

# 2D Discrete Fourier Transform (DFT)

# 2D Discrete Fourier Transform

- Fourier transform of a 2D signal defined over a discrete finite 2D grid of size  $N_x \times N_y$   
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 continuous Fourier transform

# 2D Discrete Fourier Transform

- 2D Fourier Transform

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

- 2D Discrete Fourier Transform (DFT)

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

*2D DFT is a sampled version of 2D FT.*

# 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

and  $l = 0, 1, \dots, N - 1$

$$k = 0, 1, \dots, M - 1$$

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

- 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 Discrete Fourier Transform

**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 Discrete Fourier Transform

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	$\Im[f_1(x, y) + f_2(x, y)] = \Im[f_1(x, y)] + \Im[f_2(x, y)]$ $\Im[f_1(x, y) \cdot f_2(x, y)] \neq \Im[f_1(x, y)] \cdot \Im[f_2(x, y)]$
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	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.

# 2D Discrete Fourier Transform

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 Discrete Fourier Transform

Some useful FT pairs:

$$\text{Impulse} \quad \delta(x, y) \Leftrightarrow 1$$

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

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

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

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

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

# Periodicity

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

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

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

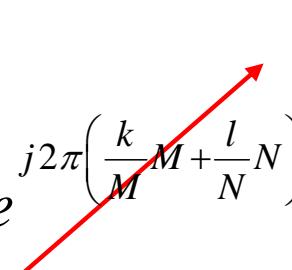
1

$$= F[k,l]$$

# Periodicity

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

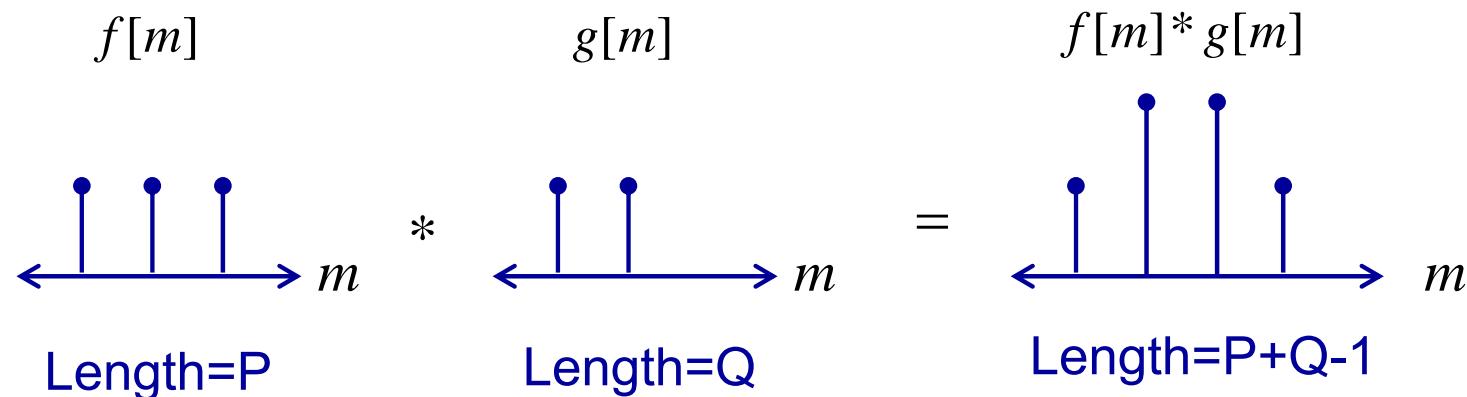
$$f[m, n] = \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} F[k, l] e^{j2\pi \left( \frac{k}{M}m + \frac{l}{N}n \right)}$$

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


# Convolution

- Be careful about the convolution property!

$$f[m] * g[m] \Leftrightarrow F[k]G[k]$$

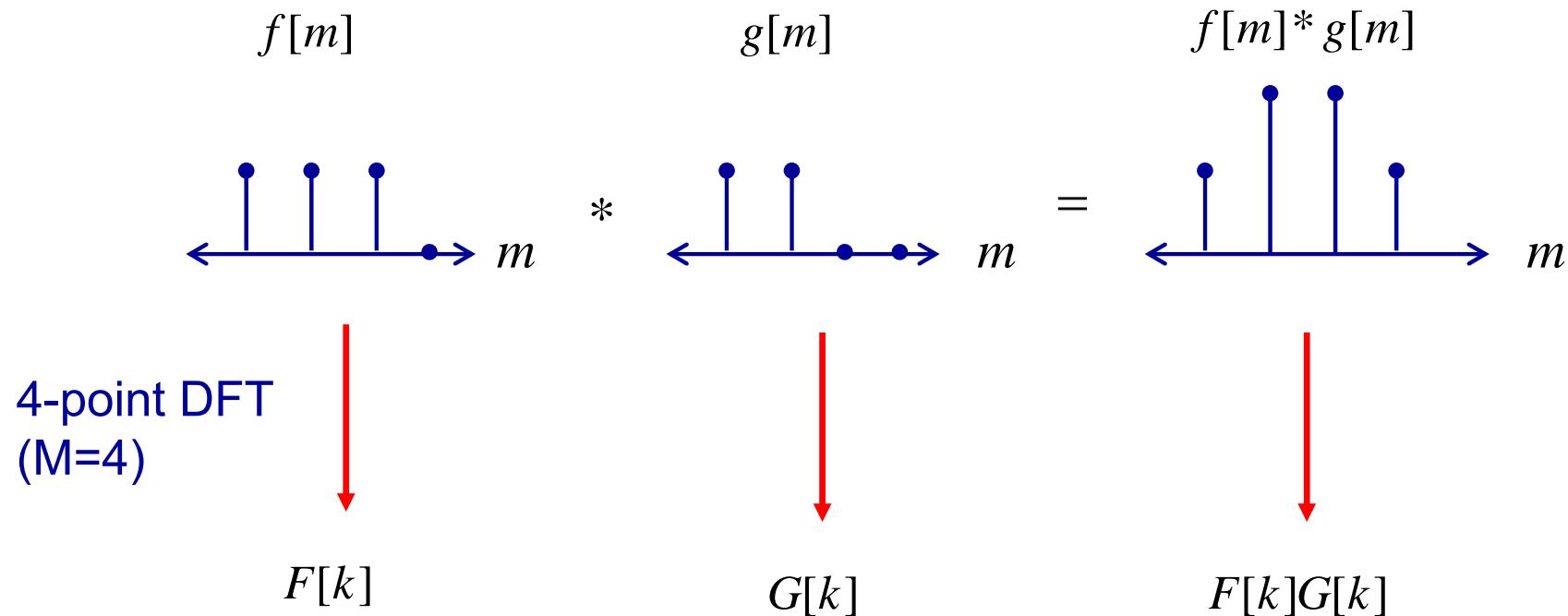


For the convolution property to hold, M must be *greater than or equal* to P+Q-1.

# Convolution

$$f[m] * g[m] \Leftrightarrow F[k]G[l]$$

- Zero padding



## 2D DCT

- Separable product (equivalently, a composition) of DCTs along each dimension

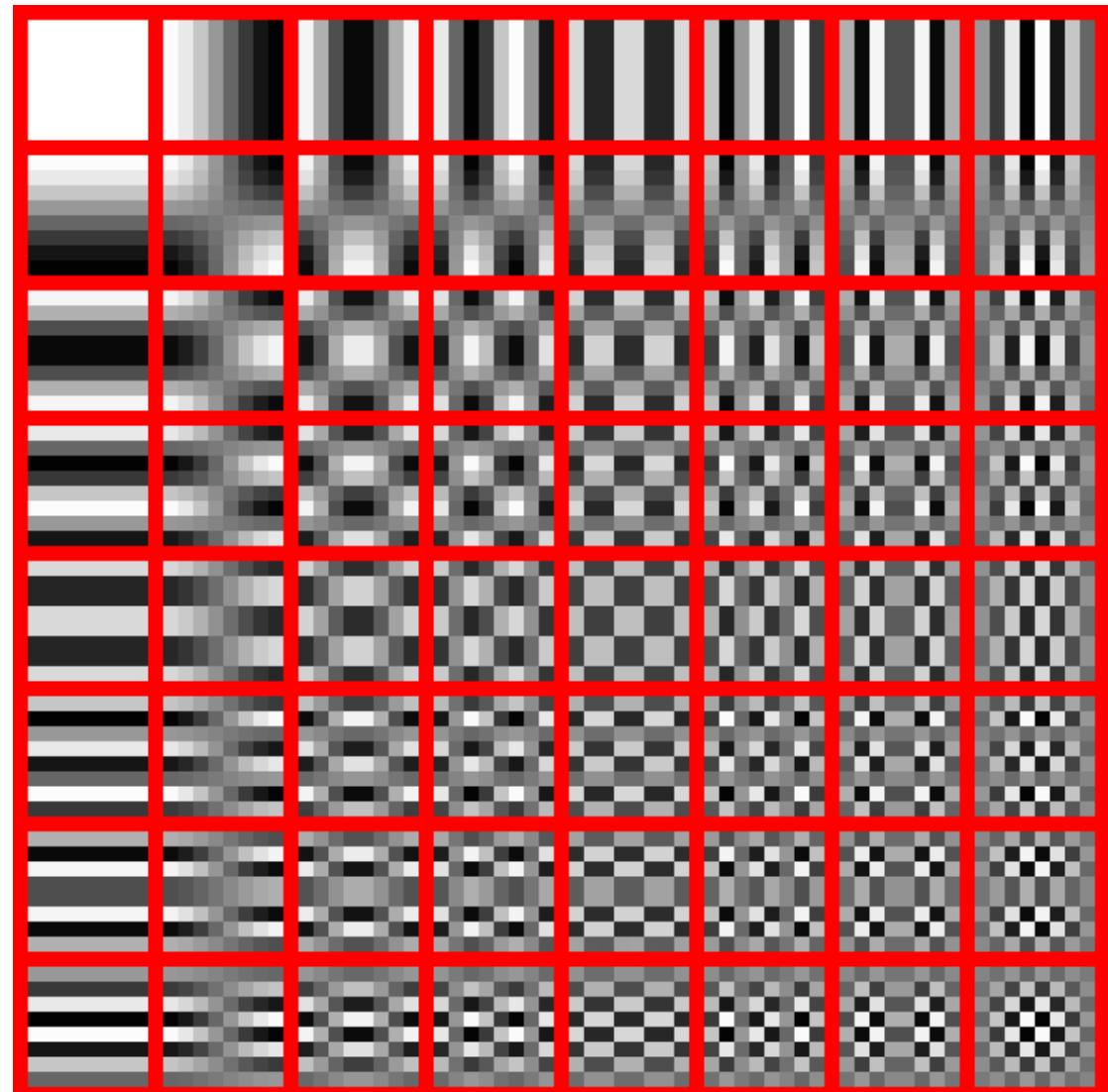
$$X_{k_1, k_2} = \sum_{n_1=0}^{N_1-1} \sum_{n_2=0}^{N_2-1} x_{n_1, n_2} \cos\left[\frac{\pi}{N_1}\left(n_1 + \frac{1}{2}\right)k_1\right] \cos\left[\frac{\pi}{N_2}\left(n_2 + \frac{1}{2}\right)k_2\right]$$

- *Row-column* algorithm
- 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

# DCT: basis functions

*Block-based transform*

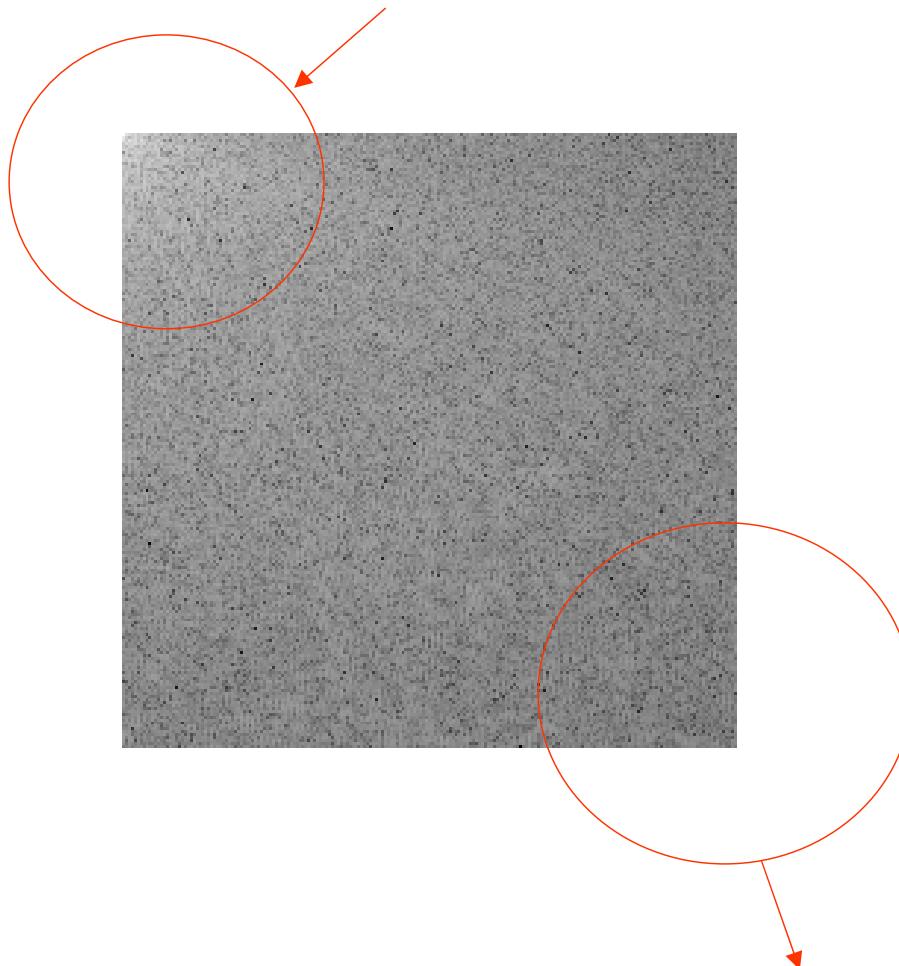
Block size  
 $N_1=N_2=8$



# DCT: example

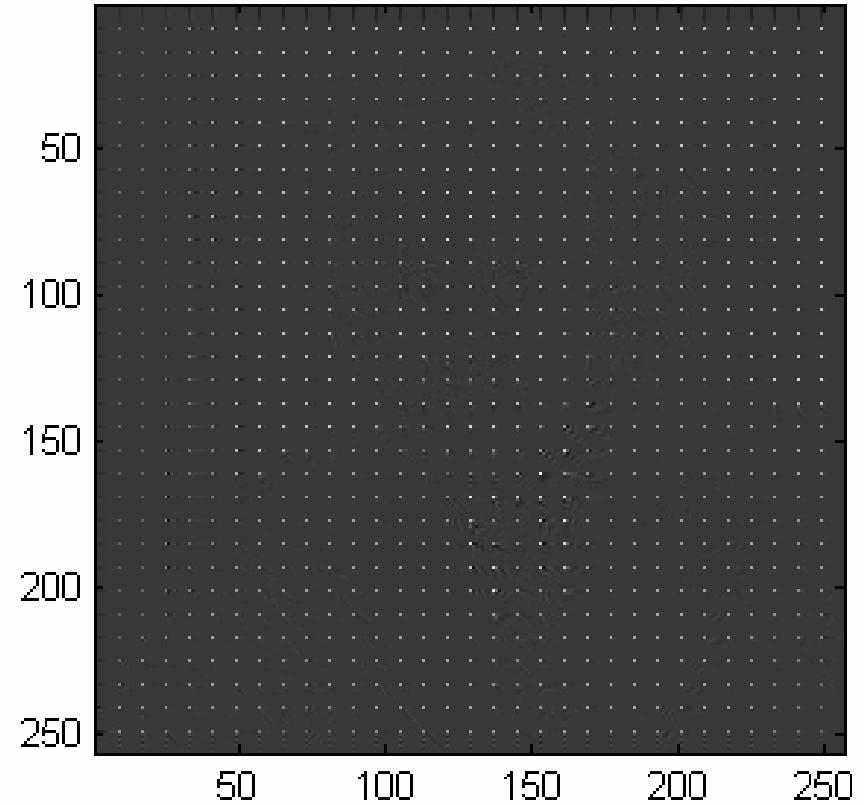
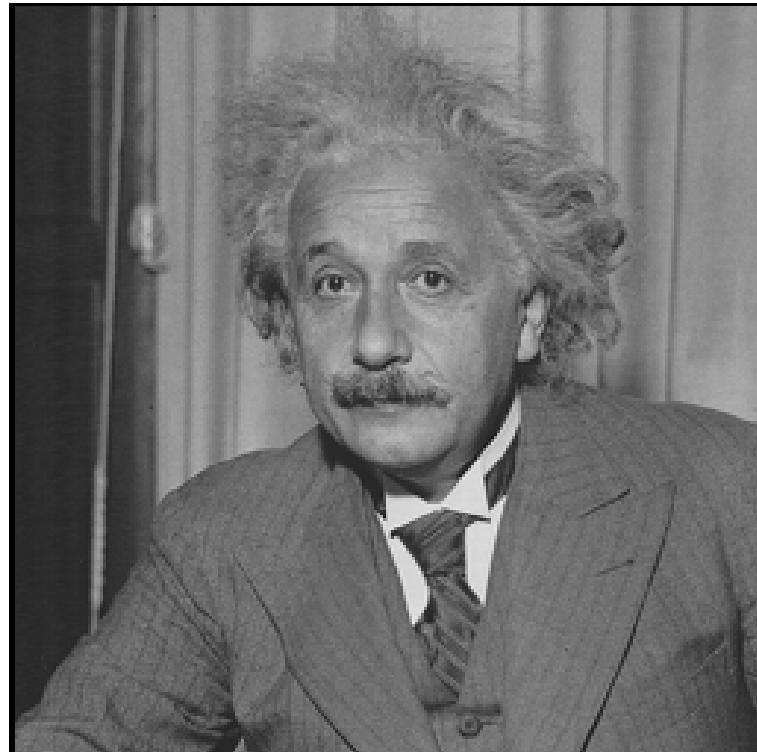


Low-frequency components

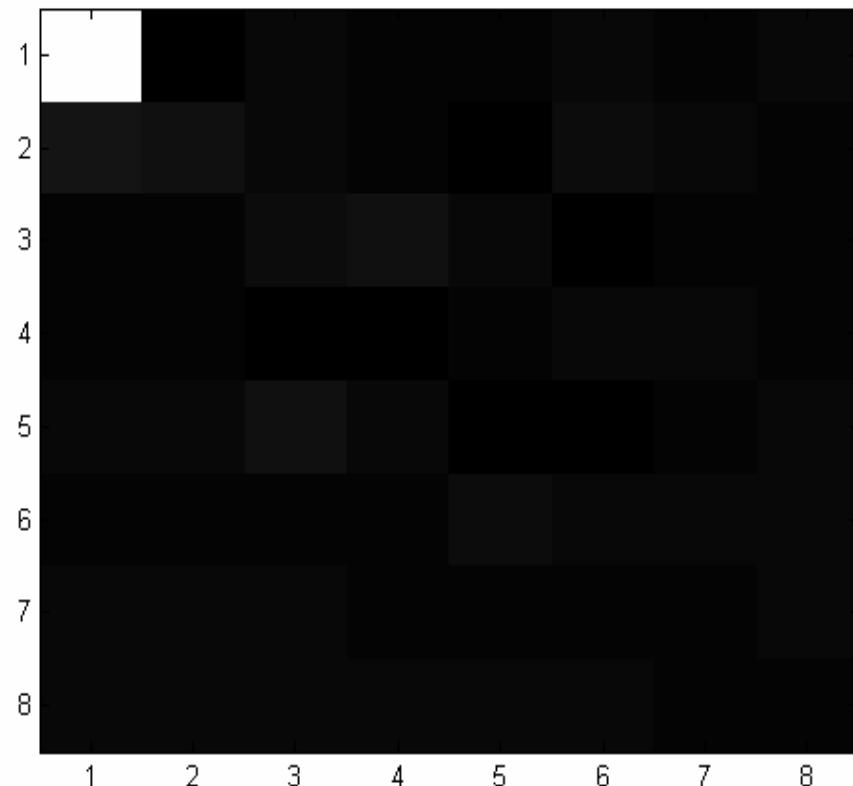
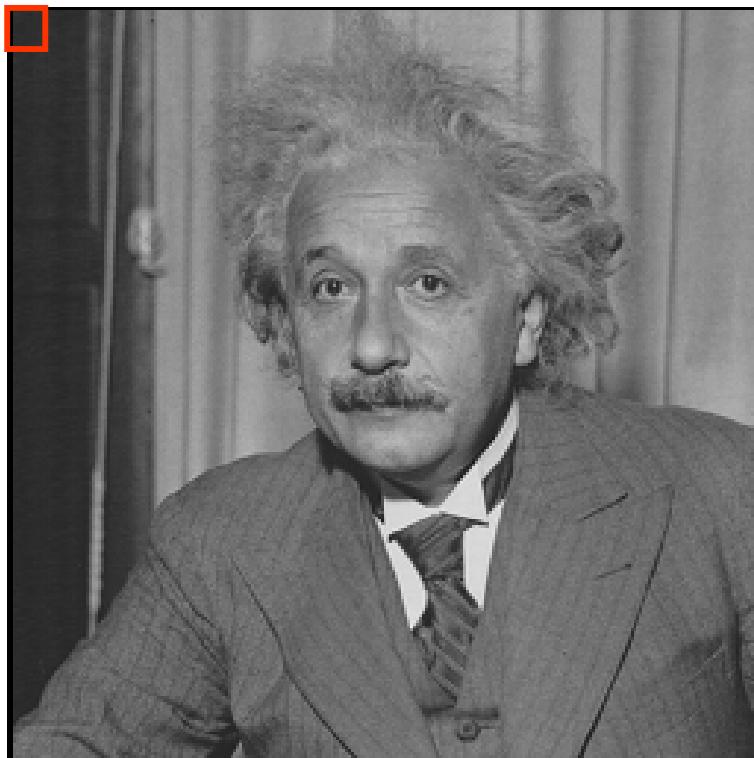


High-frequency components

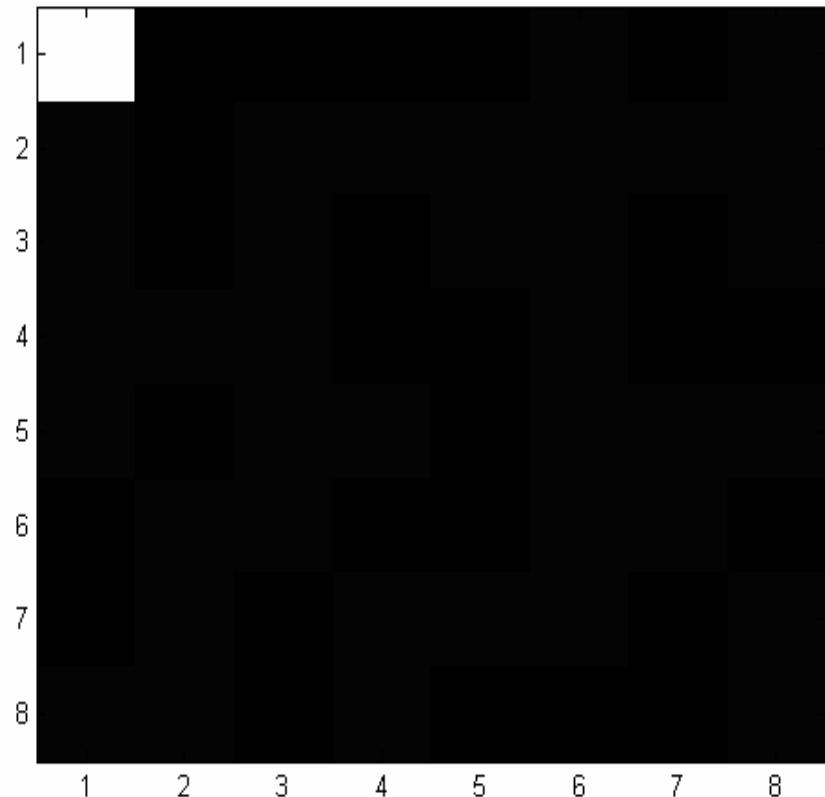
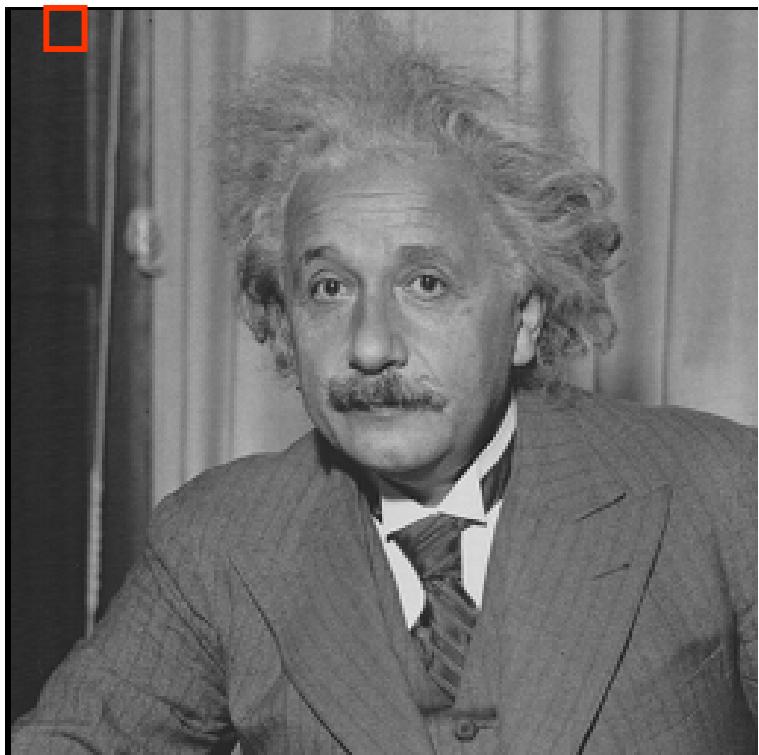
# Block-based DCT



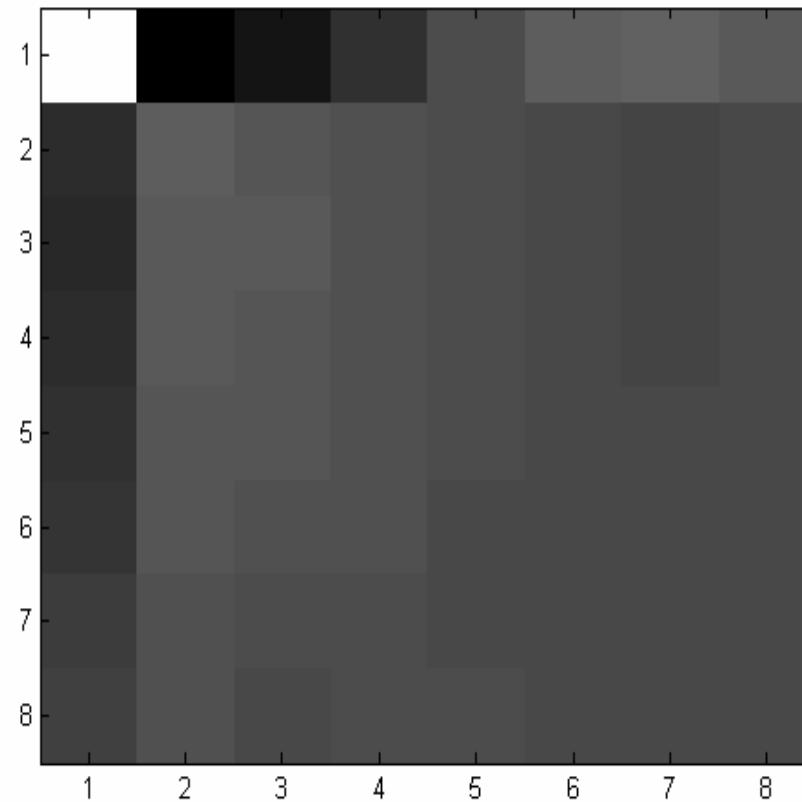
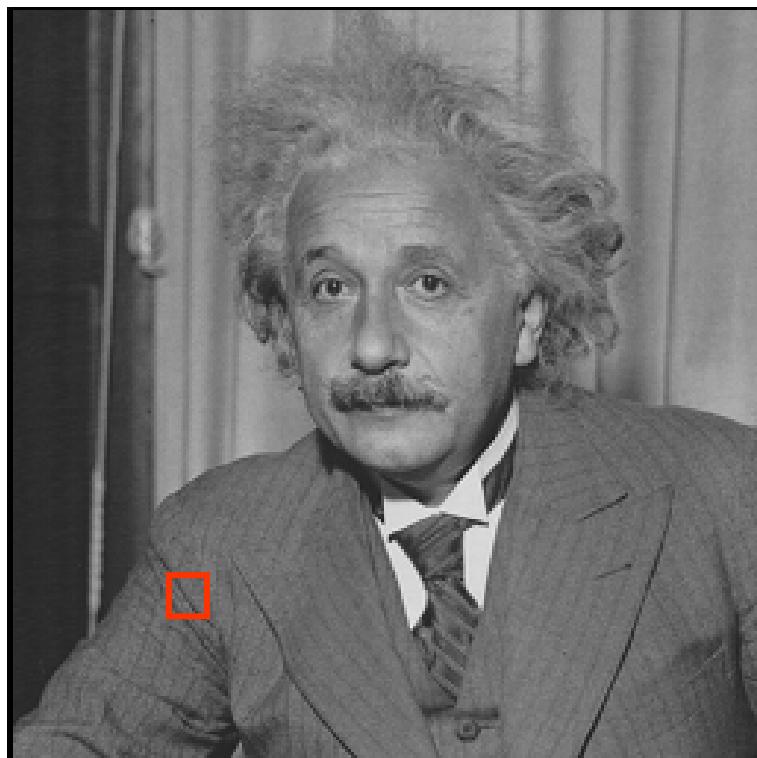
# Block-based DCT



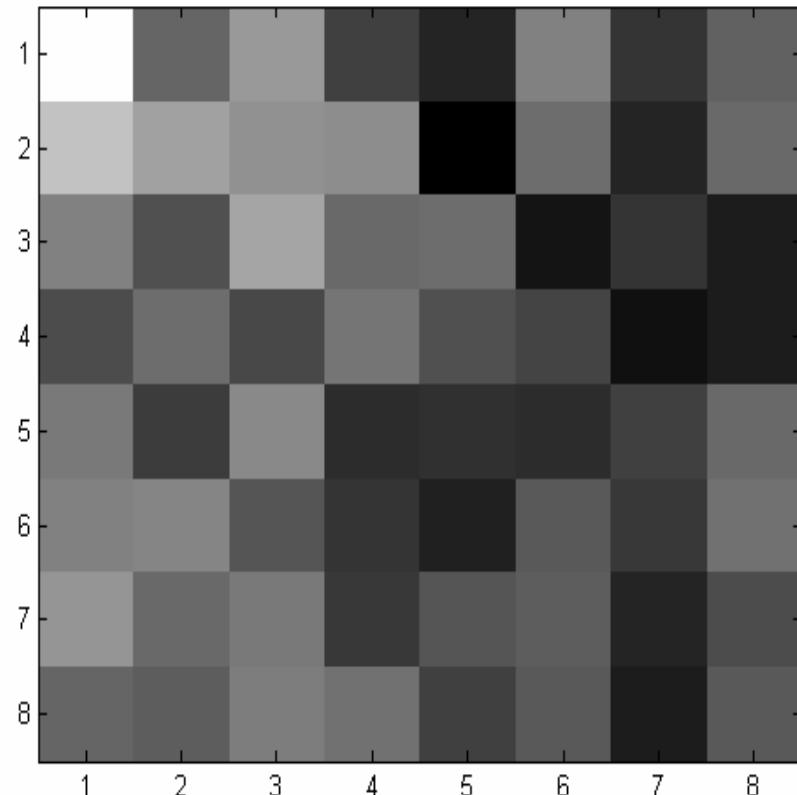
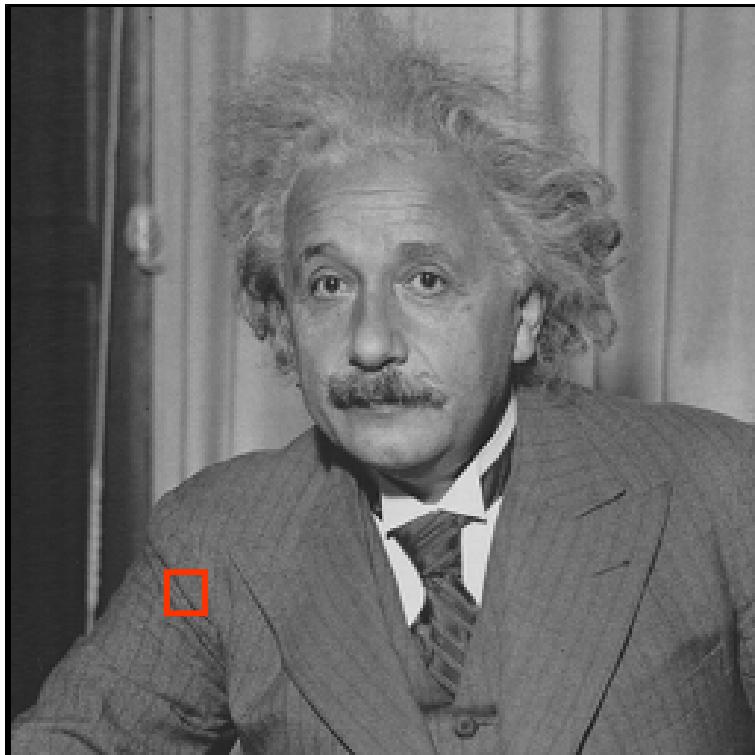
# Block-based DCT



# Block-based DCT



# Block-based DCT

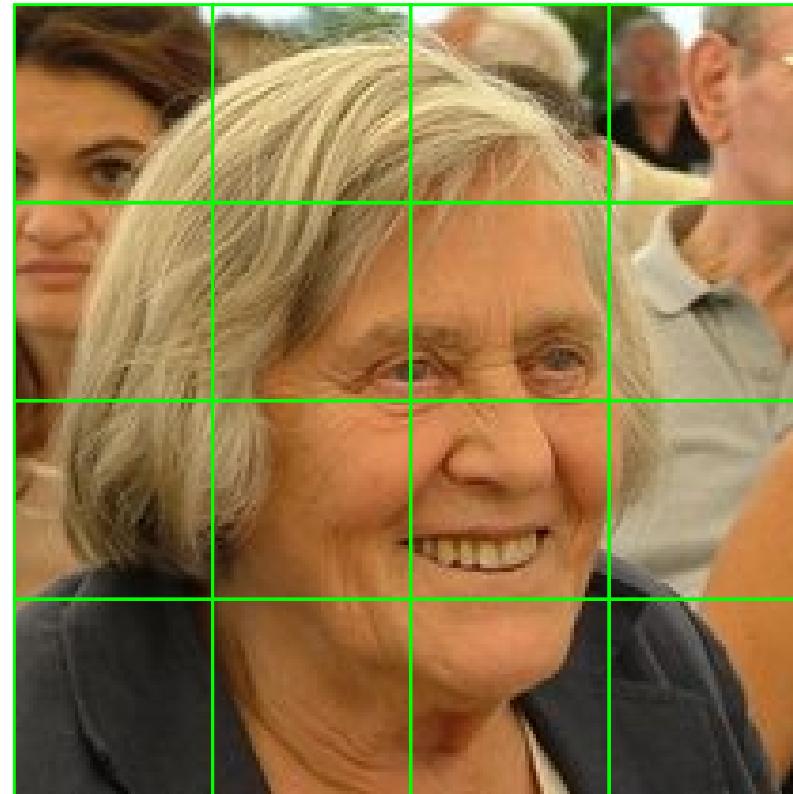


The mean value has been subtracted before calculating the DCT. This is usual in coding applications and allows to outline the value of the coefficients corresponding to non-zero spatial frequencies

# JPEG preview

*Block-based DCT*

Block size  
 $N_1=N_2=8$



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

# Appendix

# Appendix: Impulse Train

- The Fourier Transform of a comb function is

$$\begin{aligned} F\left(\text{comb}_{M,N}[m,n]\right) &= \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} \text{comb}_{M,N}[m,n] e^{-j2\pi(um+vn)} \\ &= \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} \left[ \sum_{k=-\infty}^{\infty} \sum_{l=-\infty}^{\infty} \delta[m-kM, n-lN] \right] e^{-j2\pi(um+vn)} \\ &= \sum_{k=-\infty}^{\infty} \sum_{l=-\infty}^{\infty} \left[ \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} \delta[m-kM, n-lN] e^{-j2\pi(um+vn)} \right] \\ &= \sum_{k=-\infty}^{\infty} \sum_{l=-\infty}^{\infty} \left[ e^{-j2\pi(ukM+vlN)} \right] \end{aligned}$$

## Impulse Train (cont'd)

- The Fourier Transform of a comb function is

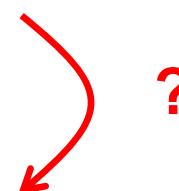
$$F\left(comb_{M,N}[m,n]\right) = \sum_{k=-\infty}^{\infty} \sum_{l=-\infty}^{\infty} \left[ e^{-j2\pi(ukM+vlN)} \right]$$

$$= \sum_{k=-\infty}^{\infty} \sum_{l=-\infty}^{\infty} \left[ 1 e^{-j2\pi((uM)k+(vN)l)} \right]$$

(Fourier Trans. of 1)

$$= \sum_{k=-\infty}^{\infty} \sum_{l=-\infty}^{\infty} \delta(uM - k, vN - l)$$

$$= \frac{1}{MN} \sum_{k=-\infty}^{\infty} \sum_{l=-\infty}^{\infty} \delta\left(u - \frac{k}{M}, v - \frac{l}{N}\right)$$



# Impulse Train (cont'd)

## ■ Proof

$$\begin{aligned} \int_{-\infty}^{\infty} \sum_{k=-\infty}^{\infty} \delta(uM - k) F(u) du &= \sum_{k=-\infty}^{\infty} \int_{-\infty}^{\infty} \delta(uM - k) F(u) du \\ &= \sum_{k=-\infty}^{\infty} \frac{1}{M} \int_{-\infty}^{\infty} \delta(v - k) F\left(\frac{v}{M}\right) dv \\ &= \frac{1}{M} \sum_{k=-\infty}^{\infty} F\left(\frac{k}{M}\right) = \frac{1}{M} \sum_{k=-\infty}^{\infty} \int_{-\infty}^{\infty} \delta(v - \frac{k}{M}) F(v) dv \\ &= \int_{-\infty}^{\infty} \frac{1}{M} \sum_{k=-\infty}^{\infty} \delta(u - \frac{k}{M}) F(u) du \end{aligned}$$