	Spatial Domain [†]		Frequency Domain [†]
1)	f(x, y) real	\Leftrightarrow	$F^*(u,v) = F(-u,-v)$
2)	f(x, y) imaginary	\Leftrightarrow	$F^*(-u, -v) = -F(u, v)$
3)	f(x, y) real	\Leftrightarrow	R(u, v) even; $I(u, v)$ odd
4)	f(x, y) imaginary	\Leftrightarrow	R(u, v) odd; $I(u, v)$ even
5)	f(-x, -y) real	\Leftrightarrow	$F^*(u,v)$ complex
6)	f(-x, -y) complex	\Leftrightarrow	F(-u, -v) complex
7)	$f^*(x, y)$ complex	\Leftrightarrow	$F^*(-u-v)$ complex
8)	f(x, y) real and even	\Leftrightarrow	F(u, v) real and even
9)	f(x, y) real and odd	\Leftrightarrow	F(u, v) imaginary and odd
10)	f(x, y) imaginary and even	\Leftrightarrow	F(u, v) imaginary and even
11)	f(x, y) imaginary and odd	\Leftrightarrow	F(u, v) real and odd
12)	f(x, y) complex and even	\Leftrightarrow	F(u, v) complex and even
13)	f(x, y) complex and odd	\Leftrightarrow	F(u, v) complex and odd

¹³⁾ f(x, y) complex and odd $\Leftrightarrow F(u, v)$ complex and odd

Recall that x, y, u, and v are discrete (integer) variables, with x and u in the range [0, M-1], and y, and v in the range [0, M-1]. To say that a complex function is even means that its real and imaginary parts are even, and similarly for an odd complex function.

TABLE 4.1 Some symmetry properties of the 2-D DFT and its inverse. R(u, v) and I(u, v) are the real and imaginary parts of F(u, v), respectively. The term *complex* indicates that a function has nonzero real and imaginary parts.

Figura 5: Propiedades de simetría de la DFT 2D

Name	Expression(s)
1) Discrete Fourier transform (DFT) of $f(x, y)$	$F(u, v) = \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) e^{-j2\pi(ux/M + vy/N)}$
 Inverse discrete Fourier transform (IDFT) of F(u, v) 	$f(x,y) = \frac{1}{MN} \sum_{u=0}^{M-1} \sum_{v=0}^{N-1} F(u,v) e^{j2\pi(ux/M + vy/N)}$
3) Polar representation	$F(u,v) = F(u,v) e^{j\phi(u,v)}$
4) Spectrum	$ F(u, v) = [R^{2}(u, v) + I^{2}(u, v)]^{1/2}$ R = Real(F); I = Imag(F)
5) Phase angle	$\phi(u, v) = \tan^{-1} \left[\frac{I(u, v)}{R(u, v)} \right]$
6) Power spectrum	$P(u,v) = F(u,v) ^2$
7) Average value	$\overline{f}(x,y) = \frac{1}{MN} \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x,y) = \frac{1}{MN} F(0,0)$

TABLE 4.2 Summary of

Summary of DFT definitions and corresponding expressions.

(Continued)

Figura 7: Resumen: definiciones de la DFT 2D

Name	Expression(s)
8) Periodicity (k ₁ and k ₂ are integers)	$F(u, v) = F(u + k_1 M, v) = F(u, v + k_2 N)$ = $F(u + k_1 M, v + k_2 N)$
9) Convolution	$f(x, y) = f(x + k_1 M, y) = f(x, y + k_2 N)$ $= f(x + k_1 M, y + k_2 N)$ $f(x, y) \star h(x, y) = \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} f(m, n)h(x - m, y - n)$ $= \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} f(m, n)h(x - m, y - n)$
10) Correlation	$f(x, y) \approx h(x, y) = \sum_{m=0}^{\infty} \sum_{n=0}^{\infty} f^{*}(m, n) h(x + m, y + n)$
11) Separability	The 2-D DFT can be computed by computing 1-D DFT transforms along the rows (columns) of the image, followed by 1-D transforms along the columns (rows) of the result. See Section 4.11.1.
12) Obtaining the inverse Fourier transform using a forward transform algorithm.	$MNf^*(x, y) = \sum_{u=0}^{M-1} \sum_{v=0}^{N-1} F^*(u, v) e^{-j2\pi(ux/M+vy/N)}$ This equation indicates that inputting $F^*(u, v)$ into an algorithm that computes the forward transform (right side of above equation) yields $MNf^*(x, y)$. Taking the complex conjugate and dividing by MN gives the desired inverse. See Section 4.11.2.

TABLE 4.2 (Continued)

Figura 8: Resumen: definiciones de la DFT 2D

Name	DFT Pairs
Symmetry properties	See Table 4.1
2) Linearity	$af_1(x, y) + bf_2(x, y) \Leftrightarrow aF_1(u, v) + bF_2(u, v)$
3) Translation (general)	$f(x, y) e^{j2\pi(u_0x/M + v_0y/N)} \Leftrightarrow F(u - u_0, v - v_0)$ $f(x - x_0, y - y_0) \Leftrightarrow F(u, v) e^{-j2\pi(ux_0/M + vy_0/N)}$
4) Translation to center of the frequency rectangle, (M/2, N/2)	$f(x, y)(-1)^{x+y} \Leftrightarrow F(u - M/2, v - N/2)$ $f(x - M/2, y - N/2) \Leftrightarrow F(u, v)(-1)^{u+v}$
5) Rotation	$f(r, \theta + \theta_0) \Leftrightarrow F(\omega, \varphi + \theta_0)$ $x = r \cos \theta y = r \sin \theta u = \omega \cos \varphi v = \omega \sin \varphi$
6) Convolution theorem [†]	$\begin{split} f(x,y) \bigstar h(x,y) &\Leftrightarrow F(u,v) H(u,v) \\ f(x,y) h(x,y) &\Leftrightarrow F(u,v) \bigstar H(u,v) \end{split}$
	(Continued)

TABLE 4.3

Summary of DFT pairs. The closed-form expressions in 12 and 13 are valid only for continuous variables. They can be used with discrete variables by sampling the closed-form, continuous expressions.

Figura 9: Pares conocidos de la DFT 2D

	Name	DFT Pairs				
7)	Correlation theorem [†]	$f(x, y) \stackrel{.}{\Rightarrow} h(x, y) \Leftrightarrow F^{*}(u, v) H(u, v)$ $f^{*}(x, y)h(x, y) \Leftrightarrow F(u, v) \stackrel{.}{\Rightarrow} H(u, v)$				
8)	Discrete unit impulse	$\delta(x, y) \Leftrightarrow 1$				
9)	Rectangle	$\operatorname{rect}[a,b] \Leftrightarrow ab \frac{\sin(\pi ua)}{(\pi ua)} \frac{\sin(\pi vb)}{(\pi vb)} e^{-j\pi(ua+vb)}$				
10)	Sine	$\sin(2\pi u_0 x + 2\pi v_0 y) \Leftrightarrow$				
		$j\frac{1}{2}\Big[\delta(u+Mu_0,v+Nv_0)-\delta(u-Mu_0,v-Nv_0)\Big]$				
11)	Cosine	$\cos(2\pi u_0 x + 2\pi v_0 y) \Leftrightarrow$				
		$\frac{1}{2} \left[\delta(u + Mu_0, v + Nv_0) + \delta(u - Mu_0, v - Nv_0) \right]$				
denc	The following Fourier transform pairs are derivable only for continuous variables, denoted as before by t and z for spatial variables and by μ and ν for frequency variables. These results can be used for DFT work by sampling the continuous forms.					
12) Differentiation (The expressions on the right		$ \left(\frac{\partial}{\partial t} \right)^m \left(\frac{\partial}{\partial z} \right)^n f(t, z) \Leftrightarrow (j2\pi\mu)^m (j2\pi\nu)^n F(\mu, \nu) $ $ \frac{\partial^m f(t, z)}{\partial r^m} \Leftrightarrow (j2\pi\mu)^m F(\mu, \nu); \frac{\partial^n f(t, z)}{\partial r^n} \Leftrightarrow (j2\pi\nu)^n F(\mu, \nu) $				
assume that $\frac{\partial t^m}{\partial t^m} = (f2\pi\mu) P(\mu, \nu), \frac{\partial z^n}{\partial z^n} = (f2\pi\nu) P(\mu, \nu)$						
13) (Gaussian	$A2\pi\sigma^2 e^{-2\pi^2\sigma^2(t^2+z^2)} \Leftrightarrow Ae^{-(\mu^2+\nu^2)/2\sigma^2}$ (A is a constant)				

[†]Assumes that the functions have been extended by zero padding. Convolution and correlation are associative, commutative, and distributive,

TABLE 4.3 (Continued)

