

# 2D Discrete Fourier Transform (DFT)

# Outline

- Circular and linear convolutions
- 2D DFT
- 2D DCT
- Properties
- Other formulations
- Examples

# 2D Discrete Fourier Transform

- Fourier transform of a 2D signal defined over a discrete finite 2D grid of size  $M \times N$

or equivalently

- Fourier transform of a 2D set of samples forming a bidimensional sequence
- As in the 1D case, 2D-DFT, though a self-consistent transform, can be considered as a mean of calculating the transform of a 2D sampled signal defined over a discrete grid.
- The signal is periodized along both dimensions and the 2D-DFT can be regarded as a sampled version of the 2D DTFT

# 2D Discrete Fourier Transform (2D DFT)

- 2D Fourier (discrete time) Transform (DTFT) [Gonzalez]

$$F(u, v) = \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} f[m, n] e^{-j2\pi(um+vn)}$$

a-periodic signal  
periodic transform

- 2D Discrete Fourier Transform (DFT)

$$F[k, l] = \frac{1}{MN} \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} f[m, n] e^{-j2\pi\left(\frac{k}{M}m + \frac{l}{N}n\right)}$$

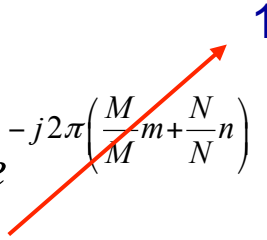
**periodized** signal  
periodic and **sampled**  
transform

2D DFT can be regarded as a sampled version of 2D DTFT.

## 2D DFT: Periodicity

- A [M,N] point DFT is periodic with period [M,N]
  - Proof

$$F[k, l] = \frac{1}{MN} \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} f[m, n] e^{-j2\pi \left( \frac{k}{M}m + \frac{l}{N}n \right)}$$

$$\begin{aligned} F[k + M, l + N] &= \frac{1}{MN} \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} f[m, n] e^{-j2\pi \left( \frac{k+M}{M}m + \frac{l+N}{N}n \right)} \\ &= \frac{1}{MN} \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} f[m, n] e^{-j2\pi \left( \frac{k}{M}m + \frac{l}{N}n \right)} e^{-j2\pi \left( \frac{M}{M}m + \frac{N}{N}n \right)} \\ &= F[k, l] \end{aligned}$$


(In what follows: spatial coordinates=k,l, frequency coordinates: u,v)

## 2D DFT: Periodicity

- Periodicity

$$F[u, v] = F[u + mM, v] = F[u, v + nN] = F[u + mM, v + nN]$$

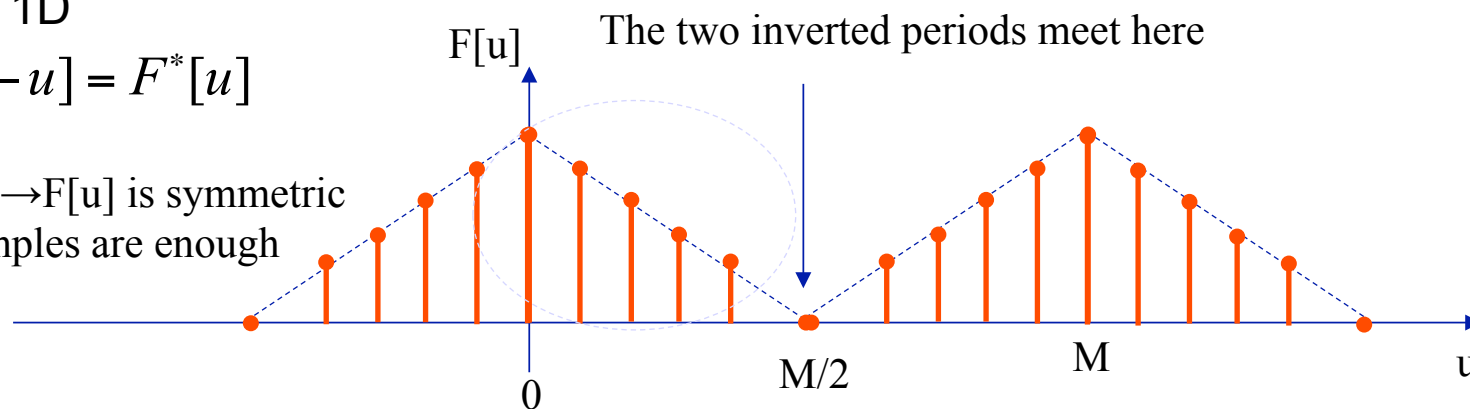
$$f[k, l] = f[k + mM, l] = f[k, l + nN] = f[k + mM, l + nN]$$

- This has important consequences on the implementation and energy compaction property

- 1D

$$F[N - u] = F^*[u]$$

$f[k]$  real  $\rightarrow$   $F[u]$  is symmetric  
 $M/2$  samples are enough



# Periodicity: 1D

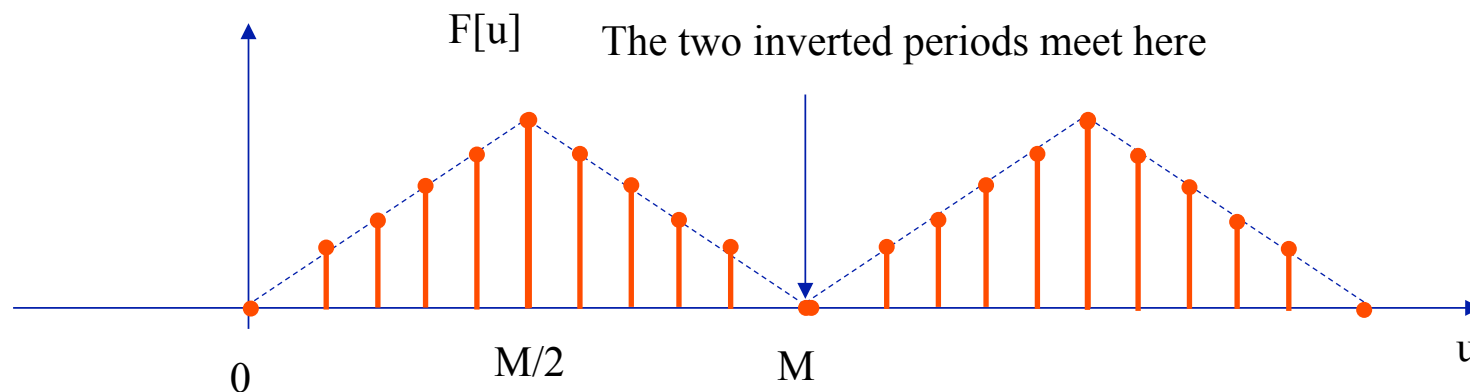
$$f[k] \leftrightarrow F[u]$$

$$f[k] e^{j2\pi \frac{u_0 k}{M}} \leftrightarrow F[u - u_0]$$

$$u_0 = \frac{M}{2} \rightarrow e^{j2\pi \frac{u_0 k}{M}} = e^{j2\pi \frac{Mk}{2M}} = e^{j\pi k} = (-1)^k$$

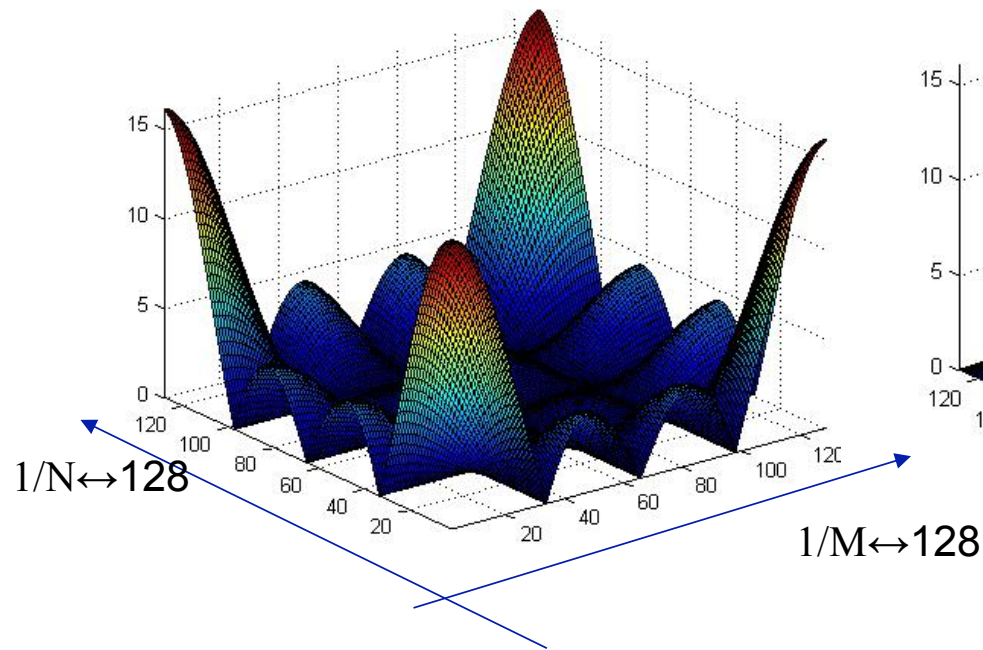
$$(-1)^k f[k] \leftrightarrow F[u - \frac{M}{2}]$$

changing the sign of every other sample of the DFT puts  $F[0]$  at the center of the interval  $[0, M]$

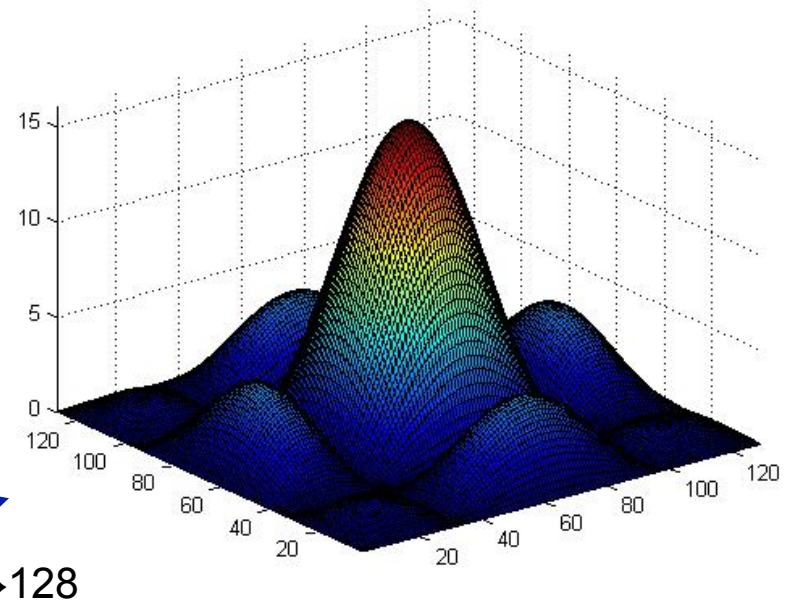


It is more practical to have one complete period positioned in  $[0, M-1]$

# Periodicity in 2D



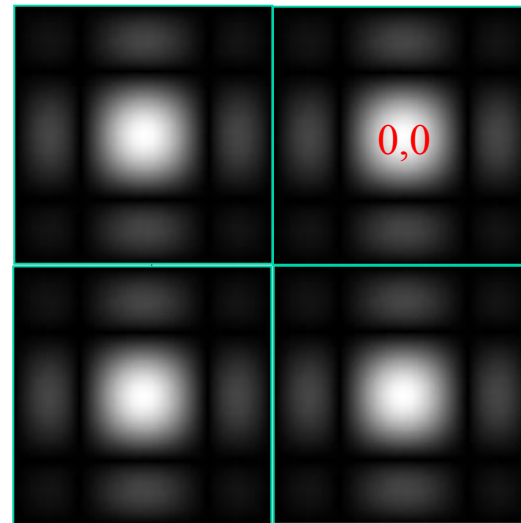
I 4 semiperiodi si incontrano ai vertici



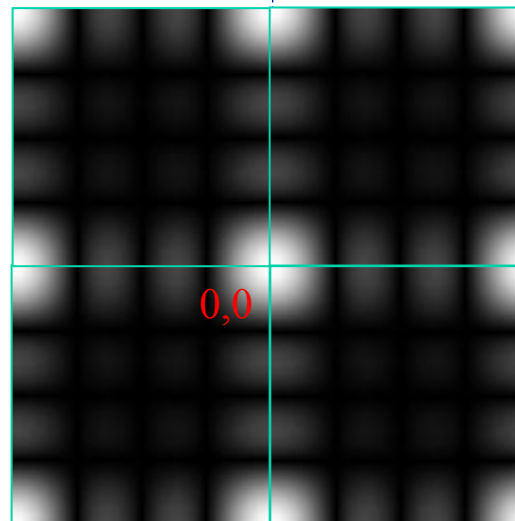
I 4 semiperiodi si incontrano al centro



# Periodicity



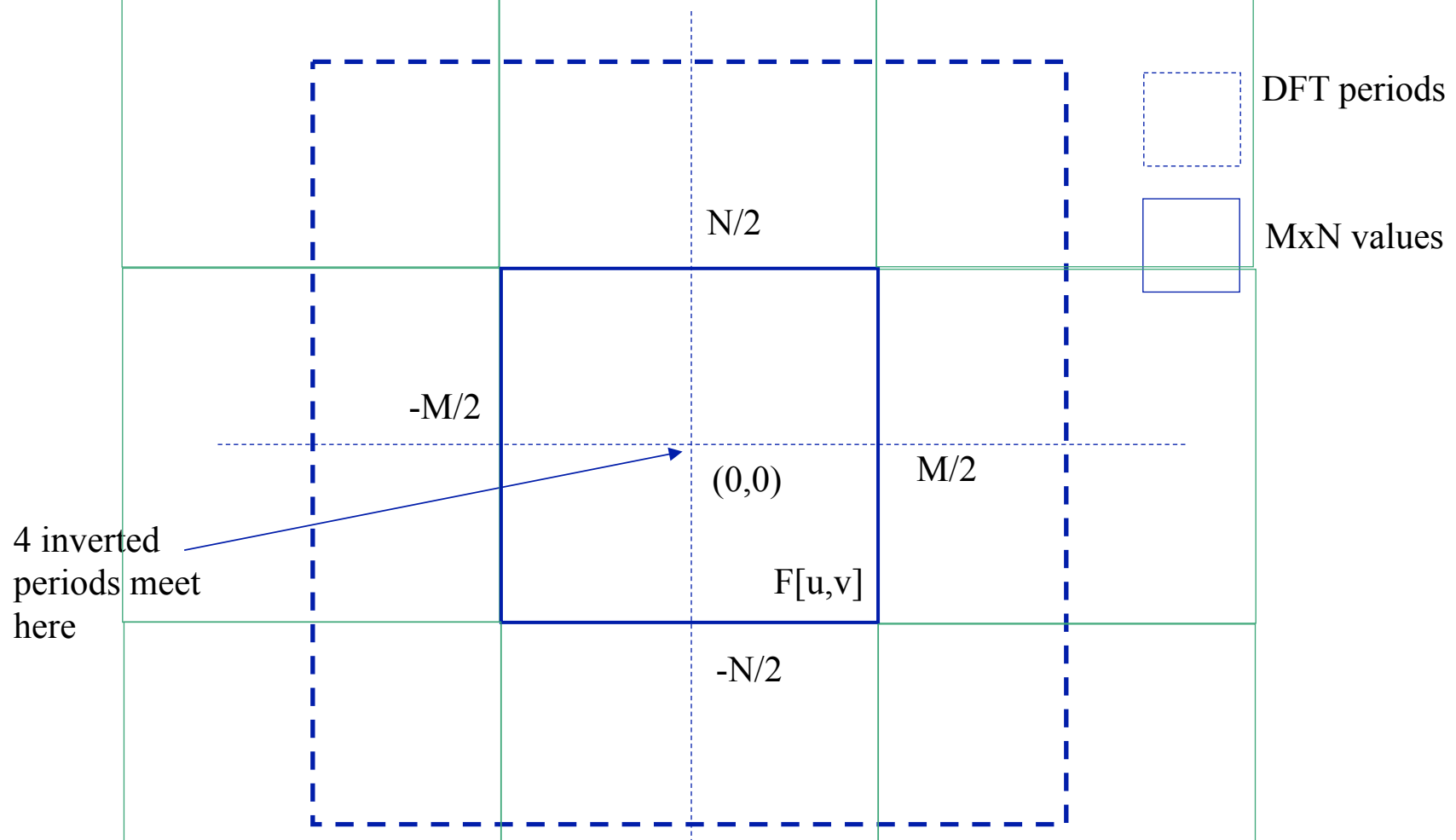
fftshift(fft2)



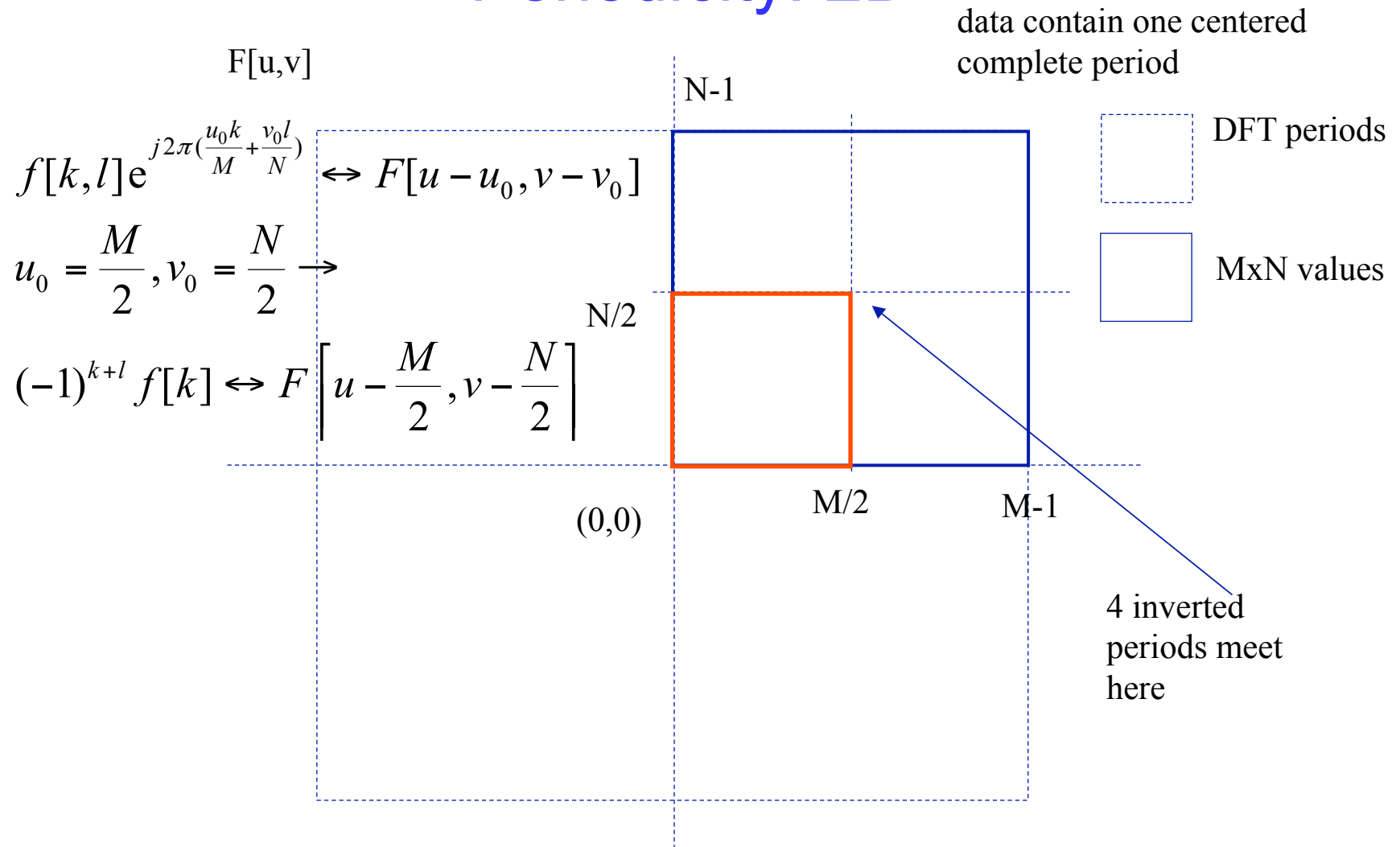
fft2

0,127=1/M,1/N

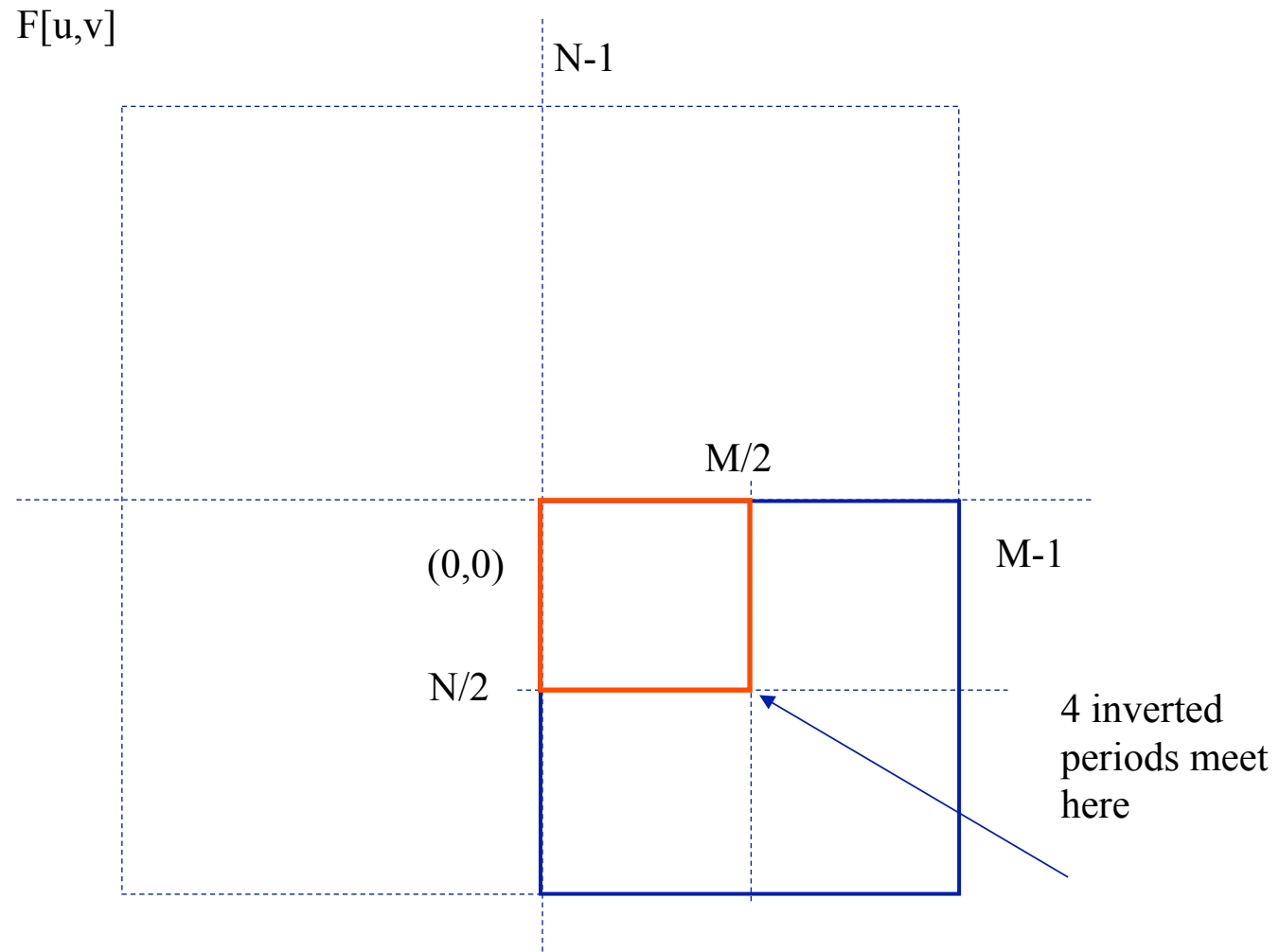
# Periodicity: 2D



# Periodicity: 2D



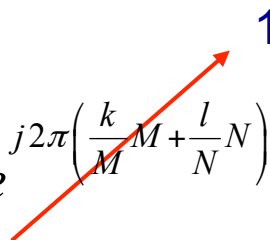
# Periodicity: 2D



# Periodicity in spatial domain

- $[M,N]$  point inverse DFT is periodic with period  $[M,N]$

$$f[m, n] = \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} F[k, l] e^{j2\pi \left( \frac{k}{M}m + \frac{l}{N}n \right)}$$

$$\begin{aligned} f[m + M, n + N] &= \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} F[k, l] e^{j2\pi \left( \frac{k}{M}(m+M) + \frac{l}{N}(n+N) \right)} \\ &= \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} F[k, l] e^{j2\pi \left( \frac{k}{M}m + \frac{l}{N}n \right)} e^{j2\pi \left( \frac{k}{M}M + \frac{l}{N}N \right)} \\ &= f[m, n] \end{aligned}$$


# Angle and phase spectra

$$F[u, v] = |F[u, v]| e^{j\Phi[u, v]}$$

$$|F[u, v]| = \left[ \operatorname{Re}\{F[u, v]\}^2 + \operatorname{Im}\{F[u, v]\}^2 \right]^{1/2} \quad \text{modulus (amplitude spectrum)}$$

$$\Phi[u, v] = \arctan \left[ \frac{\operatorname{Im}\{F[u, v]\}}{\operatorname{Re}\{F[u, v]\}} \right] \quad \text{phase}$$

$$P[u, v] = |F[u, v]|^2 \quad \text{power spectrum}$$

For a real function

$$F[-u, -v] = F^*[u, v] \quad \text{conjugate symmetric with respect to the origin}$$

$$|F[-u, -v]| = |F[u, v]|$$

$$\Phi[-u, -v] = -\Phi[u, v]$$

## Translation and rotation

$$f[k, l] e^{j2\pi\left(\frac{m}{M}k + \frac{n}{N}l\right)} \leftrightarrow F[u - m, v - l]$$

$$f[k - m, l - n] \leftrightarrow F[u, v] e^{-j2\pi\left(\frac{m}{M}k + \frac{n}{N}l\right)}$$

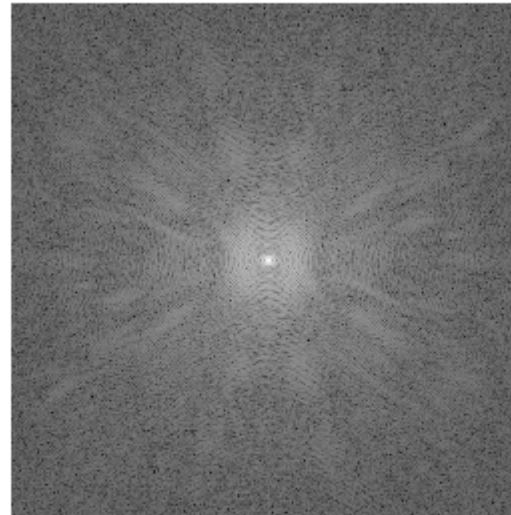
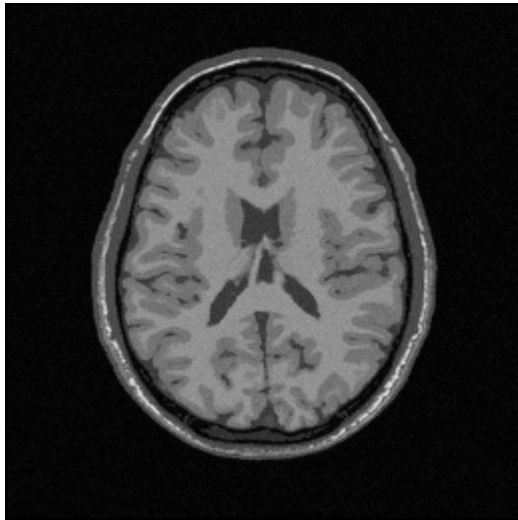
$$\begin{cases} k = r \cos \vartheta \\ l = r \sin \vartheta \end{cases} \quad \begin{cases} u = \omega \cos \varphi \\ l = \omega \sin \varphi \end{cases}$$
$$f[r, \vartheta + \vartheta_0] \leftrightarrow F[\omega, \varphi + \vartheta_0]$$

Rotations in spatial domain correspond equal rotations  
in Fourier domain

## mean value

$$F[0,0] = \frac{1}{NM} \sum_{n=0}^{N-1} \sum_{m=0}^{M-1} f[n,m]$$

DC coefficient





# Separability

- The discrete two-dimensional Fourier transform of an image array is defined in series form as

$$F[k, l] = \frac{1}{MN} \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} f[m, n] e^{-j2\pi \left( \frac{k}{M}m + \frac{l}{N}n \right)}$$

- inverse transform

$$f[m, n] = \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} F[k, l] e^{j2\pi \left( \frac{k}{M}m + \frac{l}{N}n \right)}$$

- Because the transform kernels are separable and symmetric, the two dimensional transforms can be computed as sequential row and column one-dimensional transforms.
- The basis functions of the transform are complex exponentials that may be decomposed into sine and cosine components.

## 2D DFT: summary

**TABLE 4.1**  
Summary of some important properties of the 2-D Fourier transform.

Property	Expression(s)
Fourier transform	$F(u, v) = \frac{1}{MN} \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) e^{-j2\pi(ux/M+vy/N)}$
Inverse Fourier transform	$f(x, y) = \sum_{u=0}^{M-1} \sum_{v=0}^{N-1} F(u, v) e^{j2\pi(ux/M+vy/N)}$
Polar representation	$F(u, v) =  F(u, v)  e^{-j\phi(u, v)}$
Spectrum	$ F(u, v)  = [R^2(u, v) + I^2(u, v)]^{1/2}, \quad R = \text{Real}(F) \text{ and } I = \text{Imag}(F)$
Phase angle	$\phi(u, v) = \tan^{-1} \left[ \frac{I(u, v)}{R(u, v)} \right]$
Power spectrum	$P(u, v) =  F(u, v) ^2$
Average value	$\bar{f}(x, y) = F(0, 0) = \frac{1}{MN} \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y)$
Translation	$f(x, y) e^{j2\pi(u_0 x/M + v_0 y/N)} \Leftrightarrow F(u - u_0, v - v_0)$ $f(x - x_0, y - y_0) \Leftrightarrow F(u, v) e^{-j2\pi(ux_0/M + vy_0/N)}$ <p>When <math>x_0 = u_0 = M/2</math> and <math>y_0 = v_0 = N/2</math>, then</p> $f(x, y) (-1)^{x+y} \Leftrightarrow F(u - M/2, v - N/2)$ $f(x - M/2, y - N/2) \Leftrightarrow F(u, v) (-1)^{u+v}$

## 2D DFT: summary

Conjugate symmetry	$F(u, v) = F^*(-u, -v)$ $ F(u, v)  =  F(-u, -v) $
Differentiation	$\frac{\partial^n f(x, y)}{\partial x^n} \Leftrightarrow (ju)^n F(u, v)$ $(-jx)^n f(x, y) \Leftrightarrow \frac{\partial^n F(u, v)}{\partial u^n}$
Laplacian	$\nabla^2 f(x, y) \Leftrightarrow -(u^2 + v^2)F(u, v)$
Distributivity	$\mathfrak{F}[f_1(x, y) + f_2(x, y)] = \mathfrak{F}[f_1(x, y)] + \mathfrak{F}[f_2(x, y)]$ $\mathfrak{F}[f_1(x, y) \cdot f_2(x, y)] \neq \mathfrak{F}[f_1(x, y)] \cdot \mathfrak{F}[f_2(x, y)]$
Scaling	$af(x, y) \Leftrightarrow aF(u, v), f(ax, by) \Leftrightarrow \frac{1}{ ab } F(u/a, v/b)$
Rotation	$x = r \cos \theta \quad y = r \sin \theta \quad u = \omega \cos \varphi \quad v = \omega \sin \varphi$ $f(r, \theta + \theta_0) \Leftrightarrow F(\omega, \varphi + \theta_0)$
Periodicity	$F(u, v) = F(u + M, v) = F(u, v + N) = F(u + M, v + N)$ $f(x, y) = f(x + M, y) = f(x, y + N) = f(x + M, y + N)$
Separability	<p>See Eqs. (4.6-14) and (4.6-15). Separability implies that we can compute the 2-D transform of an image by first computing 1-D transforms along each row of the image, and then computing a 1-D transform along each column of this intermediate result. The reverse, columns and then rows, yields the same result.</p>

## 2D DFT: summary

Property	Expression(s)
Computation of the inverse Fourier transform using a forward transform algorithm	$\frac{1}{MN} f^*(x, y) = \frac{1}{MN} \sum_{u=0}^{M-1} \sum_{v=0}^{N-1} F^*(u, v) e^{-j2\pi(ux/M + vy/N)}$ <p>This equation indicates that inputting the function <math>F^*(u, v)</math> into an algorithm designed to compute the forward transform (right side of the preceding equation) yields <math>f^*(x, y)/MN</math>. Taking the complex conjugate and multiplying this result by <math>MN</math> gives the desired inverse.</p>
Convolution <sup>†</sup>	$f(x, y) * h(x, y) = \frac{1}{MN} \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} f(m, n) h(x - m, y - n)$
Correlation <sup>†</sup>	$f(x, y) \circ h(x, y) = \frac{1}{MN} \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} f^*(m, n) h(x + m, y + n)$
Convolution theorem <sup>†</sup>	$f(x, y) * h(x, y) \Leftrightarrow F(u, v) H(u, v);$ $f(x, y) h(x, y) \Leftrightarrow F(u, v) * H(u, v)$
Correlation theorem <sup>†</sup>	$f(x, y) \circ h(x, y) \Leftrightarrow F^*(u, v) H(u, v);$ $f^*(x, y) h(x, y) \Leftrightarrow F(u, v) \circ H(u, v)$

## 2D DFT: summary

Some useful FT pairs:

*Impulse*  $\delta(x, y) \Leftrightarrow 1$

*Gaussian*  $A\sqrt{2\pi}\sigma e^{-2\pi^2\sigma^2(x^2+y^2)} \Leftrightarrow Ae^{-(u^2+v^2)/2\sigma^2}$

*Rectangle*  $\text{rect}[a, b] \Leftrightarrow ab \frac{\sin(\pi ua)}{(\pi ua)} \frac{\sin(\pi vb)}{(\pi vb)} e^{-j\pi(ua+vb)}$

*Cosine*  $\cos(2\pi u_0 x + 2\pi v_0 y) \Leftrightarrow$   
 $\frac{1}{2} [\delta(u + u_0, v + v_0) + \delta(u - u_0, v - v_0)]$

*Sine*  $\sin(2\pi u_0 x + 2\pi v_0 y) \Leftrightarrow$   
 $j \frac{1}{2} [\delta(u + u_0, v + v_0) - \delta(u - u_0, v - v_0)]$

<sup>†</sup> Assumes that functions have been extended by zero padding.

other formulations

# 2D Discrete Fourier Transform

- 2D Discrete Fourier Transform (DFT)

$$F[k, l] = \frac{1}{MN} \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} f[m, n] e^{-j2\pi \left( \frac{k}{M}m + \frac{l}{N}n \right)}$$

where  $l = 0, 1, \dots, N - 1$   
 $k = 0, 1, \dots, M - 1$

- Inverse DFT

$$f[m, n] = \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} F[k, l] e^{j2\pi \left( \frac{k}{M}m + \frac{l}{N}n \right)}$$

# 2D Discrete Fourier Transform

- It is also possible to define DFT as follows

$$F[k, l] = \frac{1}{\sqrt{MN}} \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} f[m, n] e^{-j2\pi \left( \frac{k}{M}m + \frac{l}{N}n \right)}$$

where  $k = 0, 1, \dots, M - 1$   
 $l = 0, 1, \dots, N - 1$

- Inverse DFT

$$f[m, n] = \frac{1}{\sqrt{MN}} \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} F[k, l] e^{j2\pi \left( \frac{k}{M}m + \frac{l}{N}n \right)}$$



# 2D Discrete Fourier Transform

- Or, as follows

$$F[k, l] = \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} f[m, n] e^{-j2\pi \left( \frac{k}{M}m + \frac{l}{N}n \right)}$$

where  $k = 0, 1, \dots, M-1$  and  $l = 0, 1, \dots, N-1$

- Inverse DFT

$$f[m, n] = \frac{1}{MN} \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} F[k, l] e^{j2\pi \left( \frac{k}{M}m + \frac{l}{N}n \right)}$$

## 2D DFT

- The discrete two-dimensional Fourier transform of an image array is defined in series form as

$$\mathcal{F}(u, v) = \frac{1}{N} \sum_{j=0}^{N-1} \sum_{k=0}^{N-1} F(j, k) \exp\left\{\frac{-2\pi i}{N}(uj + vk)\right\}$$

- inverse transform

$$F(j, k) = \frac{1}{N} \sum_{u=0}^{N-1} \sum_{v=0}^{N-1} \mathcal{F}(u, v) \exp\left\{\frac{2\pi i}{N}(uj + vk)\right\}$$

# 2D DCT

Discrete Cosine Transform

# 2D DCT

- based on most common form for 1D DCT

$$C(u) = \alpha(u) \sum_{x=0}^{N-1} f(x) \cos \left[ \frac{\pi(2x+1)u}{2N} \right], \quad u, x=0, 1, \dots, N-1$$

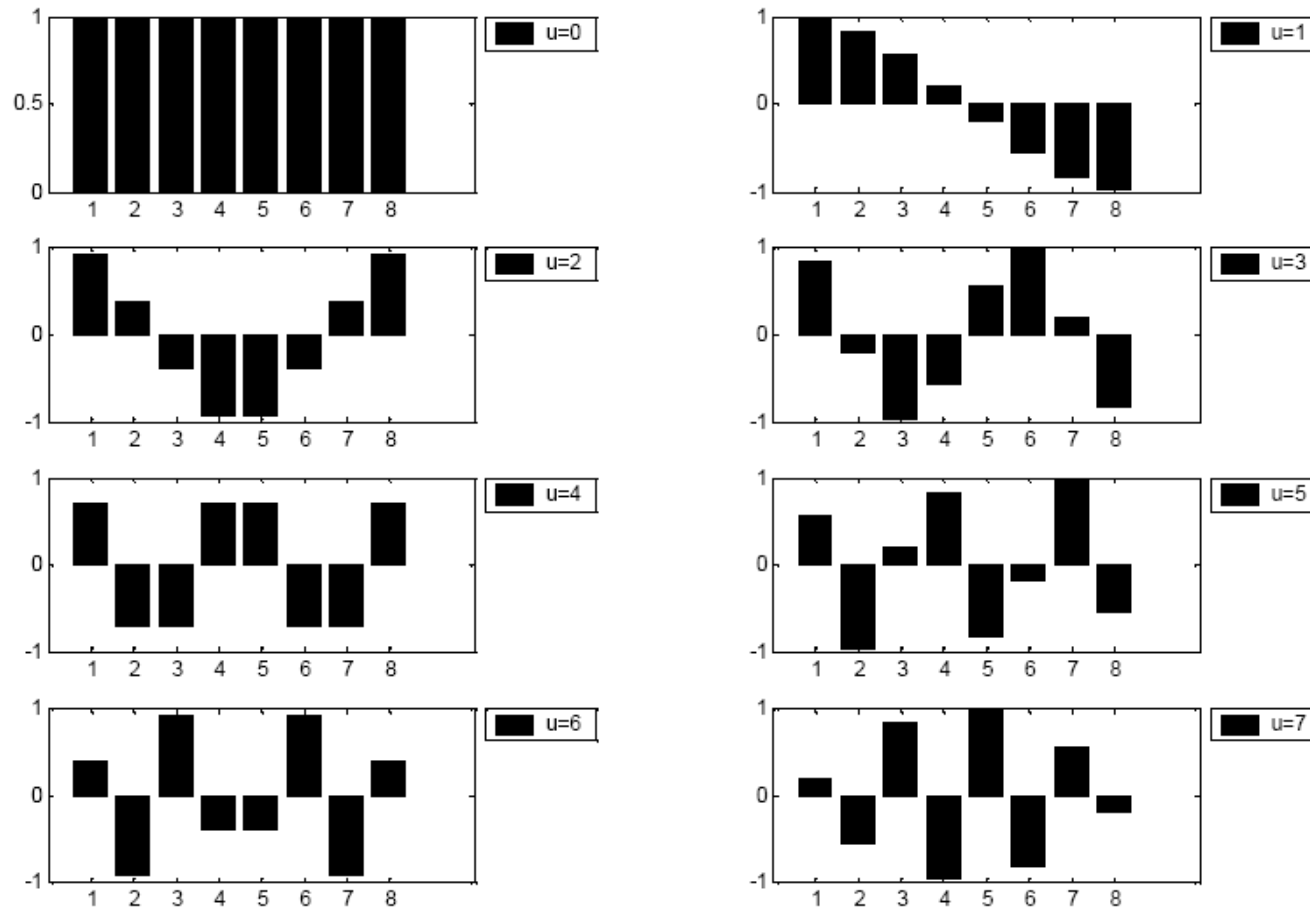
$$f(x) = \sum_{u=0}^{N-1} \alpha(u) C(u) \cos \left[ \frac{\pi(2x+1)u}{2N} \right],$$

$$\alpha(u) = \begin{cases} \sqrt{\frac{1}{N}} & \text{for } u = 0 \\ \sqrt{\frac{2}{N}} & \text{for } u \neq 0. \end{cases}$$

$$C(u=0) = \sqrt{\frac{1}{N}} \sum_{x=0}^{N-1} f(x). \quad \text{“mean” value}$$

# 1D basis functions

Figure 1



Cosine basis functions are orthogonal

## 2D DCT

- Corresponding 2D formulation

direct 
$$C(u, v) = \alpha(u)\alpha(v) \sum_{x=0}^{N-1} \sum_{y=0}^{N-1} f(x, y) \cos\left[\frac{\pi(2x+1)u}{2N}\right] \cos\left[\frac{\pi(2y+1)v}{2N}\right],$$

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

$$\alpha(u) = \begin{cases} \sqrt{\frac{1}{N}} & \text{for } u = 0 \\ \sqrt{\frac{2}{N}} & \text{for } u \neq 0. \end{cases}$$

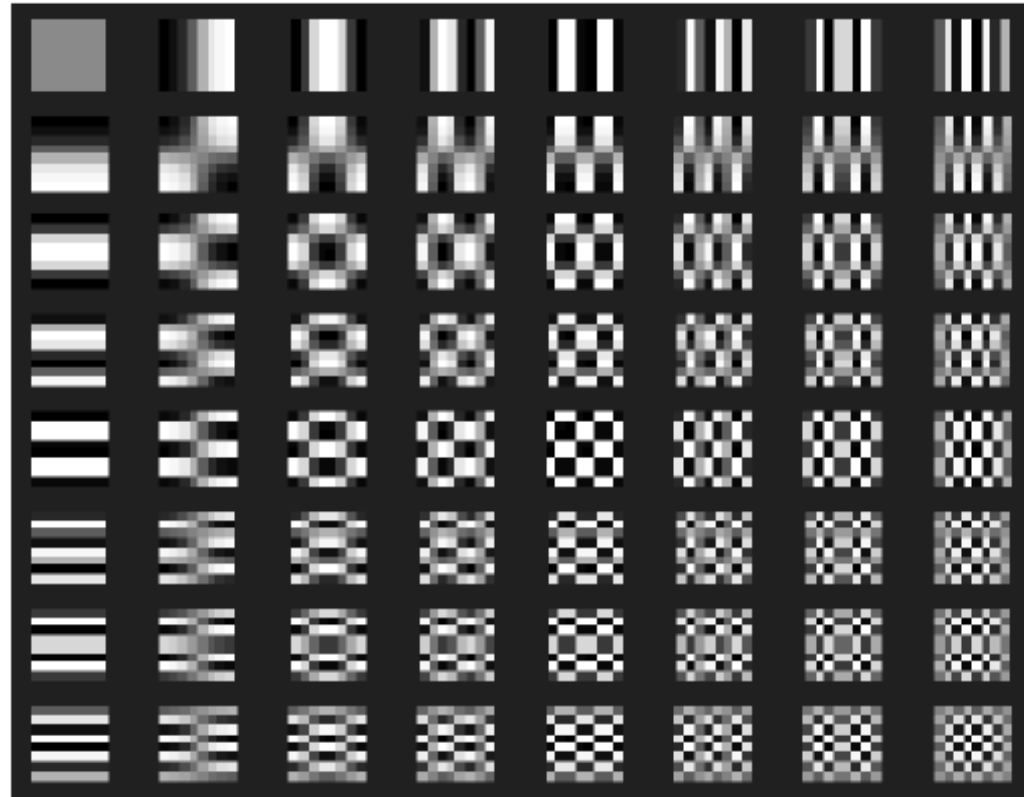
inverse 
$$f(x, y) = \sum_{u=0}^{N-1} \sum_{v=0}^{N-1} \alpha(u)\alpha(v) C(u, v) \cos\left[\frac{\pi(2x+1)u}{2N}\right] \cos\left[\frac{\pi(2y+1)v}{2N}\right],$$

## 2D basis functions

- The 2-D basis functions can be generated by multiplying the horizontally oriented 1-D basis functions (shown in Figure 1) with vertically oriented set of the same functions.
- The basis functions for  $N = 8$  are shown in Figure 2.
  - The basis functions exhibit a progressive increase in frequency both in the vertical and horizontal direction.
  - The top left basis function assumes a constant value and is referred to as the *DC coefficient*.

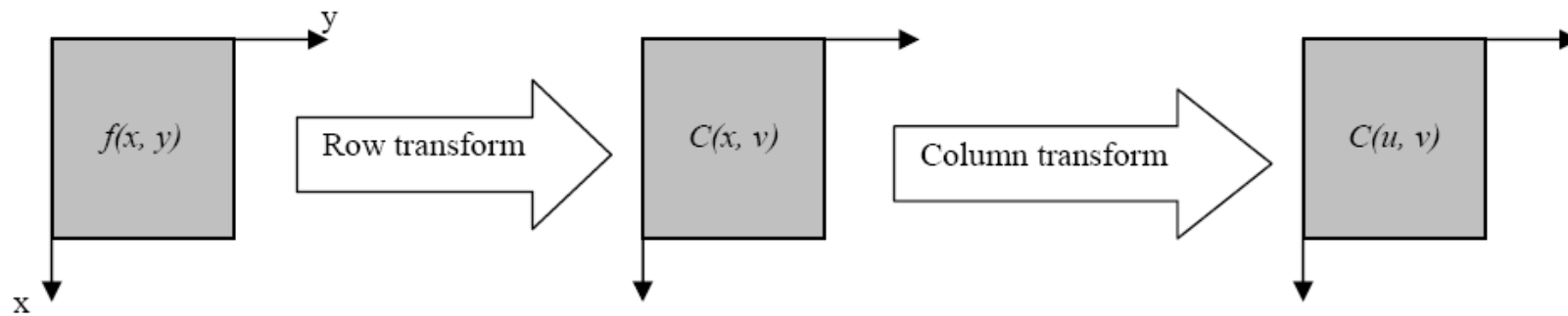
# 2D DCT basis functions

Figure 2





# Separability



The inverse of a multi-dimensional DCT is just a separable product of the inverse(s) of the corresponding one-dimensional DCT , e.g. the one-dimensional inverses applied along one dimension at a time

# Block-based implementation

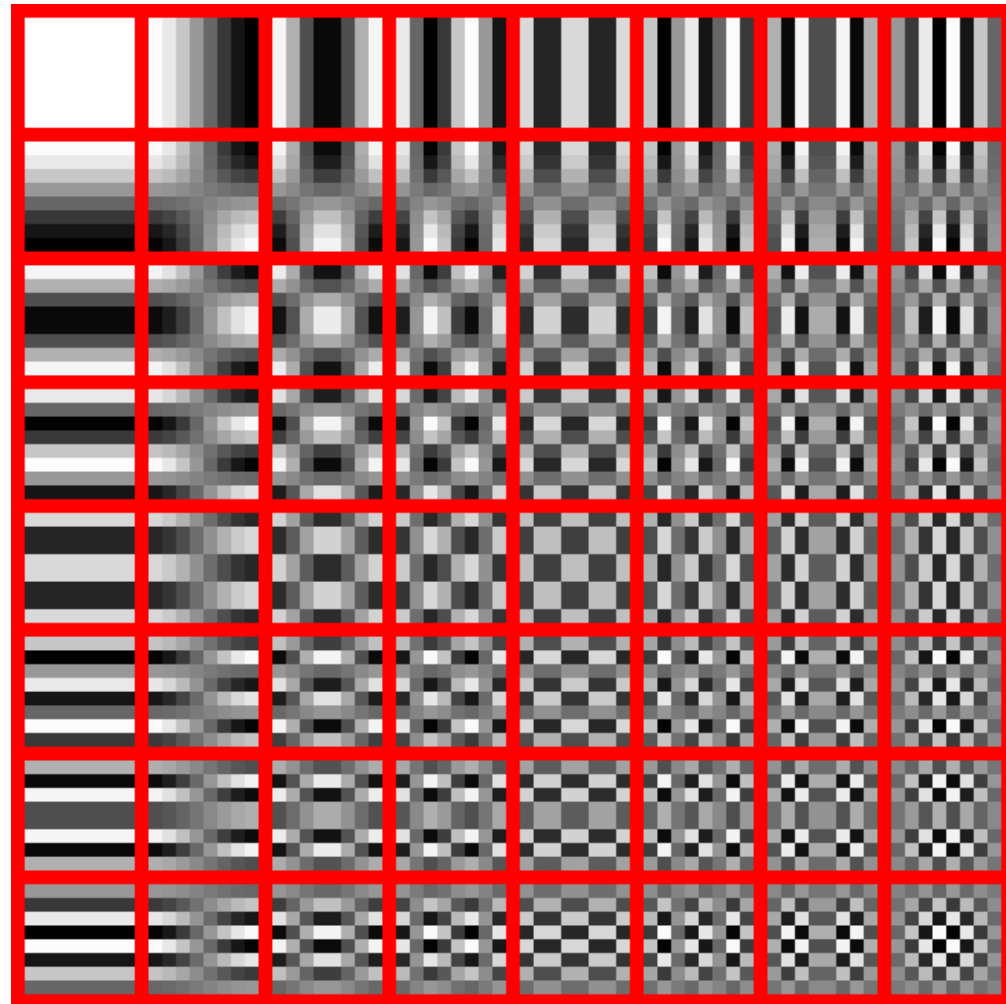
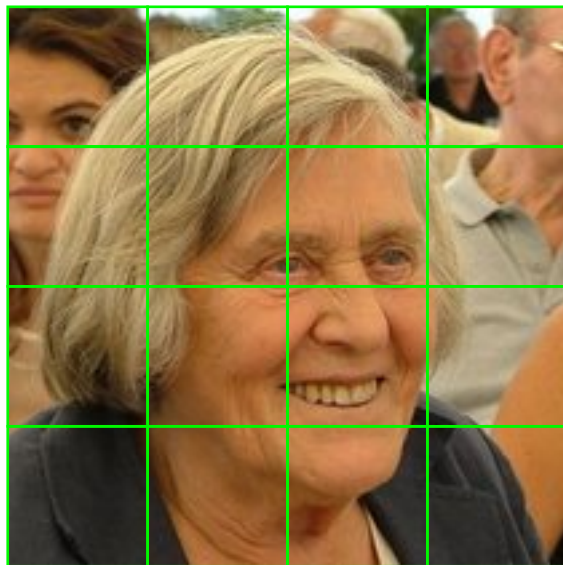
Basis function

*Block-based* transform

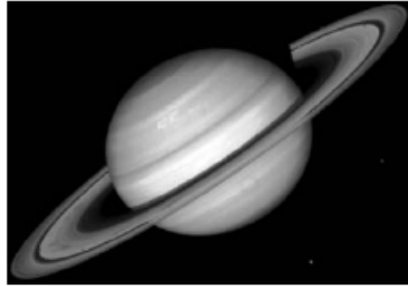
Block size

$N=M=8$

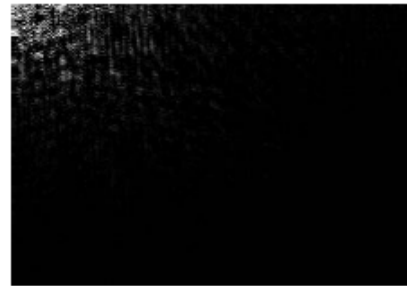
The source data (8x8) is transformed to a linear combination of these 64 frequency squares.



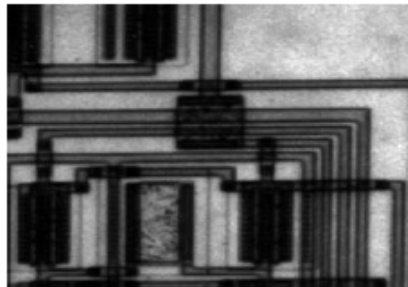
# Energy compaction



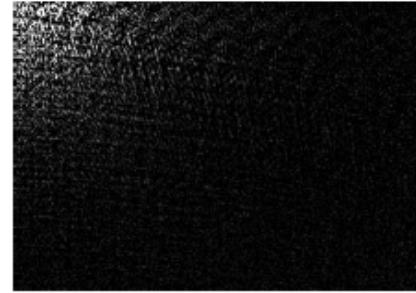
(a)



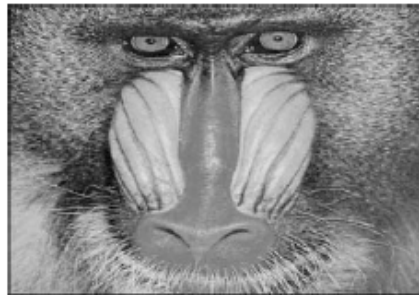
(b)



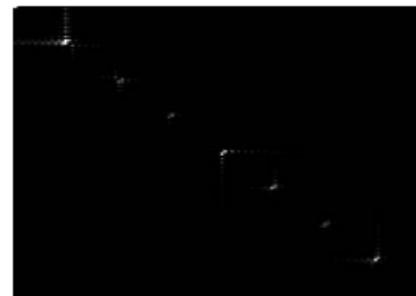
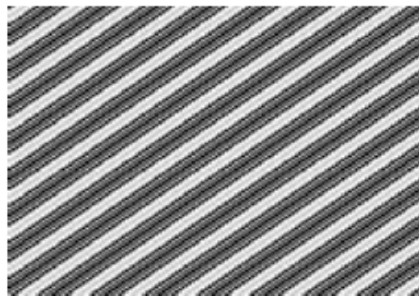
# Energy compaction



(d)



(e)



# Appendix

- Euler's formula

$$A(j, k; u, v) = \exp\left\{\frac{-2\pi i}{N}(uj + vk)\right\} = \cos\left\{\frac{2\pi}{N}(uj + vk)\right\} - i \sin\left\{\frac{2\pi}{N}(uj + vk)\right\}$$

$$B(j, k; u, v) = \exp\left\{\frac{2\pi i}{N}(uj + vk)\right\} = \cos\left\{\frac{2\pi}{N}(uj + vk)\right\} + i \sin\left\{\frac{2\pi}{N}(uj + vk)\right\}$$