



Digital Signal
Processing- Lecture 8

Topics to be covered:

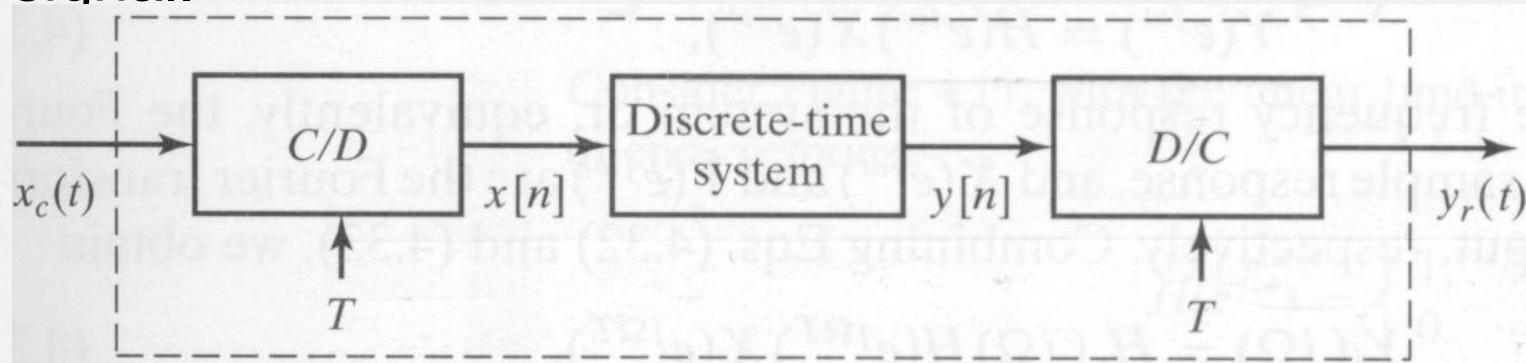
- Sampling of continuous time signal

Sampling of Continuous-Time Signals

- Introduction
- Periodic Sampling
- Frequency-Domain Representation of Sampling
- Reconstruction of a Bandlimited Signal from its Samples
- Discrete-Time Processing of Continuous-Time signals

Introduction

- Continuous-time signal processing can be implemented through a process of **sampling**, **discrete-time processing**, and the subsequent **reconstruction** of a continuous-time signal.

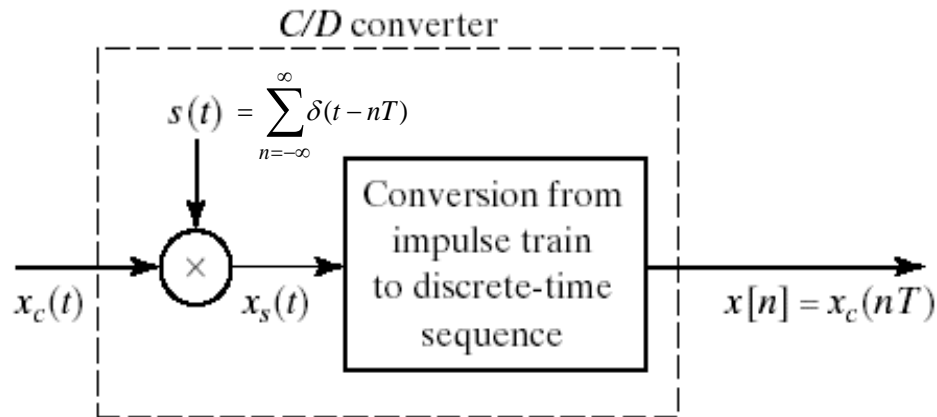


$$x[n] = x_c(nT), \\ -\infty < n < \infty$$

T: sampling period
 $f_s = 1/T$: sampling frequency

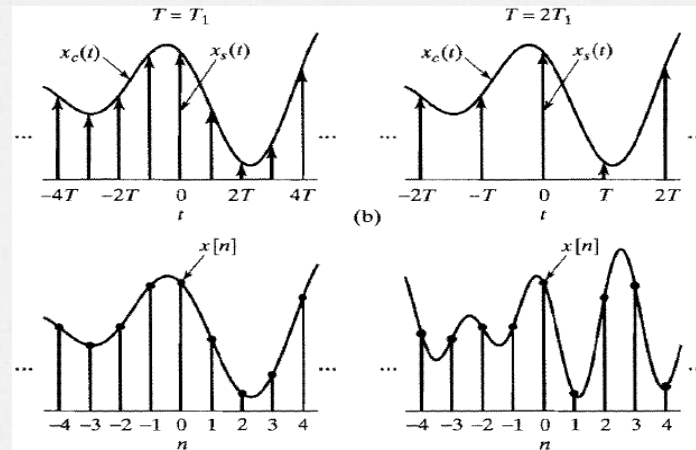
$$\Omega_s = 2\pi/T, \quad (\text{rad} / \text{s})$$

Continuous-time signal



Periodic Sampling

T: sampling period



Frequency-Domain Representation of Sampling

$$s(t) = \sum_{n=-\infty}^{\infty} \delta(t - nT)$$

T : sample period; $f_s = 1/T$: sample rate
 $\Omega_s = 2\pi/T$: sample rate

$$x_s(t) = x_c(t)s(t) = x_c(t) \sum_{n=-\infty}^{\infty} \delta(t - nT) = \sum_{n=-\infty}^{\infty} x_c(nT) \delta(t - nT)$$

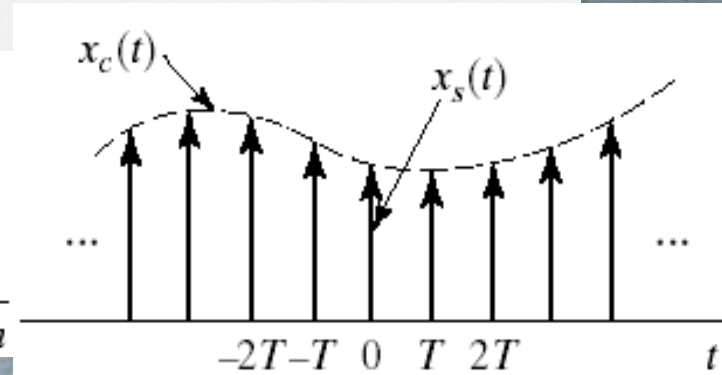
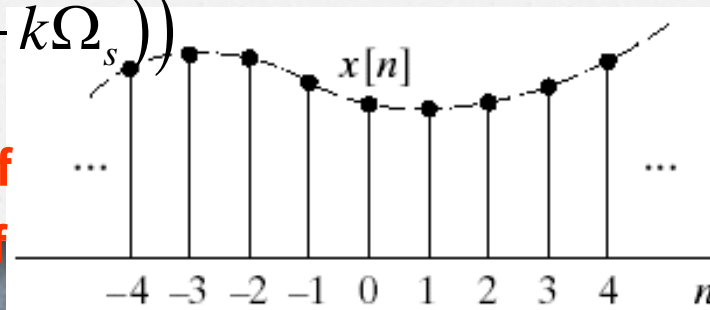
$$x[n] = x_c(t) \big|_{t=nT} = x_c(nT) \quad S(j\Omega) = \frac{2\pi}{T} \sum_{k=-\infty}^{\infty} \delta(\Omega - k\Omega_s)$$

$$X_s(j\Omega) = \frac{1}{2\pi} X_c(j\Omega) * S(j\Omega) = \frac{1}{2\pi} \int_{-\infty}^{\infty} S(j\theta) X_c(j(\Omega - \theta)) d\theta$$

$$= \frac{1}{2\pi} \int_{-\infty}^{\infty} \frac{2\pi}{T} \sum_{k=-\infty}^{\infty} \delta(\theta - k\Omega_s) X_c(j(\Omega - \theta)) d\theta = \frac{1}{T} \sum_{k=-\infty}^{\infty} \int_{-\infty}^{\infty} \delta(\theta - k\Omega_s) X_c(j(\Omega - \theta)) d\theta$$

$$= \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c(j(\Omega - k\Omega_s))$$

Representation of $X_s(j\Omega)$ in terms of $X(e^{j\omega})$

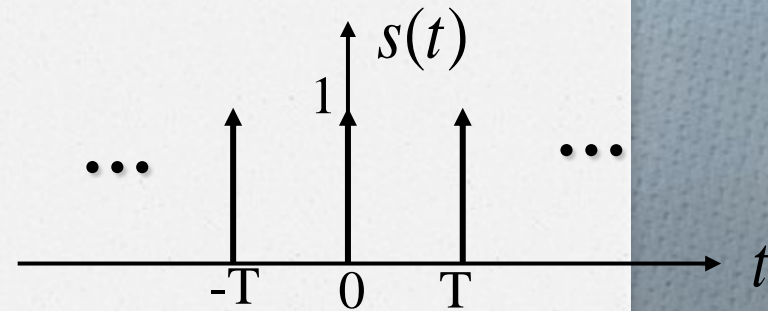


$$S(j\Omega) = \frac{2\pi}{T} \sum_{k=-\infty}^{\infty} \delta(\Omega - k\Omega_s)$$

T: sample period; **fs=1/T**:sample rate;**Ωs=2π/T**:sample rate

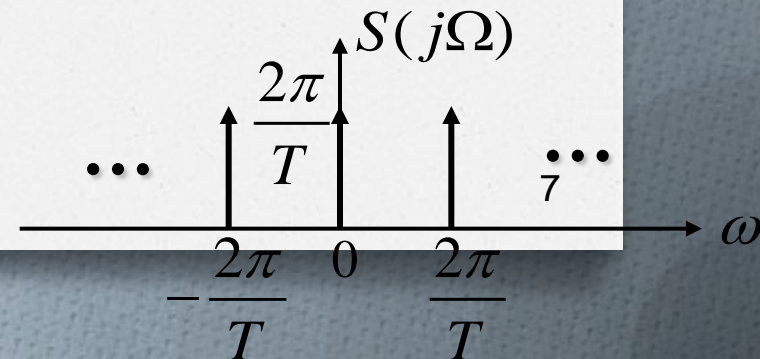
$$s(t) = \sum_{n=-\infty}^{\infty} \delta(t - nT) = \sum_{n=-\infty}^{\infty} a_k e^{jk\Omega_s t} = \frac{1}{T} \sum_{n=-\infty}^{\infty} e^{jk\Omega_s t}$$

$$a_k = \frac{1}{T} \int_{-T/2}^{+T/2} \delta(t) e^{-jk\Omega_s t} dt = \frac{1}{T}$$



$$e^{jk\Omega_s t} \xleftrightarrow{F} 2\pi\delta(\Omega - k\Omega_s)$$

$$S(j\Omega) = \frac{2\pi}{T} \sum_{k=-\infty}^{\infty} \delta(\Omega - k\Omega_s)$$



Representation of $X(e^{j\omega})$ in terms of $X_s(j\Omega)$, $X_c(j\Omega)$

$$x_s(t) = x_c(t)s(t) = x_c(t) \sum_{n=-\infty}^{\infty} \delta(t - nT) = \sum_{n=-\infty}^{\infty} x_c(nT) \delta(t - nT)$$

$$X_s(j\Omega) = \int_{-\infty}^{\infty} \sum_{n=-\infty}^{\infty} x_c(nT) \delta(t - nT) e^{-j\Omega t} dt$$

$$= \sum_{k=-\infty}^{\infty} x_c(nT) e^{-j\Omega T n} \quad x[n] = x_c(nT)$$

$\Omega T = \omega$

$$X(e^{j\omega}) = \sum_{n=-\infty}^{\infty} x_c(nT) e^{-j\omega n} = X(e^{j\Omega T})$$

DTFT

$$= X_s(j\Omega) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c(j(\Omega - k\Omega_s))$$

$$\Omega_s = \frac{2\pi}{T}$$

Representation of $X(e^{j\omega})$ in terms of $X_s(j\Omega)$, $X_c(j\Omega)$

$$X(e^{j\omega}) = X(e^{j\Omega T}) = X_s(j\Omega) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c(j(\Omega - k\Omega_s))$$

DTFT

$$\Omega = \omega / T$$

Continuous FT

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

$$\text{if } X_c(j\Omega) = 0, \quad \Omega \geq \frac{\pi}{T}$$

$$\text{then } X(e^{j\omega}) = \frac{1}{T} X_c\left(j\frac{\omega}{T}\right)$$

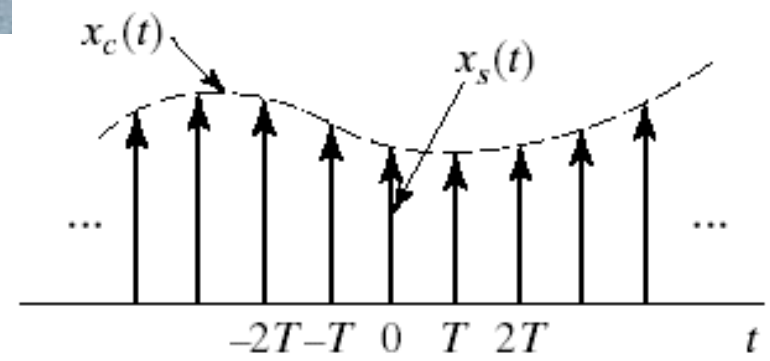
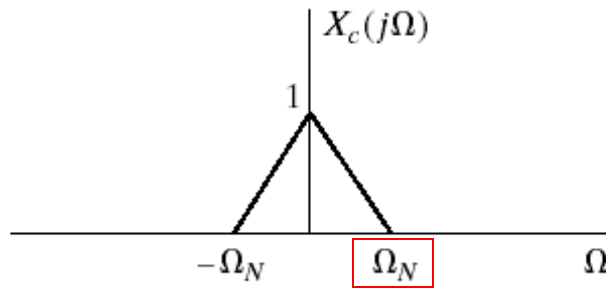
$$|\omega| < \pi$$

Nyquist Sampling Theorem

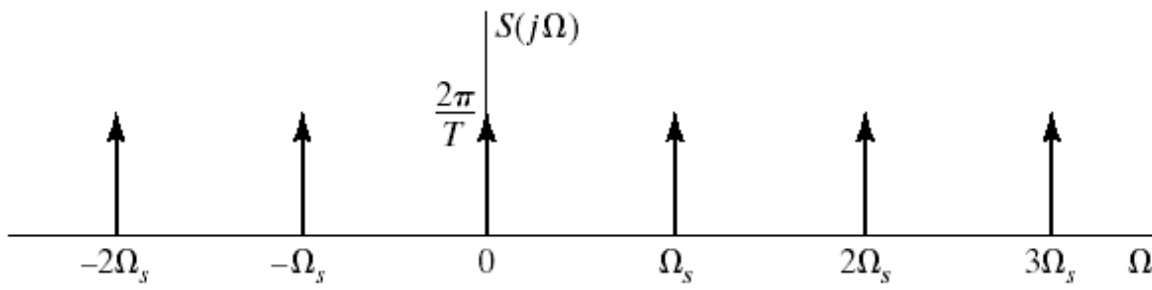
- Let $x_c(t)$ be a bandlimited signal with $X_c(j\Omega) = 0$ for $|\Omega| \geq \Omega_N$. Then $x_c(t)$ is uniquely determined by its samples $x[n] = x_c(nT)$, $n = 0, \pm 1, \pm 2, \dots$, if

$$\Omega_s = \frac{2\pi}{T} \geq 2\Omega_N$$

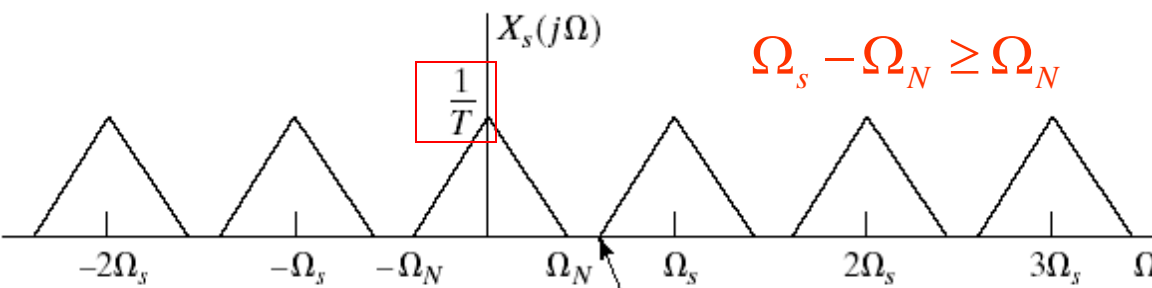
- The frequency Ω_N is commonly referred as the *Nyquist frequency*.
- The frequency $2\Omega_N$ is called the *Nyquist rate*.



frequency spectrum of ideal sample signal



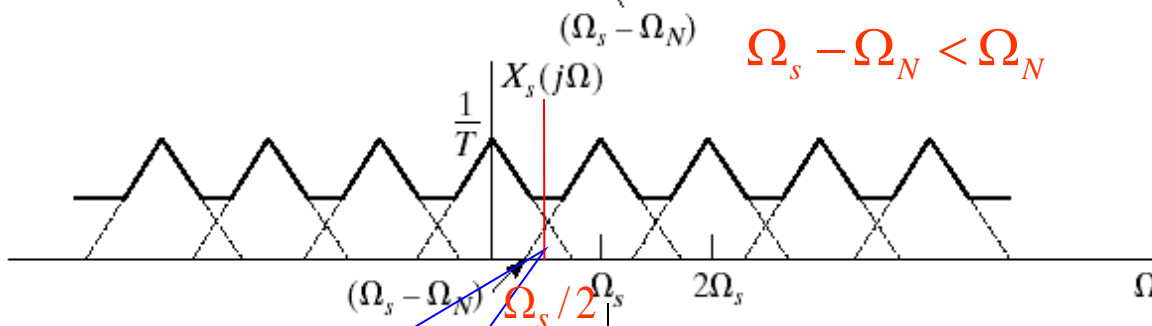
$$X_s(j\Omega) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c(j(\Omega - k\Omega_s))$$



$$\Omega_s - \Omega_N \geq \Omega_N$$

No aliasing

$$X(e^{j\omega}) = X_s(j\Omega) |_{\Omega=\omega/T}$$



$$\Omega_s - \Omega_N < \Omega_N$$

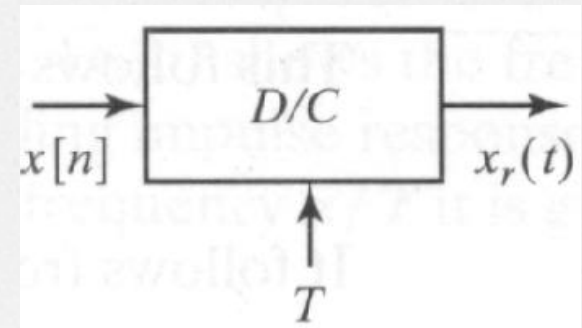
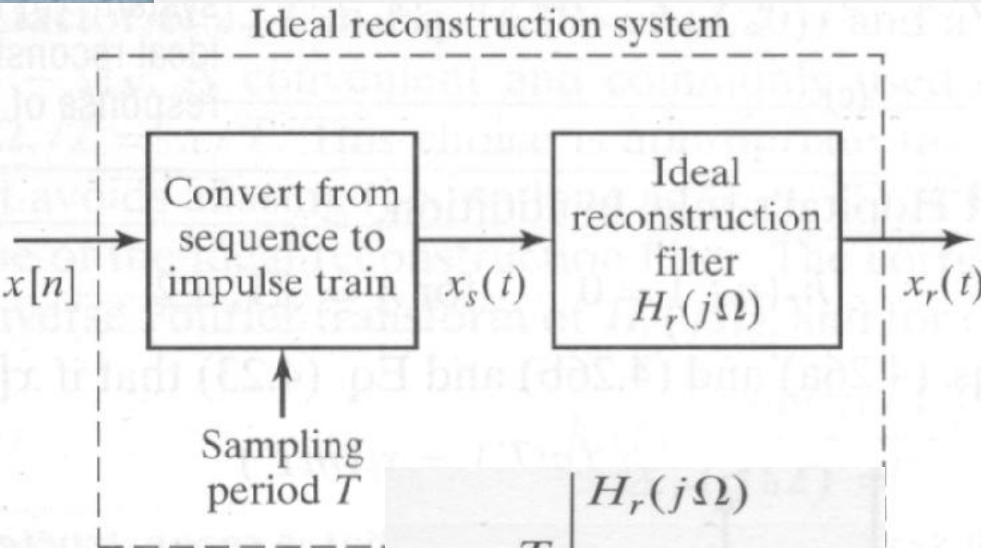
$$= \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c(j(\omega - k2\pi)/T)$$

aliasing frequency

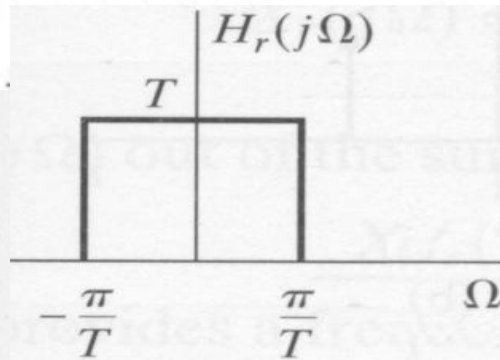


aliasing

Reconstruction of a Bandlimited Signal from its Samples



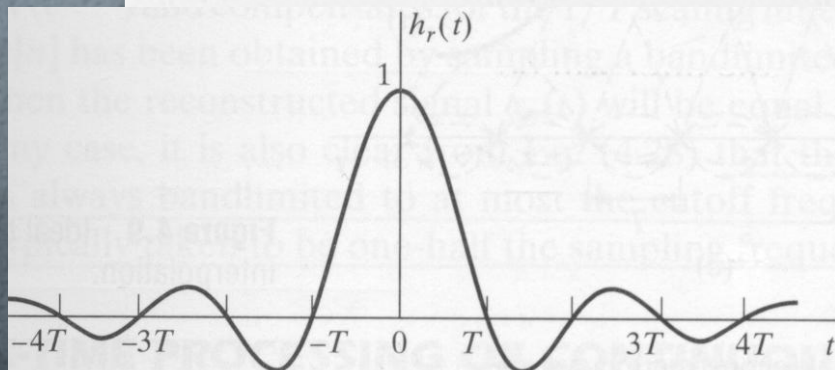
Gain: T



$$h_r(t) = \frac{\sin(\pi t/T)}{\pi t/T}$$

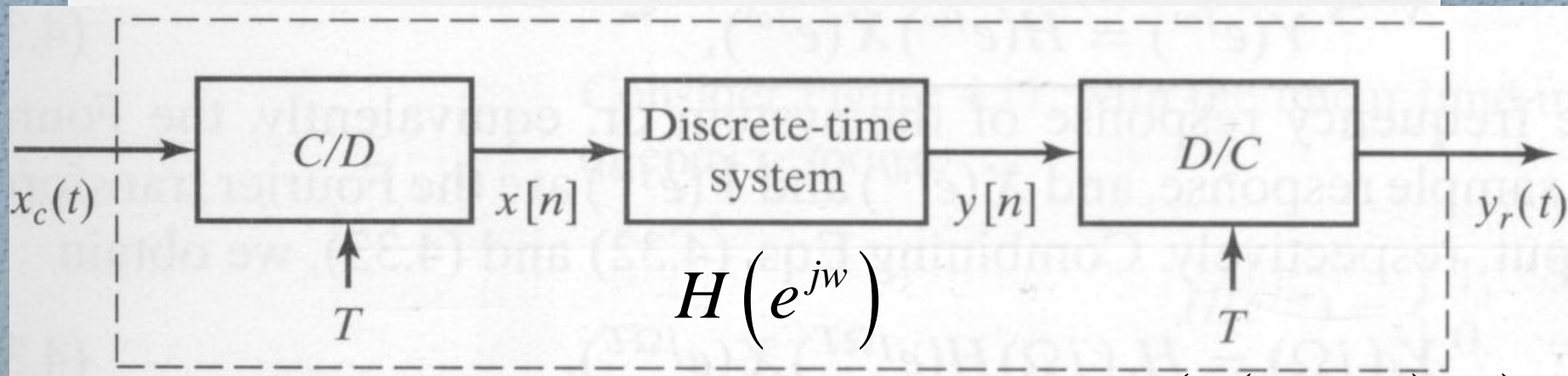
$$x_r(t) = \sum_{n=-\infty}^{\infty} x[n] h_r(t - nT)$$

$$= \sum_{n=-\infty}^{\infty} x[n] \frac{\sin[\pi(t - nT)/T]}{\pi(t - nT)/T}$$



$$X_r(j\Omega) = H_r(j\Omega) X(e^{j\Omega T})$$

Discrete-Time Processing of Continuous-Time signals



$$x[n] = x_c(nT) \quad y_r(t) = \sum_{n=-\infty}^{\infty} y[n] \frac{\sin(\pi(t - nT)/T)}{\pi(t - nT)/T}$$

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

$$= \begin{cases} TY(e^{j\Omega T}), & |\Omega| < \frac{\pi}{T} \\ 0, & \text{otherwise} \end{cases}$$

C/D Converter

- Output of C/D Converter

$$x[n] = x_c(nT)$$

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

D/C Converter

○ Output of D/C Converter

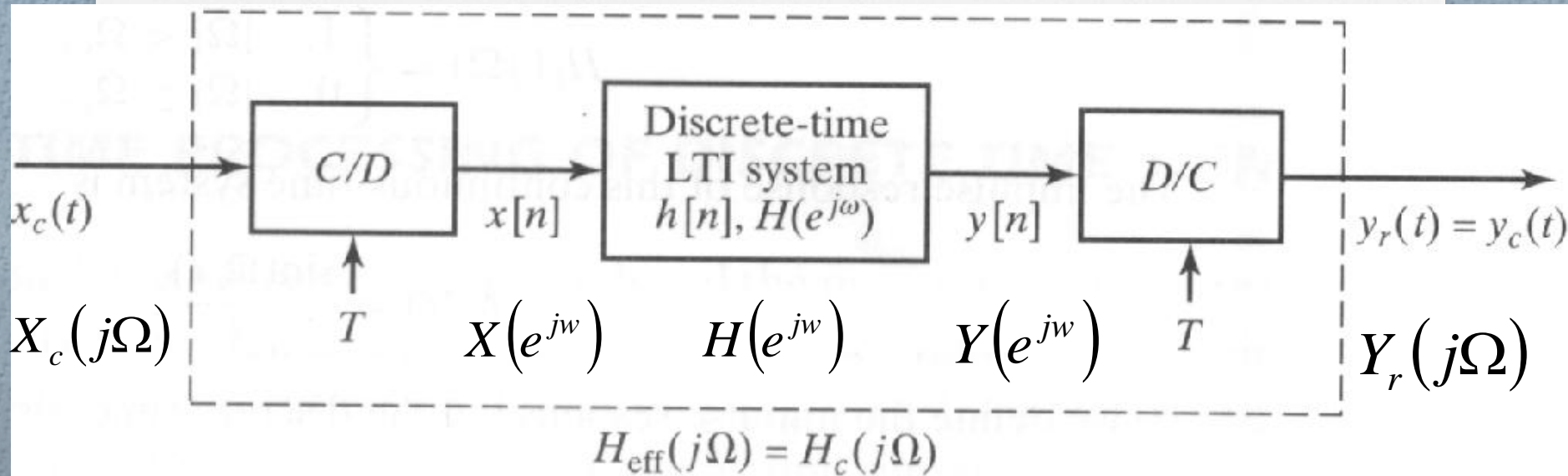
$$y_r(t) = \sum_{n=-\infty}^{\infty} y[n] \frac{\sin(\pi(t - nT)/T)}{\pi(t - nT)/T}$$

$$Y_r(j\Omega) = H_r(j\Omega)Y(e^{j\Omega T})$$

$$= \begin{cases} TY(e^{j\Omega T}), & |\Omega| < \frac{\pi}{T} \\ 0, & \text{otherwise} \end{cases}$$

$$H_r(j\Omega) = \begin{cases} T, & |\Omega| < \frac{\pi}{T} \\ 0, & \text{otherwise} \end{cases}$$

Linear Time-Invariant Discrete-Time Systems



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

$$Y_r(j\Omega) = H_r(j\Omega)H(e^{j\Omega T})X(e^{j\Omega T})$$

$$= H_r(j\Omega)H(e^{j\Omega T})\frac{1}{T}\sum_{k=-\infty}^{\infty}X_c\left(j\left(\Omega - \frac{2\pi k}{T}\right)\right) = \begin{cases} H(e^{j\Omega T})X_c(j\Omega), & |\Omega| < \frac{\pi}{T} \\ 0, & |\Omega| \geq \frac{\pi}{T} \end{cases}$$

Linear and Time-Invariant

- Linear and time-invariant system behavior depends on two factors:
- First, the discrete-time system must be linear and time invariant.
- Second, the input signal must be bandlimited, and the sampling rate must be high enough to satisfy Nyquist Sampling Theorem.

$$Y_r(j\Omega) = H_r(j\Omega)H(e^{j\Omega T})X(e^{j\Omega T})$$

$$= H_r(j\Omega)H(e^{j\Omega T})\frac{1}{T}\sum_{k=-\infty}^{\infty}X_c\left(j\left(\Omega - \frac{2\pi k}{T}\right)\right)$$

If $X_c(j\Omega) = 0$ for $|\Omega| \geq \pi/T$, $H_r(j\Omega) = \begin{cases} T, & |\Omega| < \frac{\pi}{T} \\ 0, & \text{otherwise} \end{cases}$

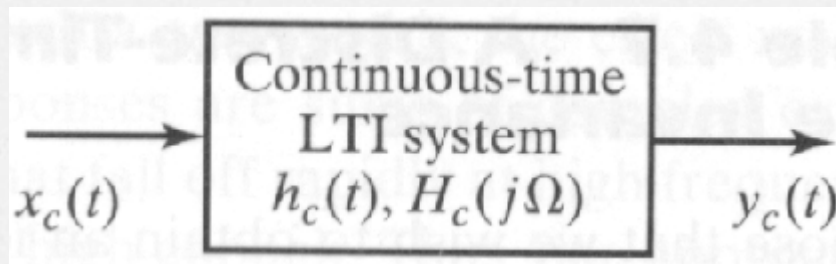
$$Y_r(j\Omega) = \begin{cases} H(e^{j\Omega T})X_c(j\Omega), & |\Omega| < \frac{\pi}{T} \\ 0, & |\Omega| \geq \frac{\pi}{T} \end{cases}$$

$$Y_r(j\Omega) = H_{eff}(j\Omega)X_c(j\Omega)$$

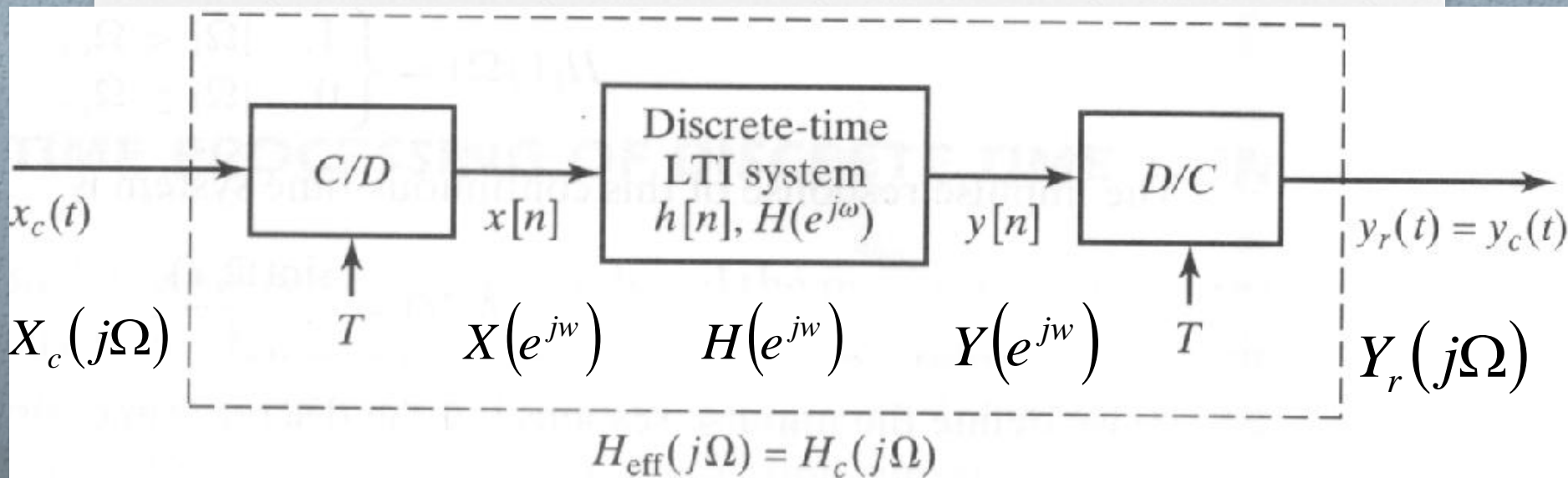
$$H_{eff}(j\Omega) = \begin{cases} H(e^{j\Omega T}), & |\Omega| < \frac{\pi}{T} \\ 0, & |\Omega| \geq \frac{\pi}{T} \end{cases}$$

Impulse Invariance

Given:



Design: $H(e^{j\omega}) \leftarrow H_c(j\Omega)$, $h[n] \leftarrow h_c(nT)$



$$h[n] = Th_c(nT)$$

$$H_c(j\Omega) = H_{\text{eff}}(j\Omega) = \begin{cases} H(e^{j\Omega T}), & |\Omega| < \frac{\pi}{T} \\ 0, & |\Omega| \geq \frac{\pi}{T} \end{cases}$$

impulse-invariant version of the continuous-time system

Impulse Invariance

Two constraints

1. $H(e^{j\omega}) = H_c(j\omega/T), \quad |\omega| < \pi$

2. T is chosen such that

$$H_c(j\Omega) = 0, \quad |\Omega| \geq \pi/T$$

$$\Omega_c < \pi/T$$

$$h[n] = Th_c(nT)$$

The *discrete-time system* is called *an impulse-invariant version* of the *continuous-time system*

$$h[n] = h_c(nT) \longrightarrow X(e^{j\omega}) = \frac{1}{T} X_c\left(j\frac{\omega}{T}\right)$$

$$h[n] = Th_c(nT) \longleftarrow X(e^{j\omega}) = X_c\left(j\frac{\omega}{T}\right) \quad |\omega| < \pi$$