

Digital Signal Processing

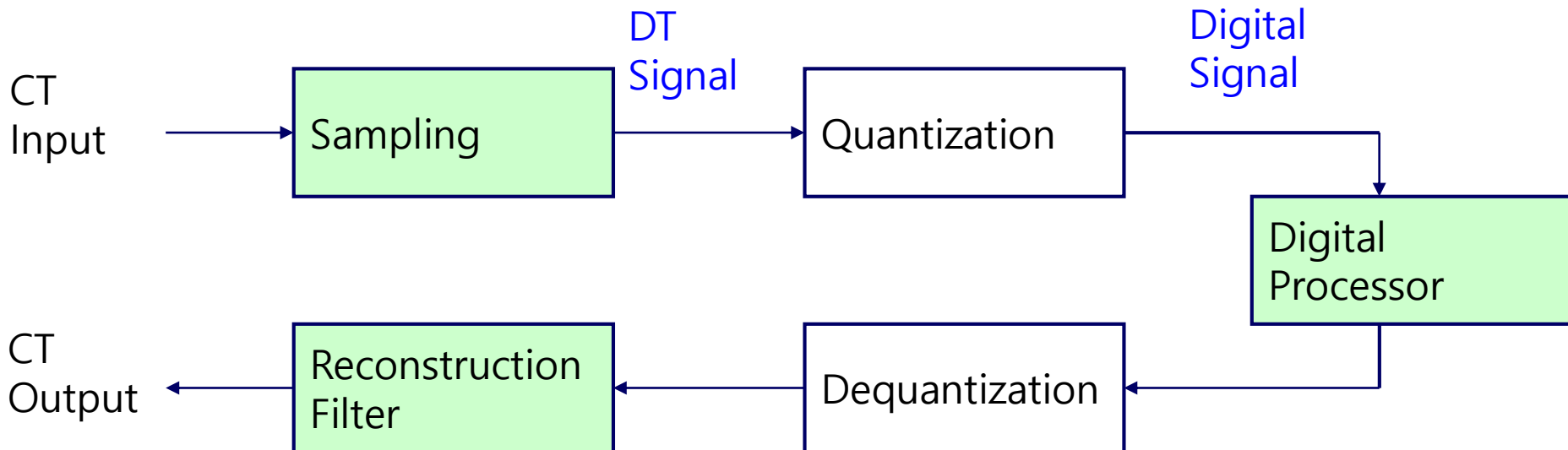
Chap 4. Sampling of Continuous-Time Signals

Chang-Su Kim

Digital Processing of Continuous-Time Signals

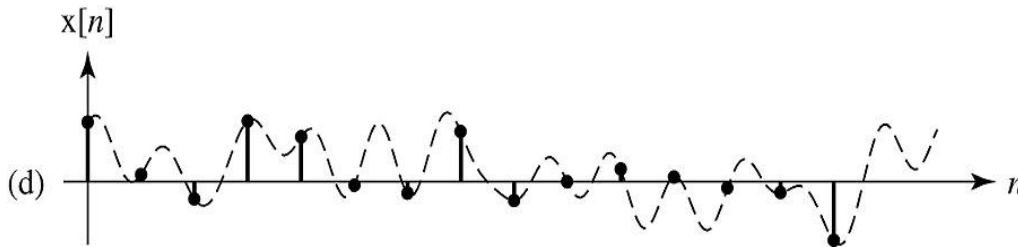
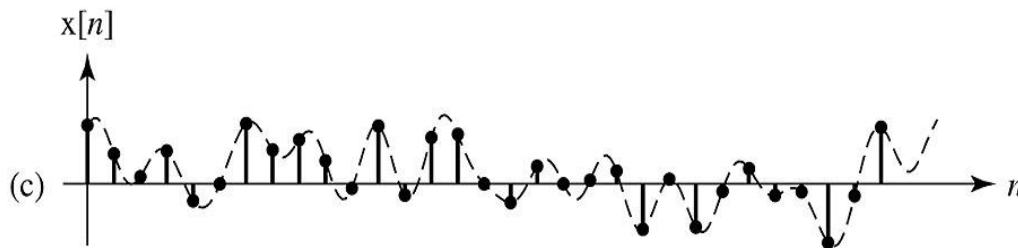
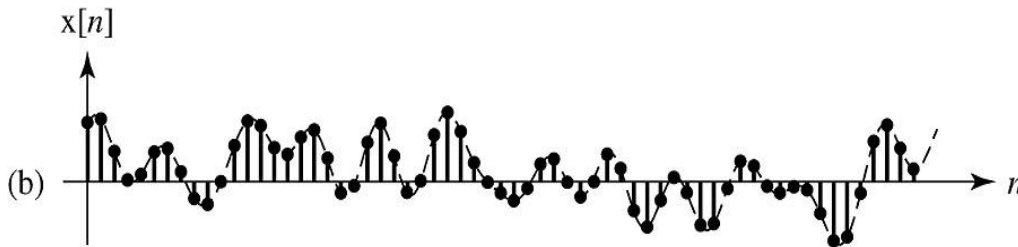
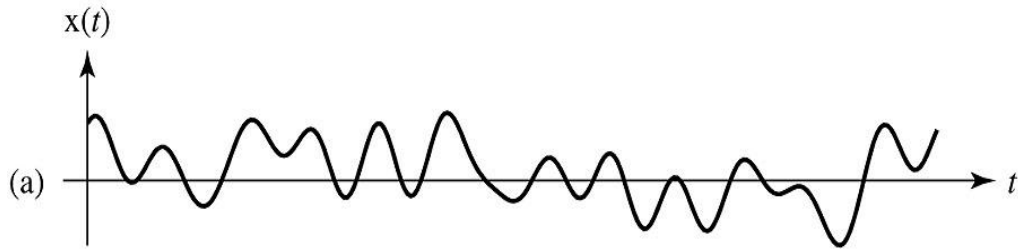
- Digital processing of a CT signal involves three basic steps
 1. Conversion of the CT signal into a DT signal
 2. Processing of the DT signal
 3. Conversion of the processed DT signal back into a CT signal

Conceptual Block diagram



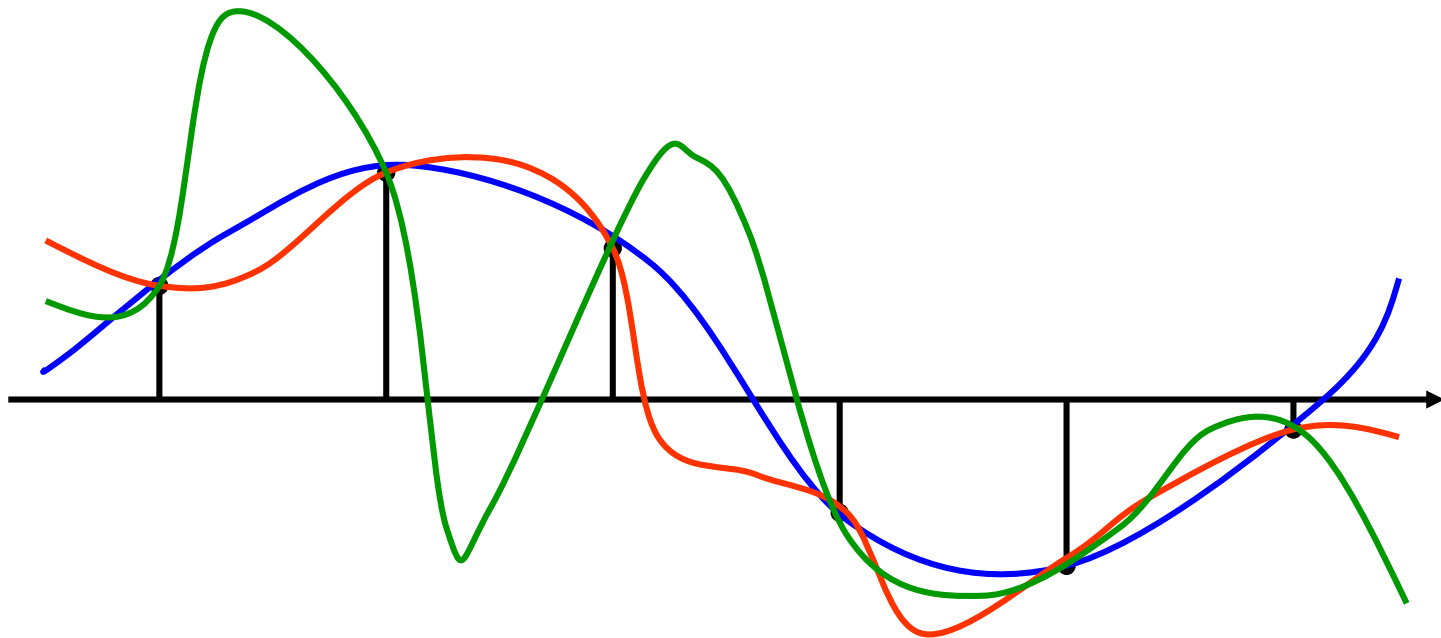
Sampling

Sampling



- Sampling is a procedure to extract a DT signal from a CT signals
- (b), (c), (d) are obtained by sampling (a)
- Is (b) enough to represent (a)?
- What is the adequate sampling rate to represent a given CT signal without information loss?

In general, DT signal cannot represent CT signal perfectly



Are these sample enough to reconstruct the original blue curve?

Continuous-Time Fourier Transform

- CTFT Formulae

- Forward transform

$$X(j\Omega) = \int_{-\infty}^{\infty} x(t)e^{-j\Omega t} dt$$

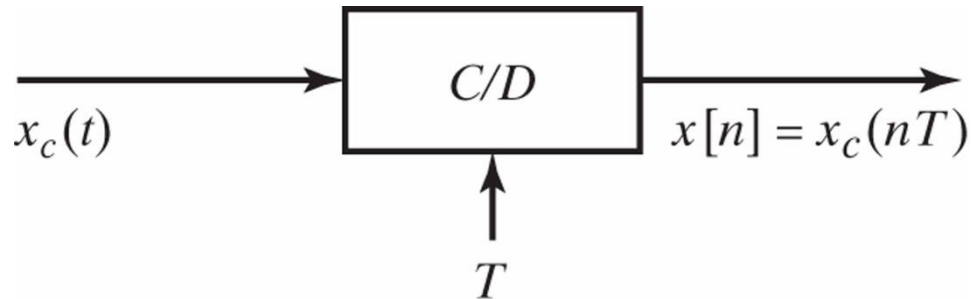
- Inverse transform

$$x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(j\Omega)e^{j\Omega t} d\Omega$$

- We will use a number of properties of CTFT without proofs

- They are studied in the course *Signals and Systems*

Periodic Sampling



- C/D (continuous-time to discrete-time) converter
- $x[n] = x_c(nT)$, $-\infty < n < \infty$.
 - T : sampling period
 - $\Omega_s = \frac{2\pi}{T}$ (or $f_s = \frac{1}{T}$) : sampling frequency

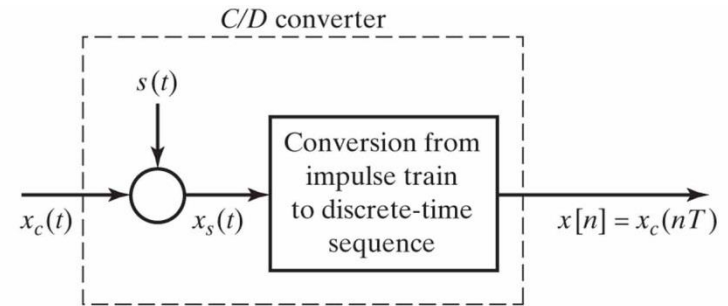
Periodic Sampling

- Conceptually, it is easier to introduce an impulse train for the C/D conversion

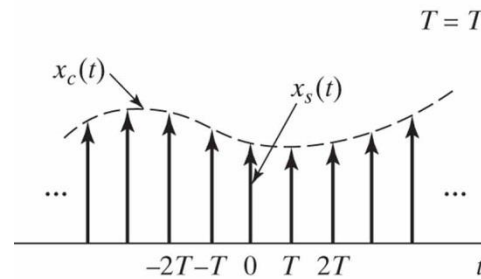
$$s(t) = \sum_{n=-\infty}^{\infty} \delta(t - nT)$$

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

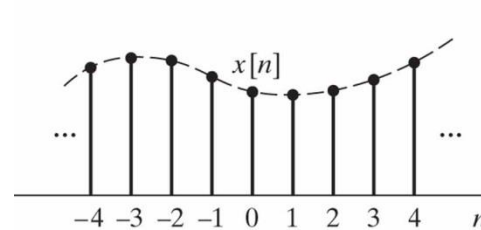
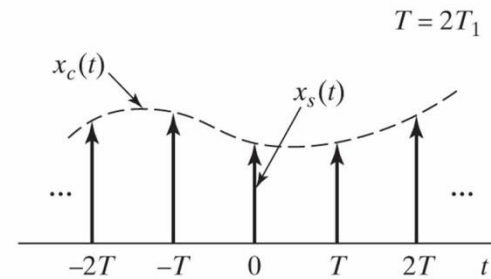
- $x_s(t)$ and $x[n]$ have the same information
 - Given $x_s(t)$, we can make $x[n]$.
 - Given $x[n]$, we can make $x_s(t)$.



(a)



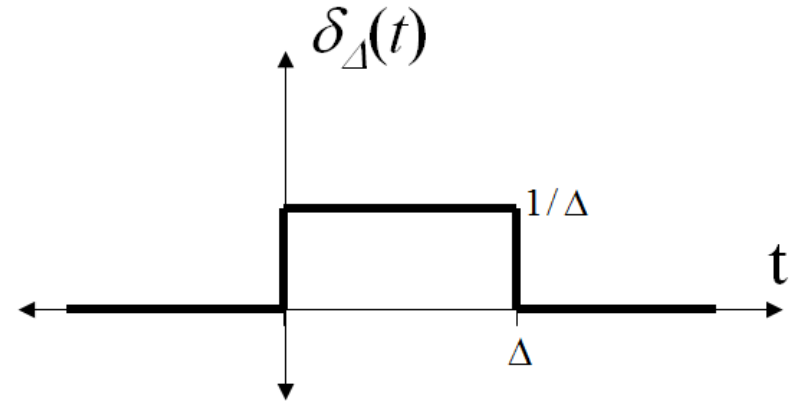
(b)



(c)

- Approximated unit impulse

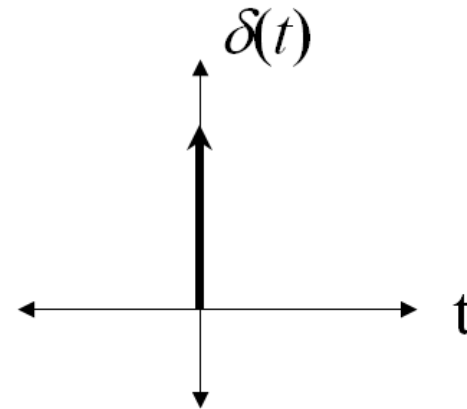
$$\delta_{\Delta}(t) = \frac{du_{\Delta}(t)}{dt} = \begin{cases} \frac{1}{\Delta}, & 0 \leq t < \Delta \\ 0, & \text{otherwise} \end{cases}$$



- Unit Impulse:

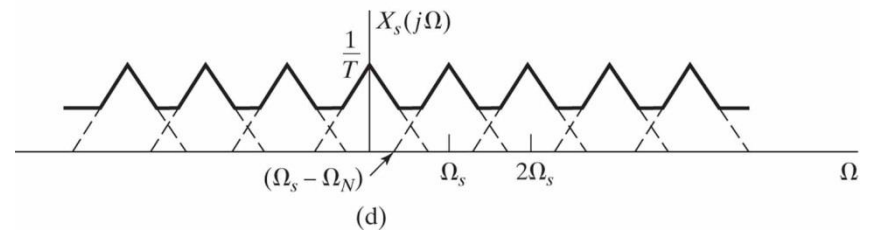
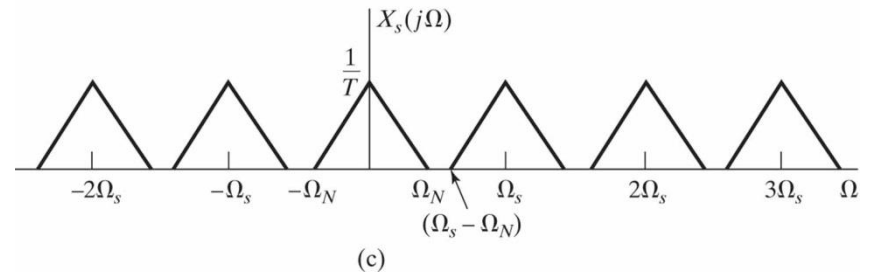
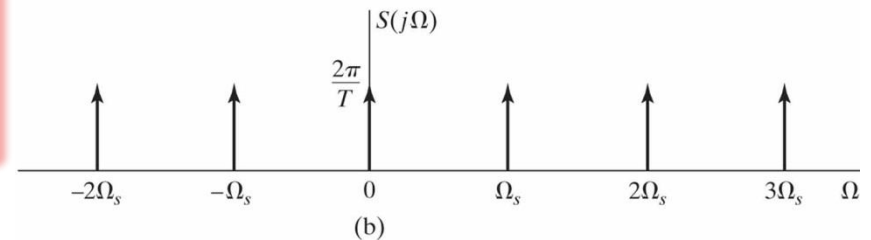
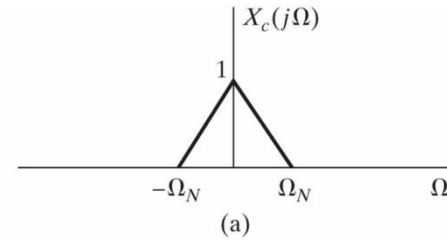
$$\delta(t) = \lim_{\Delta \rightarrow 0} \delta_{\Delta}(t) = \begin{cases} \infty, & t = 0 \\ 0, & t \neq 0 \end{cases}$$

$$\int_{-a}^b \delta(t) dt = 1 \quad \text{for any } a > 0 \text{ and } b > 0.$$



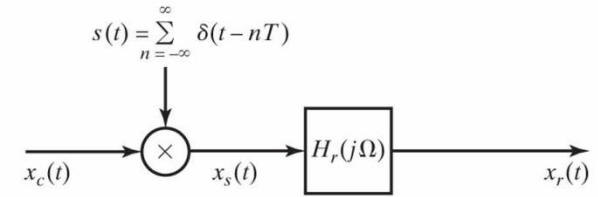
Frequency-Domain Representation of Sampling

- $$S(j\Omega) = \frac{2\pi}{T} \sum_{k=-\infty}^{\infty} \delta(\Omega - k\Omega_s)$$
- $$X_s(j\Omega) = \frac{1}{2\pi} X_c(j\Omega) * S(j\Omega)$$
$$= \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c(j(\Omega - k\Omega_s))$$

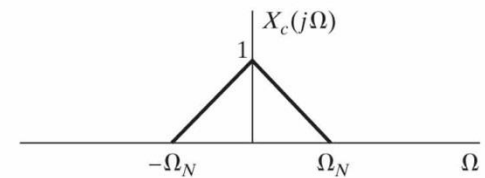


Recovery of $x_c(t)$ from $x_s(t)$

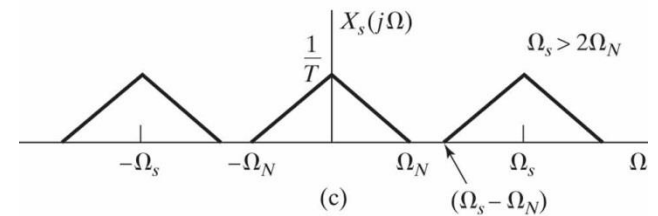
- If you can recover $x_c(t)$ from $x_s(t)$, you can recover $x_c(t)$ from $x[n]$.
- Recovery is possible through an ideal low-pass filter when $\Omega_s > 2\Omega_N$.



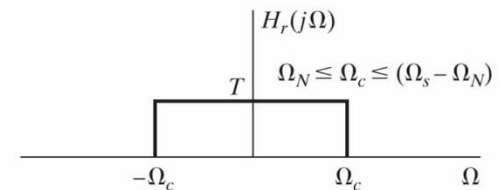
(a)



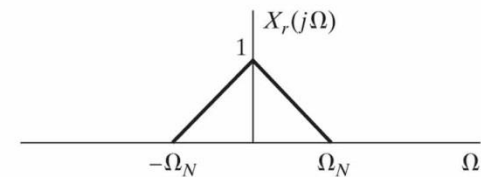
(b)



(c)



(d)



(e)

Nyquist-Shannon Sampling Theorem

Let $x_c(t)$ be a band-limited signal with

$$X_c(j\Omega) = 0 \text{ for } |\Omega| \geq \Omega_N.$$

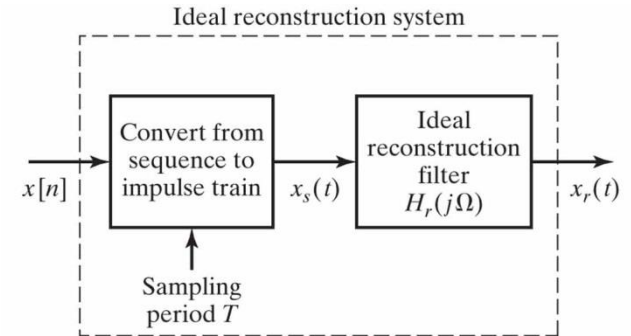
Then $x_c(t)$ is uniquely determined by its samples $x[n] = x_c(nT)$, $-\infty < n < \infty$, if

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

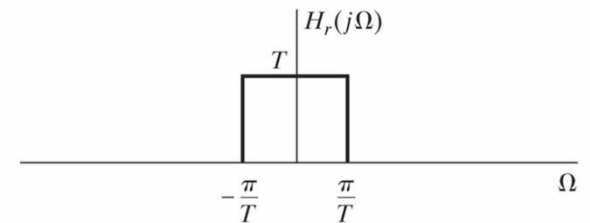
- $2\Omega_N$ is called the Nyquist rate.
- Under certain conditions, a CT signal can be completely represented by and recoverable from samples
- A low-pass signal can be reconstructed from samples, if the sampling rate is high enough. Because it is a low-pass signal, the change between two close samples is constrained (or expected).

Recovery of $x_c(t)$ from $x_s(t)$

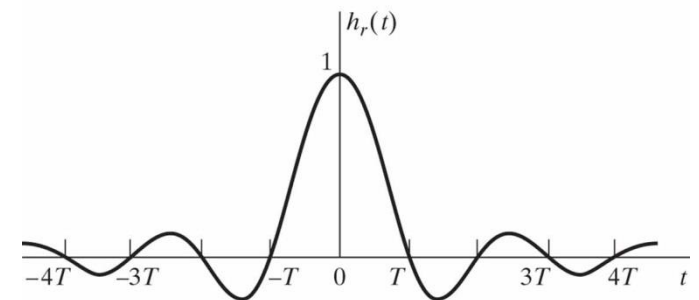
- $$h_r(t) = \frac{\sin(\frac{\pi t}{T})}{\frac{\pi t}{T}}$$
- $$x_s(t) = \sum_{n=-\infty}^{\infty} x[n] \delta(t - nT)$$
- $$x_r(t) = \sum_{n=-\infty}^{\infty} x[n] \frac{\sin(\frac{\pi(t-nT)}{T})}{\frac{\pi(t-nT)}{T}}$$



(a)



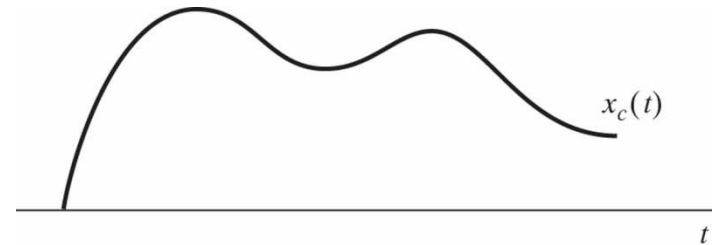
(b)



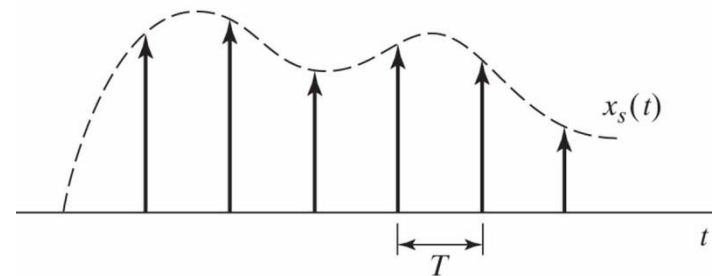
(c)

Recovery of $x_c(t)$ from $x[n]$

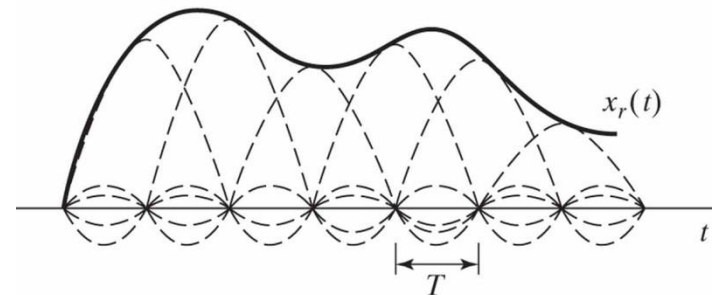
$$x_r(t) = \sum_{n=-\infty}^{\infty} x[n] \frac{\sin\left(\frac{\pi(t - nT)}{T}\right)}{\frac{\pi(t - nT)}{T}}$$



(a)



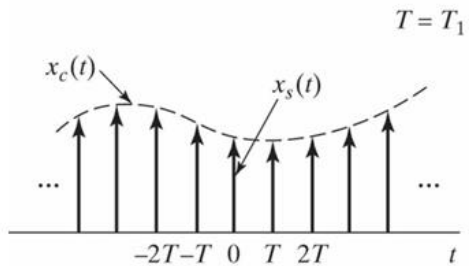
(b)



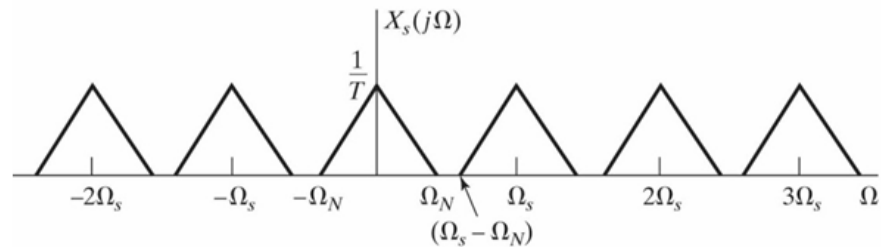
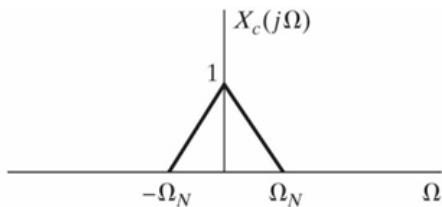
(c)

Summary

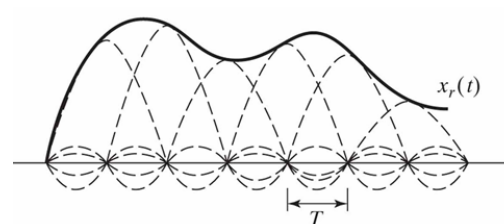
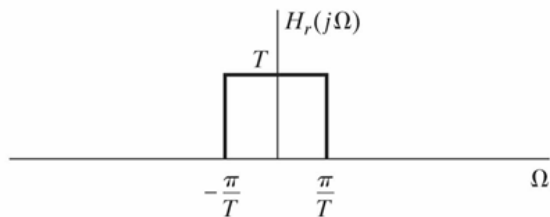
- $s(t) = \sum_{n=-\infty}^{\infty} \delta(t - nT), \quad x_s(t) = x_c(t)s(t)$



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



- $X_r(j\Omega) = X_S(j\Omega)H_r(j\Omega), \quad x_r(t) = \sum_{n=-\infty}^{\infty} x[n] \frac{\sin(\frac{\pi(t-nT)}{T})}{\frac{\pi(t-nT)}{T}}$



Frequency-Domain Relationship between $x[n]$ and $x_s(t)$

- Relationship between $X(e^{j\omega})$ and $X_s(j\Omega)$

$$X(e^{j\omega}) = X_s\left(j\frac{\omega}{T}\right)$$

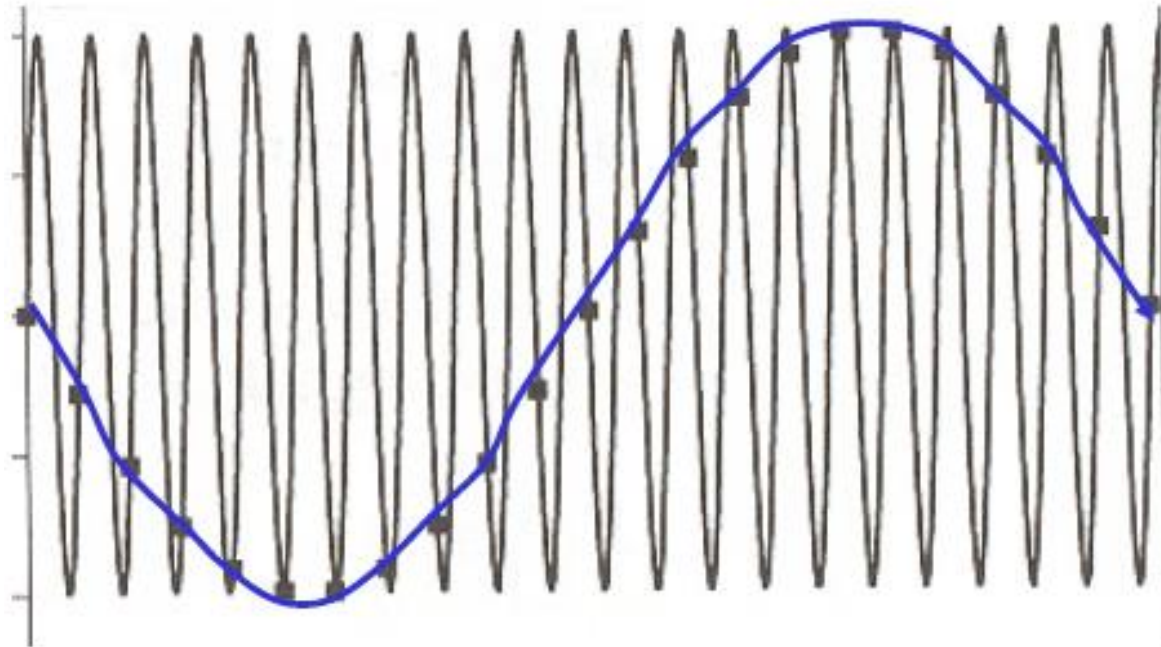
$$X_s(j\Omega) = X(e^{j\Omega T})$$

- Recall that $X(e^{j\omega})$ is always periodic

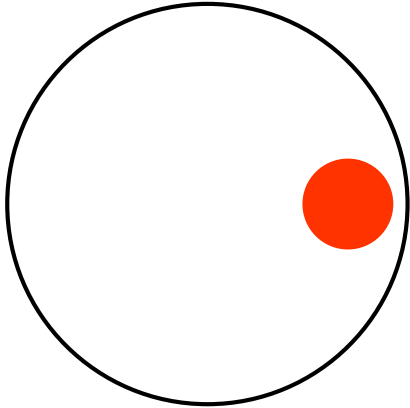
Aliasing

Undersampling Causes Aliasing

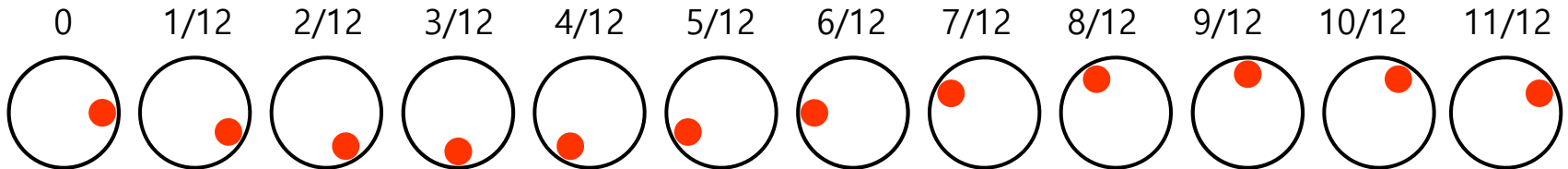
- Undersampling: sampling rate is less than Nyquist rate



Undersampling Causes Aliasing

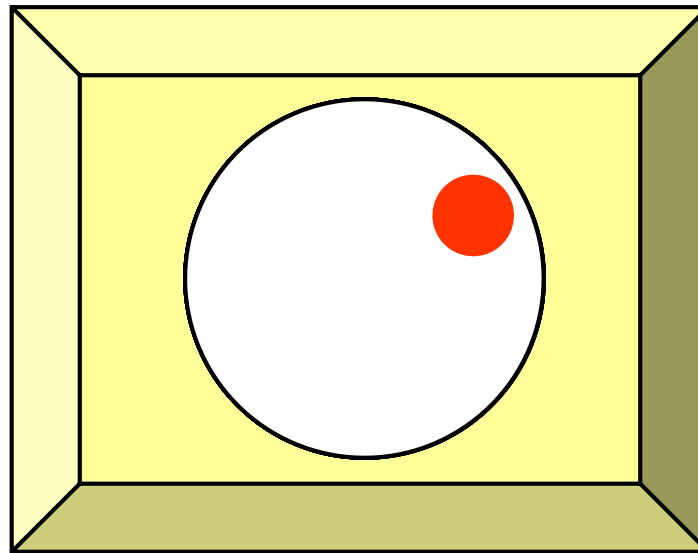
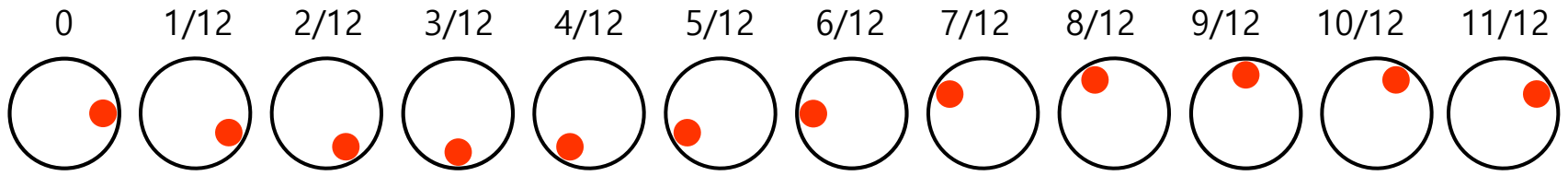


- Rotating disk
 - 1 rotation/second
- To avoid aliasing, it should be motion-pictured with at least 2 frames/s.



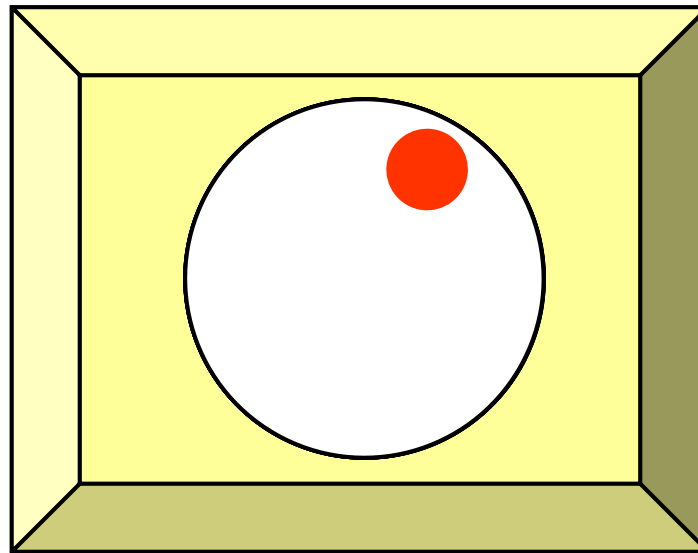
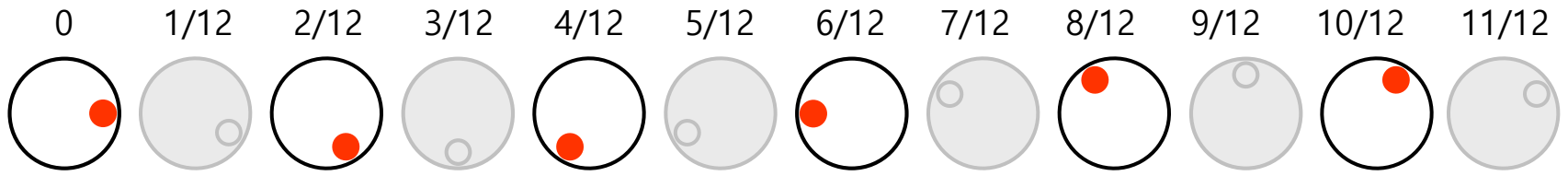
Undersampling Causes Aliasing

- 12 frames/s



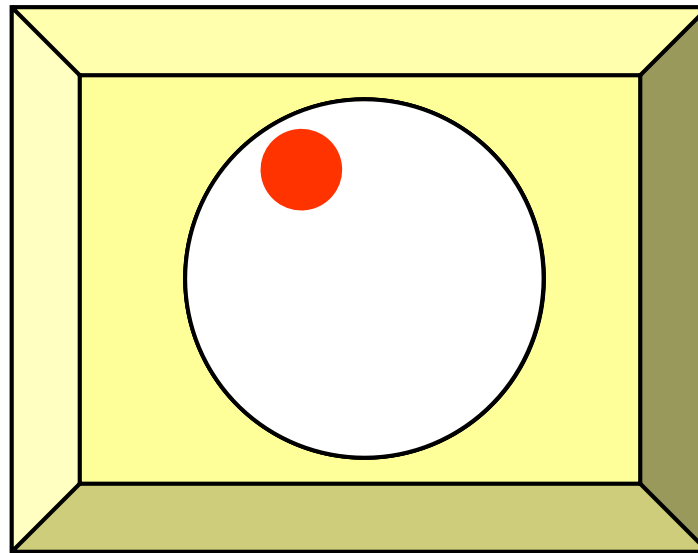
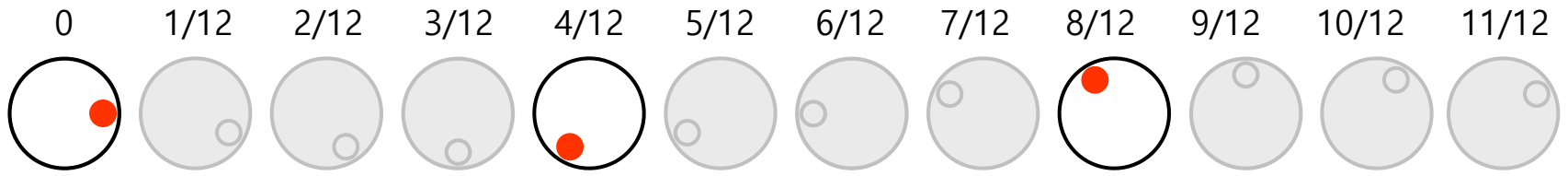
Undersampling Causes Aliasing

- 6 frames/s



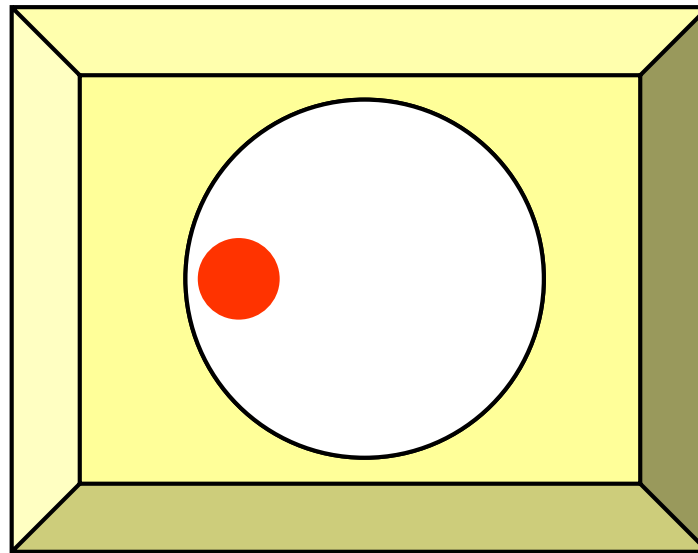
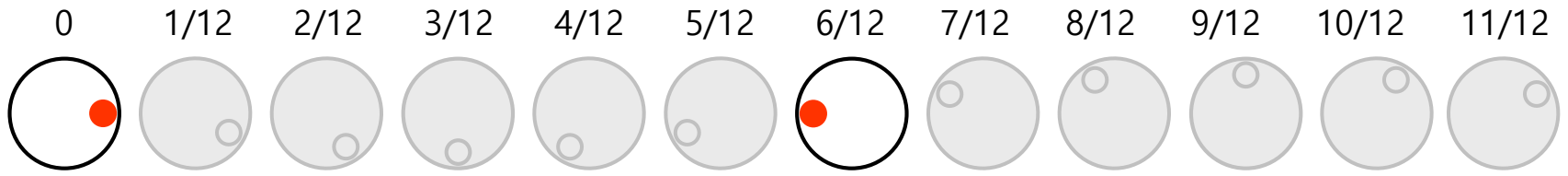
Undersampling Causes Aliasing

- 3 frames/s



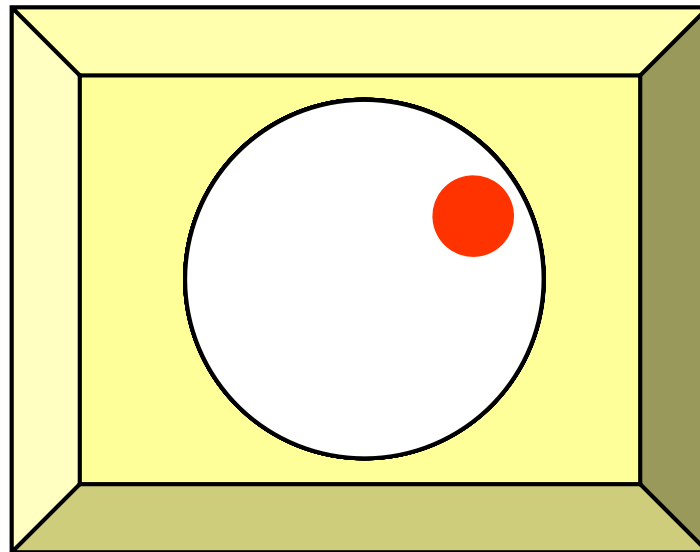
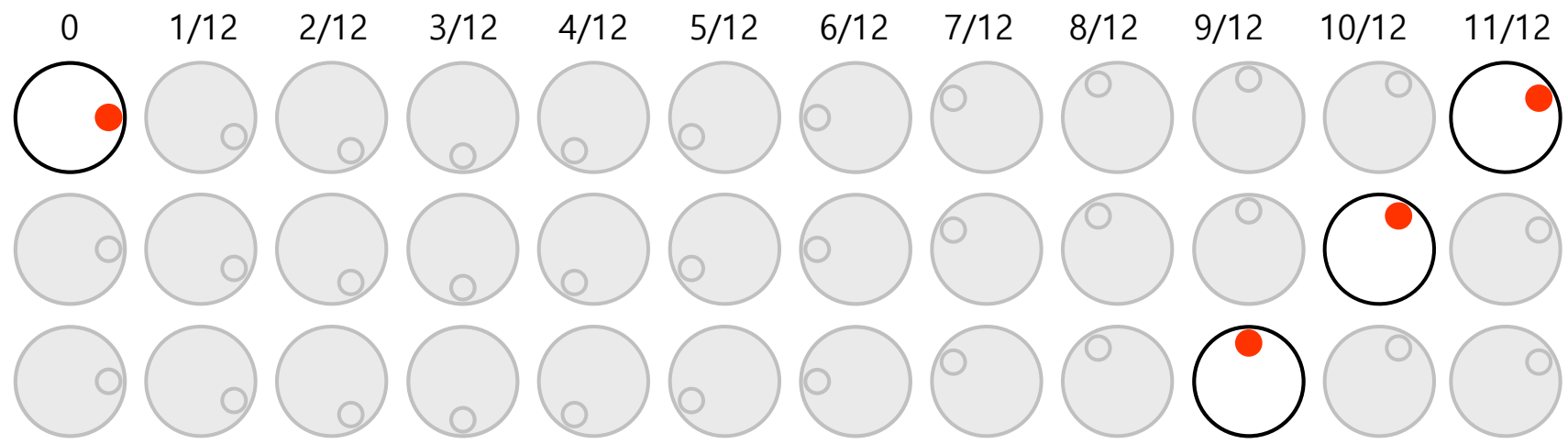
Undersampling Causes Aliasing

- 2 frames/s



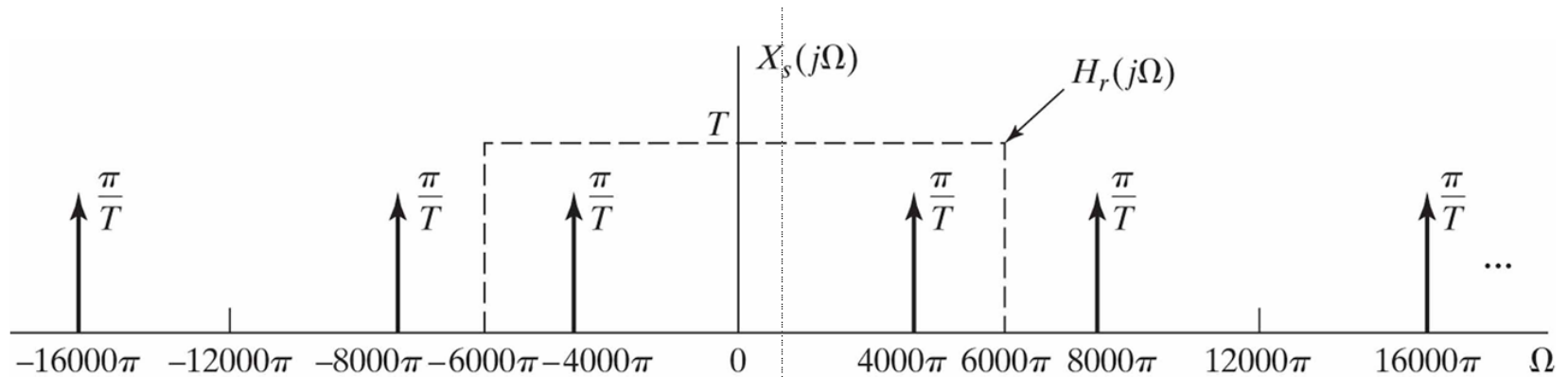
Undersampling Causes Aliasing

- $12/11 = 1.09$ frames/s



Examples

- $x_c(t) = \cos(4000\pi t)$, $T = 1/6000$.



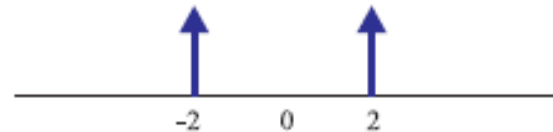
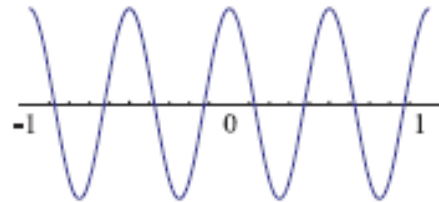
- $1 \leftrightarrow 2\pi\delta(\Omega)$
- $\cos(\Omega_0 t) \leftrightarrow \pi(\delta(\Omega - \Omega_0) + \delta(\Omega + \Omega_0))$
- $\sin(\Omega_0 t) \leftrightarrow \frac{\pi}{j}(\delta(\Omega - \Omega_0) - \delta(\Omega + \Omega_0))$

Examples

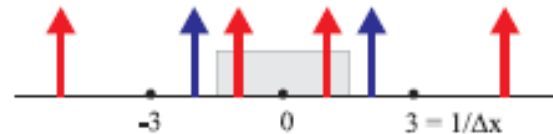
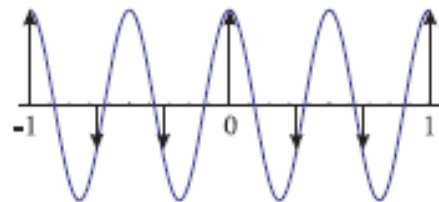
- $x_c(t) = \cos(16000\pi t)$, $T = 1/6000$.

Examples

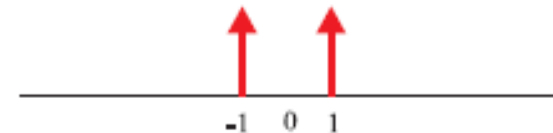
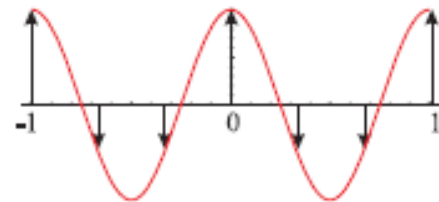
Original Signal: $f(x) = \cos 4\pi x$



Sampling: $\Delta x = 1/3$

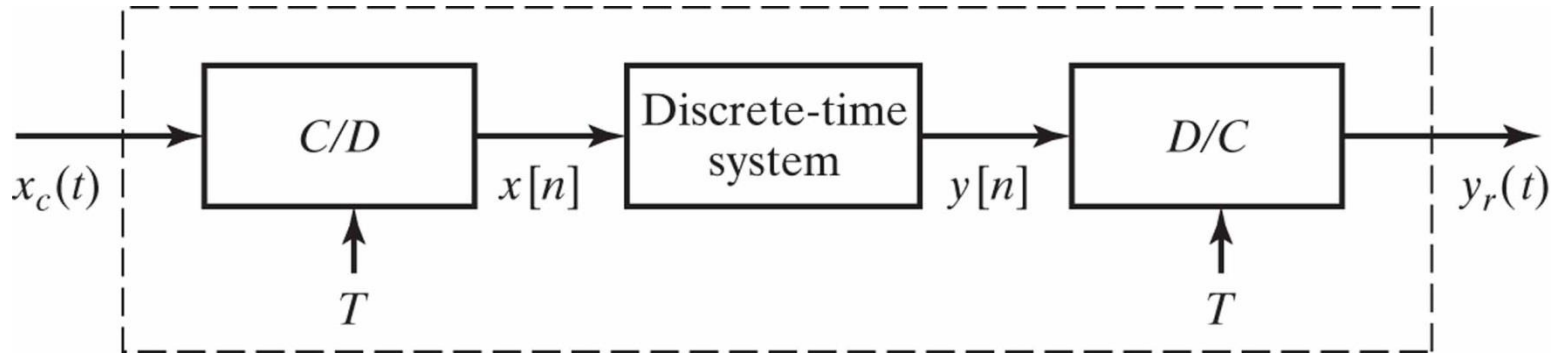


Reconstructed Signal: $f(x) = \cos 2\pi x$



DT Processing of CT Signals

C/D and D/C conversions



- C/D

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

- D/C

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

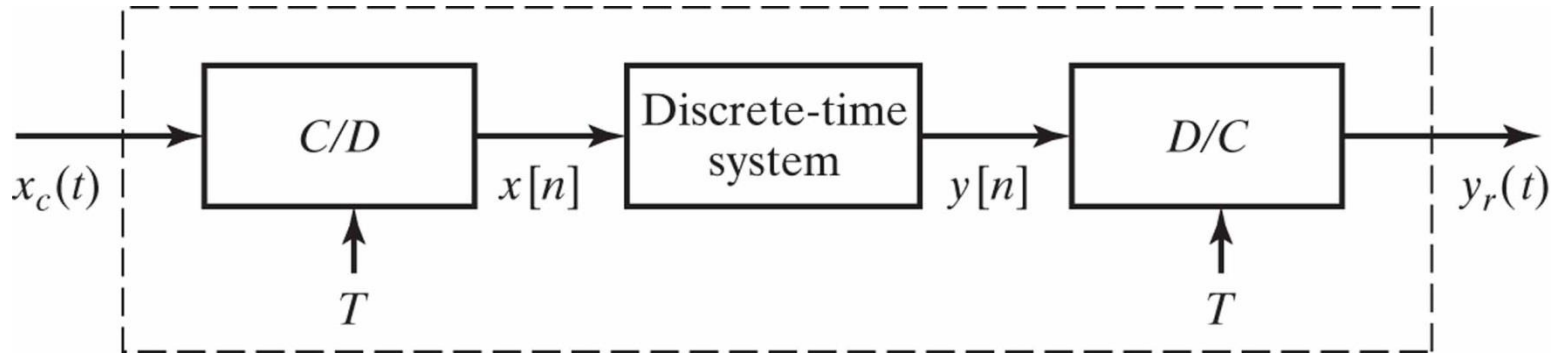
Relationship between

$x[n]$ and $x_s(t)$

$$X(e^{j\omega}) = X_s \left(j \frac{\omega}{T} \right),$$

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

Overall System



- Effective Frequency Response

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

- Assumptions

- $x_c(t)$ is band-limited
- $\frac{2\pi}{T}$ satisfies the Nyquist rate

Relationship between

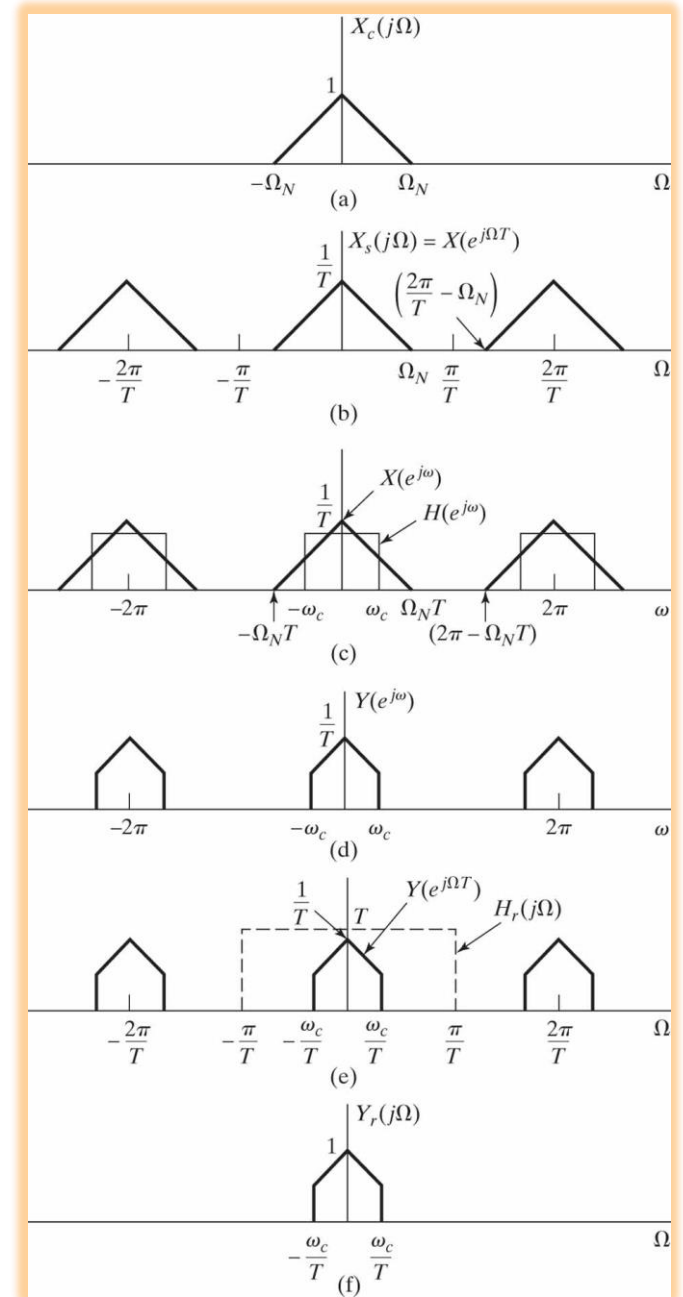
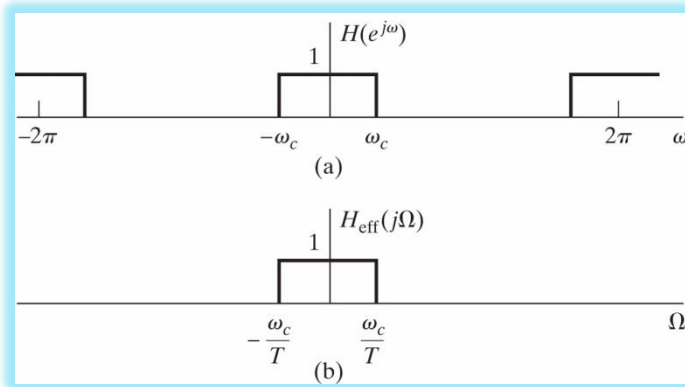
$x[n]$ and $x_s(t)$

$$X(e^{j\omega}) = X_s\left(j\frac{\omega}{T}\right),$$

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

Example 1

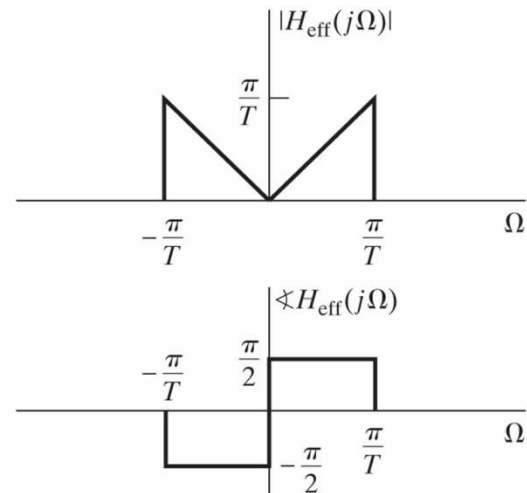
- $$H(e^{j\omega}) = \begin{cases} 1, & |\omega| < \omega_c \\ 0, & \omega_c < |\omega| < \pi \end{cases}$$



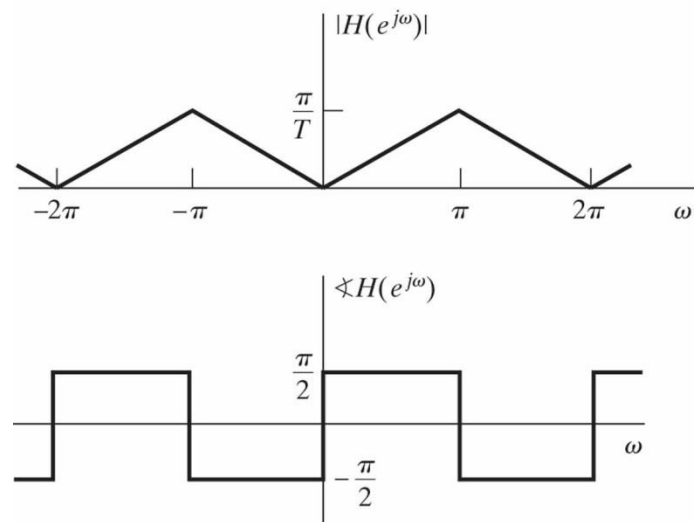
Example 2

- $$y_c(t) = \frac{d}{dt} x_c(t)$$

$$\Rightarrow h[n] = \begin{cases} 0, & n = 0 \\ \frac{(-1)^n}{nT}, & n \neq 0 \end{cases}$$

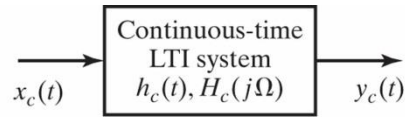


(a)

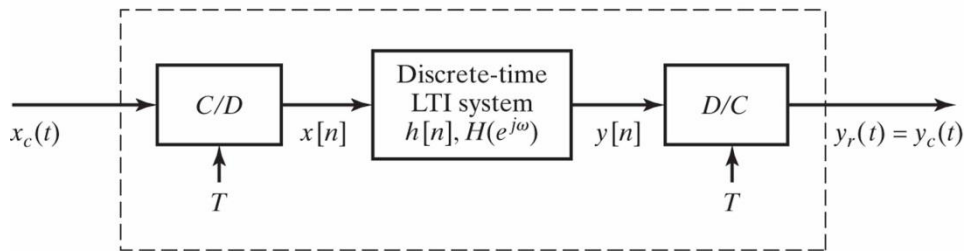


(b)

Impulse Invariance



(a)



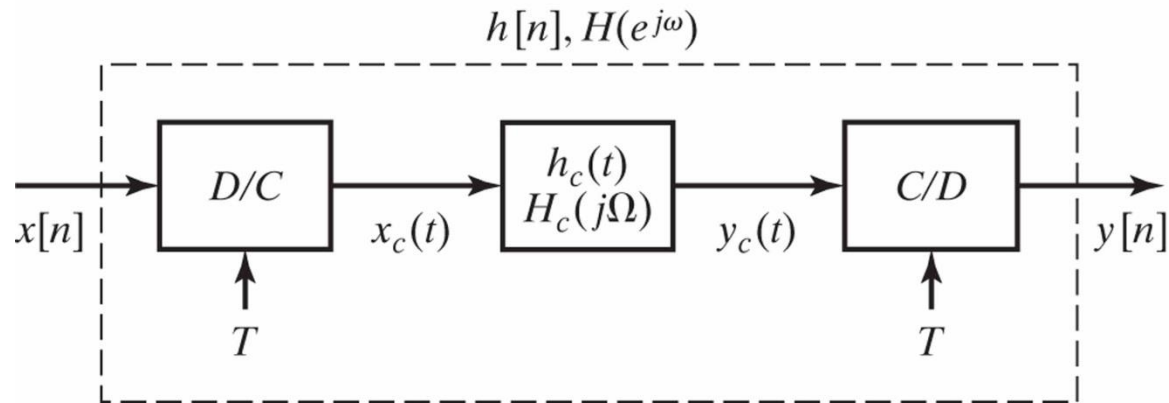
$$H_{\text{eff}}(j\Omega) = H_c(j\Omega)$$

(b)

- $h[n] = Th_c(nT)$
- Example
 - Ideal lowpass filter $h[n]$ with cutoff frequency ω_c

CT Processing of DT Signals

CT Processing of DT Signals



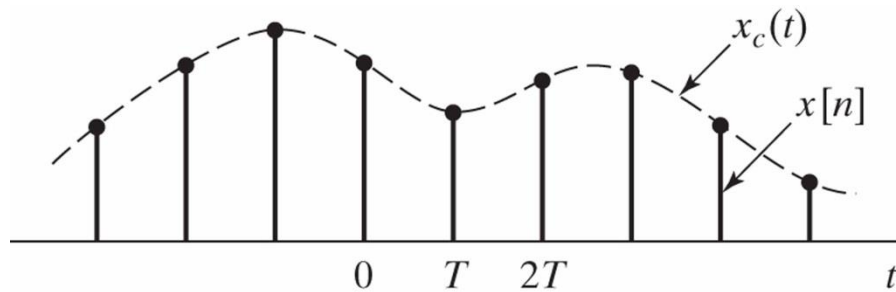
- It is rarely used, but provides a useful interpretation of some DT systems
- Main results

$$H(e^{j\omega}) = H_c\left(j\frac{\omega}{T}\right), \quad |\omega| < \pi$$

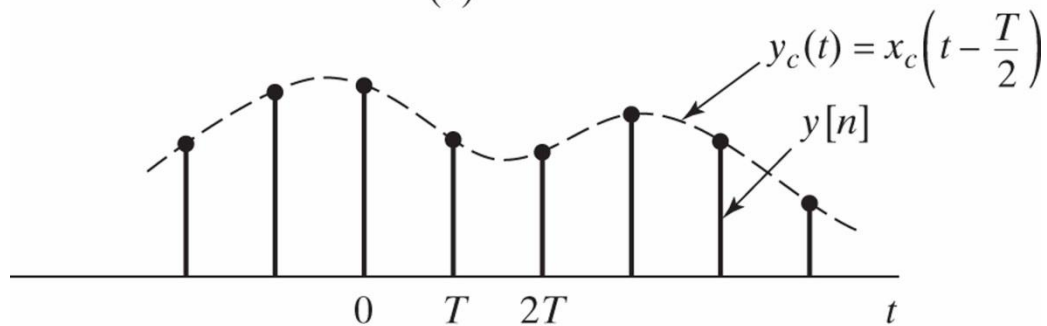
$$H_c(j\Omega) = H(e^{j\Omega T}), \quad |\Omega| < \frac{\pi}{T}$$

Example – Fractional Delay

- $H(e^{j\omega}) = e^{-j\omega\Delta}$, $|\omega| < \pi$
 $\Rightarrow h[n] = \frac{\sin \pi(n - \Delta)}{\pi(n - \Delta)}$



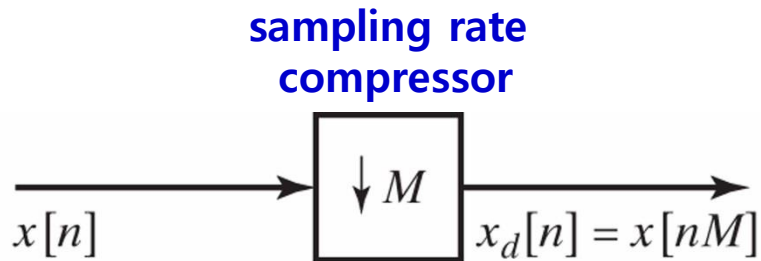
(a)



(b)

Changing Sampling Rate

Reducing Sampling Rate by an Integer Factor M



Sampling period T

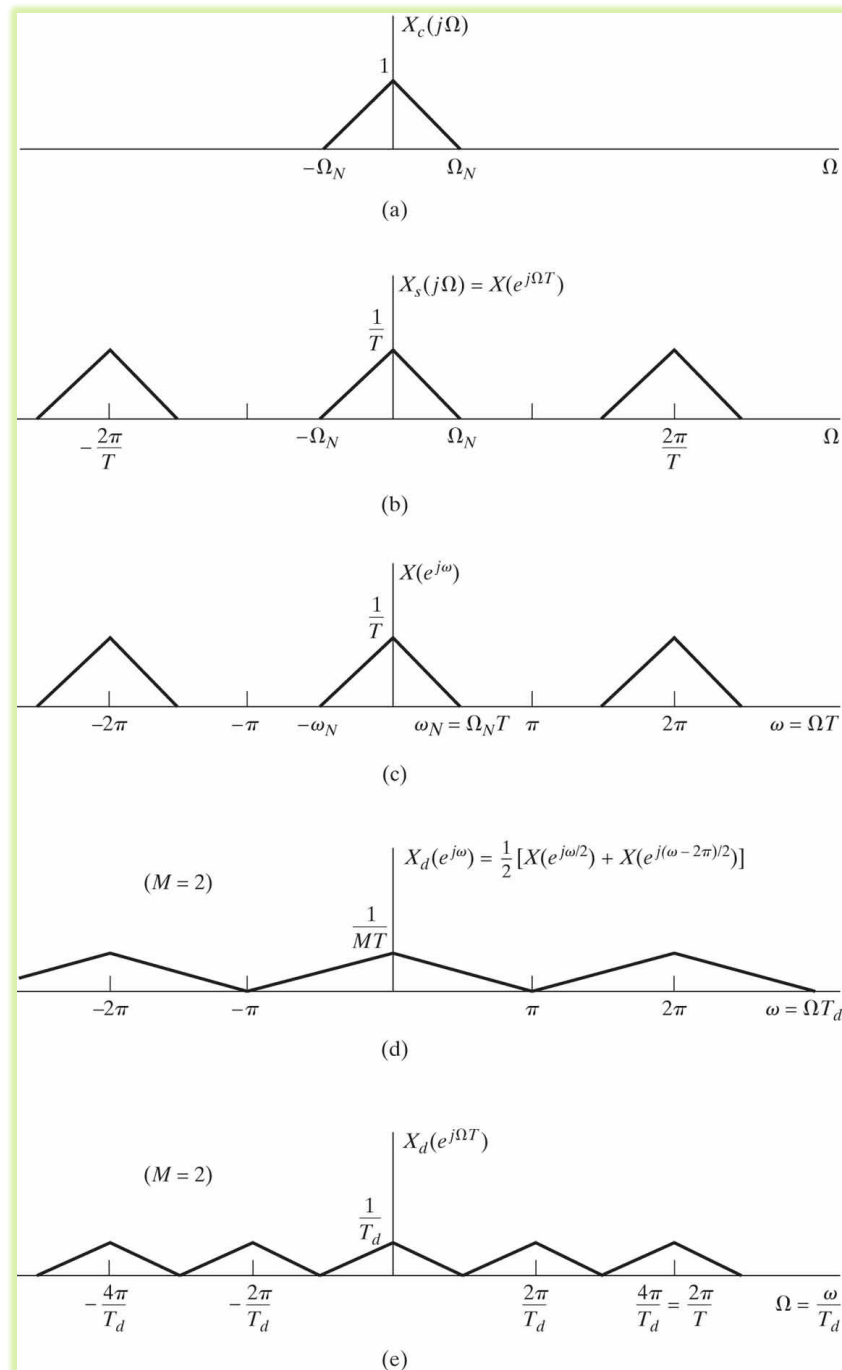
Sampling period $T_d = MT$

- Time domain

$$x_d[n] = x[nM]$$

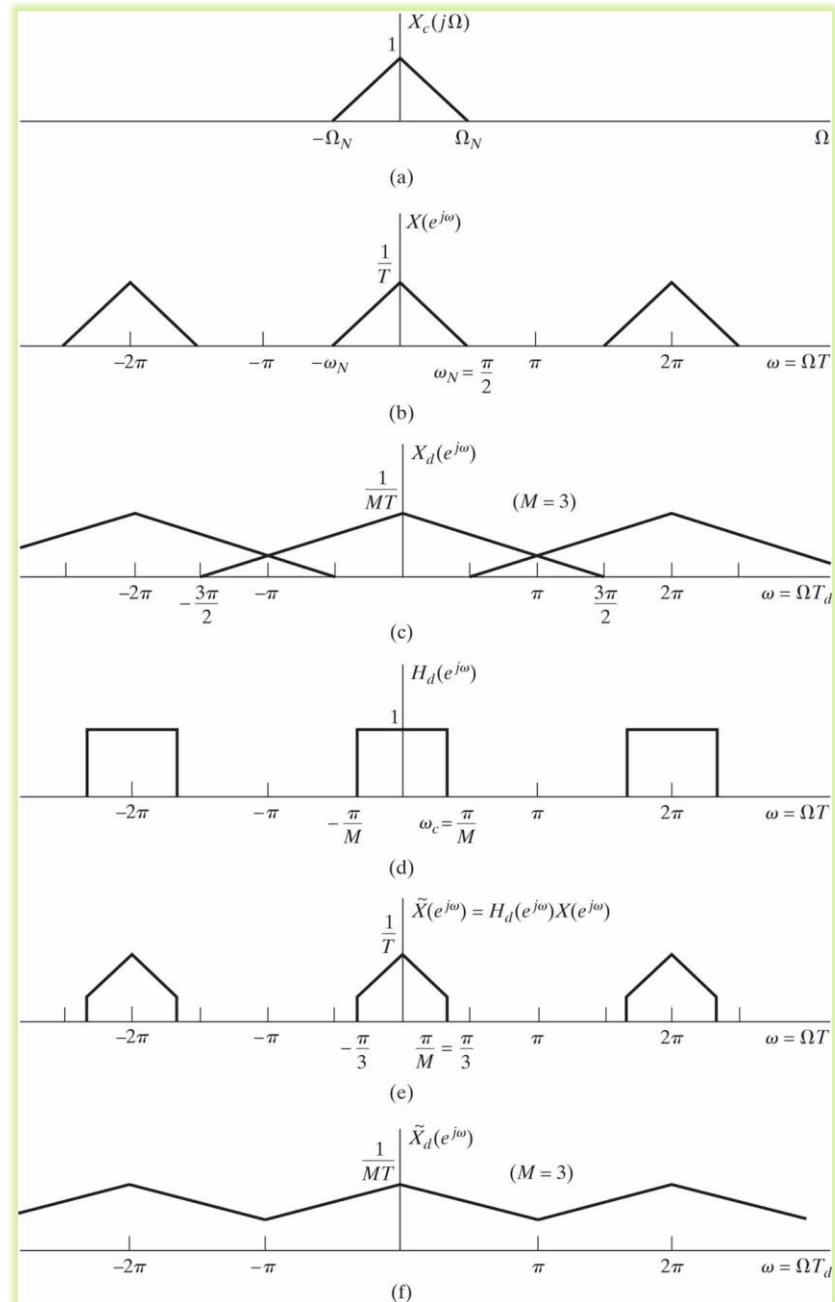
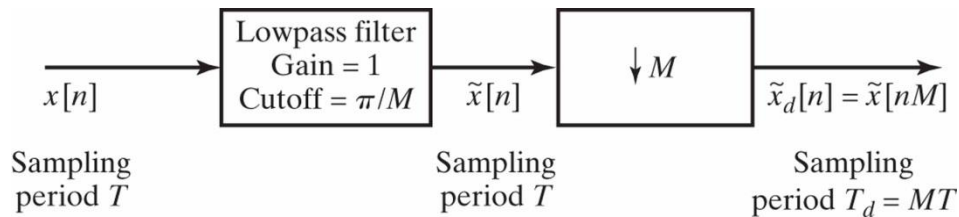
- Frequency domain

$$X_d(e^{j\omega}) = \frac{1}{M} \sum_{k=0}^{M-1} X(e^{j(\frac{\omega}{M} - \frac{2\pi k}{M})})$$



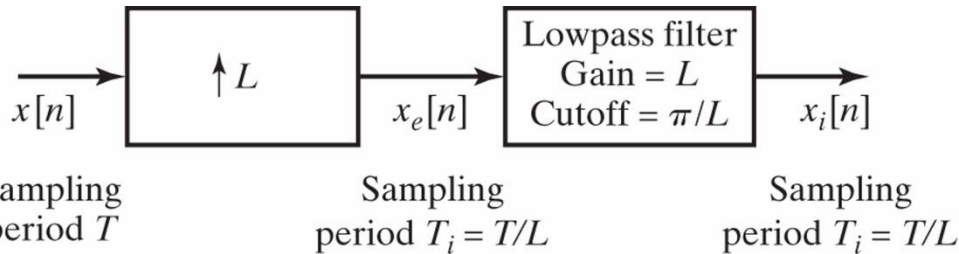
Reducing Sampling Rate by an Integer Factor M

- To avoid aliasing, we need
 - $X(e^{j\omega}) = 0$ if $\omega_N < |\omega| < \pi$
 - $\omega_N < \frac{\pi}{M}$
- Anti-aliasing filter can be used



Increasing Sampling Rate by an Integer Factor L

sampling rate expander



- Input and output

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

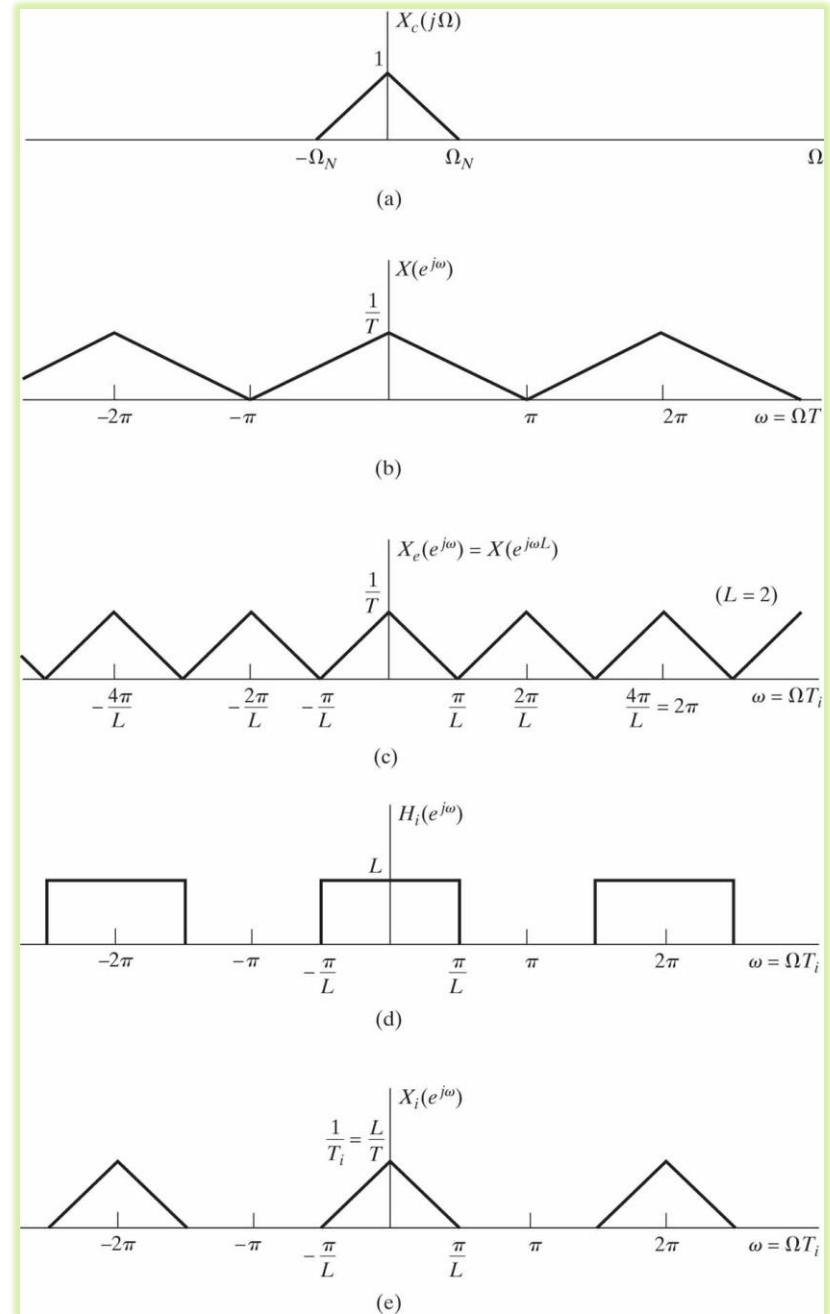
$$x_i[n] = x_c\left(n\frac{T}{L}\right)$$

- Intermediate signal

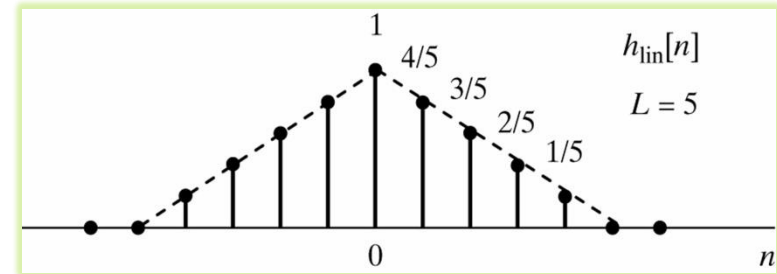
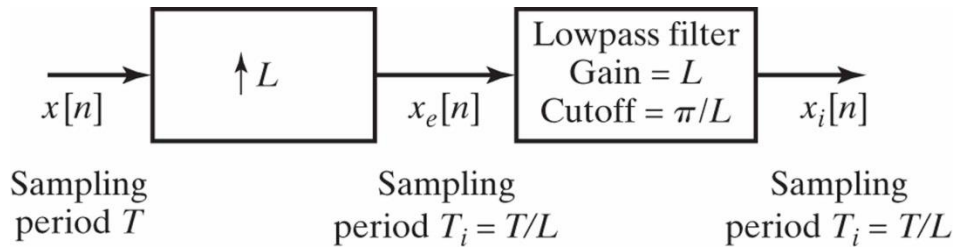
$$x_e[n] = \begin{cases} x\left[\frac{n}{L}\right] & \text{if } n \text{ is a multiple of } L \\ 0 & \text{otherwise} \end{cases}$$

- Output in terms of input

$$x_i[n] = \sum_{k=-\infty}^{\infty} x[k] \frac{\sin \frac{\pi(n-kL)}{L}}{\frac{\pi(n-kL)}{L}}$$



Ideal and Linear Interpolation Filters

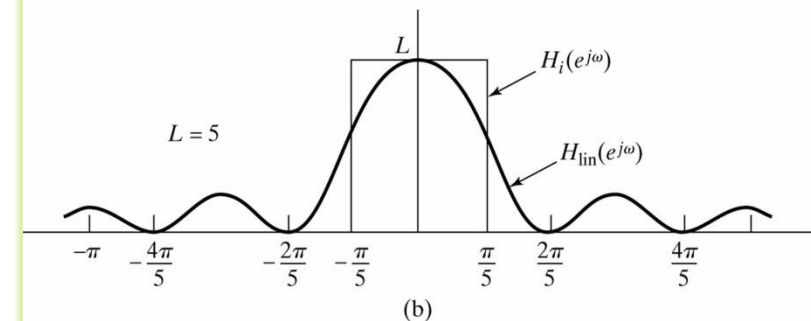
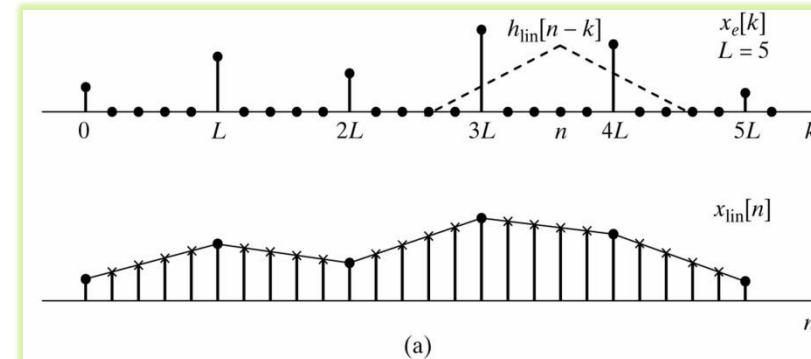


- $x_e[n] = \sum_{k=-\infty}^{\infty} x[k] \delta[n - kL]$
- $x_i[n] = x_e[n] * h_i[n] = \sum_{k=-\infty}^{\infty} x[k] h_i[n - kL]$
- Ideal filter

$$h_i[n] = \frac{\sin \frac{\pi n}{L}}{\frac{\pi n}{L}}$$

- Linear filter

$$h_{lin}[n] = \begin{cases} 1 - \frac{|n|}{L}, & -L \leq n \leq L \\ 0, & \text{otherwise} \end{cases}$$



Changing Sampling Rate by a Noninteger Factor

