



# Chap2

# Discrete-time signals and systems

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# Signals

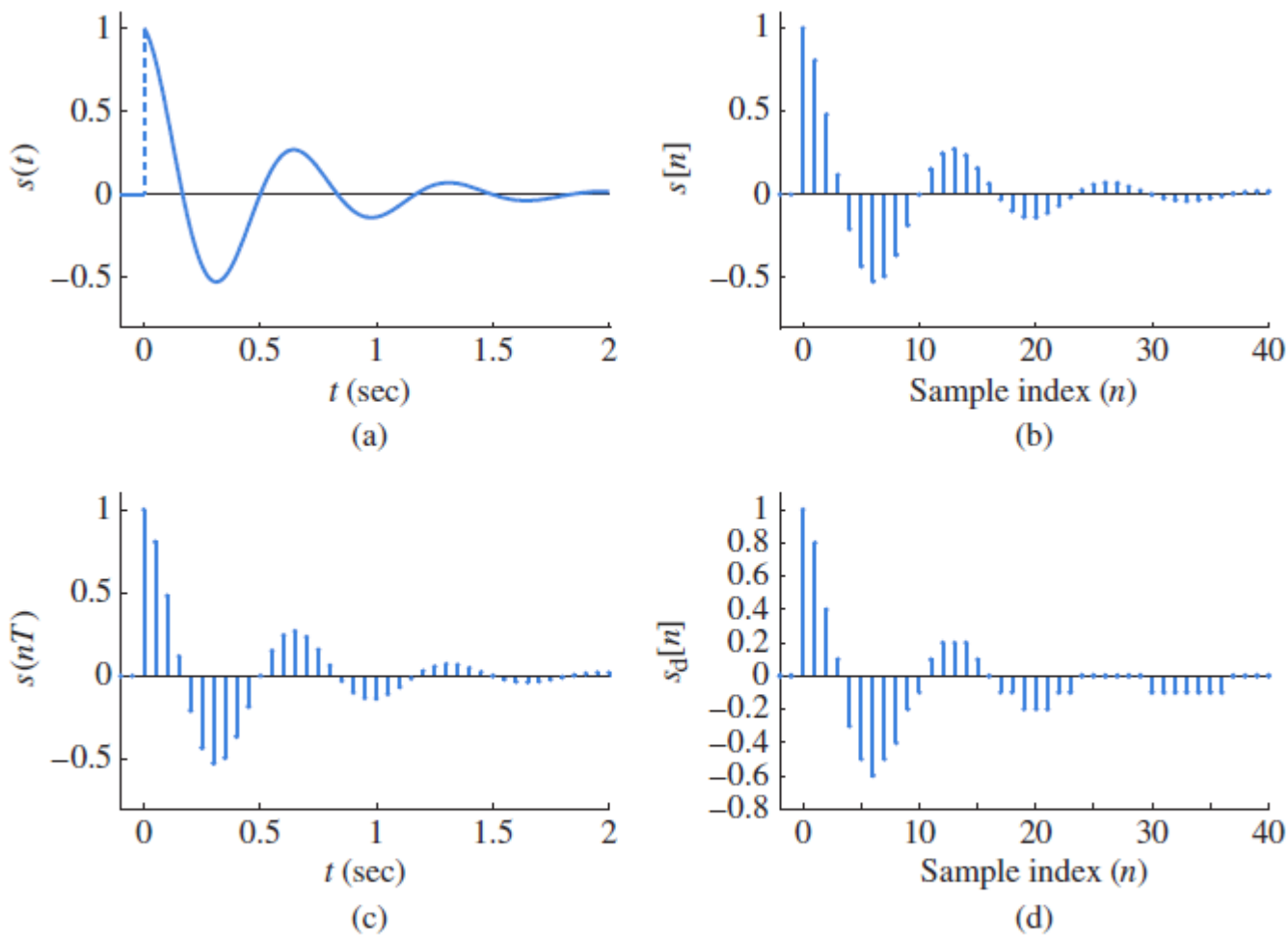
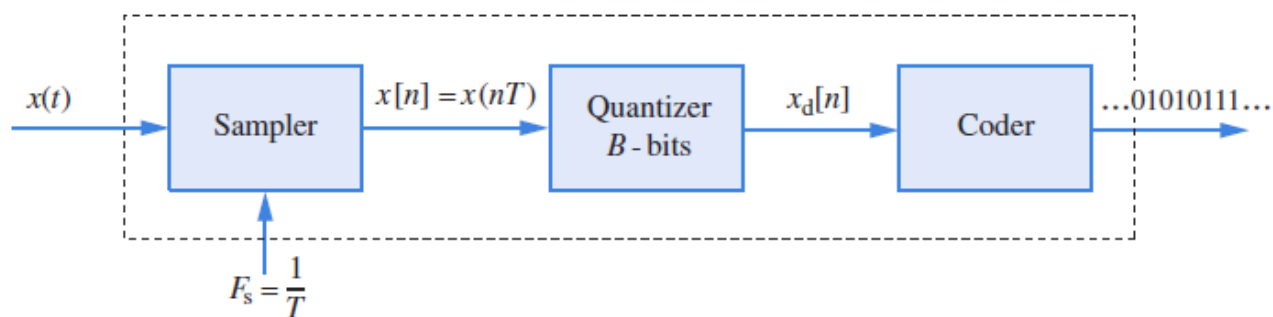
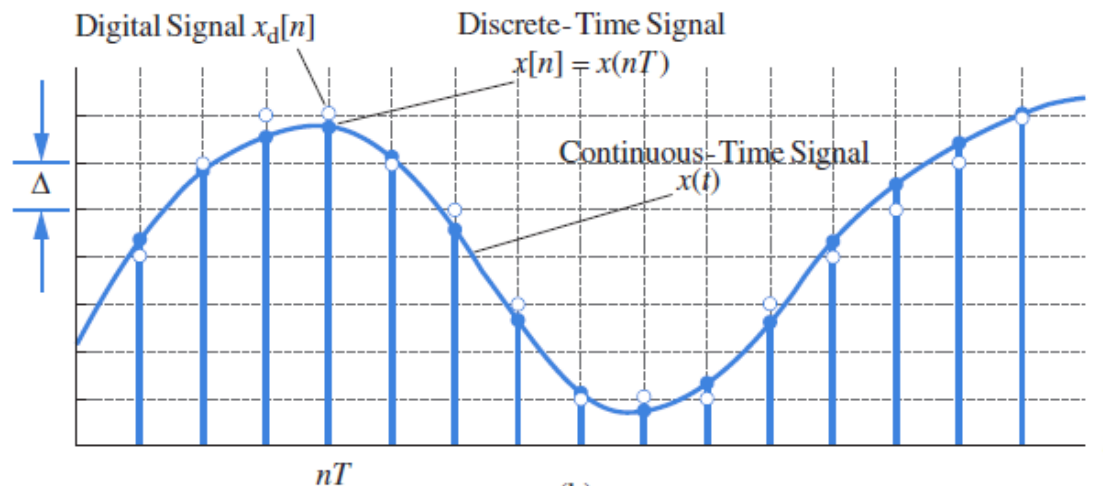


Figure 1.5 Plots illustrating the graphical representation of continuous-time signals (a), discrete-time signals (b) and (c), and digital signals (d).

# Analog-to-digital conversion



(a)

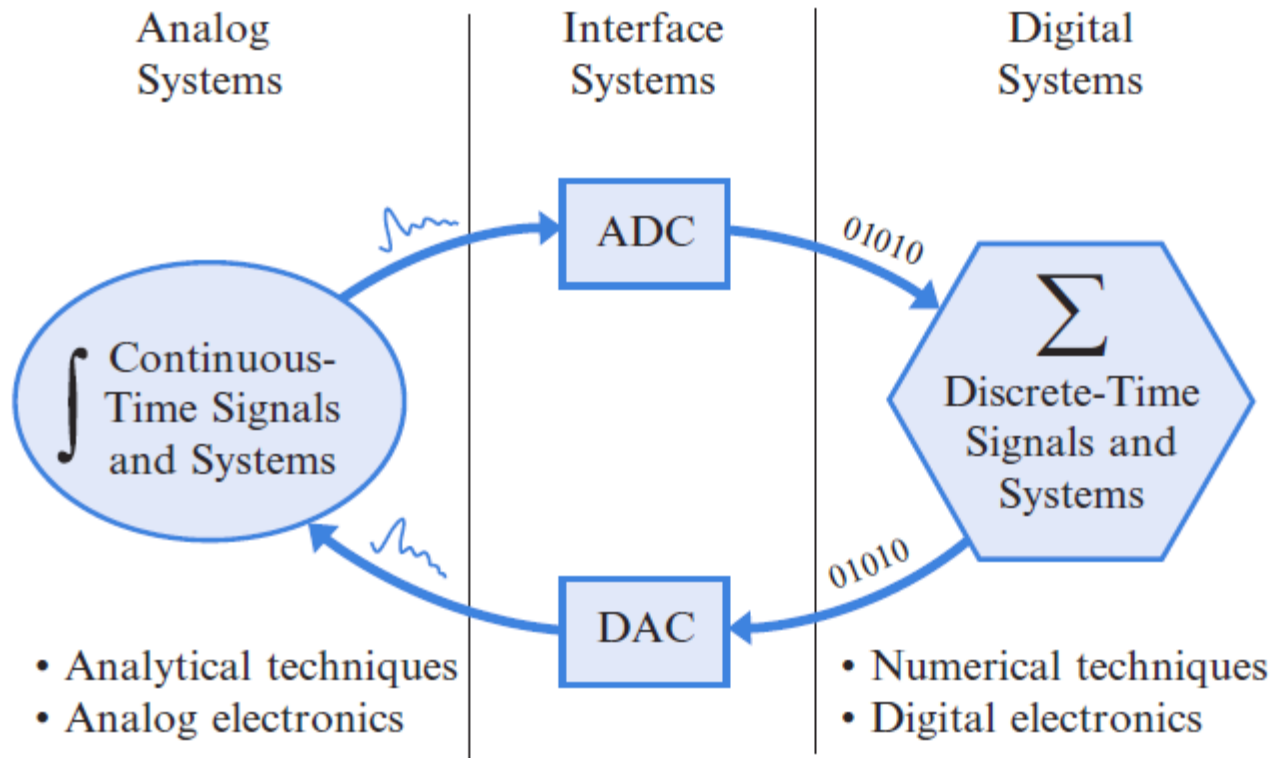


(b)

Figure 1.8 (a) Block diagram representation of the analog-to-digital conversion process. (b) Examples of the signals  $x(t)$ ,  $x[n]$ , and  $x_d[n]$  involved in the process. The amplitude of  $x[n]$  is known with infinite precision, whereas the amplitude of  $x_d[n]$  is known with finite precision  $\Delta$  (quantization step or resolution).



# Systems



- DSP enables
- Complex algorithm
  - Reliable computing

Figure 1.11 The three classes of system: analog systems, digital systems, and interface systems from analog-to-digital and digital-to-analog.

# Application example

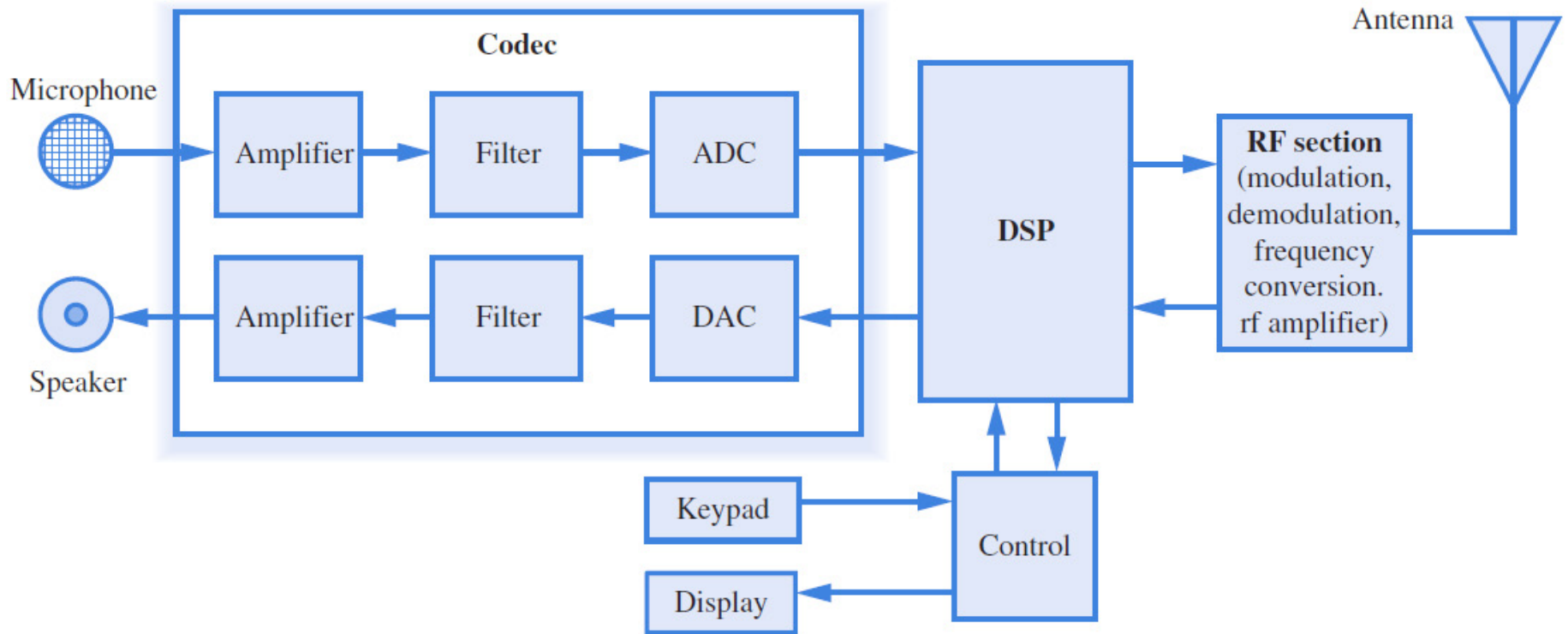


Figure 1.14 Simplified block diagram of a digital cellular phone.



# Chap 2 Discrete-time signals and systems

- 2.1 Discrete-time signals
- 2.3 Discrete-time systems
- 2.4 Convolution of LTI systems
- 2.5 Properties of LTI systems
- 2.9 FIR spatial filters
- 2.11 Continuous-time LTI systems



# Discrete-time sampling

Sampling:  
 $x[n] = x(nT)$

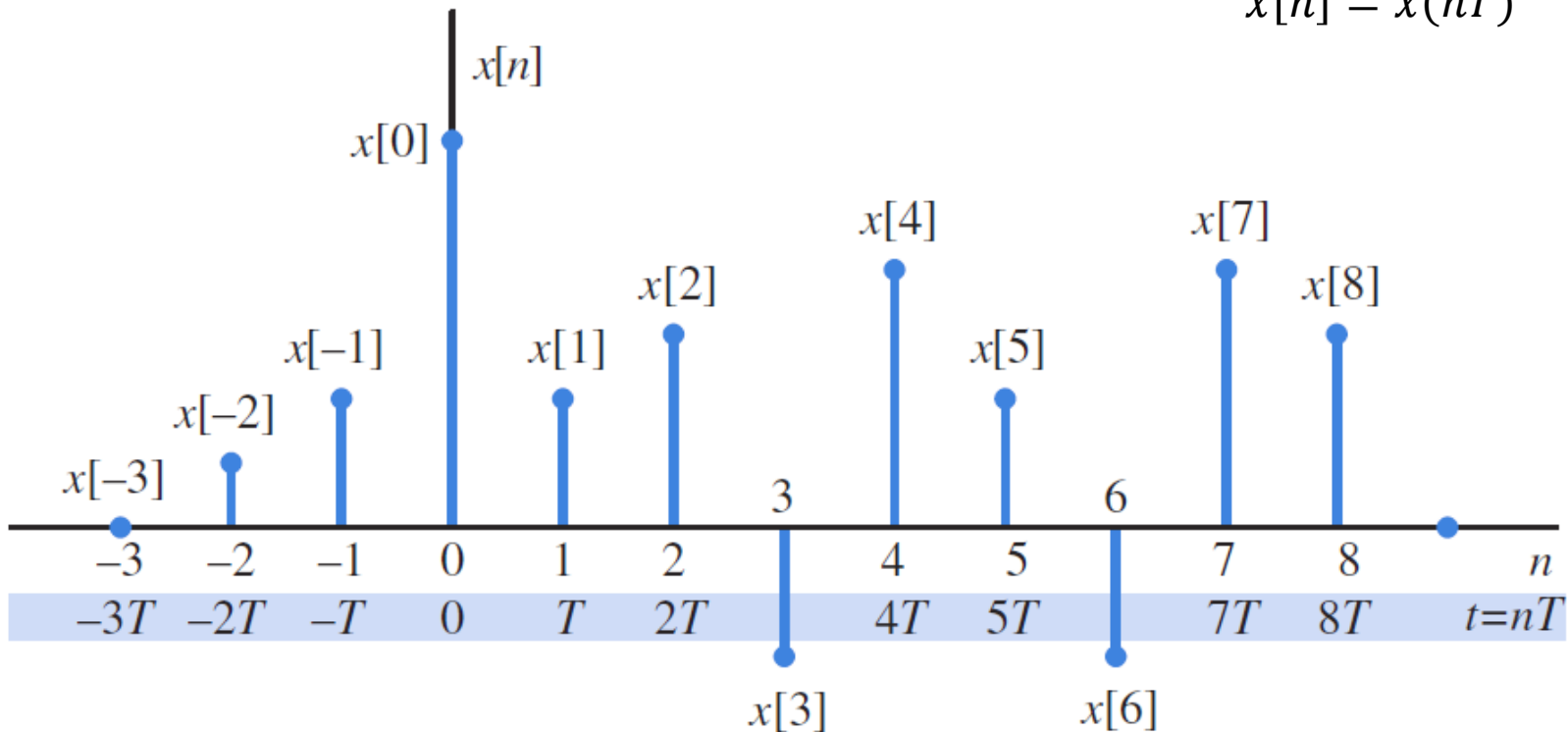


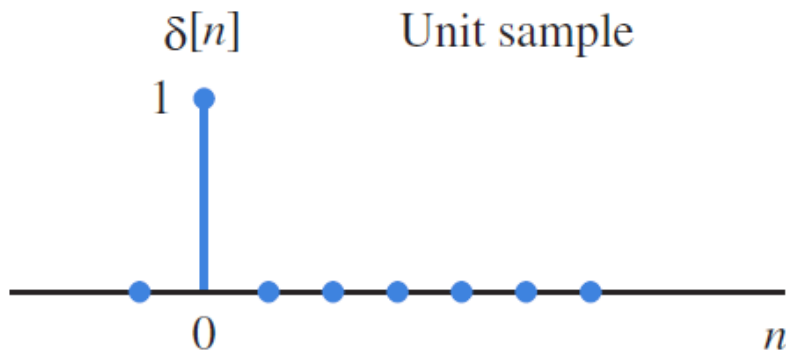
Figure 2.1 Representation of a sampled signal.



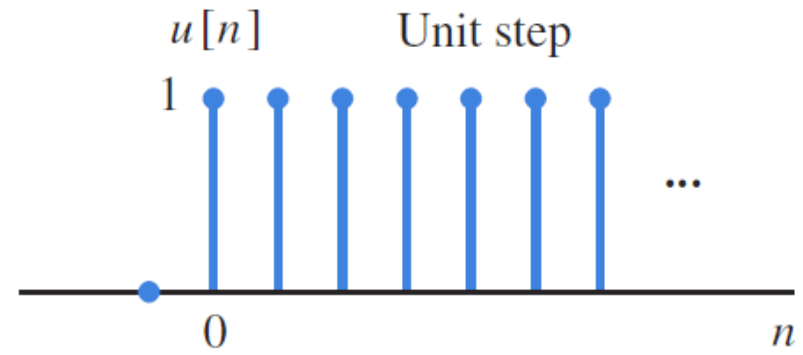
# Elementary discrete-time signals

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

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



(a)

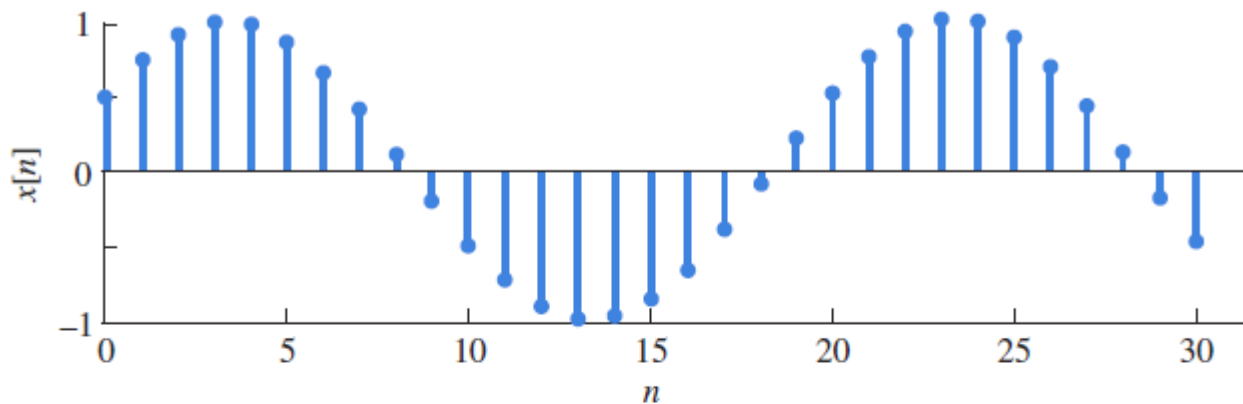


(b)

Figure 2.2 Some elementary discrete-time signals.

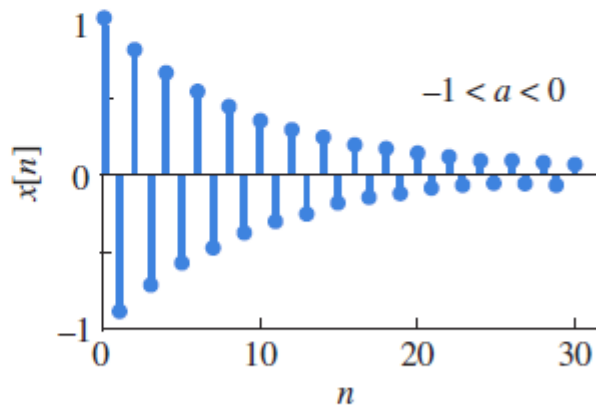
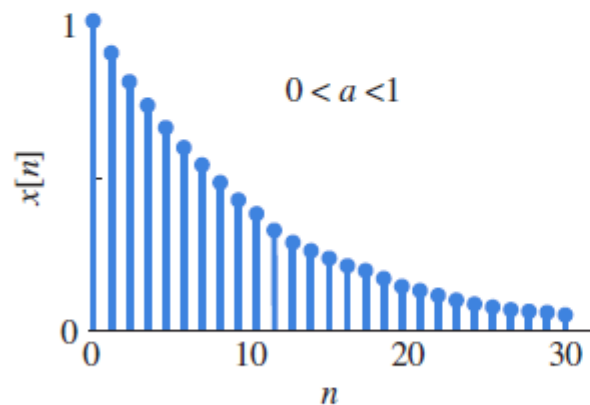


# Other examples



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

(a)



$$x[n] \triangleq Aa^n$$

(b)

Figure 2.3 Examples of a discrete-time sinusoidal signal (a), and two real exponential sequences (b).



# Discrete-time systems

$$x[n] \xrightarrow{\mathcal{H}} y[n] \quad \text{or} \quad y[n] = \mathcal{H}\{x[n]\}$$

Operator



Figure 2.5 Block diagram representation of a discrete-time system.



# Causality and stability

**Definition 2.1** A system is called *causal* if the present value of the output does not depend on future values of the input, that is,  $y[n_0]$  is determined by the values of  $x[n]$  for  $n \leq n_0$ , only.

**Definition 2.2** A system is said to be *stable*, in the Bounded-Input Bounded-Output (BIBO) sense, if every bounded input signal results in a bounded output signal, that is

$$|x[n]| \leq M_x < \infty \Rightarrow |y[n]| \leq M_y < \infty. \quad (2.18)$$

A signal  $x[n]$  is bounded if there exists a positive finite constant  $M_x$  such that  $|x[n]| \leq M_x$  for all  $n$ .



# Linearity

**Definition 2.3** A system is called *linear* if and only if for every real or complex constant  $a_1, a_2$  and every input signal  $x_1[n]$  and  $x_2[n]$

$$\mathcal{H}\{a_1x_1[n] + a_2x_2[n]\} = a_1\mathcal{H}\{x_1[n]\} + a_2\mathcal{H}\{x_2[n]\}, \quad (2.19)$$

for all values of  $n$ .

Example: Is a square-law system linear or nonlinear?

$$y[n] = x^2[n]$$



# Time invariance

**Definition 2.4** A system is called *time-invariant* or *fixed* if and only if

$$y[n] = \mathcal{H}\{x[n]\} \Rightarrow y[n - n_0] = \mathcal{H}\{x[n - n_0]\}, \quad (2.22)$$

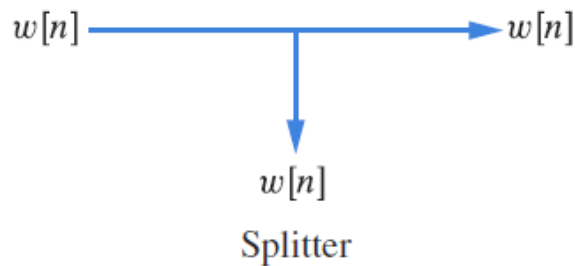
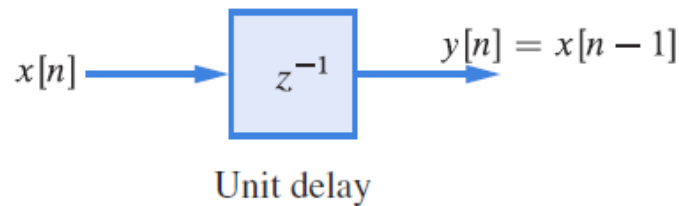
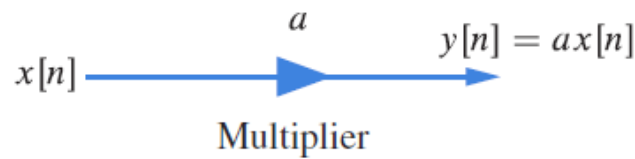
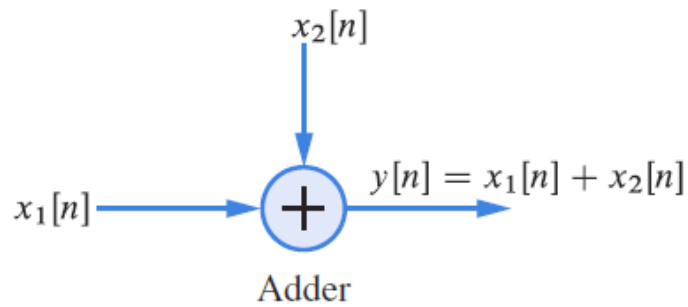
for every input  $x[n]$  and every time shift  $n_0$ . That is, a time shift in the input results in a corresponding time shift in the output.

Example: Is a downsampler linear? Time-invariant?

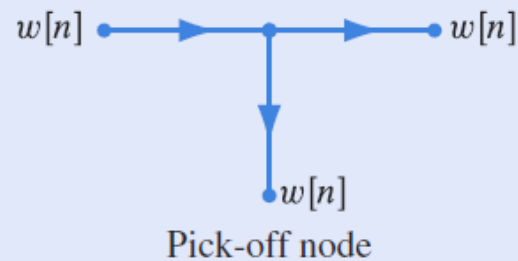
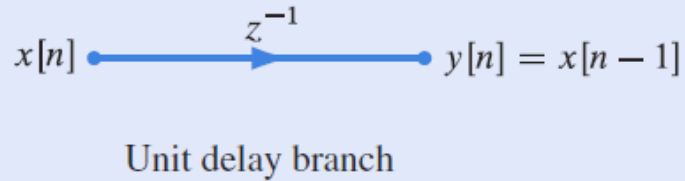
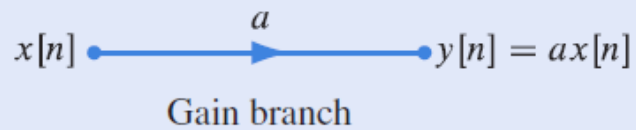
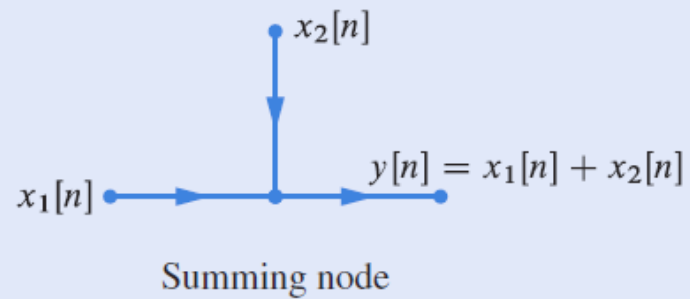
$$y[n] = \mathcal{H}\{x[n]\} = x[nM]$$



### Block Diagram Elements



### Signal Flow Graph Elements



**Figure 2.6** Basic building blocks and the corresponding signal flow graph elements for the implementation of discrete-time systems.



# SFG example

$$w[n] = x[n] + aw[n - 1], \quad (\text{input node})$$

$$y[n] = w[n] + bw[n - 1]. \quad (\text{output node})$$

$$y[n] = x[n] + bx[n - 1] + ay[n - 1]$$

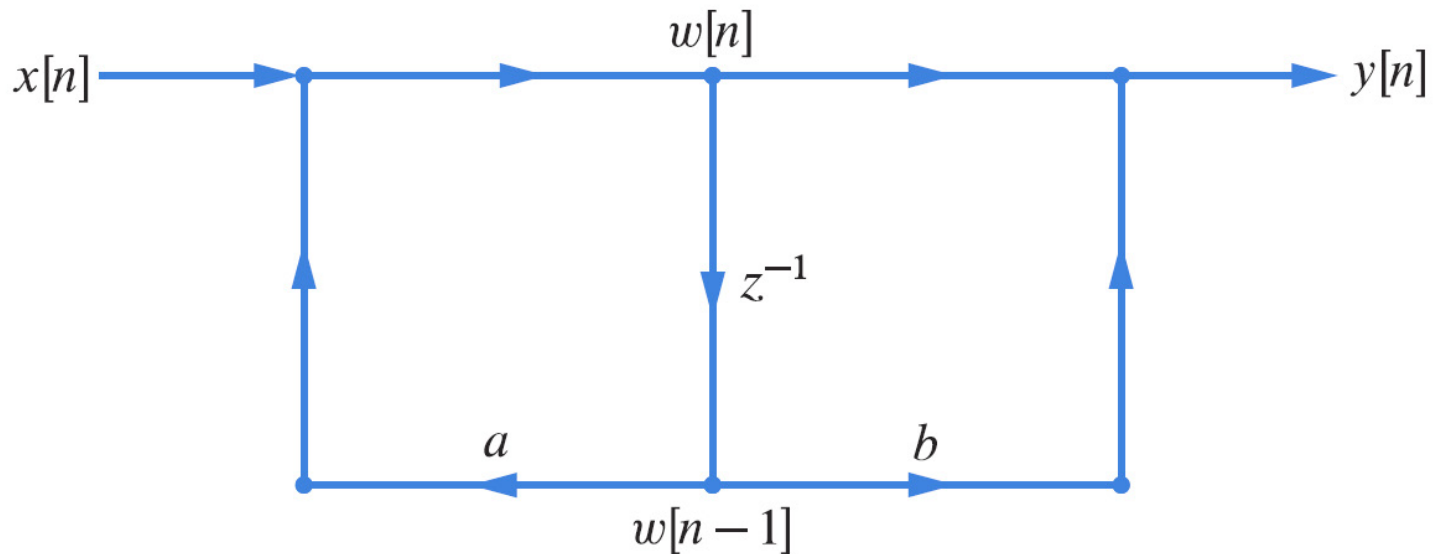


Figure 2.8 Signal flow graph of a first-order discrete-time system.



# Convolution of LTI systems

$$x[n] = \sum_{k=-\infty}^{\infty} x[k]\delta[n - k] \xrightarrow{\mathcal{H}} y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n - k]$$

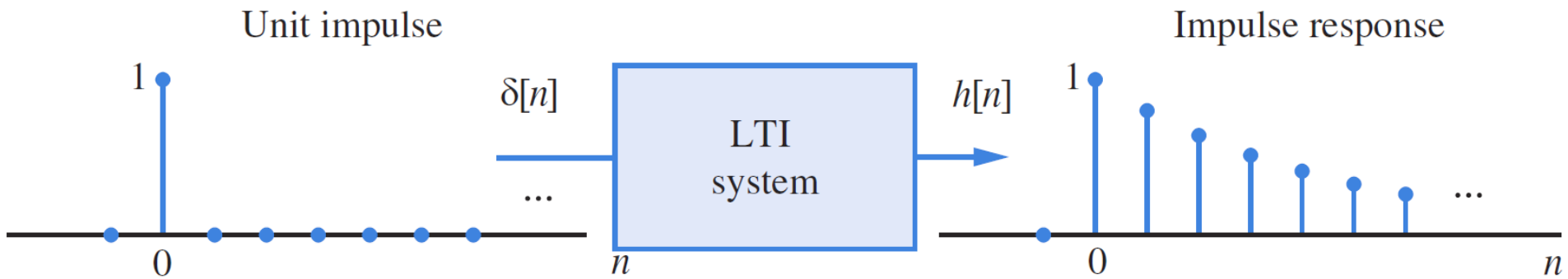


Figure 2.9 The impulse response of a linear time-invariant system.

An LTI system is represented by its impulse response.





# Finite impulse response (FIR) system

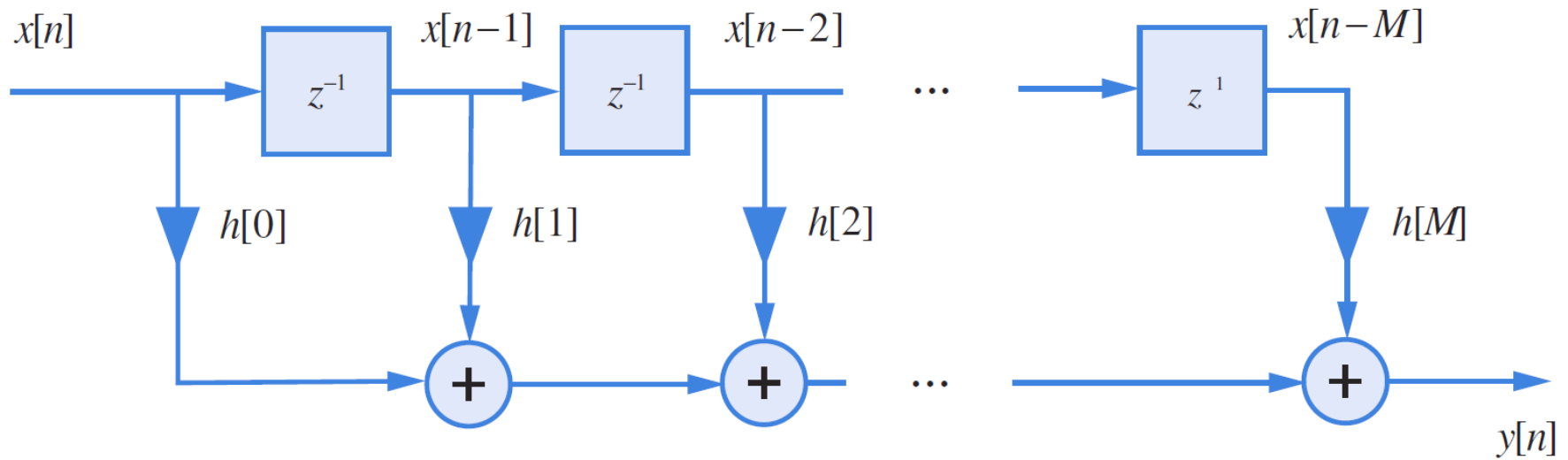


Figure 2.14 Block diagram representation of an FIR system.



# Properties of convolution

$$y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k] = \sum_{m=-\infty}^{\infty} h[m]x[n-m] = h[n] * x[n]$$

**Table 2.3** Summary of convolution properties.

Property	Formula
Identity	$x[n] * \delta[n] = x[n]$
Delay	$x[n] * \delta[n - n_0] = x[n - n_0]$
Commutative	$x[n] * h[n] = h[n] * x[n]$
Associative	$(x[n] * h_1[n]) * h_2[n] = x[n] * (h_1[n] * h_2[n])$
Distributive	$x[n] * (h_1[n] + h_2[n]) = x[n] * h_1[n] + x[n] * h_2[n]$



# Causality and stability

**Result 2.5.1** A linear time-invariant system with impulse response  $h[n]$  is causal if

$$h[n] = 0 \quad \text{for} \quad n < 0. \quad (2.50)$$

**Result 2.5.2** A linear time-invariant system with impulse response  $h[n]$  is stable, in the bounded-input bounded-output sense, if and only if the impulse response is absolutely summable, that is, if

$$\sum_{n=-\infty}^{\infty} |h[n]| < \infty. \quad (2.52)$$



# Examples of LTI response

**Table 2.4** Response of linear time-invariant systems to some test sequences.

Type of response	Input sequence		Output sequence
Impulse	$x[n] = \delta[n]$	$\xrightarrow{\mathcal{H}}$	$y[n] = h[n]$
Step	$x[n] = u[n]$	$\xrightarrow{\mathcal{H}}$	$y[n] = s[n] = \sum_{k=-\infty}^n h[k]$
Exponential	$x[n] = a^n, \text{ all } n$	$\xrightarrow{\mathcal{H}}$	$y[n] = H(a)a^n, \text{ all } n$
Complex sinusoidal	$x[n] = e^{j\omega n}, \text{ all } n$	$\xrightarrow{\mathcal{H}}$	$y[n] = H(e^{j\omega})e^{j\omega n}, \text{ all } n$

$$H(a) = \sum_{-\infty}^{\infty} h[n]a^{-n}$$

Frequency response



# FIR spatial (2D) filter

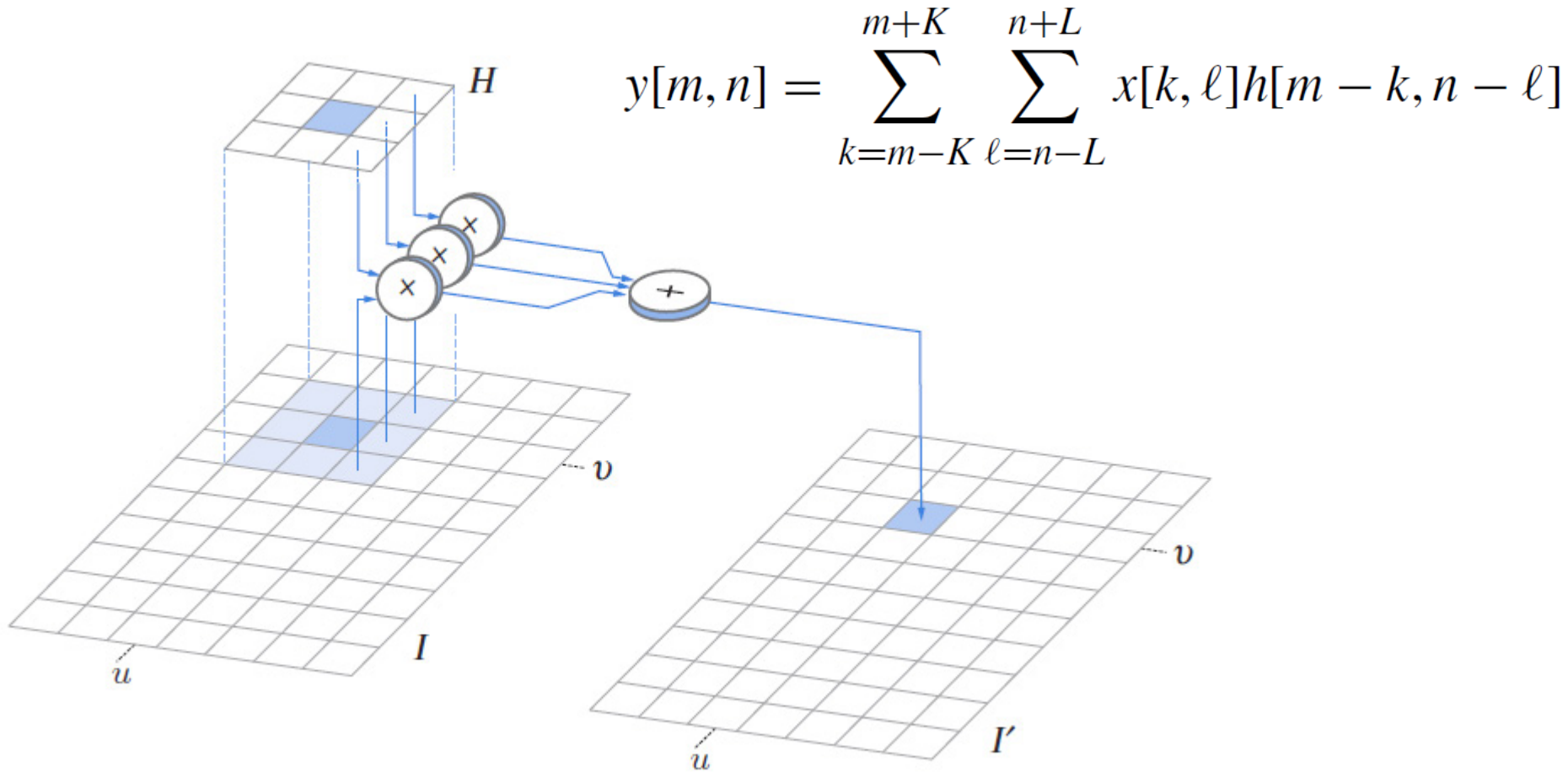
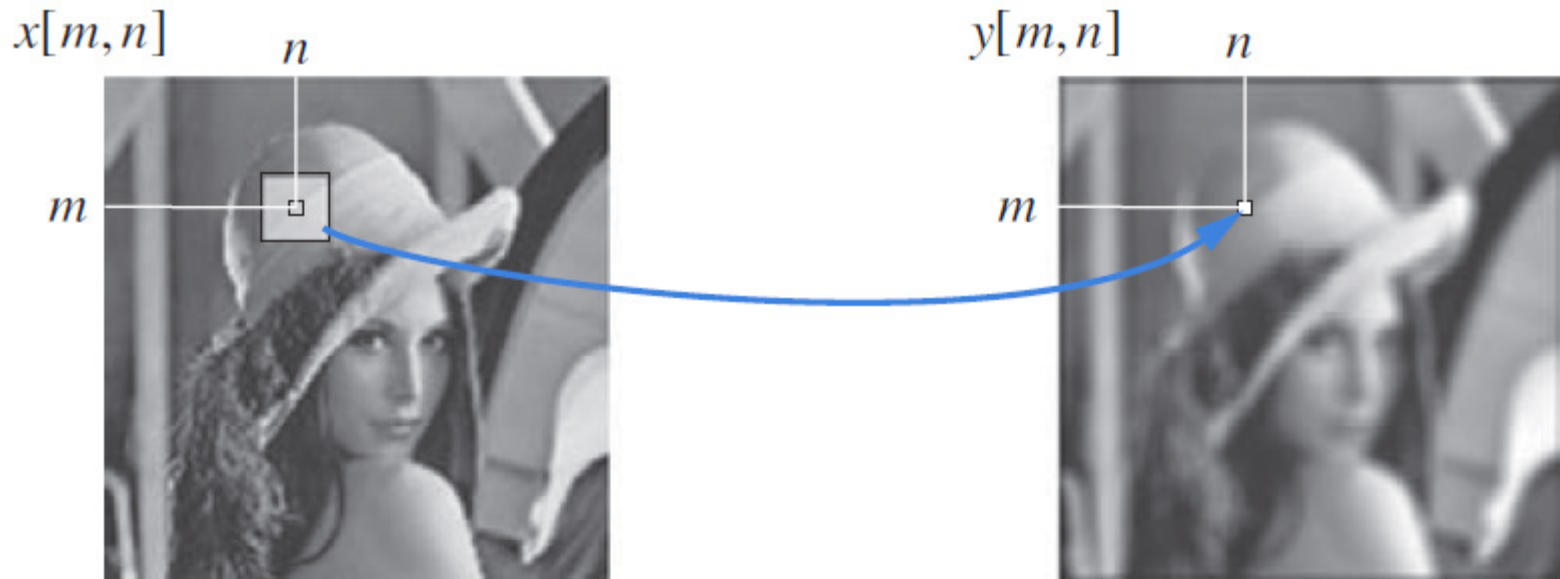


Figure 2.24 FIR spatial filter implementation.

# Example of 3x3 average filter

$$y[m, n] = \sum_{k=-1}^1 \sum_{\ell=-1}^1 \left(\frac{1}{9}\right) x[m - k, n - \ell]$$



**Figure 2.23** The FIR spatial filtering operation.



# Continuous-time LTI systems

$$\delta(t) \xrightarrow{\mathcal{H}} h(t) \quad \text{(Impulse Response)}$$

$$\delta(t - \tau) \xrightarrow{\mathcal{H}} h(t - \tau) \quad \text{(Time - Invariance)}$$

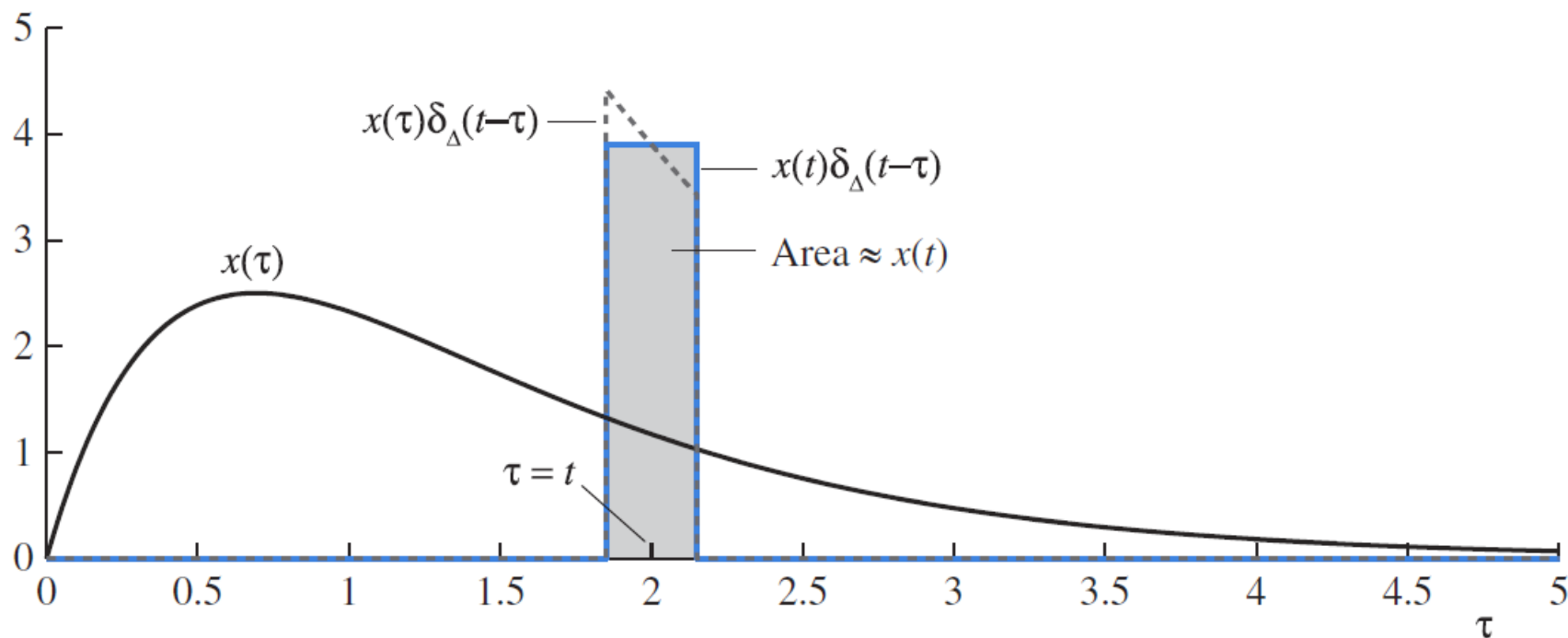
$$x(\tau)\delta(t - \tau) \xrightarrow{\mathcal{H}} x(\tau)h(t - \tau) \quad \text{(Homogeneity)}$$

$$\underbrace{\int_{-\infty}^{\infty} x(\tau)\delta(t - \tau)d\tau}_{x(t)} \xrightarrow{\mathcal{H}} \underbrace{\int_{-\infty}^{\infty} x(\tau)h(t - \tau)d\tau}_{y(t)}, \quad \text{(Additivity)}$$



# Continuous-time impulse function

$$\delta_{\Delta}(t) = \begin{cases} 1/\Delta, & -\Delta/2 < t < \Delta/2 \\ 0, & \text{otherwise} \end{cases}$$



**Figure 2.30** Interpretation of convolution by a narrow pulse as a scanning operation.