

# SIGNALS AND SYSTEMS USING MATLAB

## Chapter 11 — Fourier Analysis of Discrete-time Signals and Systems

## Chapter 11 — Fourier Analysis of Discrete-time Signals and Systems

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# Discrete-time Fourier transform (DTFT)

$$\begin{aligned} \text{DTFT} \quad X(e^{j\omega}) &= \sum_n x[n] e^{-j\omega n}, \quad -\pi \leq \omega < \pi \\ \text{IDTFT} \quad x[n] &= \frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{j\omega}) e^{j\omega n} d\omega \end{aligned}$$

- Periodic

$$X(e^{j(\omega+2\pi k)}) = \sum_n x[n] e^{-j(\omega+2\pi k)n} = X(e^{j\omega}) \quad k \text{ integer}$$

- Sampling and DTFT

$$X_s(e^{j\omega}) = \mathcal{F}[x_s(t)] = \sum_n x(nT_s) \mathcal{F}[\delta(t - nT_s)] = \sum_n x(nT_s) e^{-jn\Omega T_s}$$

- Z-transform and the DTFT

$$X_s(e^{j\omega}) = X(z)|_{z=e^{j\omega}}, \quad UC \subset ROC$$

- Eigenvalues and the DTFT

LTI system, input  $x[n] = e^{j\omega_0 n}$ , the steady-state output

$$y[n] = \sum_k h[k] x[n-k] = \sum_k h[k] e^{j\omega_0(n-k)} = e^{j\omega_0 n} H(e^{j\omega_0})$$

$$H(e^{j\omega_0}) = \sum_k h[k] e^{-j\omega_0 k}, \quad \text{DTFT}[h[n]]$$

Example: Non-causal  $x[n] = \alpha^{|n|}$  with  $|\alpha| < 1$

$$\begin{aligned} X(z) &= \sum_{n=0}^{\infty} \alpha^n z^{-n} + \sum_{m=0}^{\infty} \alpha^m z^m - 1 \\ &= \frac{1 - \alpha^2}{1 - \alpha(z + z^{-1}) + \alpha^2}, \quad \text{ROC: } |\alpha| < |z| < \frac{1}{|\alpha|} \end{aligned}$$

$$X(e^{j\omega}) = X(z)|_{z=e^{j\omega}} = \frac{1 - \alpha^2}{(1 + \alpha^2) - 2\alpha \cos(\omega)}$$

The DTFT at  $\omega = 0$  gives

$$X(e^{j0}) = \sum_{n=-\infty}^{\infty} x[n] e^{j0n} = \sum_{n=-\infty}^{\infty} \alpha^{|n|} = \frac{2}{1 - \alpha} - 1 = \frac{1 + \alpha}{1 - \alpha}$$

equivalently

$$X(e^{j0}) = \frac{1 - \alpha^2}{1 - 2\alpha + \alpha^2} = \frac{1 - \alpha^2}{(1 - \alpha)^2} = \frac{1 + \alpha}{1 - \alpha}$$

Dual pairs

$$\begin{aligned}
 \delta[n - k], \text{ integer } k &\Leftrightarrow e^{-j\omega k} \\
 e^{-j\omega_0 n}, \quad -\pi \leq \omega_0 < \pi &\Leftrightarrow 2\pi\delta(\omega + \omega_0) \\
 \sum_k X[k]e^{-j\omega_k n} &\Leftrightarrow \sum_k 2\pi X[k]\delta(\omega + \omega_k)
 \end{aligned}$$

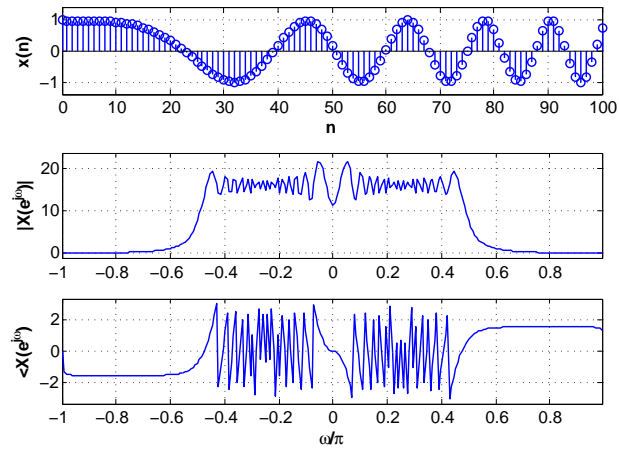
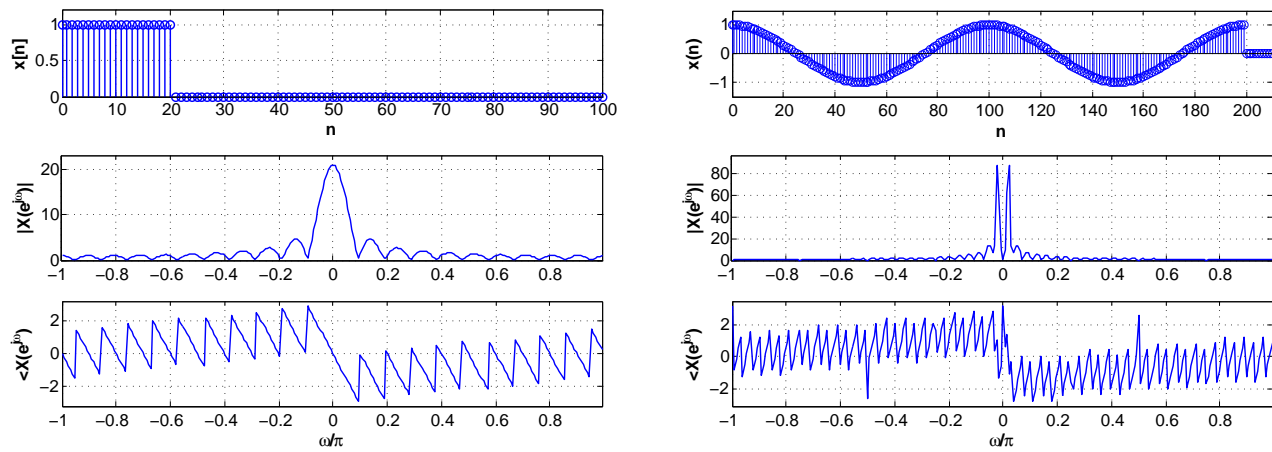
DTFT of

$$\begin{aligned}
 x[n] &= \sum_{\ell} A_{\ell} \cos(\omega_{\ell} n + \theta_{\ell}) = \sum_{\ell} 0.5 A_{\ell} (e^{j(\omega_{\ell} n + \theta_{\ell})} + e^{-j(\omega_{\ell} n + \theta_{\ell})}) \\
 X(e^{j\omega}) &= \sum_{\ell} \pi A_{\ell} [e^{j\theta_{\ell}} \delta(\omega - \omega_{\ell}) + e^{-j\theta_{\ell}} \delta(\omega + \omega_{\ell})] \quad -\pi \leq \omega < \pi
 \end{aligned}$$

Example:

$$X(e^{j\omega}) = 1 + \delta(\omega - 4) + \delta(\omega + 4) + 0.5\delta(\omega - 2) + 0.5\delta(\omega + 2) \Rightarrow$$

$$x[n] = \frac{1}{2\pi} \delta[n] + \frac{1}{0.5\pi} \cos(4n) + \frac{1}{\pi} \cos(2n)$$



*DTFT of a pulse, a windowed sinusoid and a chirp: magnitude and phase spectra for each*

# Decimation and interpolation

- $x[n]$ , band-limited to  $\pi/M$  in  $[-\pi, \pi)$  or  $|X(e^{j\omega})| = 0, |\omega| > \pi/M$  for an integer  $M > 1$ , can be **down-sampled** by a factor of  $M$  to generate a discrete-time signal

$$x_d[n] = x[Mn] \quad \text{with} \quad X_d(e^{j\omega}) = \frac{1}{M} X(e^{j\omega/M})$$

an expanded version of  $X(e^{j\omega})$ .

- A signal  $x[n]$  is **up-sampled** by a factor of  $L > 1$  to generate a signal  $x_u[n] = x[n/L]$  for  $n = \pm kL, k = 0, 1, 2, \dots$  and zero otherwise. The DTFT of  $x_u[n]$  is  $X(e^{jL\omega})$  or a compressed version of  $X(e^{j\omega})$ .

Example: Ideal low-pass filter with frequency response

$$H(e^{j\omega}) = \begin{cases} 1 & -\pi/2 \leq \omega \leq \pi/2 \\ 0 & -\pi \leq \omega < -\pi/2 \quad \text{and} \quad \pi/2 < \omega \leq \pi \end{cases}$$
$$h[n] = \frac{1}{2\pi} \int_{-\pi/2}^{\pi/2} e^{j\omega n} d\omega = \begin{cases} 0.5 & n = 0 \\ \sin(\pi n/2)/(\pi n) & n \neq 0 \end{cases}$$

Down-sampled impulse response

$$h_d[n] = h[2n] = \begin{cases} 0.5 & n = 0 \\ \sin(\pi n)/(2\pi n) = 0 & n \neq 0 \end{cases} = 0.5\delta[n]$$
$$H_d(e^{j\omega}) = \frac{1}{2} H(e^{j\omega/2}) = \frac{1}{2}, \quad -\pi \leq \omega < \pi$$

Example: Pulse  $x[n] = u[n] - u[n - 4]$  down-sampled by  $M = 2$  gives

$$x_d[n] = x[2n] = u[2n] - u[2n - 4] = u[n] - u[n - 2]$$

$$X(z) = 1 + z^{-1} + z^{-2} + z^{-3} \quad \text{ROC: whole Z-plane (except for the origin)}$$

$$\begin{aligned} X(e^{j\omega}) &= e^{-j(\frac{3}{2}\omega)} \left[ e^{j(\frac{3}{2}\omega)} + e^{j(\frac{1}{2}\omega)} + e^{-j(\frac{1}{2}\omega)} + e^{-j(\frac{3}{2}\omega)} \right] \\ &= 2e^{-j(\frac{3}{2}\omega)} \left[ \cos\left(\frac{\omega}{2}\right) + \cos\left(\frac{3\omega}{2}\right) \right] \end{aligned}$$

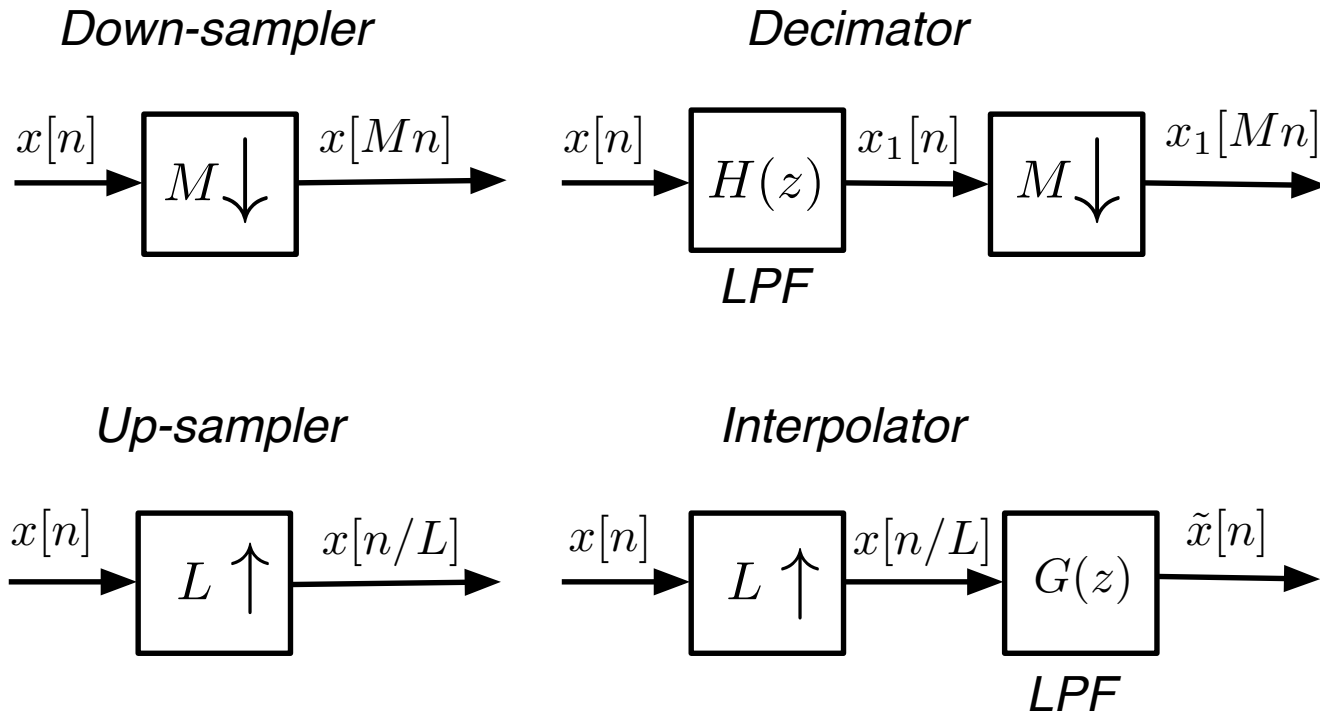
$$\begin{aligned} X_d(z) = 1 + z^{-2} &\Rightarrow X_d(e^{j\omega}) = e^{-j\omega} [e^{j\omega} + e^{-j\omega}] \\ &= 2e^{-j\omega} \cos(\omega) \end{aligned}$$

$$X_d(e^{j\omega}) \neq 0.5X(e^{j\omega/2})$$

Aliasing: maximum frequency of  $x[n]$  is not  $\pi/M = \pi/2$

Passing  $x[n]$  through ideal low-pass filter  $H(e^{j\omega})$  with cut-off frequency  $\pi/2$ , output  $x_1[n]$  has maximum frequency of  $\pi/2$  and down-sampling it with  $M = 2$  would give a signal with a DTFT  $0.5X_1(e^{j\omega/2})$

# Decimator and interpolator



*Down-sampler and decimator (top) and up-sampler and interpolator (bottom)*

$$H(e^{j\omega}) = \begin{cases} 1 & -\pi/M \leq \omega \leq \pi/M \\ 0 & \text{otherwise in } [-\pi, \pi) \end{cases}$$

$$G(e^{j\omega}) = \begin{cases} L & -\pi/L \leq \omega \leq \pi/L \\ 0 & \text{otherwise in } [-\pi, \pi) \end{cases}$$



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**Table 11.1 Discrete-time Fourier Transform (DTFT) Properties**


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Z-transform:	$x[n], X(z),  z  = 1 \in ROC$	$X(e^{j\omega}) = X(z) _{z=e^{j\omega}}$
Periodicity:	$x[n]$	$X(e^{j\omega}) = X(e^{j(\omega+2\pi k)}), \quad k \text{ integer}$
Linearity:	$\alpha x[n] + \beta y[n]$	$\alpha X(e^{j\omega}) + \beta Y(e^{j\omega})$
Time-shifting:	$x[n - N]$	$e^{-j\omega N} X(e^{j\omega})$
Frequency-shift:	$x[n]e^{j\omega_0 n}$	$X(e^{j(\omega-\omega_0)})$
Convolution:	$(x * y)[n]$	$X(e^{j\omega})Y(e^{j\omega})$
Multiplication:	$x[n]y[n]$	$\frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{j\theta})Y(e^{j(\omega-\theta)})d\theta$
Symmetry:	$x[n]$ real-valued	$ X(e^{j\omega}) $ even function of $\omega$ $\angle X(e^{j\omega})$ odd function of $\omega$
Parseval's relation:	$\sum_{n=-\infty}^{\infty}  x[n] ^2 = \frac{1}{2\pi} \int_{-\pi}^{\pi}  X(e^{j\omega}) ^2 d\omega$	

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Example: DTFT of sinusoids cannot be found from the Z-transform or from the sum defining the DTFT

- Cosine – using frequency–shift property

$$\begin{aligned}x[n] &= \cos(\omega_0 n) = 0.5(e^{j\omega_0 n} + e^{-j\omega_0 n}) \\X(e^{j\omega}) &= DTFT[0.5]_{\omega-\omega_0} + DTFT[0.5]_{\omega+\omega_0} = \pi [\delta(\omega - \omega_0) + \delta(\omega + \omega_0)]\end{aligned}$$

- Sine – using time–shift property

$$\begin{aligned}y[n] &= \sin(\omega_0 n) = \cos(\omega_0(n - \pi/(2\omega_0))) = x[n - \pi/(2\omega_0)] \\Y(e^{j\omega}) &= X(e^{j\omega})e^{-j\omega\pi/(2\omega_0)} = \pi \left[ \delta(\omega - \omega_0)e^{-j\omega\pi/(2\omega_0)} + \delta(\omega + \omega_0)e^{-j\omega\pi/(2\omega_0)} \right] \\&= \pi \left[ \delta(\omega - \omega_0)e^{-j\pi/2} + \delta(\omega + \omega_0)e^{j\pi/2} \right] = -j\pi [\delta(\omega - \omega_0) - \delta(\omega + \omega_0)]\end{aligned}$$

Example: For  $x[n] = \cos(\omega_0 n + \phi)$ ,  $-\pi \leq \phi < \pi$ ,

$$X(e^{j\omega}) = \pi [e^{-j\phi}\delta(\omega - \omega_0) + e^{j\phi}\delta(\omega + \omega_0)]$$

$$\text{magnitude } |X(e^{j\omega})| = |X(e^{-j\omega})| = \pi [\delta(\omega - \omega_0) + \delta(\omega + \omega_0)]$$

$$\text{phase } \theta(\omega) = \begin{cases} \phi & \omega = -\omega_0 \\ -\phi & \omega = \omega_0 \\ 0 & \text{otherwise} \end{cases}$$

## Example: FIR filters

$$(i) \quad h_1[n] = \sum_{k=0}^9 \frac{1}{10} \delta[n-k]$$

$$H_1(z) = \frac{1}{10} \sum_{n=0}^9 z^{-n} = 0.1 \frac{1 - z^{-10}}{1 - z^{-1}} = 0.1 \frac{z^{10} - 1}{z^9(z - 1)} = 0.1 \frac{\prod_{k=1}^9 (z - e^{j2\pi k/10})}{z^9}$$

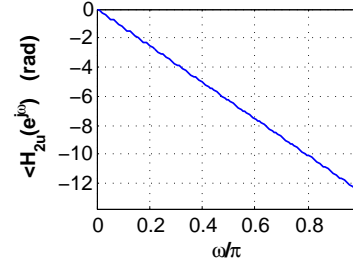
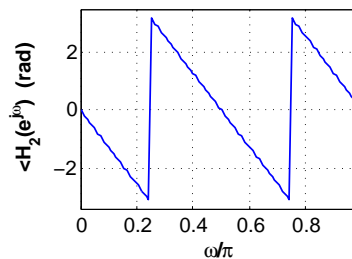
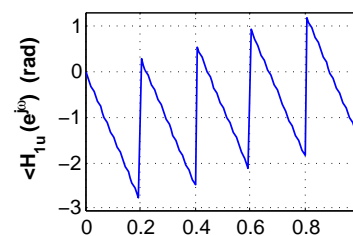
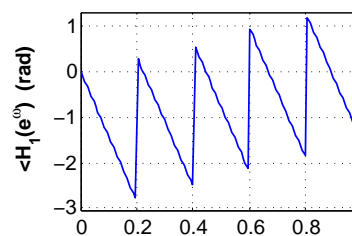
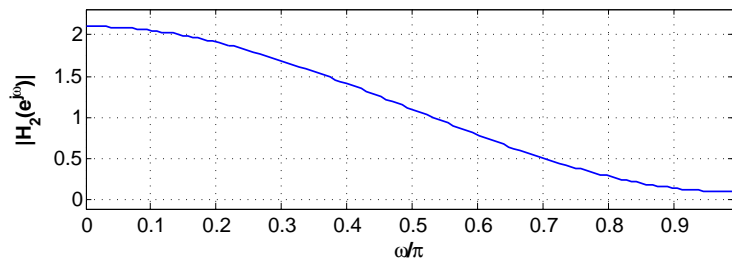
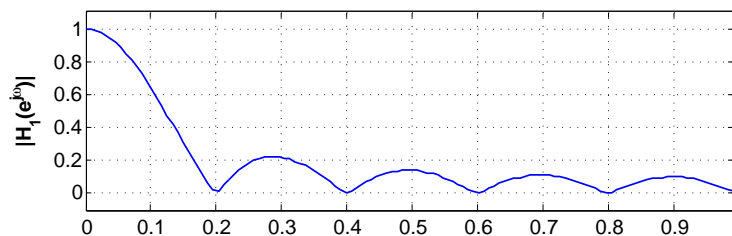
Because zeros on UC, its phase is not defined at the frequencies of the zeros (not continuous) and it cannot be unwrapped

$$(ii) \quad h_2[n] = 0.5\delta[n-3] + 1.1\delta[n-4] + 0.5\delta[n-5] \quad \text{symmetric about } n=4$$

$$H_2(z) = 0.5z^{-3} + 1.1z^{-4} + 0.5z^{-5} = z^{-4}(0.5z + 1.1 + 0.5z^{-1})$$

$$\text{frequency response } H_2(e^{j\omega}) = e^{-j4\omega}(1.1 + \cos(\omega))$$

Since  $1.1 + \cos(\omega) > 0$  for  $-\pi \leq \omega < \pi$ , the phase  $\angle H_2(e^{j\omega}) = -4\omega$ , i.e., a linear phase.



# Convolution sum

$h[n]$  impulse response of stable LTI system, output

$$y[n] = \sum_k x[k] h[n - k], \quad x[n] \text{ (input)}$$

$$Y(z) = H(z)X(z) \quad \text{ROC} : \mathcal{R}_Y = \mathcal{R}_H \cap \mathcal{R}_X$$

$$\text{UC} \subset \mathcal{R}_Y \Rightarrow Y(e^{j\omega}) = H(e^{j\omega})X(e^{j\omega}) \quad \text{or}$$

$$|Y(e^{j\omega})| = |H(e^{j\omega})||X(e^{j\omega})|$$

$$\angle Y(e^{j\omega}) = \angle H(e^{j\omega}) + \angle X(e^{j\omega})$$

Example: All-pass system or cascade systems with transfer functions

$$H_i(z) = K_i \frac{z - 1/\alpha_i}{z - \alpha_i^*} \quad |z| > |\alpha_i|, \quad i = 1, \dots, N-1, \quad |\alpha_i| < 1, \quad K_i > 0$$

For zero  $1/\alpha_i$  of  $H_i(z)$ , a pole  $\alpha_i^*$  exists

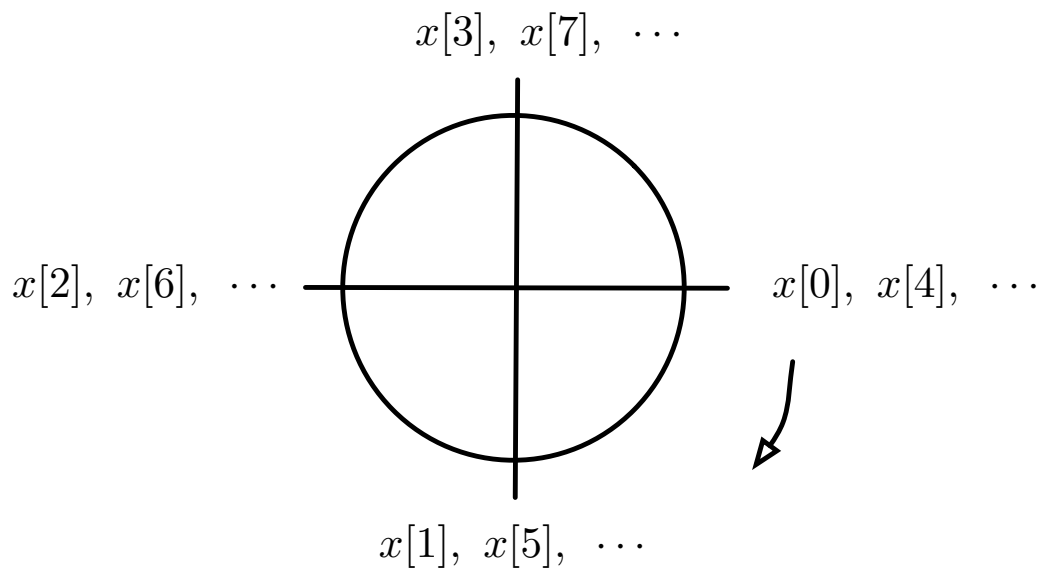
$$|H_i(e^{j\omega})|^2 = H_i(e^{j\omega})H_i^*(e^{j\omega}) = K_i^2 \frac{e^{j\omega}(e^{-j\omega} - \alpha_i)e^{-j\omega}(e^{j\omega} - \alpha_i^*)}{\alpha_i\alpha_i^*(e^{j\omega} - \alpha_i^*)(e^{-j\omega} - \alpha_i)} = \frac{K_i^2}{|\alpha_i|^2}$$

$$H(e^{j\omega}) = \prod_i H_i(e^{j\omega}) = \prod_i |\alpha_i| \frac{e^{j\omega} - 1/\alpha_i}{e^{j\omega} - \alpha_i}, \quad \Rightarrow$$

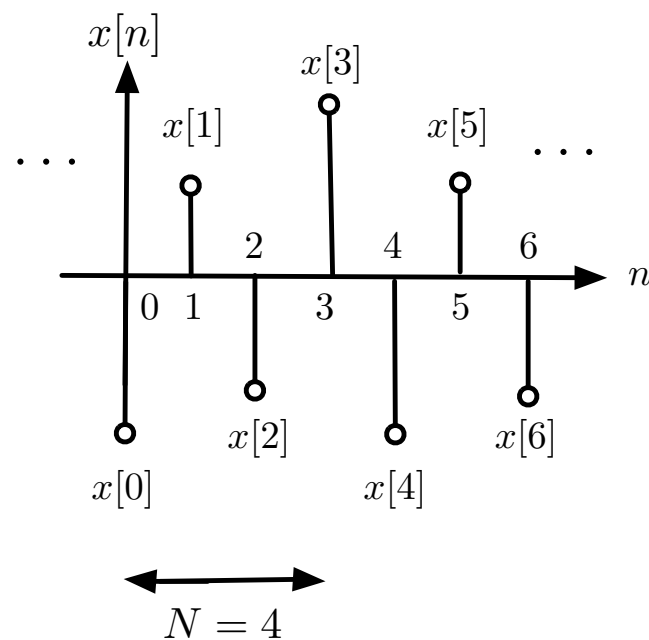
$$|H(e^{j\omega})| = \prod_i |H_i(e^{j\omega})| = 1, \quad \angle H(e^{j\omega}) = \sum_i \angle H_i(e^{j\omega})$$

$$Y(e^{j\omega}) = |X(e^{j\omega})| e^{j(\angle X(e^{j\omega}) + \angle H(e^{j\omega}))}$$

# Fourier series



(a)



(b)

*Circular (a) and linear (b) representations of a periodic discrete-time signal  $x[n]$*

Periodic signal  $x[n]$  of fundamental period  $N$

$$\text{Fourier series } x[n] = \sum_{k=k_0}^{k_0+N-1} X[k] e^{j\frac{2\pi}{N}kn}$$

Fourier series coefficients

$$X[k] = \frac{1}{N} \sum_{n=n_0}^{n_0+N-1} x[n] e^{-j\frac{2\pi}{N}kn}$$

- Connection with the Z-transform

$$x_1[n] = x[n](u[n] - u[n - N]) \text{ period of } x[n]$$

$$\mathcal{Z}(x_1[n]) = \sum_{n=0}^{N-1} x[n] z^{-n} \quad \text{ROC: whole Z-plane, except for origin}$$

$$X[k] = \frac{1}{N} \sum_{n=0}^{N-1} x[n] e^{-j\frac{2\pi}{N}kn} = \frac{1}{N} \mathcal{Z}(x_1[n]) \Big|_{z=e^{j\frac{2\pi}{N}k}}$$

Example: Periodic  $x[n]$ , fundamental period  $N = 20$ , and period  $x_1[n] = u[n] - u[n - 10]$

$$X[k] = \frac{z^{-5}(z^5 - z^{-5})}{20z^{-0.5}(z^{0.5} - z^{-0.5})} \Big|_{z=e^{j\frac{2\pi}{20}k}} = \frac{e^{-j9\pi k/20} \sin(\pi k/2)}{20 \sin(\pi k/20)}$$

# DTFT of periodic signals

$$\text{Fourier series } x[n] = \sum_{k=0}^{N-1} X[k] e^{j2\pi nk/N}$$
$$X(e^{j\omega}) = \sum_{k=0}^{N-1} 2\pi X[k] \delta(\omega - 2\pi k/N) \quad -\pi \leq \omega < \pi$$

Example: Periodic signal

$$\delta_M[n] = \sum_{m=-\infty}^{\infty} \delta[n - mM], \text{ fundamental period } M$$

$$\text{DTFT: } \Delta_M(e^{j\omega}) = \sum_{m=-\infty}^{\infty} e^{-j\omega mM}$$

$$\text{Fourier series coefficients: } \Delta_M[k] = \frac{1}{M} \sum_{n=0}^{M-1} \delta[n] e^{-j2\pi nk/M} = \frac{1}{M}$$

$$\text{Fourier series } \delta_M[n] = \sum_{k=0}^{M-1} \frac{1}{M} e^{j2\pi nk/M}$$

$$\text{DTFT: } \Delta_M(e^{j\omega}) = \frac{2\pi}{M} \sum_{k=0}^{M-1} \delta\left(\omega - \frac{2\pi k}{M}\right) \quad -\pi \leq \omega < \pi$$

$x[n]$  periodic of fundamental period  $N$  input of LTI system

$$x[n] = \sum_{k=0}^{N-1} X[k] e^{j(k\omega_0)n} \quad \omega_0 = \frac{2\pi}{N}$$

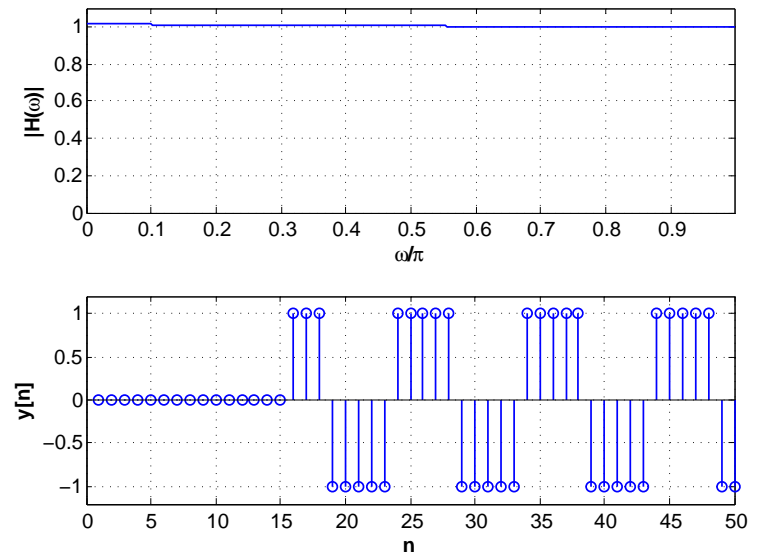
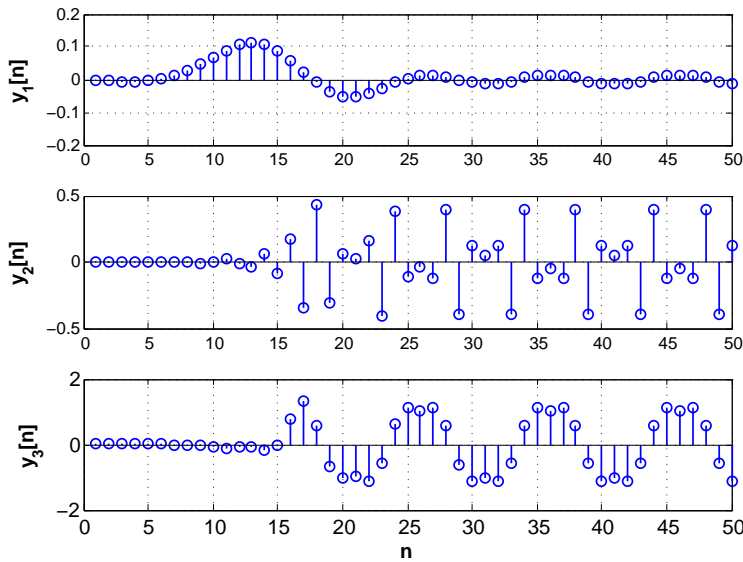
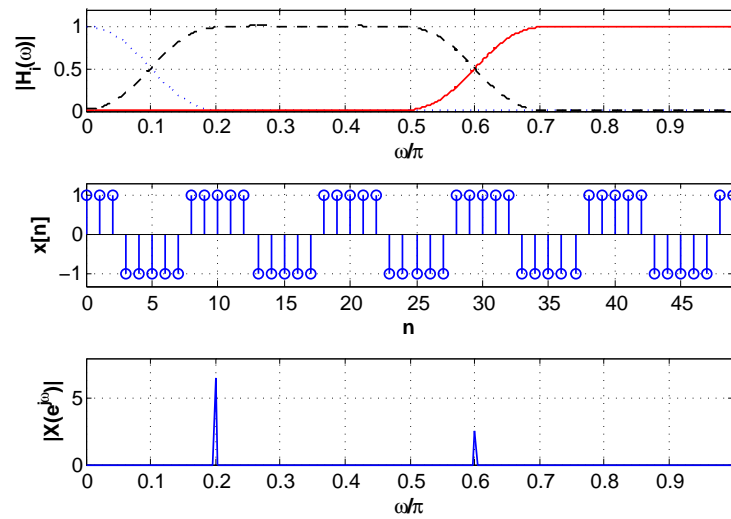
eigenfunction property of LTI systems: periodic output

$$y[n] = \sum_{k=0}^{N-1} X[k] H(e^{jk\omega_0}) e^{jk\omega_0 n} \quad \omega_0 = \frac{2\pi}{N} \text{ fundamental frequency}$$

coefficients  $Y[k] = X[k] H(e^{jk\omega_0})$

frequency response  $H(e^{jk\omega_0}) = H(z) \big|_{z=e^{jk\omega_0}}$





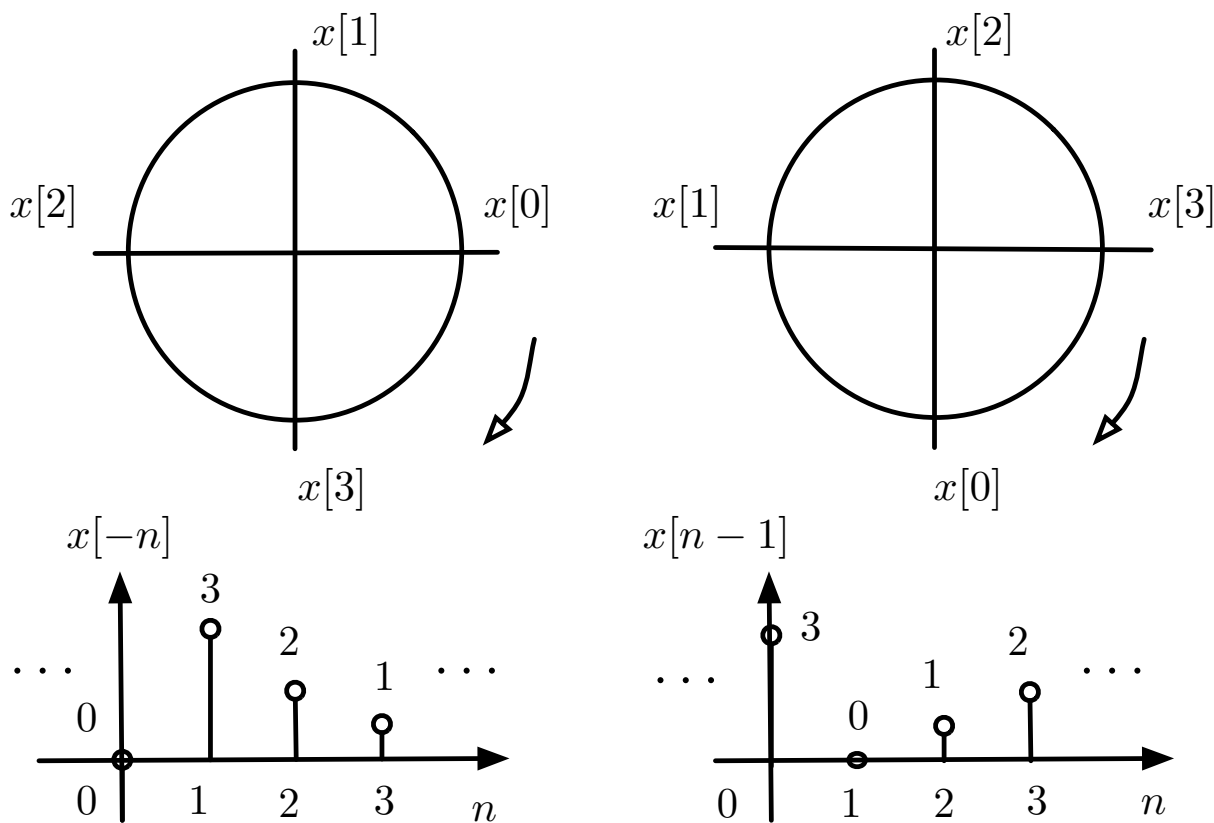
*Spectral analyzer: magnitude response of low-pass, band-pass and high-pass filters, input signal and its magnitude spectrum (top, center figure). Outputs of filters (bottom left), overall magnitude response of the bank of filters (all-pass filter) and overall response.*

# Circular shifting

$$x[n - M] \Leftrightarrow X[k]e^{-j2\pi Mk/N} \text{ FS coefficients}$$

Example: Linear shift vs circular shift

$x[n]$  periodic, of fundamental period  $N = 4$  with period  $x_1[n] = n, n = 0, \dots, 3$ .



*Circular representation of  $x[-n]$  and  $x[n-1]$*

# Periodic convolution

Periodic signals  $x[n]$  and  $y[n]$  of the same fundamental period  $N$

$$x[n]y[n] \Leftrightarrow \sum_{m=0}^{N-1} X[m]Y[k-m], \quad 0 \leq k \leq N-1 \quad (\text{periodic convolution})$$
$$(\text{periodic convolution}) \quad \sum_{m=0}^{N-1} x[m]y[n-m], \quad 0 \leq n \leq N-1 \Leftrightarrow NX[k]Y[k]$$

Example: Multiplication of the Fourier series

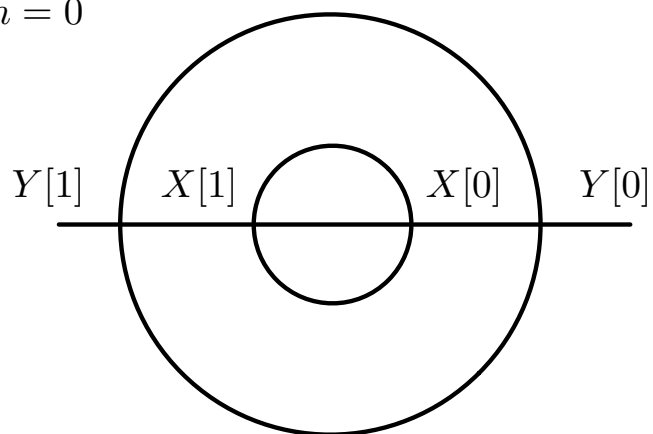
$$x[n] = X[0] + X[1]e^{j\omega_0 n}, \quad y[n] = Y[0] + Y[1]e^{j\omega_0 n} \quad \omega_0 = 2\pi/N = \pi$$

$$x[n]y[n] = \underbrace{(X[0]Y[0] + X[1]Y[1])}_{V[0]} + \underbrace{(X[0]Y[1] + X[1]Y[0])}_{V[1]} e^{j\omega_0 n}$$

Using the periodic convolution formula we have that

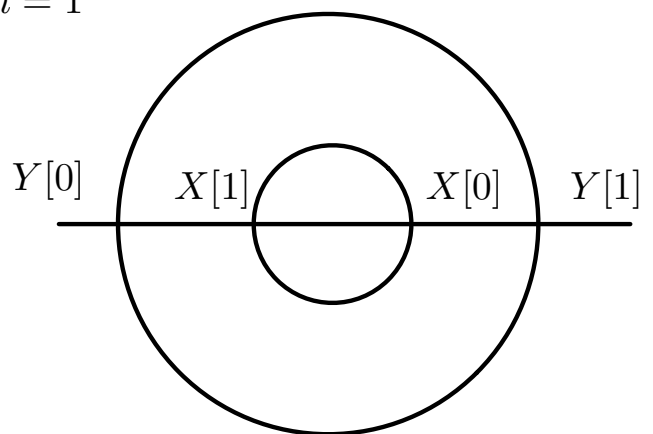
$$V[0] = \sum_{k=0}^1 X[k]Y[-k] = X[0]Y[0] + X[1]Y[-1] = X[0]Y[0] + X[1]Y[2-1]$$
$$V[1] = \sum_{k=0}^1 X[k]Y[1-k] = X[0]Y[1] + X[1]Y[0]$$

$$m = 0$$



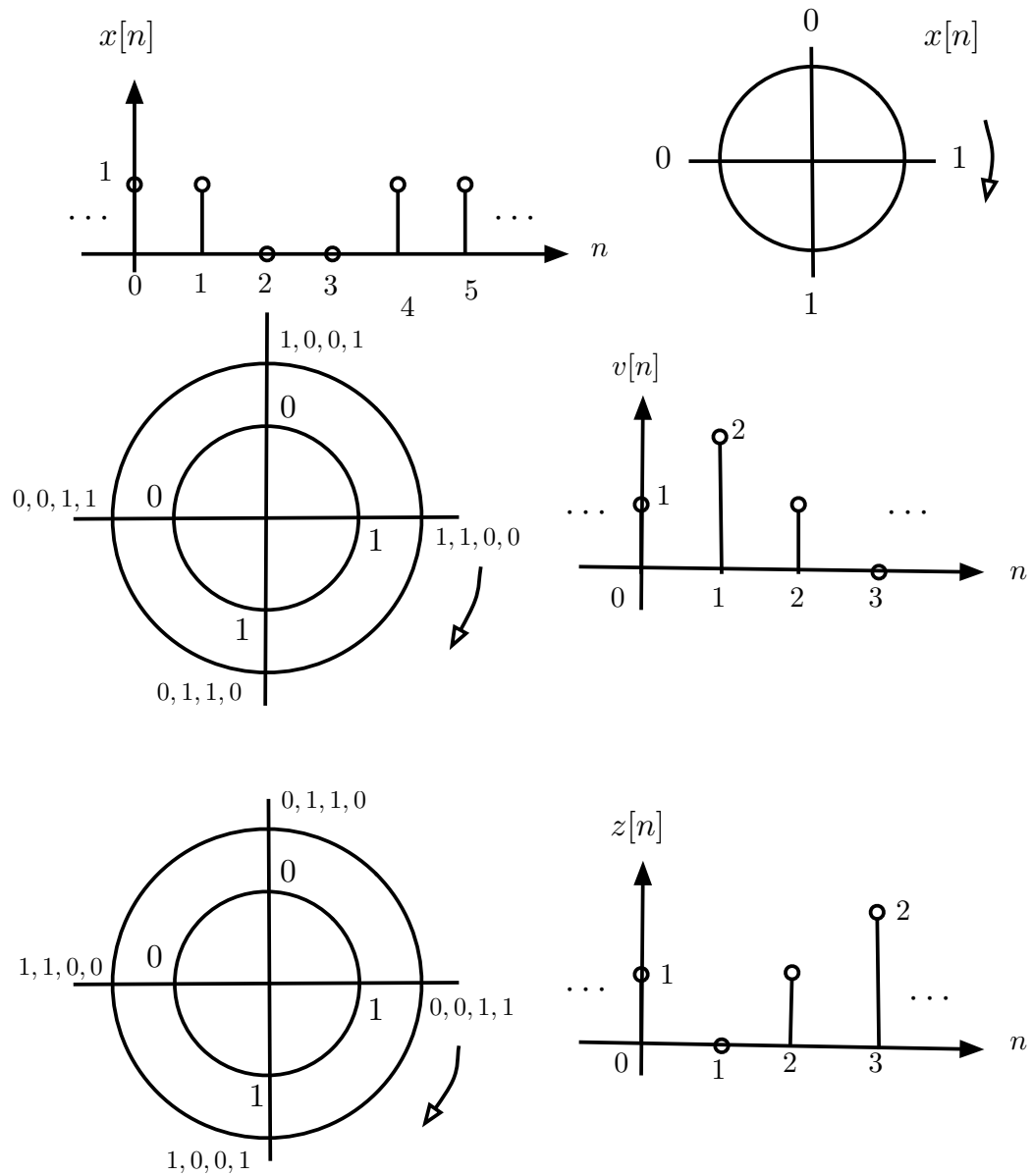
$$V[0] = X[0]Y[0] + X[1]Y[1]$$

$$m = 1$$



$$V[1] = X[0]Y[1] + X[1]Y[0]$$

*Periodic convolution of the Fourier series coefficients  $\{X[k]\}$  and  $\{Y[k]\}$ .*



Linear and circular representations of  $x[n]$  (top). Periodic convolution of  $x[n]$  with itself to get  $v[n]$ , and linear representation of  $v[n]$  (middle); periodic convolution of  $x[n]$  and  $y[n] = x[n - 2]$ , the result is  $z[n] = v[n - 2]$  represented linearly (bottom).

- Periodic signals

$x[n]$  periodic, of fundamental period  $N$

$$\text{DFT } X[k] = \sum_{n=0}^{N-1} x[n] e^{-j2\pi nk/N} \quad 0 \leq k \leq N-1$$

$$\text{IDFT } x[n] = \frac{1}{N} \sum_{k=0}^{N-1} X[k] e^{j2\pi nk/N} \quad 0 \leq n \leq N-1$$

$X[k]$ ,  $x[n]$  periodic of the same fundamental period  $N$

- Fourier series and DFT

$$\text{periodic signal } \tilde{x}[n] \Rightarrow \text{FS: } \tilde{x}[n] = \sum_{k=0}^{N-1} \tilde{X}[k] e^{j\omega_0 nk} \quad 0 \leq n \leq N-1$$

FS coefficients

$$\tilde{X}[k] = \frac{1}{N} \mathcal{Z}[\tilde{x}_1[n]]|_{z=e^{jk\omega_0}} = \frac{1}{N} \sum_{n=0}^{N-1} \tilde{x}[n] e^{-j\omega_0 nk}, \quad 0 \leq k \leq N-1, \quad \omega_0 = 2\pi/N$$

$$\text{period } \tilde{x}_1[n] = \tilde{x}[n]W[n], \quad W[n] = u[n] - u[n-N]$$

$$\text{DFT } X[k] = N\tilde{X}[k]$$

Aperiodic signal  $y[n]$  of finite length  $N$ :

- Choose  $L \geq N$ , the length of the DFT, to be fundamental period of periodic extension  $\tilde{y}[n]$  having  $y[n]$  as a period with padded zeros if necessary
- Find DFT of  $\tilde{y}[n]$ ,

$$\tilde{y}[n] = \frac{1}{L} \sum_{k=0}^{L-1} \tilde{Y}[k] e^{j2\pi nk/L} \quad 0 \leq n \leq L-1$$

and IDFT

$$\tilde{Y}[k] = \sum_{n=0}^{L-1} \tilde{y}[n] e^{-j2\pi nk/L} \quad 0 \leq k \leq L-1$$

- DFT of  $y[n]$ :  $Y[k] = \tilde{Y}[k]$  for  $0 \leq k \leq L-1$ , and

IDFT of  $Y[k]$ :  $y[n] = \tilde{y}[n]W[n]$ ,  $0 \leq n \leq L-1$ ,

$W[n] = u[n] - u[n-L]$  is a rectangular window of length  $L$ .

# DFT via Fast Fourier Transform (FFT)

Given finite length  $x[n]$  or period of periodic signal

- DFT is efficiently computed using FFT algorithm
- Causal aperiodic signal: inputting  $\{x[n], n = 0, 1, \dots, N - 1\}$  into FFT gives  $\{X[k], k = 0, 1, \dots, N - 1\}$  or DFT of  $x[n]$  using FFT of length  $L = N$   
For  $L > N$  DFT attach  $L - N$  zeros at the end of the above sequence
- Non-causal aperiodic signal:  $\{x[n], n = -n_0, \dots, 0, 1, \dots, N - n_0 - 1\}$  use periodic extension to get

$$\underbrace{x[0] \ x[1] \ \cdots \ x[N - n_0 - 1]}_{\text{causal samples}} \quad \underbrace{x[-n_0] \ x[-n_0 + 1] \ \cdots \ x[-1]}_{\text{non-causal samples}}$$

$L > N$  DFT: zeros between the causal and non-causal components can be attached

$$\underbrace{x[0] \ x[1] \ \cdots \ x[N - n_0 - 1]}_{\text{causal samples}} \quad 0 \ 0 \ \cdots \ 0 \ 0 \quad \underbrace{x[-n_0] \ x[-n_0 + 1] \ \cdots \ x[-1]}_{\text{non-causal samples}}$$

- Periodic signal:  $x[n]$  periodic of fundamental period  $N$  choose  $L = N$  (or a multiple of  $N$ ) to find DFT  $X[k]$  using FFT  
If  $L = MN$  (several periods) divide the obtained DFT by  $M$



- Frequency resolution

- $x[n]$  periodic of fundamental period  $N$ , non-zero frequency components exist only at harmonic frequencies  $\{2\pi k/N\}$

- $x[n]$  aperiodic, the number of frequency components depend on length  $L$  of DFT

To increase number of frequencies considered or **frequency resolution** of the DFT

- Aperiodic signal: increase number of samples in signal without distorting the signal by **padding with zeros**
- Periodic signal: **consider several periods and divide the DFT by number of periods used**

- **Frequency scales** N-DFT of  $x[n]$  of length  $N$ , is sequence of complex values  $X[k]$  for  $k = 0, 1, \dots, N - 1$ , or the following equivalent frequency scale

$$[0, 2\pi/N, \dots, 2\pi(N-1)/N] \text{ (rad)}$$

$$[-\pi, -(N-2)\pi/N, \dots, \pi - 2\pi/N] \text{ (rad)}$$

$$[-1, -(N-2)/N, \dots, 1 - 2/N]$$

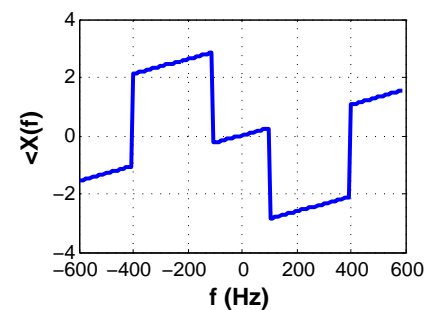
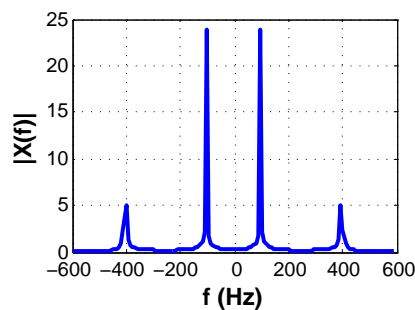
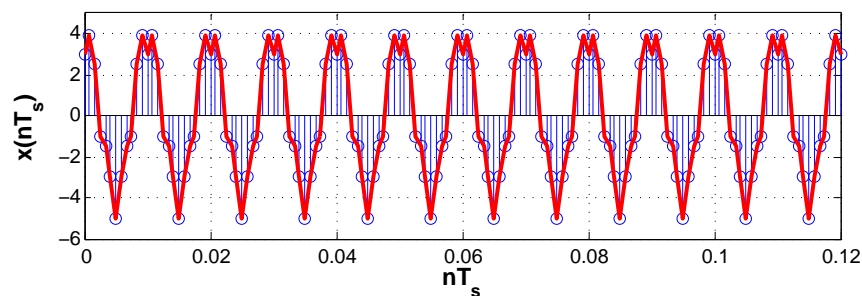
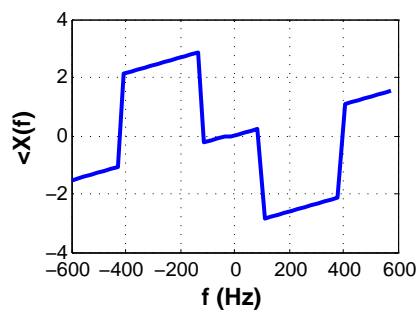
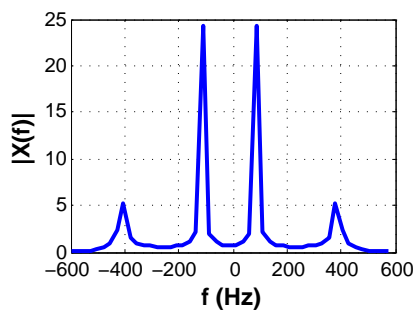
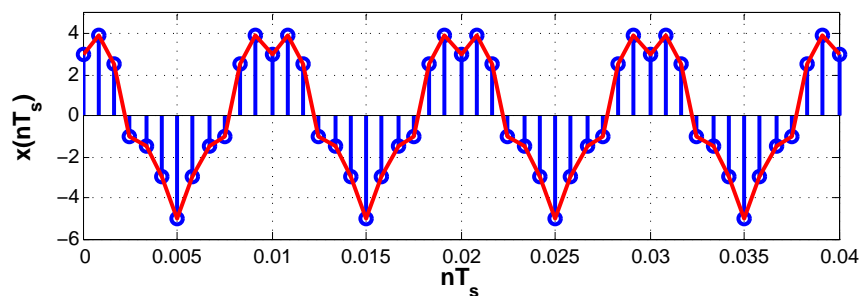
For sampled signals

$T_s$ , sampling period,  $f_s$  sampling frequency

$$\Omega = \frac{\omega}{T_s} = \omega f_s \text{ (rad/sec)} \quad \text{or} \quad f = \frac{\omega}{2\pi T_s} = \frac{\omega f_s}{2\pi} \text{ (Hz)}$$

giving scales

$$[-\pi f_s, \dots, \pi f_s] \text{ (rad/sec)} \text{ and } [-f_s/2, \dots, f_s/2] \text{ (Hz)}$$



Computation of the FFT of a periodic signal using 4 and 12 periods to improve the frequency resolution of the FFT. Notice that both magnitude and phase responses look alike, but when we use 12 periods these spectra look sharper due to the increase in the number of frequency components added.

# Linear and circular convolution

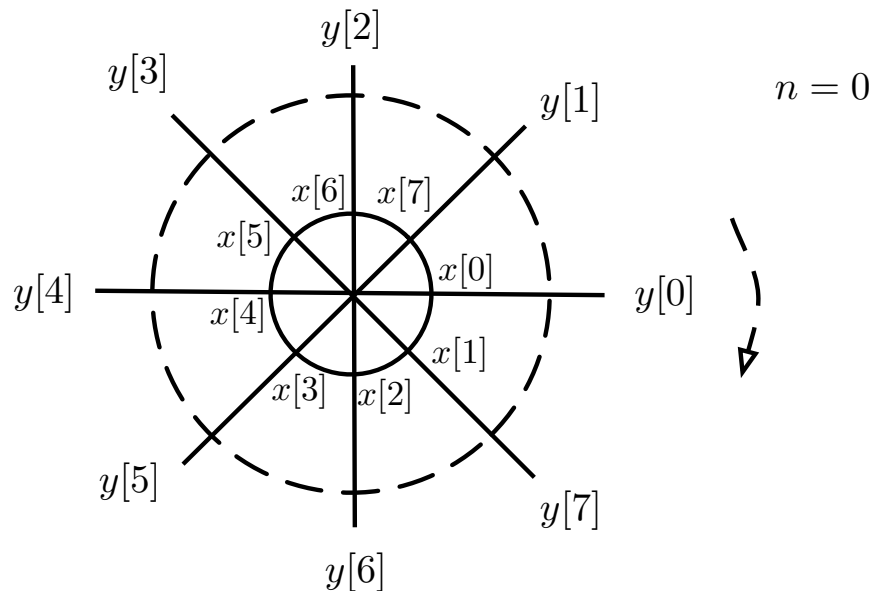
$x[n]$ , of length  $M$ , input of an LTI system with impulse response  $h[n]$  of length  $K$

- Linear convolution
  - Find DFTs  $X[k]$  and  $H[k]$  of length  $L \geq M + K - 1$  for  $x[n]$  and  $h[n]$
  - Multiply them to get  $Y[k] = X[k]H[k]$ .
  - Find the inverse DFT of  $Y[k]$  of length  $L$  to obtain  $y[n]$ .

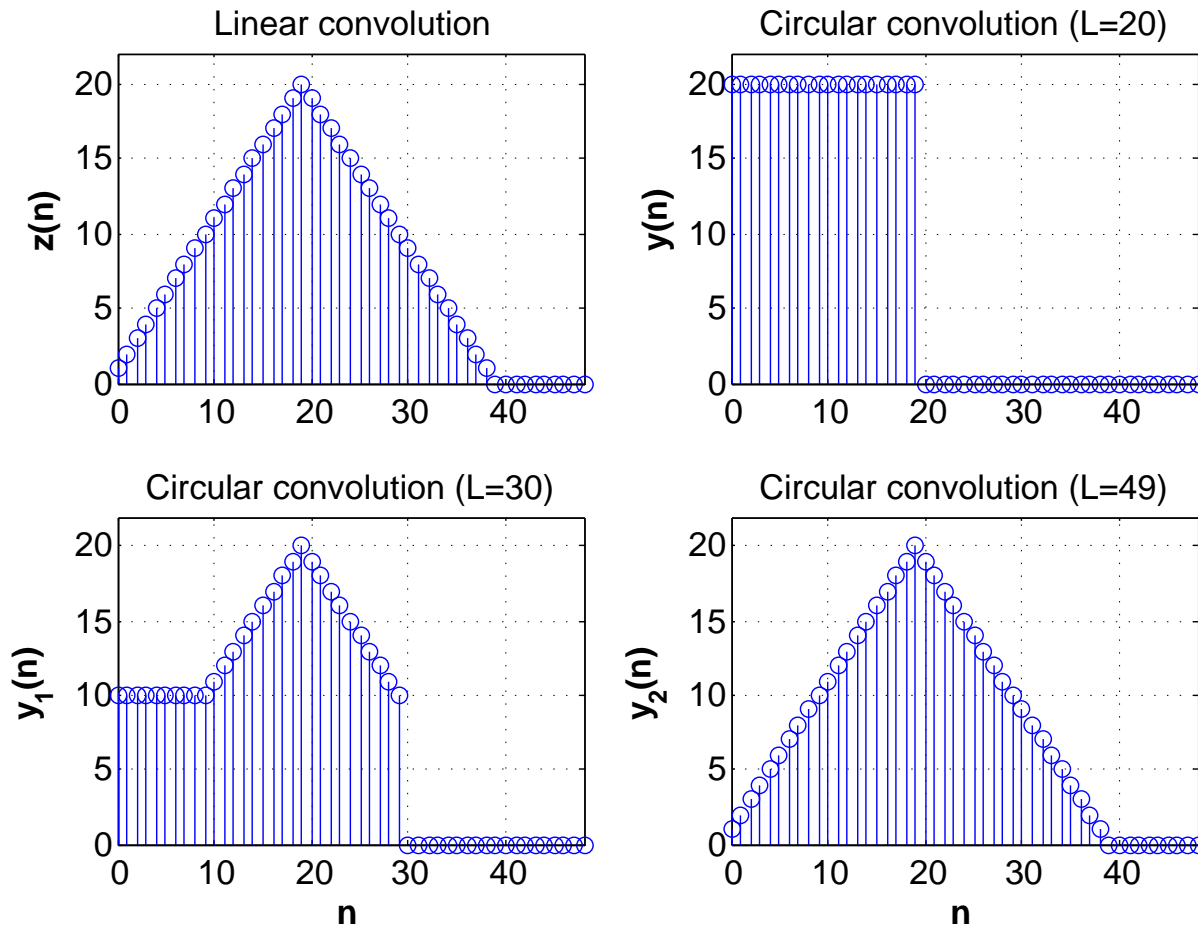
Computationally efficient using FFT

- Linear vs circular convolutions

$$Y[k] = X[k]H[k] \Leftrightarrow y[n] = (x \otimes_L h)[n] \text{ circular convolution}$$
$$\text{If } L \geq M + K - 1 \Rightarrow y[n] = (x \otimes_L h)[n] = (x * h)[n]$$



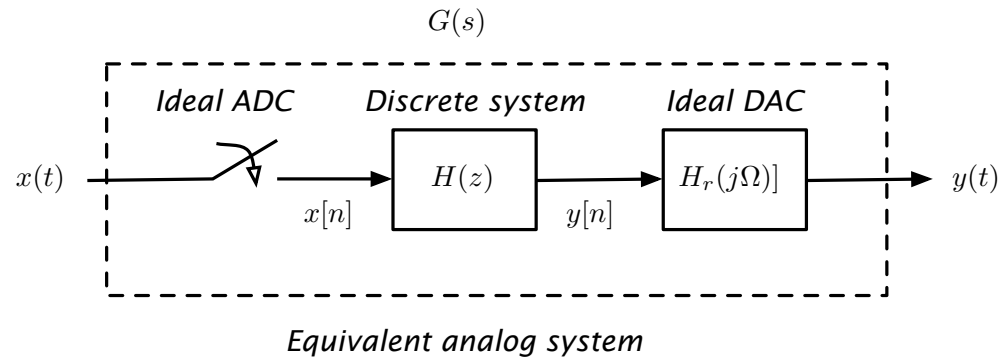
Circular convolution of length  $L = 8$  of  $x[n]$  and  $y[n]$



*Circular vs linear convolutions: Top-left: linear convolution of  $x[n]$  with itself. Top-right and bottom-left: circular convolutions of  $x[n]$  with itself of length  $L < 2N - 1$ . Bottom-right: circular convolution of  $x[n]$  with itself of length  $L > 2N - 1$  coinciding with the linear convolution.*

# The Fast Fourier Transform (FFT) algorithm

- Discrete and continuous-time signals can be processed discretely using FFT



*Discrete processing of analog signals using A/D and D/A converters.  $G(s)$  is the transfer function of the overall system, while  $H(z)$  is the transfer function of the discrete-time system.*

- Duality of DFT and IDFT – consider  $x[n]$  complex

$$X[k] = \sum_{n=0}^{N-1} x[n] W_N^{kn} \quad k = 0, \dots, N-1, \quad W_N = e^{-j2\pi/N}$$
$$x[n] = \frac{1}{N} \sum_{k=0}^{N-1} X[k] W_N^{-kn} \quad n = 0, \dots, N-1$$

- Complexity of algorithm:
  - Total number of additions and multiplications:** direct calculation of  $X[k]$ ,  $k = 0, \dots, N-1$ , from DFT requires  $N \times N$  complex multiplications, and  $N \times (N-1)$  complex additions.
  - Storage:**  $\{X[k]\}$  are complex requiring  $2N^2$  locations in memory

# Radix-2 FFT decimation-in-time algorithm

- Fundamental principle of “Divide and Conquer”
- **Periodicity:**  $W_N^{nk}$  periodic in  $n$  and  $k$  of fundamental period  $N$

$$W_N^{nk} = \begin{cases} W_N^{(n+N)k} \\ W_N^{n(k+N)} \end{cases}$$

- **Symmetry:**

$$[W_N^{nk}]^* = W_N^{(N-n)k} = W_N^{n(N-k)}$$

- **Decimation-in-time**

$$X[k] = \sum_{n=0}^{N-1} x[n] W_N^{kn} = \sum_{n=0}^{N/2-1} \left[ x[2n] W_N^{k(2n)} + x[2n+1] W_N^{k(2n+1)} \right]$$

$$W_N^{k(2n)} = e^{-j2\pi(2kn)/N} = e^{-j2\pi kn/(N/2)} = W_{N/2}^{kn}$$

$$W_N^{k(2n+1)} = W_N^k W_{N/2}^{kn}$$

$$X[k] = \sum_{n=0}^{N/2-1} x[2n] W_{N/2}^{kn} + W_N^k \sum_{n=0}^{N/2-1} x[2n+1] W_{N/2}^{kn} = Y[k] + W_N^k Z[k]$$

Computation of  $X[k]$  :

$$\begin{aligned} X[k] &= Y[k] + W_N^k Z[k] & k = 0, \dots, (N/2) - 1 \\ X[k + N/2] &= Y[k + N/2] + W_N^{k+N/2} Z[k + N/2] \\ &= Y[k] - W_N^k Z[k] & k = 0, \dots, N/2 - 1 \end{aligned}$$

Matrix form

$$\mathbf{X}_N = \begin{bmatrix} \mathbf{I}_{N/2} & \mathbf{\Omega}_{N/2} \\ \mathbf{I}_{N/2} & -\mathbf{\Omega}_{N/2} \end{bmatrix} \begin{bmatrix} \mathbf{Y}_{N/2} \\ \mathbf{Z}_{N/2} \end{bmatrix} = \mathbf{A}_1 \begin{bmatrix} \mathbf{Y}_{N/2} \\ \mathbf{Z}_{N/2} \end{bmatrix}$$

$\mathbf{I}_{N/2}$  unit matrix,  $\mathbf{\Omega}_{N/2}$  diagonal matrix with entries  $\{W_N^k, k = 0, \dots, N/2 - 1\}$

If  $N = 2^\gamma$ , repeating above process

$$\mathbf{X}_N = \left[ \prod_{i=1}^{\gamma} \mathbf{A}_i \right] \mathbf{P}_N \mathbf{x} \quad \mathbf{x} = [x[0], \dots, x[N-1]]^T$$

$\mathbf{P}_N$  permutation matrix

Number of operations of the order of  $N \log_2 N = \gamma N \ll$  the original number of order  $N^2$ .

Example: Decimation-in-time FFT algorithm for  $N = 4$

Direct computation of DFT in matrix form

$$\begin{bmatrix} X[0] \\ X[1] \\ X[2] \\ X[3] \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & W_4^1 & W_4^2 & W_4^3 \\ 1 & W_4^2 & 1 & W_4^2 \\ 1 & W_4^3 & W_4^2 & W_4^1 \end{bmatrix} \begin{bmatrix} x[0] \\ x[1] \\ x[2] \\ x[3] \end{bmatrix}$$

$$W_4^4 = W_4^{4+0} = e^{-j2\pi 0/4} = W_4^0 = 1$$

$$W_4^6 = W_4^{4+2} = e^{-j2\pi 2/4} = W_4^2$$

$$W_4^9 = W_4^{4+4+1} = e^{-j2\pi 1/4} = W_4^1$$

Number of real multiplications is  $16 \times 4$  and of real additions is  $12 \times 2 + 16 \times 2$  giving a total of 120 operations

In matrix form

$$\begin{bmatrix} X[0] \\ X[1] \\ \dots \\ X[2] \\ X[3] \end{bmatrix} = \begin{bmatrix} 1 & 0 & \vdots & 1 & 0 \\ 0 & 1 & \vdots & 0 & W_4^1 \\ \dots & \dots & \dots & \dots & \dots \\ 1 & 0 & \vdots & -1 & 0 \\ 0 & 1 & \vdots & 0 & -W_4^1 \end{bmatrix} \begin{bmatrix} Y[0] \\ Y[1] \\ \dots \\ Z[0] \\ Z[1] \end{bmatrix} = \mathbf{A}_1 \begin{bmatrix} Y[0] \\ Y[1] \\ Z[0] \\ Z[1] \end{bmatrix}$$



Repeating process

$$\begin{bmatrix} Y[0] \\ Y[1] \\ \dots \\ Z[0] \\ Z[1] \end{bmatrix} = \begin{bmatrix} 1 & 1 & \vdots & 0 & 0 \\ 1 & -1 & \vdots & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & \vdots & 1 & 1 \\ 0 & 0 & \vdots & 1 & -1 \end{bmatrix} \begin{bmatrix} x[0] \\ x[2] \\ \dots \\ x[1] \\ x[3] \end{bmatrix} = \mathbf{A}_2 \begin{bmatrix} x[0] \\ x[2] \\ x[1] \\ x[3] \end{bmatrix}.$$

The scrambled  $\{x[n]\}$  entries can be written

$$\begin{bmatrix} x[0] \\ x[2] \\ x[1] \\ x[3] \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x[0] \\ x[1] \\ x[2] \\ x[3] \end{bmatrix} = \mathbf{P}_4 \begin{bmatrix} x[0] \\ x[1] \\ x[2] \\ x[3] \end{bmatrix}$$

finally giving

$$\begin{bmatrix} X[0] \\ X[1] \\ X[2] \\ X[3] \end{bmatrix} = \mathbf{A}_1 \mathbf{A}_2 \mathbf{P}_4 \begin{bmatrix} x[0] \\ x[1] \\ x[2] \\ x[3] \end{bmatrix}$$

Number of complex additions and multiplications is now 10 (2 complex multiplications and 8 complex additions) not counting multiplications by 1 or  $-1$  Computation of the Inverse DFT

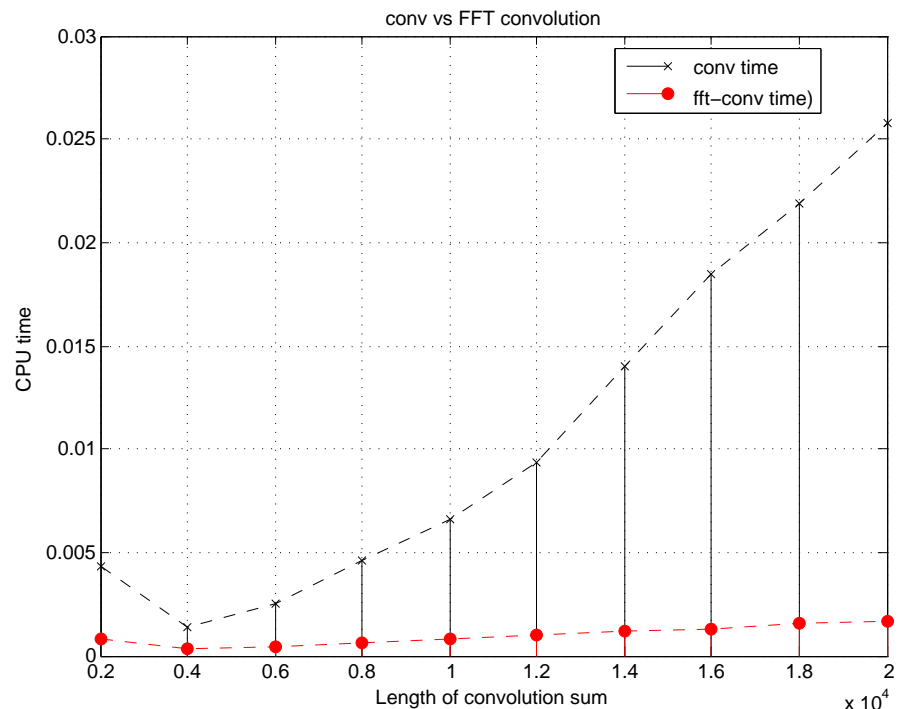
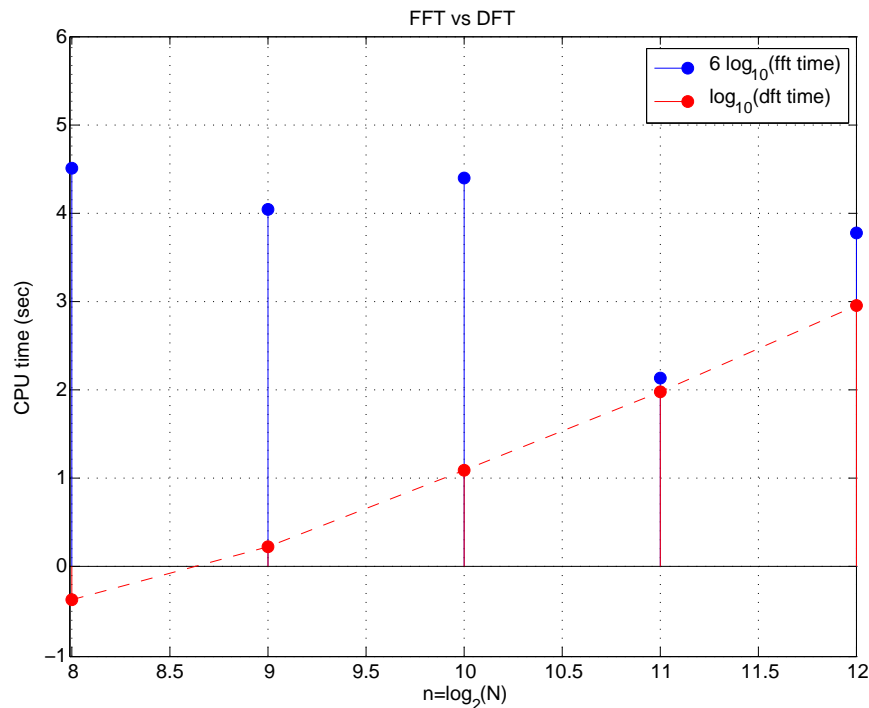
# Inverse FFT

Assuming  $x[n]$  complex

$$Nx^*[n] = \sum_{k=0}^{N-1} X^*[k] W^{nk}$$

use FFT algorithm of  $\{X^*[k]\}$  to find  $Nx^*[n]$ , compute its complex conjugate and divide by  $N$

Example: FFT and direct computation, convolution



Left: execution times for the fft and the dft functions, in logarithmic scale, used in computing the DFT of sequences of ones of increasing length  $N = 256$  to  $4096$  (corresponding to  $n = 8, \dots, 12$ ). The CPU time for the FFT is multiplied by  $10^6$ . Right: comparison of execution times of convolution of sequence of ones with itself using MATLAB's conv function and a convolution sum implemented with FFT.

## 2D Discrete Fourier Transform

- Discrete 2D sequence  $\tilde{x}[m, n]$  periodic with  $(N_1, N_2) \rightarrow$  Fourier series

$$\tilde{x}[m, n] = \frac{1}{N_1 N_2} \sum_{k=0}^{N_1-1} \sum_{\ell=0}^{N_2-1} \tilde{X}(k, \ell) e^{j2\pi(mk/N_1 + n\ell/N_2)}$$

- Discrete Fourier coefficients periodic with the same period:  $(N_1, N_2)$  ;

$$\tilde{X}(k, \ell) = \sum_{m=0}^{N_1-1} \sum_{n=0}^{N_2-1} \tilde{x}[m, n] e^{-j2\pi(mk/N_1 + n\ell/N_2)}$$

correspond to the harmonic frequencies  $(2\pi k/N_1, 2\pi \ell/N_2)$  for  $k = 0, \dots, N_1$  and  $\ell = 0, \dots, N_2$ .

- A non-periodic signal  $x[m, n]$  with a finite support  $\Rightarrow$  assume  $x[m, n]$  is one period of an extended periodic signal  $\tilde{x}[m, n]$
- Find Fourier coefficients  $\tilde{X}(k, \ell)$  of  $\tilde{x}[m, n]$ , use one period of them as 2D-DFT  $X(k, \ell)$  of  $x[m, n]$

**Example:** 2D-DFT  $X(k, \ell)$  and its inverse of  $x[m, n]$  with finite support in the first quadrant  $0 \leq m \leq M - 1$ ,  $0 \leq n \leq N - 1$

- Take  $N_1 \geq M$  and  $N_2 \geq N$  (adding zeros to  $x[m, n]$  when  $N_1 > M$  and  $N_2 > N$ )
- Extended periodic signal  $\tilde{x}[m, n]$  of rectangular period  $(N_1, N_2)$

$$\tilde{x}[m, n] = \begin{cases} x[m, n] & 0 \leq m \leq M - 1, 0 \leq n \leq N - 1 \\ 0 & M \leq m \leq N_1 - 1, N \leq n \leq N_2 - 1. \end{cases}$$

- The discrete Fourier series of  $\tilde{x}[m, n]$  has coefficients  $\{\tilde{X}(k, \ell)\}$
- Two-dimensional Discrete Fourier Transform (2D-DFT)  $\{X(k, \ell)\}$  of the non-periodic signal  $x[m, n]$
- Inverse DFT of  $\{X(k, \ell)\}$ ,  $0 \leq k \leq N_1 - 1$ ,  $0 \leq \ell \leq N_2 - 1$  will give the original signal padded with zeros
- To improve resolution  $\Rightarrow$  make  $N_1$  and  $N_2$  much larger than  $M$  and  $N$
- For periodic signals  $\Rightarrow$  take  $N_1 = uM_1$  and  $N_2 = vM_2$  multiples of the periods
- Computation of the 2D-DFT is done efficiently by 2D-FFT algorithm

- 2D convolution of  $x[m, n]$ ,  $0 \leq m \leq M - 1$ ,  $0 \leq n \leq N - 1$  and  $h[m, n]$ ,  $0 \leq m \leq K - 1$ ,  $0 \leq n \leq L - 1$  gives

$$y[m, n] = (x * h)[m, n], \quad 0 \leq m \leq M + K - 1, \quad 0 \leq n \leq N + L - 1$$

- Take 2D-DFT of size  $\geq (M + K) \times (N + L)$  of  $x[m, n]$  and  $h[m, n] \Rightarrow$

$$y[m, n] = (x * h)[m, n] \Leftrightarrow Y(k, \ell) = X(k, \ell)H(k, \ell)$$

- 2D sequence  $x[m, n]$  with 2D-DFT  $X(k, \ell)$  (size  $N_1 \times N_2$ )  $\Rightarrow$  shifting in space

$$\text{2D-DFT}(x[m - u, n - v]) = e^{-j2\pi uk/N_1} e^{-j2\pi v\ell/N_2} X(k, \ell)$$

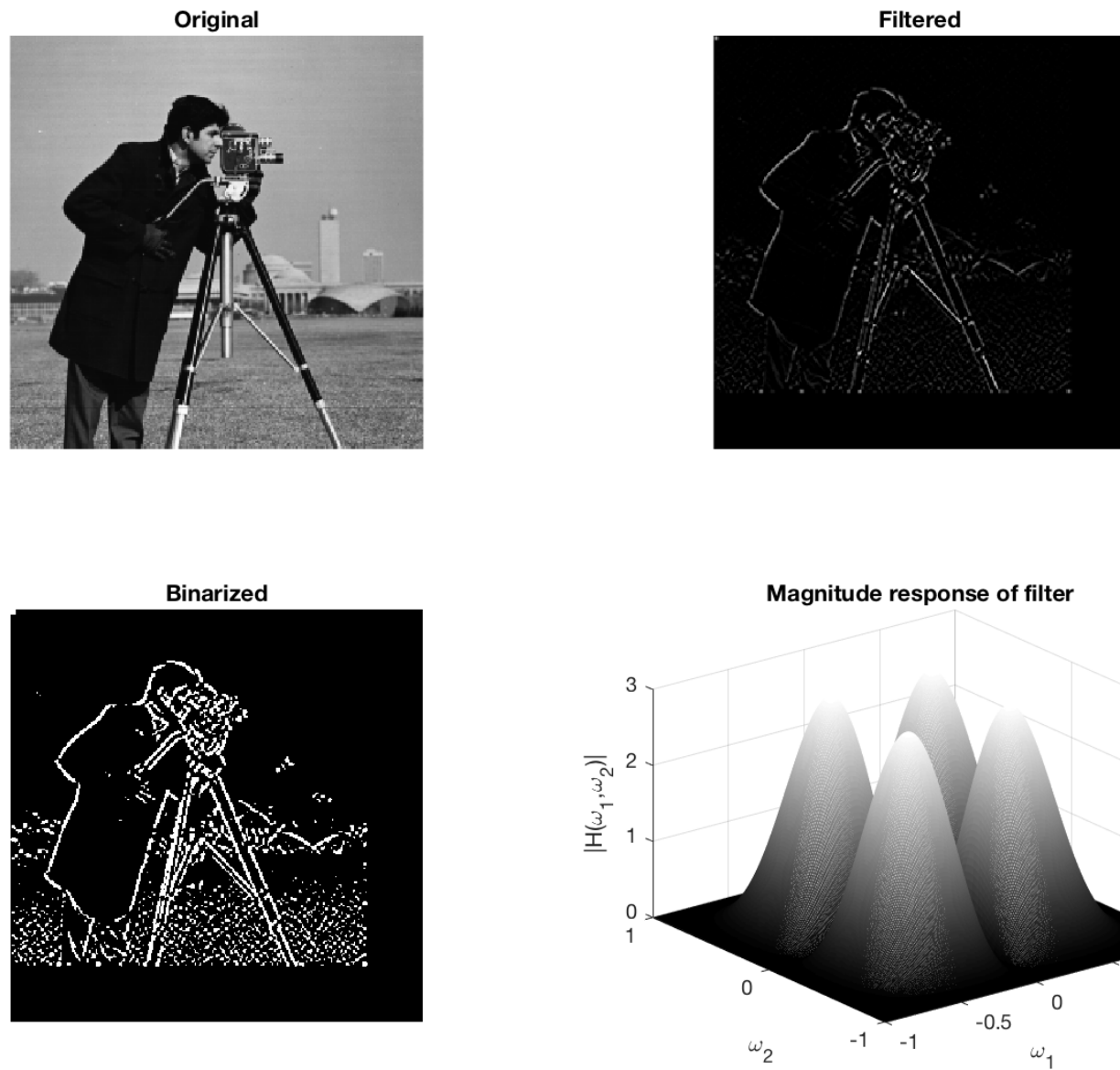
**Example:** Filtering the image *cameraman* with

$$h_1[m, n] = \frac{1}{3} (\delta[m, n] + 2\delta[m-1, n] + \delta[m-2, n] - \delta[m, n-2]) \\ - \frac{1}{3} (2\delta[m-1, n-2] - \delta[m-2, n-2])$$

and the result filtered with

$$h_2[m, n] = \frac{1}{3} (\delta[m, n] - \delta[m-2, n] + 2\delta[m, n-1] - 2\delta[m-2, n-1]) \\ - \frac{1}{3} (\delta[m, n-2] - \delta[m-2, n-2])$$

- Use the 2D-FFT of size: first filtering  $(256 + 3 - 1) \times (256 + 3 - 1) = 258 \times 258$
- Second filtering:  $(258 + 3 - 1) \times (258 + 3 - 1) \Rightarrow$  choose  $300 \times 300$
- Make filtered image into binary: by making it positive then thresholding it so that values bigger than 1.1 are converted into 1 and the rest are zero



**Figure:** Cascade filtering and binarization of the image cameraman. Clockwise from top left: original and filtered images, magnitude response of cascaded filter and binarized image.

- Transform domain coding → real valued coefficients: DCT

$$C_x(k) = \left(\frac{2}{N}\right)^{1/2} \sum_{n=0}^{N-1} \Lambda(n) x[n] \cos\left(\frac{\pi k}{2N}(2n+1)\right)$$
$$\Lambda(n) = \begin{cases} 1/\sqrt{2} & n=0 \\ 1 & \text{otherwise} \end{cases}$$

- Extension to 2D: 2D-DCT

$$C_x(k, \ell) = \left(\frac{4}{MN}\right)^{1/2} \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} \Lambda(m, n) x[m, n] \cos\left(\frac{\pi k}{2M}(2m+1)\right) \cos\left(\frac{\pi \ell}{2N}(2n+1)\right)$$

- $\Lambda(m, n) = \Lambda(m)\Lambda(n)$



**Example:** Compression of *cameraman* image: 2D-DCT coefficients bigger than 0.1 in absolute value

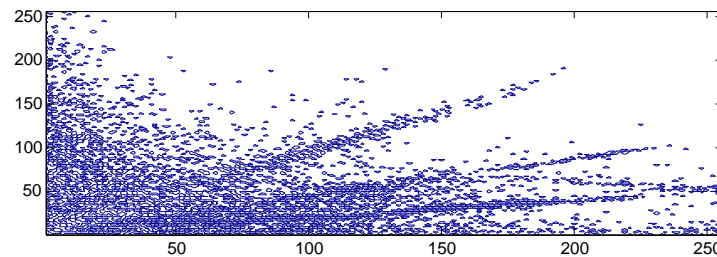
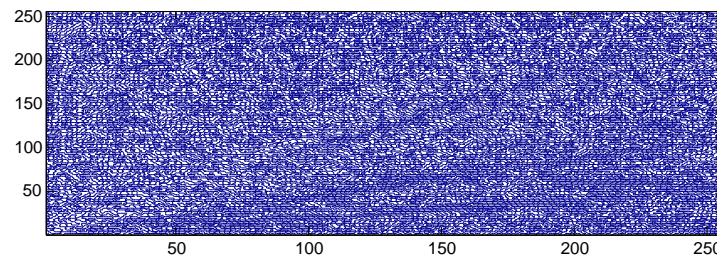


Figure: Compression using thresholded DCT values: original image (left), compressed image (center), support of original 2D-DCT values (top right) and thresholded 2D-DCT values (bottom right).