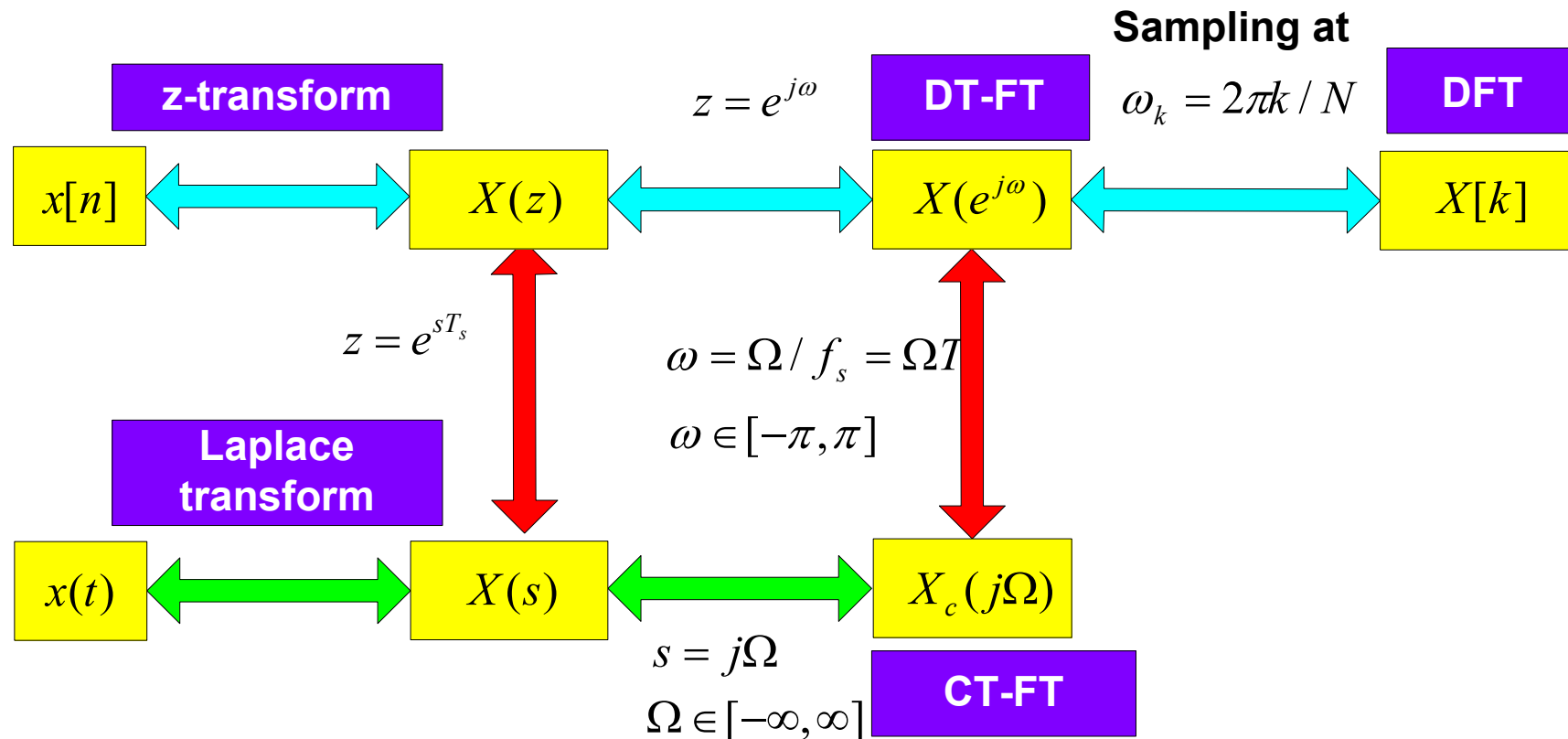


RELATIONSHIP BETWEEN THE TRANSFORMATIONS



If $x(t)$ is bandlimited and the sampling rate is greater than the Nyquist rate, then $x(t)$ can be recovered from $x[n]$. Anti-aliasing filter with bandwidth f_{\max} has to be applied to $x(t)$ to avoid aliasing.

$$X(e^{j\omega}) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c\left(j\frac{\omega}{T} - j\frac{2\pi k}{T}\right) \quad (10.1)$$

Cascade form of LTI system

Factor the numerator and denominator polynomials of $H(z)$ as

$$H(z) = A \frac{\prod_{k=1}^{M_1} (1 - g_k z^{-1}) \prod_{k=1}^{M_2} (1 - h_k z^{-1})(1 - h_k^* z^{-1})}{\prod_{k=1}^{N_1} (1 - c_k z^{-1}) \prod_{k=1}^{N_2} (1 - d_k z^{-1})(1 - d_k^* z^{-1})}$$

The **first-order factors** represent real zeros at g_k and c_k real poles, and the **second-order factors** represent **complex conjugate pairs of zeros** at h_k and h_k^* and complex conjugate pairs of poles at d_k and d_k^* .

A modular structure is usually preferred and is obtained by combining pairs of real factors and complex conjugate pairs onto second-order factors.

$$H(z) = \prod_{k=1}^{N_s} \frac{b_{0k} + b_{1k} z^{-1} + b_{2k} z^{-2}}{1 - a_{1k} z^{-1} - a_{2k} z^{-2}}$$

where $N_s = \lfloor (N + 1) / 2 \rfloor$ is the largest integer contained in $(N + 1) / 2$.

We can implement a cascade structure with a minimum number of multiplications and a minimum number of delay elements if we use the **direct form II** structure (interchange the feedforward and feedback parts of the 2nd order section and combine the delays) for each second-order section.

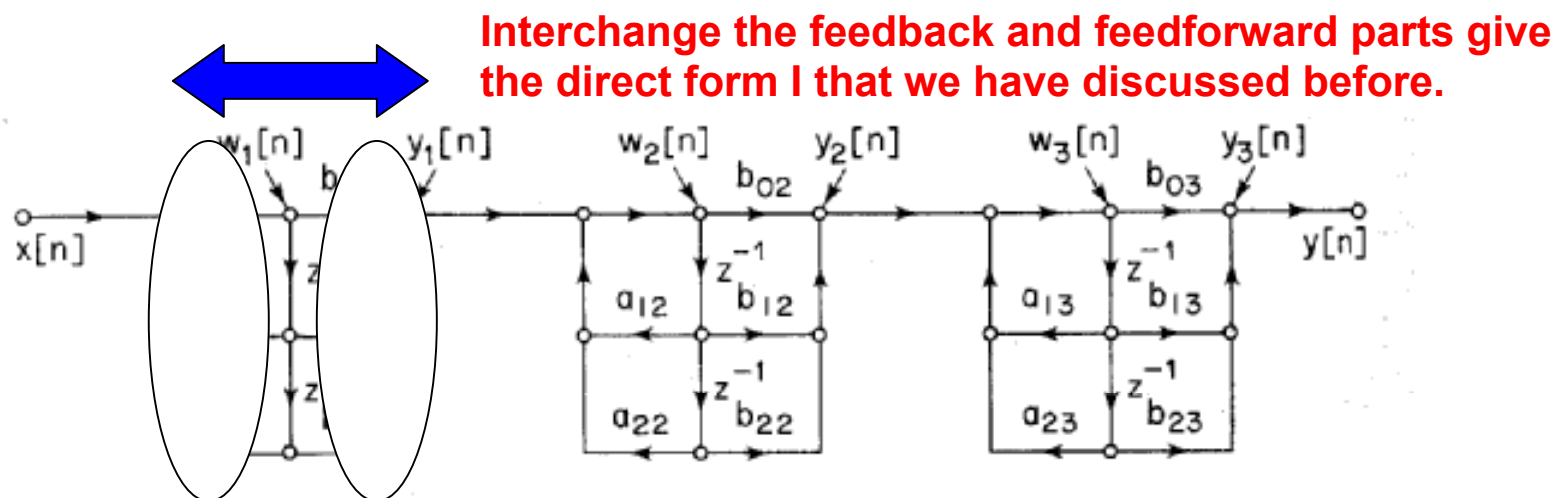


Figure 6.14 Cascade structure for a sixth-order system with a direct form II realization of each second-order subsystem.

A variety of equivalent systems can be obtained by pairing the poles and zeros in different ways and by ordering the second-order sections in different ways.

THE DISCRETE-TIME AND DISCRETE FOURIER TRANSFORMS

CONTENTS

FREQUENCY RESPONSE OF LTI SYSTEMS

DISCRETE-TIME FOURIER TRANSFORM

DISCRETE FOURIER TRANSFORM

9. Frequency response of LTI systems

The response of an LTI system to a sinusoidal input is sinusoidal with the same frequency and the amplitude and phase are determined by the LTI system.

Consider a sinusoidal input sequence: $x[n] = e^{j\omega n}$. The output of an LTI system with an impulse response $h[n]$ is

$$y[n] = \sum_{k=-\infty}^{\infty} h[k]e^{j\omega(n-k)} = e^{j\omega n} \left[\sum_{k=-\infty}^{\infty} h[k]e^{-j\omega k} \right] = e^{j\omega n} H(e^{j\omega})$$

where
$$H(e^{j\omega}) = \sum_{k=-\infty}^{\infty} h[k]e^{-j\omega k} \quad (9.1)$$

is called the frequency response of the system.

■ (9.1) is the discrete-time Fourier transform (DT-FT) of $h[n]$.

- The DT-FT is obtained by substituting $z = e^{j\omega}$ in $H(z)$, i.e. evaluating the z-transform at the unit circle.
- A **sufficient condition for convergence** of the frequency response (DT-FT of $h[n]$) is that **$h[n]$ is absolutely summable**. In other words, the system must be **stable** (c.f. (6.1)). Consequently, the **ROC of $H(z)$ covers the unit circle**.

In general,

$$Y(e^{j\omega}) = H(e^{j\omega})X(e^{j\omega}), \quad (8.7)$$

Example: Transient and steady-state responses.

Consider a LTI system defined by

$$y[n] = a_1 y[n-1] + b_0 x[n]. \quad \text{causal 用以确定ROC} \quad (\text{e1})$$

Taking the z-transform on both sides, one gets

$$Y(z) = a_1 z^{-1} Y(z) + b_0 X(z). \quad (\text{e2})$$

The transfer function is then given by

$$H(z) = \frac{Y(z)}{X(z)} = \frac{b_0}{1 - a_1 z^{-1}}. \text{ ROC } |z| > |a_1|. \quad (\text{e3})$$

Taking the inverse z-transform and noting that the system (e1) is causal, one gets the impulse response $h[n]$ as follows

$$h[n] = b_0 a_1^n u[n]. \quad (\text{e4})$$

Assuming $|a_1| < 1$, so that the frequency response of the system exists (the system is stable). It is then obtained by evaluating the z-transform in (e3) on the unit circle:

$$H(e^{j\omega}) = H(z) \big|_{z=e^{j\omega}} = \frac{b_0}{1 - a_1 e^{-j\omega}}. \quad (\text{e5})$$

Suppose that the input to the system is a sinusoids starting at $n=0$:

$$x[n] = e^{j\omega_0 n} u[n]. \quad (\text{e6})$$

We want to determine the output of the system. Taking the z-transform of (e6), one gets

$$X(z) = \frac{1}{1 - e^{j\omega_0} z^{-1}}. \quad \text{ROC } |z| > 1. \quad (\text{e7})$$

Using the convolution theorem, the z-transform of the output is the product of $H(z)$ and $X(z)$:

$$Y(z) = H(z)X(z) = \frac{b_0}{1 - a_1 z^{-1}} \cdot \frac{1}{1 - e^{j\omega_0} z^{-1}}. \quad \text{ROC } |z| > 1. \quad (\text{e8})$$

Using partial fraction expansion, we have

$$Y(z) = \frac{A_0}{1 - a_1 z^{-1}} + \frac{A_1}{1 - e^{j\omega_0} z^{-1}}. \quad \text{ROC } |z| > 1, \quad (\text{e9})$$

where $A_0 = \frac{b_0 a_1}{a_1 - e^{j\omega_0}}$ and $A_1 = \frac{b_0}{1 - a_1 e^{-j\omega_0}}$. Taking the inverse z-transform,

one get the desired output:

$$y[n] = \overset{\text{正比于系统冲击响应}}{A_0 a_1^n u[n]} + \overset{\text{正比于输入响应}}{A_1 e^{jn\omega_0} u[n]}$$

transient response steady state response

(e10)

The **first term** is proportional to the impulse response $h[n]$, which is **solely determined by the poles of the systems**. For stable system, it dies down as n tends to infinity. Therefore it is called the **transient response**. **The second**

term is proportional to the input complex exponential and $A_1 = \frac{b_0}{1 - a_1 e^{-j\omega_0}}$ is

the **frequency response of the system at $\omega = \omega_0$** . It is called the **steady-state component** of the output.

Exercise:

If $b_0=5$, $a_1=-0.8$, and $\omega_0 = 2\pi / 10$. Determine the transient and steady state component of the above example.

[Answer:**Transient:**

$$y_t[n] = \left(\frac{-4}{-0.8 - e^{j0.2\pi}} \right) (-0.8)^n u[n].$$

Steady-state:

$$y_s[n] = \left(\frac{5}{1 + 0.8e^{-j0.2\pi}} \right) e^{j0.2n} u[n].]$$

9.1 Magnitude and Phase responses

In general, $H(e^{j\omega})$ is **complex** and it can be expressed in terms of its **real and imaginary parts** as:

$$H(e^{j\omega}) = H_R(e^{j\omega}) + jH_I(e^{j\omega}). \quad (9.2)$$

or in terms of **magnitude and phase** as

$$H(e^{j\omega}) = |H(e^{j\omega})| e^{j\angle H(e^{j\omega})}. \quad (9.3)$$

$|H(e^{j\omega})|$ is the magnitude response or the gain of the system and $\angle H(e^{j\omega})$ is the phase response or **phase shift** of the system.

9.1.1 Magnitude and Phase responses

The magnitude and phase responses can also be expressed in terms of the poles and zeros as follows:

$$H(z) \Big|_{z=e^{j\Omega}} = \frac{b_0 \prod_{k=1}^M (1 - c_k z^{-1})}{a_0 \prod_{k=1}^N (1 - d_k z^{-1})} \Bigg|_{z=e^{j\Omega}} = \frac{b_0 \prod_{k=1}^M (1 - c_k e^{-j\omega})}{a_0 \prod_{k=1}^N (1 - d_k e^{-j\omega})} \quad (9.4)$$

c_k and d_k are the zeros and poles of $H(z)$.

■ The magnitude-squared function is

$$\left| H(e^{j\omega}) \right|^2 = H(e^{j\omega}) H^*(e^{j\omega}) = \left(\frac{b_0}{a_0} \right)^2 \frac{\prod_{k=1}^M (1 - c_k e^{-j\omega})(1 - c_k^* e^{j\omega})}{\prod_{k=1}^N (1 - d_k e^{-j\omega})(1 - d_k^* e^{j\omega})} \quad (9.5)$$

Log magnitude of $H(e^{j\omega})$ (in decibels dB) or gain in dB as follows:

$$20 \log_{10} |H(e^{j\omega})| = 20 \log_{10} \left| \frac{b_0}{a_0} \right| + \sum_{k=1}^M 20 \log_{10} |1 - c_k e^{-j\omega}| - \sum_{k=1}^N 20 \log_{10} |1 - d_k e^{-j\omega}| \quad (9.6)$$

Zero dB corresponds to $|H(e^{j\omega})| = 1$ while $|H(e^{j\omega})| = 10^m$ is 20m dB.

$20 \log_{10} |H(e^{j\omega})|$ is negative when $|H(e^{j\omega})| < 1$.

■ **The phase response is**

$$\angle H(e^{j\omega}) = \angle \left(\frac{b_0}{a_0} \right) + \sum_{k=1}^M \angle (1 - c_k e^{-j\omega}) - \sum_{k=1}^N \angle (1 - d_k e^{-j\omega}) \quad (9.7)$$

The principal value of the phase function can be computed as:

$$\text{ARG}[H(e^{j\omega})] = \arctan \left[\frac{H_R(e^{j\omega})}{H_I(e^{j\omega})} \right] \quad (9.8)$$

Since

$$-\pi < \text{ARG}[H(e^{j\omega})] < \pi \quad (9.9)$$

$$\angle H(e^{j\omega}) = \text{ARG}[H(e^{j\omega})] + 2\pi \cdot r(\omega), \quad (9.10)$$

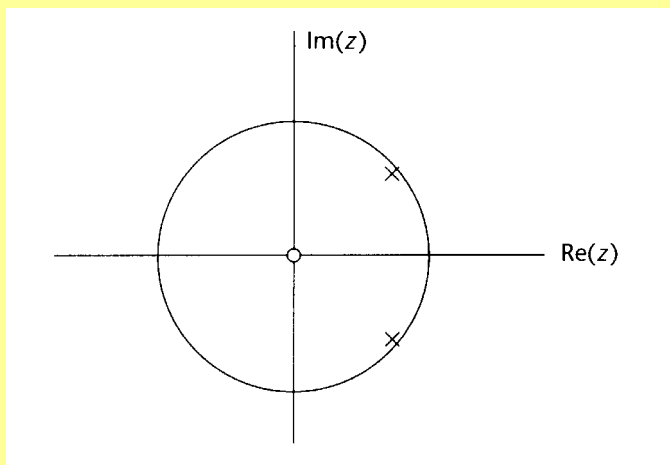
where $r(\omega)$ is a positive or negative integer that can be different at each value of ω . Alternatively, $\text{ARG}[H(e^{j\omega})]$ can be obtained from taking the principal value of each term in (5.3.5):

$$\begin{aligned} \text{ARG}[H(e^{j\omega})] = & \text{ARG}\left[\frac{b_0}{a_0}\right] + \sum_{k=1}^M \text{ARG}[1 - c_k e^{-j\omega}] \\ & - \sum_{k=1}^N \text{ARG}[1 - d_k e^{-j\omega}] + 2\pi \cdot r(\omega) \end{aligned} \quad (9.11)$$

Except at the discontinuities of $\text{ARG}[H(e^{j\omega})]$ corresponding to jumps between $+\pi$ and $-\pi$.

Examples

Frequency response of a LTI system with poles at $0.95\angle \pm 45^\circ$ and a zero at the origin. Sampling period is 1ms (sampling rate 1kHz).



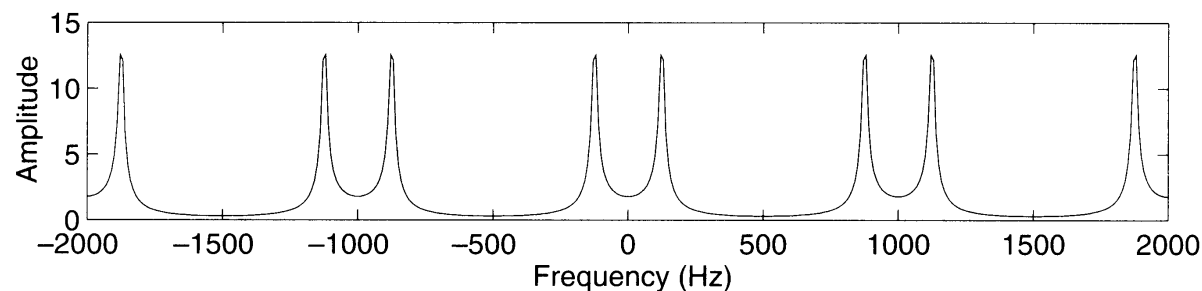
Pole-zero plot

X – poles

O – zero

Note the peaks in the frequency response near the poles.

Fig. 14.19
Predicted
amplitude
response

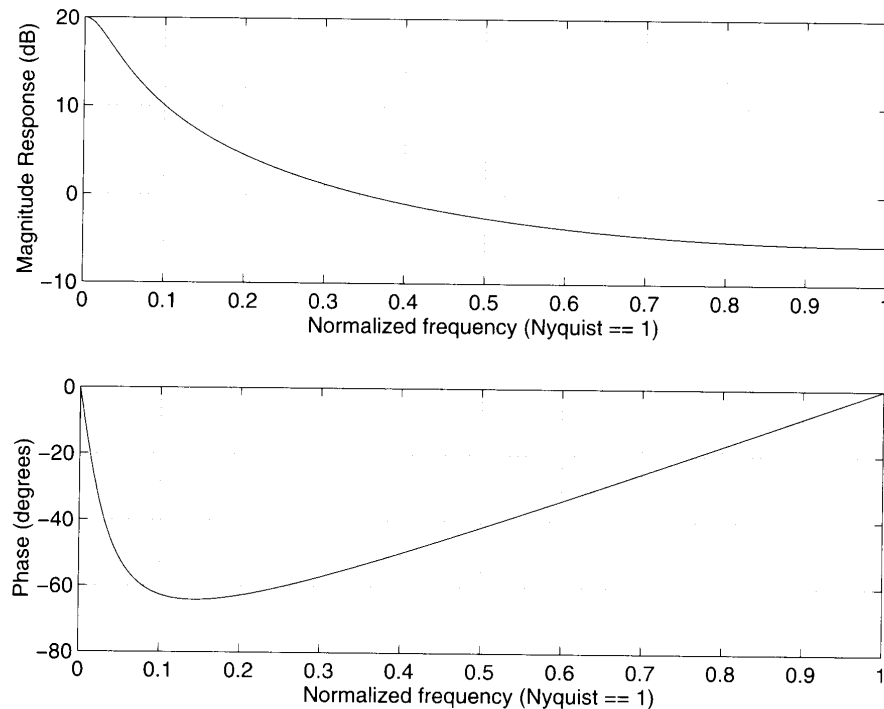


$$\omega = -\pi \quad \omega = \pi$$

Examples 14.7 and 14.8 (in textbook)

$H(z) = 1/(1 - 0.9z^{-1})$. The sampling frequency is 10 kHz.

Fig. 14.28
Frequency
response plot
using MATLAB

**MATLAB COMMANDS**

```
b=[1 0];  
a=[1 -0.9];  
subplot(2,2,1);  
zplane(b,a)  
freqz(b,a)
```

$$\omega = \pi$$

$$\Omega = 2\pi \times 5000$$

9.3 Discrete-time Fourier transform (DT-FT)

The **discrete-time Fourier transform** pair is defined as follows:

$$X(e^{j\omega}) = \sum_{n=-\infty}^{\infty} x[n]e^{-j\omega n} \quad \text{(DT-FT)} \quad (9.12)$$

$$x[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{j\omega})e^{j\omega n} d\omega \quad \text{(Inverse DT-FT)} \quad (9.13)$$

- The DT-FT represents the frequency components of $x[n]$ at digital radian frequency ω .
- The inverse DT-FT synthesizes the sequence $x[n]$ from infinitesimally small complex sinusoids of form.

$$\frac{1}{2\pi} X(e^{j\omega})e^{j\omega n} d\omega$$

To show that they are inverses of each other, substitute (9.13) into (9.12), we obtain

$$\tilde{x}[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} \left[\sum_{m=-\infty}^{\infty} x[m] e^{-j\omega m} \right] e^{j\omega n} d\omega = \sum_{m=-\infty}^{\infty} x[m] \left[\frac{1}{2\pi} \int_{-\pi}^{\pi} e^{j\omega(n-m)} d\omega \right] \quad (9.14)$$

Since

$$\frac{1}{2\pi} \int_{-\pi}^{\pi} e^{j\omega(n-m)} d\omega = \begin{cases} 1, & m = n \\ 0, & m \neq n \end{cases} = \delta[n - m],$$

we obtain the desired result.

$$\tilde{x}[n] = \sum_{m=-\infty}^{\infty} x[m] \delta[n - m] = x[n]$$

Relationship between continuous time Fourier transform and DT-FT (see section 10, eqn. (10.1)).

9.4 Discrete Fourier transform (DFT)

Consider the DT-FT of a **finite length sequence**: $x[n]$, $n=0,\dots,N-1$.

$$X(e^{j\omega}) = \sum_{n=0}^{N-1} x[n]e^{-j\omega n} \quad (\text{DT-FT})$$

Sampling Ω regularly at $\omega_k = 2\pi k / N$ (spacing $2\pi / N$), $k=0,1,\dots,N-1$, we obtain the **discrete Fourier transform (DFT)**.

$$X[k] = X(e^{j\omega_k}) = \sum_{n=0}^{N-1} x[n]e^{-j(2\pi nk / N)} \quad (9.15) \quad (\text{DFT})$$

Examples: Compute the DFT of the following sequence

$$x[n] = \begin{cases} 1 & n = 0, \dots, 4 \\ 0 & n = 5, \dots, N-1 \end{cases}, \quad X[k] = \sum_{n=0}^4 e^{-j(2\pi nk / N)} = \frac{1 - e^{-j5(2\pi k / N)}}{1 - e^{-j(2\pi k / N)}}$$

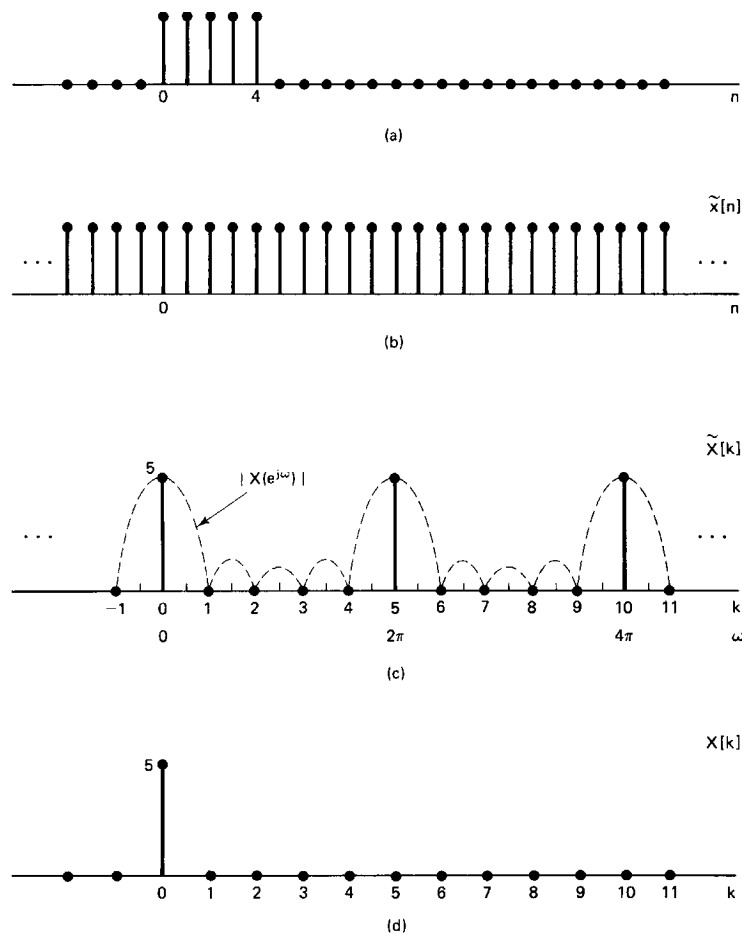


Figure 8.10 Illustration of the DFT. (a) Finite-length sequence $x[n]$. (b) Periodic sequence $\tilde{x}[n]$ formed from $x[n]$ with period $N = 5$. (c) Fourier series coefficients $\tilde{X}[k]$ for $\tilde{x}[n]$. To emphasize that the Fourier series coefficients are samples of the Fourier transform, $|X(e^{j\omega})|$ is also shown. (d) DFT of $x[n]$.

N=5

■ DFT as uniform samples of the DT-FT in the frequency domain (spacing $2\pi / N$).

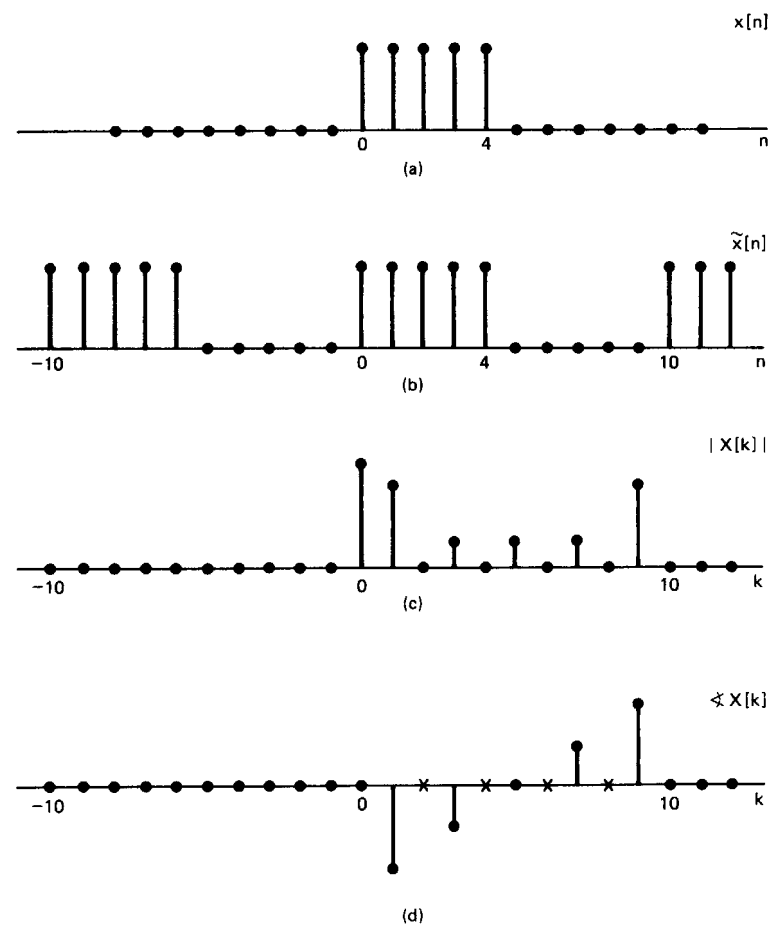


Figure 8.11 Illustration of the DFT. (a) Finite-length sequence $x[n]$. (b) Periodic sequence $\tilde{x}[n]$ formed from $x[n]$ with period $N = 10$. (c) DFT magnitude. (d) DFT phase. (x's indicate indeterminate value.)

N=11

9.4.1 Inverse Discrete Fourier transform (IDFT)

DFT is an **orthogonal transformation** and it has a simple inversion formula (the **inverse DFT**):

$$x[n] = \frac{1}{N} \sum_{k=0}^{N-1} X[k] e^{j(2\pi nk / N)} \quad (9.16)$$

- It is identical to the DFT, except for the scaling factor (1/N) and the kernel $e^{j(2\pi nk / N)}$ (note the negative sign in the DFT).
- DFT supports a kind of convolution called “**circular convolution**” and it can be used to compute discrete-time convolution (i.e. real-time filtering using FIR filters). Fast algorithms for DFT called **fast Fourier transform** (FFT) with order $O(N \log_2 N)$ arithmetic complexity are available and they found many applications. (MATLAB COMMAND: **X=fft(x, N)**, N – transform length).

Exercises:

1. By substituting (9.15) into (9.16), verify that (9.16) is the inverse of the

DFT [Hint: use the identity: $x[n] = \sum_{k=0}^{N-1} e^{j(2\pi k / N)(n-m)} = N \cdot \delta[n - m]$].

2. Verify (9.16) and compute its inverse DFT.
3. Show that a finite length sequence of length N can be represented by its uniform samples of DT-FT at $\omega_k = 2\pi k / N$, $k=0,1,\dots,N-1$. [Hint: the DFT is reversible with the inverse transformation given by the IDFT.]