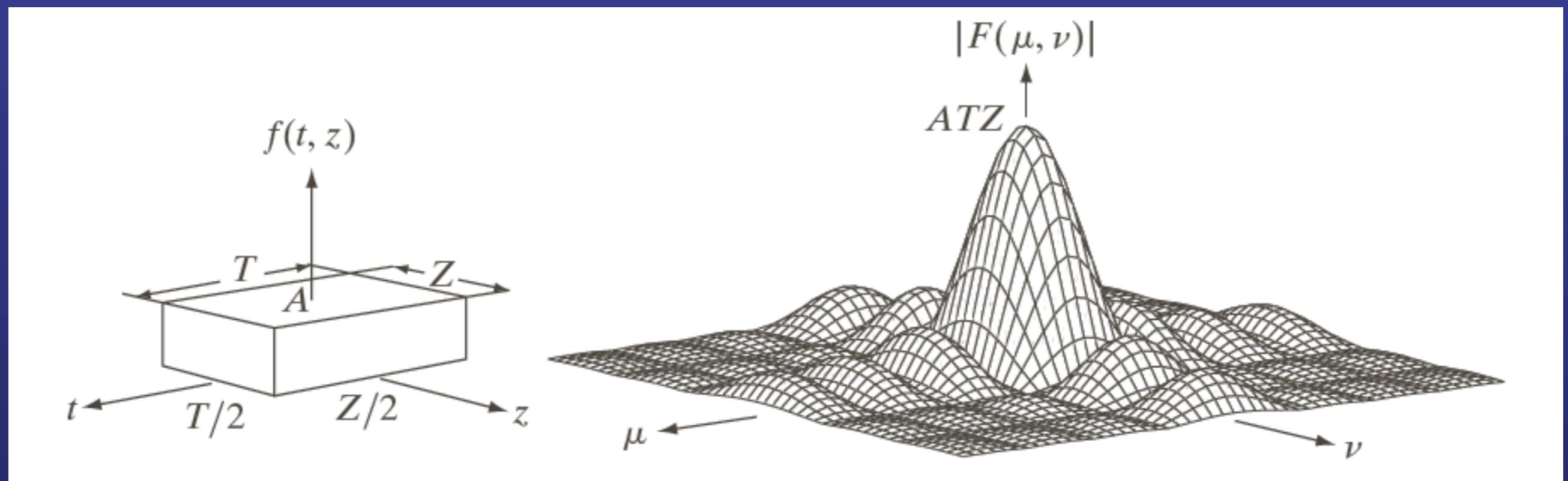


# Image Enhancement in the Frequency Domain

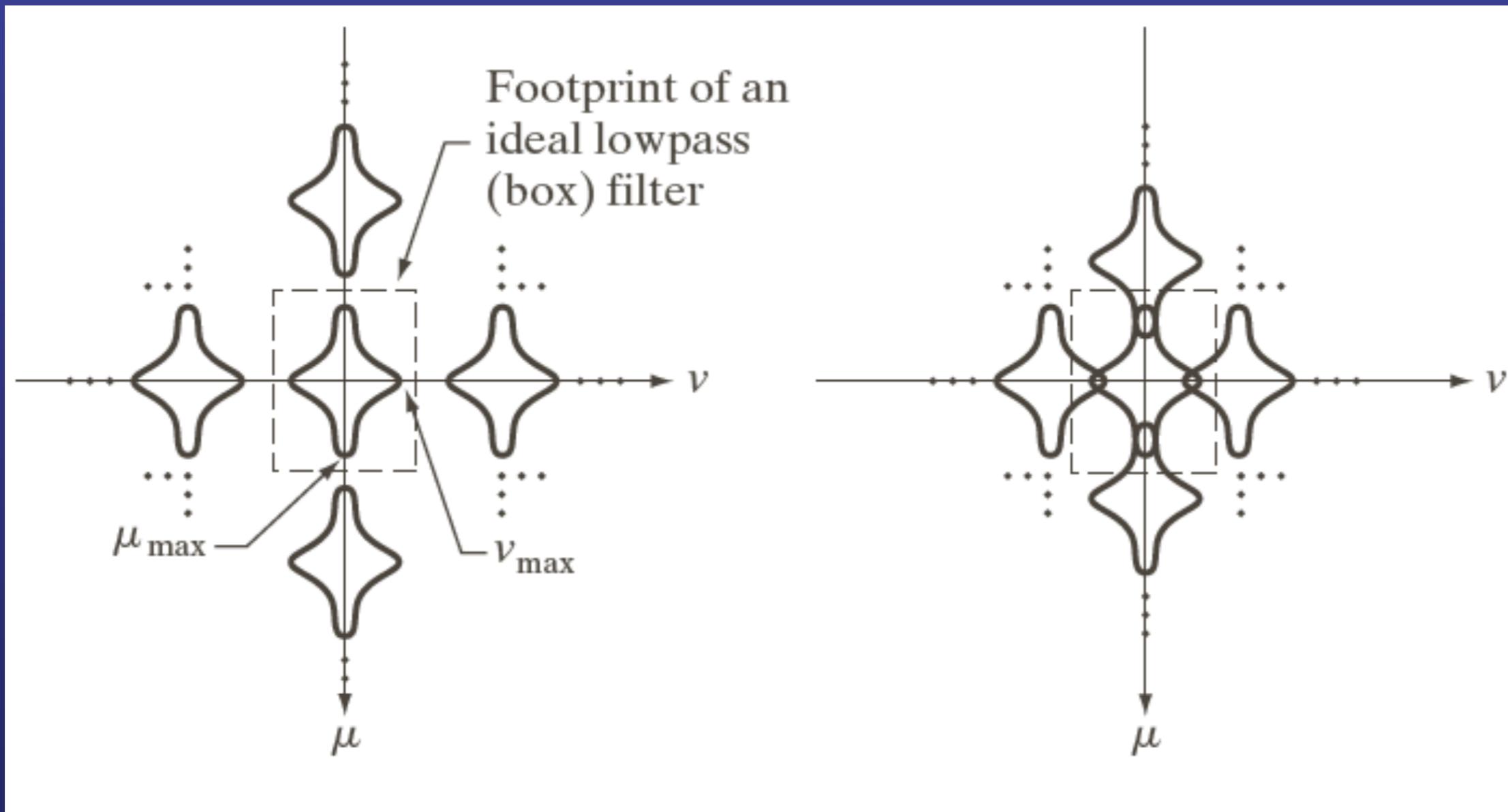
柯正雯

# 2D Sinc function



# over sampled vs. under sampled

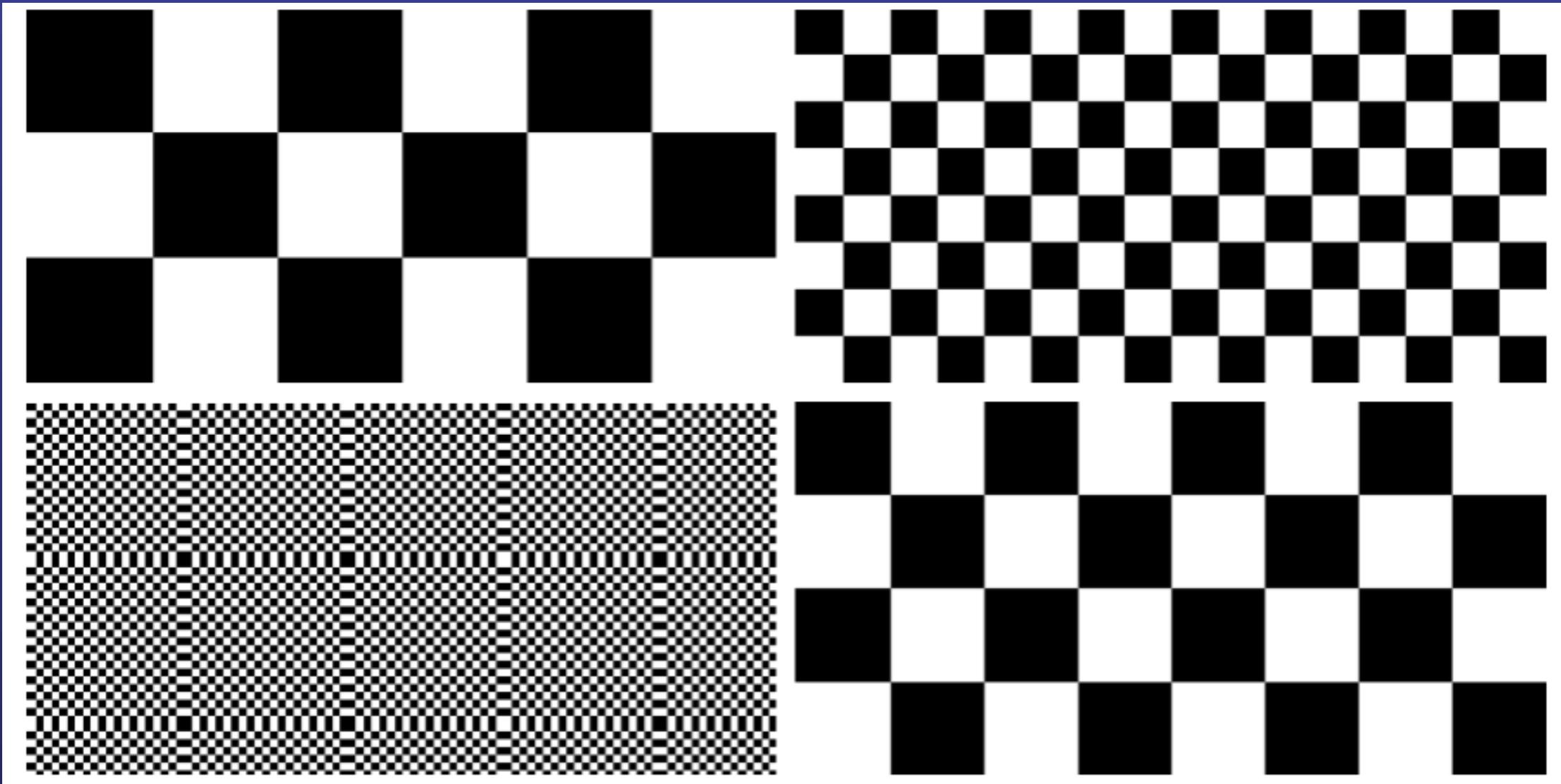
---



# Aliasing

---

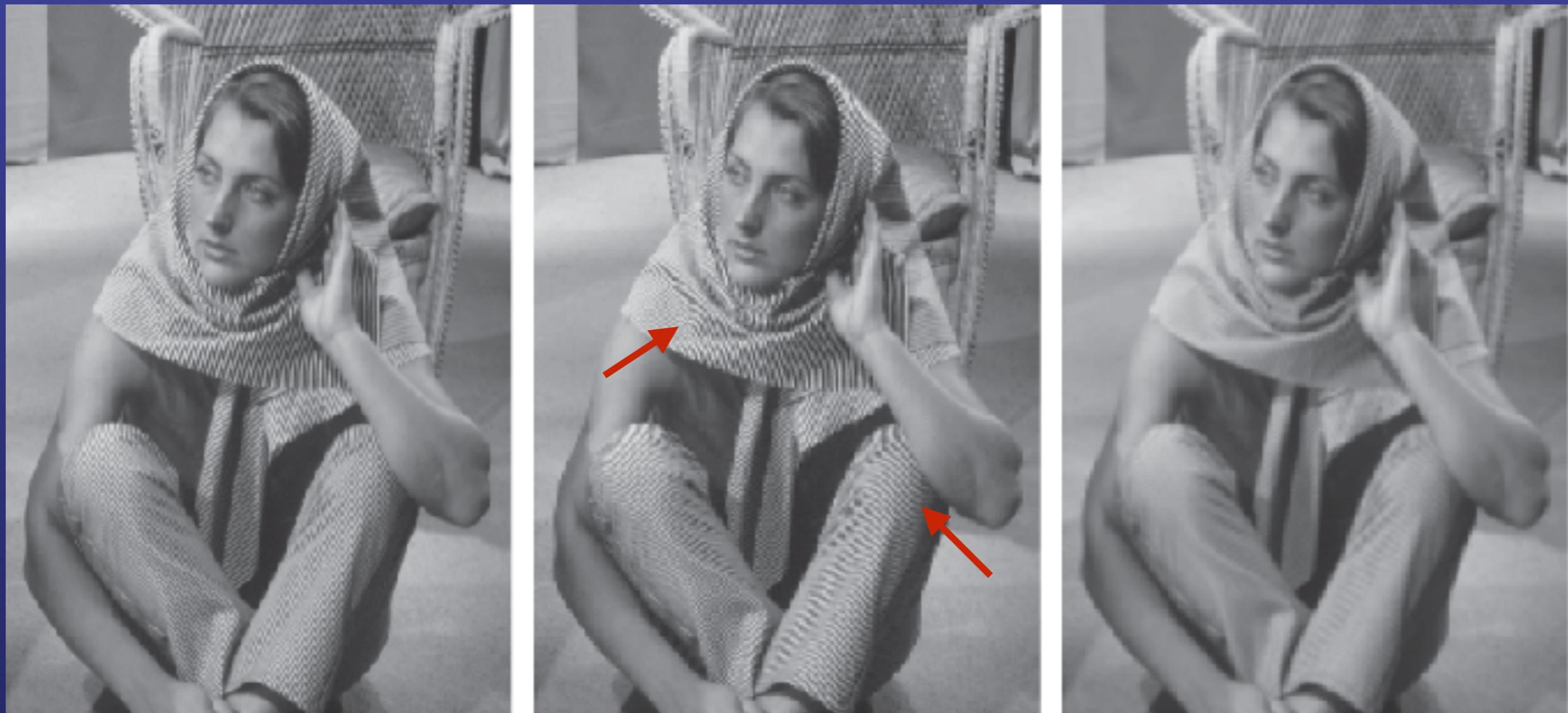
Normal Images



Aliased Images

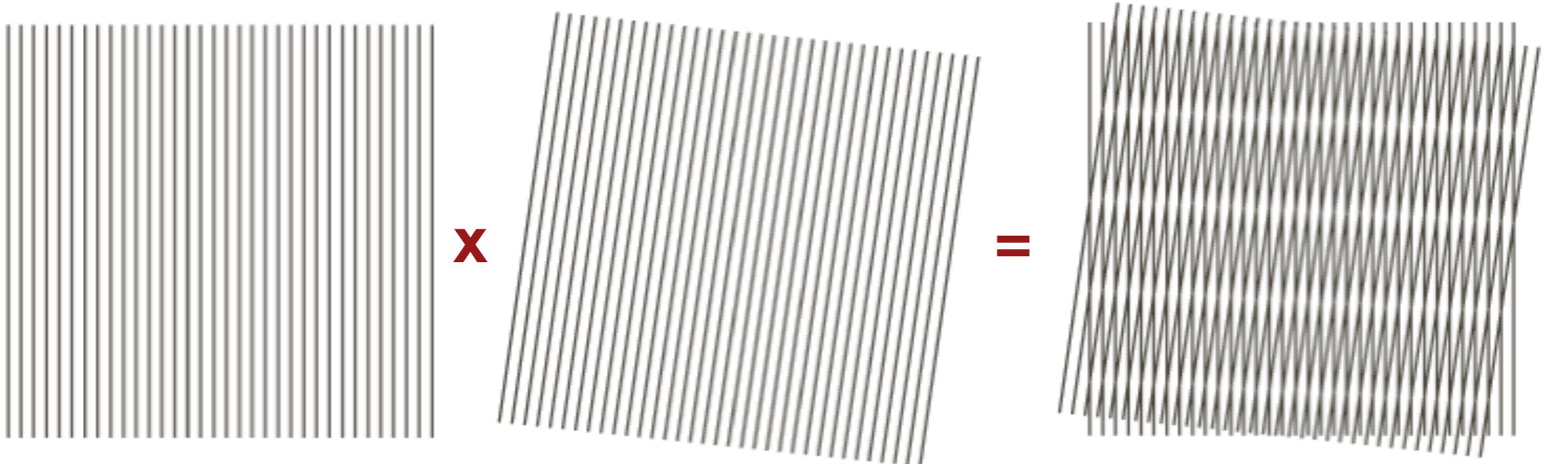
# 2D Sampling & Aliasing

---



# Moiré Patterns

---



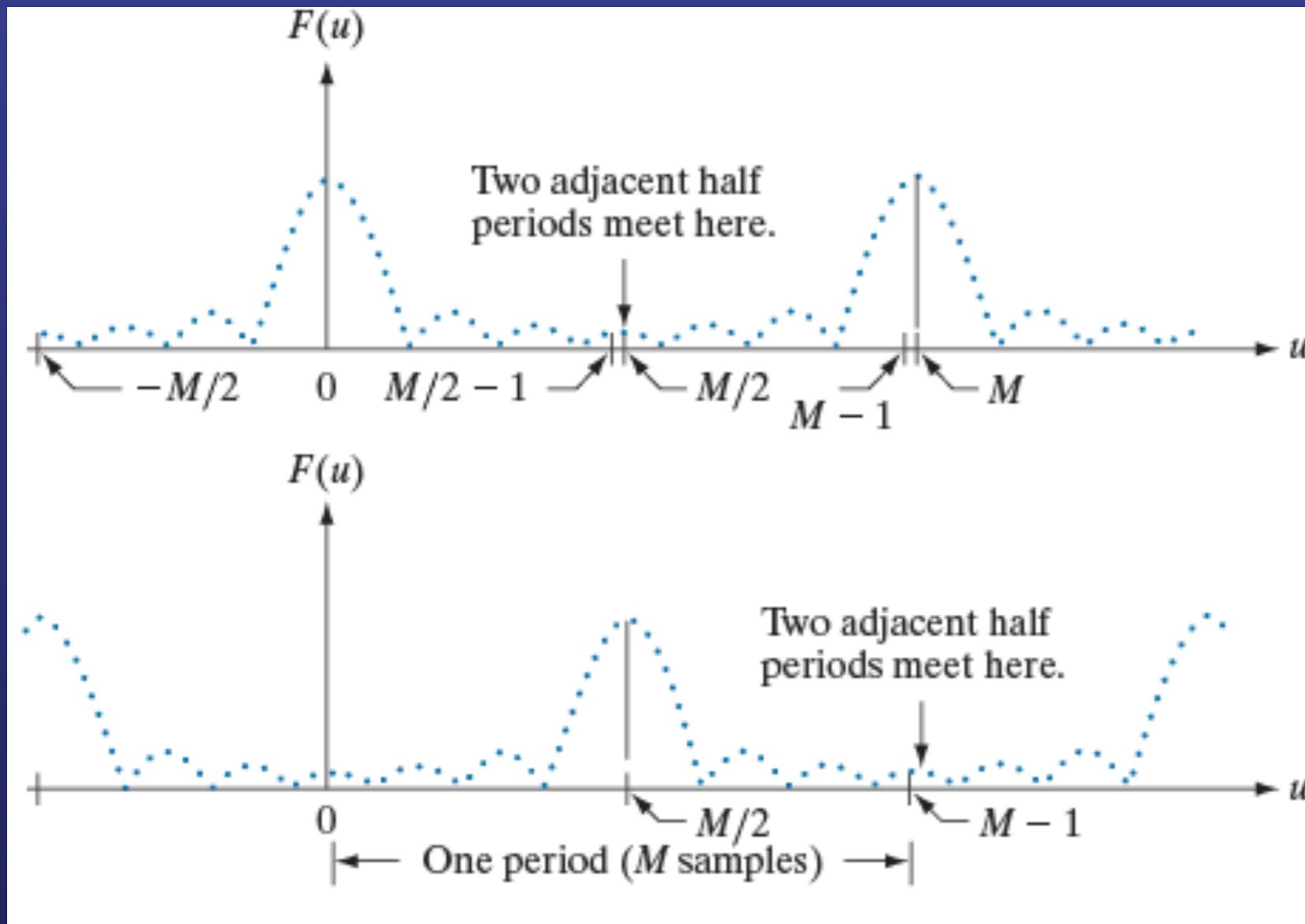
# Moiré Patterns

---



# Periodicity

$$f(x) e^{j2\pi(u_0x/M)} \Leftrightarrow F(u - u_0)$$

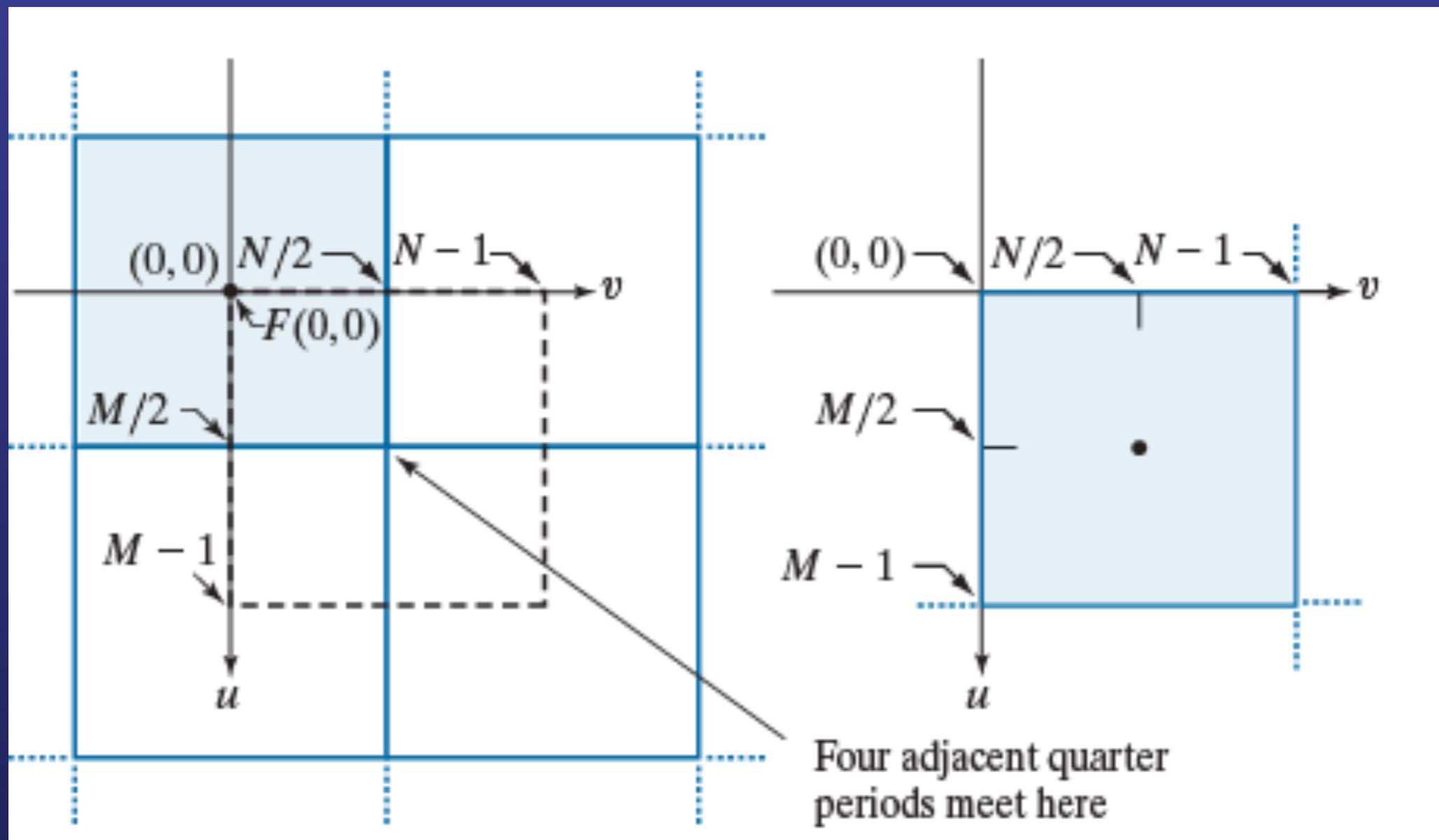


$$u_0 = M/2$$

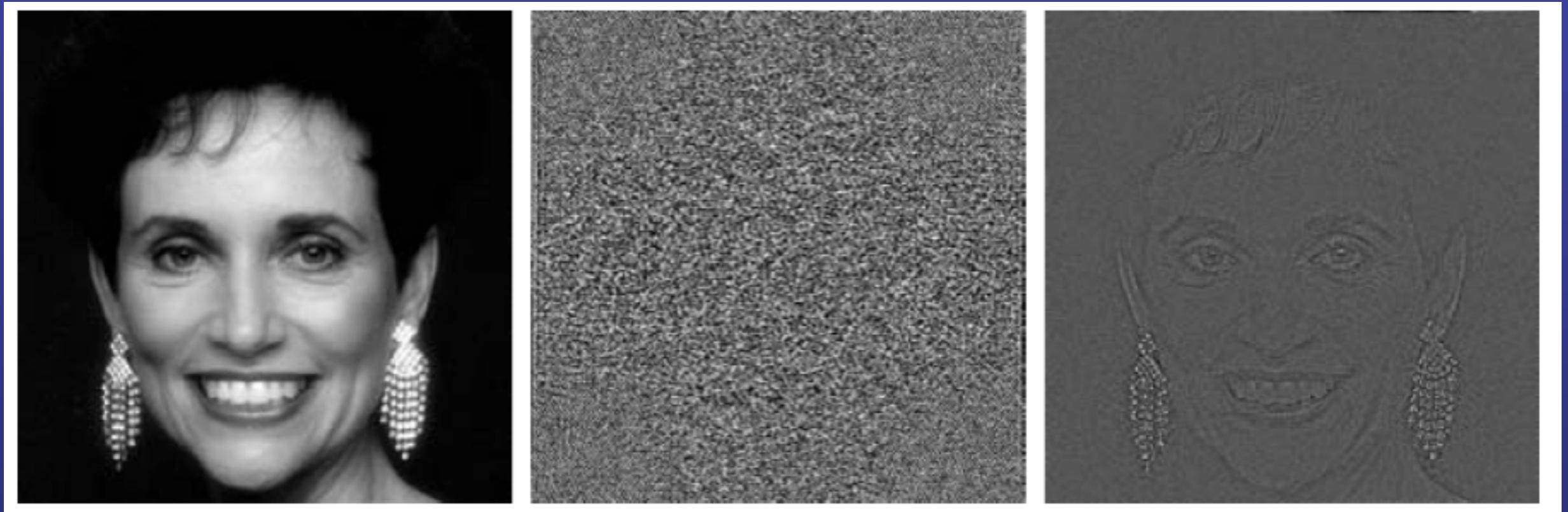
$$\begin{aligned} f(x) e^{j\pi x} \\ = f(x) (-1)^x \end{aligned}$$

# Periodicity

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



# Phase Images



$$f(x, y)$$

$$\phi(u, v)$$

$$g(x, y)$$

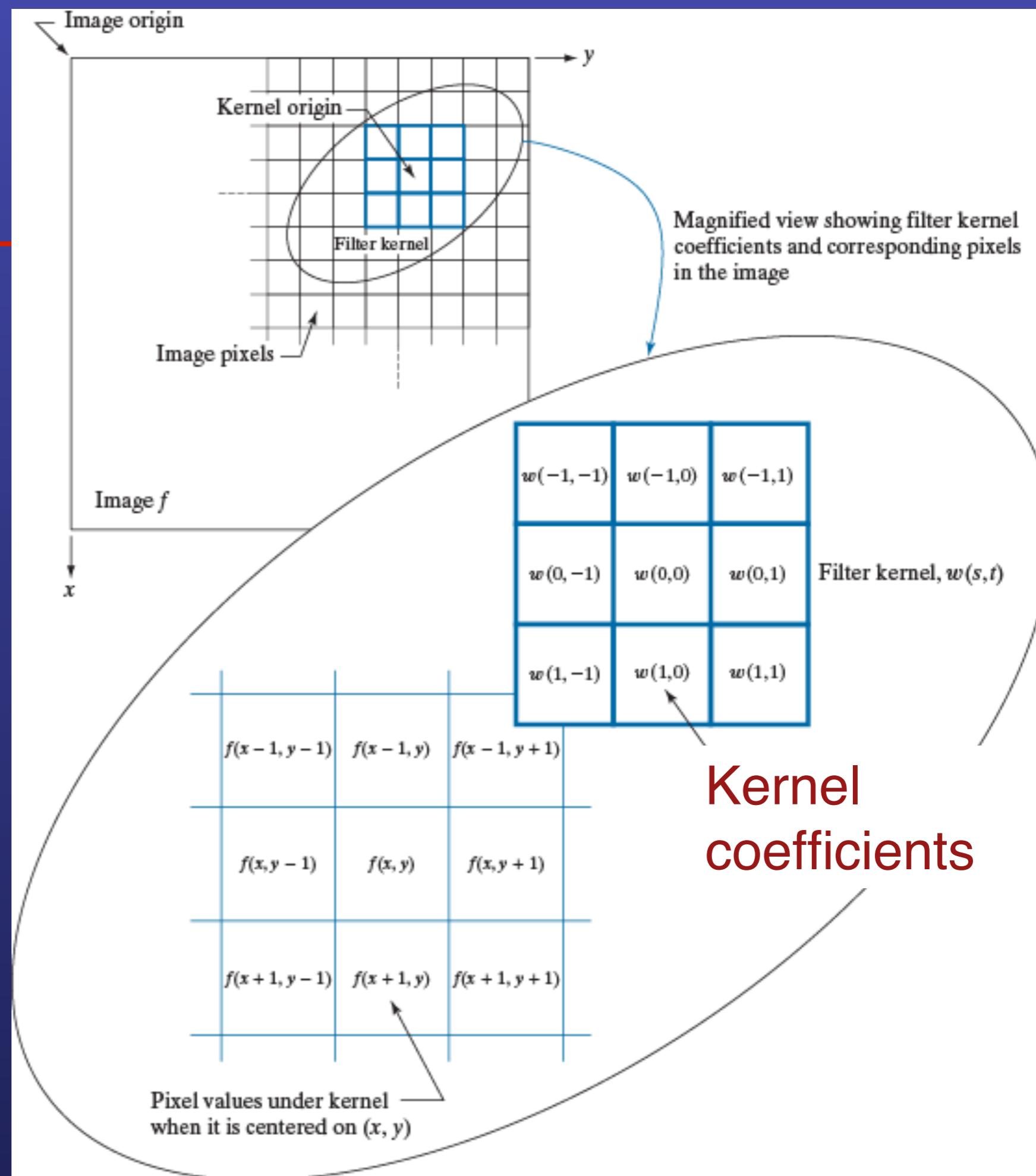
$$F(u, v)$$

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

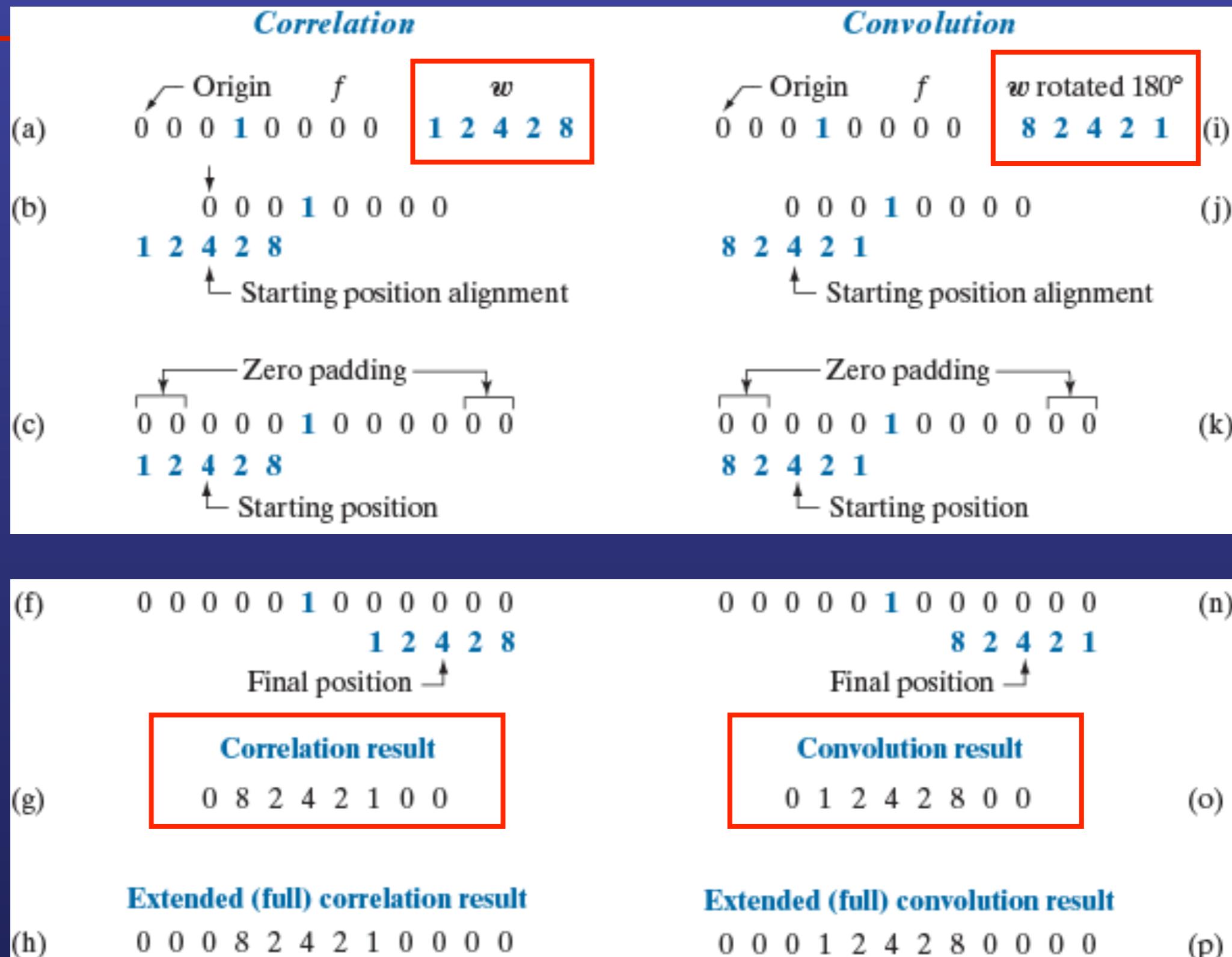


$$G(u, v) = 1 e^{j\phi(u, v)}$$

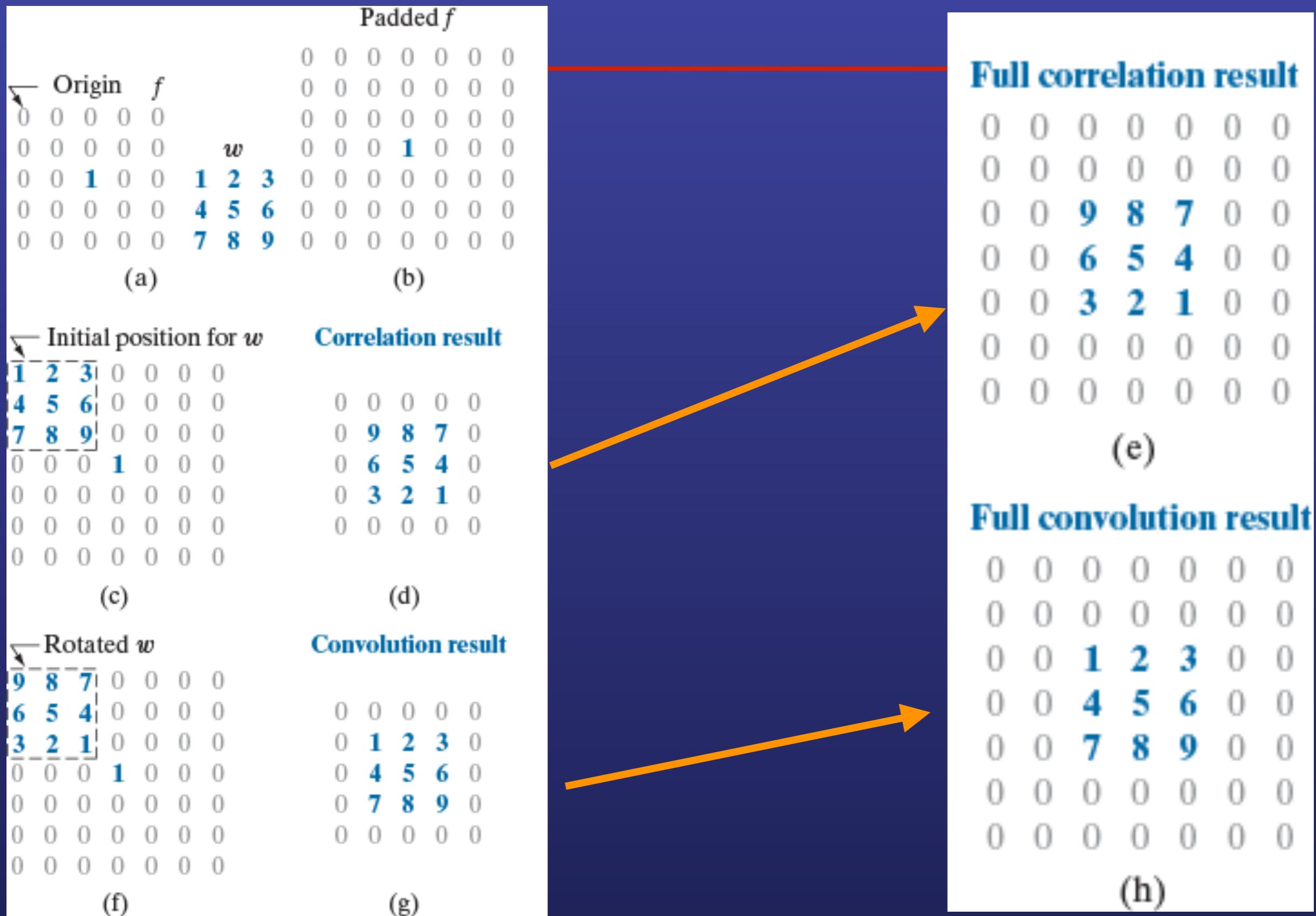
# Spatial Convolution



# Spatial Convolution

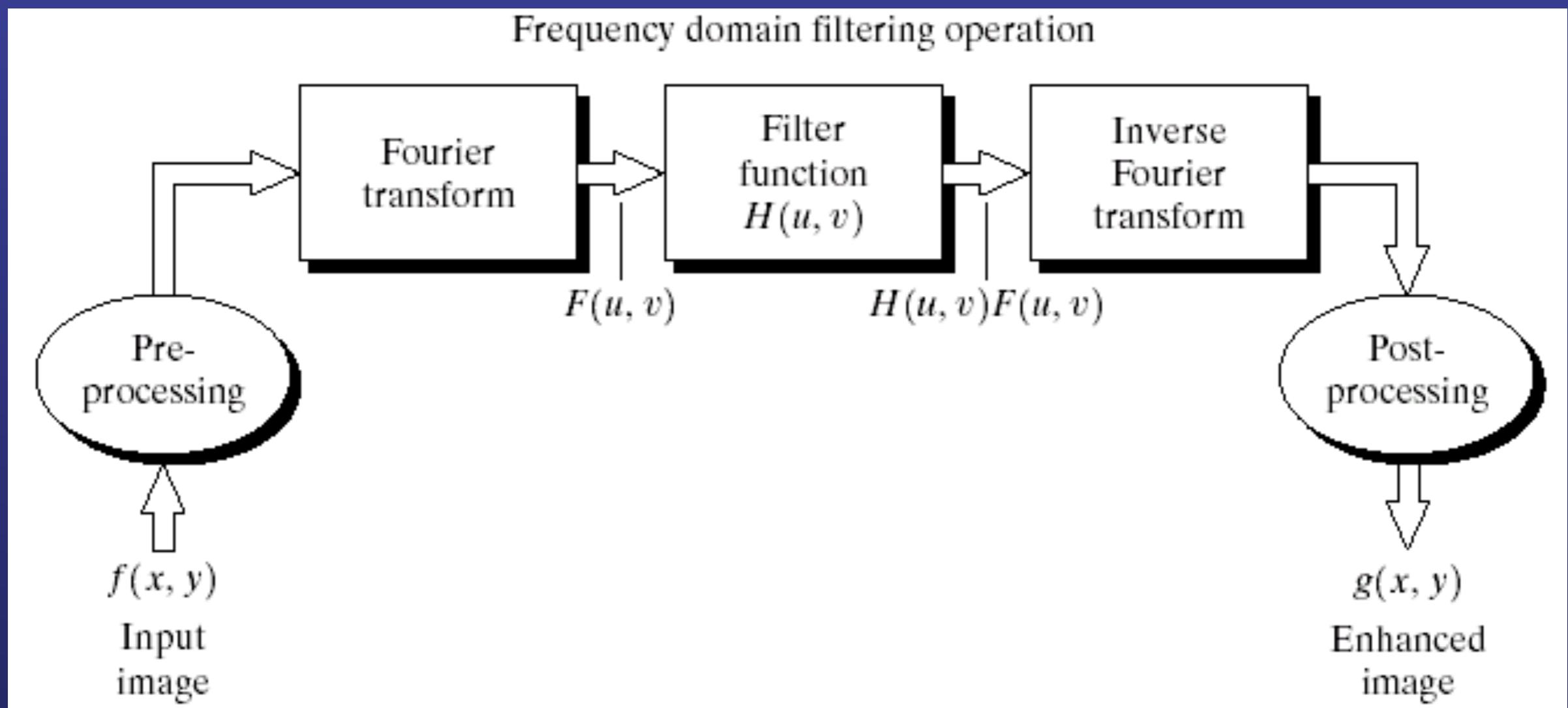


# Spatial Convolution



# Basic scheme of filtering

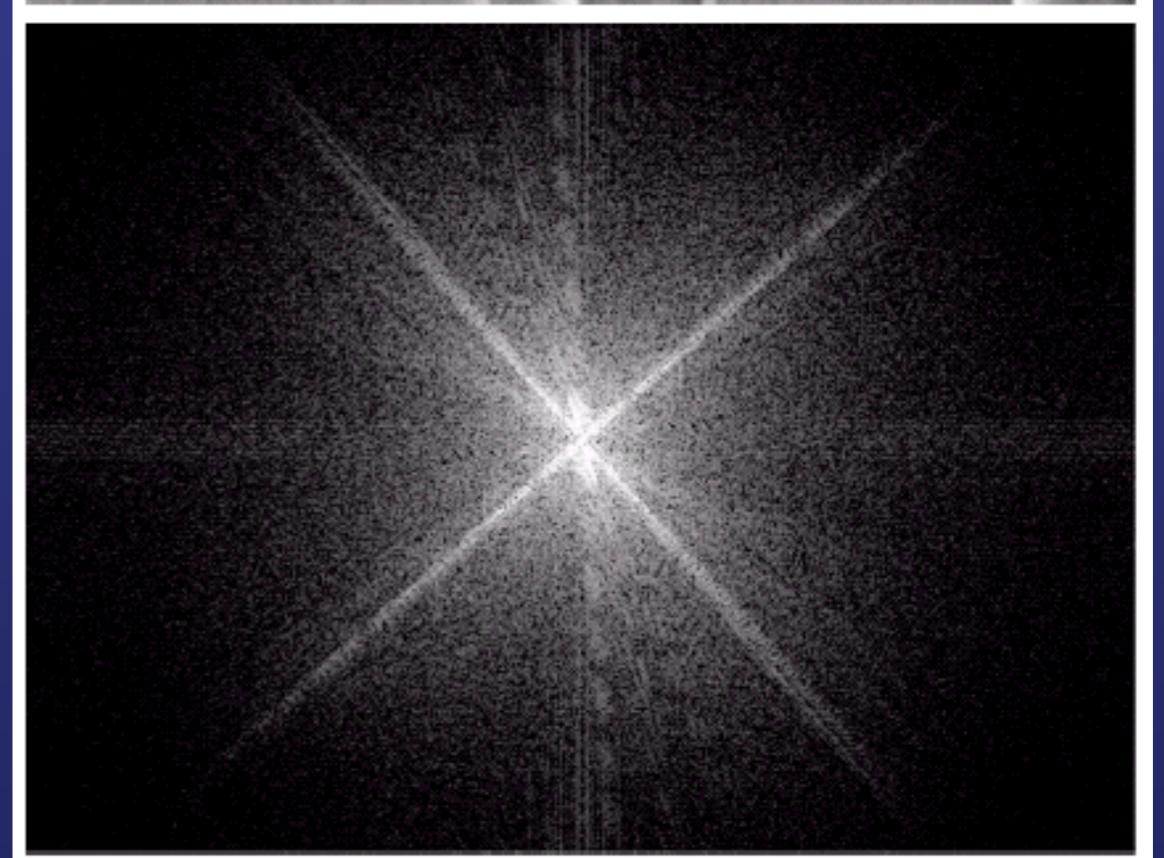
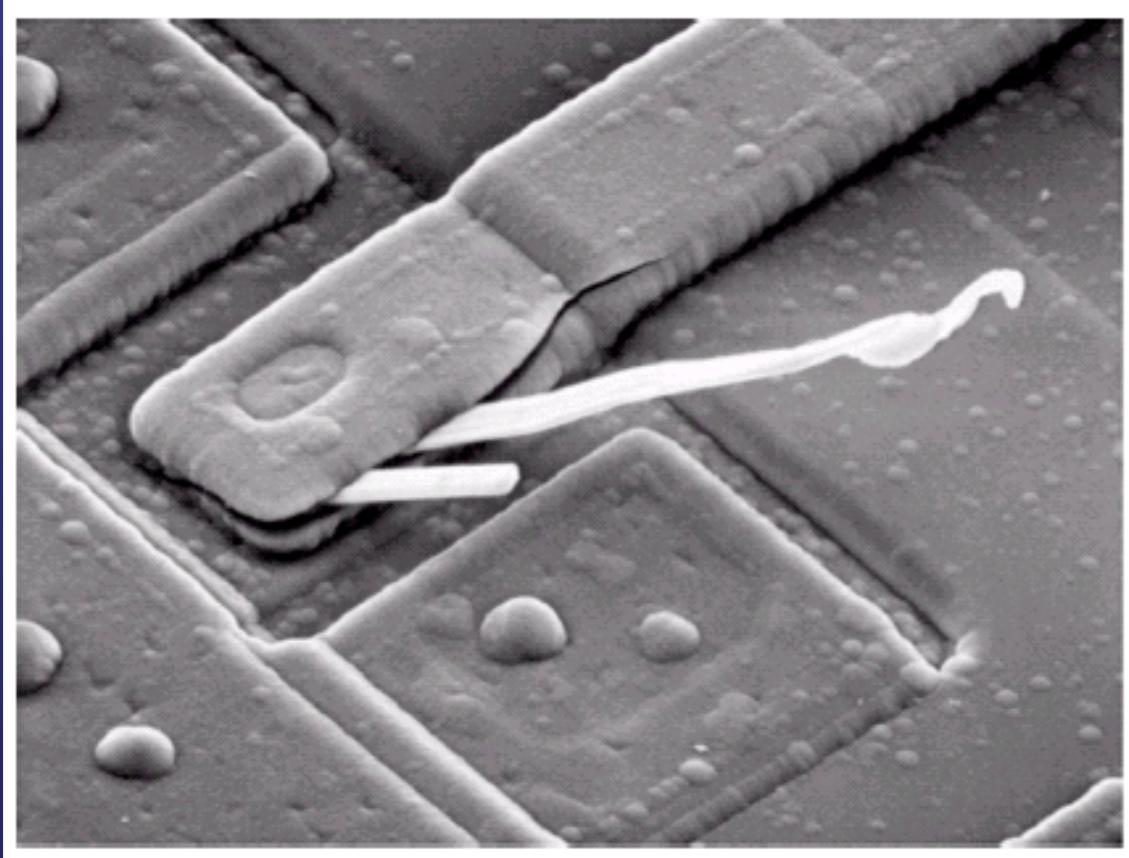
---



# Basic filter

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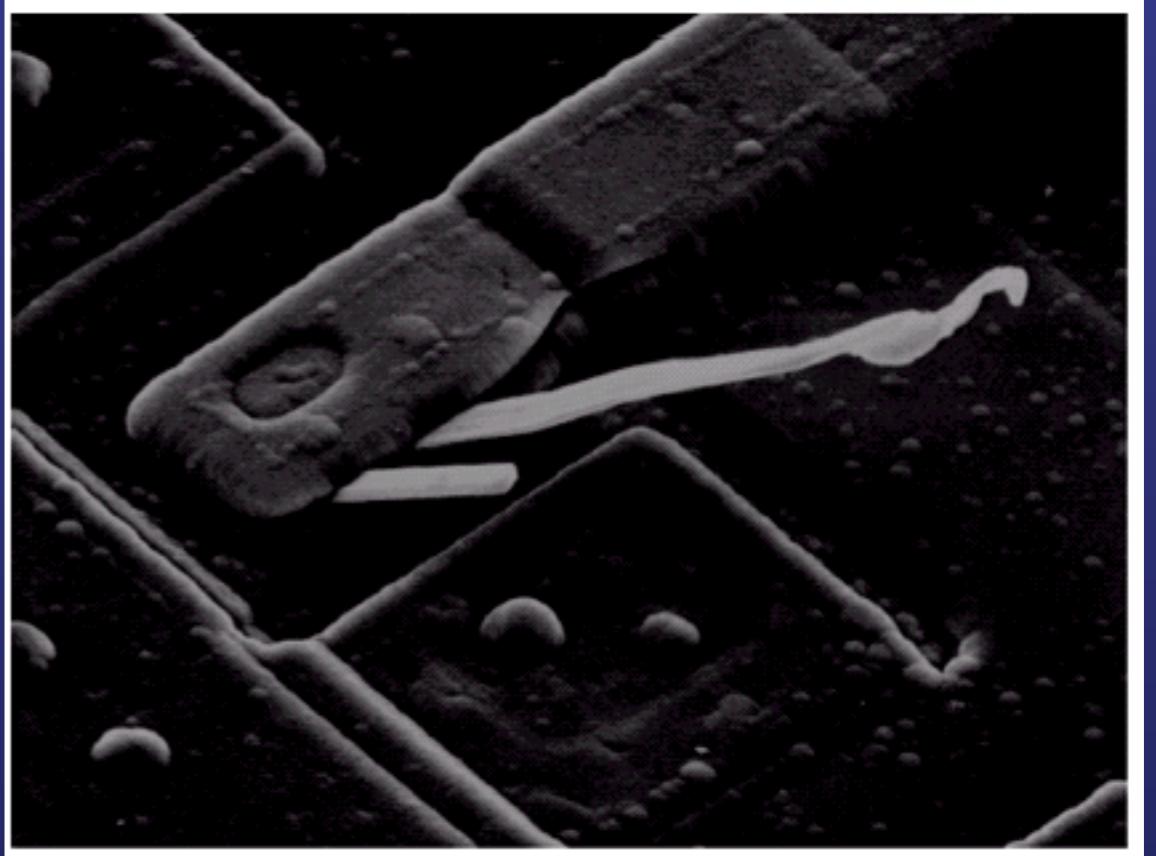
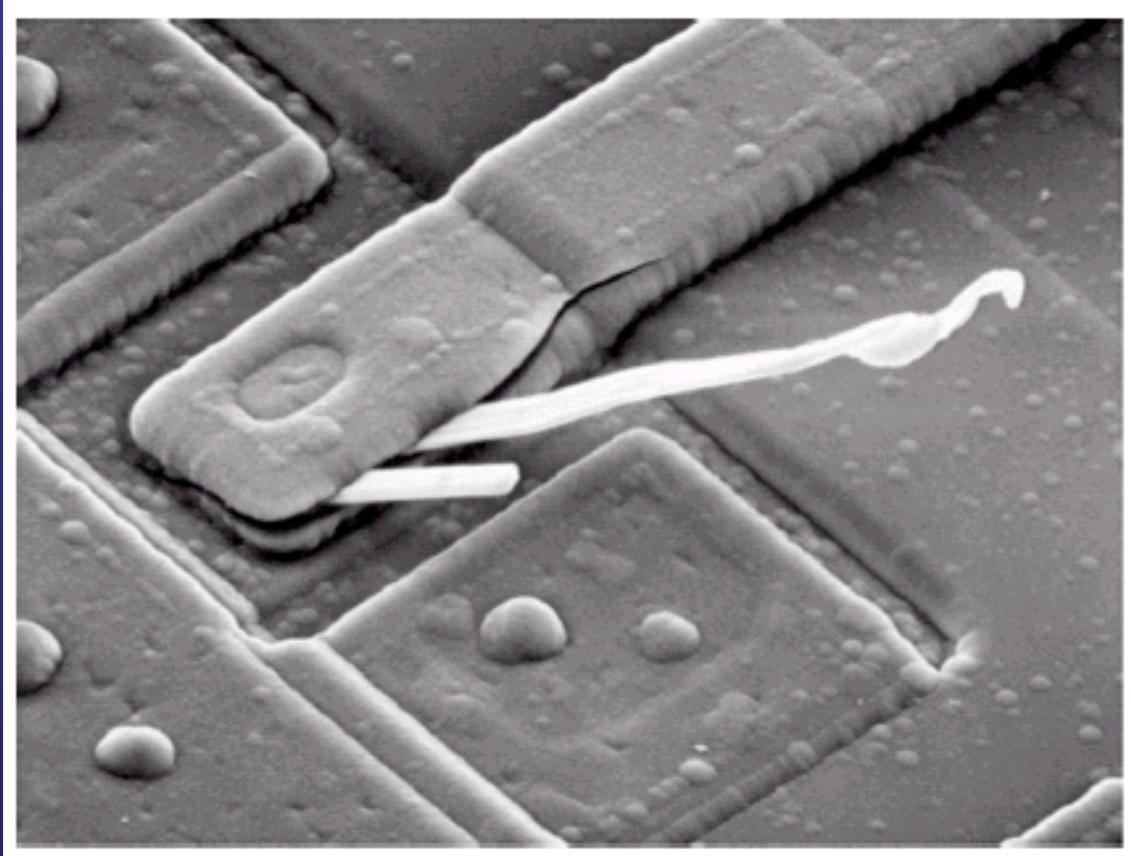
$$H(u,v) = \begin{cases} 0, & \text{if } (u,v) = (M/2, N/2) \\ 1, & \text{otherwise.} \end{cases}$$



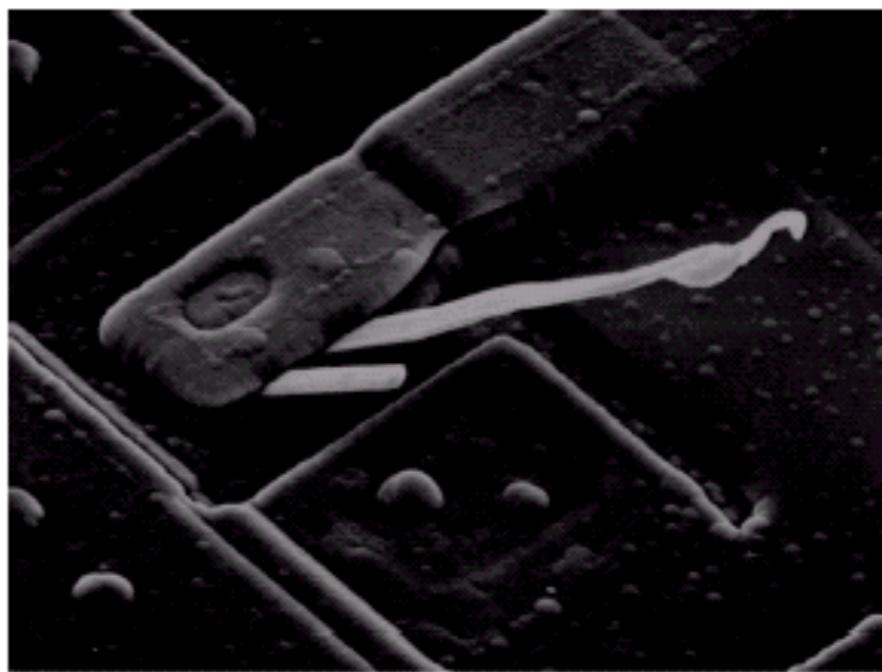
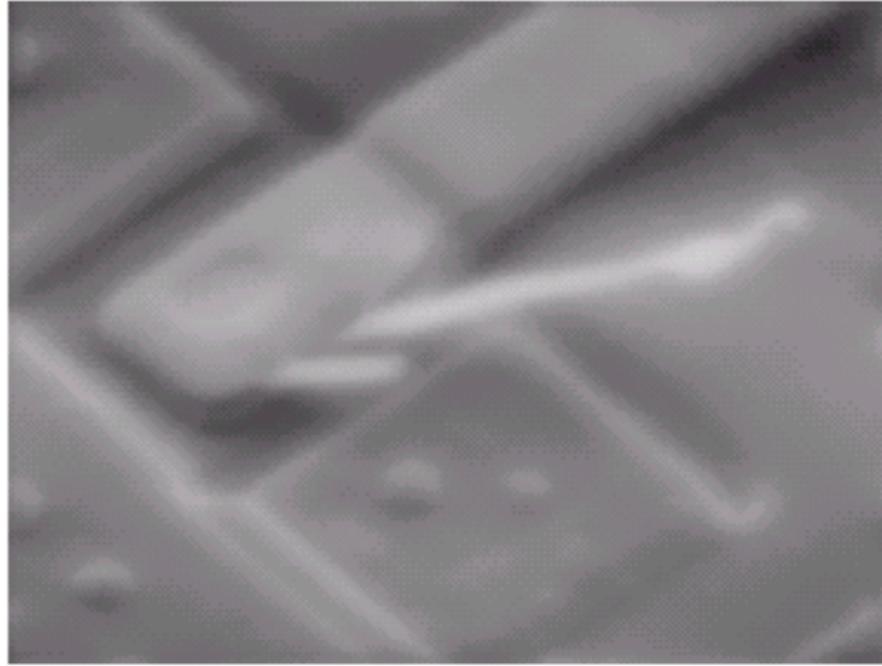
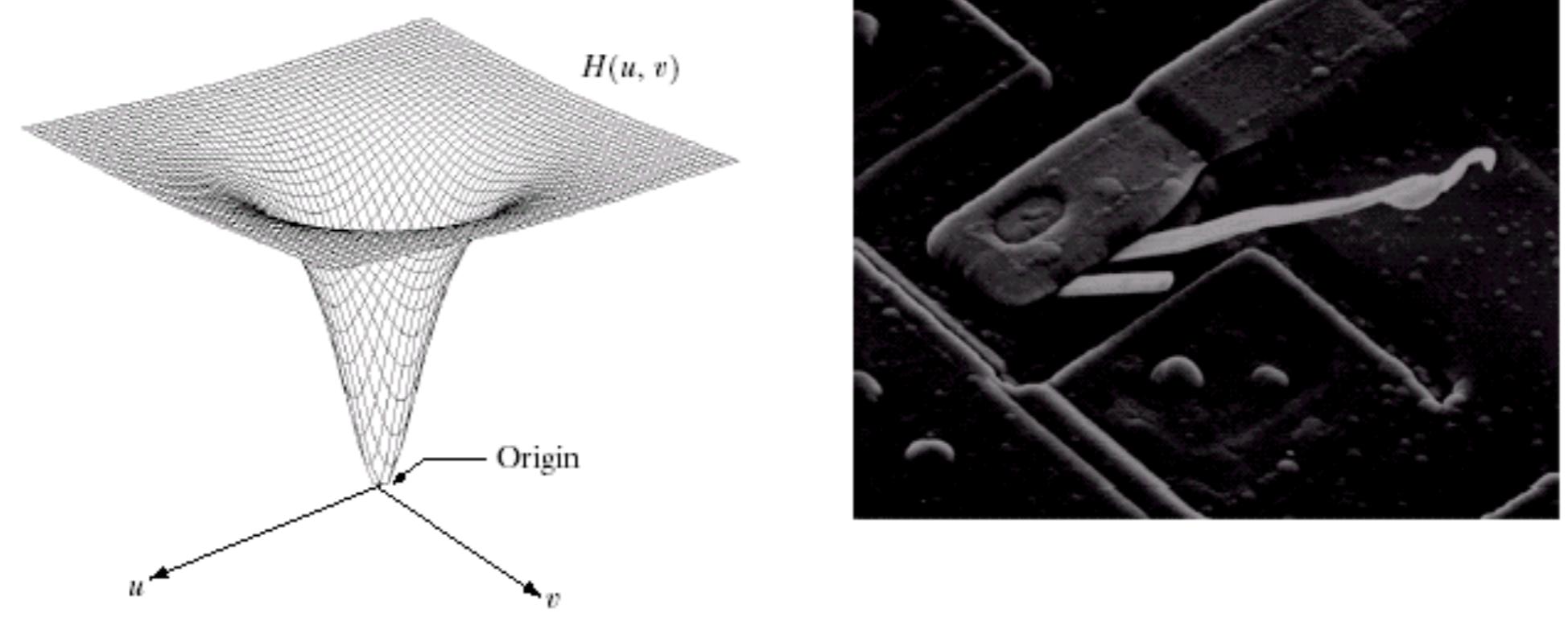
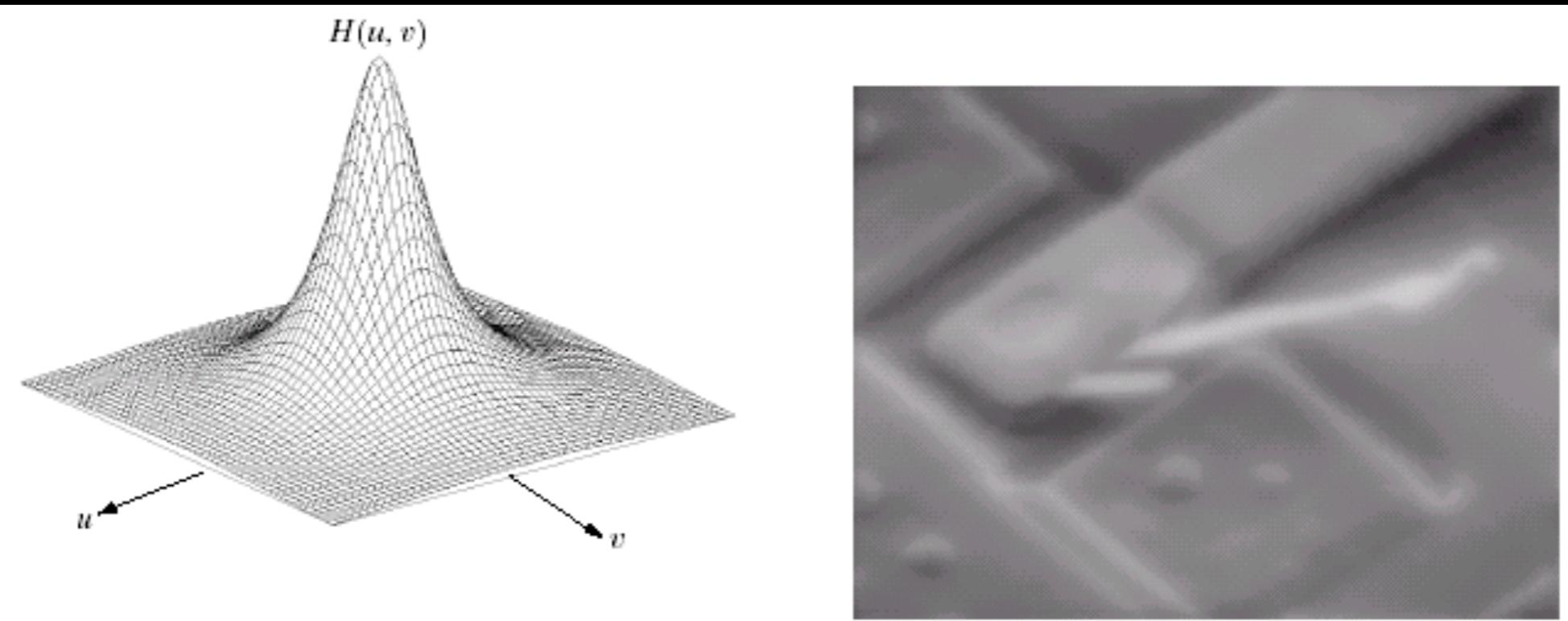
# Basic filter

---

$$H(u,v) = \begin{cases} 0, & \text{if } (u,v) = (M/2, N/2) \\ 1, & \text{otherwise.} \end{cases}$$



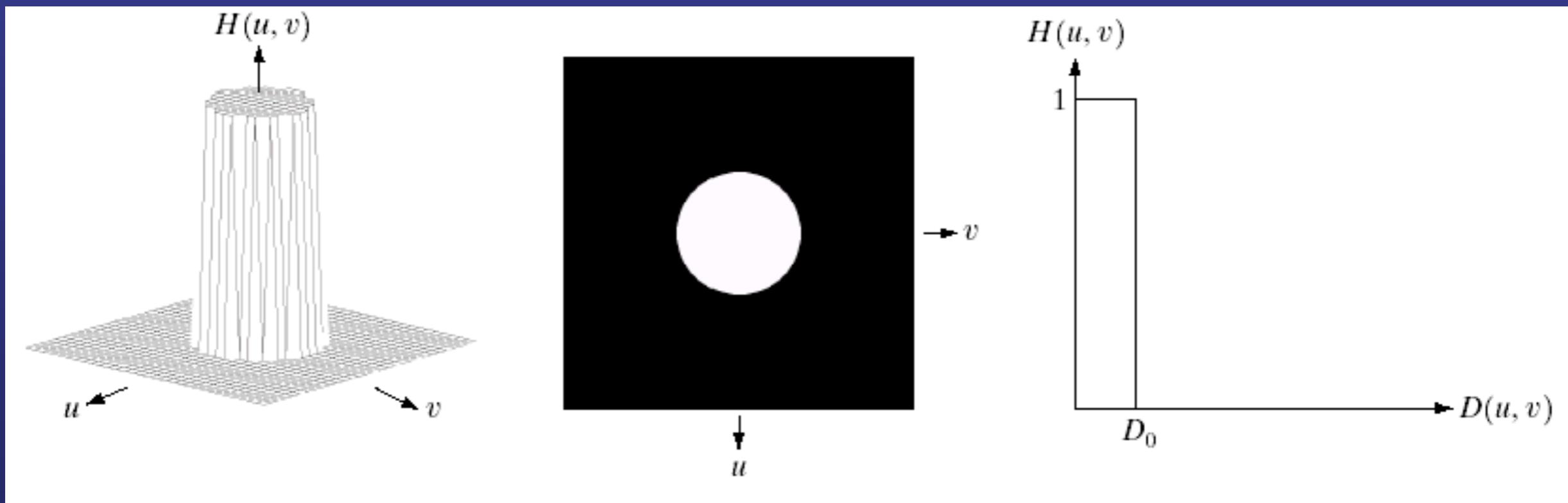
# Filtering



# Ideal Low-pass Filter

---

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

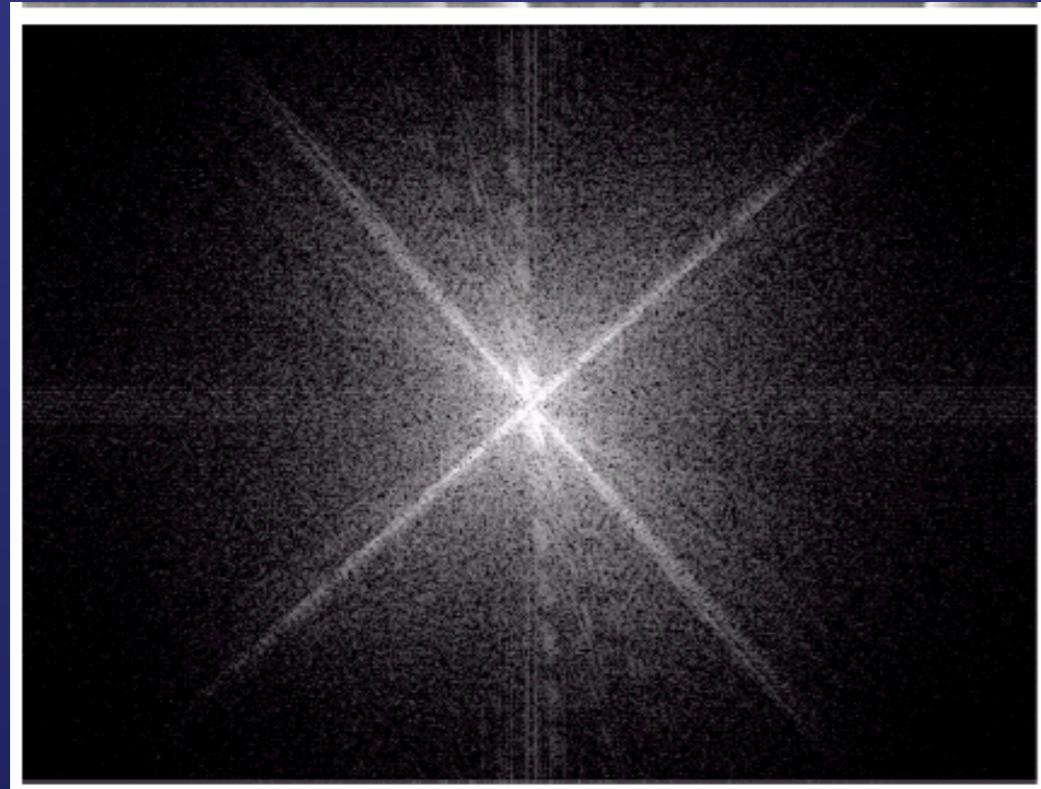
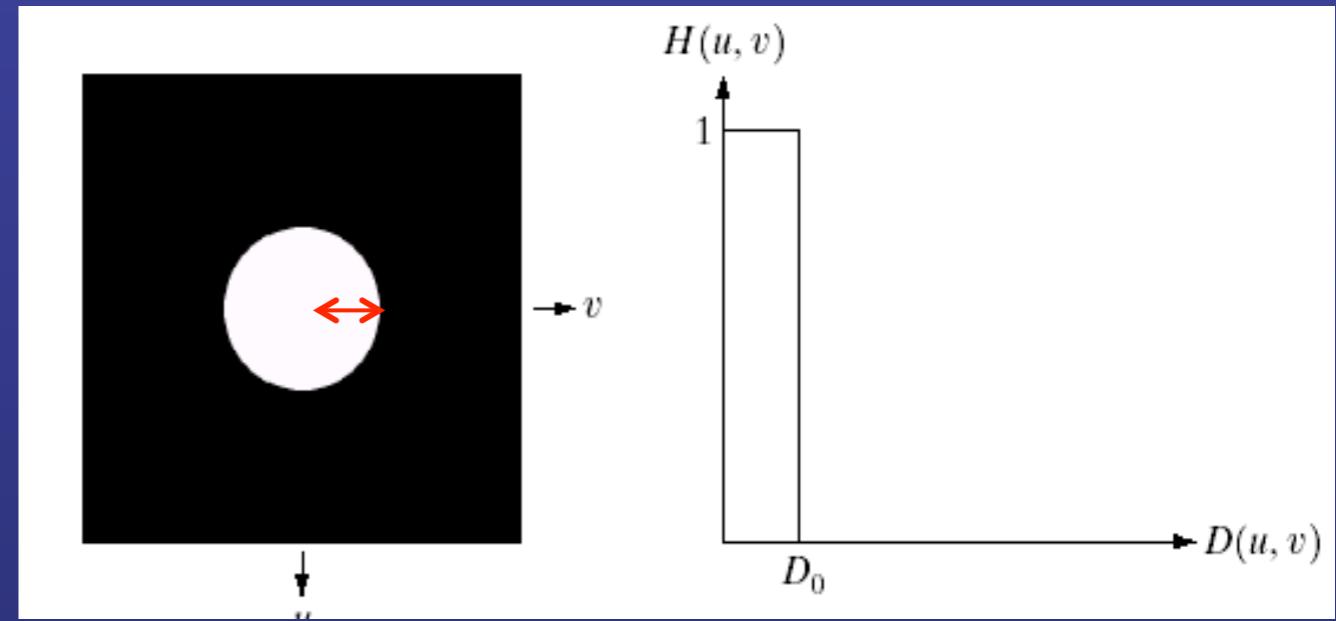


$$D(u,v) = [(u - M/2)^2 + (v - N/2)^2]^{1/2}$$

# Percentage of Power

$D_0 \rightarrow$  cutoff frequency

$$\alpha = 100 \left[ \sum_u \sum_v P(u, v) / P_T \right]$$

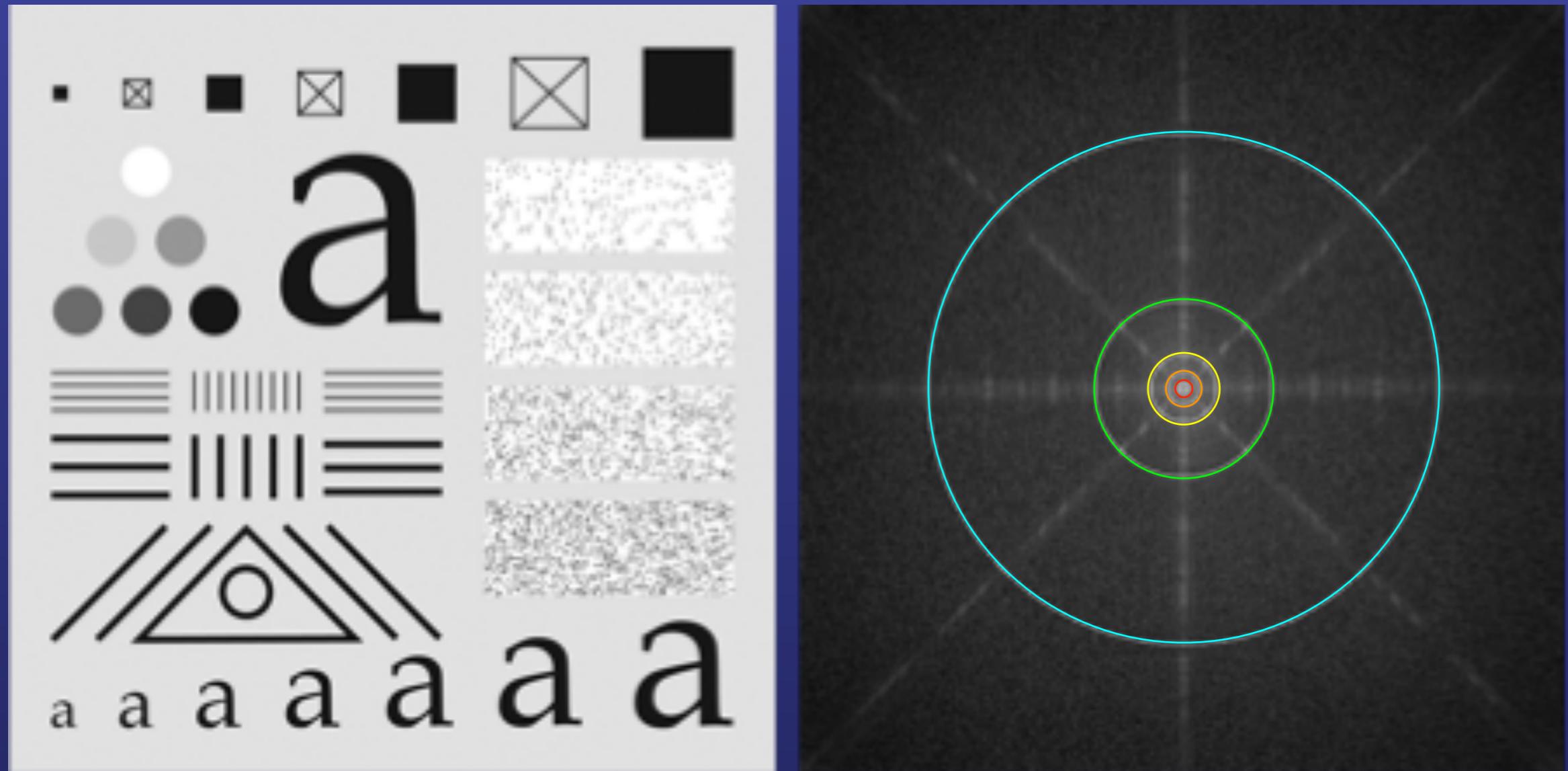


$$P_T = \sum_{u=0}^{P-1} \sum_{v=0}^{Q-1} P(u, v)$$

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

# Percentage of Power

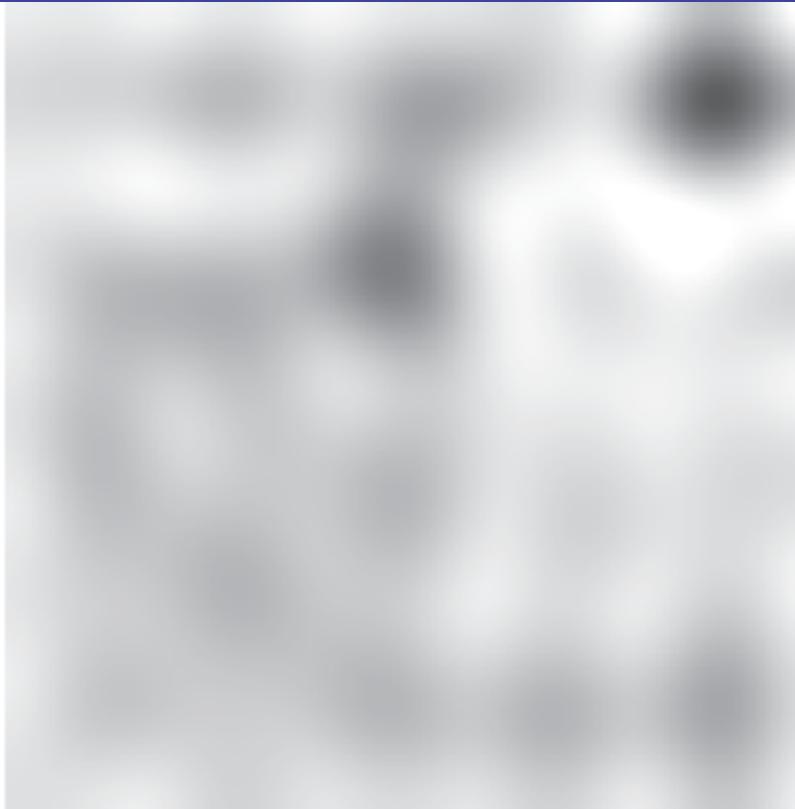
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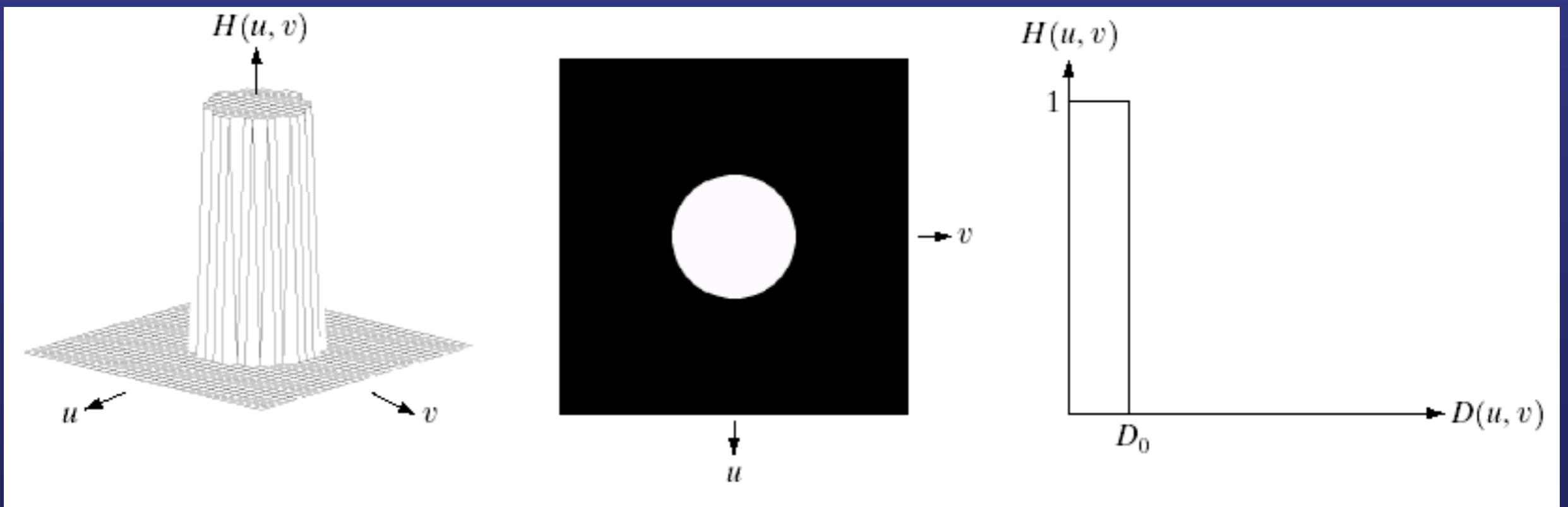
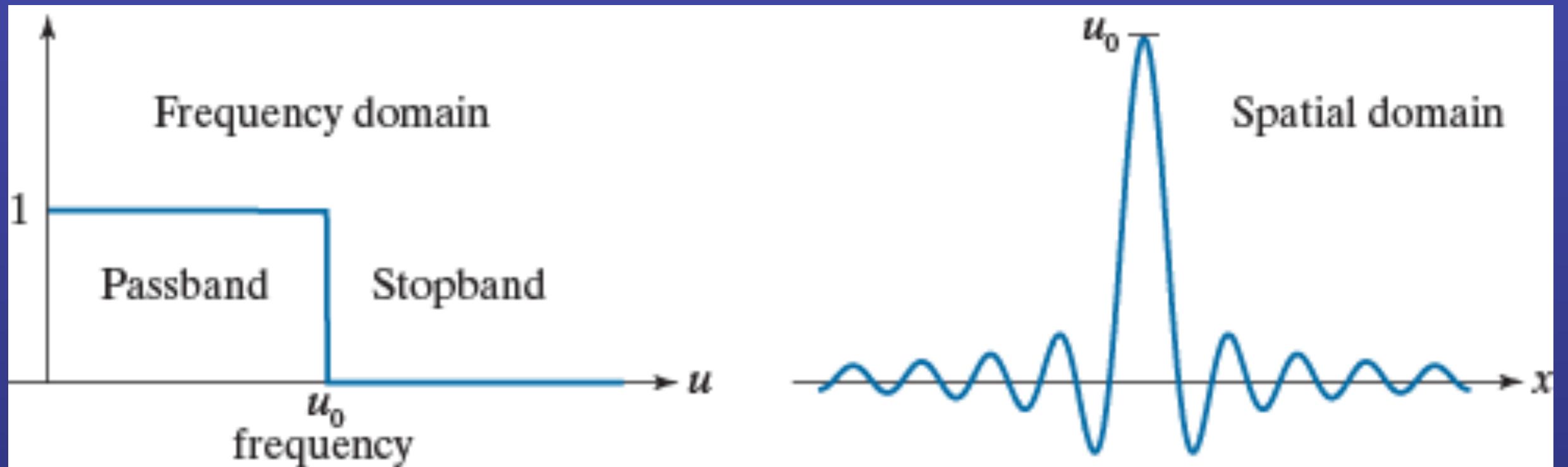


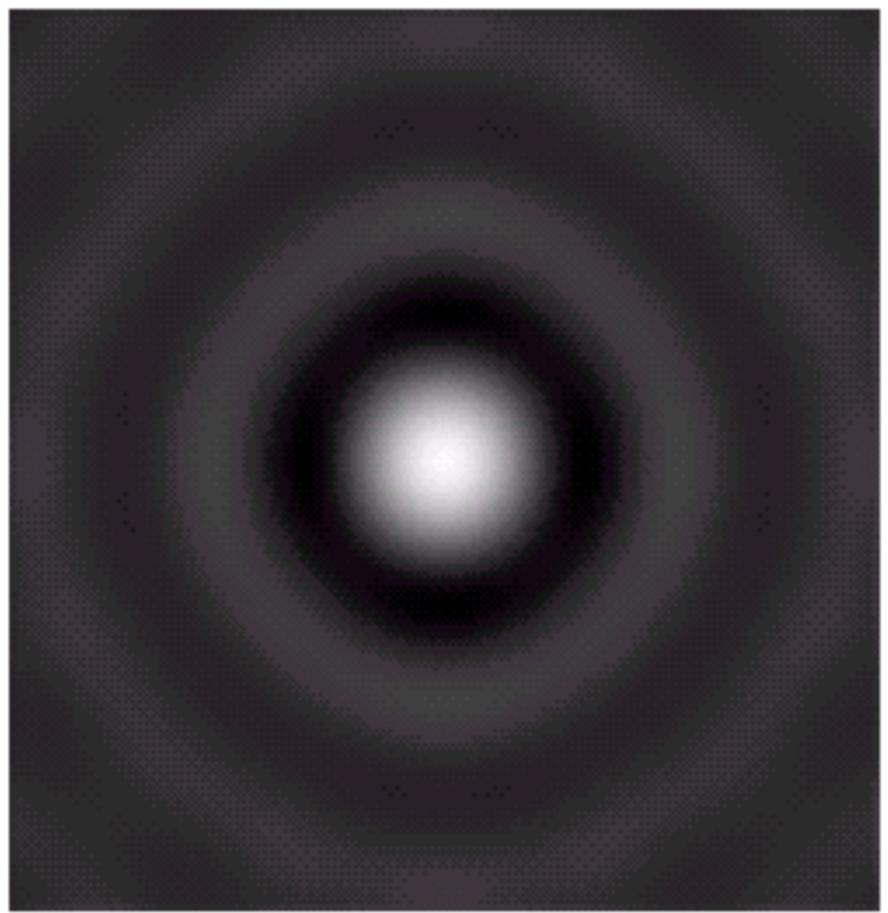
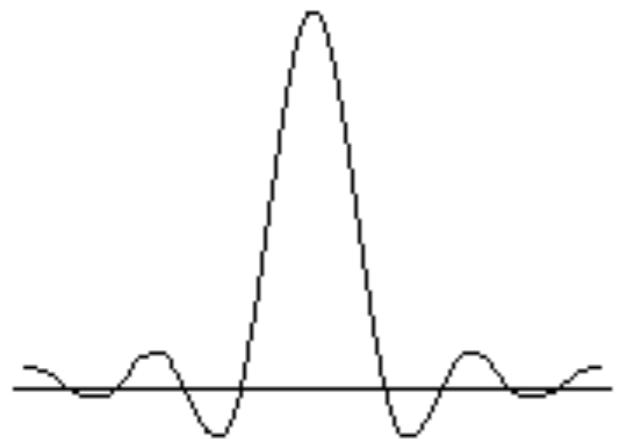
Radii: 10, 30, 60, 160, 460

Power: 86.9%, 92.8%, 95.1%, 97.6%, 99.4%

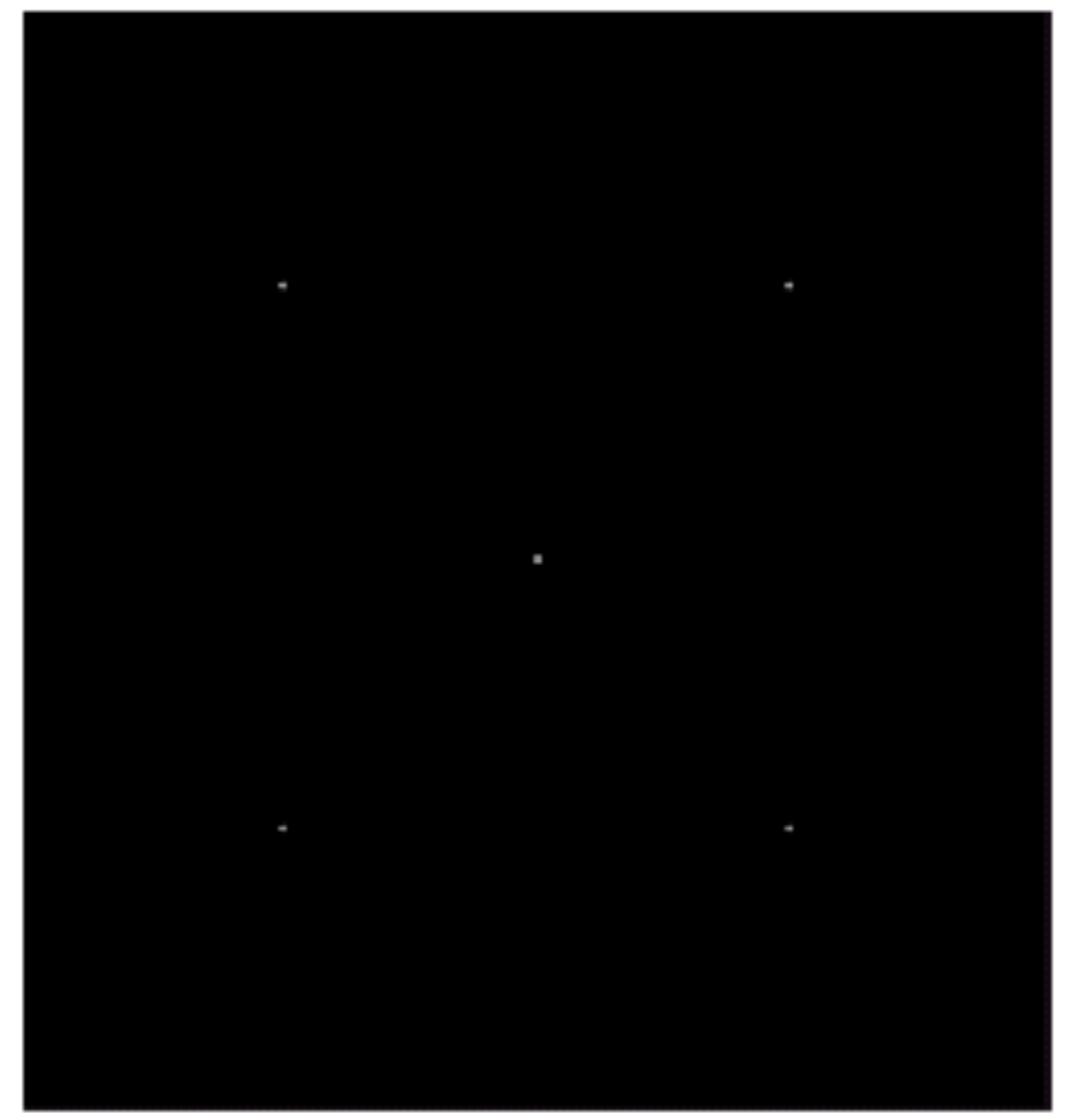
Notice the “ringing” in b-f

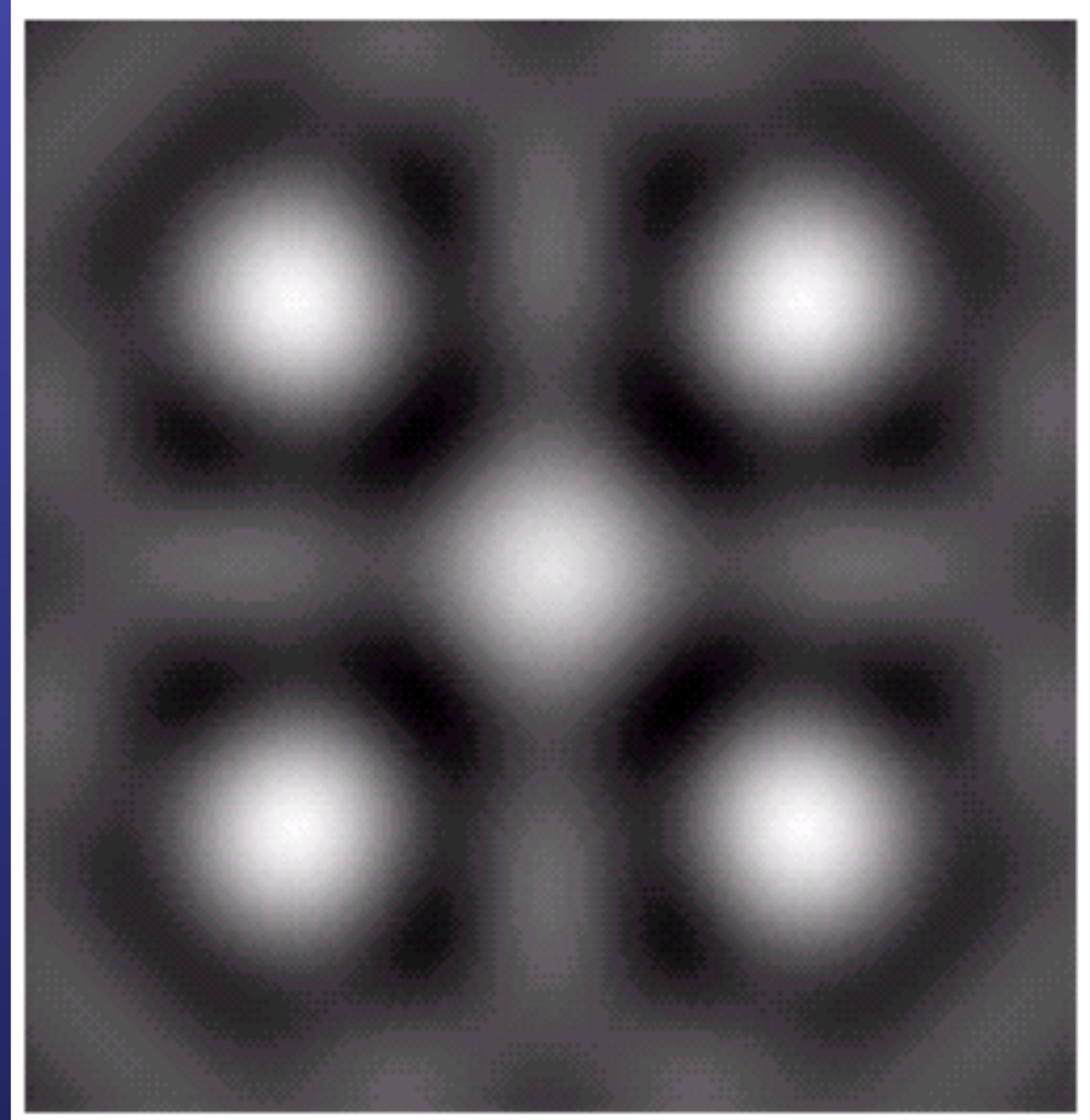
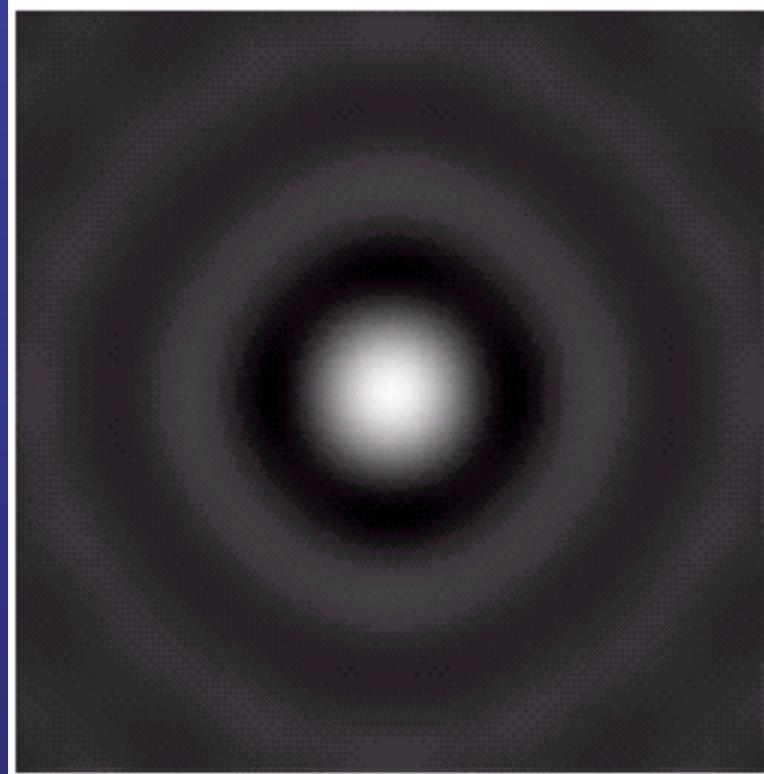
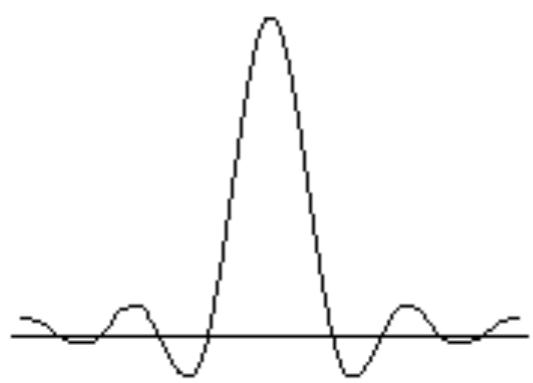






\*



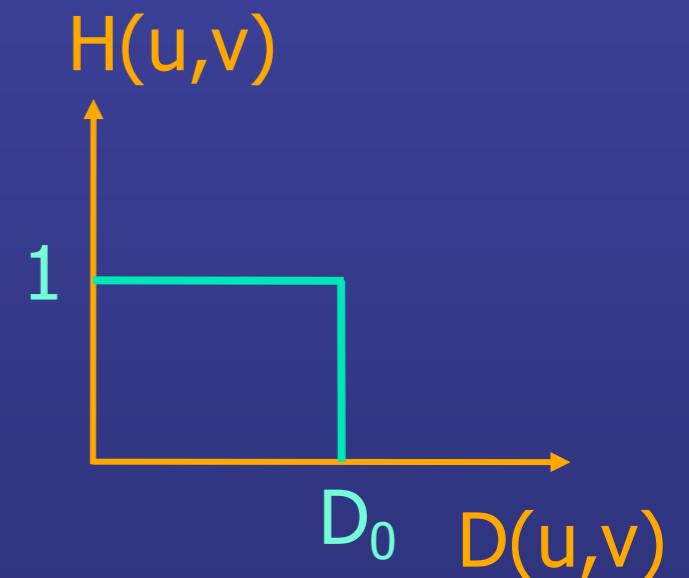


# Butterworth Lowpass filter

## ■ ILPF

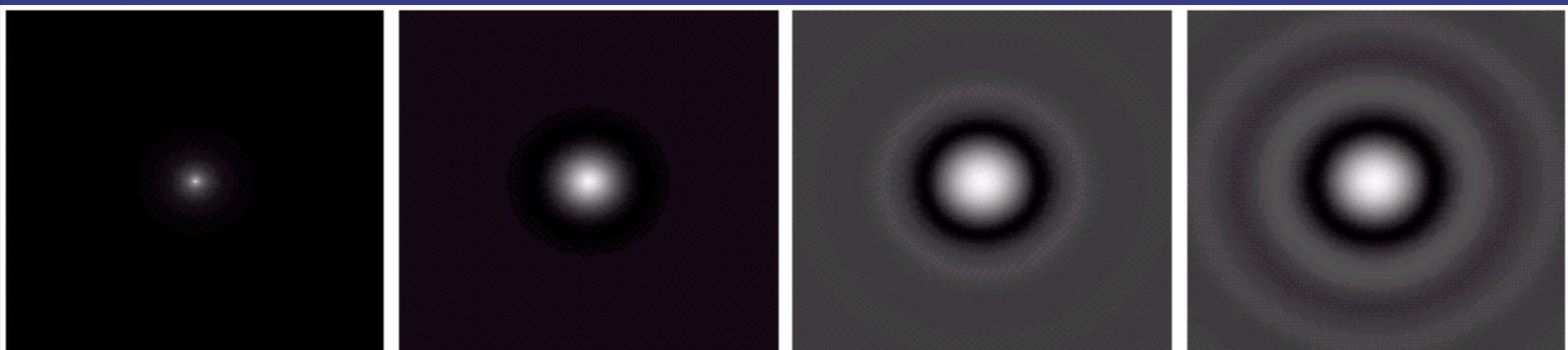
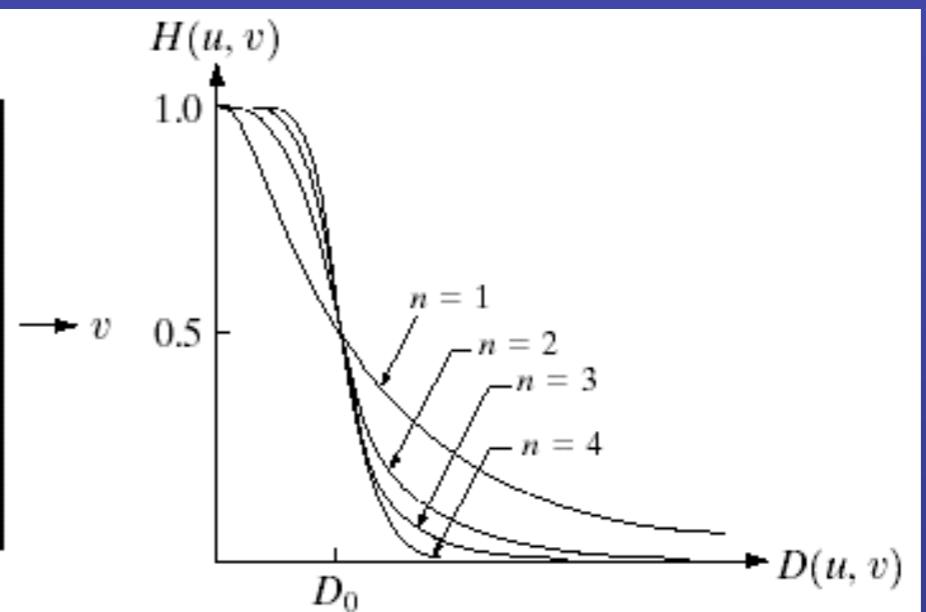
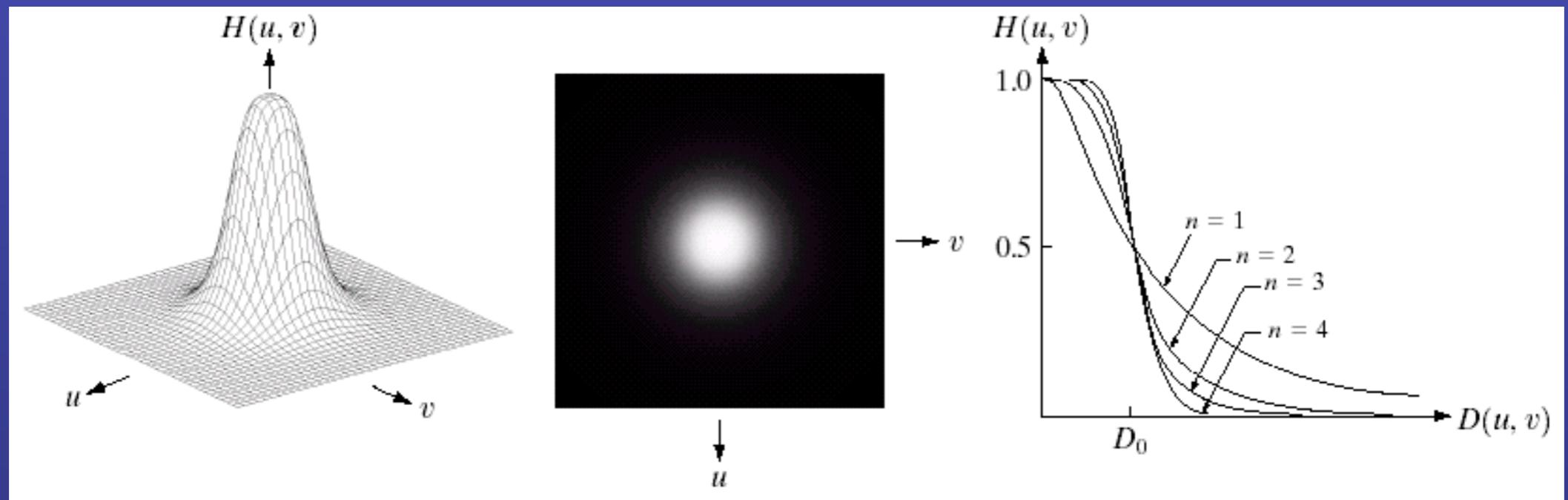
$$G(u,v) = H(u,v)F(u,v)$$

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



## ■ Butterworth

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



a | b | c | d

**FIGURE 4.16** (a)–(d) Spatial representation of BLPFs of order 1, 2, 5, and 20, and corresponding gray-level profiles through the center of the filters (all filters have a cutoff frequency of 5). Note that ringing increases as a function of filter order.

# Butterworth Lowpass filter

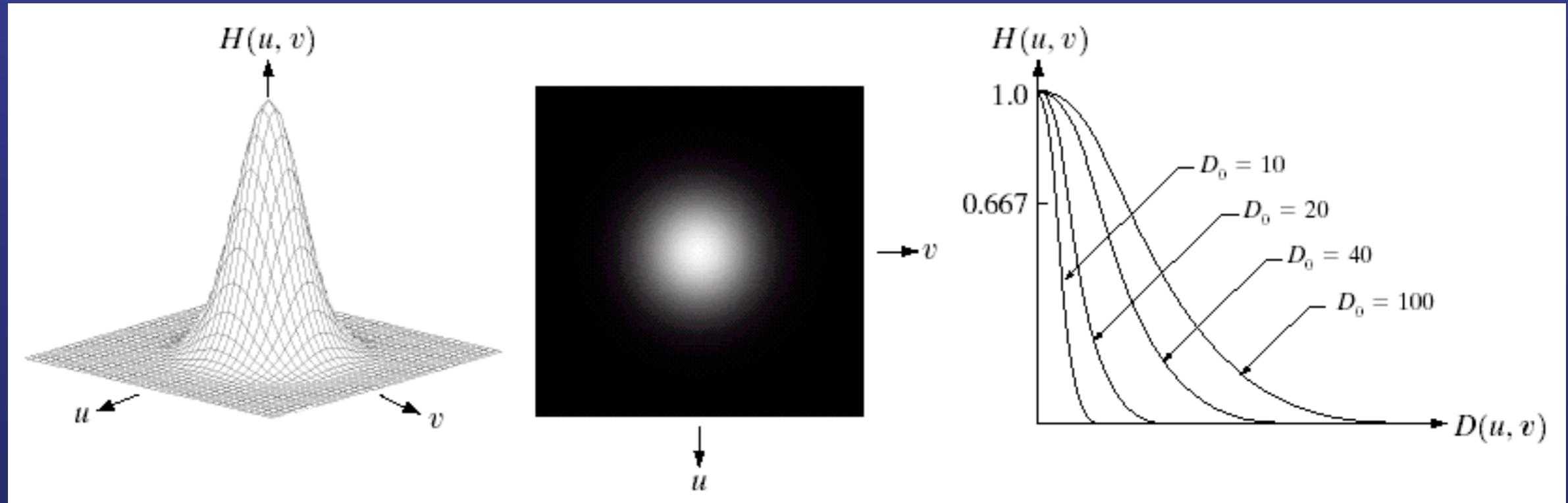
n=2.25



# Gaussian Lowpass Filter

---

$$H(u, v) = \exp(-D^2(u, v)/2D_0^2)$$



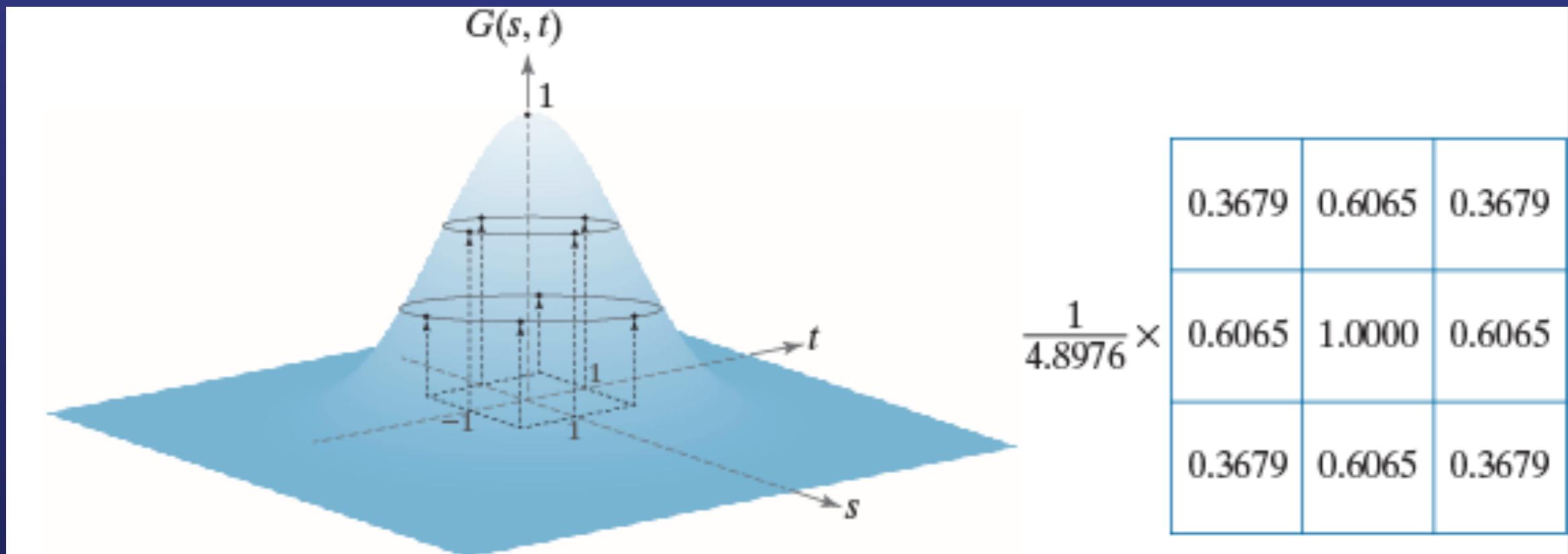
*Gaussian  $\xrightarrow{\mathcal{F}}$  Gaussian*

# Gaussian Lowpass Filter

$$H(u, v) = \exp(-D^2(u, v)/2D_o^2)$$



Gaussian kernel in spatial domain



# Gaussian Lowpass Filter

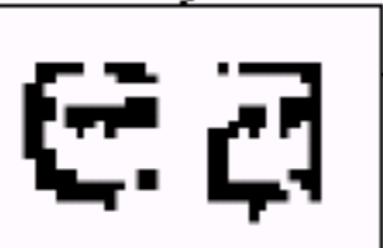
- No ringing
- Good for medical imaging



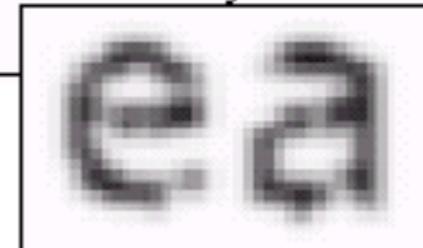
# Another examples

---

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.



# Another examples

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# High Pass Filters

---

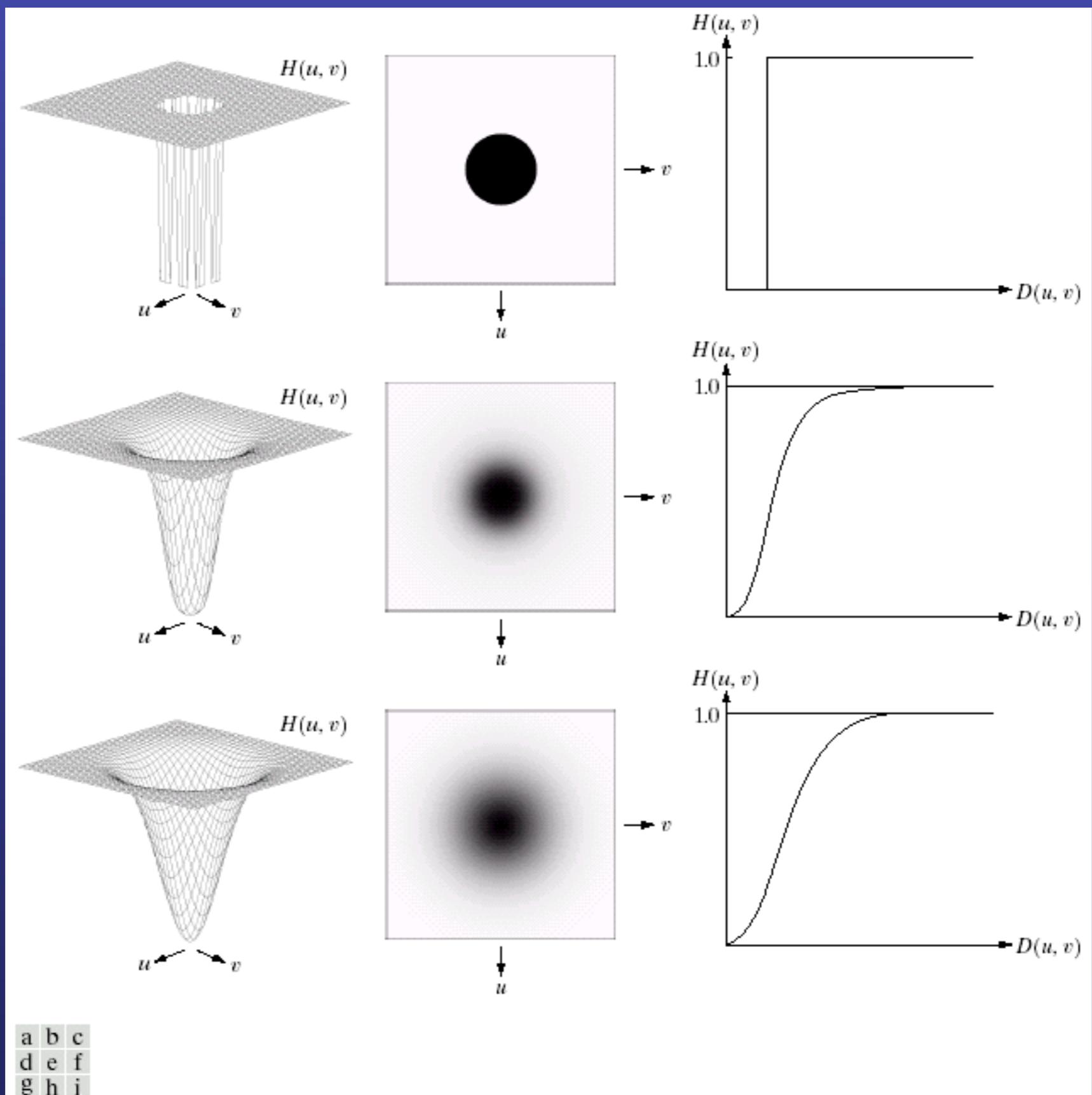
$$H_{hp}(u,v) = 1 - H_{lp}(u,v)$$

## ■ IHPF

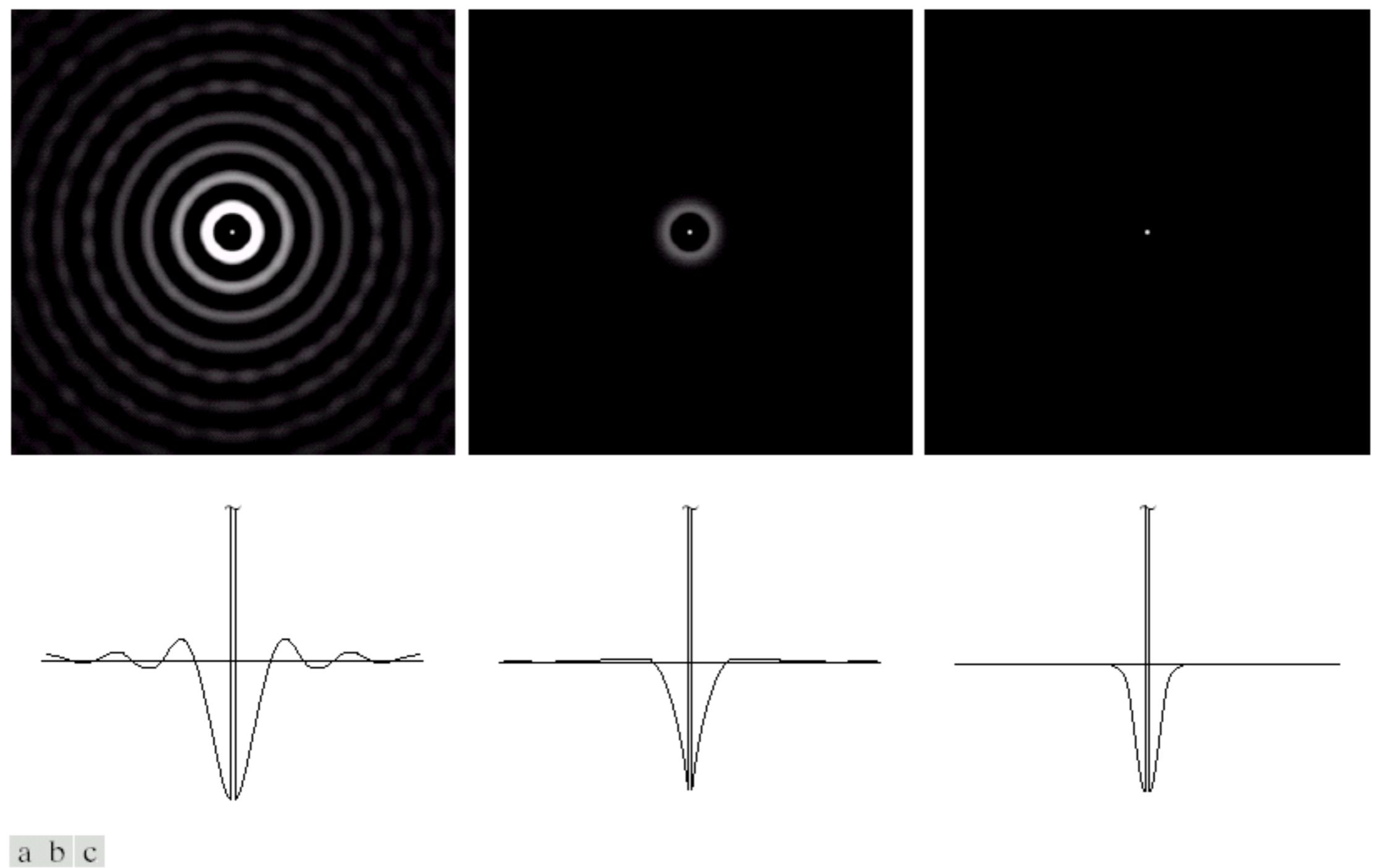
$$H(u,v) = \begin{cases} 0, & \text{if } D(u,v) \leq D_o \\ 1, & \text{if } D(u,v) > D_o \end{cases}$$

## ■ Butterworth Filter

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

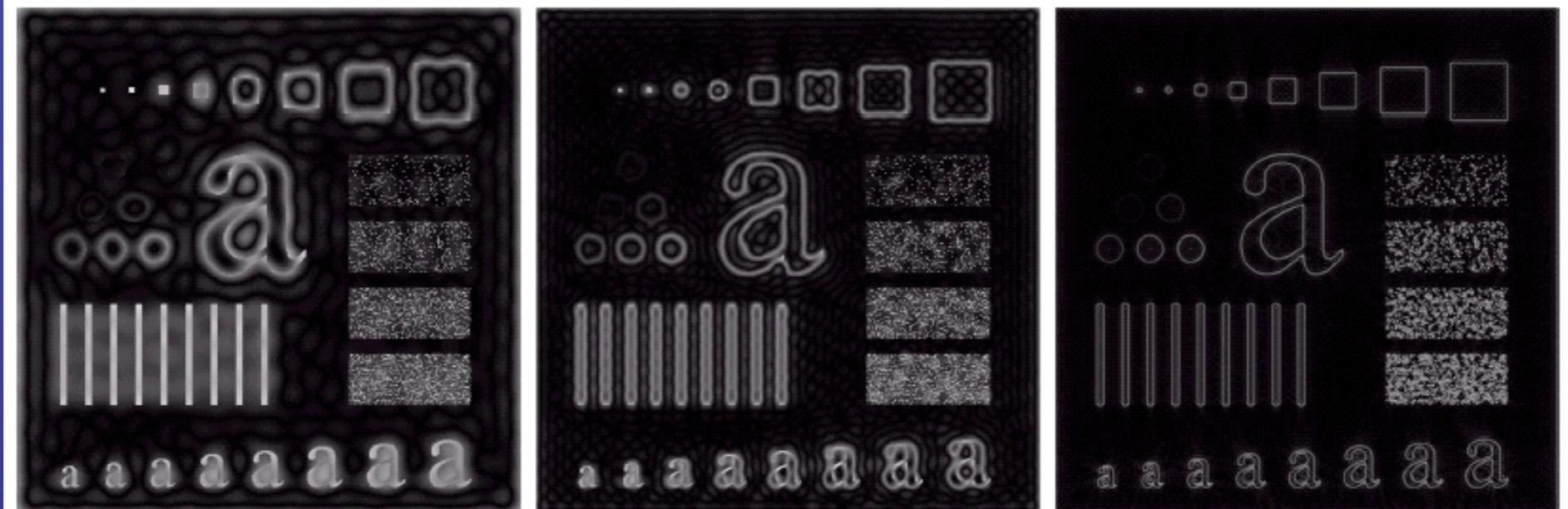


**FIGURE 4.22** Top row: Perspective plot, image representation, and cross section of a typical ideal highpass filter. Middle and bottom rows: The same sequence for typical Butterworth and Gaussian highpass filters.

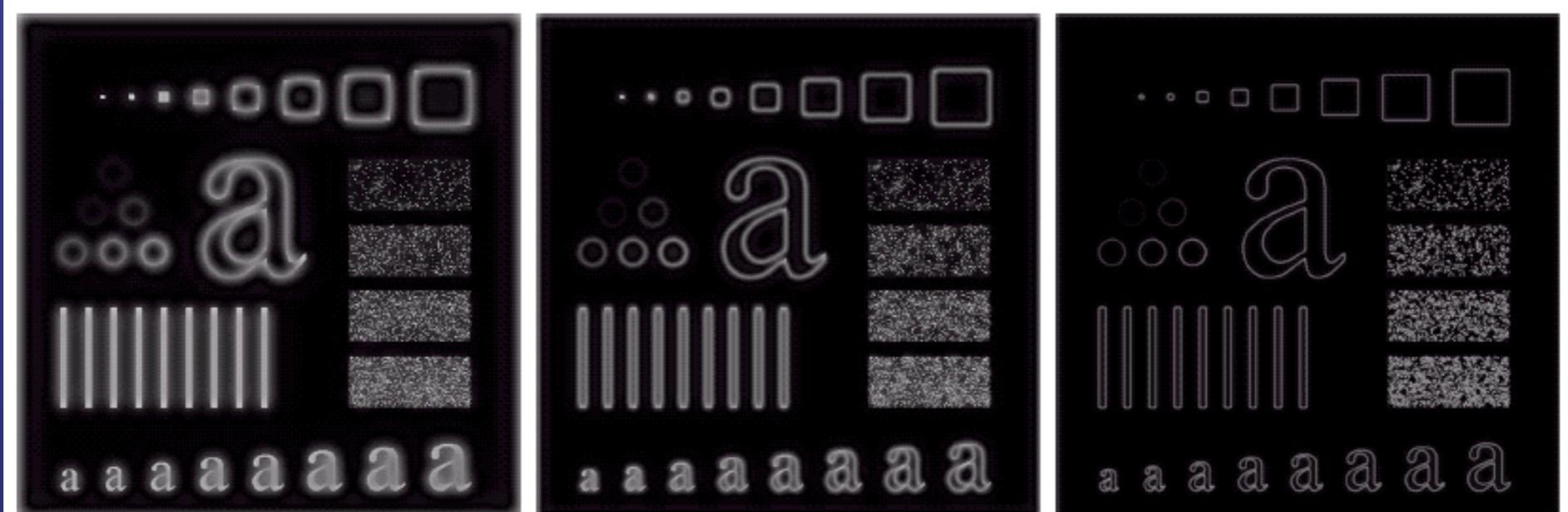


**FIGURE 4.23** Spatial representations of typical (a) ideal, (b) Butterworth, and (c) Gaussian frequency domain highpass filters, and corresponding gray-level profiles.

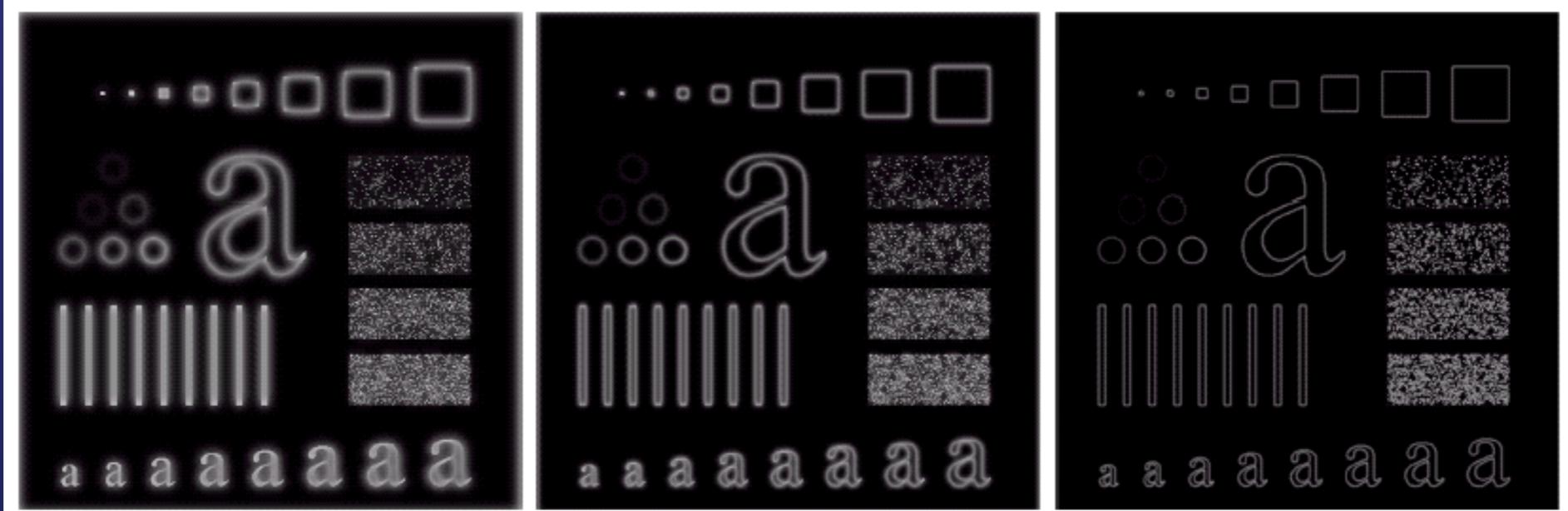
IHPF



BHPF



GHPF



# Homomorphic Filtering

---

$$f(x, y) = i(x, y)r(x, y)$$

$$0 < i(x, y) < \infty \quad \text{illumination}$$

$$0 < r(x, y) < 1 \quad \text{reflectance}$$

$$F\{f(x, y)\} \neq F\{i(x, y)\}F\{r(x, y)\}$$

# Homomorphic Filtering

---

$$\begin{aligned} \text{let } z(x,y) &= \ln f(x,y) \\ &= \ln i(x,y) + \ln r(x,y) \end{aligned}$$

$$\therefore Z(u,v) = I_{\ln} + R_{\ln}$$

$$S(u,v) = H(u,v)Z(u,v)$$

$$= HI_{\ln} + HR_{\ln}$$

$$\begin{aligned} \therefore s(x,y) &= F^{-1}\{HI_{\ln}\} + F^{-1}\{HR_{\ln}\} \\ &= i'(x,y) + r'(x,y) \end{aligned}$$

# Homomorphic Filtering

---

$$g(x,y) = e^{s(x,y)} = e^{i'(x,y)} \times e^{r'(x,y)}$$
$$= i_o(x,y)r_o(x,y)$$

$$i_o(x,y) = e^{i'(x,y)}$$

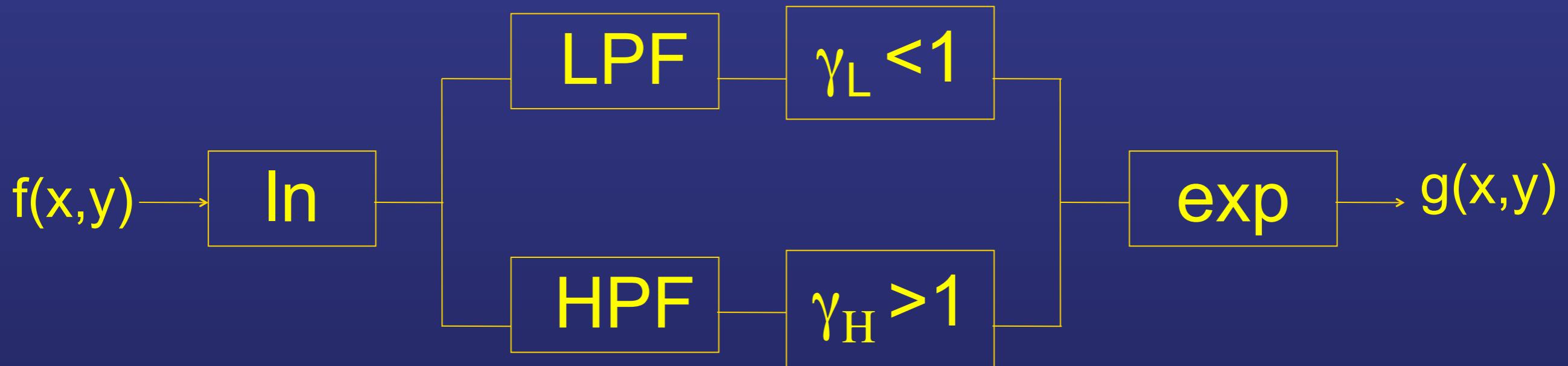
$$i'(x,y) \rightarrow F^{-1}\{H(u,v)I_{ln}(u,v)\}$$

$$I_{ln}(u,v) \rightarrow F\{\ln i(x,y)\}$$

# Homomorphic Filtering

---

- In practice:  $i \rightarrow$  slow varying  $\rightarrow$  LPF  
 $r \rightarrow$  fast varying  $\rightarrow$  HPF



# Homomorphic Filtering

