

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.,

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

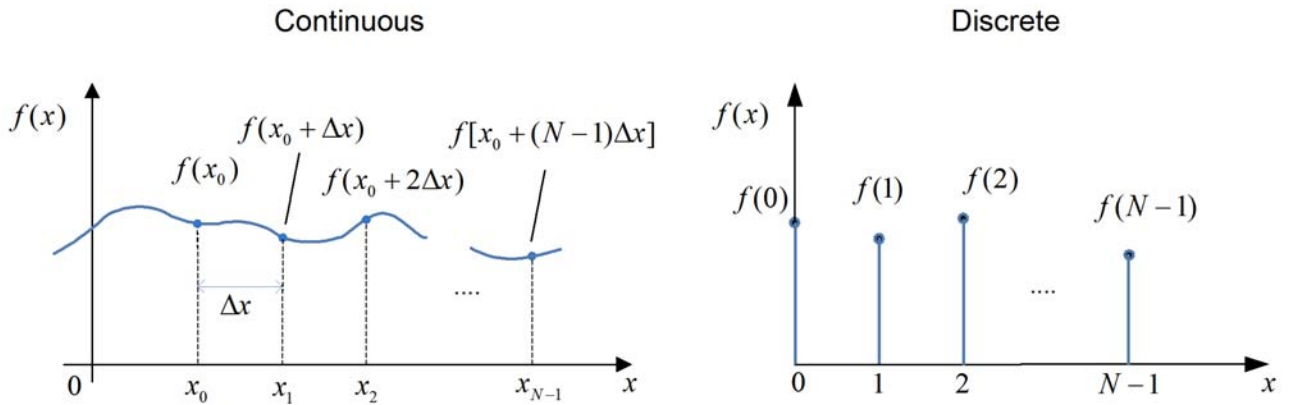
$$f(1) = f(x_0 + \Delta x)$$

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

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

...

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



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.,

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

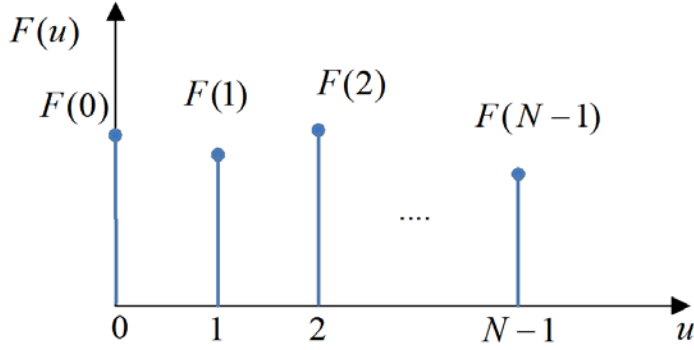
$$F(1) = F(\Delta u)$$

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

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

...

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



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.

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

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

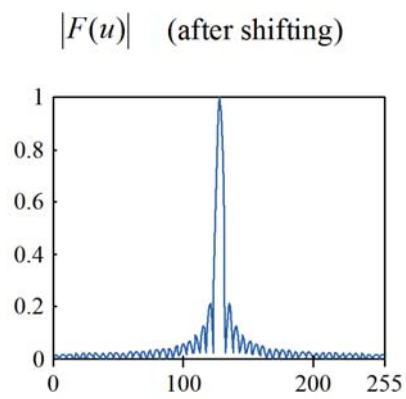
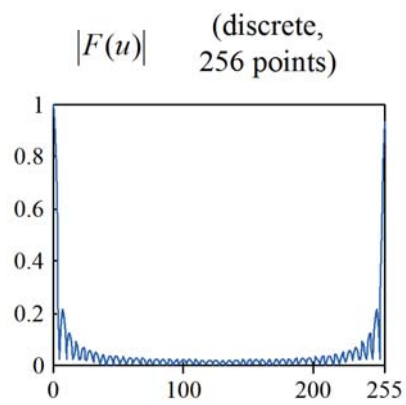
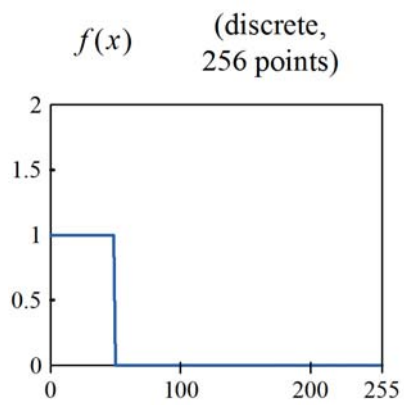
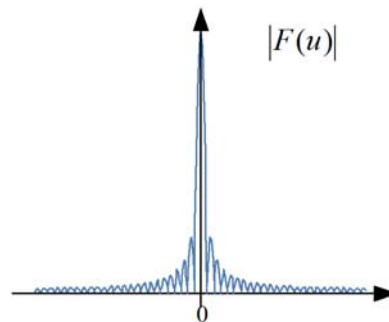
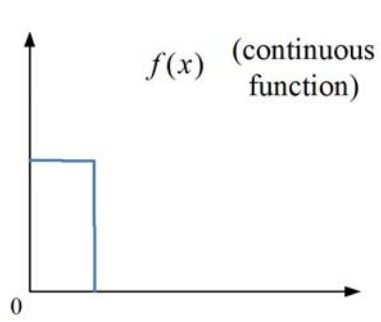
$$\text{Power spectrum: } P(u, v) = |F(u, v)|^2 = R^2(u, v) + I^2(u, v)$$

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 $MN \log_2 M \log_2 N$ operations. Suppose $M = N = 2^9$. Then

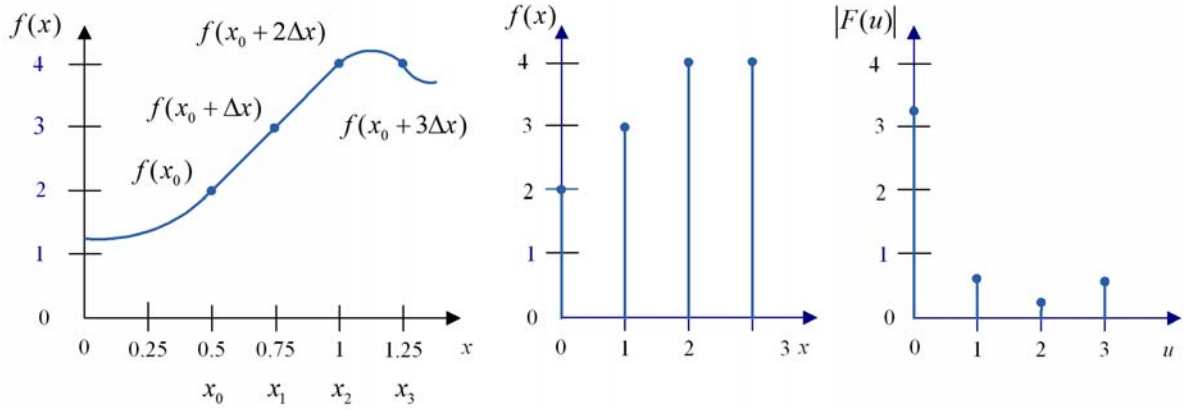
$$\text{Direct DFT: } M^2N^2 = 69 \times 10^9$$

$$\text{FFT: } MN \log_2 M \log_2 N = 21 \times 10^6$$

Example (1D DFT)



Example (1D DFT)



Sampling takes place at $x_0 = 0.5, x_1 = 0.75, x_2 = 1.0, x_3 = 1.25$, i.e., $f(0) = 2, f(1) = 3, f(2) = 4, f(3) = 4$.

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

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

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

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

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

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

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

Similarly,

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

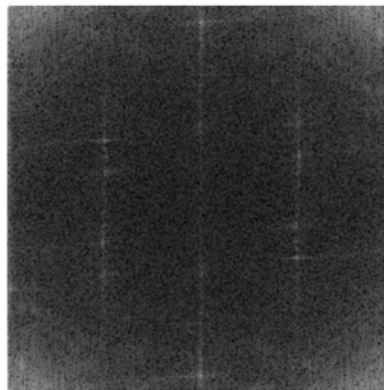
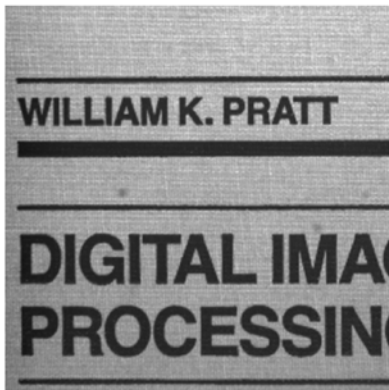
Fourier spectrum:

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

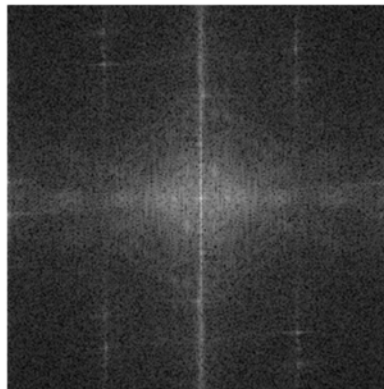
In this example,

$$N = 4, \quad \Delta x = 0.25, \quad \Delta u = 1/(N\Delta x) = 1$$

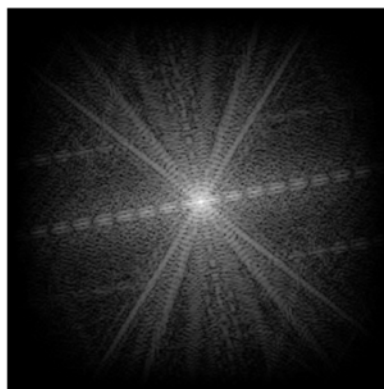
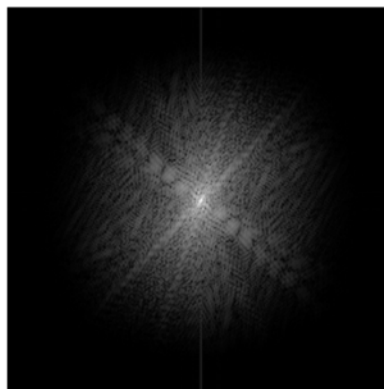
Examples of 2D DFT



Without shifting



With shifting



Some Properties of the 2D DFT

Separability

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

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

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

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

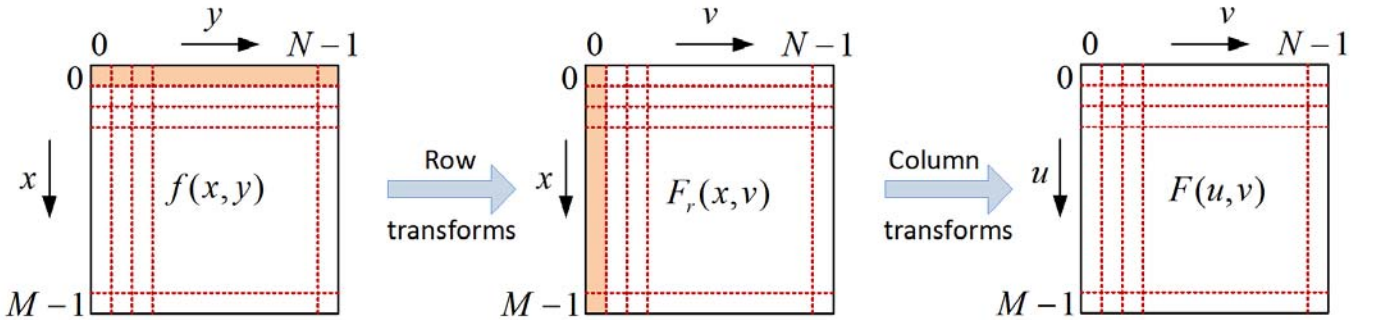
where

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

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).

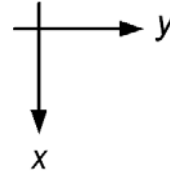
For each value of x , the expression inside the brackets is a 1D transform, with frequency values $v = 0, 1, \dots, N - 1$. Therefore the 2D function $F(x, v)$ is obtained by taking a transform along each row of $f(x, y)$. The desired result $F(u, v)$ is then obtained by taking a transform along each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

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

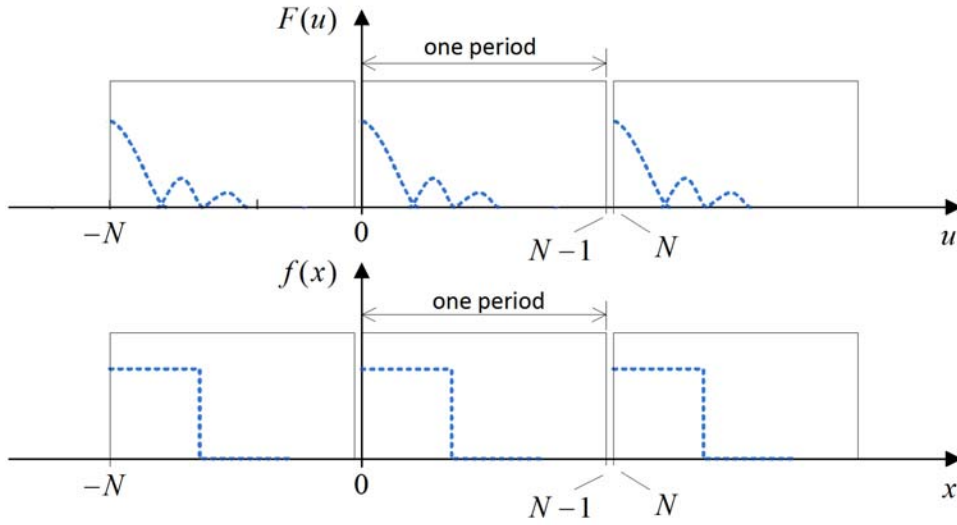
Periodicity and Conjugate Symmetry

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:

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

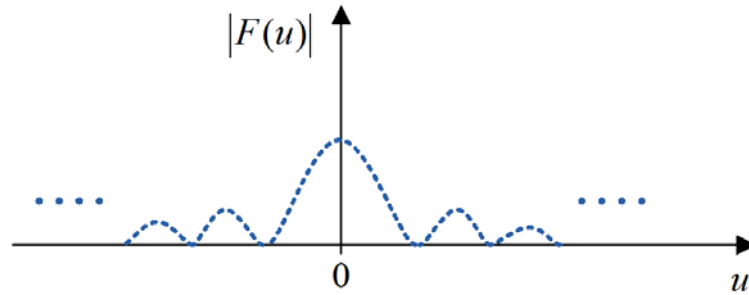


Furthermore, if $f(x)$ is real, the DFT also exhibits conjugate symmetry:

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

The DFT magnitude is then symmetrical about $u = 0$ since

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



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

$$|F(u)| = |F(-u)| \quad (\text{symmetry}) \quad (21)$$

$$= |F(-u + N)| \quad (\text{periodicity}) \quad (22)$$

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

$$|F(0)| = |F(N)|$$

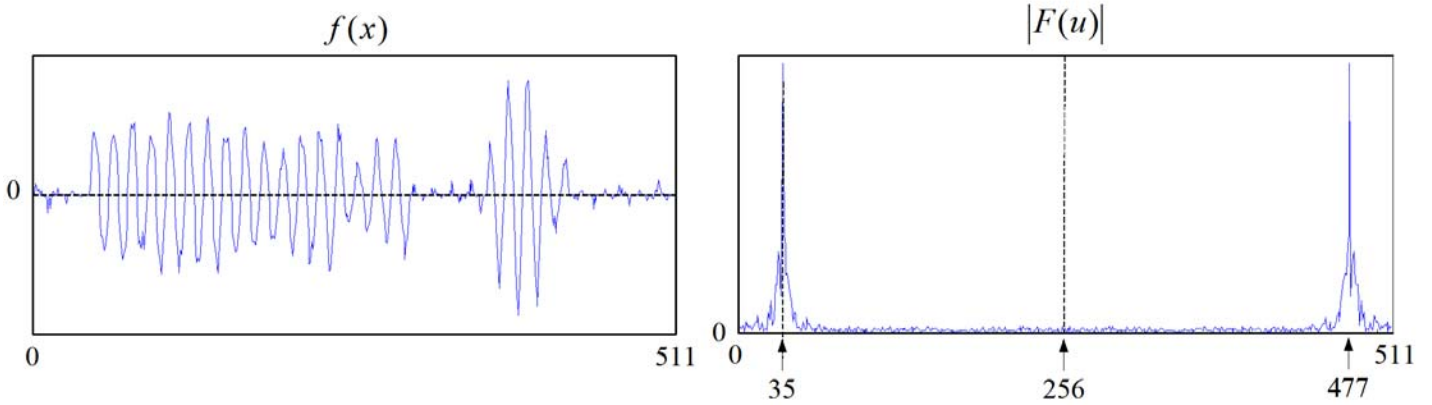
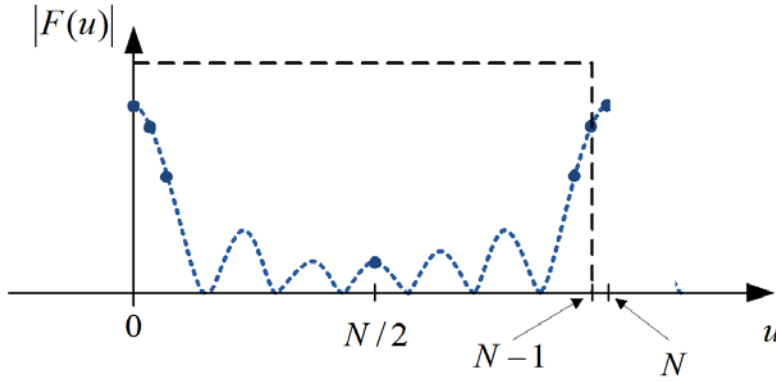
$$|F(1)| = |F(N - 1)|$$

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

...

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

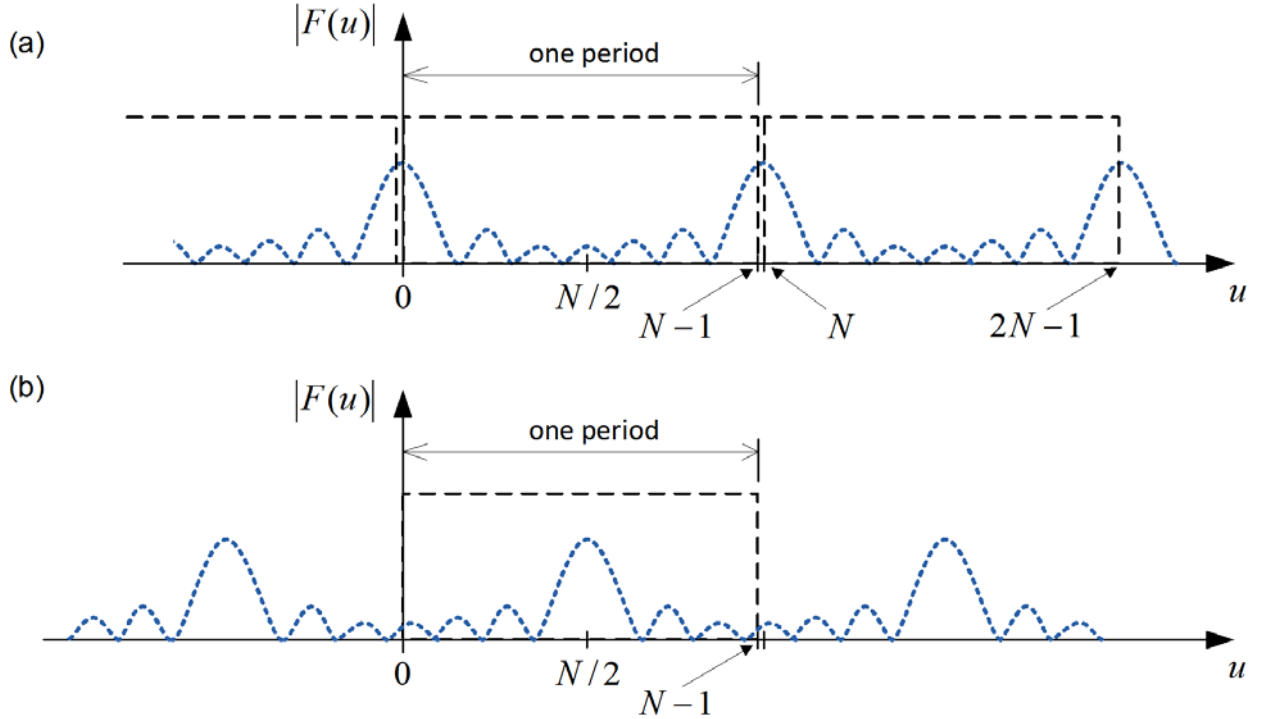
i.e., $|F(u)|$ is symmetrical about $u = N/2$.



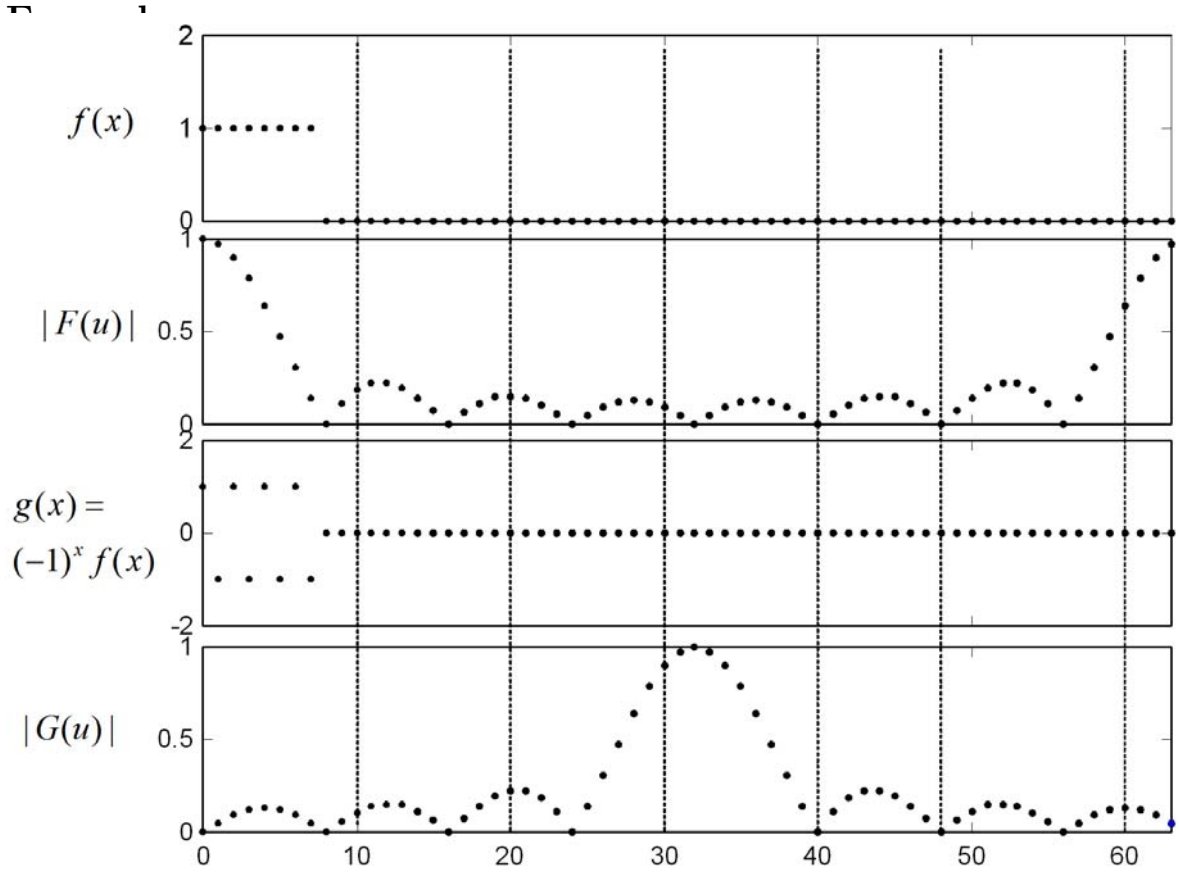
Cecause of the periodicity property, the values of $|F(u)|$ in this window are repeated along the u axis.

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,

$$F(u, v) = F(u + kN, v + lN) \quad k, l = 0, \pm 1, \pm 2, \dots \quad (23)$$

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

$$F(u, v) = F^*(-u, -v) \quad (24)$$

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

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

In the window $u, v = 0, 1, \dots, N - 1$, $|F(u, v)|$ is symmetrical about $(u, v) = (N/2, N/2)$, i.e.,

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

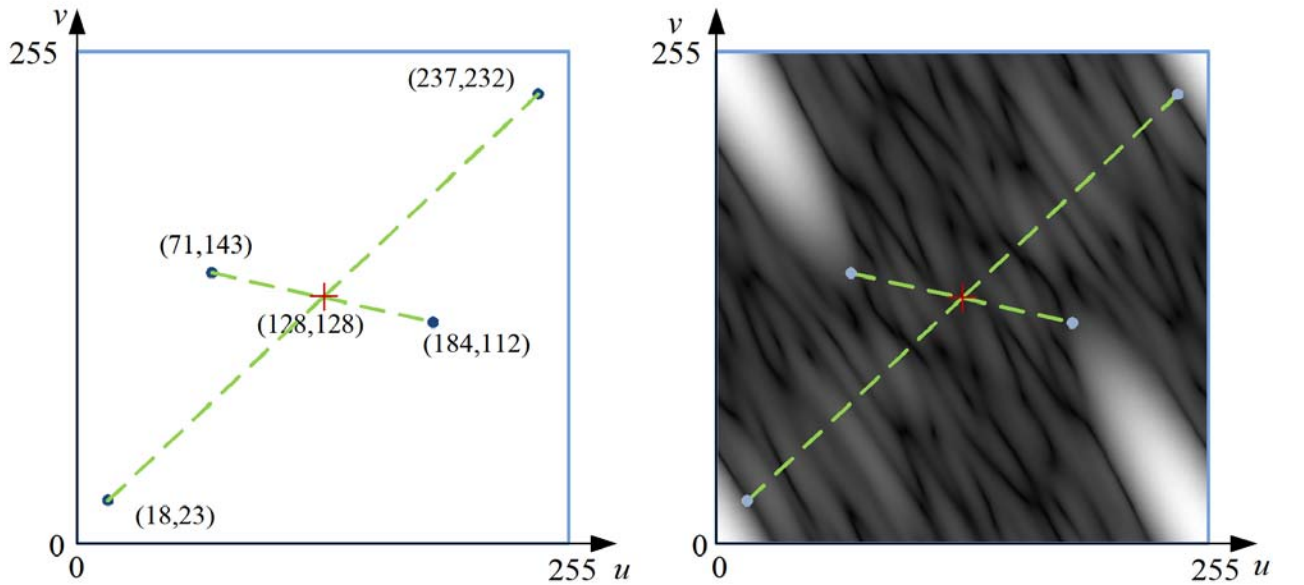
To move the origin of the transform to the point $(u, v) = (N/2, N/2)$, we multiply $f(x, y)$ by $(-1)^{x+y}$ prior to taking the transform, i.e.,

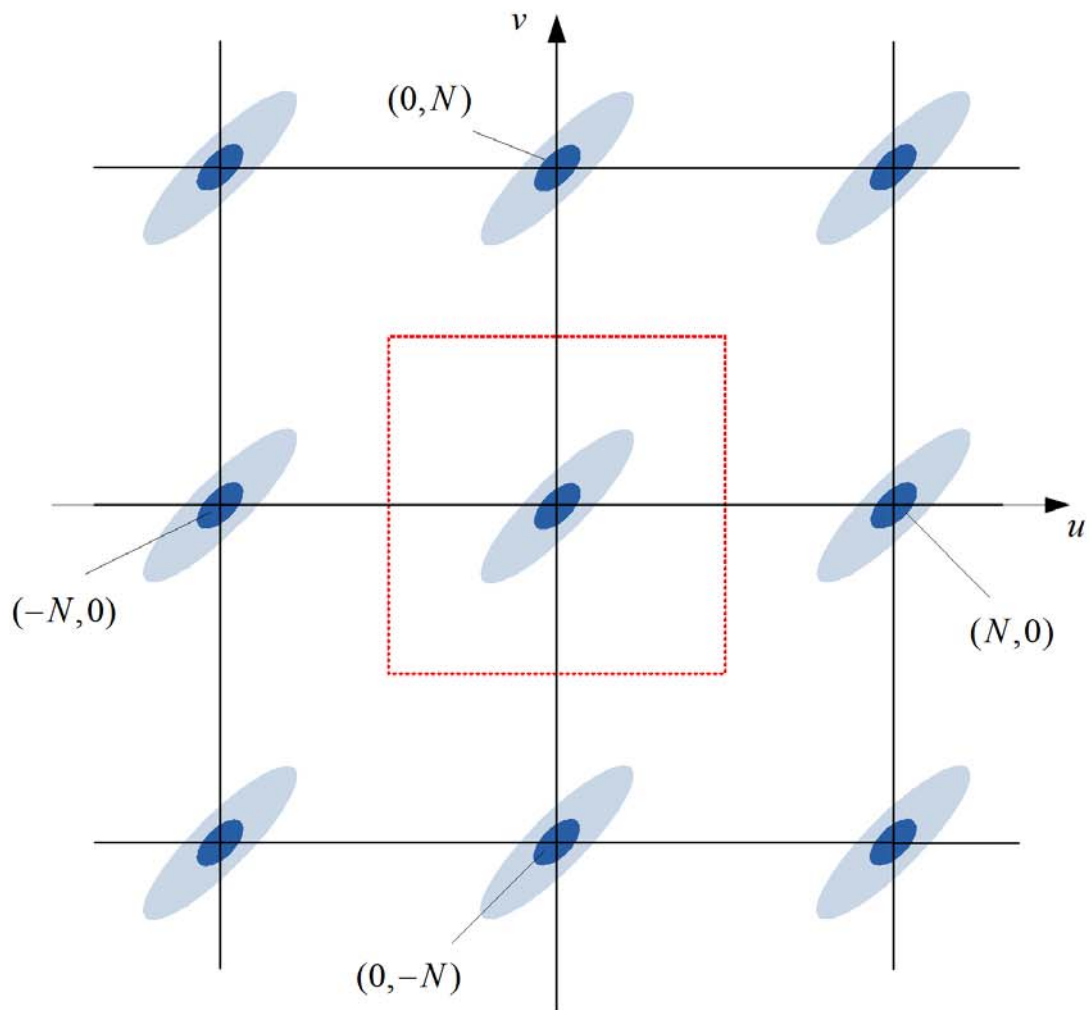
$$f'(x, y) = f(x, y) (-1)^{x+y}$$

Example

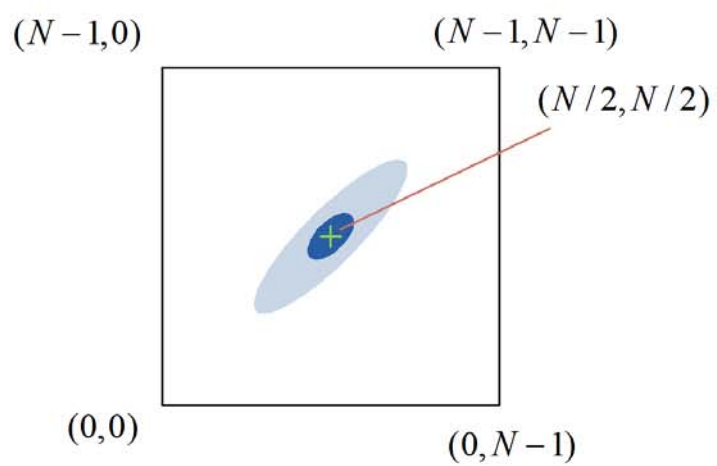
Consider a square image of size $N = 256$. $|F(u, v)|$ is symmetrical about $(u, v) = (128, 128)$.

$$\begin{aligned} |F(u, v)| &= |F(255 - u, 255 - v)| \\ |F(18, 23)| &= |F(237, 232)| \\ |F(71, 143)| &= |F(184, 112)| \end{aligned}$$





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



$|F(u, v)|$ centred at $(N/2, N/2)$

