

# Notes

**Assignments**  
**7.21, 7.23**

**Tutorial problems**  
**7.25, 7.37, 7.40**

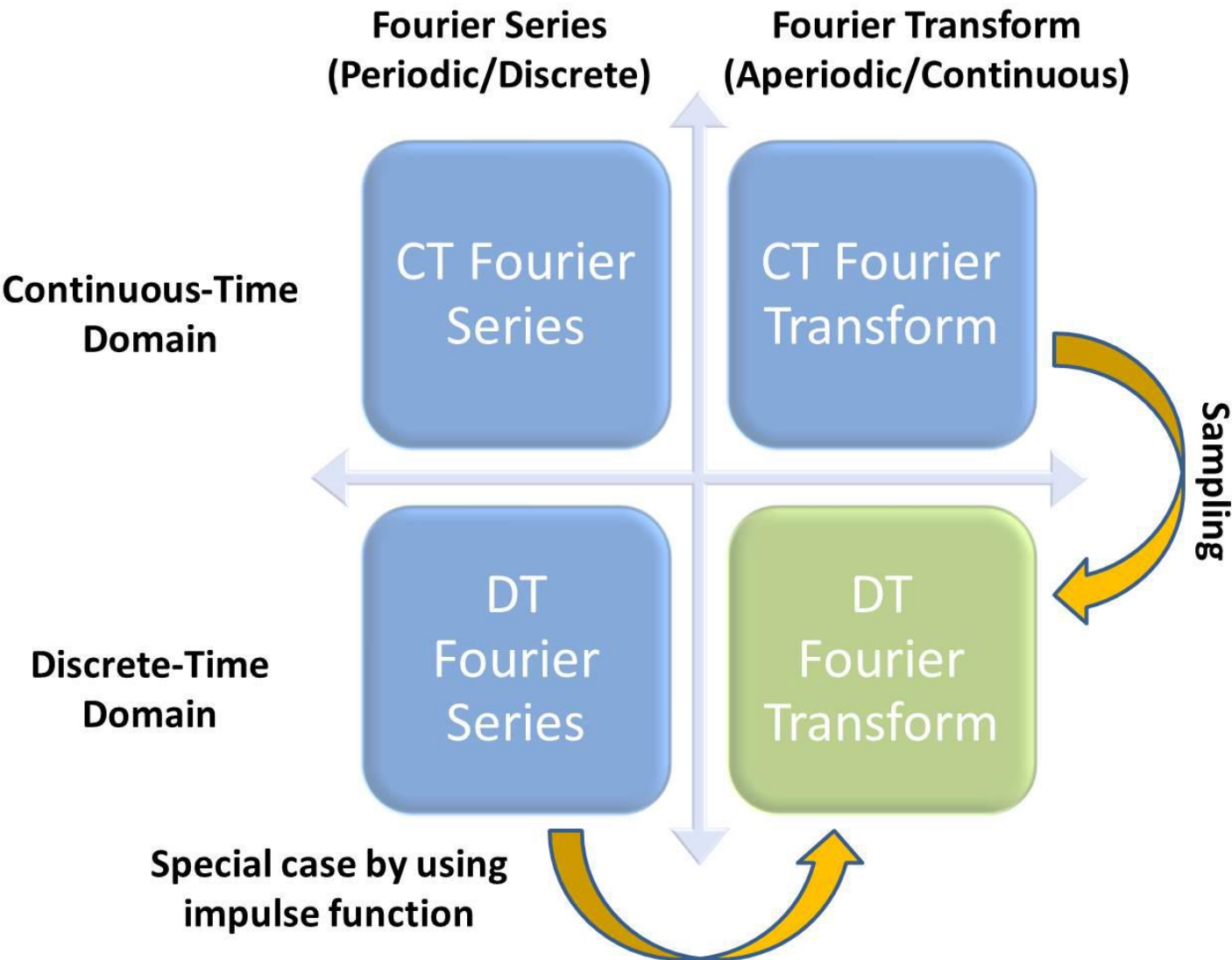


南方科技大学 SUSTC

# Chapter 7: Sampling

# Frequency domain

Time domain



# CTFT Properties: Multiplication Property

thus if

$$x(t) * y(t) \longleftrightarrow X(j\omega) \cdot Y(j\omega)$$

then the other way  
around is also true

$$\begin{aligned} x(t) \cdot y(t) &\longleftrightarrow \frac{1}{2\pi} X(j\omega) * Y(j\omega) \\ &= \frac{1}{2\pi} \int_{-\infty}^{+\infty} X(j\theta) Y(j(\omega - \theta)) d\theta \end{aligned}$$

— A consequence of *Duality*

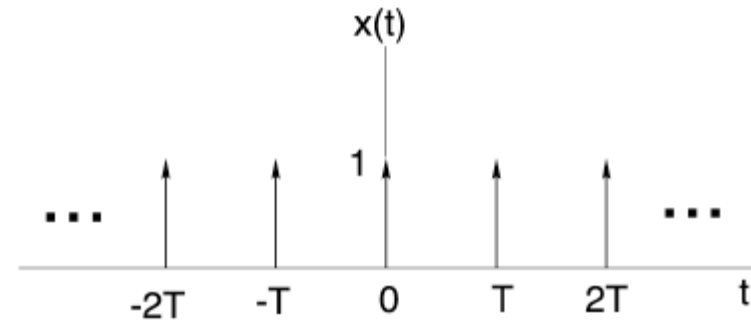
## Example 4.8

$$x(t) = \sum_{n=-\infty}^{+\infty} \delta(t - nT) \quad \text{— sampling function}$$

$$x(t) \xleftrightarrow{\text{FS}} a_k = \frac{1}{T} \int_{-T/2}^{T/2} x(t) e^{-jk\omega_o t} dt = \frac{1}{T}$$

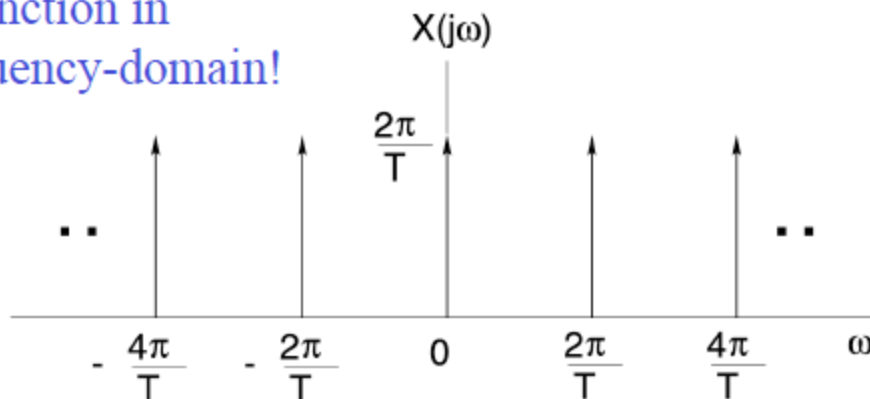
$$\Downarrow x(t) = \sum_{k=-\infty}^{+\infty} a_k e^{jk\omega_o t} = \sum_{k=-\infty}^{+\infty} \frac{1}{T} e^{jk\omega_o t}$$

$$X(j\omega) = \sum_{k=-\infty}^{+\infty} \frac{2\pi}{T} \delta\left(\omega - \underbrace{\frac{k2\pi}{T}}_{k\omega_o}\right)$$



$\omega_s = 2\pi / T$  : sampling frequency

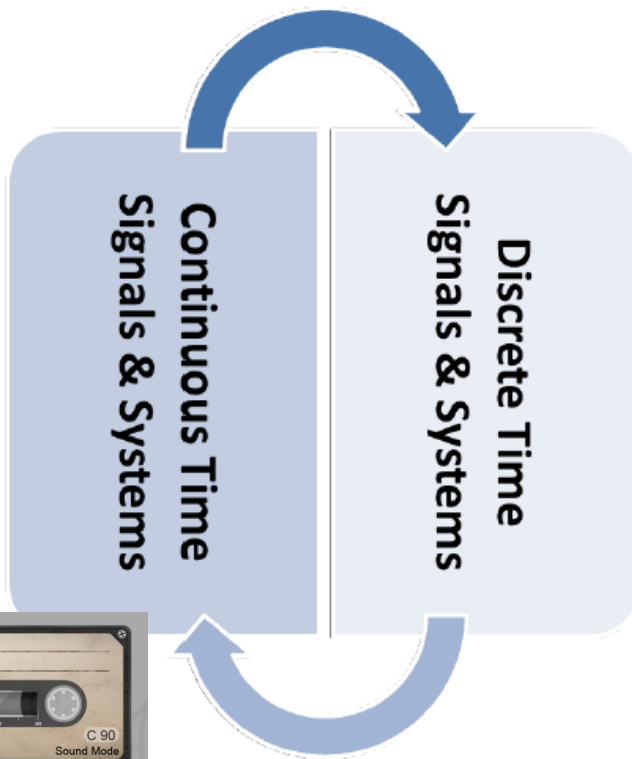
Same function in  
the frequency-domain!



Note in this case, periodic  
in both time domain (with  
a period  $T$ ) and frequency  
domain (with a period  
 $2\pi/T$ )

# Introduction

**Sampling:** to facilitate digital processing via computers or chips



**Any lossless conversion?**

**Process CT signals with DT systems?**



**Interpolation:** to present the output of digital processing

# Example: Video recording

- **Signal to be sampled:** real scene (continuous-time signals)
- **Sampling:** record by camera with a rate of 24, 25 or 30 frames per second
- **Sampled signal:** video tapes, mp4 files, avi files and etc. (discrete-time signals)
- **Reconstruction:** watch via eyes and interpret in the brain
- In our consciousness, the real scene can be reconstructed without information loss



# Outline

- Sampling is a general procedure to generate **DT signals** from **CT signals**, where information of the original signals can be kept
- **Core sampling theory:**
  - ◆ Impulse train, zero-order hold, 1st-order hold, etc.
  - ◆ Analysis in frequency domain
  - ◆ Nyquist rate
- **Undersampling:** Aliasing
- **Application:** process continuous-time signals discretely
- **More sampling techniques:** decimation, downsampling and upsampling

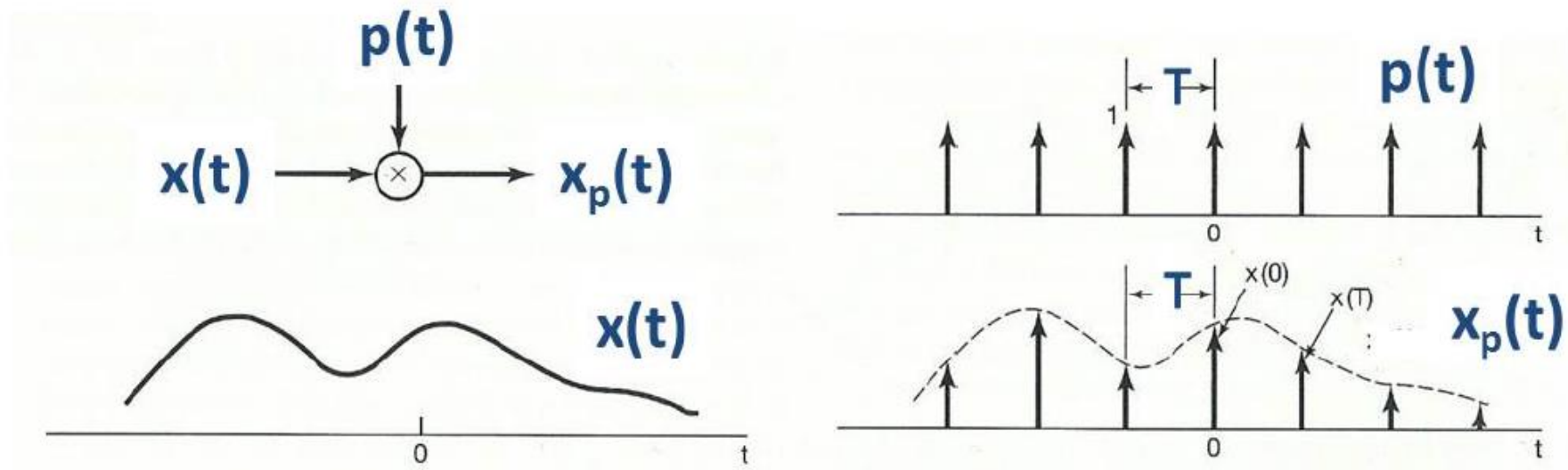


## Note on digital signal

- Need be both discrete time (DT) and digital value
  - ◆ By sampling: sampling rate
  - ◆ By quantization: how many bits
  - ◆ Could be implemented by analogy-to-digital (A/D) converter

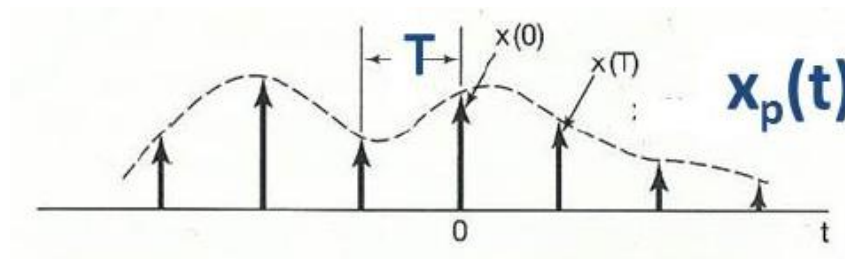
# Impulse-train sampling

- Mathematically, sampling can be represented by multiplication



- Sampling function:  $p(t) = \sum_{n=-\infty}^{\infty} \delta(t - nT)$
- Sampling period:  $T$
- Sampling:

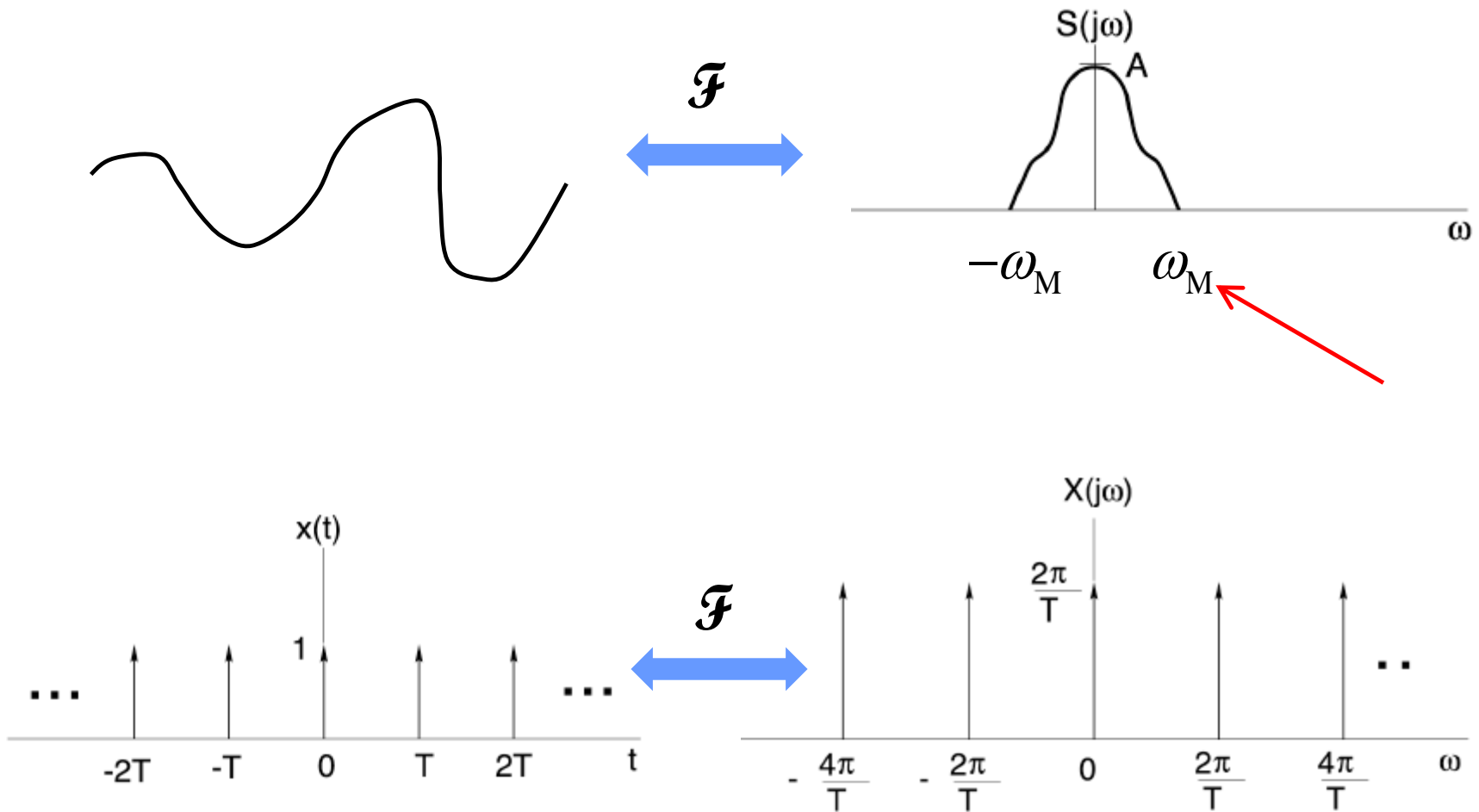
$$x_p(t) = x(t) \times p(t) = x(t) \times \sum_{n=-\infty}^{\infty} \delta(t - nT) = \sum_{n=-\infty}^{\infty} x(nT) \delta(t - nT)$$



Sampling discards most of points in the original signals.

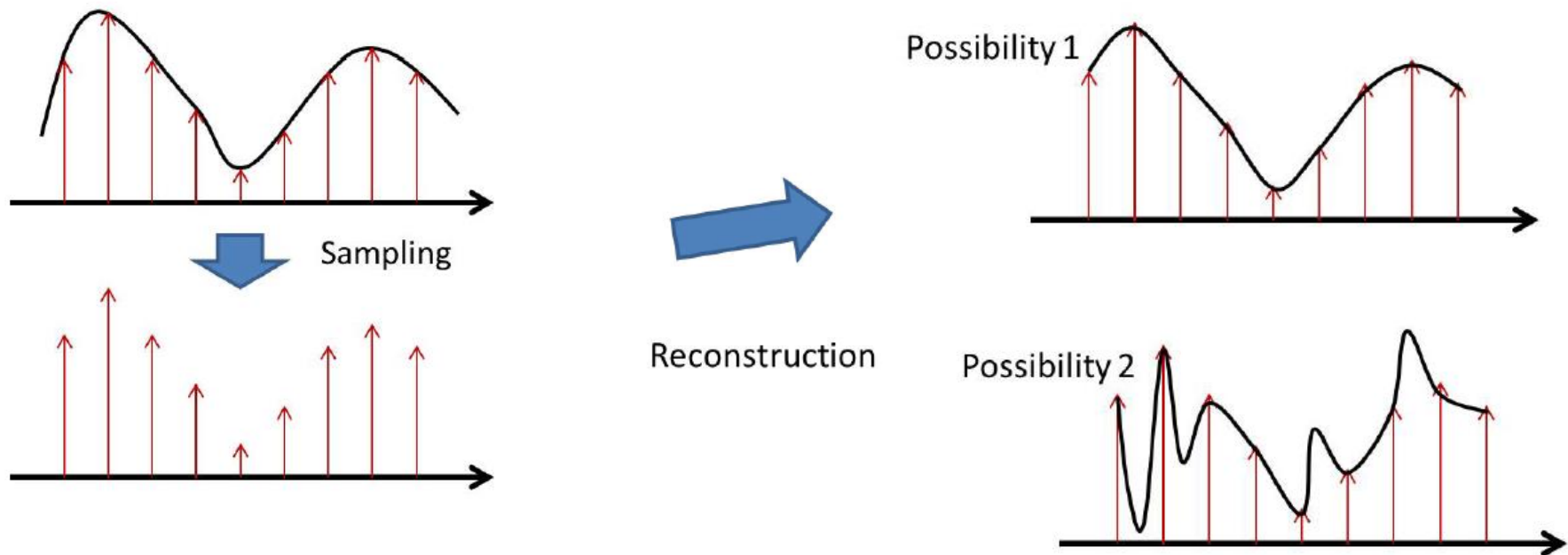
Is there any information loss in sampling?  
Or can we perfectly reconstruct the original signal?

# Two important frequencies



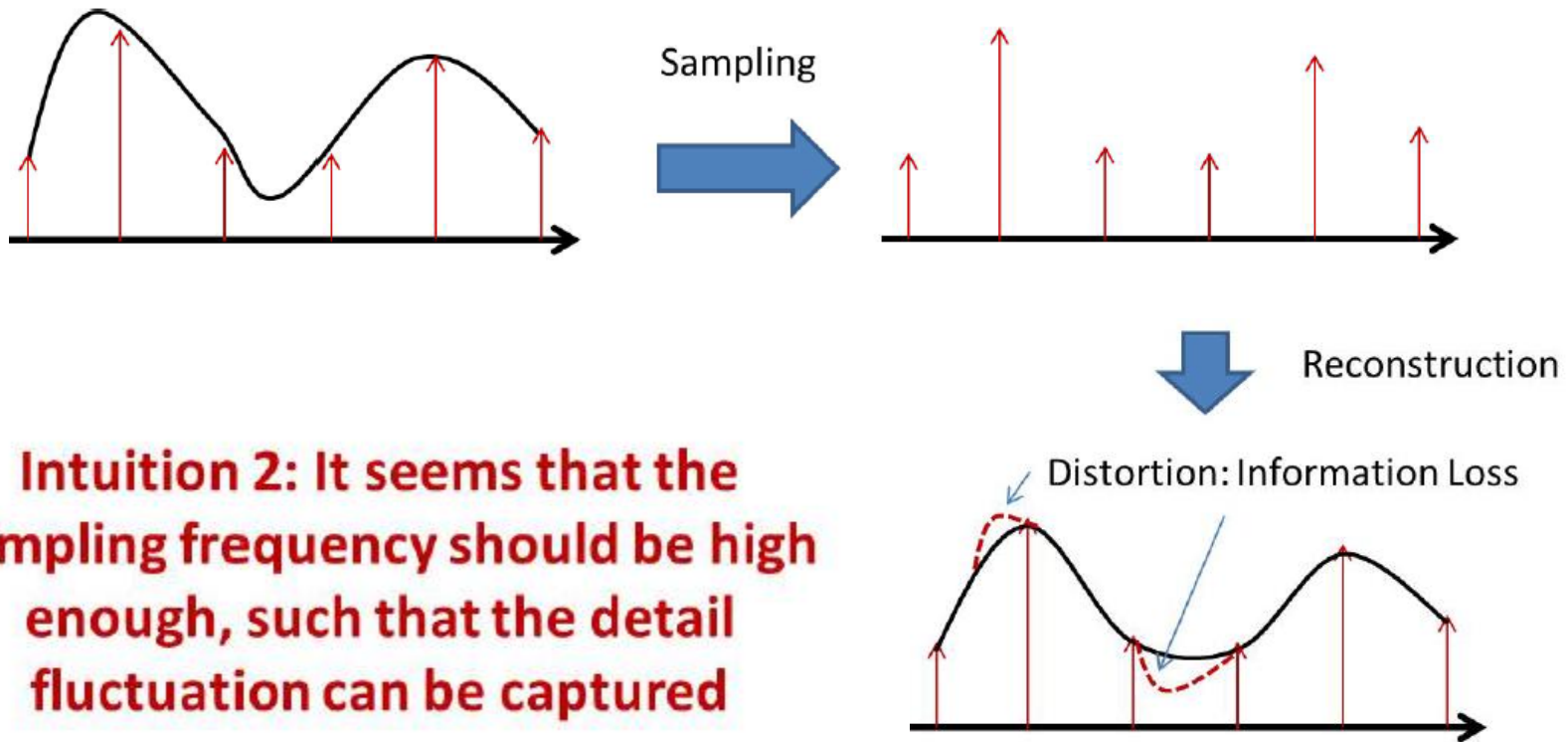
$$\omega_s = \frac{2\pi}{T}$$

# Observation (1/2)



**Intuition 1: It seems that we need a smooth interpolation**

## Observation (2/2)



**Intuition 2: It seems that the sampling frequency should be high enough, such that the detail fluctuation can be captured**

- Sampling: the frequency should be high enough
- Reconstruction: the interpolation should be smooth enough

# Frequency analysis (1/2)

- Theoretical tool: continuous-time Fourier transform
- Principle:

$$x(t) \times p(t) \Leftrightarrow \frac{1}{2\pi} X(j\omega) * P(j\omega)$$

- Fourier series of  $p(t)$ :

$$\begin{aligned} a_k &= \frac{1}{T} \int_{-T/2}^{T/2} \sum_{n=-\infty}^{\infty} \delta(t - nT) e^{-jk\omega_s t} dt \quad \text{where} \quad \omega_s = \frac{2\pi}{T} \\ &= \frac{1}{T} \int_{-T/2}^{T/2} \delta(t) e^{-jk\omega_s t} dt = \frac{1}{T} \end{aligned}$$

- Fourier Transform of  $p(t)$ :

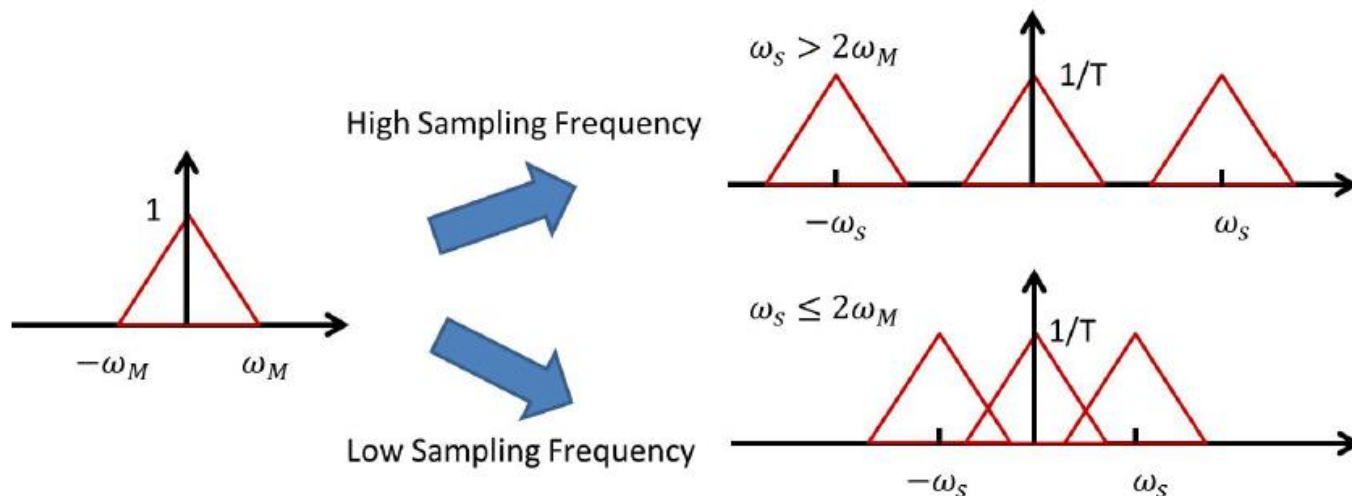
$$P(j\omega) = 2\pi a_k \sum_{k=-\infty}^{+\infty} \delta(\omega - k\omega_s) = \frac{2\pi}{T} \sum_{k=-\infty}^{+\infty} \delta(\omega - k\omega_s)$$

## Frequency analysis (2/2)

- Fourier transform of sampled signal  $x_p(t)$ :

$$\begin{aligned} X_p(j\omega) &= \frac{1}{2\pi} \int_{-\infty}^{\infty} X(j\theta) P(j(\omega - \theta)) d\theta \\ &= \frac{1}{T} \sum_{k=-\infty}^{\infty} X(j(\omega - k\omega_s)) \end{aligned}$$

- Sampling: the Fourier transform of input signal is repeated with period  $\omega_s$

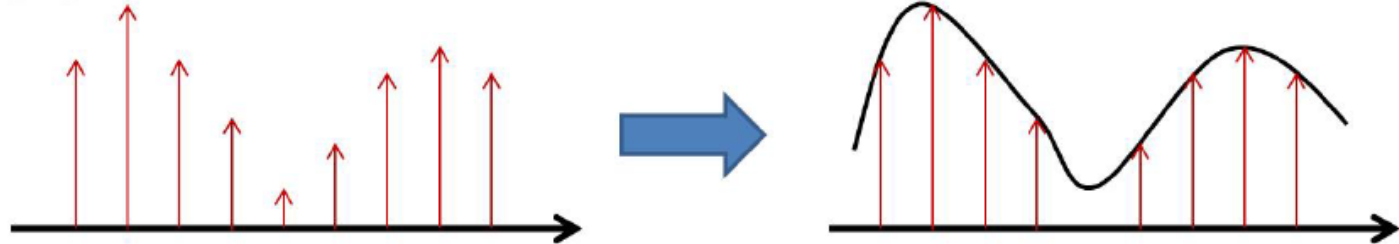




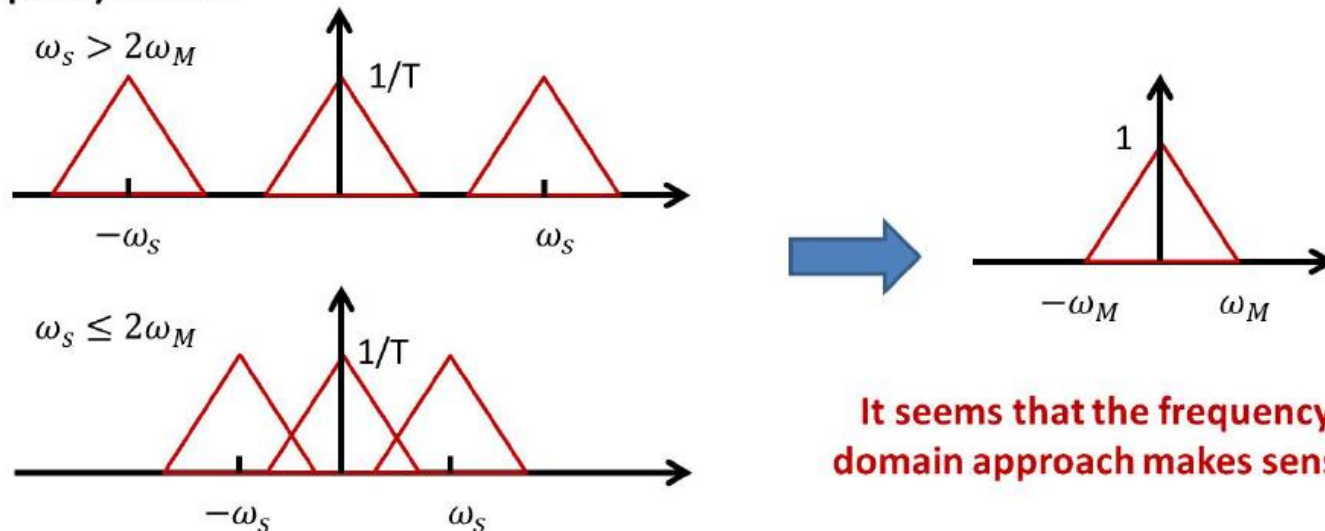
# Reconstruction problem

- Given the sampled signal, can we perfectly reconstruct the signal before sampling?

Time Domain



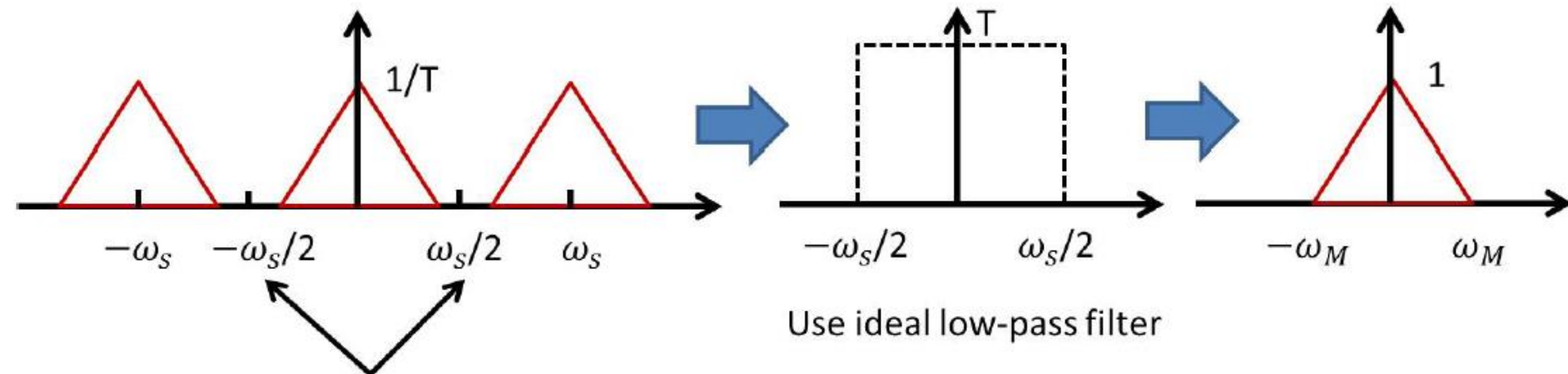
Frequency Domain



It seems that the frequency domain approach makes sense

# Reconstruction (1/2)

- Scenario of  $\omega_s > 2\omega_M$

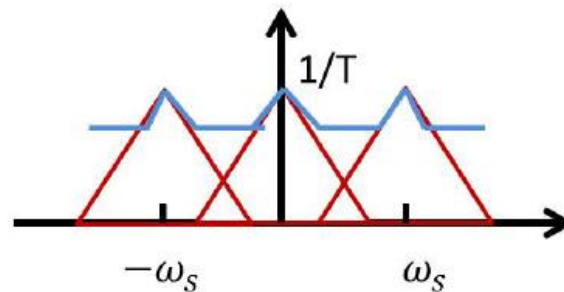


Use ideal low-pass filter

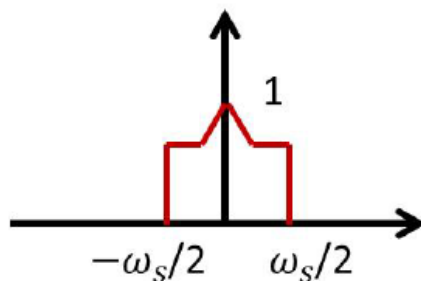
The spectrum of desired signal is within  
 $\left(-\frac{\omega_s}{2}, \frac{\omega_s}{2}\right) \rightarrow$  No overlapping

# Reconstruction (2/2)

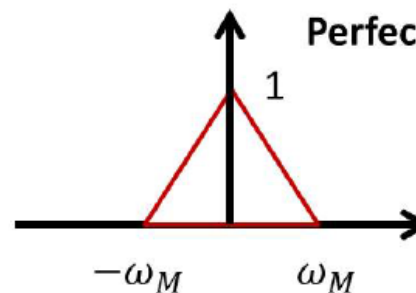
- Scenario of  $\omega_s \leq 2\omega_M$



Since  $\omega_s \leq 2\omega_M$ , we don't know the frequency range of the desired signal  
Sampling on the following signals can generate the same result:



OR



Perfect reconstruction is impossible

**Observation: the original signal  $x(t)$  can be Uniquely and perfectly reconstructed from  $x(nT)$  only when  $\omega_s > 2\omega_M$**

# Sampling theorem

## Sampling Theorem

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

$$X(j\omega) = 0 \text{ for } |\omega| > \omega_M.$$

Then,  $x(t)$  is uniquely determined by its samples  $x(nT)$  or  $x_p(t)$  if

$$\omega_s = \frac{2\pi}{T} > 2\omega_M,$$

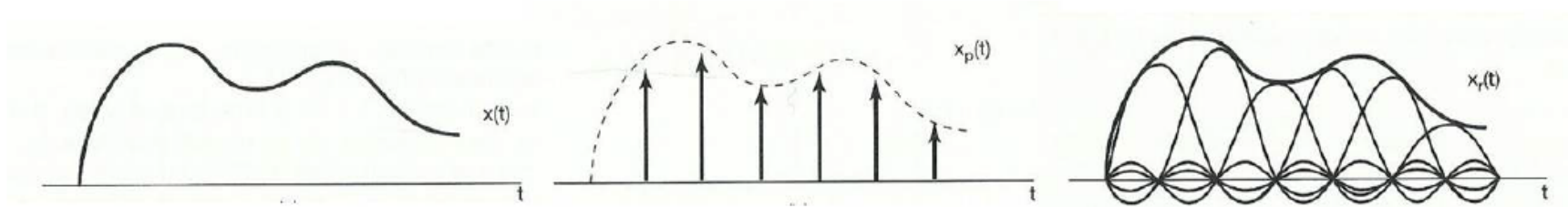
where  $2\omega_M$  is referred to as the *Nyquist rate*.

- Questions:

- ▶ How about  $\omega_s = 2\omega_M$ ?
- ▶ Sampling on band-pass signals

# Signal reconstruction: Interpolation

- If  $\omega_s > 2\omega_M$ , original signal can be perfectly reconstructed by ideal low-pass filter.
- Time domain interpretation of lowpass filtering



$$\begin{aligned}
 x_r(t) &= x_p(t) * h(t) = \sum_{n=-\infty}^{+\infty} x(nT)h(t - nT) \\
 &= \sum_{n=-\infty}^{+\infty} x(nT) \frac{\sin \frac{\omega_s}{2}(t - nT)}{\frac{\omega_s}{2}(t - nT)} = \sum_{n=-\infty}^{+\infty} x(nT) \text{sinc}\left(\frac{t - nT}{T}\right)
 \end{aligned}$$

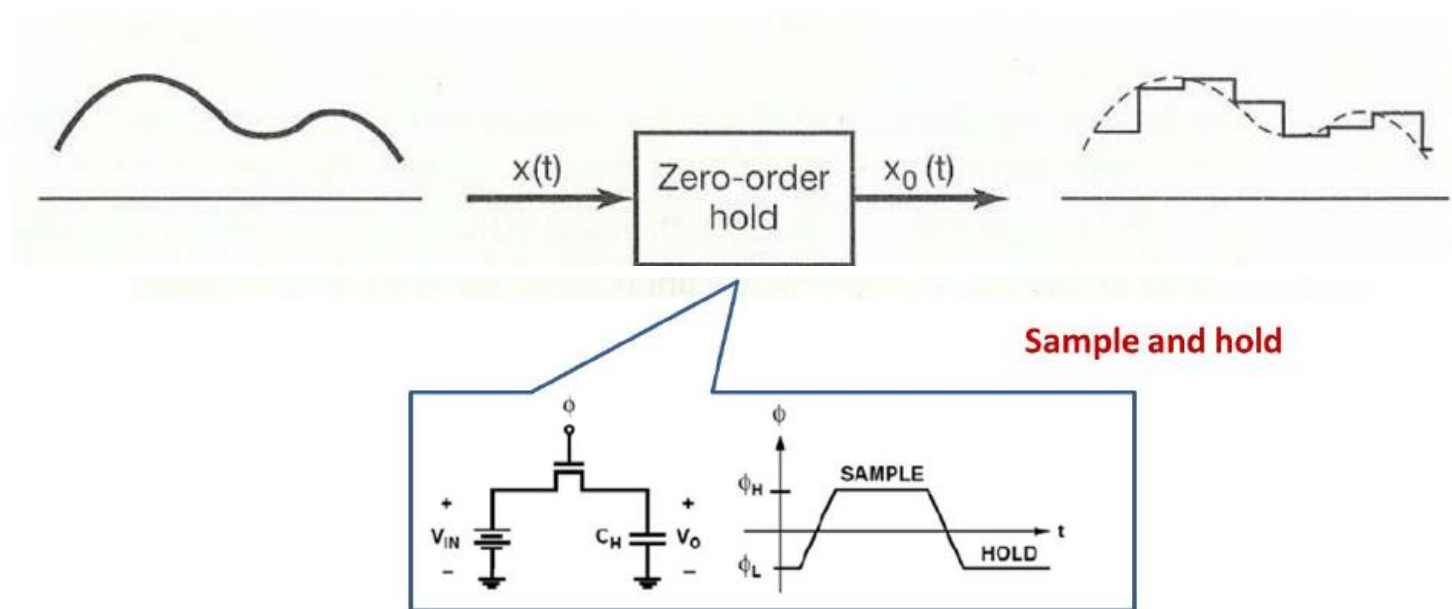
- Ideal lowpass filtering: interpolation with sinc function

# Zero-order hold

- It's difficult to generate ideal impulse chain in practical implementation.

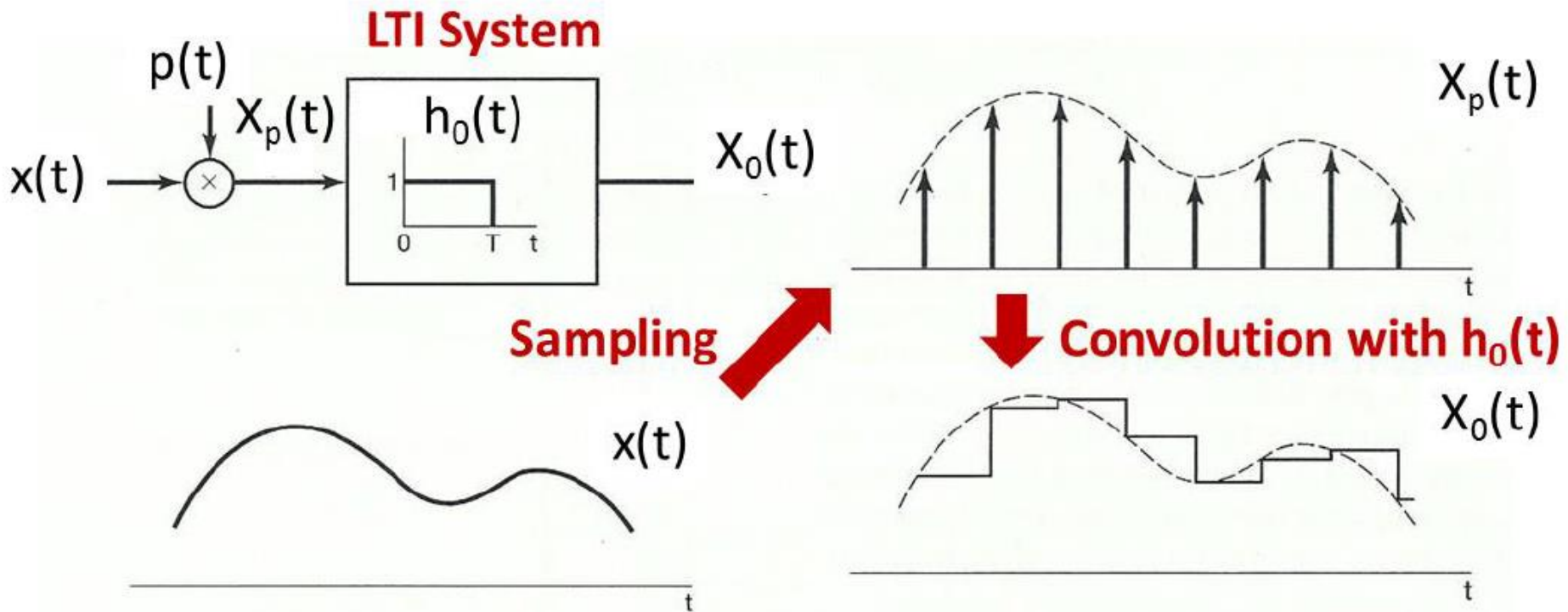
$$x_p(t) = \sum_{n=-\infty}^{\infty} x(nT)\delta(t - nT)$$

- Alternative approach: zero-order hold



- How to interpret the system of "zero-order hold" mathematically?

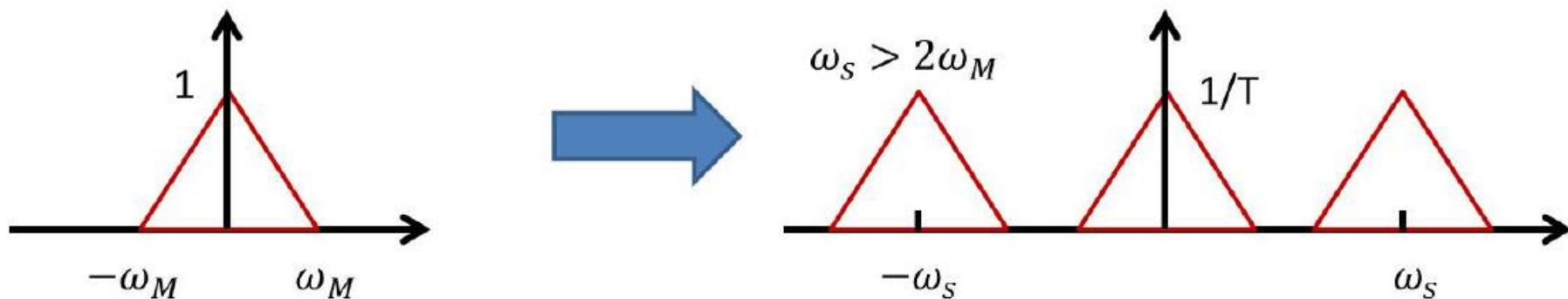
# Interpretation of zero-order hold



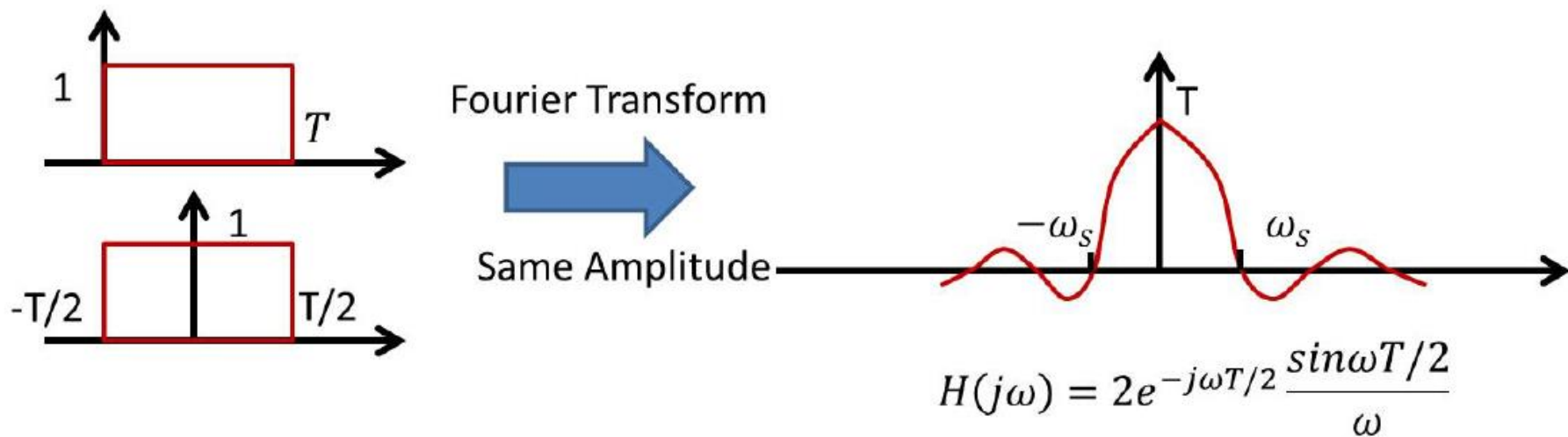
- Zero-order hold: sampling + interpolation with rectangular impulse response
- An approximation of the signal to be sampled.

# Frequency analysis (1/2)

- Step 1: Impulse-train sampling

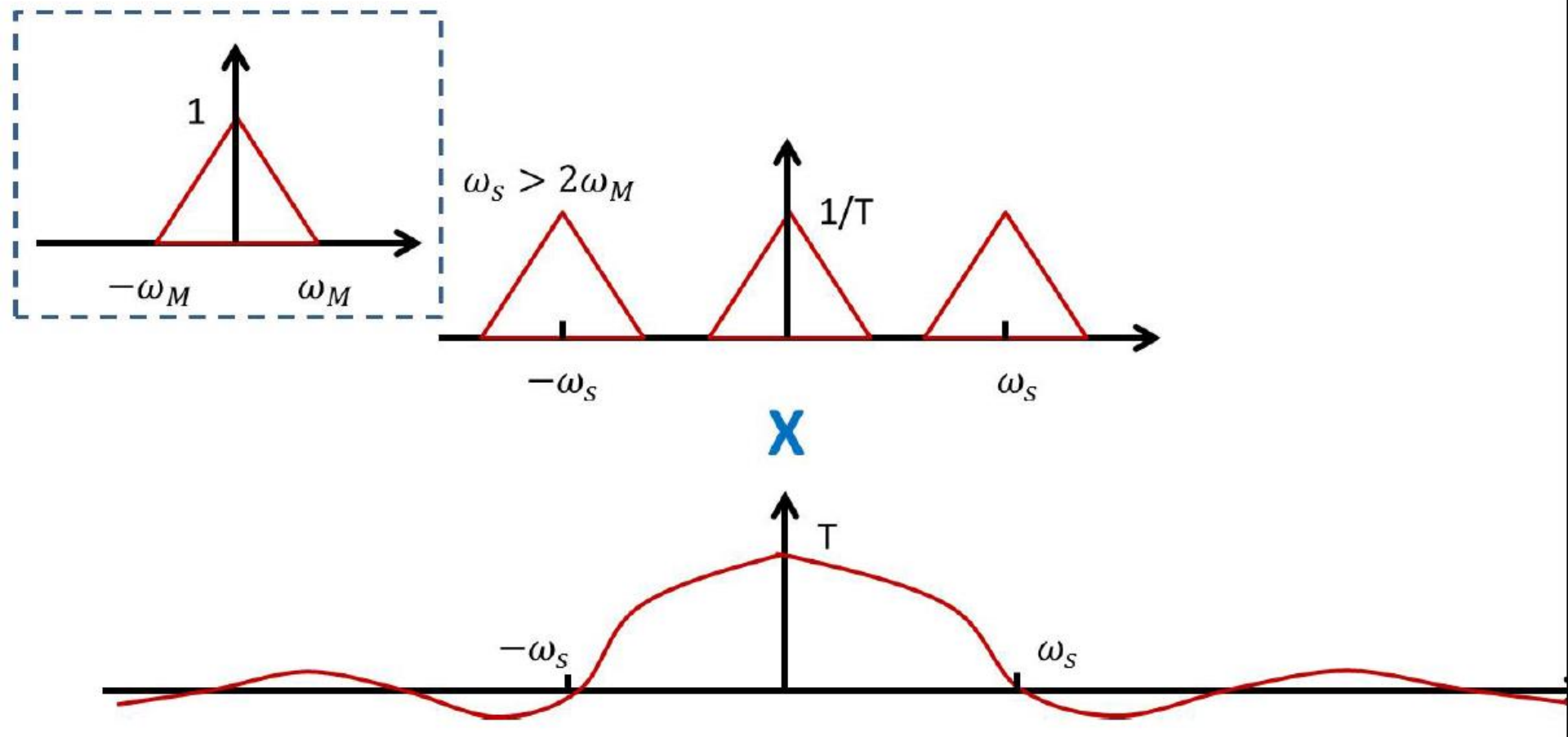


- Step 2: Frequency response of  $h_0(t)$

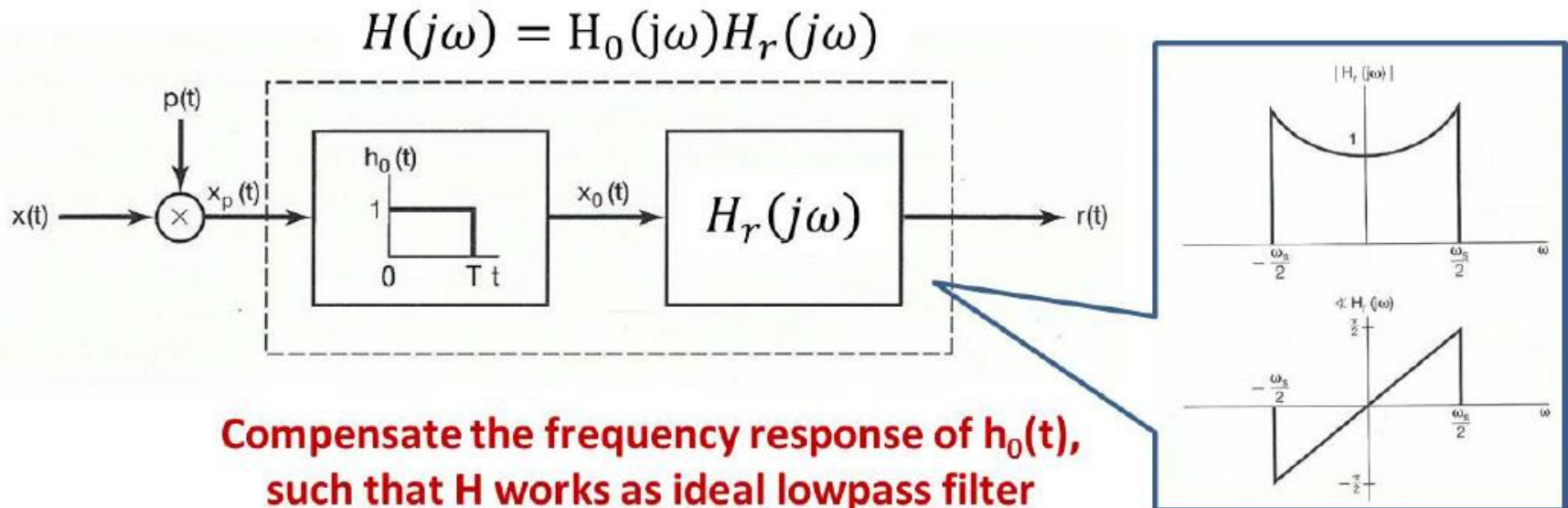




## Frequency analysis (2/2)



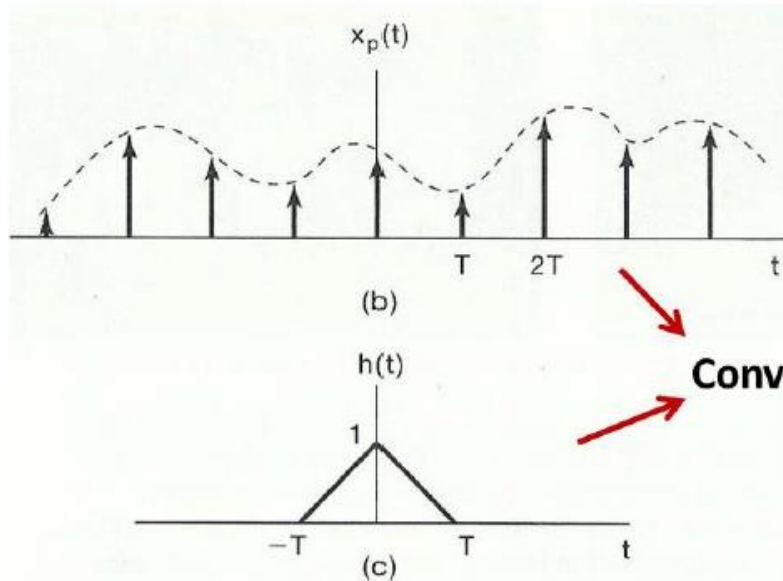
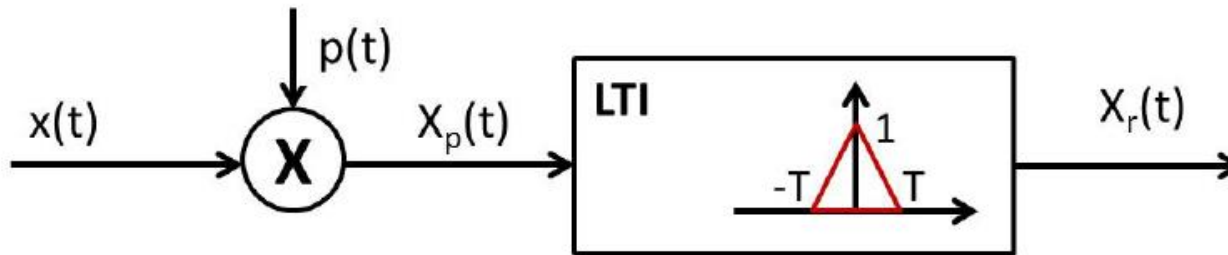
# Reconstruction



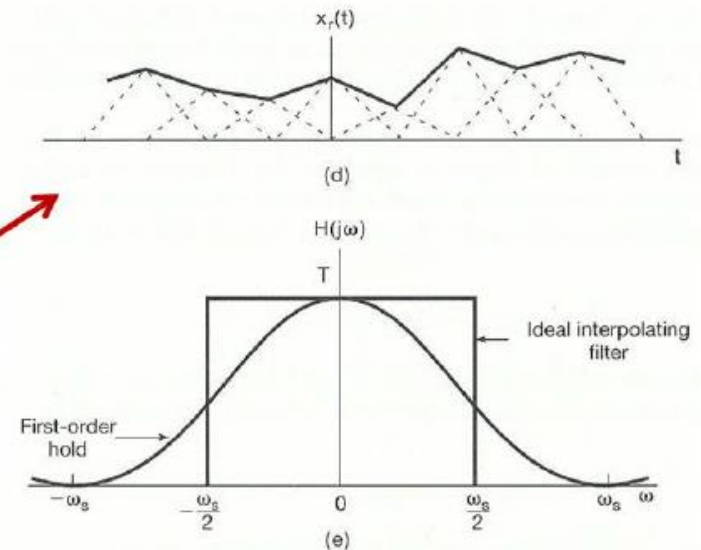
$$H_r(j\omega) = e^{j\omega T/2} \frac{\omega T}{2 \sin \omega T/2} \quad -\frac{\omega_s}{2} \leq \omega \leq \frac{\omega_s}{2}$$

- $H(j\omega)$  should be an ideal low-pass filter from  $-\omega_s/2$  to  $\omega_s/2$

# First-order hold

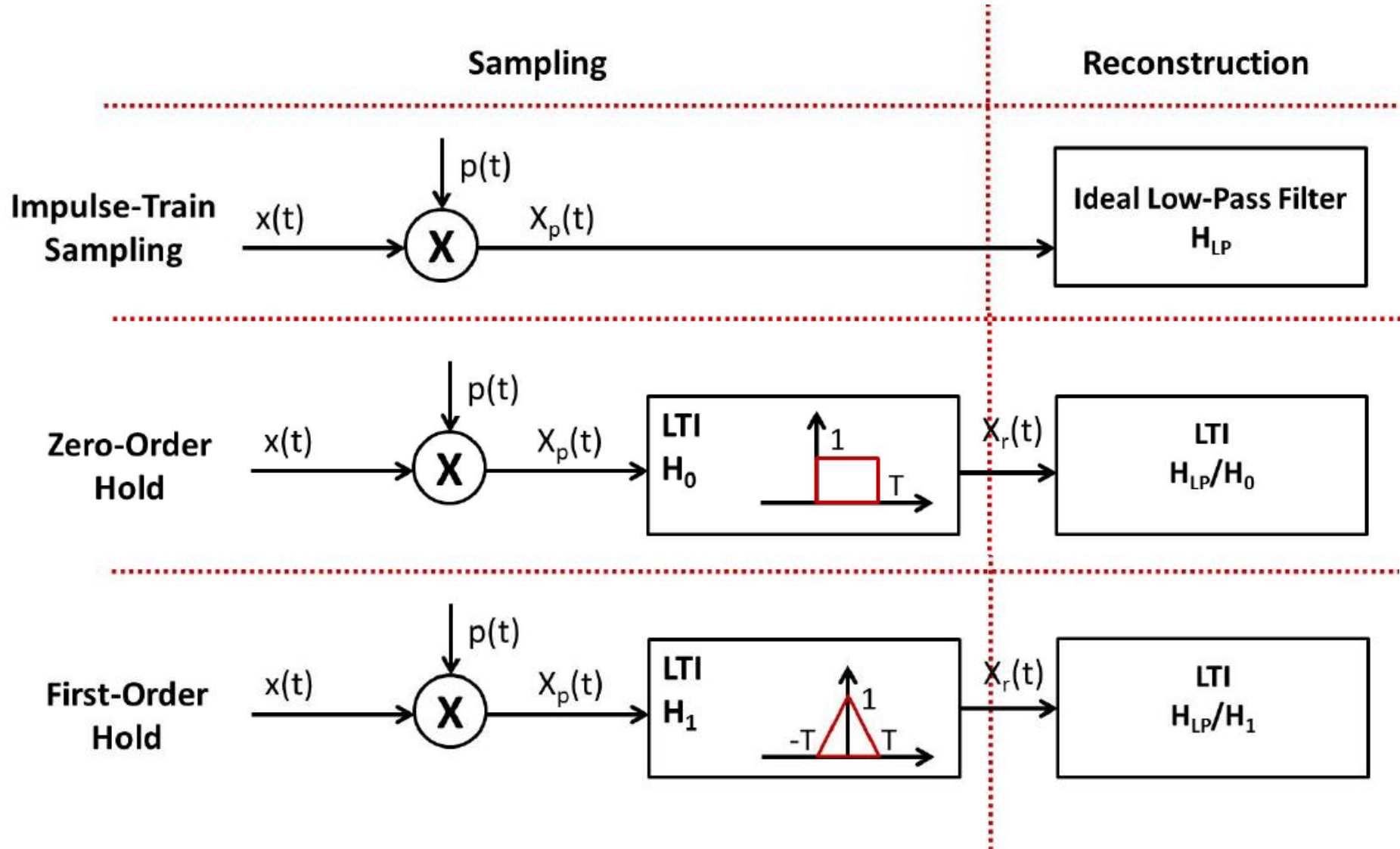


Convolution



- First-order hold: sampling + interpolation with triangular wave
- How to reconstruct?

# Summary: Sampling approaches



# Problem

## Problem (7.36)

Let  $x(t)$  be a band-limited signal such that  $X(j\omega) = 0$  for  $|\omega| \geq \pi/T$ .

(a) If  $x(t)$  is sampled using a sampling period  $T$ , determine an interpolating function  $g(t)$  such that

$$\frac{dx(t)}{dt} = \sum_{n=-\infty}^{\infty} x(nT)g(t - nT).$$

(b) Is the function  $g(t)$  unique?

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

$$x(t) = x_p(t) * h(t) \quad h(t) = \frac{\sin(\pi t / T)}{\pi t / T}$$

$$\frac{dx(t)}{dt} = x_p(t) * \frac{dh(t)}{dt} = x_p(t) * g(t) = \sum_{n=-\infty}^{\infty} x(nT) g(t - nT)$$

**Therefore:**

$$g(t) = \frac{dh(t)}{dt} = \frac{\cos(\pi t / T)}{t} - \frac{T \sin(\pi t / T)}{\pi t^2}$$