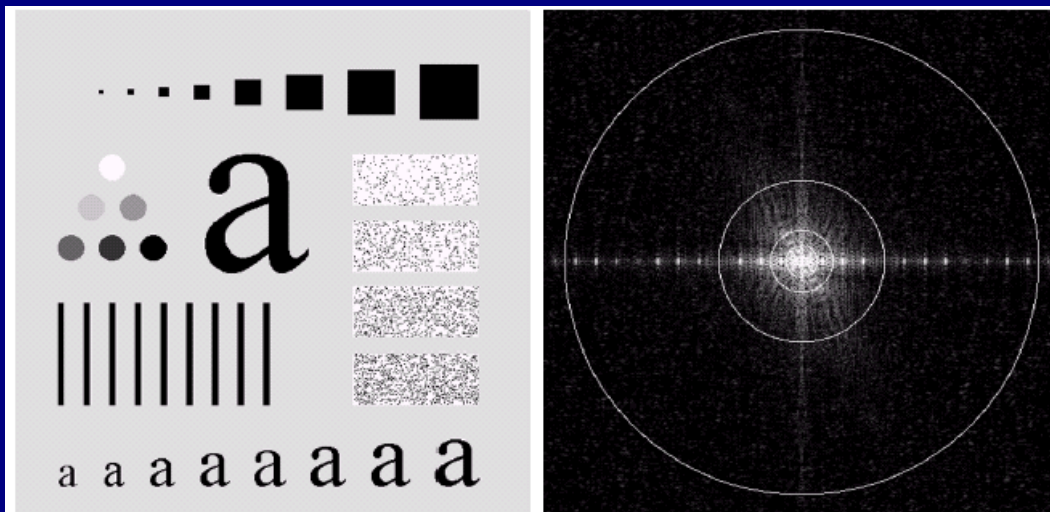
high-frequency
mask

2D-DFT Domain Filter Design

- Ideal lowpass, bandpass and highpass



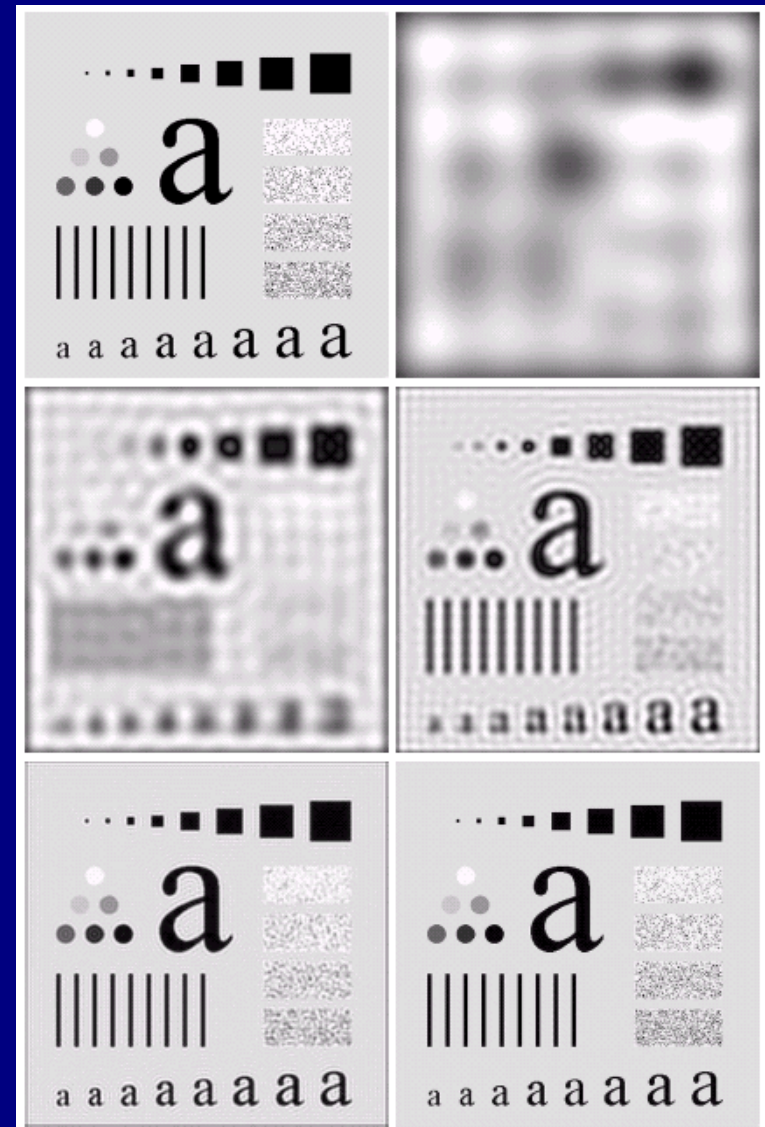
2D-DFT Domain Filter Design



a b

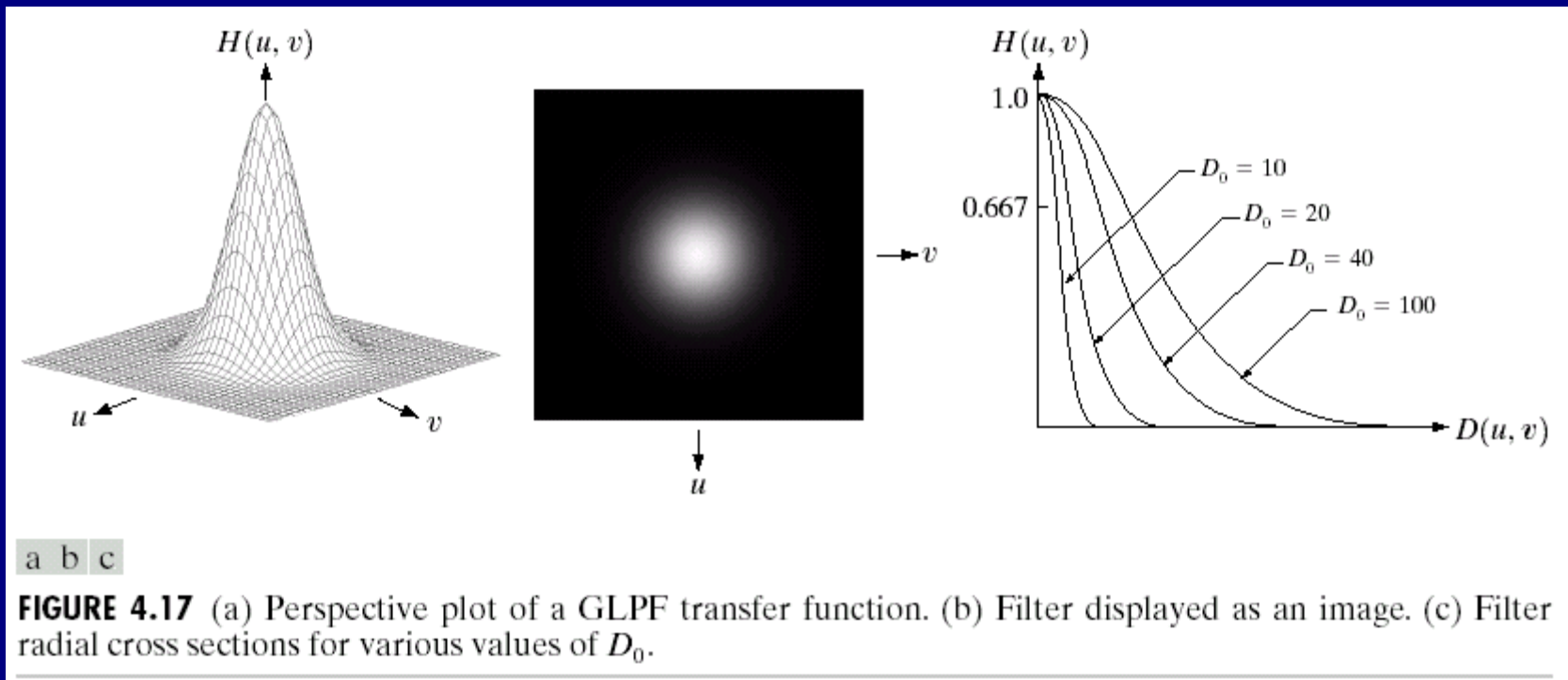
FIGURE 4.11 (a) An image of size 500×500 pixels and (b) its Fourier spectrum. The superimposed circles have radii values of 5, 15, 30, 80, and 230, which enclose 92.0, 94.6, 96.4, 98.0, and 99.5% of the image power, respectively.

Ideal lowpass filtering with cutoff frequencies set at radii values of 5, 15, 30, 80, and 230, respectively

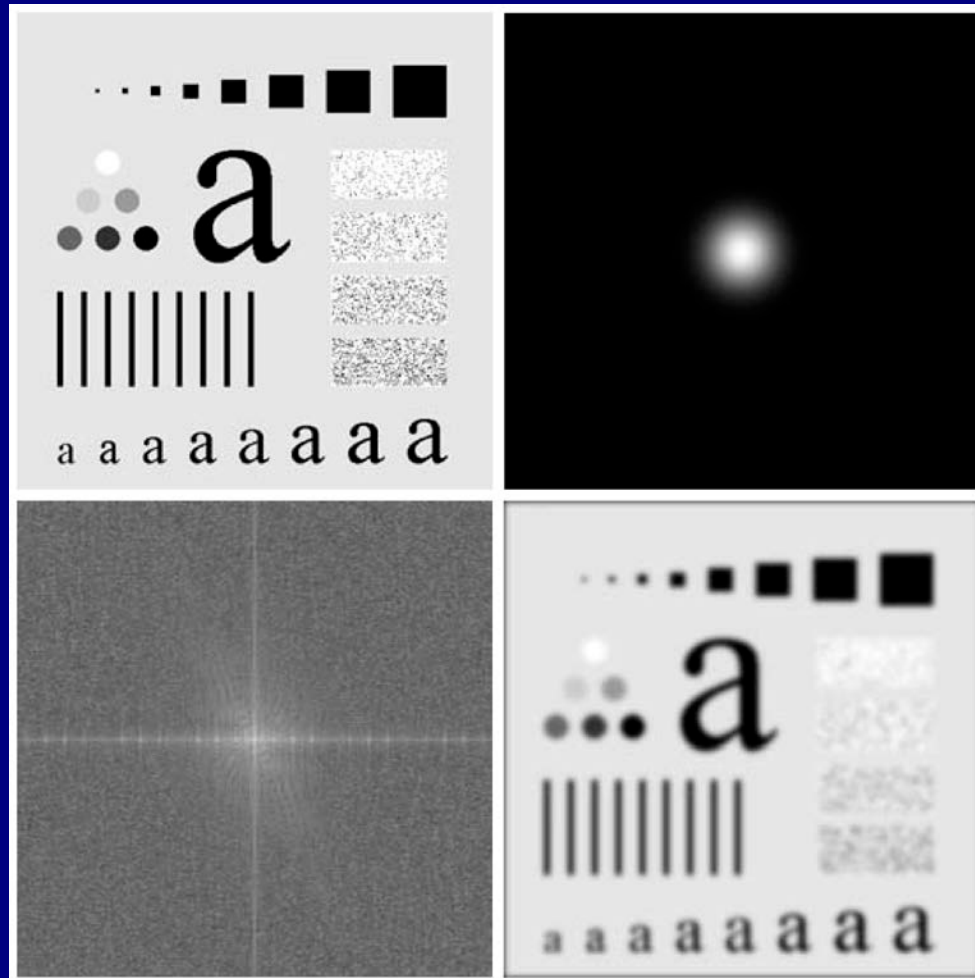


2D-DFT Domain Filter Design

- Gaussian lowpass



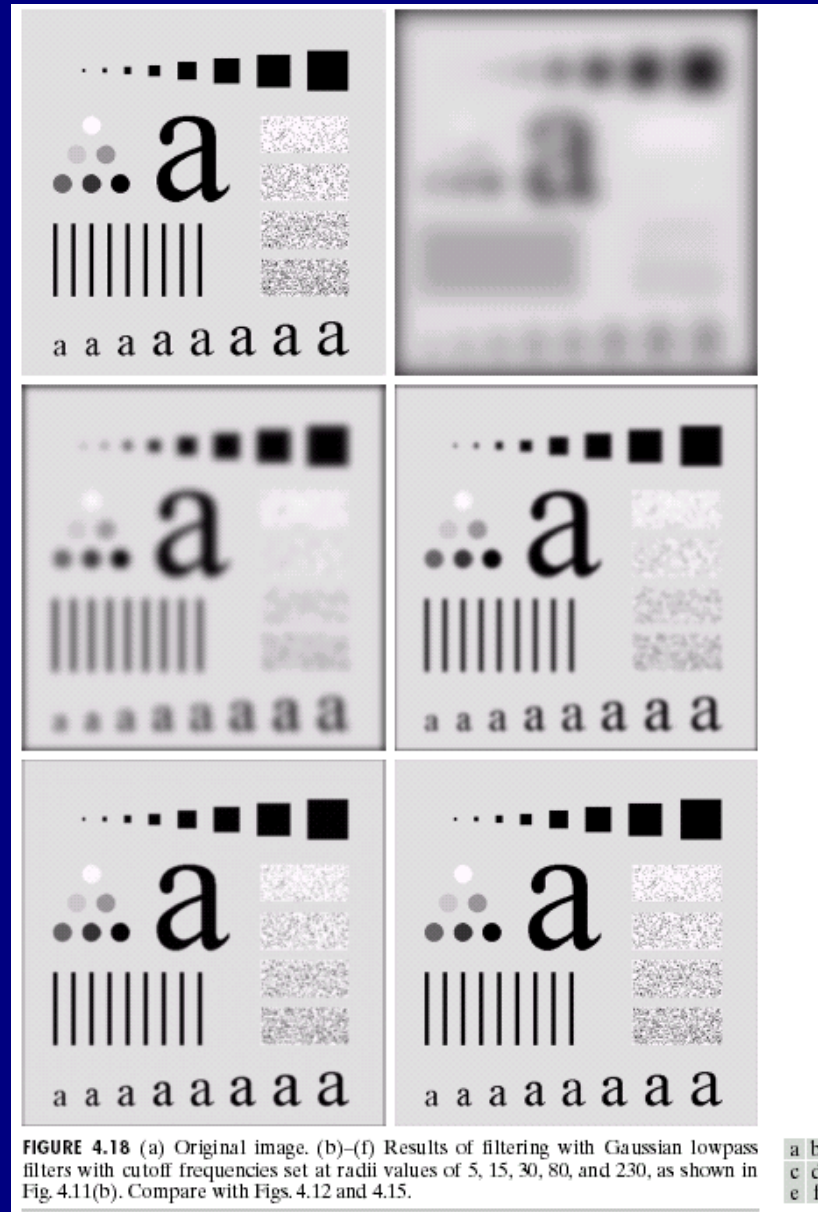
2D-DFT Domain Filter Design



Effect of Gaussian lowpass filter

2D-DFT Domain Filter Design

Effect of
Gaussian
lowpass
filter



From [Gonzalez & Woods] ²⁴

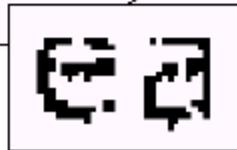
2D-DFT Domain Filter Design

a b

FIGURE 4.19

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

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.

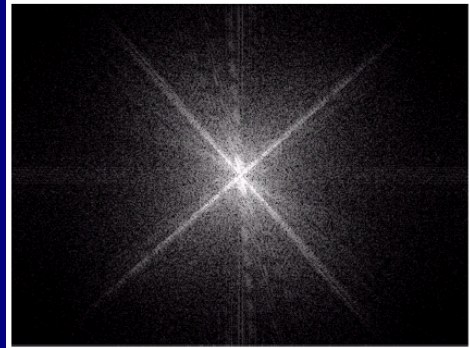
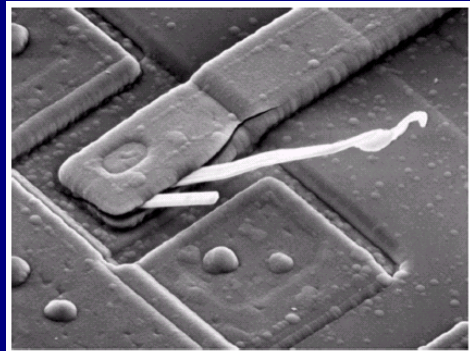


Historically, certain computer programs were written using only two digits rather than four to define the applicable year. Accordingly, the company's software may recognize a date using "00" as 1900 rather than the year 2000.



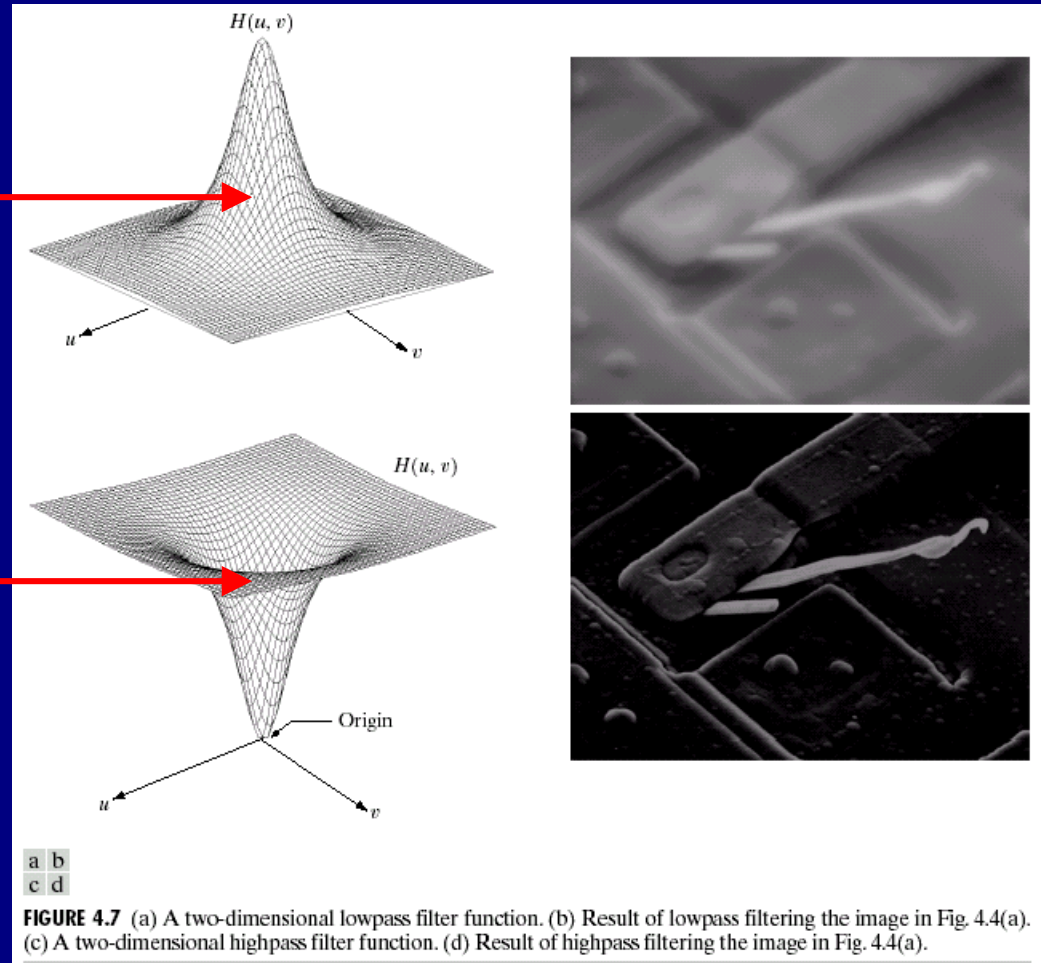
Effect of Gaussian lowpass filter

2D-DFT Domain Filter Design



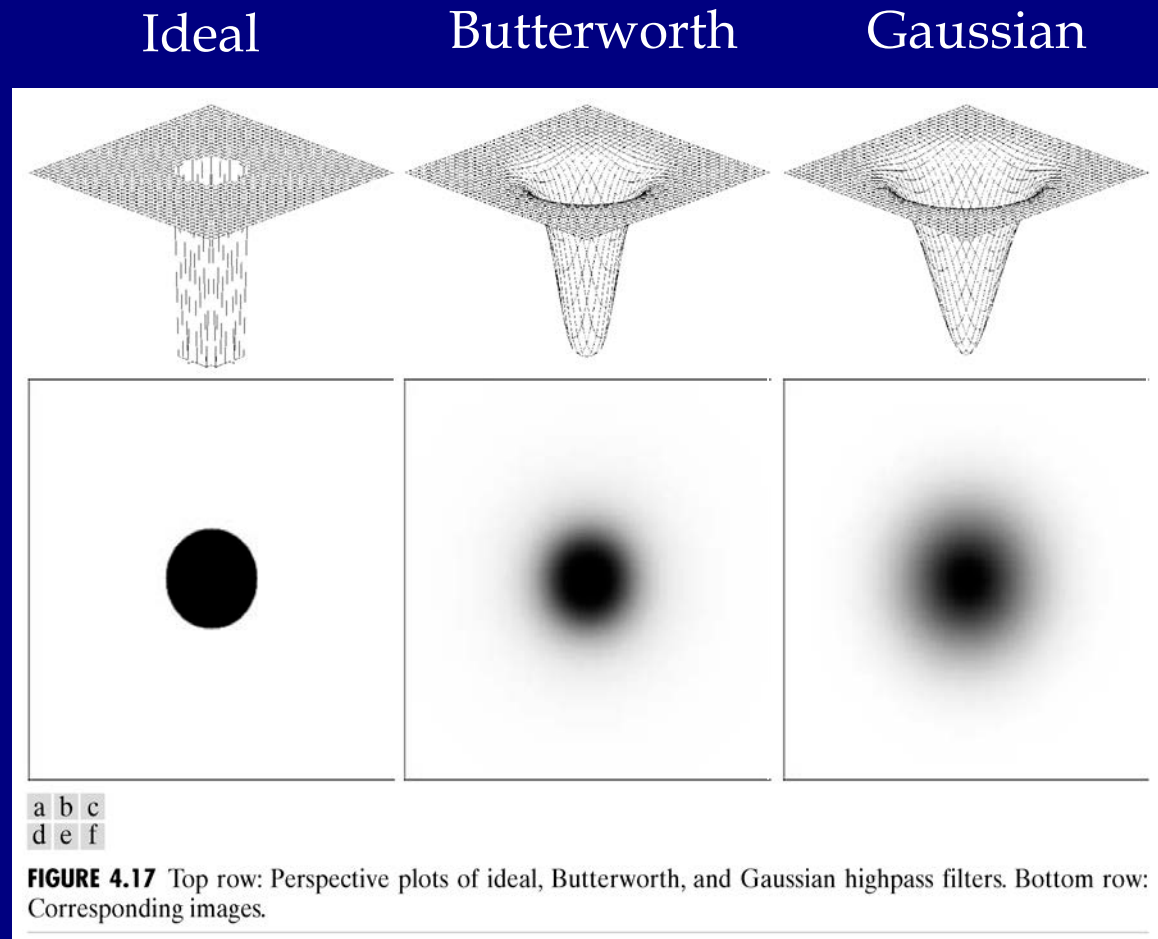
Gaussian
lowpass
filtering

Gaussian
highpass
filtering

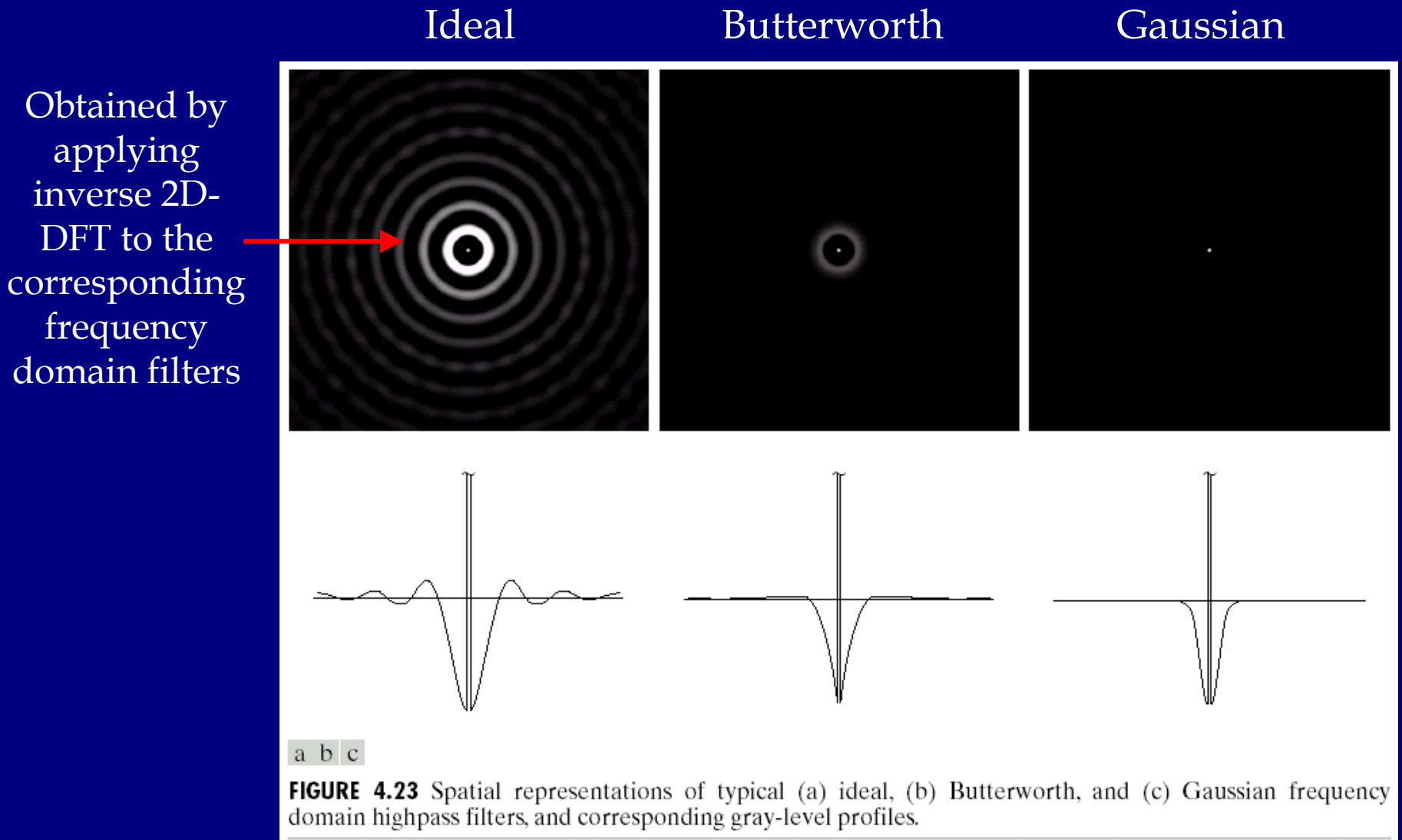


2D-DFT Domain Filter Design

- Choices of highpass filters



2D-DFT Domain Filter Design

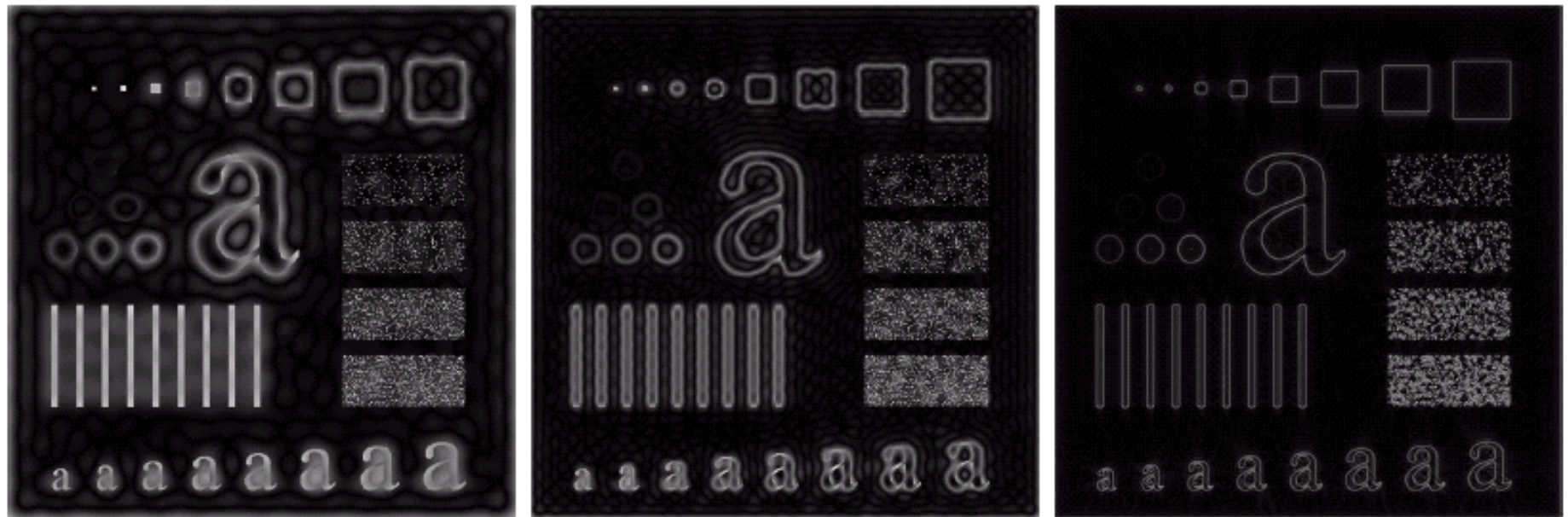


2D-DFT Domain Filter Design

Ideal

Butterworth

Gaussian



a b c

FIGURE 4.24 Results of ideal highpass filtering the image in Fig. 4.11(a) with $D_0 = 15, 30$, and 80 , respectively. Problems with ringing are quite evident in (a) and (b).

2D-DFT Domain Filter Design

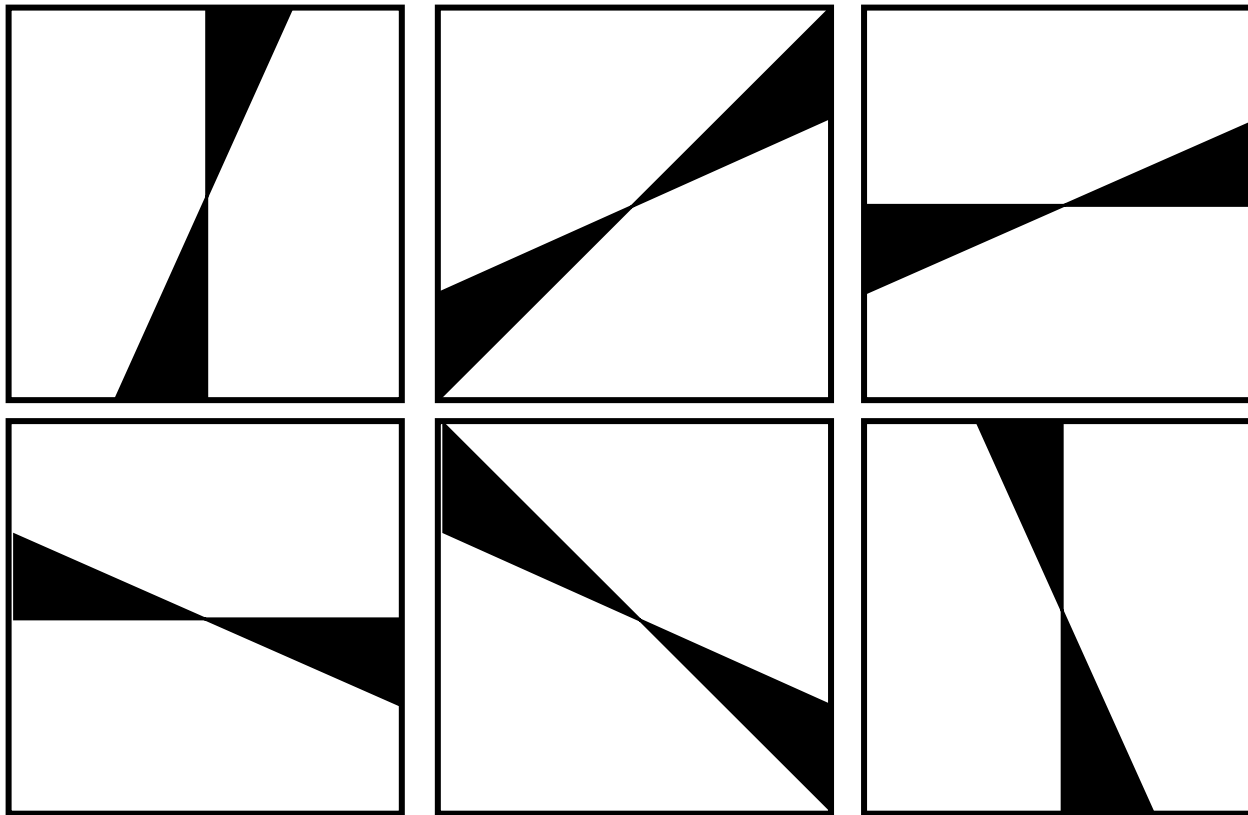
Gaussian filter with different width



FIGURE 4.26 Results of highpass filtering the image of Fig. 4.11(a) using a GHPF of order 2 with $D_0 = 15$, 30, and 80, respectively. Compare with Figs. 4.24 and 4.25.

2D-DFT Domain Filter Design

- Orientation selective filters



2D-DFT Domain Filter Design

- Narrowband Filtering

by combining radial and orientation selection

