

Chapter 3. Image Transforms

- Fourier Transform of 1 and 2 continuous
- Discrete Fourier Transform
- Properties of 2-D Fourier Transform
- FFT
- Other Transforms

Transform theory: key role in image processing

→ 2-D transforms to image enhancement, restoration, encoding, ...

3.1 Introduction to the Fourier Transform

- Notation of FT pair:

$$F\{f(x)\} = F(u) = \int_{-\infty}^{\infty} f(x) e^{-j2\pi xu} dx$$

- inverse FT

$$F^{-1}\{F(u)\} = f(x) = \int_{-\infty}^{\infty} F(u) e^{-j2\pi ux} du$$

if $f(x)$ is continuous and integrable and $F(u)$ is integrable

- $f(x) : \text{real} \xrightarrow{FT} F(u) : \text{complex}$

$$F(u) = R(u) + jI(u)$$

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

where $|F(u)| = [R^2(u) + I^2(u)]^{1/2} : \text{Fourier spectrum (magnitude)}$

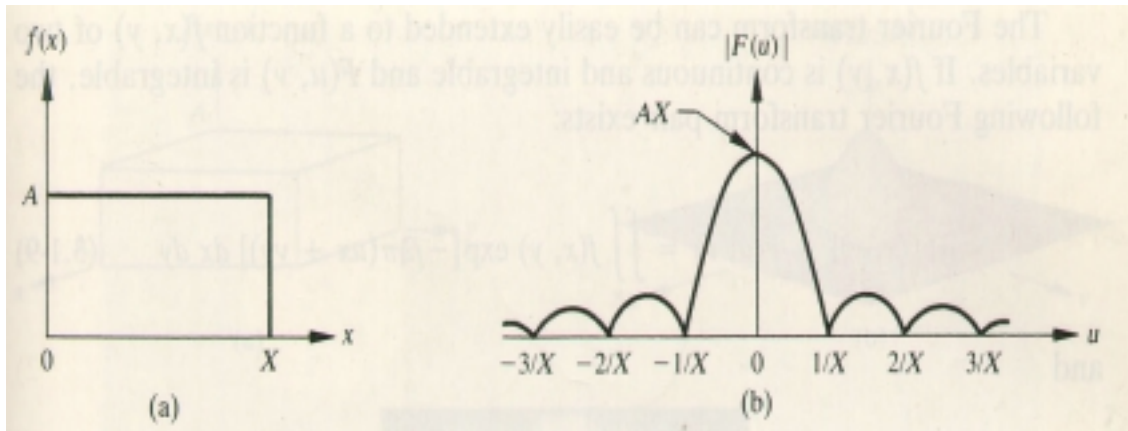
$$\phi(u) = \tan^{-1} \frac{I(u)}{R(u)} \quad : \text{phase}$$

- square of spectrum :

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

$$= R^2(u) + I^2(u) \quad : \text{power spectrum of } f(x) \quad (\text{spectral density})$$

- example:



- 2-D FT pair:
- extension of 1-D FT;

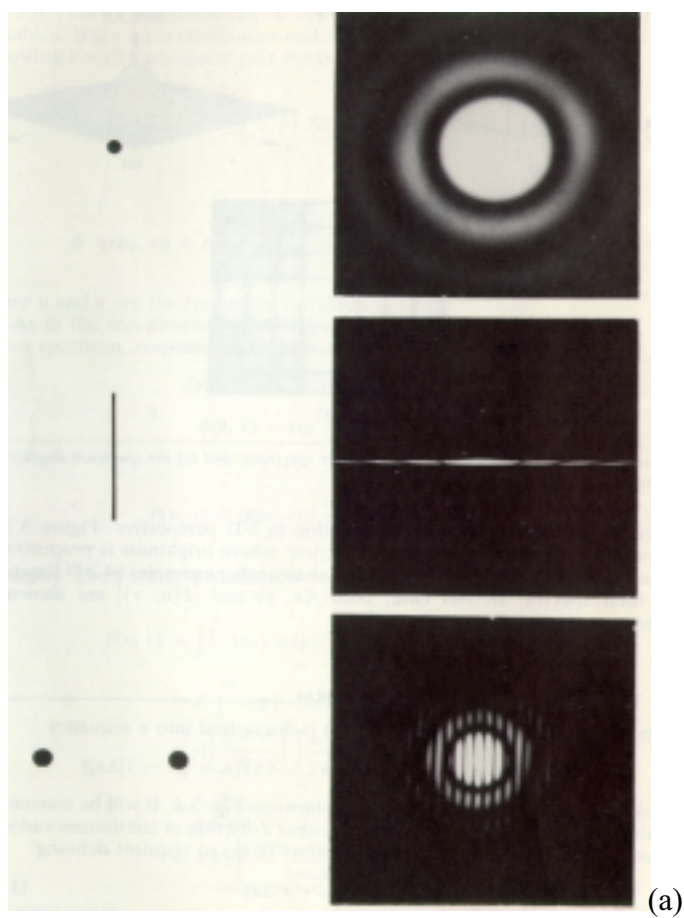
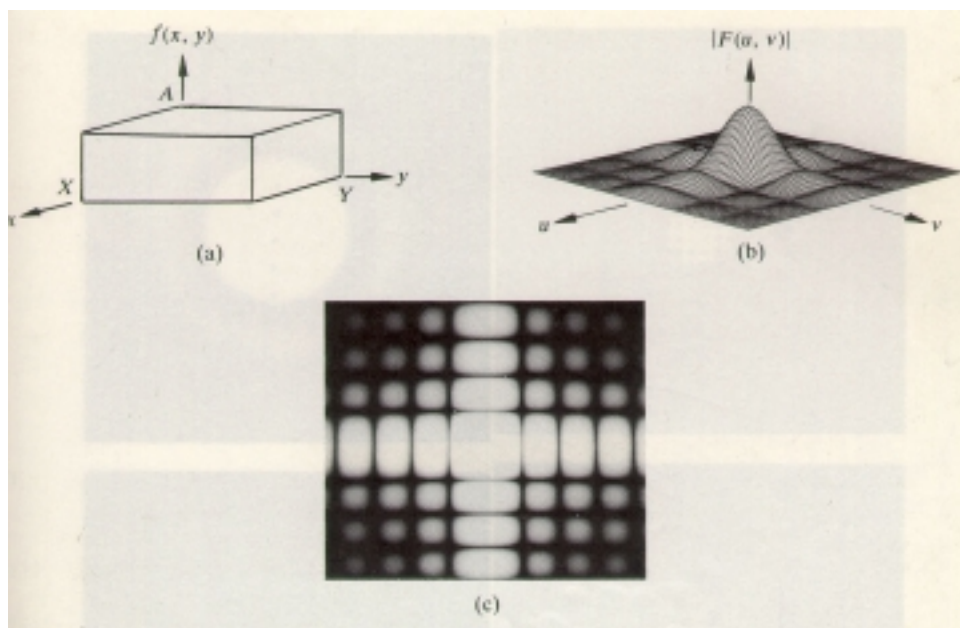
$$F\{f(x, y)\} = F(u, v) = \iint_{-\infty}^{\infty} f(x, y) e^{-j2\pi(ux+vy)} dx dy$$

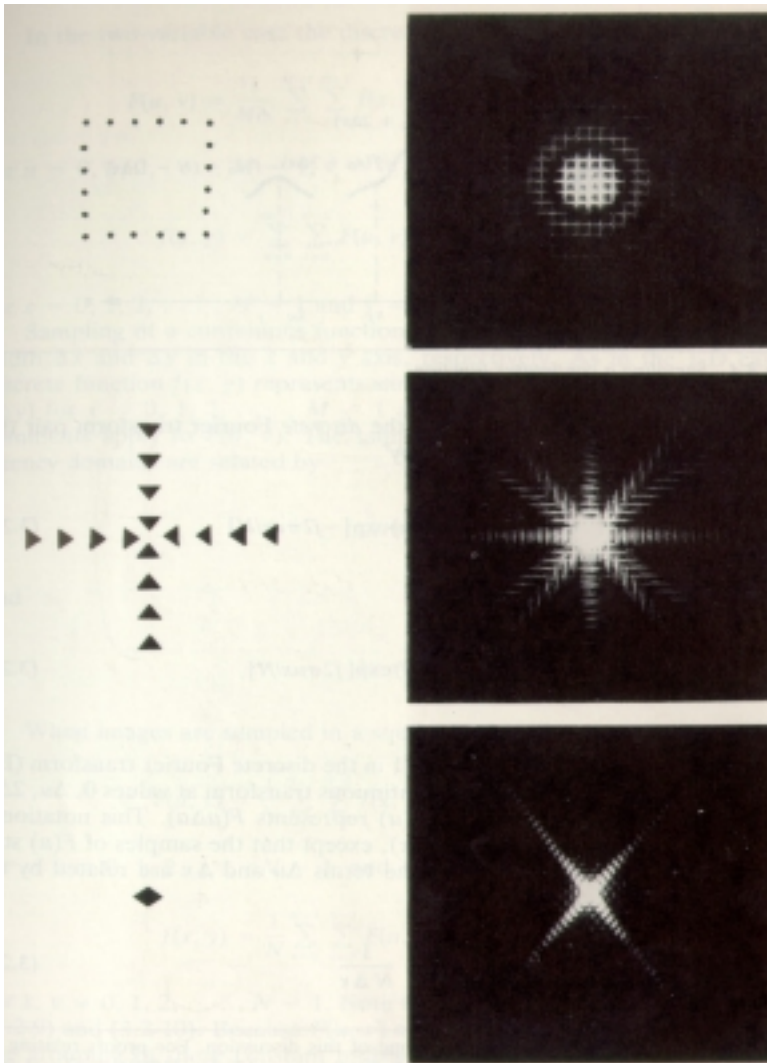
$$F^{-1}\{F(u, v)\} = f(x, y) = \iint_{-\infty}^{\infty} F(u, v) e^{+j2\pi(ux+vy)} du dv$$

- spectrum, phase and power spectrum:

- $|F(u, v)| = [R^2(u, v) + I^2(u, v)]^{1/2}$; Fourier Spectrum
- $\phi(u, v) = \tan^{-1} \frac{I(u, v)}{R(u, v)}$; Phase Spectrum
- $P(u, v) = |F(u, v)|^2 = R^2(u, v) + I^2(u, v)$; Power Spectrum

Example:





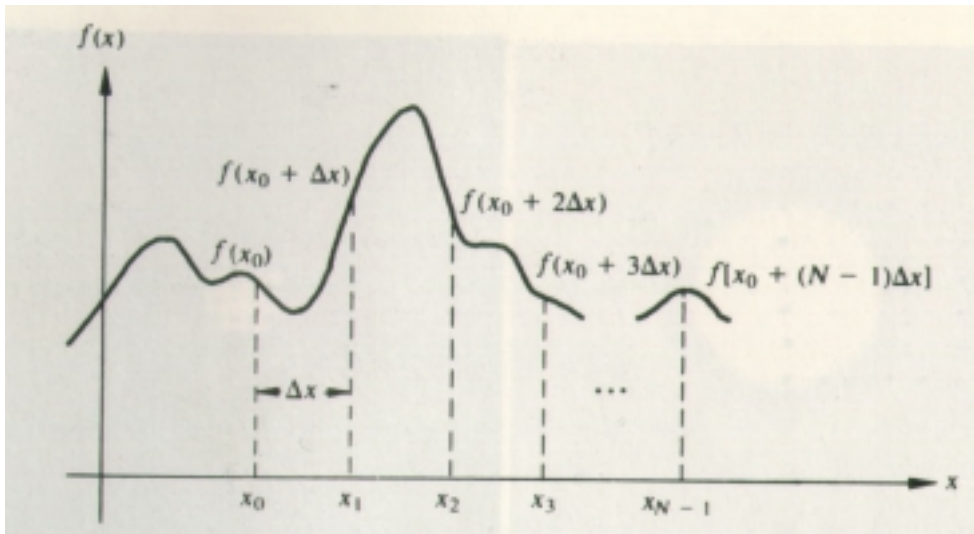
3.2 The Discrete Fourier Transform (DFT)

- Sampling:

$$f(x) \rightarrow \{f(x_0), f(x_0 + \Delta x), \dots, f(x_0 + (N + 1)\Delta x)\}$$

uniformly spaced samples

$$f(0), f(1), \dots, f(N - 1)$$



- DFT pair:

$$F(u) = \frac{1}{N} \sum_{x=0}^{N-1} f(x) \exp(-j2\pi ux / N)$$

$$\text{for } u = 0, 1, \dots, N-1$$

$$f(x) = \sum_{u=0}^{N-1} F(u) \exp(j2\pi ux / N)$$

$$\text{for } x = 0, 1, \dots, N-1$$

- values $u=0, 1, \dots, N-1$

: correspond to samples of continuous transform at values $0, \Delta u, 2\Delta u, \dots, (N-1)\Delta u$

$$F(u) \rightarrow F(u\Delta u)$$

$$\Delta u = \frac{1}{N\Delta x}$$

- 2-D DFT:

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

$$\text{for } u = 0, 1, \dots, M-1 \quad v = 0, 1, \dots, N-1$$

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

$$\text{for } u = 0, 1, \dots, N-1 \quad v = 0, 1, \dots, N-1$$

- sampling : 2-D grid

$$\Delta u = \frac{1}{M\Delta x}, \quad \Delta v = \frac{1}{N\Delta y}$$

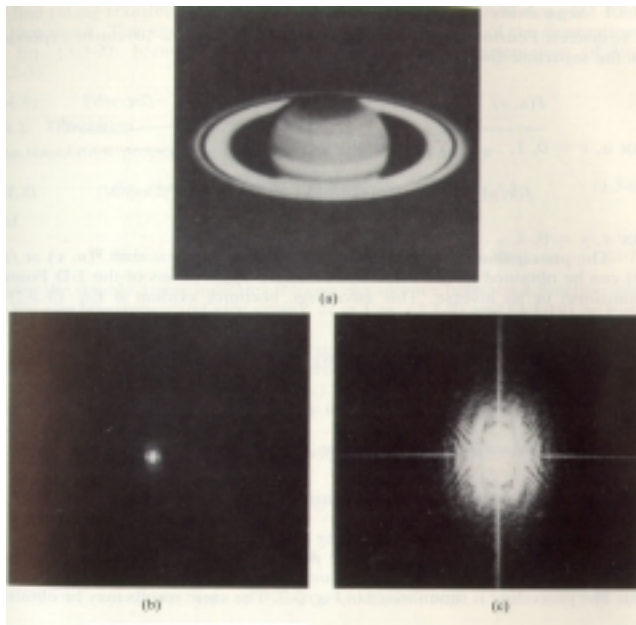
- for square array, $M=N$;

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

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

3.3 Some Properties of the 2-D FT

- dynamic range of Fourier spectra :
 - much higher than the typical display device
 - range : $[0, 2.5 \times 10^6]$



- only the brightest parts are visible on the display screen
- log compression ;

$$D(u, v) = c \log[1 + |F(u, v)|]$$

✓

Scaling constant

3.3.1 Separability

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

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

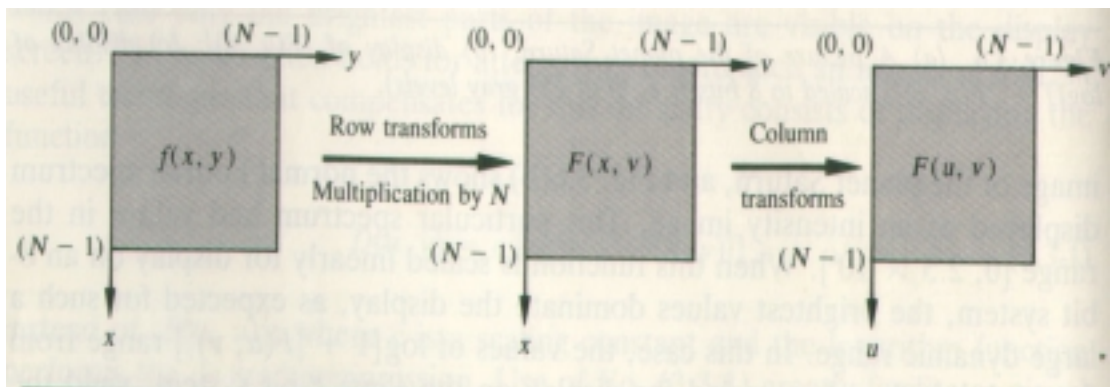
: obtained in two steps by successive applications of 1-D FT

or

$$F(u, v) = \frac{1}{N} \sum_{x=0}^{N-1} F(x, v) \exp(-j2\pi ux / N)$$

where
$$F(x, v) = N \left[\frac{1}{N} \sum_{y=0}^{N-1} f(x, y) \exp(-j2\pi vy / N) \right]$$

: 1-D transform



- 2-D FT :
 - i. 1-D transform along each row of $f(x,y)$
 - ii. multiply by N
 - iii. 1-D transform along each column of $F(x,v)$

3.3.2 Translation

- $f(x, y) \exp(j2\pi (u_0 x + v_0 y) / N) \leftrightarrow F(u - u_0, v - v_0)$

$$f(x - x_0, y - y_0) \leftrightarrow F(u, v) \exp(-j2\pi (u_0 x + v_0 y) / N)$$

- $u_0 = v_0 = N/2$;

$$\exp(j2\pi (u_0 x + v_0 y) / N) = e^{j\pi (x+y)} = (-1)^{x+y}$$

$$\rightarrow f(x, y)(-1)^{x+y} \leftrightarrow F(u - \frac{N}{2}, v - \frac{N}{2})$$

3.3.3 Periodicity and Conjugate Symmetry

- periodicity : DFT and its inverse are periodic with period N :

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

- conjugate symmetry :

- if $f(x,y)$ is real,

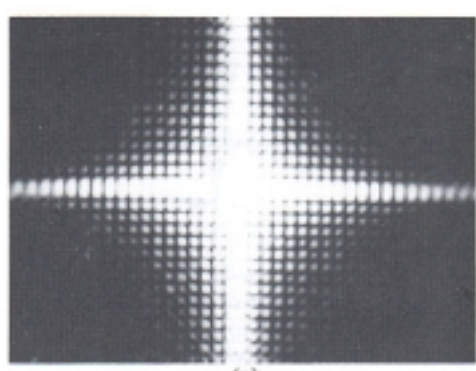
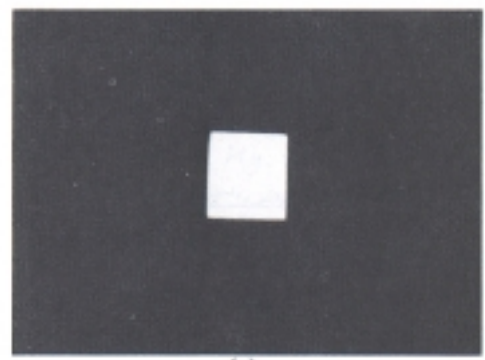
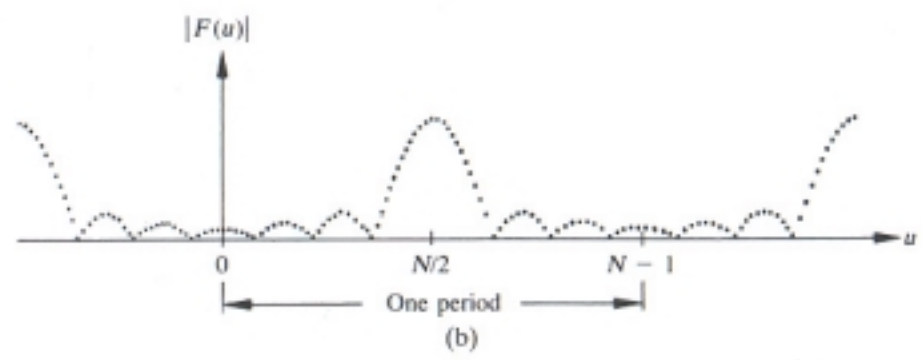
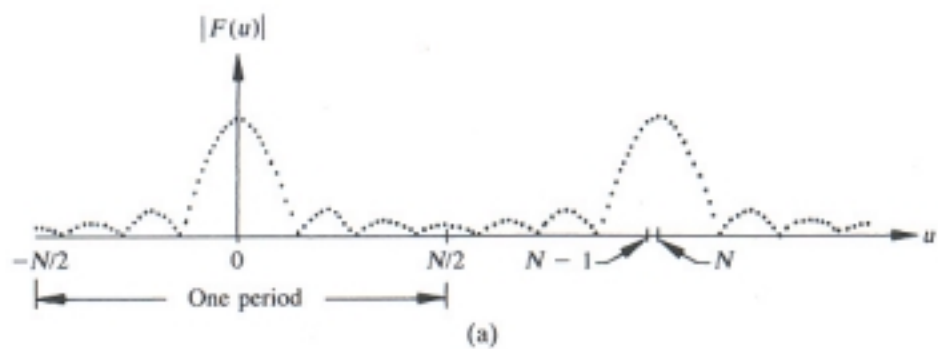
$$F(u, v) = F^* (-u, -v)$$

or $|F(u, v)| = |F(-u, -v)|$: centered on the origin

- display of one full period :

: move the origin of the transform to the point $u = N/2$

$$f(x)(-1)^x$$



3.3.4 Rotation

- in polar coordinate :

$$x = r \cos \theta, \quad y = r \sin \theta, \quad u = \omega \cos \phi, \quad v = \omega \sin \phi$$

$$(x, y) \rightarrow (r, \theta)$$

$$F(u, v) \rightarrow F(\omega, \phi)$$

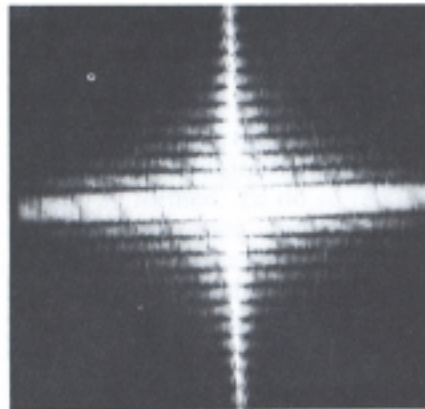
- FT pair for rotation function :

- $f(r, \theta + \theta_0) \leftrightarrow F(\omega, \phi + \theta_0)$

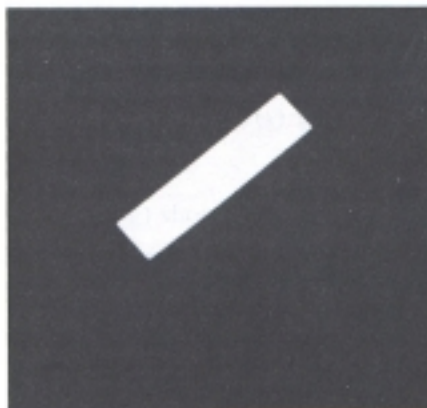
- rotating $f(x, y)$ by angle $\theta_0 \rightarrow$ rotating $F(u, v)$ by angle θ_0



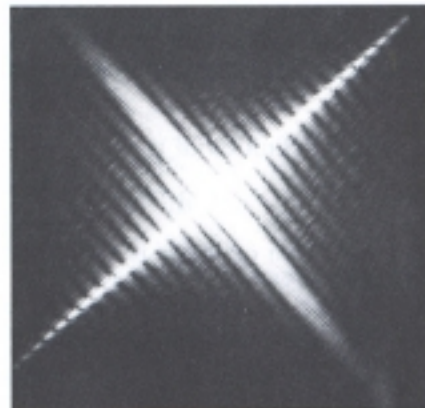
(a)



(b)



(c)



(d)

3.3.5 Distributivity and scaling

- distributivity :

$$F\{f_1(x, y) + f_2(x, y)\} = F\{f_1(x, y)\} + F\{f_2(x, y)\}$$

$$F\{f_1(x, y) \cdot f_2(x, y)\} \neq F\{f_1(x, y)\} \cdot F\{f_2(x, y)\}$$

- scaling :

$$F\{bf(x, y)\} \leftrightarrow bF\{f(x, y)\}$$

$$F\left\{f\left(\frac{x}{b}, \frac{y}{b}\right)\right\} \leftrightarrow \frac{1}{b^2} F\{f(x, y)\}$$

3.3.6 Average Value

- average value of 2-D discrete function

$$\bar{f} = \frac{1}{N^2} \sum_{x=0}^{N-1} \sum_{y=0}^{N-1} f(x, y)$$

$$= F(0, 0)$$

3.3.7 Laplacian

- Laplacian of 2-D function $f(x, y)$:

$$\nabla^2 f(x, y) = \frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2}$$

$$F\{\nabla^2 f(x, y)\} = -(\pi^2 u^2 + v^2) F\{f(x, y)\}$$

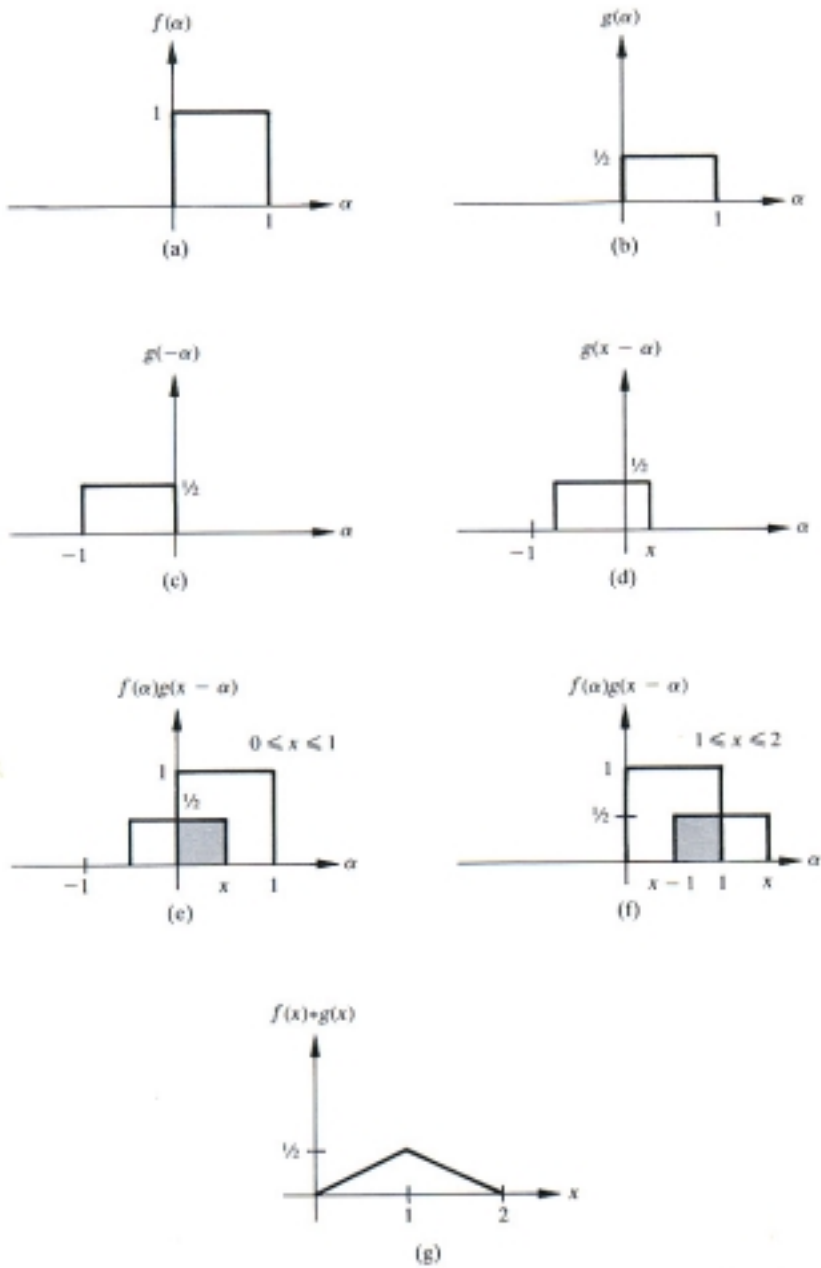
: useful for outlining edges in an image

3.3.8 Convolution and Correlation

- Convolution :

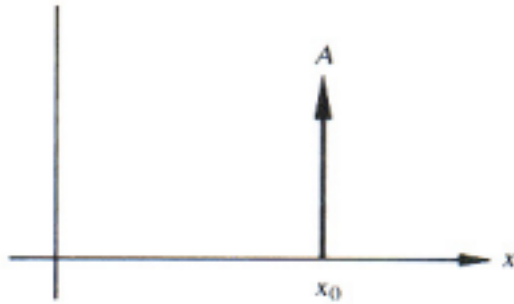
$$f(x) * g(x) = \int_{-\infty}^{\infty} f(\alpha) g(x - \alpha) d\alpha$$

ex.

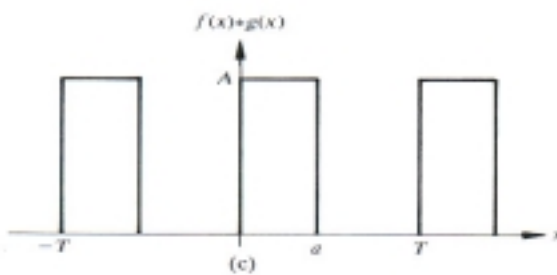
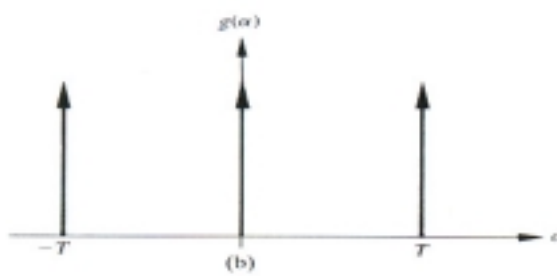
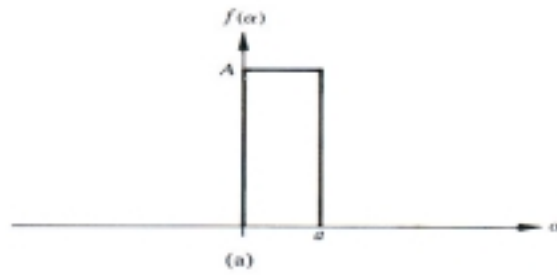


$$\int_{-\infty}^{\infty} x \delta(x - x_0) dx = x_0$$

$$\int_{-\infty}^{\infty} \delta(x - x_0) dx = \int_{x_0^-}^{x_0^+} \delta(x - x_0) dx = 1$$



- ex.



- in freq. Domain, convolution theorem ;

$$f(x) * g(x) \leftrightarrow F(u) * G(u)$$

$$f(x)g(x) \leftrightarrow F(u) * G(u)$$

- In DFT ;

$f(x), g(x)$: periodic some period M

$$M \geq A + B - 1$$

: no wraparound error

- appending zero to the samples ;

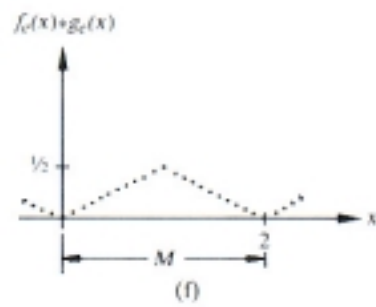
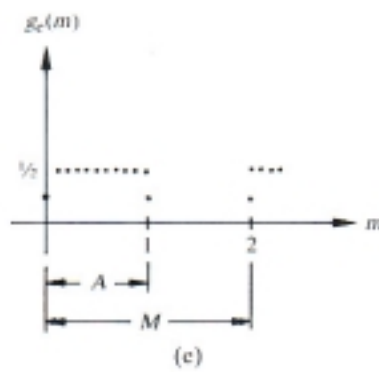
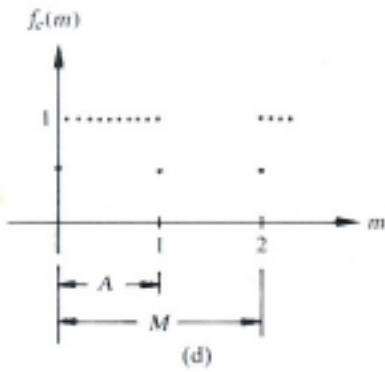
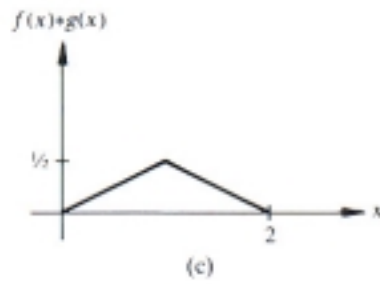
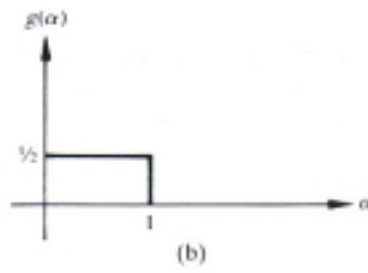
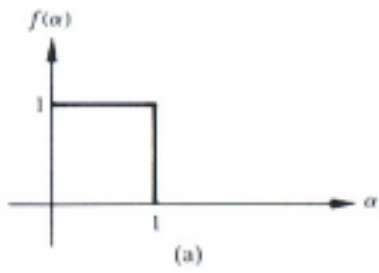
$$f_e(x) = \begin{cases} f(x) & 0 \leq x \leq A-1 \\ 0 & A \leq x \leq M-1 \end{cases}$$

$$g_e(x) = \begin{cases} g(x) & 0 \leq x \leq B-1 \\ 0 & B \leq x \leq M-1 \end{cases}$$

$$f_e(x) * g_e(x) = \frac{1}{M} \sum_{m=0}^{M-1} f_0(m) g_e(x-m)$$

for $x=0, 1, \dots, M-1$

ex.



- 2-D convolution ;

$$f(x, y) * g(x, y) = \iint_{-\infty}^{\infty} f(\alpha, \beta) g(x - \alpha, y - \beta) d\alpha d\beta$$

$$f(x, y) * g(x, y) \leftrightarrow F(u, v)G(u, v)$$

$$f(x, y)g(x, y) \leftrightarrow F(u, v) * G(u, v)$$

- array size ;

$$\begin{aligned} \checkmark \quad f(x, y) &: A \times B \\ g(x, y) &: C \times D \end{aligned}$$

✓ to avoid wraparound error

$$M \geq A + C - 1$$

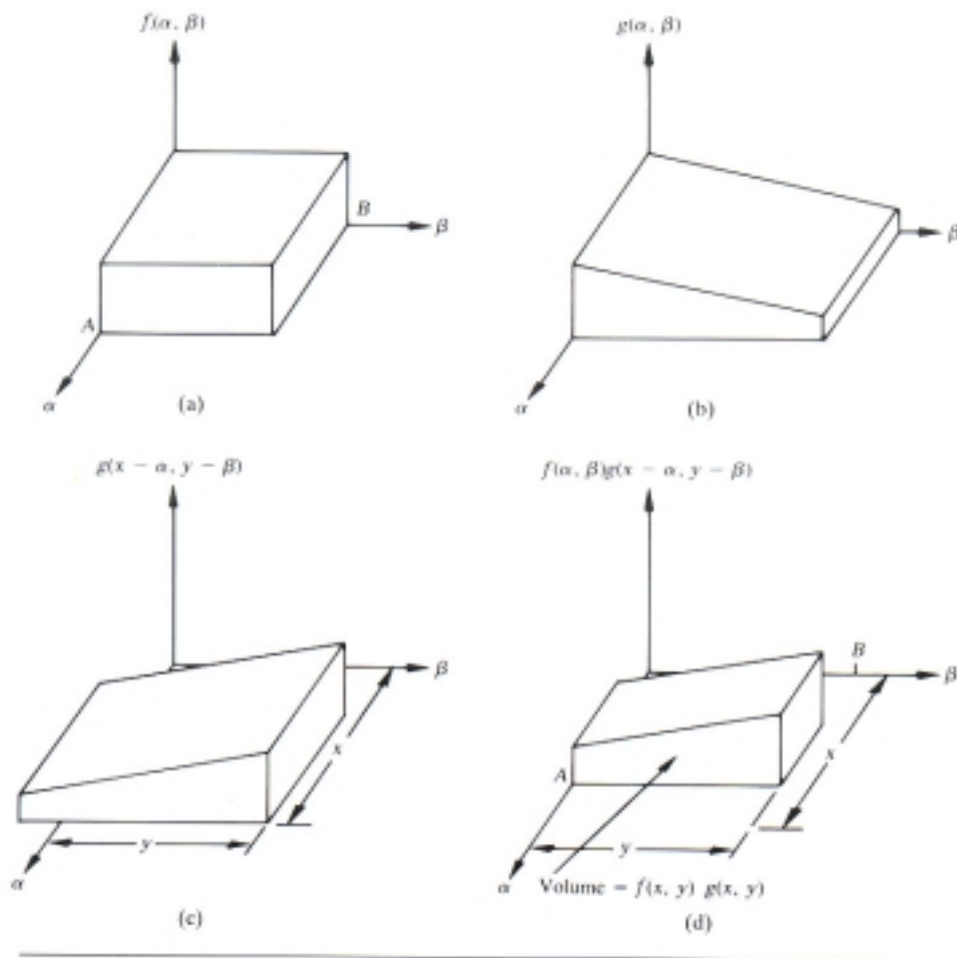
$$N \geq B + D - 1$$

✓ extending $f(x, y)$, $g(x, y)$;

$$f_e(x, y) = \begin{cases} f(x, y) & 0 \leq x \leq A-1, \text{ and } 0 \leq y \leq B-1 \\ 0 & A \leq x \leq M-1, \text{ and } B \leq y \leq N-1 \end{cases}$$

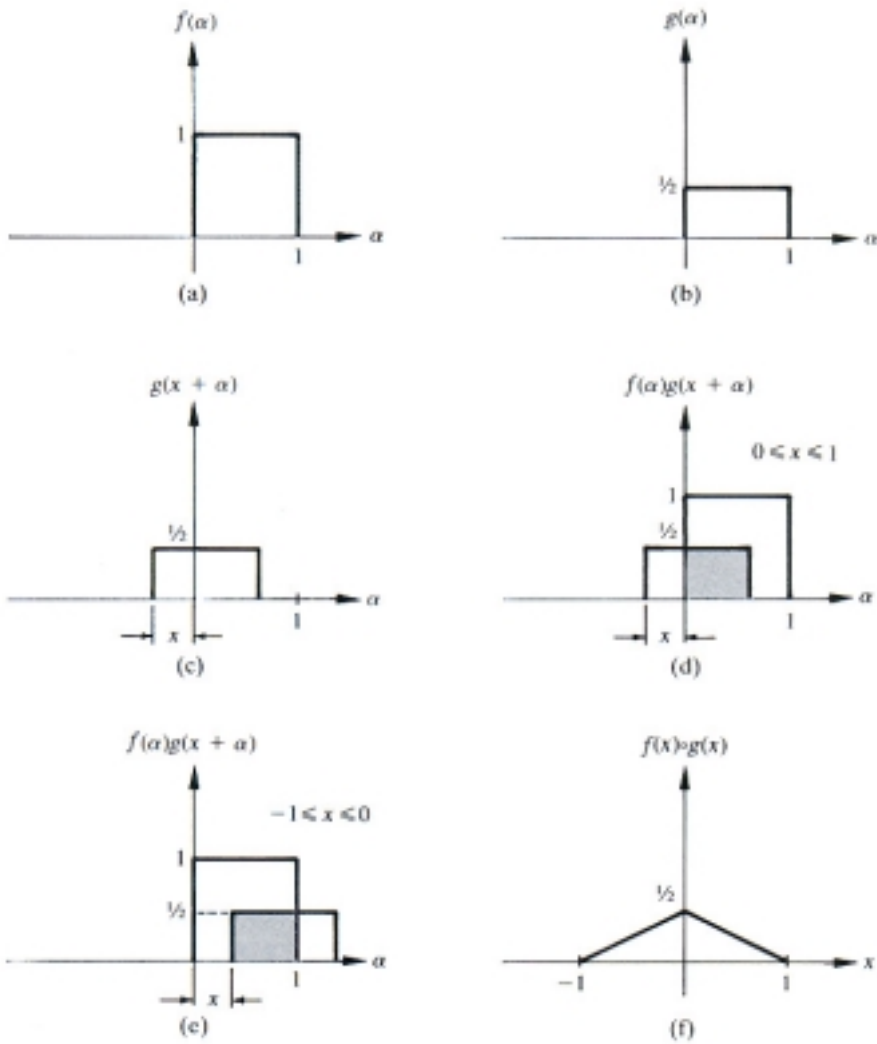
$$g_e(x, y) = \begin{cases} g(x, y) & 0 \leq x \leq C-1, \text{ and } 0 \leq y \leq D-1 \\ 0 & A \leq x \leq M-1, \text{ and } D \leq y \leq N-1 \end{cases}$$

$$f_e(x, y) * g_e(x, y) = \frac{1}{MN} \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} f_e(m, n) g_e(x-m, y-n)$$



● correlation of 2 continuous functions $f(x)$ and $g(x)$:

- $f(x) \circ g(x) = \int_{-\infty}^{\infty} f^*(\alpha)g(x + \alpha)d\alpha$
- not folded to the origin
- if $f(x)=g(x)$, autocorrelation
- if not, crosscorrelation



- discrete equivalent

$$f_e(x) \circ g_e(x) = \frac{1}{M} \sum_{m=0}^{M-1} f_e^*(m) g_e(x+m)$$

- 2-D ;

$$f_e(x, y) \circ g_e(x, y) = \frac{1}{MN} \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} f_e^*(m, n) g_e(x+m, y+n)$$

- correlation theorem ;

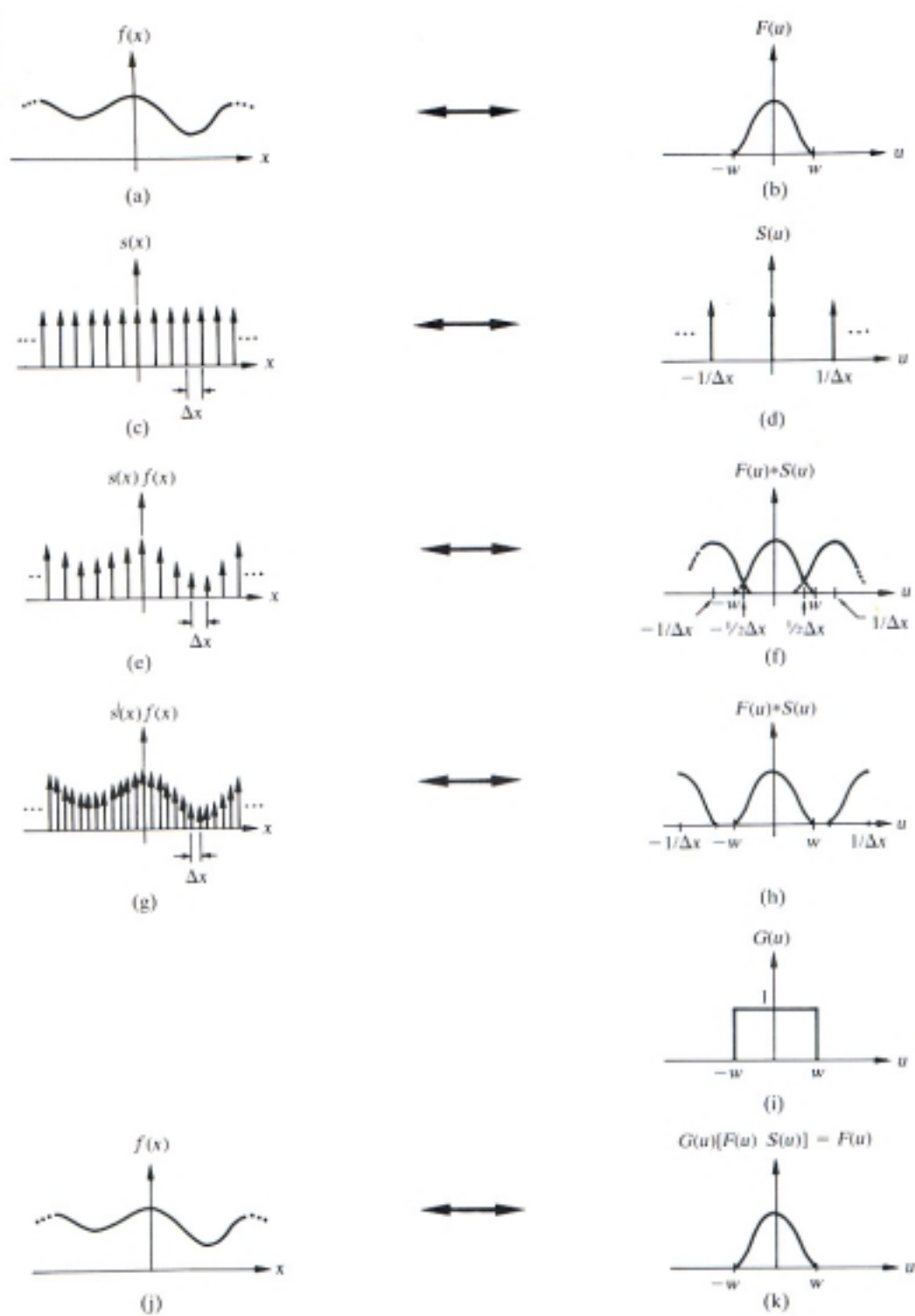
$$f(x, y) \circ g(x, y) \leftrightarrow F^*(u, v) G(u, v)$$

$$f(x, y) \circ g(x, y) \leftrightarrow F^*(u, v) G(u, v)$$

3.3.9 Sampling

1) 1-D functions

- infinite sampling interval ;



- band-limited function
- sampling function $s(x)$; comb function (or impulse train)
: a train of impulse Δx units apart
- sampling of $f(x)$: $s(x)f(x)$, $F(u)*S(u)$
: convolution theorem
- transform : periodic with period $1/\Delta x$, overlap (aliasing)
- Nyquist rate :

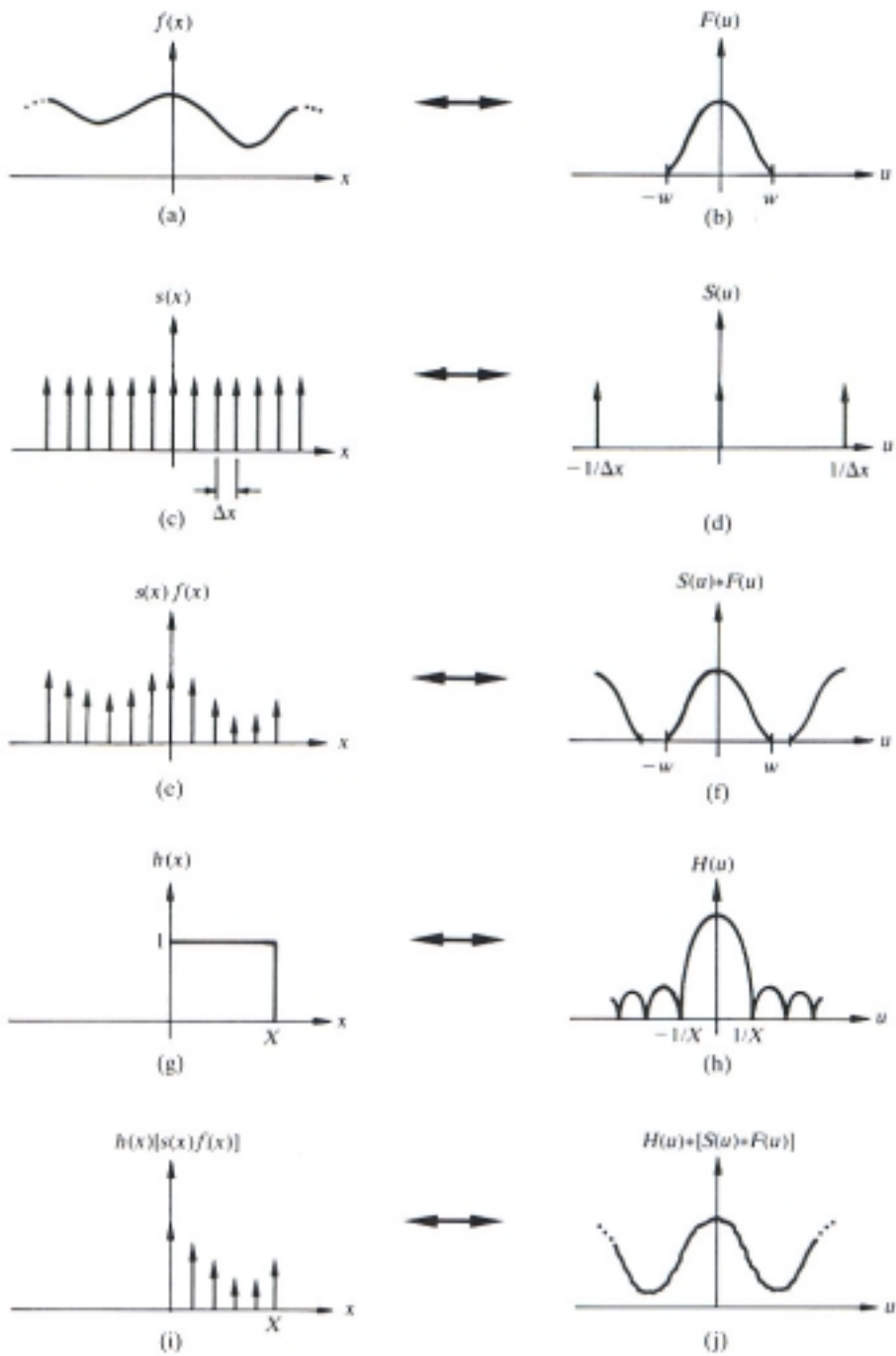
$$\Delta x \leq \frac{1}{2W} \quad \text{or} \quad \frac{1}{\Delta x} \geq 2W$$

- Whittaker-Shannon sampling theorem :

: complete recovery of a band-limited function from samples

- Finite sampling interval (practical case) :

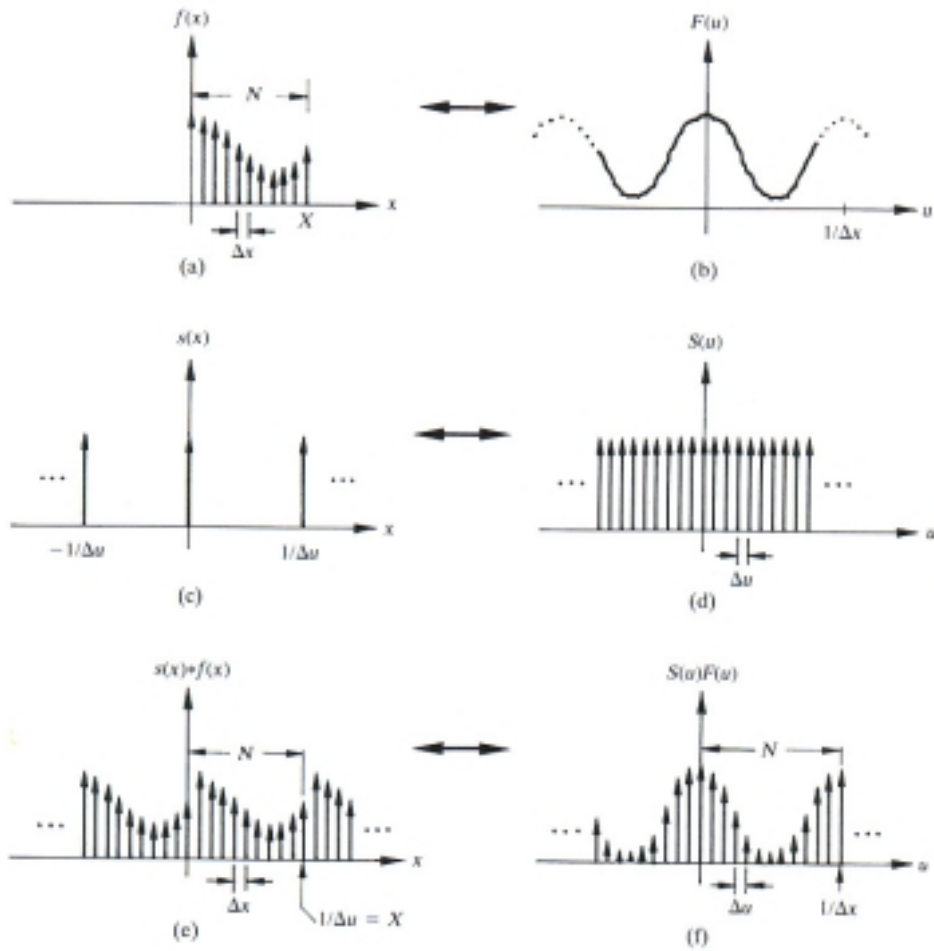
: sampling interval: $[0, X]$



- window function (rectangular)

$$h(x) = \begin{cases} 1 & 0 \leq x \leq \omega \\ 0 & \text{elsewhere} \end{cases} \quad \text{:finite sampling interval}$$

- The periodicity of DFT :



- N : samples of $f(x)$ and $F(u)$

→ in x Domain ;

$$N\Delta x = X$$

→ in freq. Domain ;

$$N\Delta u = \frac{1}{\Delta x} \rightarrow \Delta u = \frac{1}{N\Delta x}$$

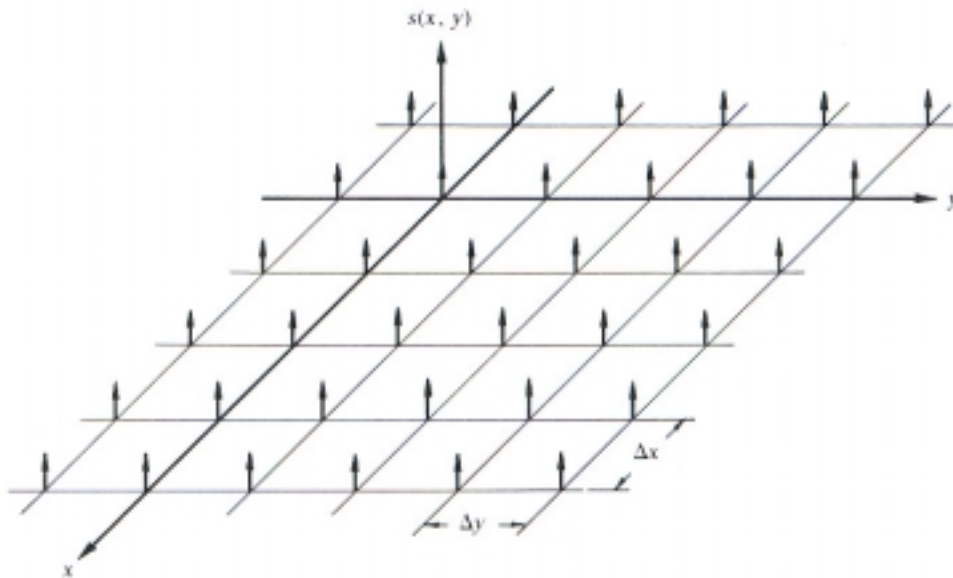
2) 2-D functions

- sampling process for 2-D functions :

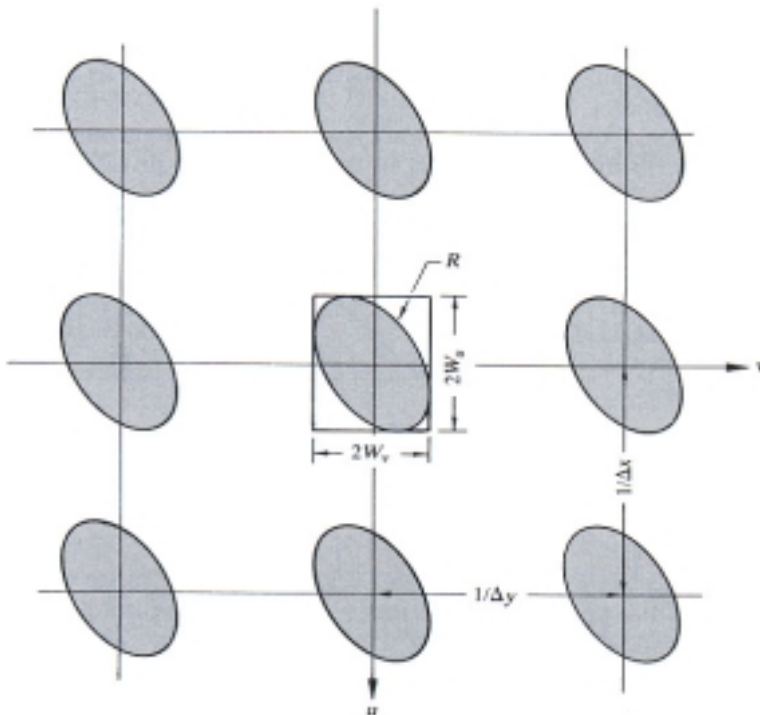
- 2-D impulse function : $\delta(x,y)$ defined as ;

$$\iint_{-\infty}^{\infty} f(x, y) \delta(x - x_0, y - y_0) dx dy = f(x_0, y_0)$$

2-D sampling function : $s(x, y)$



- sampled function : $s(x, y)f(x, y)$
- freq. Domain



- $S(u, v) * F(u, v)$

where $S(u, v)$: a train of impulses with separation $1/\Delta x$ and $1/\Delta y$ in the u

and

v directions

- $f(x, y)$: band limited
- $S(u, v) * F(u, v)$: periodic in two dimensions

● no aliasing condition :

- if $2W_u, 2W_v$: widths in the u and v directions

$$\rightarrow \frac{1}{\Delta x} \geq 2W_u, \quad \frac{1}{\Delta y} \geq 2W_v$$

● low-pass filtering :

$$G(u, v) = \begin{cases} 1 & -W_u \leq u \leq W_u \quad -W_v \leq v \leq W_v \\ 0 & \text{otherwise} \end{cases}$$

● 2-D DFT :

- for $N \times N$ images

$$\Delta u = \frac{1}{N\Delta x}, \quad \Delta v = \frac{1}{N\Delta y}$$

→ complete 2-D period will be covered by $N \times N$ uniformly spaced values in both spatial and frequency domain