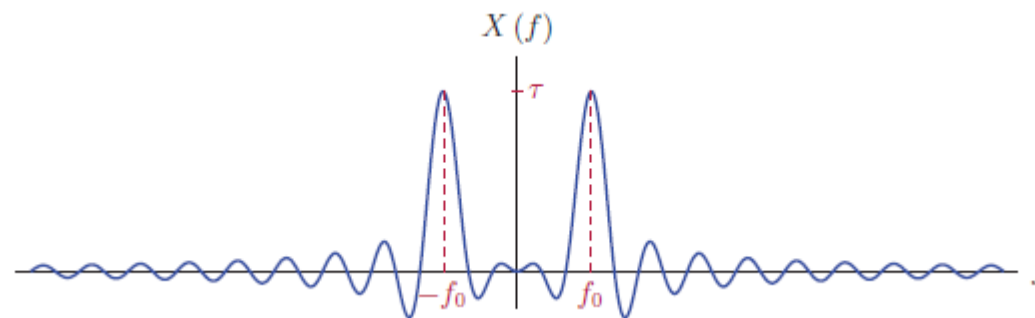


# CECC507: Signals and Systems

## Lecture Notes 2: Signal Representation and Modeling: Part B



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## Chapter 1

# Signal Representation and Modeling

1. Introduction
2. Mathematical Modeling of Signals
3. Continuous-Time Signals
4. **Discrete-Time Signals**

## Energy and power definitions

- The **energy** of a continuous time signal  $x(t)$  is given by:  $E_x = \int_{-\infty}^{\infty} |x(t)|^2 dt$

- The **average power** of a continuous time signal  $x(t)$  is given by:

**periodic** complex signal: 
$$P_x = \frac{1}{T_0} \int_{-T_0/2}^{T_0/2} |x(t)|^2 dt$$

**non-periodic** complex signal: 
$$P_x = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} |x(t)|^2 dt$$

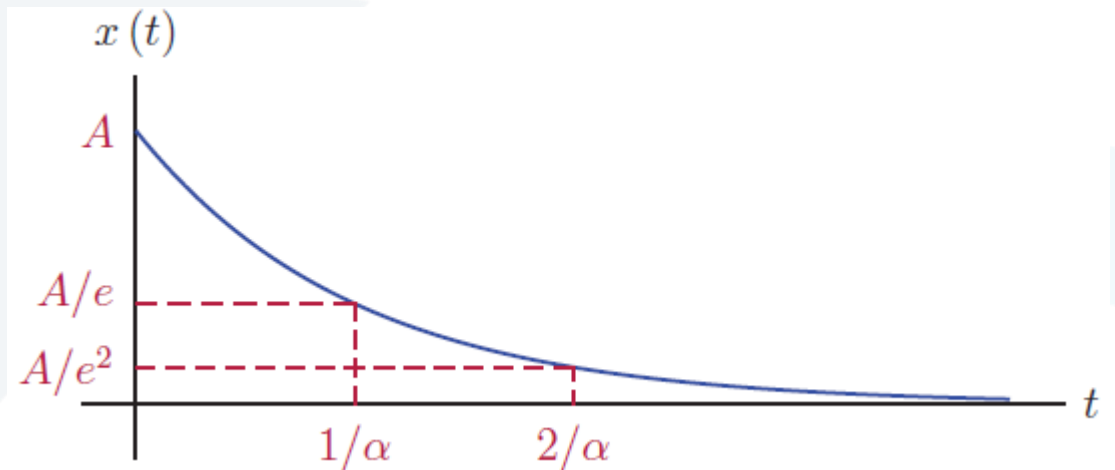
- **Energy signals** are those that have finite energy and zero power, i.e.,  $E_x < \infty$ , and  $P_x = 0$ .
- **Power signals** are those that have finite power and infinite energy, i.e.,  $E_x \rightarrow \infty$ , and  $P_x < \infty$ .

- **Example 1:** Energy of exponential signal

Compute the energy of the exponential signal (where  $\alpha > 0$ ).

$$x(t) = \begin{cases} A e^{-\alpha t} & \text{if } t \geq 0 \\ 0 & \text{otherwise} \end{cases}$$

$$E_x = \int_0^{\infty} A^2 e^{-2\alpha t} dt = \frac{A^2}{2\alpha}$$



- **Example 2:** Power of a sinusoidal signal

$$x(t) = A \sin(2\pi f_0 t + \theta)$$

$$P_x = f_0 \int_{-1/2f_0}^{1/2f_0} A^2 \sin^2(2\pi f_0 t + \theta) dt = \frac{A^2}{2}$$

## Symmetry properties

### Even and odd symmetry

- A **real-valued** signal is said to have **even symmetry** if it has the property:  $x(-t) = x(t)$  for all values of  $t$ .
- A **real-valued** signal is said to have **odd symmetry** if it has the property:  $x(-t) = -x(t)$  for all values of  $t$ .

### Decomposition into even and odd components

- Every **real-valued** signal  $x(t)$  has a **unique** representation of the form:  $x(t) = x_e(t) + x_o(t)$ ; where the signals  $x_e$  and  $x_o$  are **even** and **odd**, respectively.
- In particular, the signals  $x_e$  and  $x_o$  are given by:

$$x_e(t) = \frac{1}{2}[x(t) + x(-t)] \text{ and } x_o(t) = \frac{1}{2}[x(t) - x(-t)]$$

## Symmetry properties for complex signals

- A **complex-valued** signal is said to have **conjugate symmetric** if it has the property:  $x(-t) = x^*(t)$  for all values of  $t$ .
- A **complex-valued** signal is said to have **conjugate antisymmetric** if it has the property:  $x(-t) = -x^*(t)$  for all values of  $t$ .

## Decomposition of complex signals

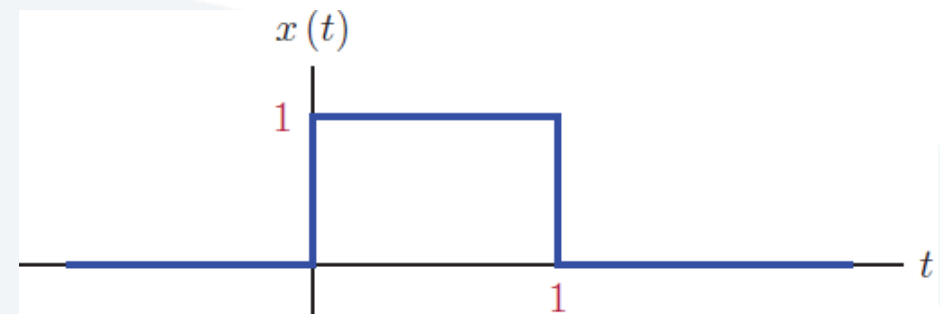
- Every **complex-valued** signal  $x(t)$  has a **unique** representation of the form:  $x(t) = x_E(t) + x_O(t)$ ; where the signals  $x_E$  and  $x_O$  are **conjugate symmetric** and **conjugate antisymmetric**, respectively.
- In particular, the signals  $x_E$  and  $x_O$  are given by:

$$x_E(t) = \frac{1}{2}[x(t) + x^*(-t)] \text{ and } x_O(t) = \frac{1}{2}[x(t) - x^*(-t)]$$

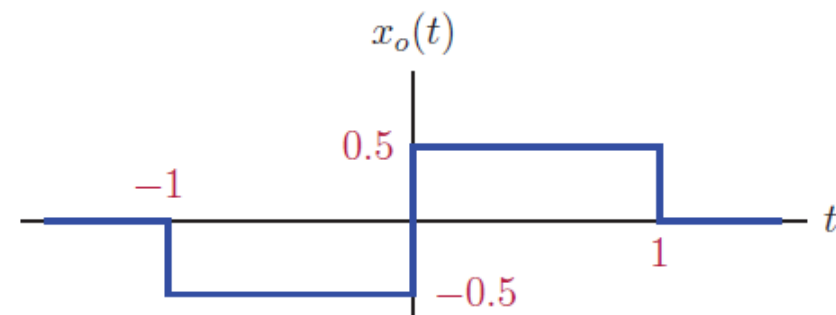
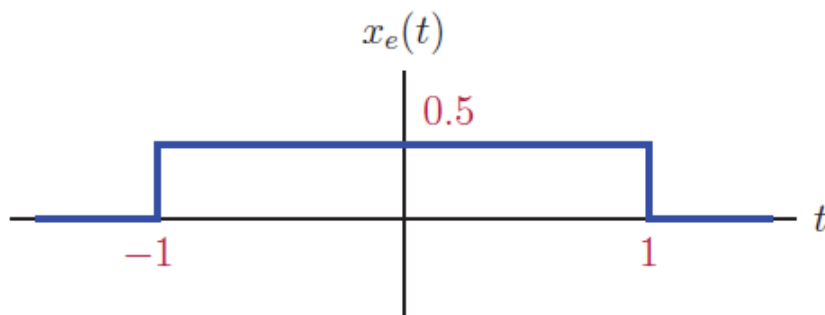
- **Example 3:** Even and odd components of a rectangular pulse

Determine the even and the odd components of the rectangular pulse signal.

$$\Pi\left(t - \frac{1}{2}\right) = \begin{cases} 1 & \text{if } 0 < t < 1 \\ 0 & \text{otherwise} \end{cases}$$

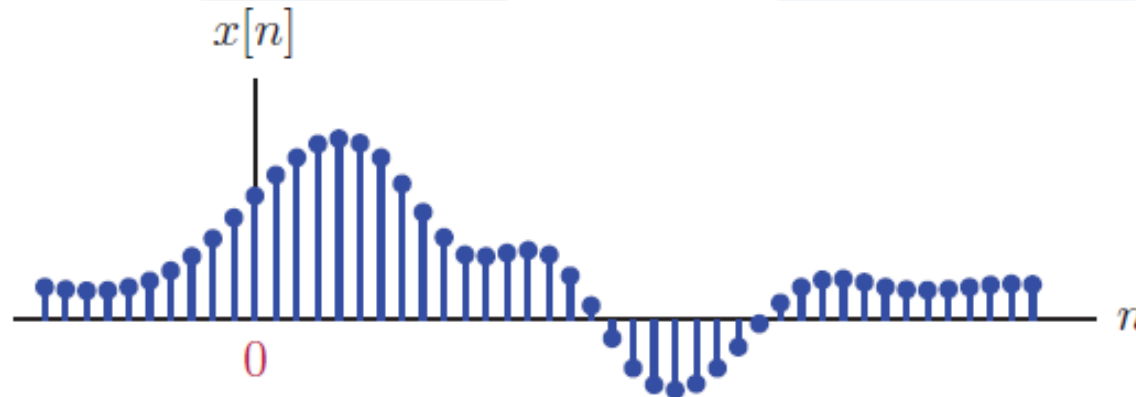


$$x_e(t) = \frac{\Pi\left(t - \frac{1}{2}\right) + \Pi\left(-t - \frac{1}{2}\right)}{2} = \frac{\Pi(t/2)}{2}, \quad x_o(t) = \frac{\Pi\left(t - \frac{1}{2}\right) - \Pi\left(-t - \frac{1}{2}\right)}{2}$$



## 4. Discrete-Time Signals

- Discrete-time signals are not defined at all time instants. they are defined only at time instants that are integer multiples of a fixed time increment  $T_s$ , that is, at  $t = nT_s$ .
- Consequently, the mathematical model for a discrete-time signal is a function  $x[n]$  in which independent variable  $n$  is an integer, and is referred to as the **sample index**.

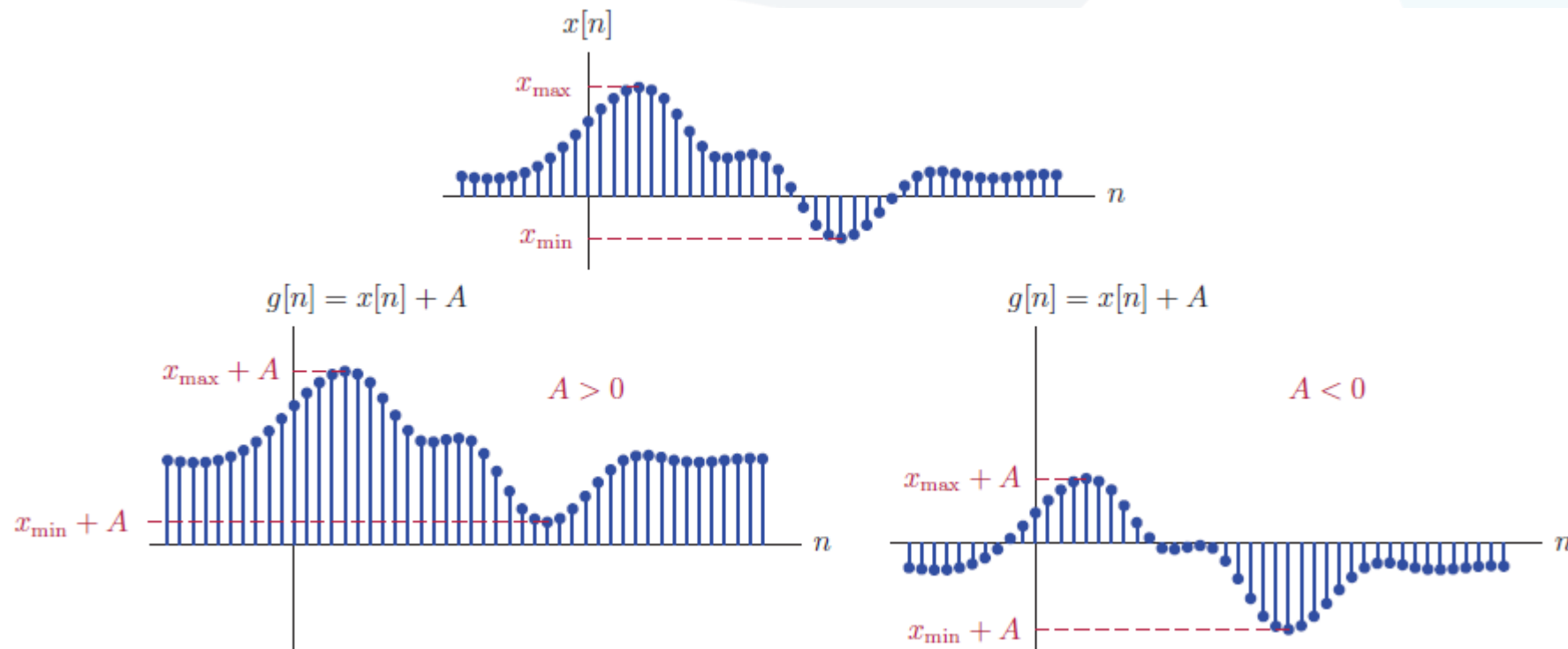




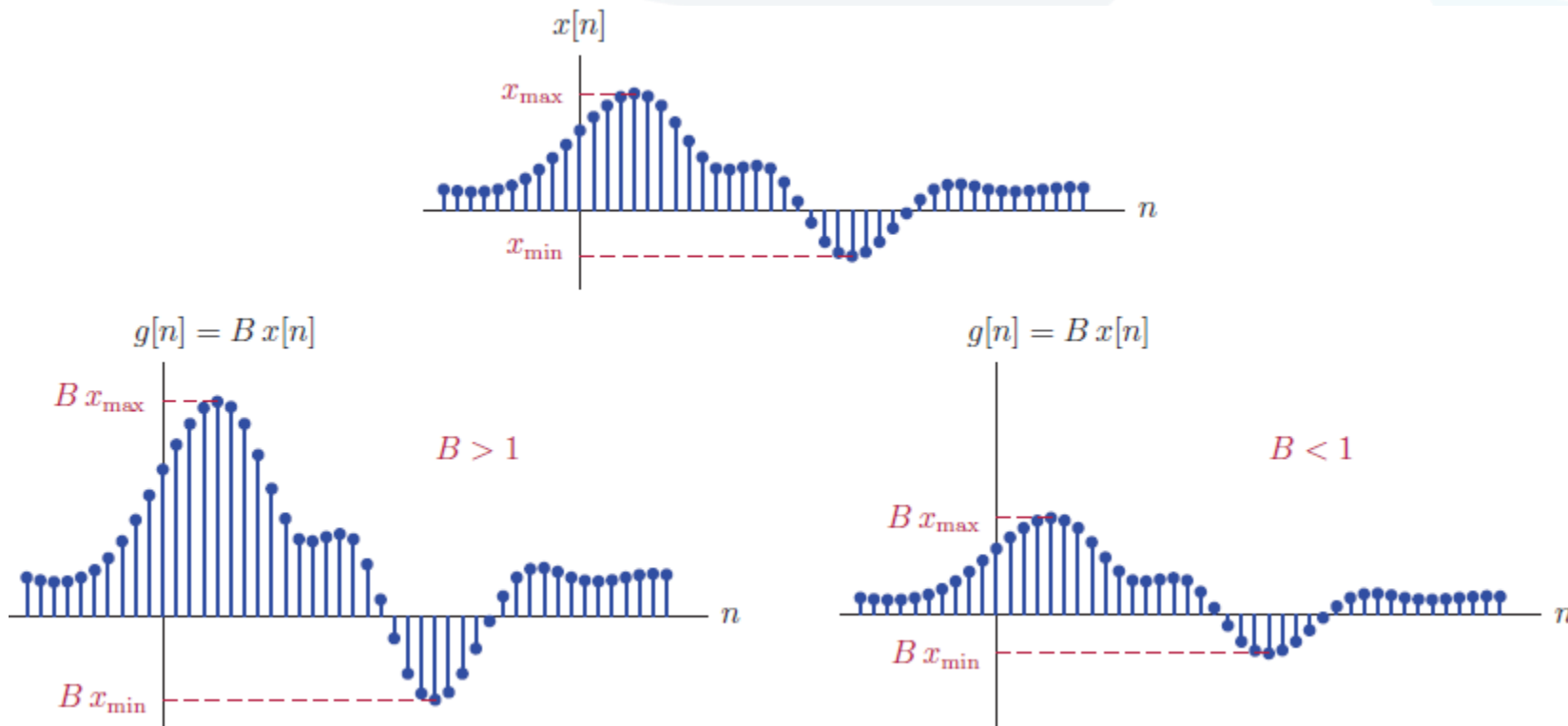
- Sometimes discrete-time signals are also modeled using mathematical functions:  $x[n] = 3\sin[0.2n]$ .
- In a discrete-time signal the time variable is discrete, yet the amplitude of each sample is continuous.
- If, in addition to limiting the time variable to the set of integers, we also limit the amplitude values to a discrete set, the resulting signal is called a **digital signal**.
- In the simplest case there are only two possible values for the amplitude of each sample, typically indicated by “0” and “1”. The corresponding signal is called a **binary signal**.

## Signal operations

- Amplitude shifting maps the input function  $x[n]$  to the output function  $g$  as given by  $g[n] = x[n] + A$ , where  $A$  is a real number.

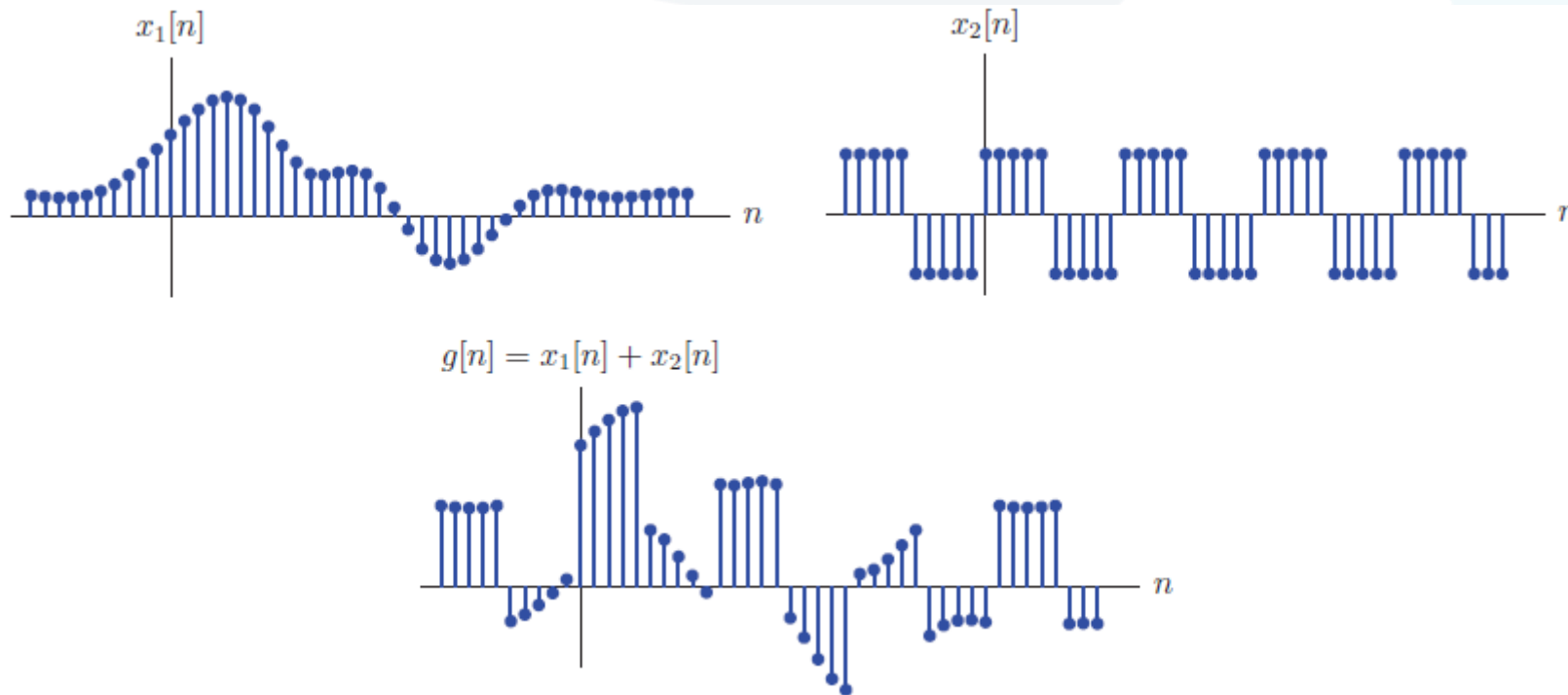


- **Amplitude scaling** maps the input function  $x$  to the output function  $g$  as given by  $g[n] = Bx[n]$ , where  $B$  is a real number.
- Geometrically, the output function  $g$  is **expanded/compressed** in amplitude.

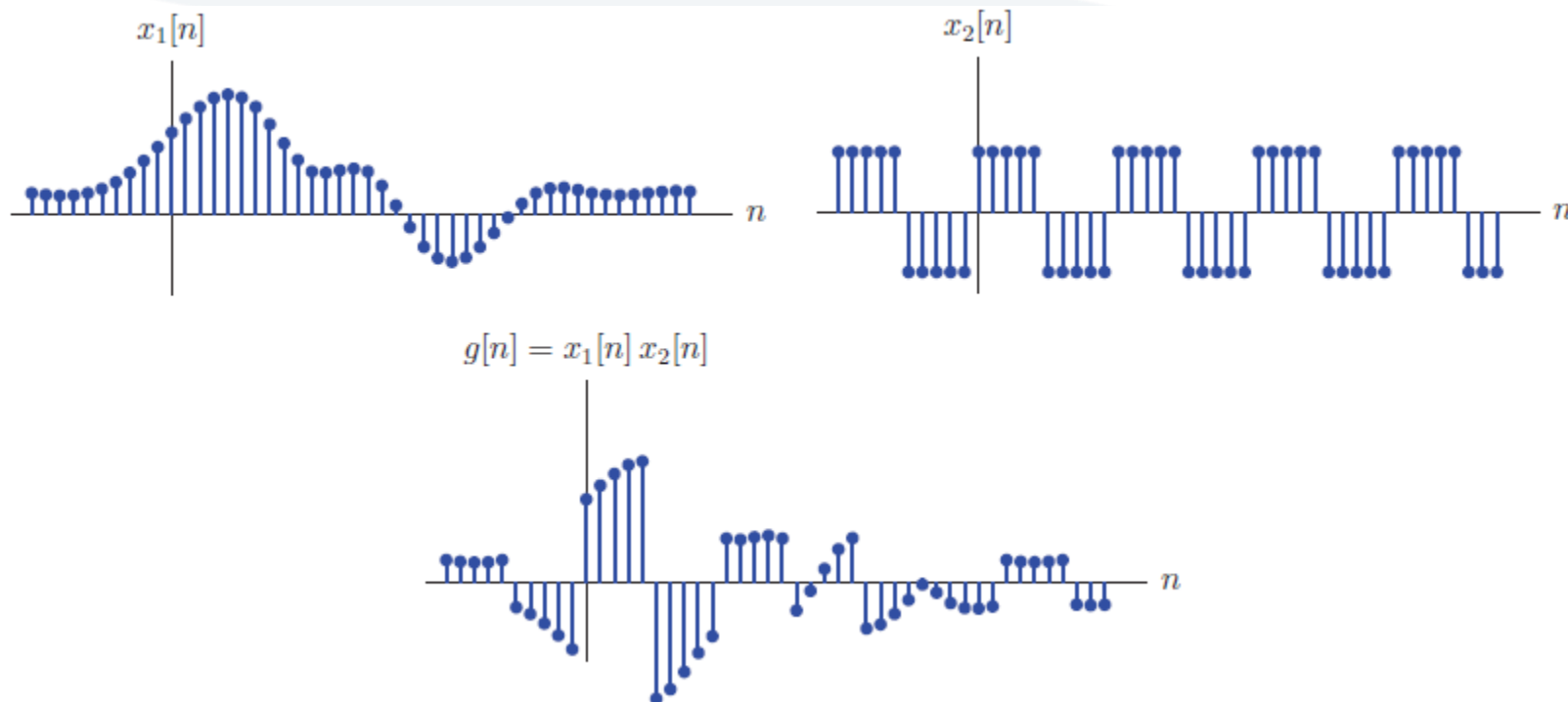


- **Addition and Multiplication** of two signals

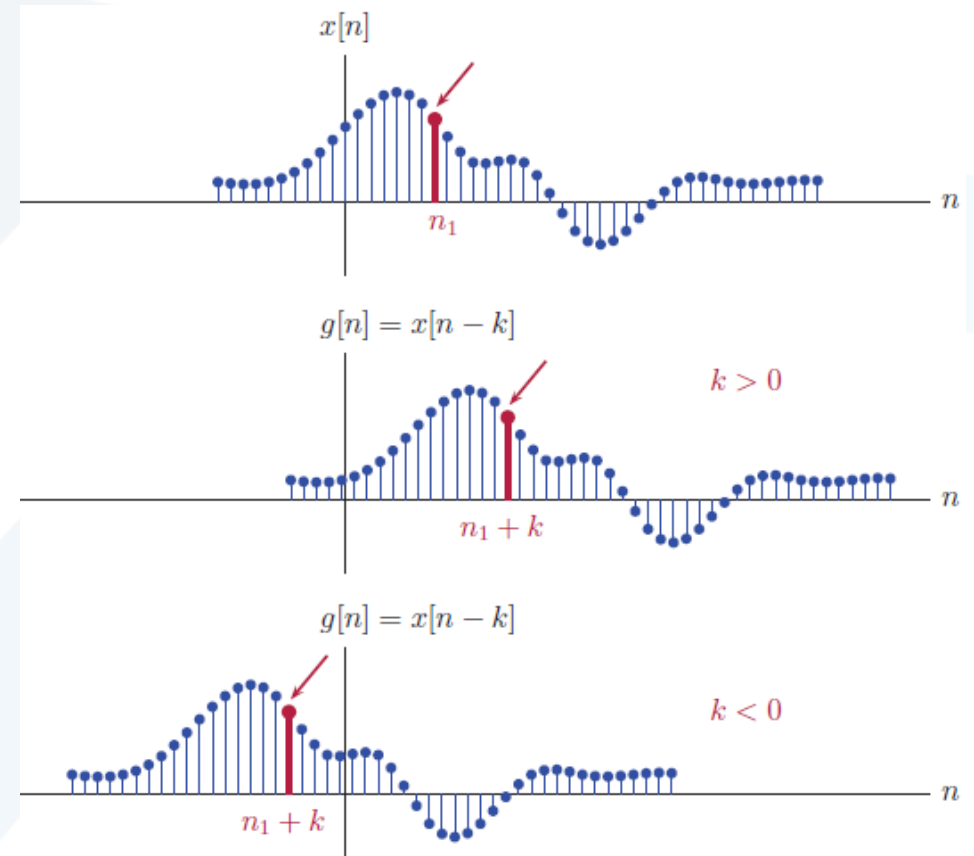
**Addition** of two signals is accomplished by adding the amplitudes of the two signals at each time instant.  $g[n] = x_1[n] + x_2[n]$ .



**Multiplication** of two signals is accomplished by multiplying the amplitudes of the two signals at each time instant.  $g[n] = x_1[n] x_2[n]$ .



- **Time shifting** (also called **translation**) maps the input signal  $x$  to the output signal  $g$  as given by:  $g[n] = x[n - k]$ ; where  $k$  is an integer.
- Such a transformation shifts the signal (to the left or right) along the time axis.
- If  $k > 0$ ,  $g$  is **shifted to the right** by  $|k|$ , relative to  $x$  (i.e., delayed in time).
- If  $k < 0$ ,  $g$  is **shifted to the left** by  $|k|$ , relative to  $x$  (i.e., advanced in time).



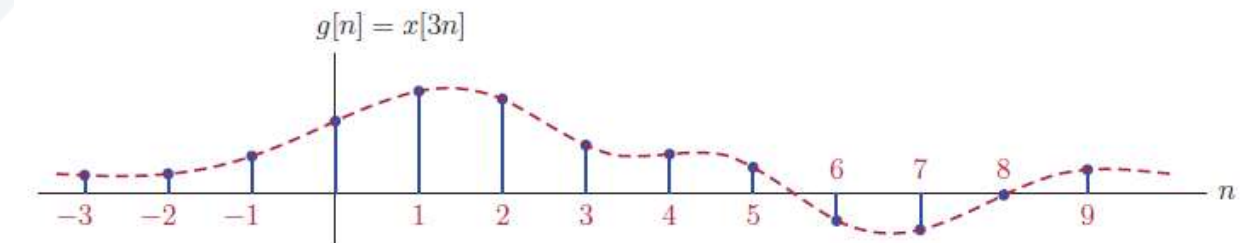
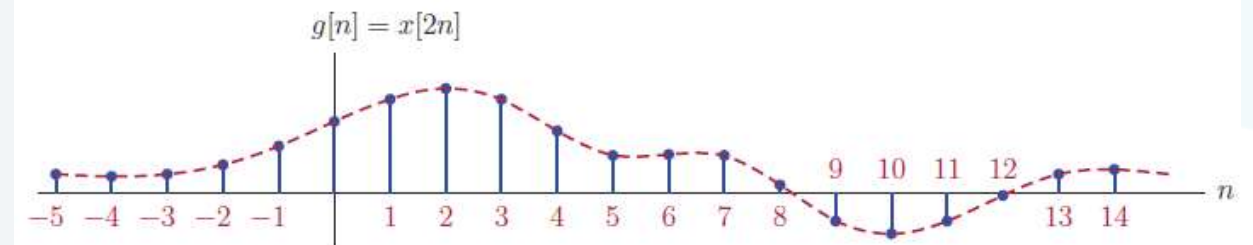
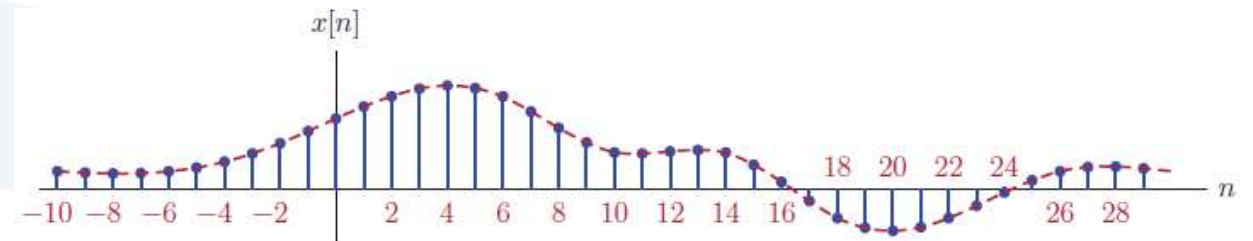
- Time scaling maps the input signal  $x$  to the output signal  $g$  as given by:

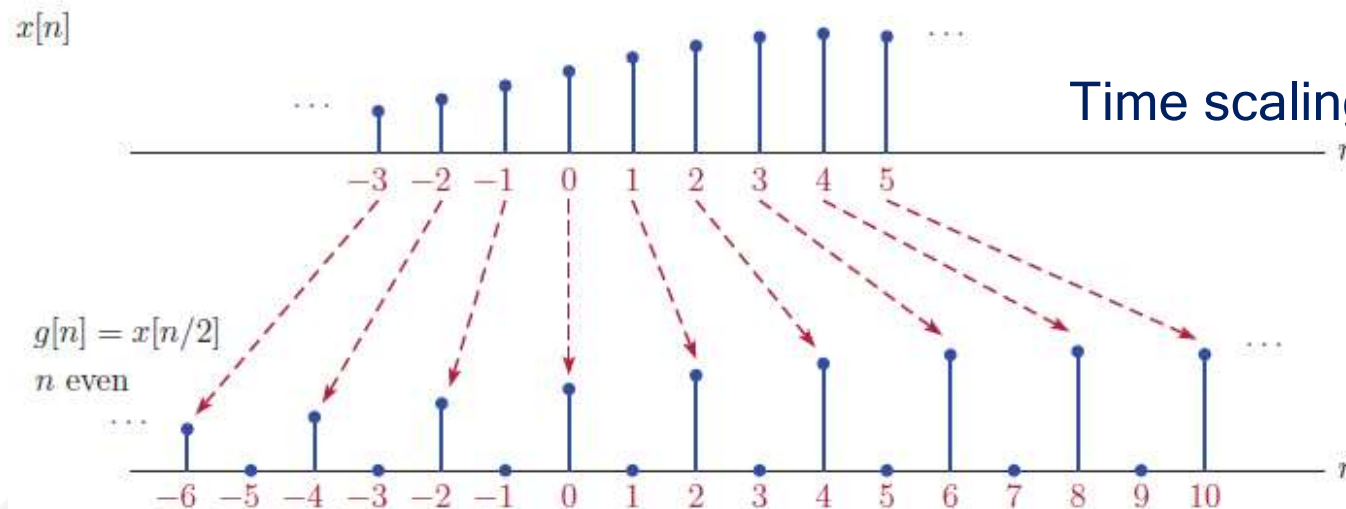
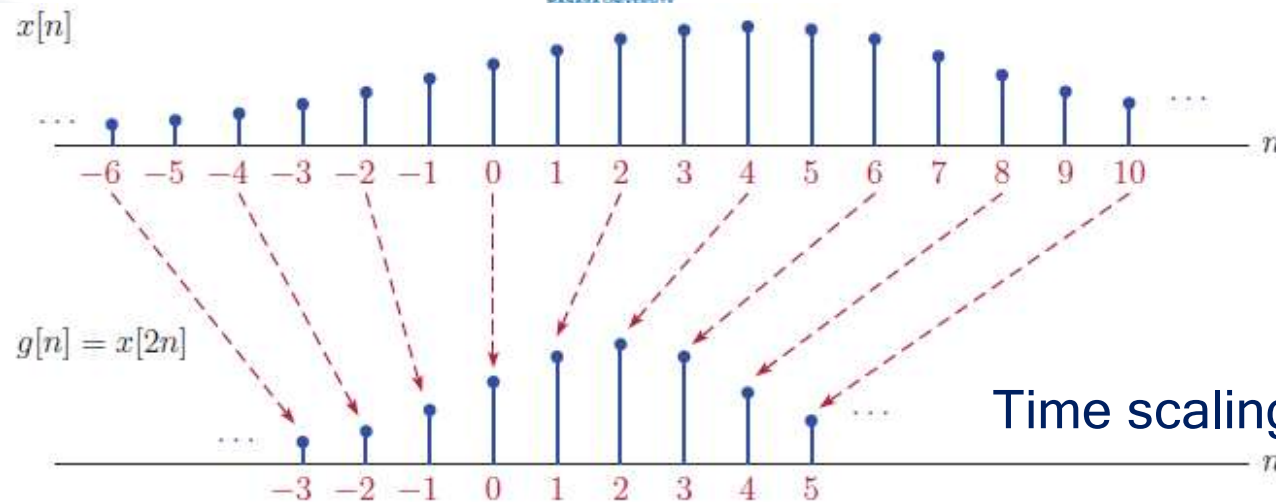
$$g[n] = x[kn]; \quad \text{downsampling}$$

and

$$g[n] = x[n/k]; \quad \text{upsampling}$$

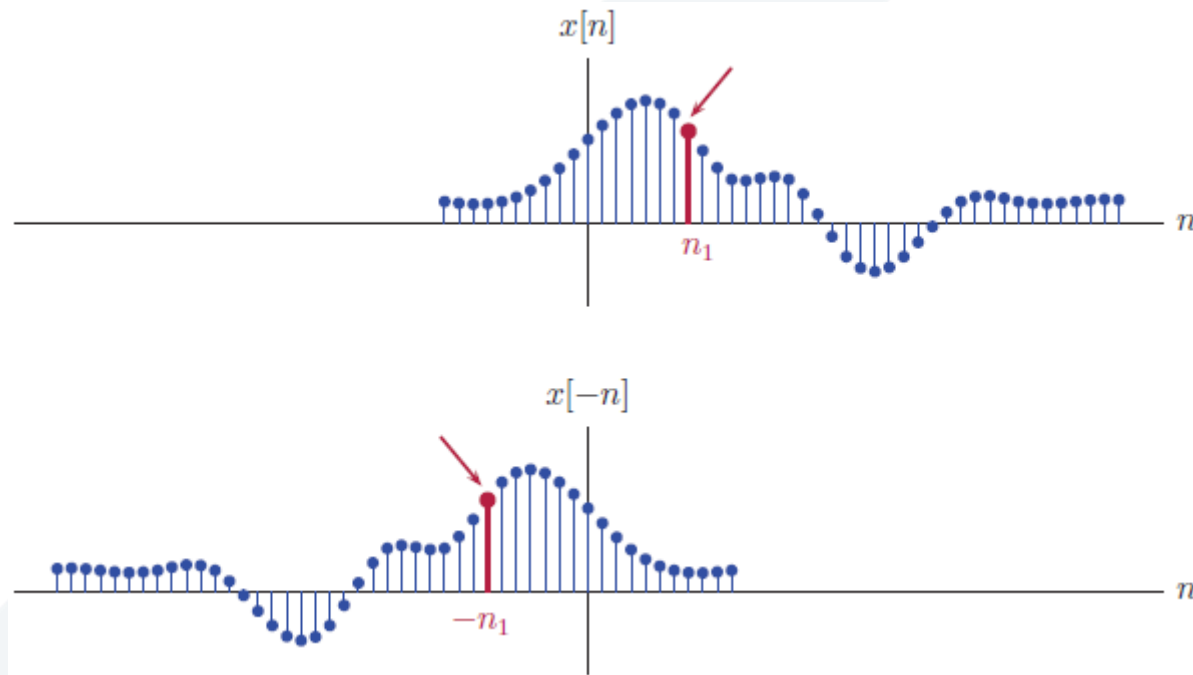
where  $k$  is a **strictly positive** integer.







- **Time reversal** (also known as **reflection**) maps the input signal  $x$  to the output signal  $g$  as given by  $g[n] = x[-n]$ .
- Geometrically, the output signal  $g$  is a reflection of the input signal  $x$  about the (vertical) line  $n = 0$ .

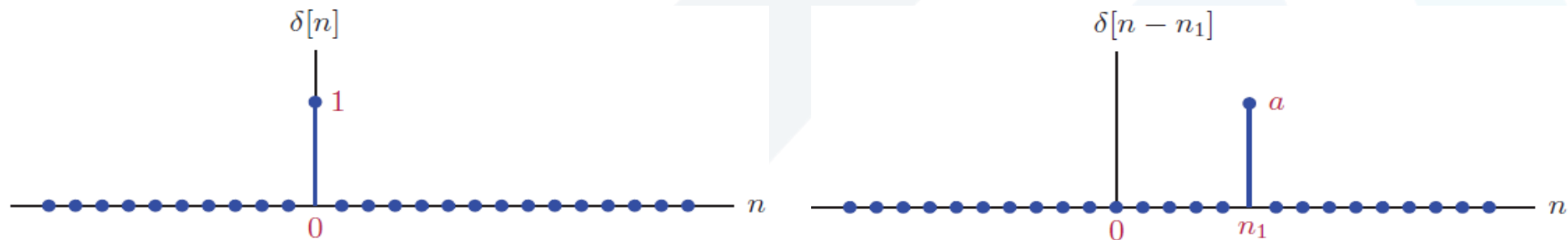


## Basic building blocks for discrete-time signals

### Unit-impulse function

- The **unit-impulse function**, denoted  $\delta$ , is defined by:

$$\delta[n] = \begin{cases} 1, & \text{if } n = 0 \\ 0, & \text{if } n \neq 0 \end{cases} \quad a\delta[n - n_1] = \begin{cases} a, & \text{if } n = n_1 \\ 0, & \text{if } n \neq n_1 \end{cases}$$

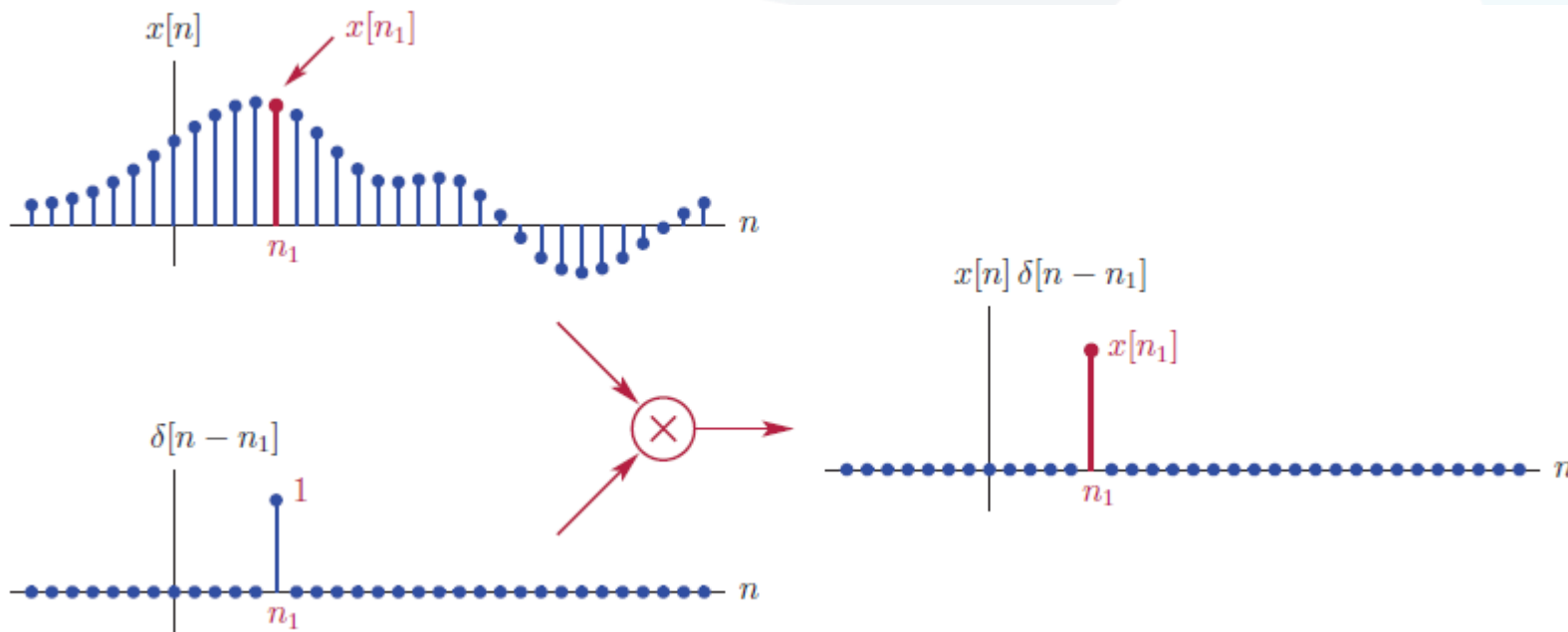


- Sampling property** of the unit-impulse function:

$$x[n]\delta[n - n_1] = x[n_1]\delta[n - n_1] = \begin{cases} x[n_1], & n = n_1 \\ 0, & n \neq n_1 \end{cases}$$

- **Sifting property** of the unit-impulse function

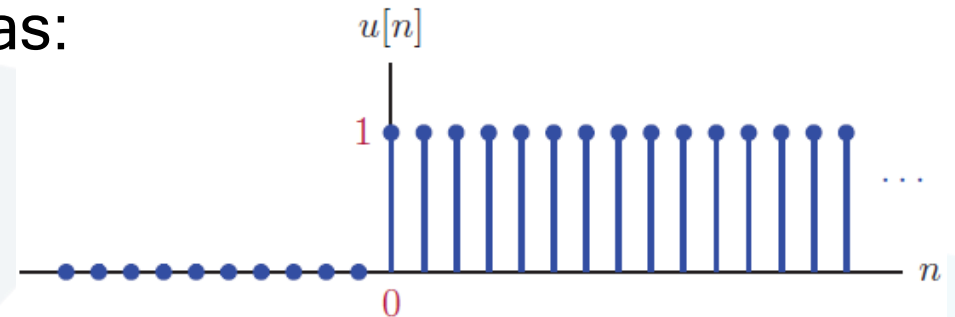
$$\sum_{n=-\infty}^{\infty} x[n] \delta[n - n_1] = x[n_1]$$



## Unit-Step Function

- The **unit-step function**, denoted  $u$ , is defined as:

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

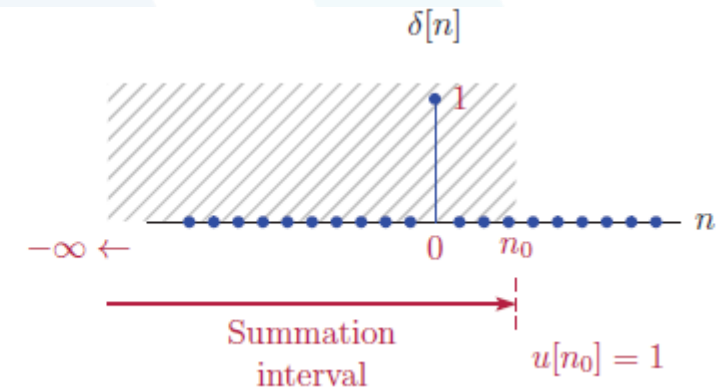
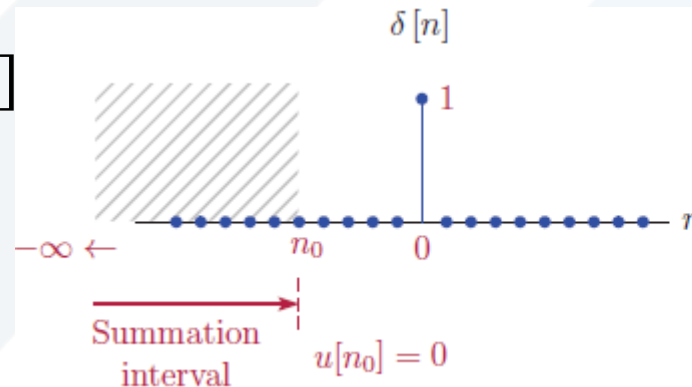


- Relationship between the unit-step function and the unit-impulse function:

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

- Conversely,  $u[n] = \sum_{k=-\infty}^n \delta[k]$

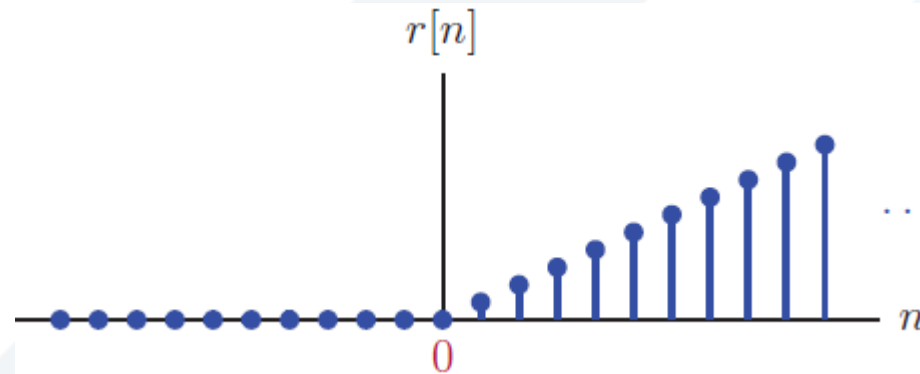
$$\text{or, } u[n] = \sum_{k=0}^{\infty} \delta[n - k]$$



## Unit-Ramp Function

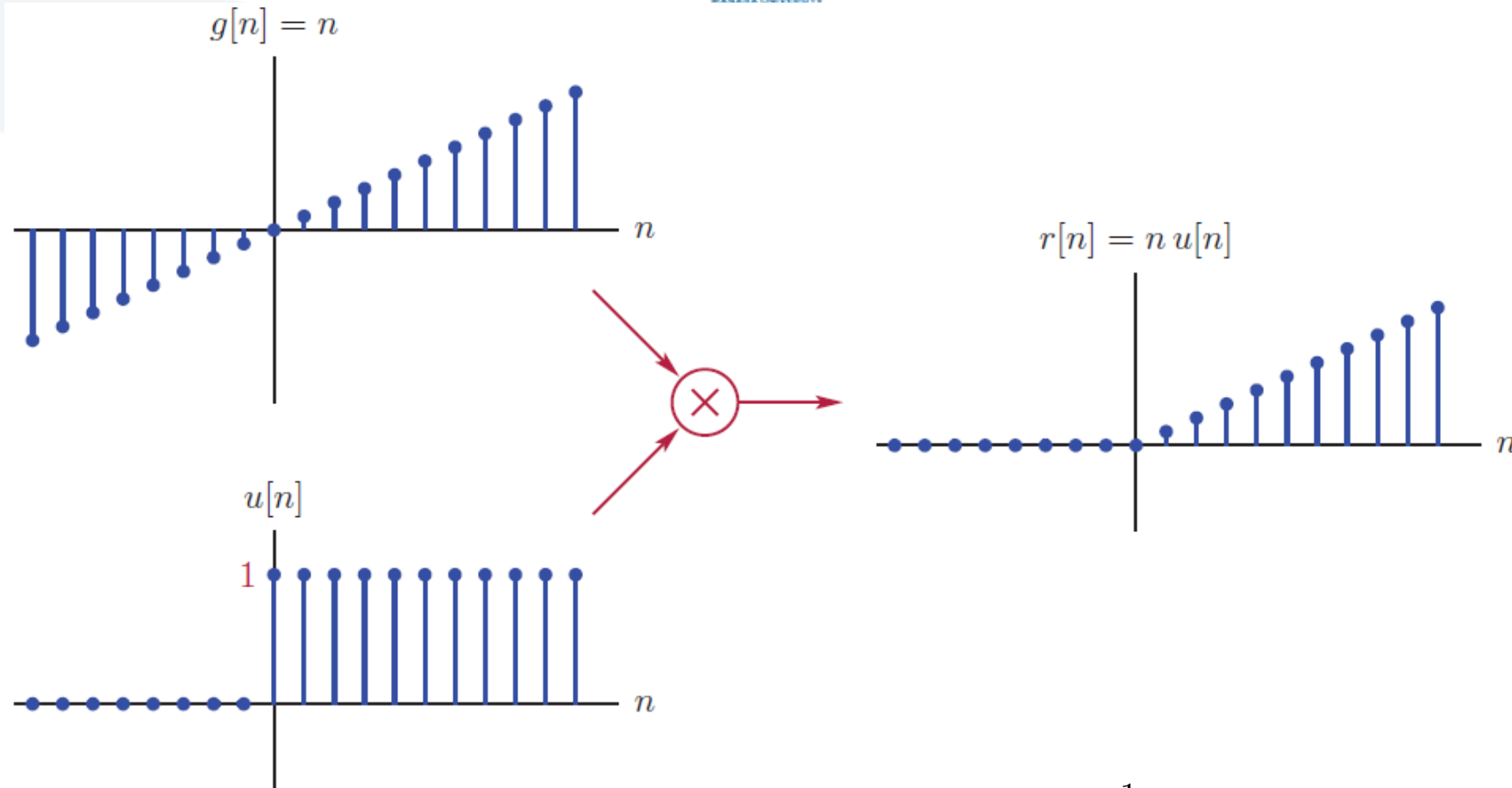
- The **unit-ramp function**, denoted  $r$ , is defined as:

$$r[n] = \begin{cases} n & \text{if } n \geq 0 \\ 0 & \text{otherwise} \end{cases}$$



- or, equivalently:

$$r[n] = nu[n]$$



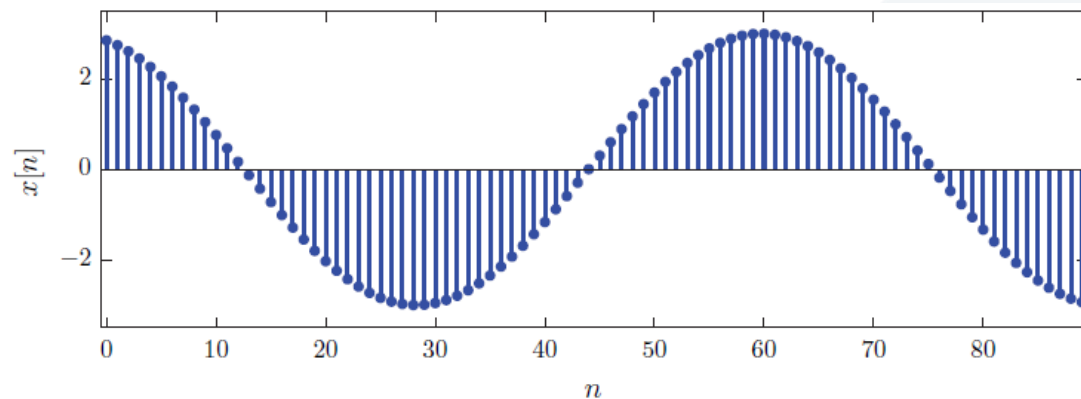
- Constructing a unit-ramp from a unit-step  $r[n] = \sum_{k=-\infty}^{n-1} u[k]$

## Sinusoidal Signal

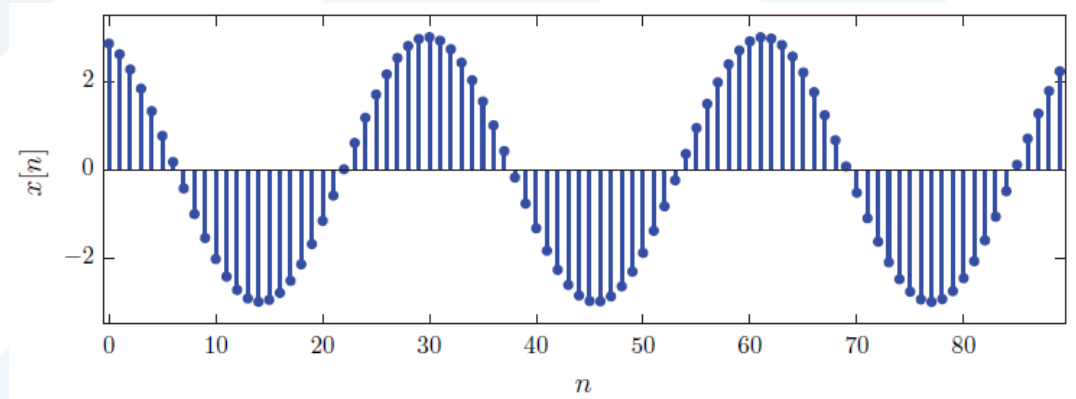
- A **discrete-time sinusoidal function** is a function of the form

$$x[n] = A \cos(\Omega_0 n + \theta)$$

where  $A$  is the **amplitude** of the signal,  $\Omega_0$  is the **angular frequency** (rad), and  $\theta$  is the initial phase angle (rad).  $\Omega_0 = 2\pi F_0$  where  $F_0$  is the **normalized frequency** (a dimensionless quantity).



$$x[n] = 3\cos(0.1n + \pi/10)$$



$$x[n] = 3\cos(0.2n + \pi/10)$$

A fundamental difference between a DT sinusoidal signal and its CT:

- For continuous-time sinusoidal signal  $x_a(t) = A\cos(\omega_0 t + \theta)$ :  $\omega_0$  is in rad/s.
- For discrete-time sinusoidal signal  $x[n] = A\cos(\Omega_0 n + \theta)$ :  $\Omega_0$  is in rad.
- Let us evaluate the amplitude of  $x_a(t)$  at time instants that are integer multiples of  $T_s$ , and construct a discrete-time signal:

$$x[n] = x_a(nT_s) = A\cos(\omega_0 T_s n + \theta) = A\cos(2\pi f_0 T_s n + \theta)$$

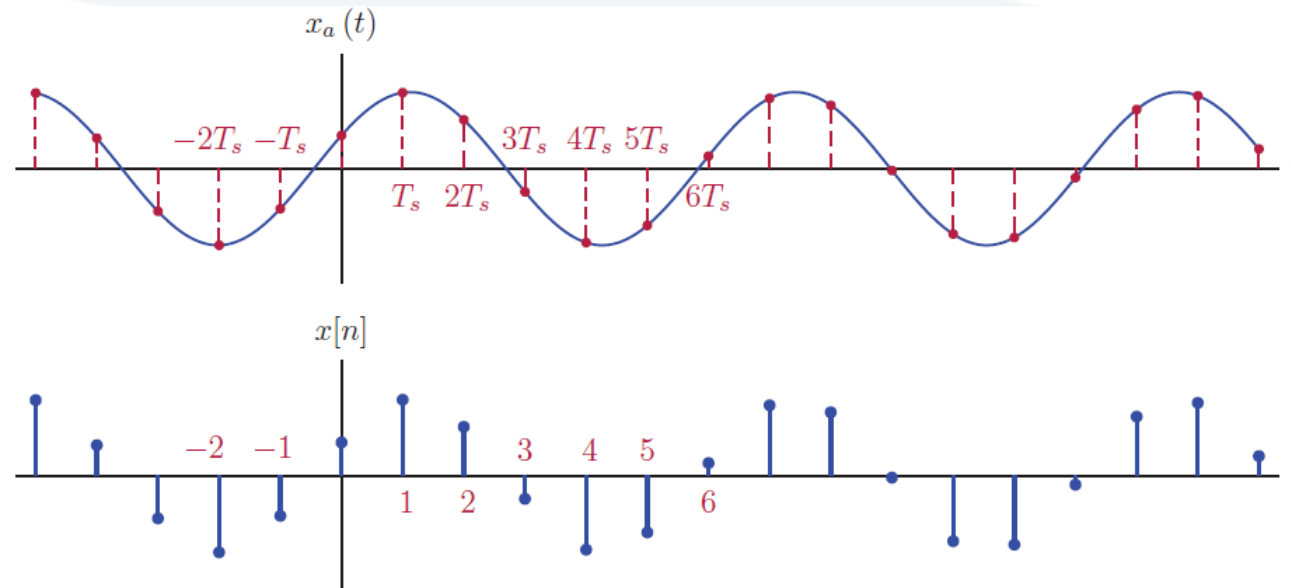
- Since the signal  $x_a(t)$  is evaluated at intervals of  $T_s$ , the number of samples taken per unit time is  $1/T_s$ .

$$x[n] = A\cos\left(2\pi \left[f_0/f_s\right]n + \theta\right) = A\cos(2\pi F_0 n + \theta)$$

- The act of constructing a discrete-time signal by evaluating a continuous-time signal at uniform intervals is called **sampling**.



- The parameters  $f_s$  and  $T_s$  are referred to as the **sampling rate** and the **sampling interval** respectively.



## Impulse decomposition for discrete-time signals

- Consider an arbitrary discrete-time signal  $x[n]$ . Let us define a new signal  $x_k[n]$  by:

$$x_k[n] = x[k]\delta[n - k] = \begin{cases} x[k], & n = k \\ 0, & n \neq k \end{cases}$$

- The signal  $x[n]$  can be reconstructed by:  $x[n] = \sum_{k=-\infty}^{\infty} x_k[n] = \sum_{k=-\infty}^{\infty} x[k]\delta[n - k]$

## Periodic discrete-time signals

- A discrete-time signal is said to be **periodic** if it satisfies:  $x[n] = x[n + N]$  for all values of the integer index  $n$  and for a specific value of  $N \neq 0$ . The parameter  $N$  is referred to as the **period** of the signal.

- The period of a periodic signal is **not unique**. That is, a signal that is periodic with period  $N$  is also periodic with period  $kN$ , for every (strictly) positive integer  $k$ ,  $x[n] = x[n + kN]$ .
- The smallest period with which a signal is periodic is called the **fundamental period**.
- The normalized **fundamental frequency** of a discrete-time periodic signal is  $F_0 = 1/N$ .

## Periodicity of discrete-time sinusoidal signals

$$\begin{aligned} A \cos(2\pi F_0 n + \theta) &= A \cos(2\pi F_0 [n + N] + \theta) \\ &= A \cos(2\pi F_0 n + 2\pi F_0 N + \theta) \end{aligned}$$

$$2\pi F_0 N = 2\pi k \Rightarrow N = k/F_0$$

$N$  must be an **integer value**

- **Example 4:** Periodicity of a discrete-time sinusoidal signal

Check the periodicity of the following discrete-time signals:

a.  $x[n] = \cos(0.2n)$

b.  $x[n] = \cos(0.2\pi n + \pi/5)$

c.  $x[n] = \cos(0.3\pi n - \pi/10)$

a.  $x[n] = \cos(0.2n)$

$$\Omega_0 = 0.2 \Rightarrow F_0 = \Omega_0/2\pi = 0.2/2\pi = 0.1/\pi \Rightarrow N = k/F_0 = 10\pi k$$

Since no value of  $k$  would produce an integer value for  $N$ , the signal is **not periodic**.

b.  $x[n] = \cos(0.2\pi n + \pi/5)$

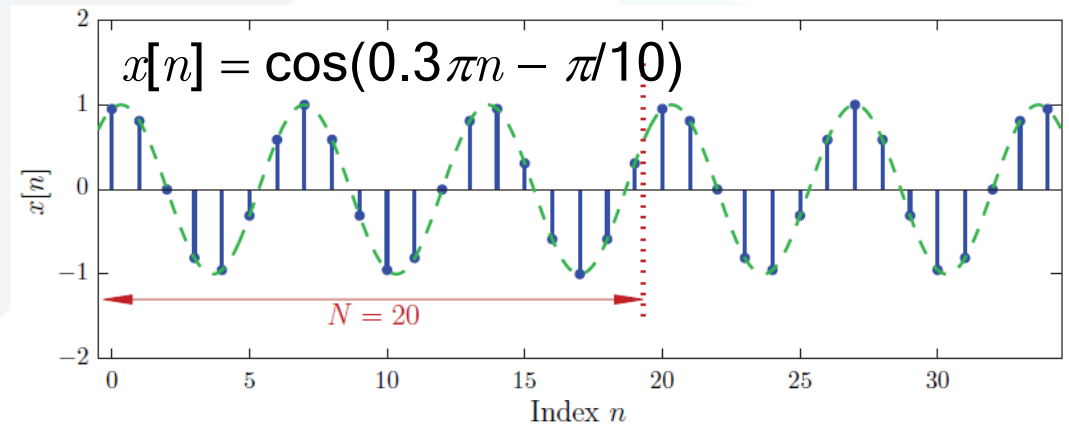
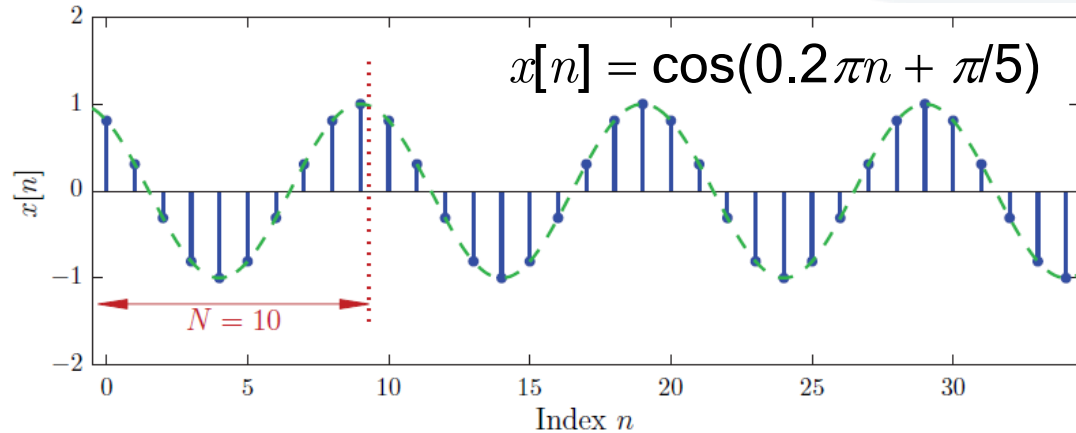
$$\Omega_0 = 0.2\pi \Rightarrow F_0 = \Omega_0/2\pi = 0.2\pi/2\pi = 0.1 \Rightarrow N = k/F_0 = 10k$$

For  $k = 1$  we have  $N = 10$  samples as the **fundamental period**.

c.  $x[n] = \cos(0.3\pi n - \pi/10)$

$$\Omega_0 = 0.3\pi \Rightarrow F_0 = \Omega_0/2\pi = 0.3\pi/2\pi = 0.15 \Rightarrow N = k/F_0 = k/0.15$$

For  $k = 3$  we have  $N = 20$  samples as the **fundamental period**.



- **Example 5:** Periodicity of a multi-tone discrete-time sinusoidal signal  
Comment on the periodicity of the two-tone discrete-time signal:

$$x[n] = 2\cos(0.4\pi n) + 1.5\sin(0.48\pi n)$$

$$x[n] = x_1[n] + x_2[n]$$

$$x_1[n] = 2\cos(\Omega_1 n)$$

$$\Omega_1 = 0.4\pi \Rightarrow F_1 = \Omega_1/2\pi = 0.4\pi/2\pi = 0.2$$

$$\Rightarrow N = k_1/F_1 = 5k_1$$

For  $k_1 = 1$  we have  $N_1 = 5$  samples as the **fundamental period**.

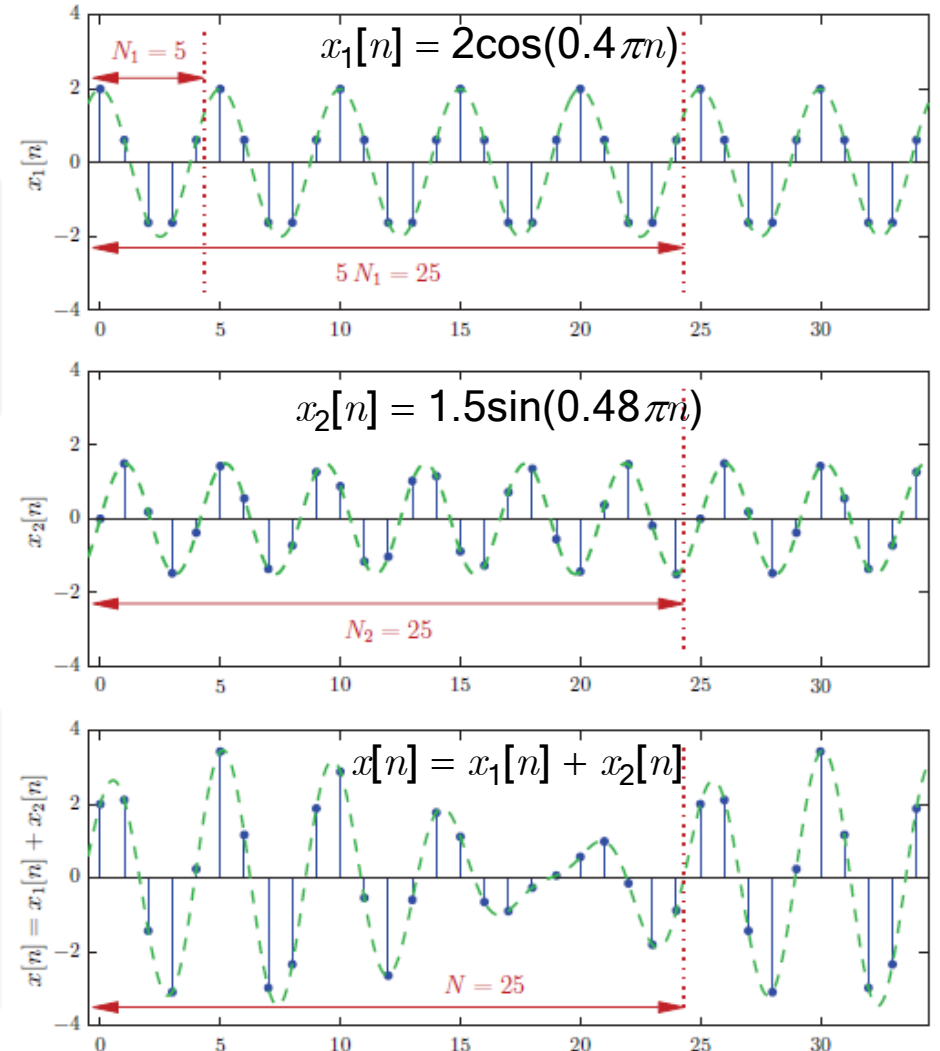
$$x_2[n] = 1.5\cos(\Omega_2 n)$$

$$\Omega_2 = 0.48\pi \Rightarrow F_2 = \Omega_2/2\pi = 0.48\pi/2\pi = 0.24$$

$$\Rightarrow N_2 = k_2/F_2 = k_2/0.24$$

For  $k_2 = 6$  we have  $N_2 = 25$  samples as the **fundamental period**.

$$\Rightarrow N = 25$$



## Energy and power definitions

- The **energy** of a discrete time signal  $x[n]$  is given by  $E_x = \sum_{n=-\infty}^{\infty} |x[n]|^2$
- A signal with finite energy is said to be an **energy signal**.
- The **average power** of a discrete time signal  $x[n]$  is given by:

periodic complex signal 
$$P_x = \frac{1}{N} \sum_{n=0}^{N-1} |x[n]|^2$$

non-periodic complex signal 
$$P_x = \lim_{M \rightarrow \infty} \frac{1}{2M+1} \sum_{n=-M}^M |x[n]|^2$$

- A signal with (nonzero) finite average power is said to be a **power signal**.

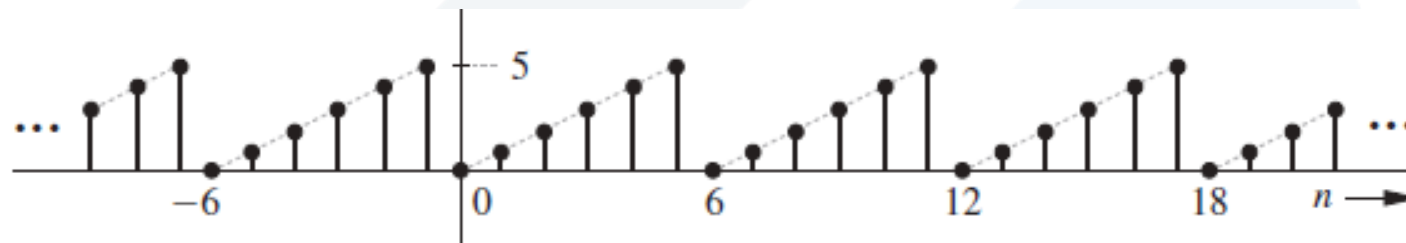
- **Example 6:** Energy of exponential signal

Determine the energy of the exponential signal  $x[n] = 0.8^n u[n]$ .

$$E_x = \sum_{n=-\infty}^{\infty} |x[n]|^2 = E_x = \sum_0^{\infty} (0.8^2)^n = \frac{1}{1 - 0.64} = \frac{1}{0.36} \approx 2.777$$

- **Example 7:** Average power of the periodic signal

Determine the normalized average power of the periodic signal



$$P_x = \frac{1}{N} \sum_{n=0}^{N-1} |x[n]|^2 = \frac{1}{6} \sum_{n=0}^5 n^2 = \frac{55}{6}$$



## Decomposition into even and odd components

### Decomposition of real signals

- Every function  $x$  has a **unique** representation of the form:  $x[n] = x_e[n] + x_o[n]$ ; where the functions  $x_e$  and  $x_o$  are **even** and **odd**, respectively.
- In particular, the functions  $x_e$  and  $x_o$  are given by
$$x_e[n] = \frac{1}{2}(x[n] + x[-n]) \text{ and } x_o[n] = \frac{1}{2}(x[n] - x[-n])$$
- The functions  $x_e$  and  $x_o$  are called the **even** part and **odd** part of  $x$ , respectively.

### Decomposition of complex signals

$$x_E[n] = \frac{1}{2}(x[n] + x^*[-n]) \text{ and } x_O[n] = \frac{1}{2}(x[n] - x^*[-n])$$