

3 – IMAGE TRANSFORMS (B)

THE DISCRETE FOURIER TRANSFORM

Suppose that a continuous function $f(x)$ is discretized into a sequence

$$\{f(x_0), f(x_0 + \Delta x), f(x_0 + 2\Delta x), \dots, f(x_0 + [N - 1]\Delta x)\}$$

by taking N samples Δx units apart.

Note that x may be used as either a discrete or continuous variable, depending on context. In the discrete case, we define

$$f(x) = f(x_0 + x\Delta x) \quad x = 0, 1, 2, \dots, N - 1$$

i.e.,

$$f(0) = f(x_0)$$

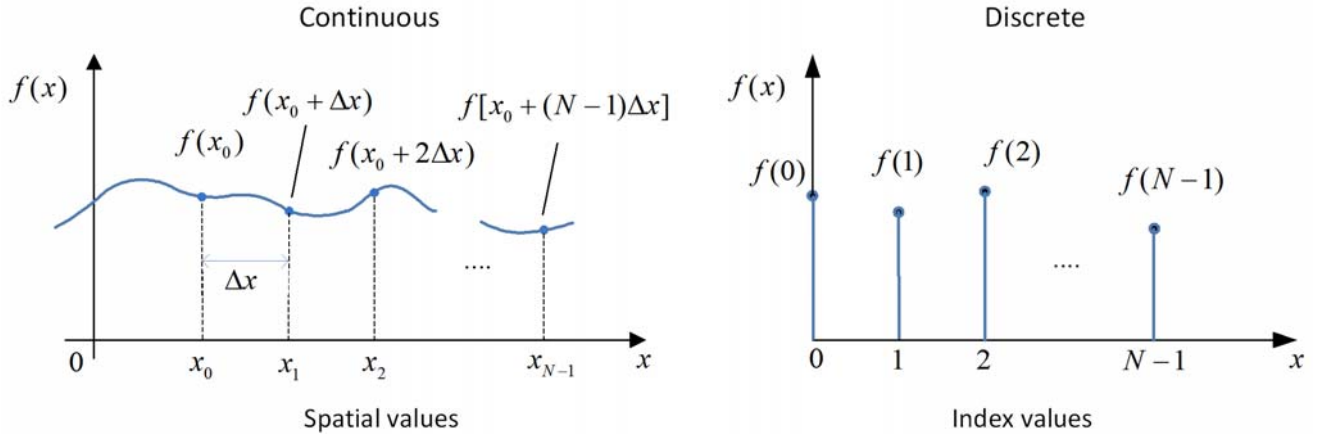
$$f(1) = f(x_0 + \Delta x)$$

$$f(2) = f(x_0 + 2\Delta x)$$

$$f(3) = f(x_0 + 3\Delta x)$$

...

$$f(N - 1) = f[x_0 + (N - 1)\Delta x]$$



The discrete Fourier transform (DFT) pair is given by

$$F(u) = \frac{1}{N} \sum_{x=0}^{N-1} f(x) \exp[-j2\pi ux/N]; \quad u = 0, 1, 2, \dots, N-1 \quad (1)$$

$$f(x) = \sum_{u=0}^{N-1} F(u) \exp[j2\pi ux/N]; \quad x = 0, 1, 2, \dots, N-1 \quad (2)$$

The terms Δu and Δx are related by

$$\Delta u = \frac{1}{N\Delta x} \quad (3)$$

$$F(u) = F(u\Delta u) \quad u = 0, 1, 2, \dots, N-1$$

i.e.,

$$F(0) = F(0)$$

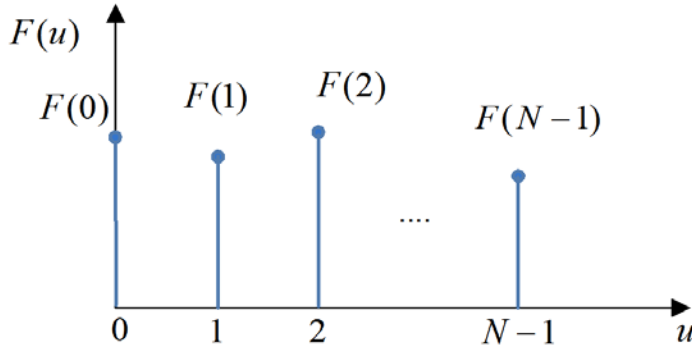
$$F(1) = F(\Delta u)$$

$$F(2) = F(2\Delta u)$$

$$F(3) = F(3\Delta u)$$

...

$$f(N-1) = f((N-1)\Delta u)$$



For example,

$$\Delta x = 2 \text{ mm}, \quad N = 500$$

$$\Delta u = \frac{1}{N\Delta x} = 1 \times 10^{-3} \text{ cycles per mm}$$

In the 2D case, the DFT pair is

$$F(u, v) = \frac{1}{MN} \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) \exp[-j2\pi(ux/M + vy/N)]; \quad (4)$$

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

and

$$f(x, y) = \sum_{u=0}^{M-1} \sum_{v=0}^{N-1} F(u, v) \exp[j2\pi(ux/M + vy/N)]; \quad (5)$$

$$\text{for } x = 0, 1, 2, \dots, M-1, y = 0, 1, 2, \dots, N-1.$$

The sampling increments in the spatial and frequency domains are related by

$$\Delta u = \frac{1}{M\Delta x}, \quad \Delta v = \frac{1}{N\Delta y} \quad (6)$$

The Fourier spectrum, phase, and power spectrum of 1D and 2D discrete functions are computed as for the continuous case.

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

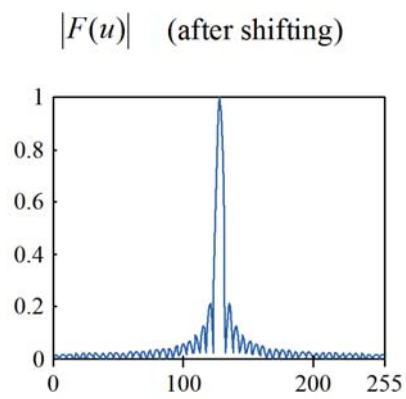
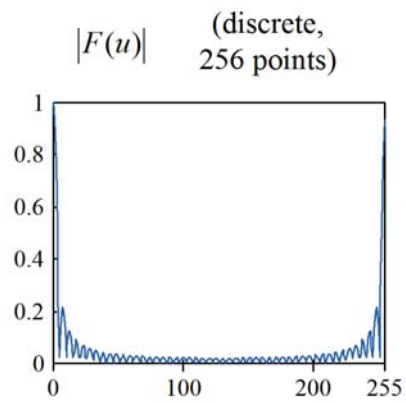
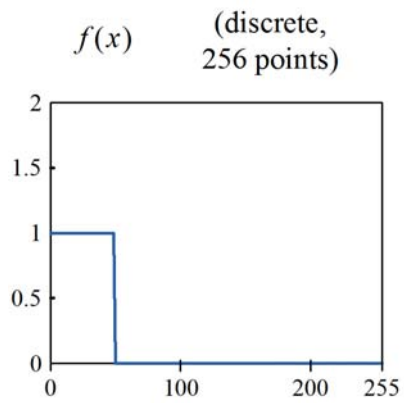
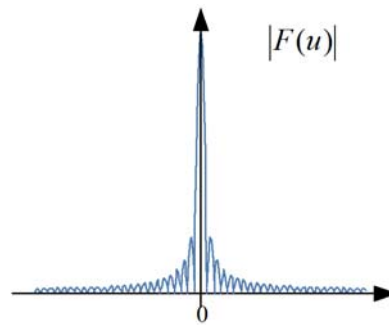
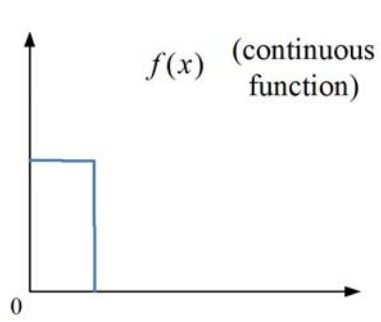
$$\text{Phase spectrum: } \phi(u, v) = \tan^{-1}[I(u, v)/R(u, v)]$$

$$\text{Power spectrum: } P(u, v) = |F(u, v)|^2 = R^2(u, v) + I^2(u, v)$$

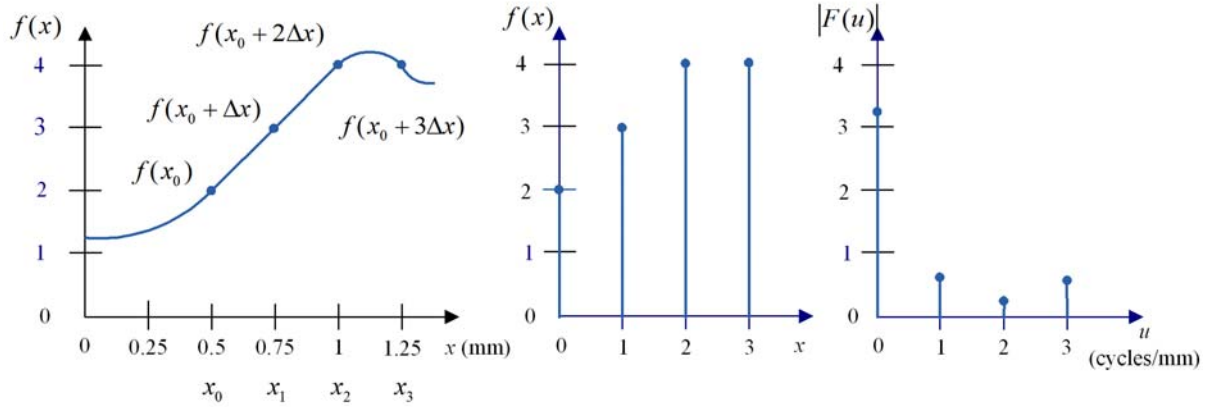
The direct computation of an N -point DFT requires of the order of N^2 operations. For an $M \times N$ array, M^2N^2 operations are required. This can be considerably reduced by the fast Fourier transform (FFT) algorithm to $MN \log_2(MN)$ operations. Suppose $M = N = 2^9$. Then

$$\begin{aligned} \text{Direct DFT:} \quad M^2N^2 &= 69 \times 10^9 \\ \text{FFT:} \quad MN \log_2(MN) &= 4.7 \times 10^6 \end{aligned}$$

Example (1D DFT)



Example (1D DFT)



Sampling takes place at

$x_0 = 0.5$ mm, $x_1 = 0.75$ mm, $x_2 = 1.0$ mm, $x_3 = 1.25$ mm, giving
 $f(0) = 2$, $f(1) = 3$, $f(2) = 4$, $f(3) = 4$.

$$F(u) = \frac{1}{N} \sum_{x=0}^{N-1} f(x) \exp[-j2\pi ux/N] = \frac{1}{4} \sum_{x=0}^3 f(x) \exp[-j2\pi ux/4]$$

$$F(0) = \frac{1}{4} \sum_{x=0}^3 f(x) \exp(0)$$

$$= \frac{1}{4} [f(0) + f(1) + f(2) + f(3)] = \frac{1}{4} [2 + 3 + 4 + 4]$$

$$= 3.25 \quad \text{or} \quad 3.25 \angle 0^\circ$$

$$F(1) = \frac{1}{4} \sum_{x=0}^3 f(x) \exp[-j2\pi x/4]$$

$$= \frac{1}{4} [f(0) \exp(0) + f(1) \exp(-j\pi/2) + f(2) \exp(-j\pi) + f(3) \exp(-j3\pi/2)]$$

$$= \frac{1}{4} (-2 + j) \quad \text{or} \quad 0.56 \angle 153^\circ$$

Similarly,

$$F(2) = -\frac{1}{4} (1 + j \times 0) \quad \text{or} \quad 0.25 \angle 180^\circ, \quad F(3) = -\frac{1}{4} (2 + j) \quad \text{or} \quad 0.56 \angle -153^\circ$$

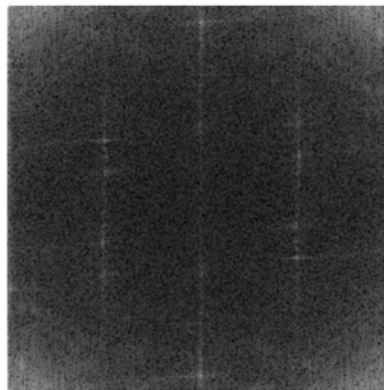
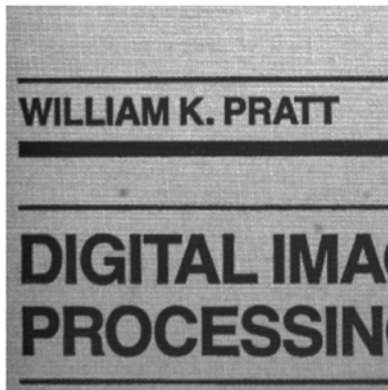
Fourier spectrum:

$$|F(0)| = 3.25, \quad |F(1)| = 0.56 \quad |F(2)| = 0.25 \quad |F(3)| = 0.56$$

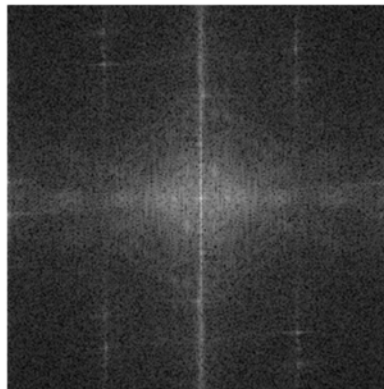
In this example,

$$N = 4, \quad \Delta x = 0.25 \text{ mm}, \quad \Delta u = 1/(N\Delta x) = 1 \text{ cycle/mm}$$

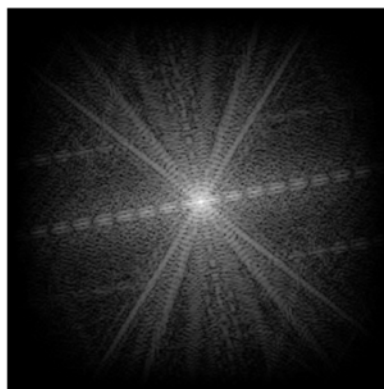
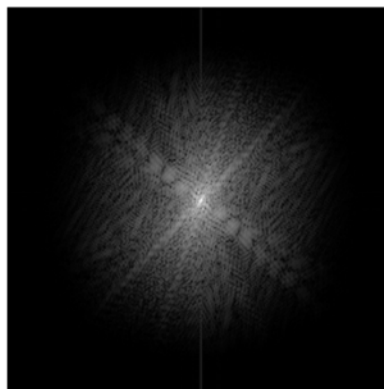
Examples of 2D DFT



Without shifting



With shifting



Some Properties of the 2D DFT

Separability

We can write Eq. (4) in the separable form

$$F(u, v) = \frac{1}{M} \sum_{x=0}^{M-1} \exp[-j2\pi ux/M] \times \left\{ \frac{1}{N} \sum_{y=0}^{N-1} f(x, y) \exp[-j2\pi vy/N] \right\} \quad (7)$$

$$= \frac{1}{M} \sum_{x=0}^{M-1} \exp[-j2\pi ux/M] \times \{F_r(x, v)\} \quad (8)$$

$$= \frac{1}{M} \sum_{x=0}^{M-1} F_r(x, v) \exp[-j2\pi ux/M] \quad (9)$$

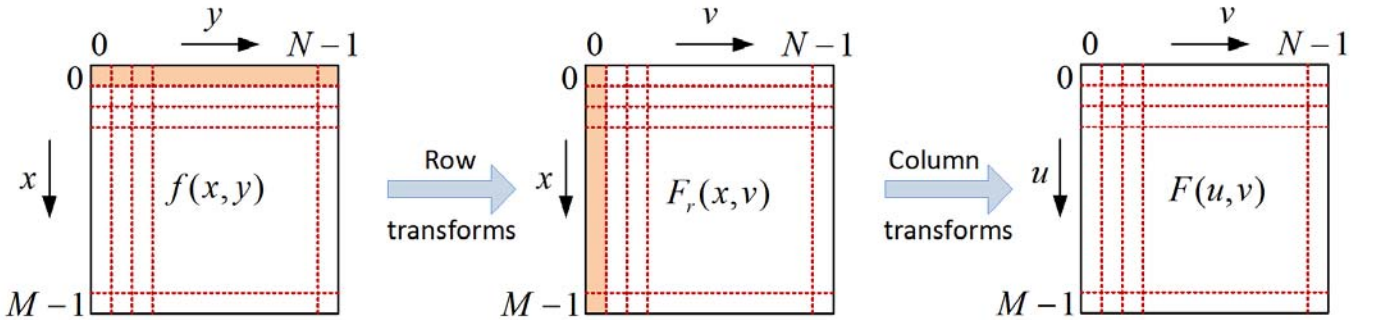
where

$$F_r(x, v) = \frac{1}{N} \sum_{y=0}^{N-1} f(x, y) \exp[-j2\pi vy/N] \quad (10)$$

Because of the separability property, $F(u, v)$ (or $f(x, y)$) can be obtained in two steps by successive applications of the 1D Fourier transform (or its inverse).

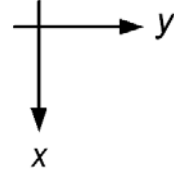
For each value of x , the expression inside the brackets is a 1D transform, with frequency values $v = 0, 1, \dots, N-1$. Therefore the 2D function $F(x, v)$ is obtained by taking a transform along each row of $f(x, y)$. The desired result $F(u, v)$ is then obtained by taking a transform along each column of $F(x, v)$.

The same results may be obtained by first taking transforms along the columns of $f(x, y)$ and then along the rows of that result.



Example

In general, for DFT computation, we will use this convention: x axis points down, y axis to the right, origin at the top left corner.



$$f(x, y) = \begin{bmatrix} 4 & 2 & 2 & 1 \\ 2 & 1 & 1 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix}$$

Taking the row transforms:

$$F_r(x, v) = \frac{1}{4} \begin{bmatrix} 9 & 2-j & 3 & 2+j \\ 4 & 1-j & 2 & 1+j \\ 2 & 1-j & 0 & 1+j \\ 1 & 1 & 1 & 1 \end{bmatrix}$$

Taking the column transforms:

$$F(u, v) = \frac{1}{16} \begin{bmatrix} 16 & 5-j3 & 6 & 5+j3 \\ 7-j3 & 0 & 3-j & 2 \\ 6 & 1-j1 & 0 & 1+j \\ 7+j3 & 2 & 3+j & 0 \end{bmatrix}$$

Average Value

The average value of the digital image $f(x, y)$ is

$$\bar{f}(x, y) = \frac{1}{MN} \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) \quad (11)$$

It is easily shown that

$$\bar{f}(x, y) = F(0, 0) \quad (12)$$

Translation

The translation property is

$$f(x - a, y - b) \leftrightarrow F(u, v) \exp[-j2\pi(ua/M + vb/N)] \quad (13)$$

For example, for $M = 100, N = 100$, and $a = 20, b = 40$,

$$f_1(x, y) = f(x - 20, y - 40) \quad (14)$$

$$F_1(u, v) = \mathcal{F}\{f(x - 20, y - 40)\} \quad (15)$$

$$= F(u, v) \exp[-j2\pi(u/5 + 2v/5)] \quad (16)$$

$$= F(u, v) \exp(-j2\pi u/5) \exp(-j4\pi v/5) \quad (17)$$

Note that a shift in $f(x, y)$ does not affect the magnitude of its Fourier transform since

$$|F(u, v) \exp[-j2\pi(ua/M + vb/N)]| = |F(u, v)| \quad (18)$$

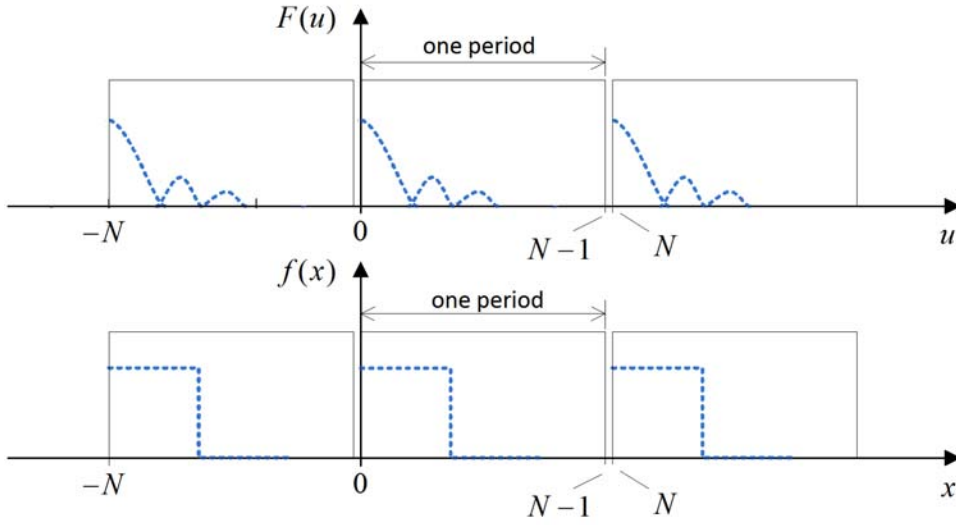
Periodicity and Conjugate Symmetry

From the definition of the DFT, we can show that $F(u)$ is periodic with period N :

$$F(u) = F(u + kN) \quad k = 0, \pm 1, \pm 2, \dots \quad (19)$$

This periodicity property also applies to the inverse of $F(u)$, i.e., $f(x)$ computed from Eq. 2 is periodic:

$$f(x) = f(x + kN) \quad k = 0, \pm 1, \pm 2, \dots \quad (20)$$

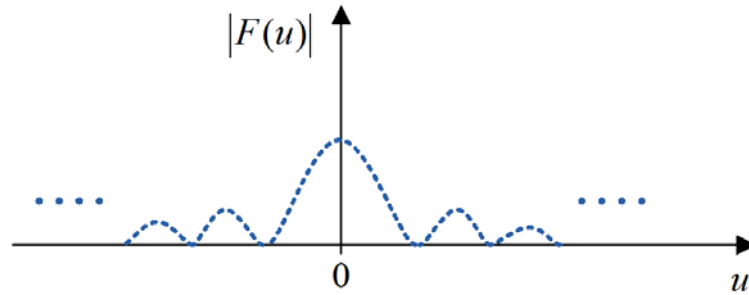


Furthermore, if $f(x)$ is real, the DFT also exhibits conjugate symmetry:

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

Furthermore, the DFT magnitude is then symmetrical about $u = 0$ since

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



Combining conjugate symmetry and periodicity,

$$F(u) = F * (-u) = F^*(-u + N) \quad (21)$$

Hence, for the DFT magnitude function $|F(u)|$, we have

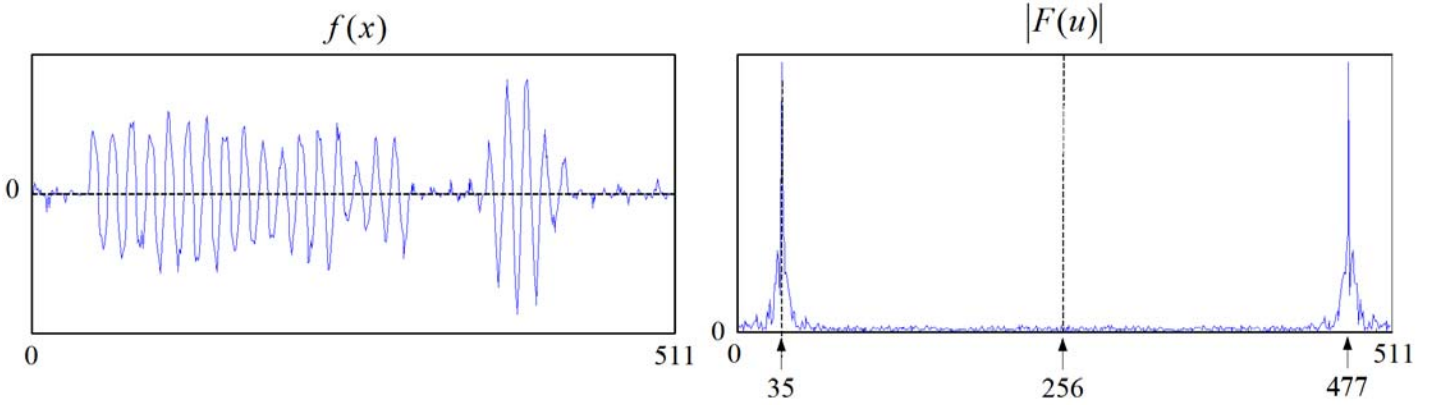
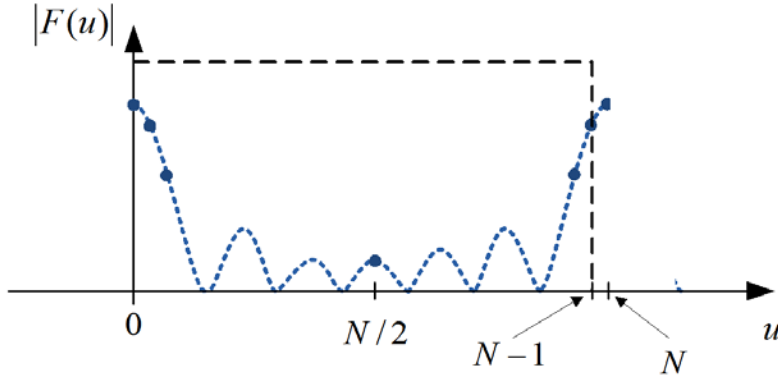
$$|F(u)| = |F(-u + N)| \quad (22)$$

In the window $u = 0, 1, \dots, N - 1$,

$$\begin{aligned} |F(0)| &= |F(N)| \\ |F(1)| &= |F(N - 1)| \\ |F(2)| &= |F(N - 2)| \\ &\dots \end{aligned}$$

$$|F(N/2)| = |F(N/2)|$$

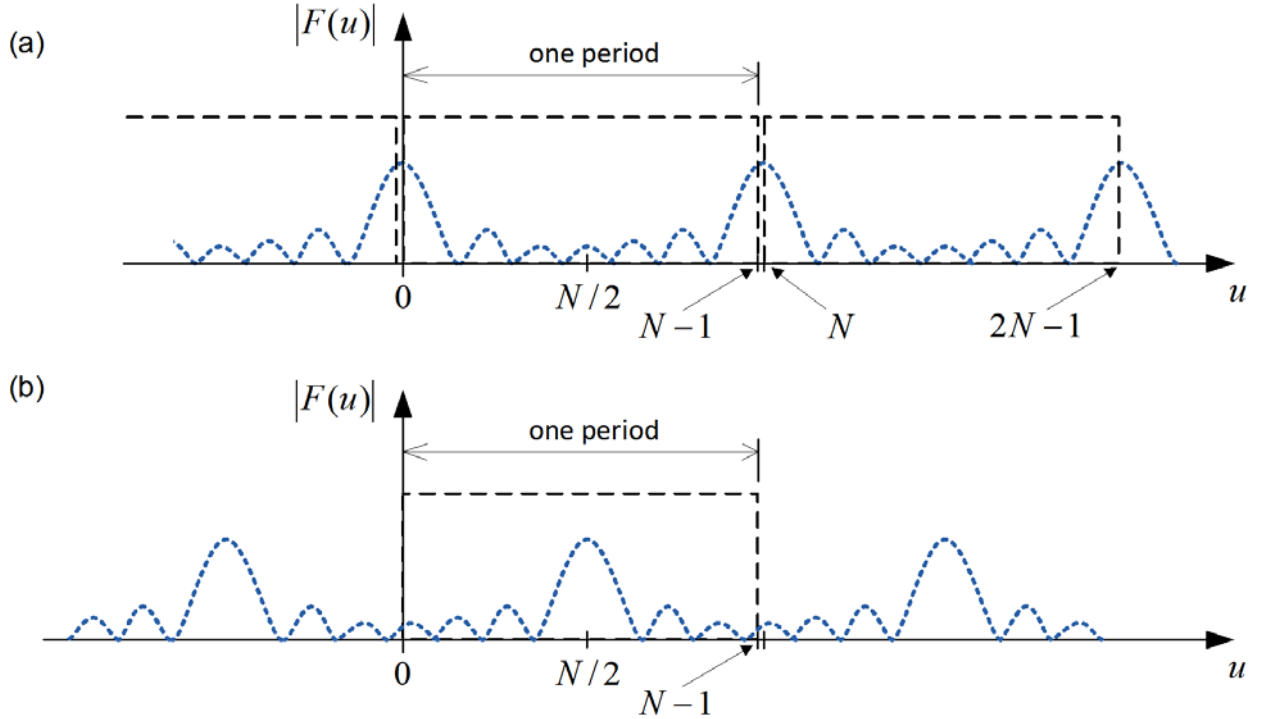
i.e., $|F(u)|$ is symmetrical about $u = N/2$.



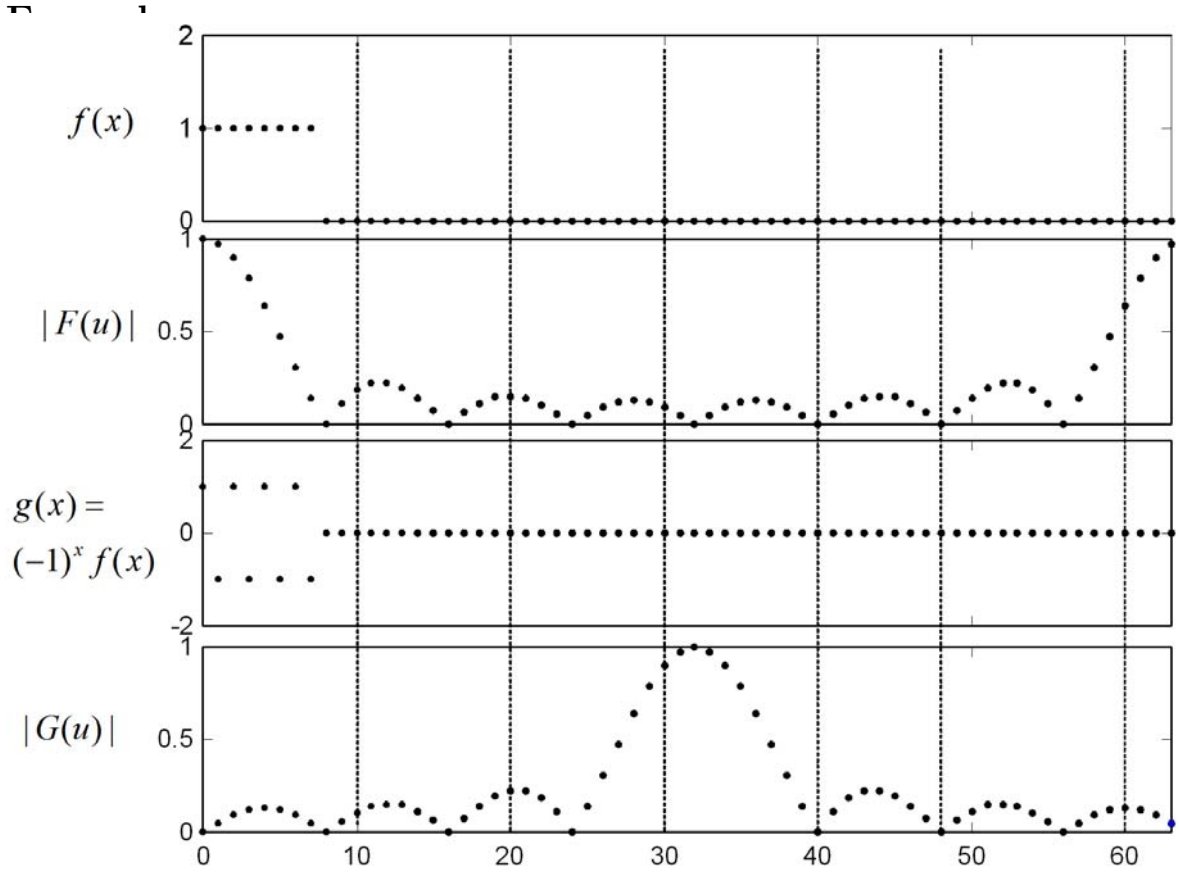
Because of the periodicity property, the values of $|F(u)|$ in this window are repeated along the u axis.

For viewing purposes, it would be better to centre $|F(u)|$, i.e., move the origin of the transform to the point $u = N/2$. This is done by multiplying $f(x)$ by $(-1)^x$ prior to taking the transform, i.e.,

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



(a) Fourier spectrum showing back-to-back half periods in the window $[0, N - 1]$;
(b) Shifted spectrum showing a full period in the same window.



Proof of Periodicity and Conjugate Symmetry

$$\begin{aligned}
 F(u) &= (1/N) \sum_{x=0}^{N-1} f(x) \exp[-j2\pi ux/N] \\
 F(u+N) &= (1/N) \sum_x f(x) \exp[-j2\pi(u+N)x/N] \\
 &= (1/N) \sum_x f(x) \exp[-j2\pi ux/N - j2\pi Nx/N] \\
 &= (1/N) \sum_x f(x) \exp[-j2\pi ux/N] \exp[-j2\pi x] \\
 &= (1/N) \sum_x f(x) \exp[-j2\pi ux/N] \quad \text{since } \exp[-j2\pi x] \equiv 1 \\
 &= F(u)
 \end{aligned}$$

$$\begin{aligned}
 F(-u) &= (1/N) \sum_x f(x) \exp[j2\pi ux/N] \\
 F^*(-u) &= (1/N) \sum_x f^*(x) \exp[-j2\pi ux/N] \\
 &= (1/N) \sum_x f(x) \exp[-j2\pi ux/N] \quad \text{for real } f \\
 &= F(u)
 \end{aligned}$$

For the 2D case involving an $N \times N$ image, both $f(x, y)$ and $F(u, v)$ are **periodic**:

$$f(x, y) = f(x + kN, y + lN) \quad k, l = 0, \pm 1, \pm 2, \dots \quad (23)$$

$$F(u, v) = F(u + kN, v + lN) \quad k, l = 0, \pm 1, \pm 2, \dots \quad (24)$$

If $f(x, y)$ is real, the DFT also exhibits conjugate symmetry:

$$F(u, v) = F^*(-u, -v) \quad (25)$$

Hence the DFT magnitude function $|F(u, v)|$ is **symmetrical about the origin**:

$$|F(u, v)| = |F(-u, -v)| \quad (26)$$

Similar to the 1D case, combining periodicity and symmetry about the origin means that in the window $u, v = 0, 1, \dots, N - 1$, $|F(u, v)|$ is symmetrical about $(u, v) = (N/2, N/2)$, i.e.,

$$|F(u, v)| = |F(N - u, N - v)| \quad (27)$$

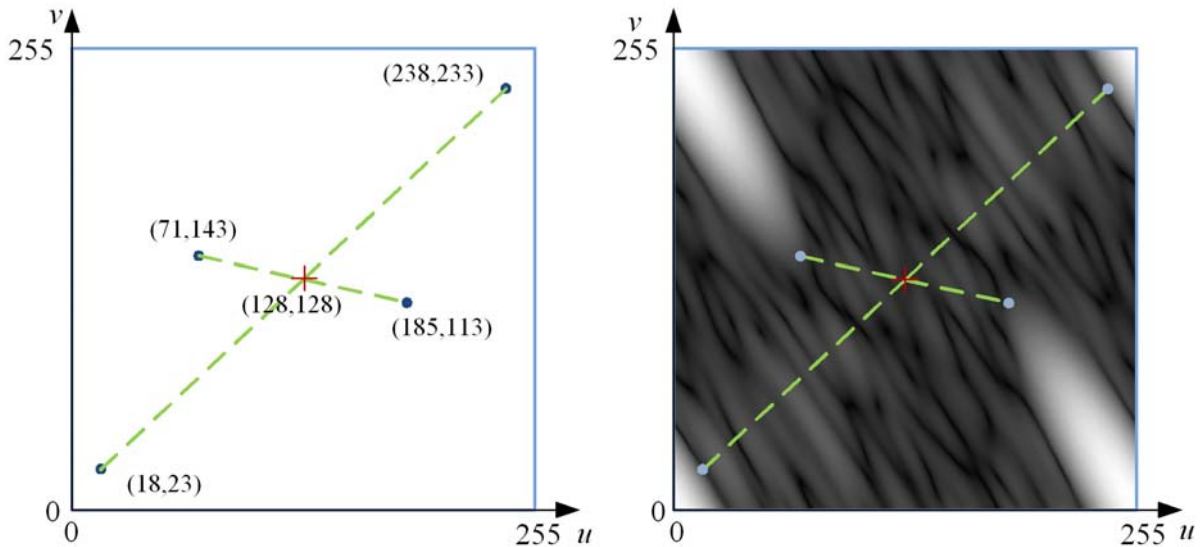
To move the origin of the transform to the point $(u, v) = (N/2, N/2)$, we multiply $f(x, y)$ by $(-1)^{x+y}$ prior to taking the transform, i.e.,

$$f'(x, y) = f(x, y) (-1)^{x+y}$$

Example

Consider a square image of size $N = 256$. $|F(u, v)|$ is symmetrical about $(u, v) = (128, 128)$.

$$\begin{aligned}|F(u, v)| &= |F(256 - u, 256 - v)| \\ |F(18, 23)| &= |F(238, 233)| \\ |F(71, 143)| &= |F(185, 113)|\end{aligned}$$



Summary:

$f(x)$ and $F(u)$ are periodic periodicity

For real $f(x)$:

$|F(u)| = |F(-u)|$ symmetry about origin

For real $f(x)$, periodicity and symmetry

$\Rightarrow |F(u)| = |F(N - u)|$ symmetry about $u = N/2$

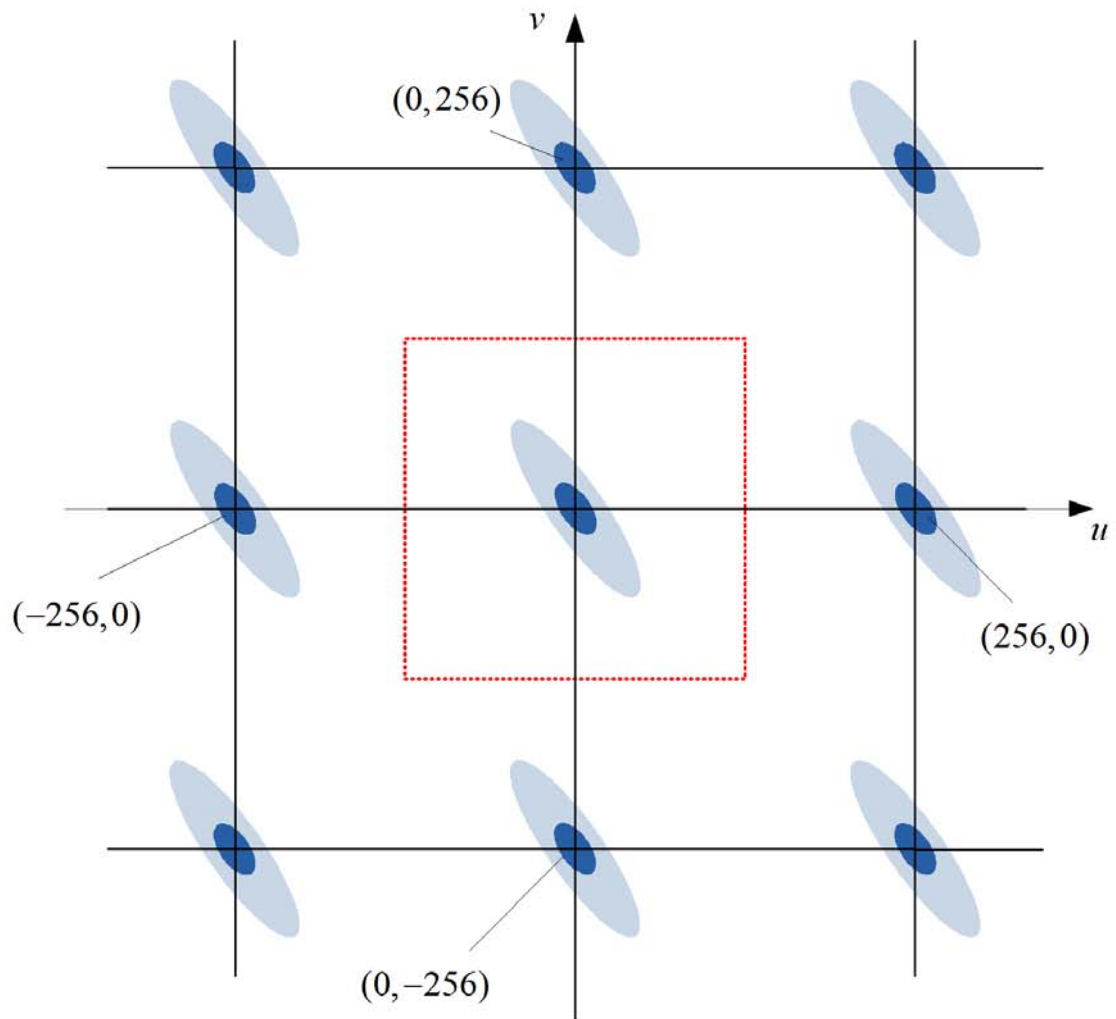
$f(x, y)$ and $F(u, v)$ are periodic periodicity

For real $f(x, y)$:

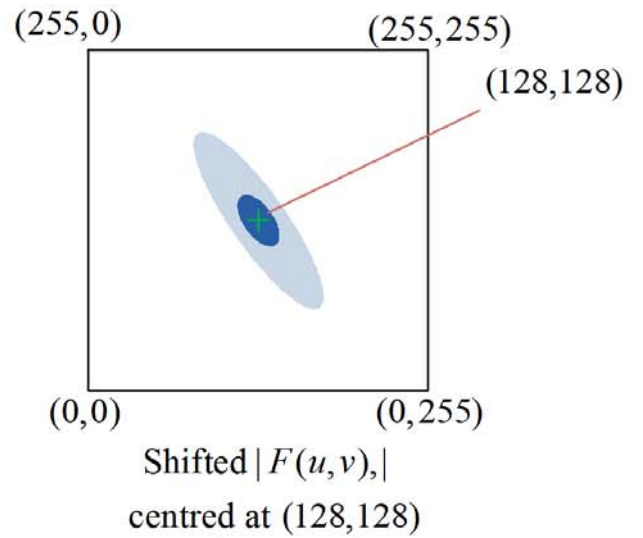
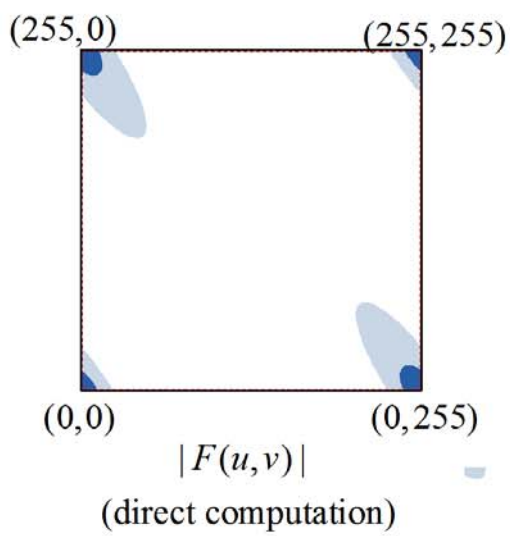
$|F(u, v)| = |F(-u, -v)|$ symmetry about origin

For real $f(x, y)$, periodicity and symmetry

$\Rightarrow |F(u, v)| = |F(N - u, N - v)|$ symmetry about $(u, v) = (N/2, N/2)$



Schematic depiction of $|F(u, v)|$ for real $f(x, y)$



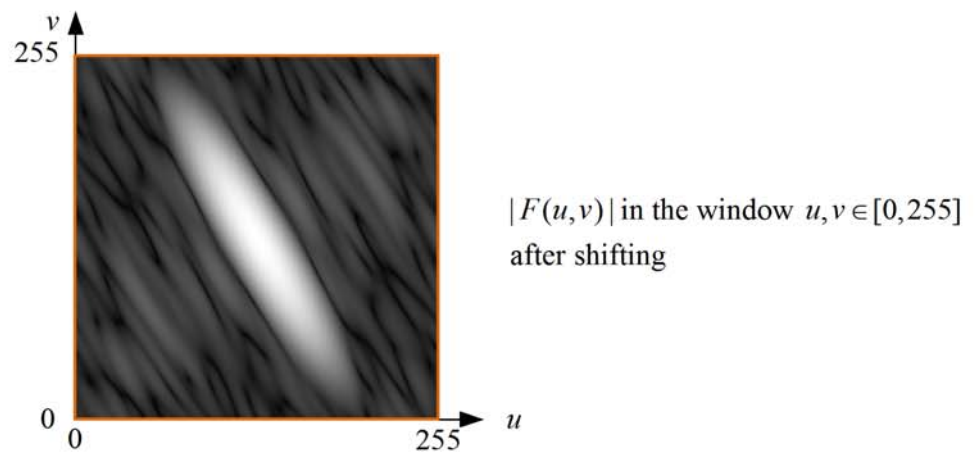
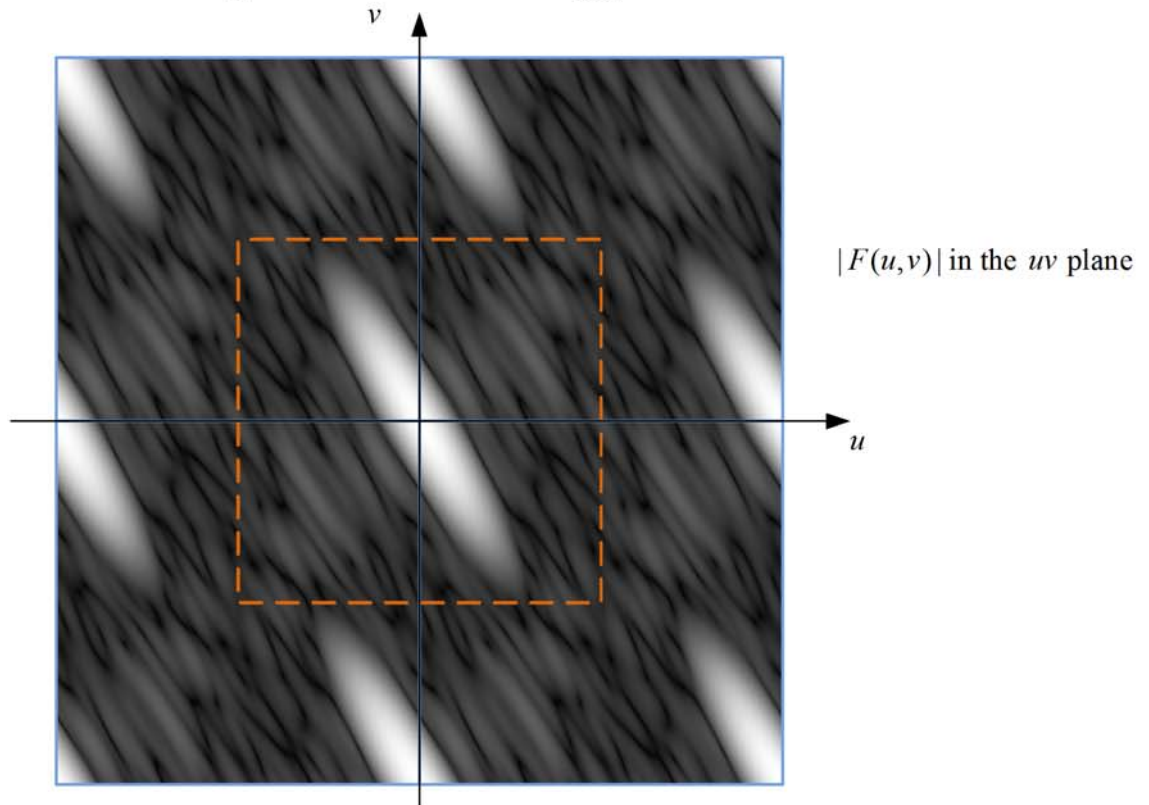
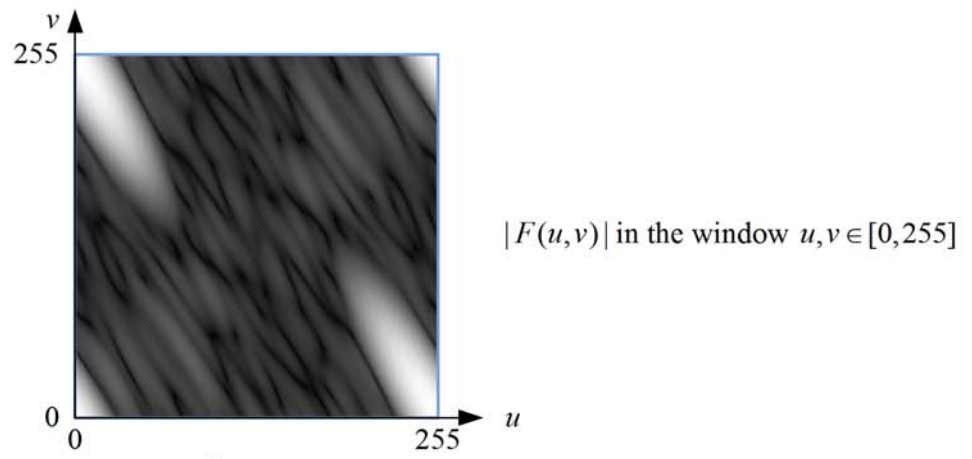


IMAGE SAMPLING

In digital image processing systems, one deals with arrays of numbers obtained by spatially sampling points of a physical image. Image samples nominally represent some physical measurements of a continuous image field, e.g., measurements of the image intensity or photographic density.

Given $f(x, y)$, which denotes a continuous, infinite-dimensional ideal image field representing the intensity of a physical image, the sampled image is given by

$$f_s(x, y) = f(x, y)s(x, y) \quad (28)$$

where $s(x, y)$, the sampling function, is represented by a 2D array of delta functions:

$$s(x, y) = \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} \delta(x - m\Delta x, y - n\Delta y) \quad (29)$$

Hence

$$f_s(x, y) = \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} f(m\Delta x, n\Delta y) \delta(x - m\Delta x, y - n\Delta y) \quad (30)$$

From Eq. (28), we use the convolution theorem to obtain the Fourier transform of $f_s(x, y)$:

$$F_s(u, v) = F(u, v) \star S(u, v) \quad (31)$$

where $S(u, v)$, the Fourier transform of $s(x, y)$, is given by

$$S(u, v) = \frac{1}{\Delta x \Delta y} \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} \delta(u - mu_s, v - nv_s) \quad (32)$$

where

$$u_s = 1/\Delta x, \quad v_s = 1/\Delta y$$

It can be shown that the Fourier transform of $f_s(x, y)$ is

$$F_s(u, v) = \frac{1}{\Delta x \Delta y} \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} F(u - mu_s, v - nv_s) \quad (33)$$

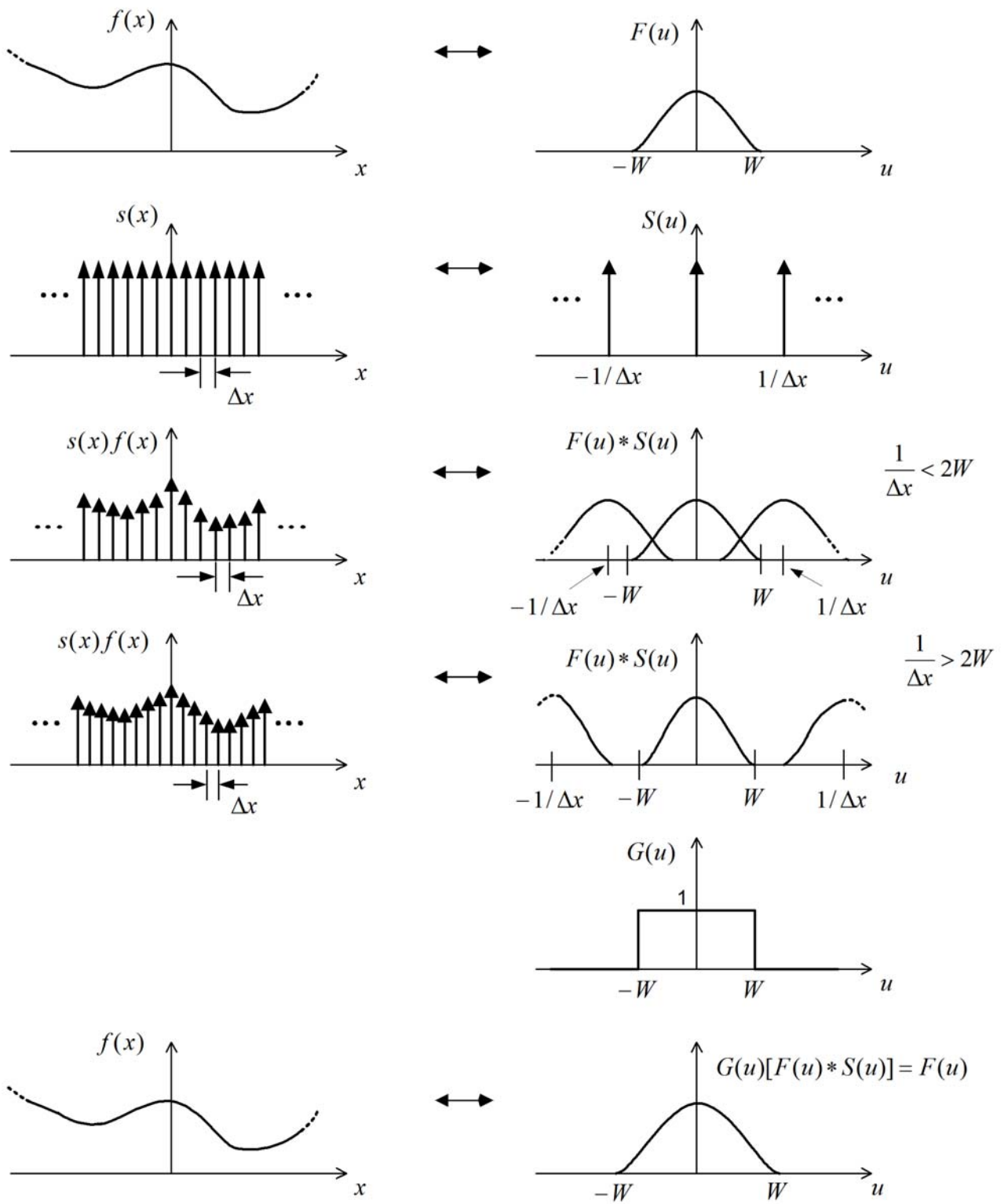


Illustration of sampling concepts

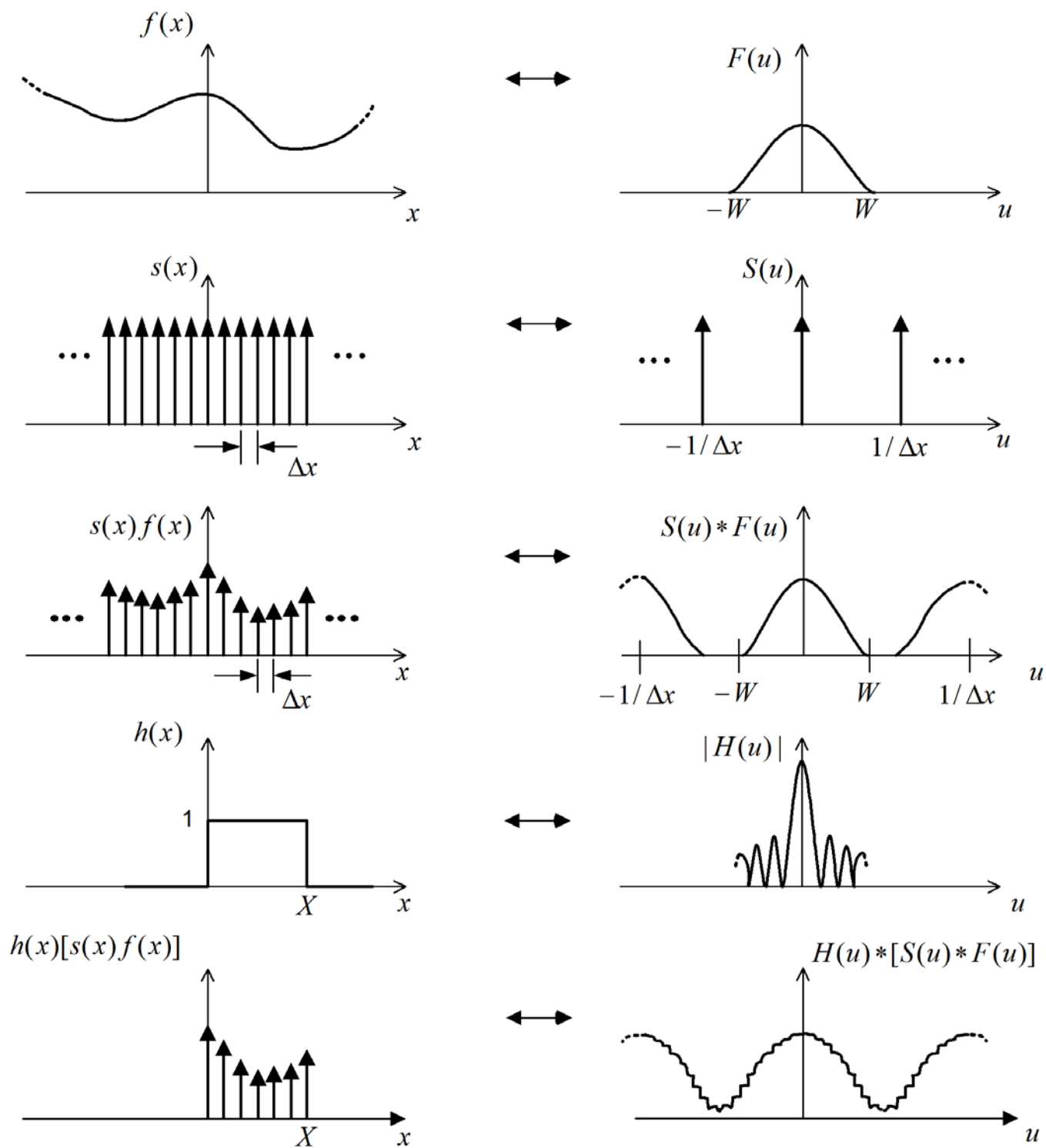
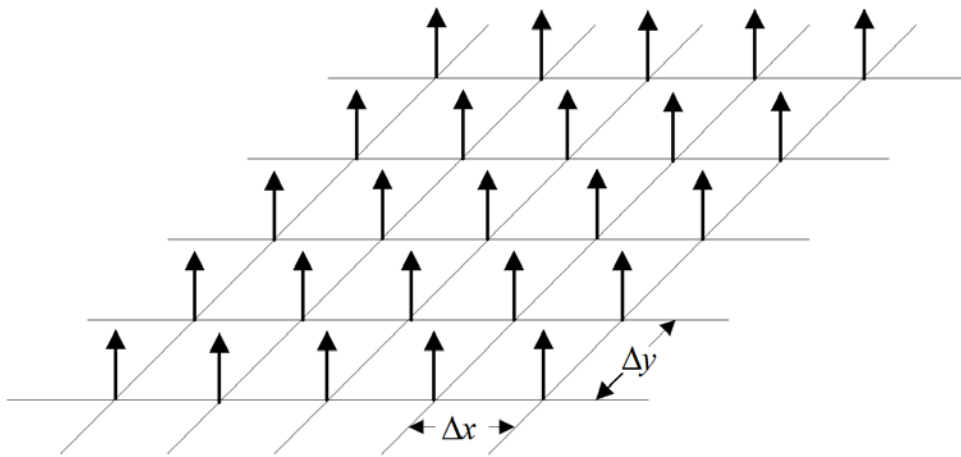
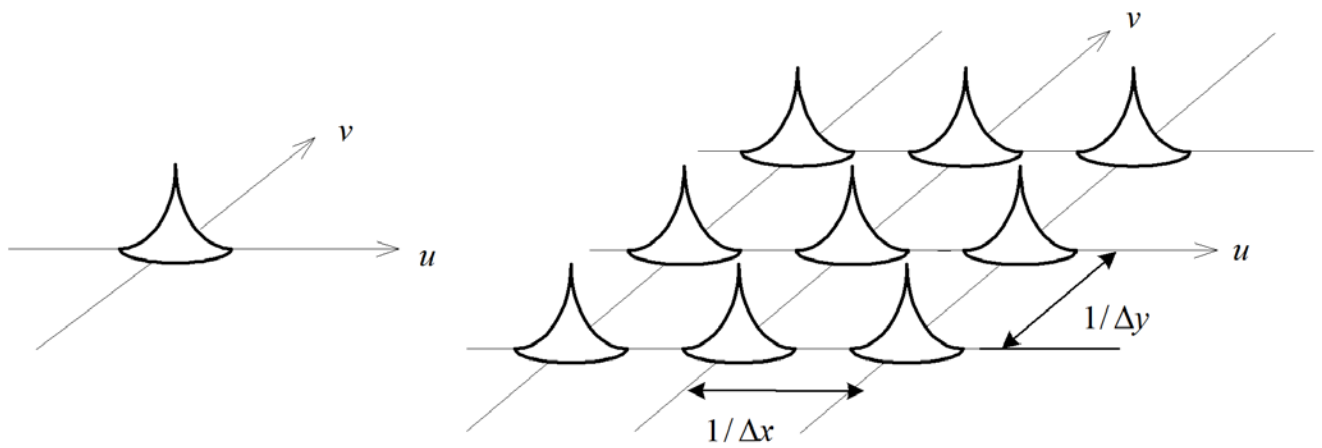


Illustration of finite-sampling concepts

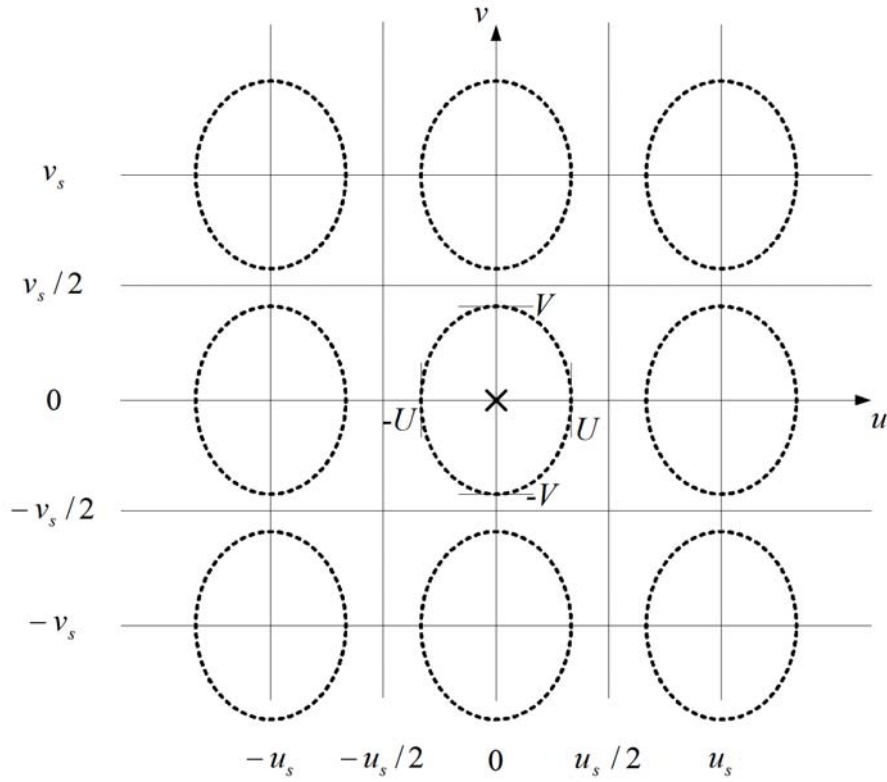


Dirac delta function sampling array.



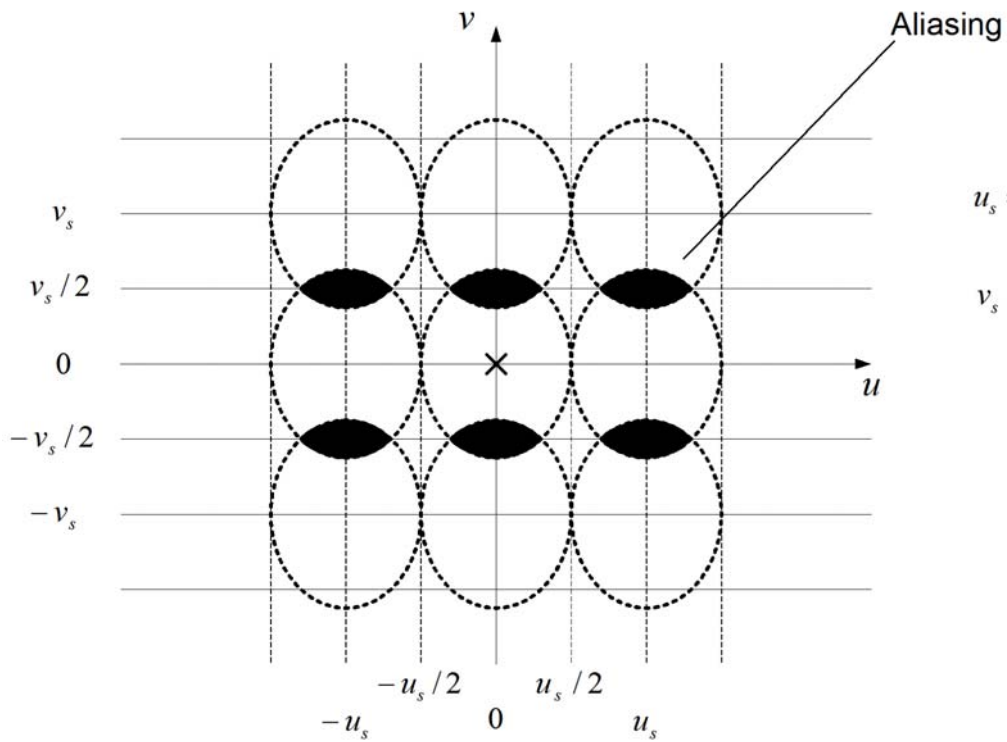
Spectrum of original image

Spectrum of sampled image



$$u_s = \frac{1}{\Delta x} \geq 2U$$

$$v_s = \frac{1}{\Delta y} \geq 2V$$



$$u_s = \frac{1}{\Delta x} = 2U$$

$$v_s = \frac{1}{\Delta y} < 2V$$

Examples

