

# Signals and Systems

## 1. Continuous-Time and Discrete-Time Signals

continuous-time signals  $x(t)$

discrete-time signals  $x[n]$

### 1.1 Signal Energy and Power

the total energy

$$E_{\infty} \triangleq \lim_{T \rightarrow \infty} \int_{-T}^T |x(t)|^2 dt = \int_{-\infty}^{\infty} |x(t)|^2 dt$$

$$E_{\infty} \triangleq \lim_{N \rightarrow \infty} \sum_{n=-N}^{+N} |x[n]|^2 = \sum_{n=-\infty}^{+\infty} |x[n]|^2$$

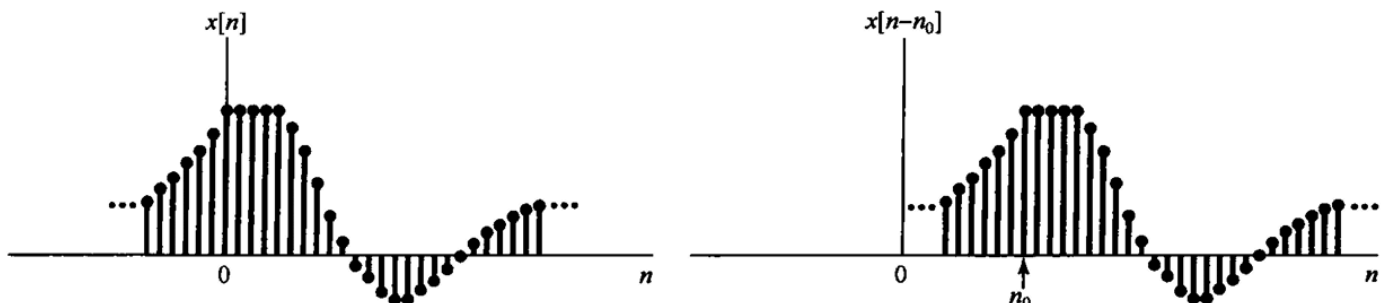
the time-averaged power

$$P_{\infty} \triangleq \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T |x(t)|^2 dt$$

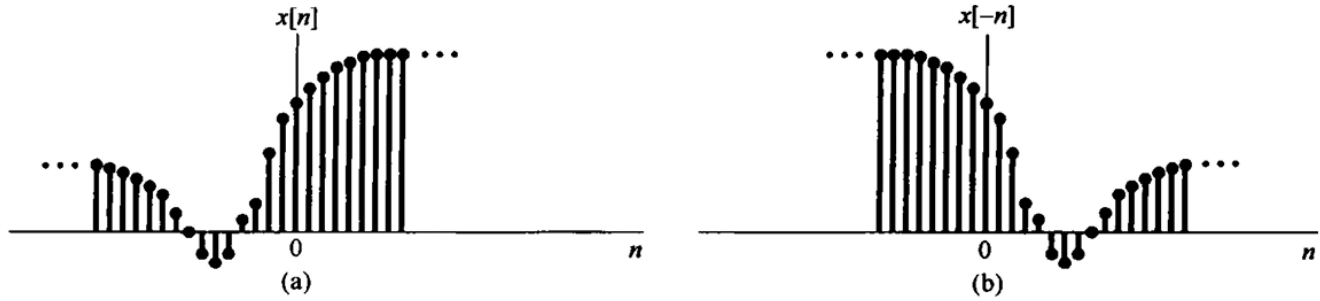
$$P_{\infty} \triangleq \lim_{N \rightarrow \infty} \frac{1}{2N+1} \sum_{n=-N}^{+N} |x[n]|^2$$

## 2. Transformations of the Independent Variable

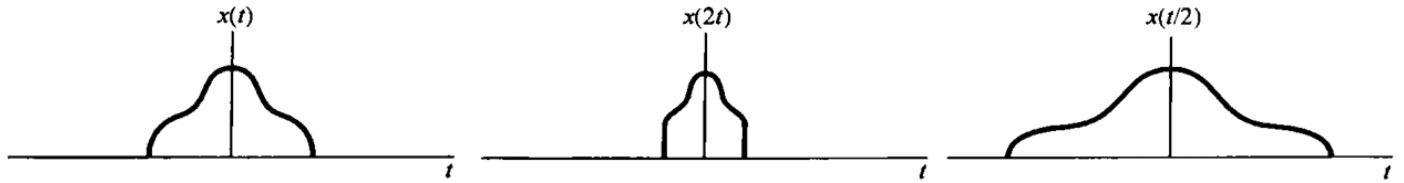
time shift  $x[n]$  and  $x[n - n_0]$



time reversal  $x[-n]$



**time scaling**  $x(2t)$  and  $x(\frac{t}{2})$



It is often of interest to determine the effect of transforming the independent variable of  $x(t)$  to  $x(\alpha t + \beta)$ , where  $\alpha$  and  $\beta$  are given numbers.

## 2.1 Periodic Signals

$$x(t) = x(t + T)$$

The **fundamental period**  $T_0$  of  $x(t)$  is the smallest positive value of  $T$  for which it holds.

$$x[n] = x[n + N]$$

The **fundamental period**  $N_0$  of  $x[n]$  is the smallest positive value of  $N$  for which it holds.

A signal that is not periodic will be referred to as an **aperiodic** signal.

## 2.2 Even and Odd Signals

**even** signals  $x(-t) = x(t)$  and  $x[-n] = x[n]$

**odd** signals  $x(-t) = -x(t)$  and  $x[-n] = -x[n]$

any signal can be broken into a sum of two signals, one of which is even and one of which is odd

the **even part** of  $x(t)$ :  $\mathcal{E}_v\{x(t)\} = \frac{1}{2}[x(t) + x(-t)]$

the **odd part** of  $x(t)$ :  $\mathcal{O}_d\{x(t)\} = \frac{1}{2}[x(t) - x(-t)]$

# 3. Exponential and Sinusoidal Signals

## 3.1 Continuous-Time Complex Exponential and Sinusoidal Signals

The continuous-time complex exponential signal is of the form  $x(t) = Ce^{at}$ .

**real exponential signals**  $C$  and  $a$  are real

**periodic complex exponential signals**  $x(t) = e^{j\omega_0 t}$

it is periodic,  $T_0 = \frac{2\pi}{\omega_0}$

**sinusoidal signals**  $x(t) = A\cos(\omega_0 t + \phi)$

**Euler's relation**  $e^{j\omega_0 t} = \cos(\omega_0 t) + j\sin(\omega_0 t)$

**fundamental frequency**  $|\omega_0|$

**harmonically related** complex exponential  $\phi_k(t) = e^{jk\omega_0 t} \quad k = 0, \pm 1, \pm 2, \dots$

**general complex exponential signals**

$C$  expressed in polar form and  $a$  in rectangular form

we have  $C = |C|e^{j\theta}$  and  $a = r + j\omega_0$

then  $Ce^{at} = |C|e^{rt}e^{j(\omega_0 t + \theta)}$

Sinusoidal signals multiplied by decaying exponentials are commonly referred to as **damped sinusoids**.

## 3.2 Discrete-Time Complex Exponential and Sinusoidal Signals

The discrete-time complex exponential signal is of the form  $x[n] = C\alpha^n$ .

**real exponential signals**  $C$  and  $\alpha$  are real

**sinusoidal signals**  $|\alpha| = 1$

then we have  $x[n] = A\cos(\omega_0 n + \phi)$

**general complex exponential signals**

we have  $C = |C|e^{j\theta}$  and  $\alpha = |\alpha|e^{j\omega_0}$

then  $C\alpha^n = |C||\alpha|^n \cos(\omega_0 n + \theta) + j|C||\alpha|^n \sin(\omega_0 n + \theta)$

$$e^{j(\omega_0 + 2\pi)n} = e^{j\omega_0 n}$$

The exponential at frequency  $\omega_0 + 2\pi$  is the same as that at frequency  $\omega_0$ , therefore, we need only consider a frequency interval of length  $2\pi$  in which to choose  $\omega_0$ .

The low-frequency discrete-time exponentials have values of  $\omega_0$  near  $2k\pi \quad k \in \mathbb{Z}$ , while the high frequencies are located near  $\omega_0 = (2k + 1)\pi \quad k \in \mathbb{Z}$

The signal  $e^{j\omega_0 n}$  is periodic if  $\frac{\omega_0}{2\pi} = \frac{m}{N}$  is a rational number and is not periodic otherwise.  
 its fundamental frequency is  $\frac{2\pi}{N} = \frac{\omega_0}{m}$

**harmonically related** complex exponential  $\phi_k[n] = e^{jk\frac{2\pi}{N}n} \quad k = 0, \pm 1, \pm 2, \dots$   
 and  $\phi_{k+N}[n] = \phi_k[n]$ , which implies that there are only  $N$  distinct periodic exponentials in the set

## 4. The Unit Impulse and Unit Step Functions

### 4.1 The Discrete-Time Unit Impulse and Unit Step Sequences

**unit impulse** or **unit sample**

$$\delta[n] = \begin{cases} 0, & n \neq 0 \\ 1, & n = 0 \end{cases}$$

**unit step**

$$u[n] = \begin{cases} 1, & n \geq 0 \\ 0, & n < 0 \end{cases}$$

There is a close relationship between the discrete-time unit impulse and unit step.

$$\delta[n] = u[n] - u[n-1]$$

$$u[n] = \sum_{m=-\infty}^n \delta[m] = \sum_{k=0}^n \delta[n-k]$$

and

$$x[n]\delta[n-n_0] = x[n_0]\delta[n-n_0]$$

### 4.2 The Continuous-Time Unit Step and Unit Impulse Functions

**unit step function**

$$u(t) = \begin{cases} 1, & t \geq 0 \\ 0, & t < 0 \end{cases}$$

**unit impulse function**

$$\delta(t) = \frac{du(t)}{dt}$$

similarly, we have

$$u(t) = \int_{-\infty}^t \delta(t) dt = \int_0^{\infty} \delta(t - \sigma) d\sigma$$

$$x(t)\delta(t - t_0) = x(t_0)\delta(t - t_0)$$

## 5. Continuous-Time and Discrete-Time Systems

**continuous-time system**  $x(t) \rightarrow y(t)$

**discrete-time systems**  $x[n] \rightarrow y[n]$

## 6. Basic System Properties

A system is said to be **memoryless** if its output for each value of the independent variable at a given time is dependent only on the input at that same time.

for example  $y(t) = x(t)$

An example of a discrete-time system **with memory** is  $y[n] = x[n - 1]$ .

A system is said to be **invertible** if distinct inputs lead to distinct outputs.

for example  $y(t) = 2x(t)$ , for which the inverse system is  $w(t) = \frac{1}{2}y(t)$

Examples of **noninvertible** systems are  $y[n] = 0$

A system is **causal** if the output at any time depends only on values of the input at the present time and in the past.

A system is **stable** if small inputs lead to responses that do not diverge.

A system is **time invariant** if the behavior and characteristics of the system are fixed over time.

A system is **linear** if:

1. The response to  $x_1(t) + x_2(t)$  is  $y_1(t) + y_2(t)$ . (additivity)

2. The response to  $ax_1(t)$  is  $ay_1(t)$ , where  $a$  is any complex constant. (homogeneity)

**superposition property**

$$x[n] = \sum_k a_k x_k[n] \rightarrow y[n] = \sum_k a_k y_k[n]$$