

# EEE336 Signal Processing and Digital Filtering

## Lecture 3 Sampling and Reconstruction

### 3\_1 Sampling Process

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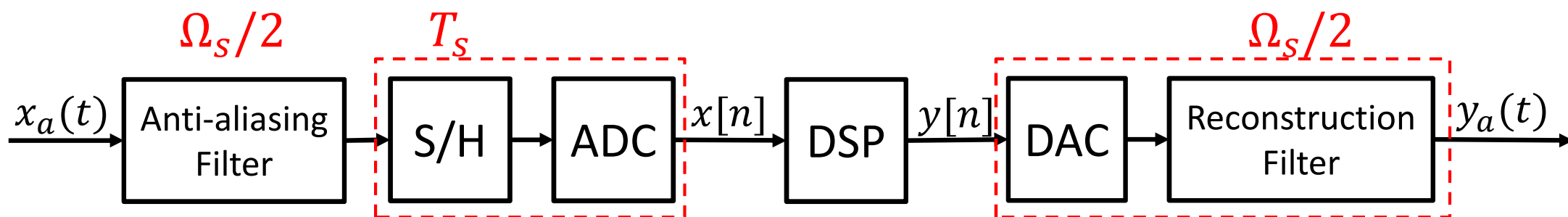
Room EE322

# Digital Processing of CT Signals

- Most signals in nature are continuous in time  
=> Need a way for “digital processing of continuous-time signals” => **SAMPLING!**

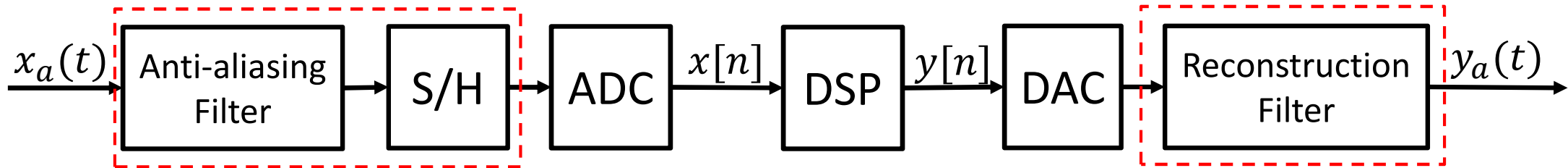


(a) Ideal data flow for the digital processing of continuous-time signals



(b) Practical data flow for the digital processing of continuous-time signals

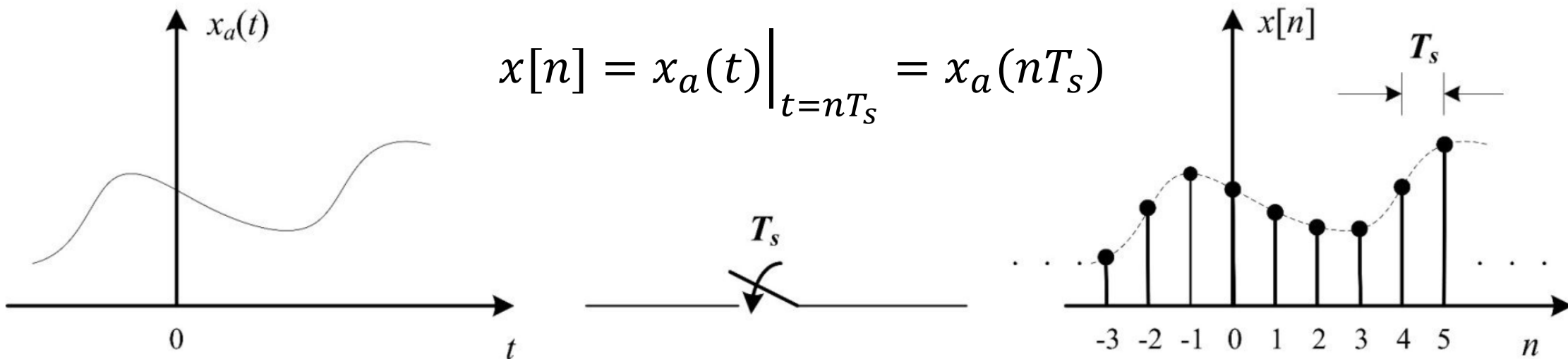
# Analog -> Digital -> Analog



- Conversion of the continuous-time signal into a discrete-time signal
  - Anti-aliasing filter – to prevent potentially detrimental effects of sampling
  - Sample & Hold – discrete in time and keep the sampling values for a while to allow the A/D converter to do its job
  - *Analog to Digital Converter (A/D) – conversion in amplitude*
- Processing of the discrete-time signal
  - Digital Signal Processing – Filter, digital processor
- Conversion of the processed discrete-time signal back into a cont.-time signal
  - *Digital to analog converter (D/A) – to obtain the continuous signal*
  - Reconstruction / smoothing filter -smooth out the signal from the D/A

# Sampling in Time domain (TD)

- A discrete-time sequence is developed by uniformly sampling the continuous-time signal  $x_a(t)$



- The time variable - time  $t$  is related to the discrete time variable  $n$  only at discrete-time instants  $t_n$

$$t_n = nT_s = \frac{n}{F_s} = \frac{2\pi n}{\Omega_s} \left\{ \begin{array}{l} T_s = 1/F_s \text{ (Sampling period, second, second/sample)} \\ F_s = 1/T_s \text{ (Sampling frequency, Hz, cycles/second)} \\ \Omega_s = 2\pi/T_s \text{ (Sampling angular frequency, radian/second)} \end{array} \right. \quad 4$$

# Sampling in Time domain (TD)

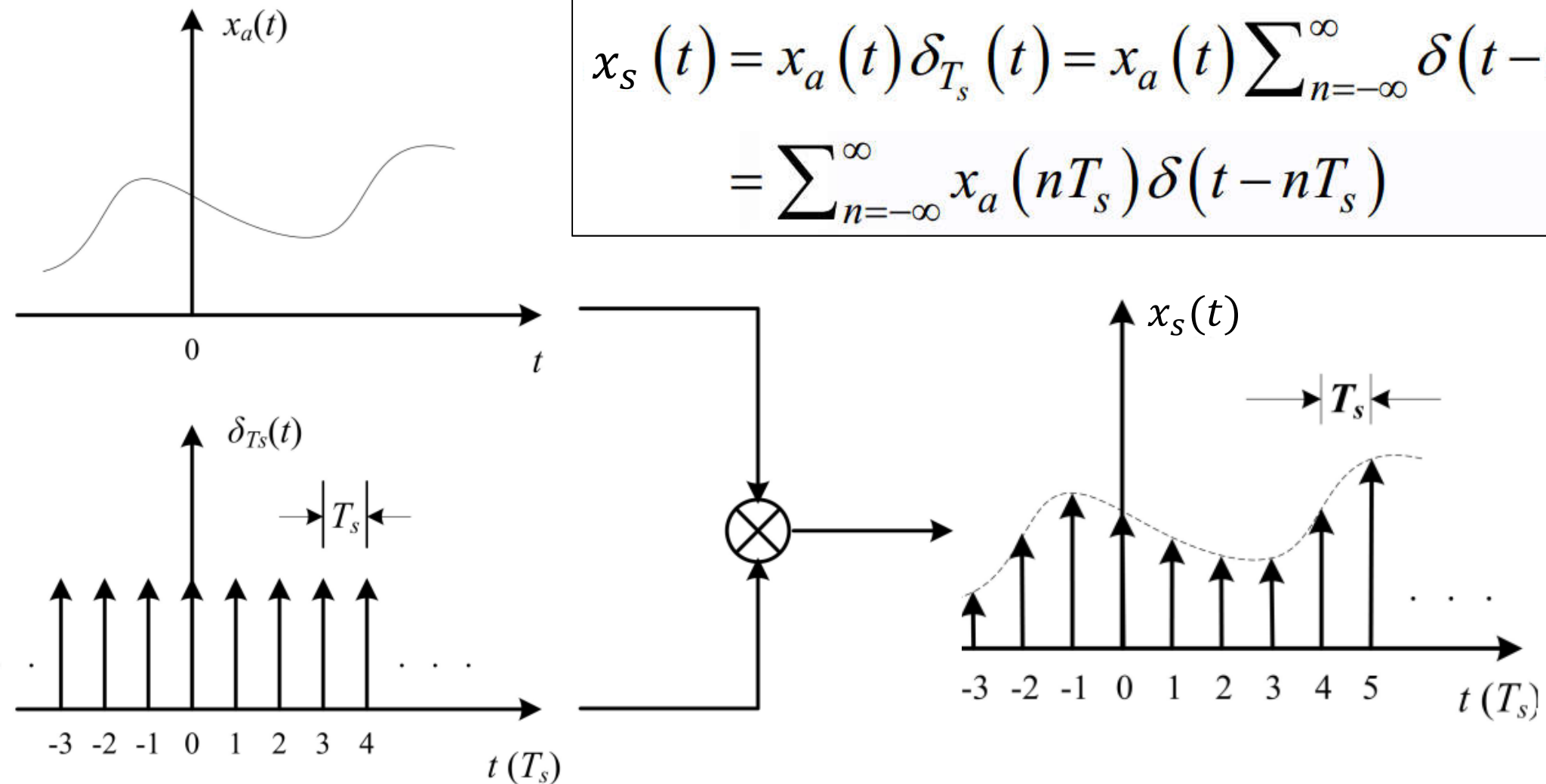
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- Consider  $x_a(t) = A\cos(\Omega_0 t + \phi)$
- Now  $x[n] = A\cos(\Omega_0 nT_s + \phi)$ 
$$= A\cos\left(\frac{2\pi\Omega_0}{\Omega_s}n + \phi\right) = A\cos(\omega_0 n + \phi) = x_a(nT)$$
  - Where  $\omega_0 = \frac{2\pi\Omega_0}{\Omega_s} = \Omega_0 T_s$
  - $\omega_0$  is the (normalized) digital angular frequency of the signal
    - Unit: radians/sample
  - $\Omega_0$  is the analog angular frequency of signal
    - Unit: radians/second
  - $\Omega_s$  is the sampling analog angular frequency
    - Unit: radians/second

# Sampling in Time domain (TD)

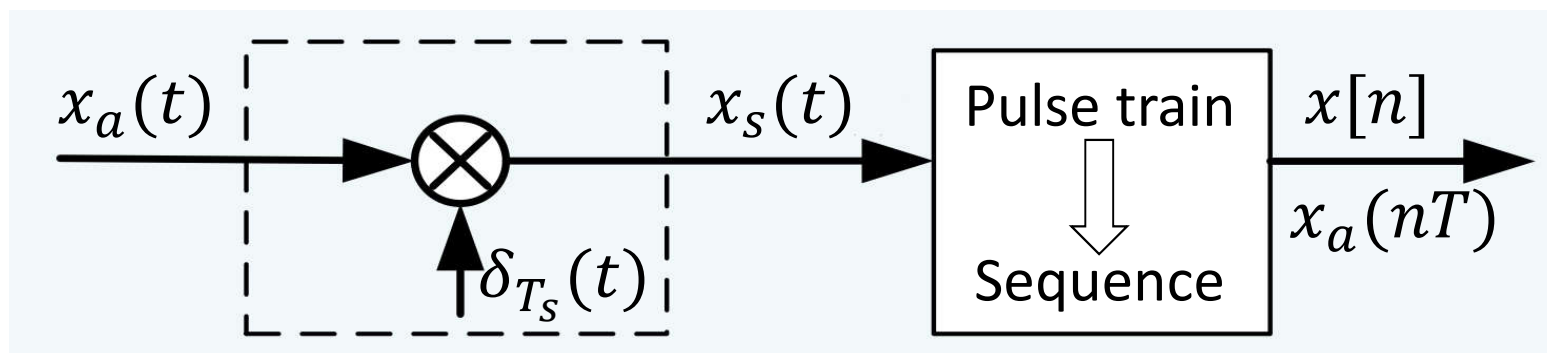
- In mathematics, the periodic sampling is modelled as the multiplication of continuous signal and impulse train

$$\begin{aligned}x_s(t) &= x_a(t) \delta_{T_s}(t) = x_a(t) \sum_{n=-\infty}^{\infty} \delta(t - nT_s) \\ &= \sum_{n=-\infty}^{\infty} x_a(nT_s) \delta(t - nT_s)\end{aligned}$$

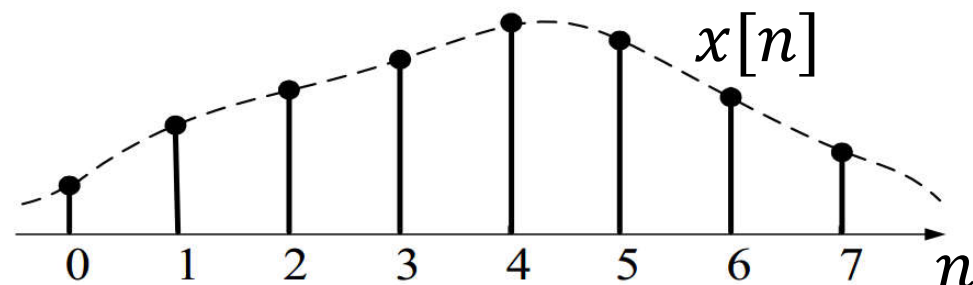
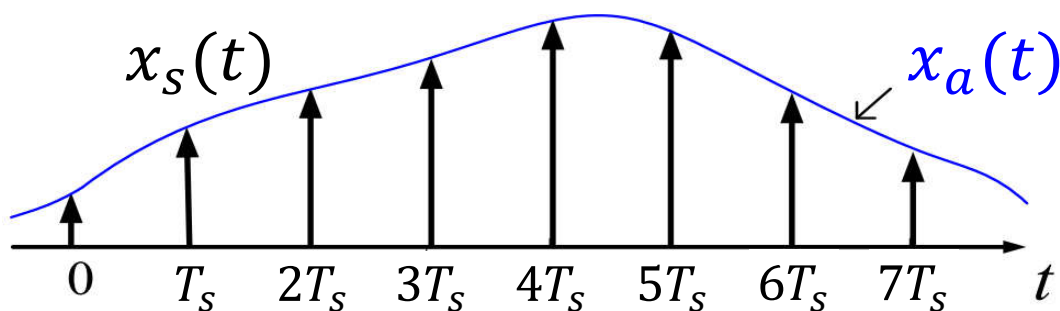


# Sampling in Time domain (TD)

- The system to convert the continuous-time (CT) signal  $x_a(t)$  to a discrete-time (DT) signal  $x[n]$  is shown:

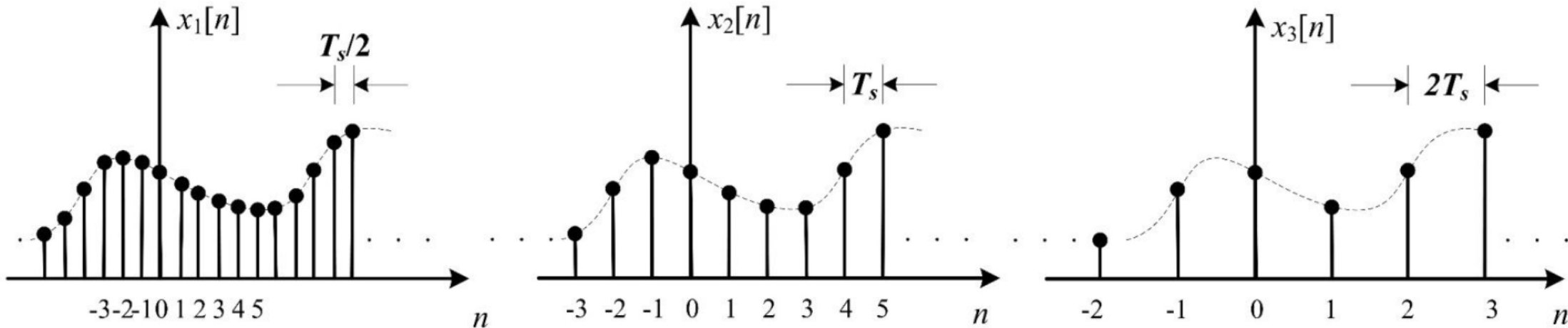


$$x_s(t) = x_a(t)\delta_{T_s}(t) = \sum_{n=-\infty}^{\infty} x_a(nT_s)\delta(t - nT_s)$$



# How signal changed after sampling

- In time domain, continuous  $\rightarrow$  discrete



- Different sampling rates, different details
  - More samples = higher sampling rate/frequency = more detail  
= more information kept = more resource occupation
  - Less samples = lower sampling rate/frequency = less details  
= more information loss = less resource occupation
- How to choose the sampling period/rate?

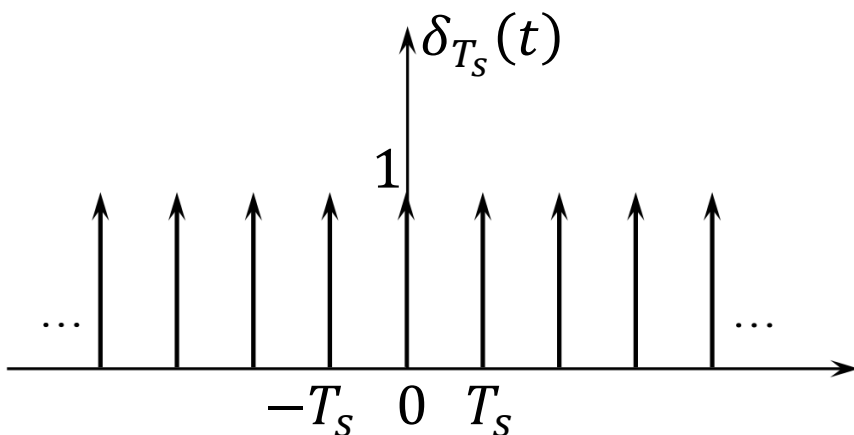




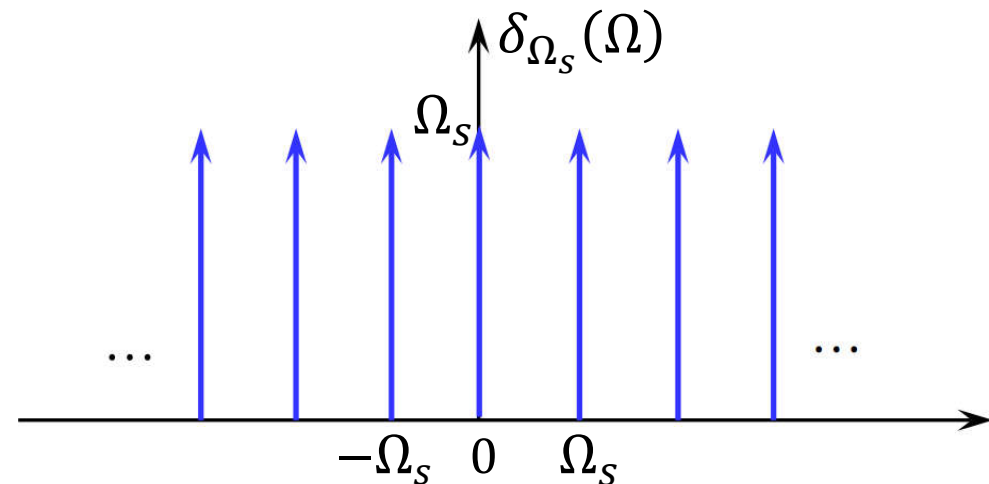
# Frequency domain analyses

- Review: CTFT of a pulse train  $\delta_{T_s}(t)$

$$\delta_{T_s}(t) = \sum_{n=-\infty}^{\infty} \delta(t - nT_s) \xleftrightarrow[\Omega_s = 2\pi f_s = \frac{2\pi}{T_s}]{\text{CTFT}} \delta_{\Omega_s}(\Omega) = \Omega_s \sum_{k=-\infty}^{\infty} \delta(\Omega - k\Omega_s)$$



TD: Time domain



FD: Frequency domain

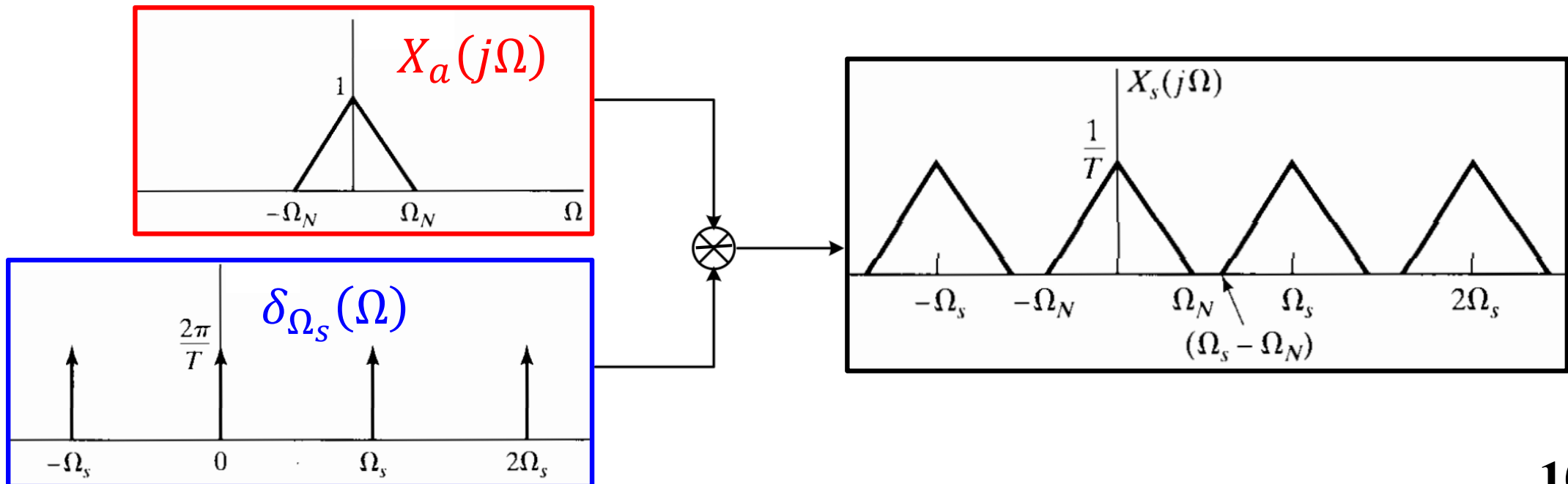
# Sampling in Frequency domain (FD)

- In TD: multiplication between  $x_a(t)$  and  $\delta_{T_s}(t)$

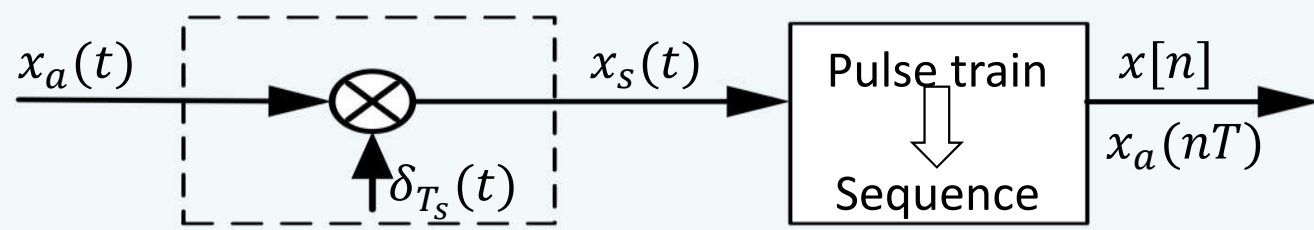
$$x_s(t) = x_a(t) \cdot \delta_{T_s}(t)$$

- In FD: convolution between  $X_a(j\Omega)$  and  $\delta_{\Omega_s}(\Omega)$

$$X_s(j\Omega) = \frac{1}{2\pi} X_a(j\Omega) * \Omega_s \sum_{k=-\infty}^{\infty} \delta(\Omega - k\Omega_s) = \frac{1}{T_s} \sum_{k=-\infty}^{\infty} X_a[j(\Omega - k\Omega_s)]$$



# Sampling in FD



- An alternative expression of  $X_s(j\Omega)$  is:

$$x_s(t) = \sum_{n=-\infty}^{\infty} x_a(nT_s) \delta(t - nT_s)$$

CTFT  
↕

$$X_s(j\Omega) = \sum_{n=-\infty}^{\infty} x_a(nT_s) e^{-j\Omega T_s n}$$

Since:

$$\left. \begin{aligned} x[n] &= x_a(nT_s) \\ X(e^{j\omega}) &= \sum_{n=-\infty}^{\infty} x[n] e^{-j\omega n} \end{aligned} \right\} X_s(j\Omega) = X(e^{j\omega}) \Big|_{\omega=\Omega T_s} = X(e^{j\Omega T_s})$$

Recall

$$X_s(j\Omega) = \frac{1}{T_s} \sum_{k=-\infty}^{\infty} X_a[j(\Omega - k\Omega_s)]$$

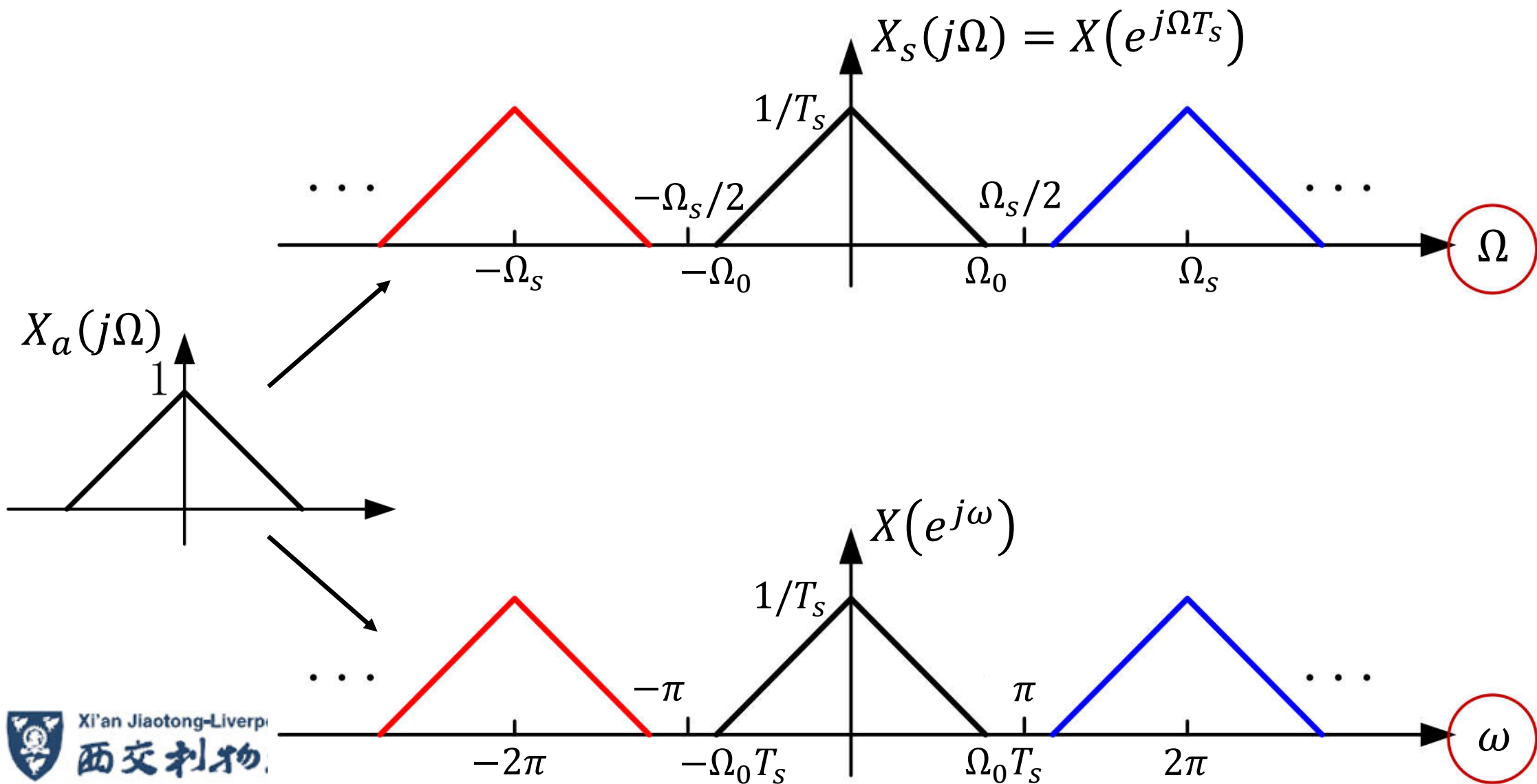
$$X(e^{j\Omega T_s}) = \frac{1}{T_s} \sum_{k=-\infty}^{\infty} X_a[j(\Omega - k\Omega_s)]$$

$$X(e^{j\omega}) = \frac{1}{T_s} \sum_{k=-\infty}^{\infty} X_a \left[ j \left( \frac{\omega}{T_s} - \frac{2\pi k}{T_s} \right) \right]$$



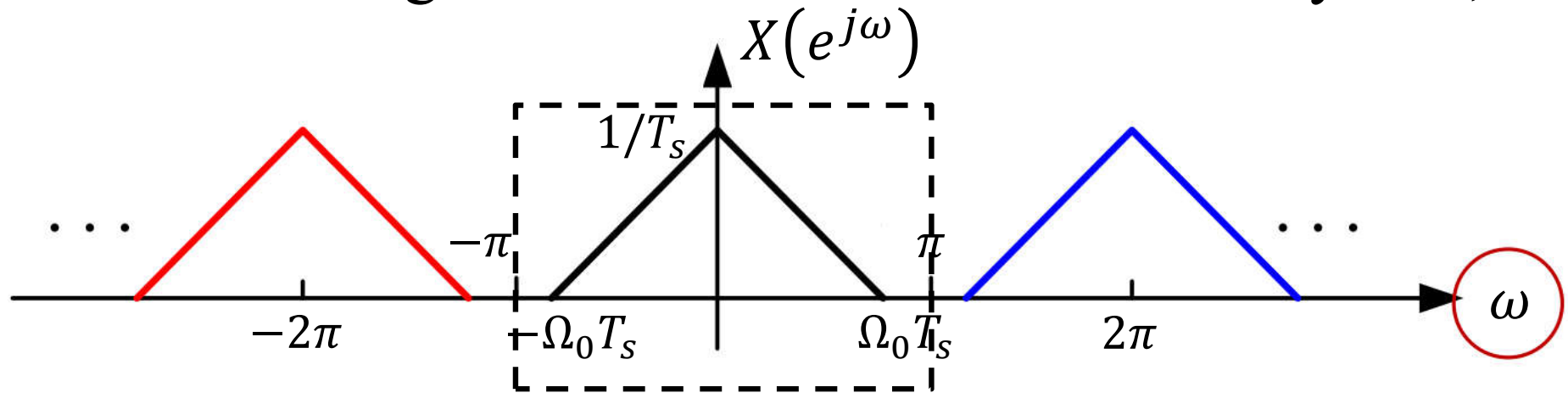
# Sampling in Frequency domain (FD)

- Discretization in TD  $\Rightarrow$  Periodicity in FD



# Recovery of the CT signal

- The spectrum of the sampled signal contains all the information of the original CT signal
  - So the CT signal can be recovered without any loss;



- But a condition needs to be satisfied:

$$\Omega_0 T_s \leq \pi \iff 2\Omega_0 \leq \Omega_s$$

- The Nyquist theorem!

## 3\_1 Wrap up

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- What is sampling process?
  - The first step to convert a continuous-time signal to a discrete-time signal;
- In time domain: multiplication the CT signal to a pulse train, then convert the modulated pulse train to sequence;
- In frequency domain: copy and shift (create infinite replica) the spectrum of the CT signal;
  - The CT signal can be recovered from the sampled signal if Nyquist theorem is satisfied, i.e.,  $2\Omega_0 \leq \Omega_s$

# EEE336 Signal Processing and Digital Filtering

## Lecture 3 Sampling and Reconstruction

### 3\_2 Aliasing

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# An Example of Sampling

*In Time Domain*

- Three continuous-time signals are sampled at  $T_s = 0.1$  s

$$g_1(t) = \cos(6\pi t)$$

$$F_s = 10 \text{ Hz}$$

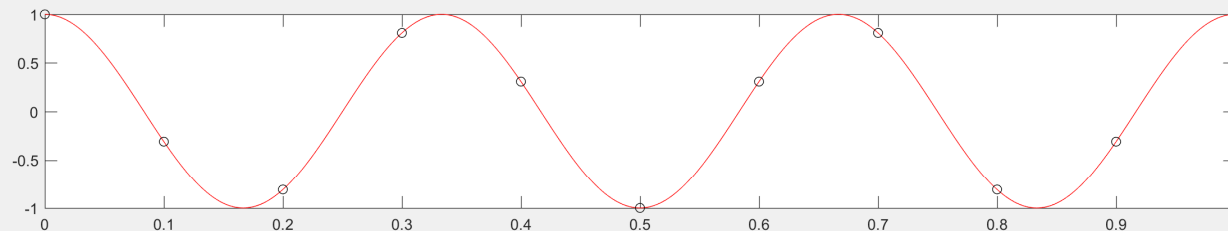
$$g_2(t) = \cos(14\pi t)$$

$$\Omega_s = 20\pi$$

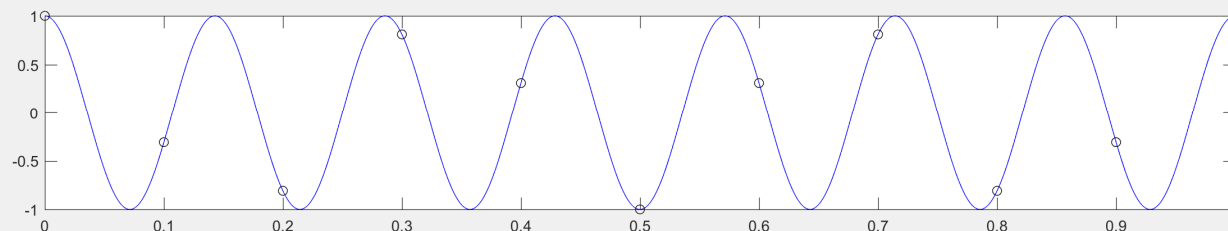
$$g_3(t) = \cos(26\pi t)$$

- Generating the following sequence as shown in figure

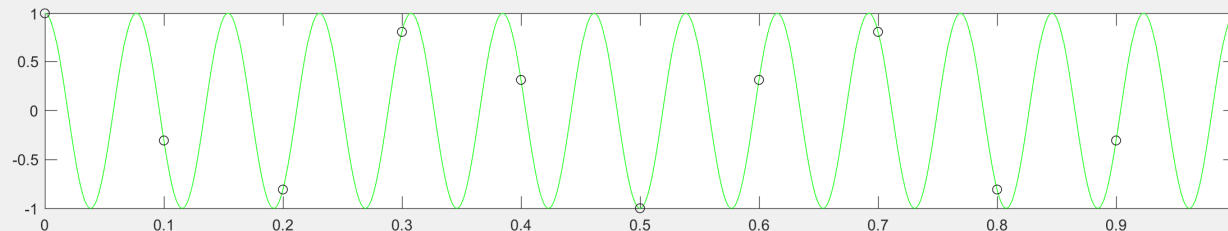
$$g_1(t) = \cos(6\pi t)$$



$$g_2(t) = \cos(14\pi t)$$



$$g_3(t) = \cos(26\pi t)$$

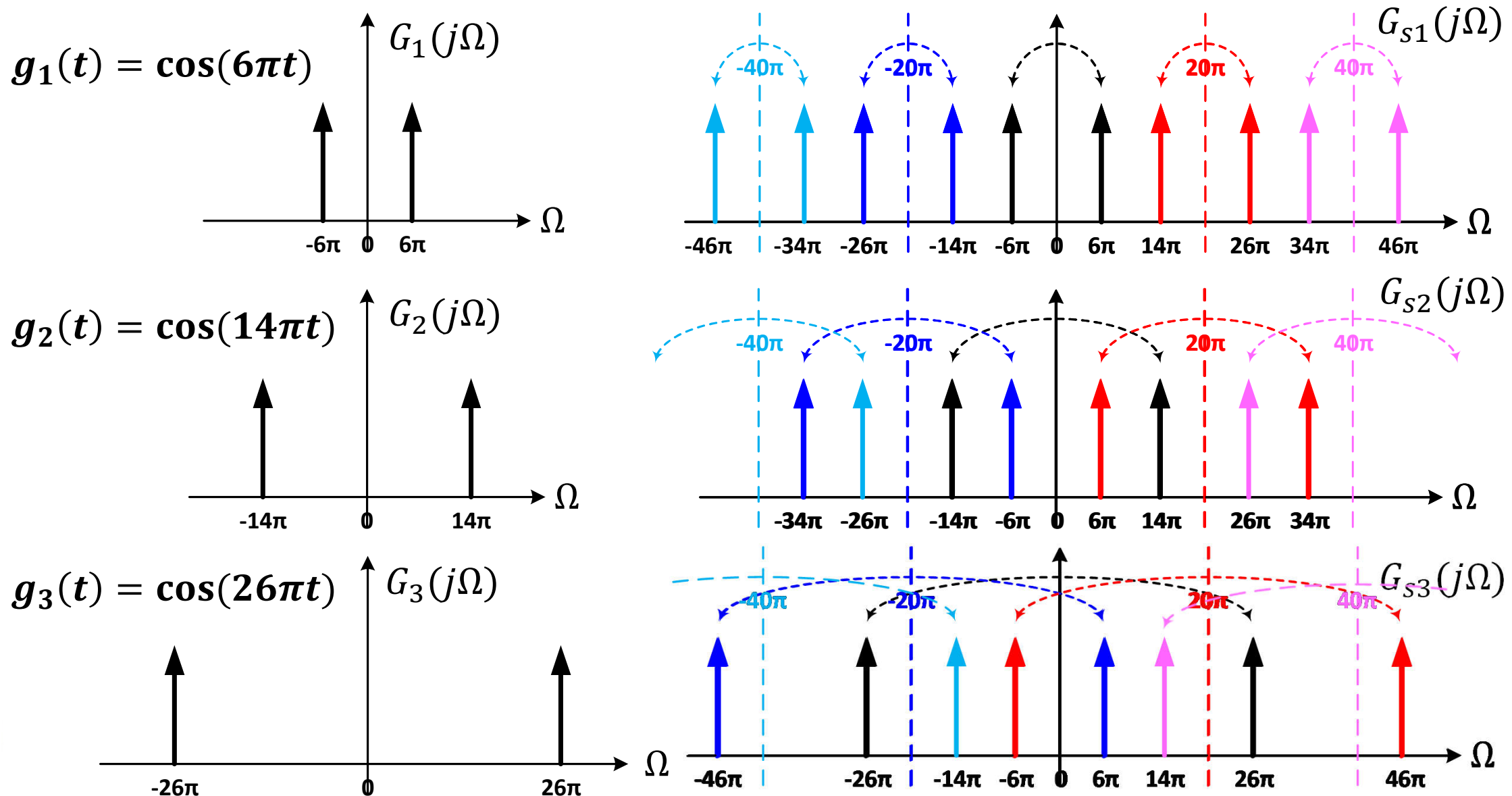




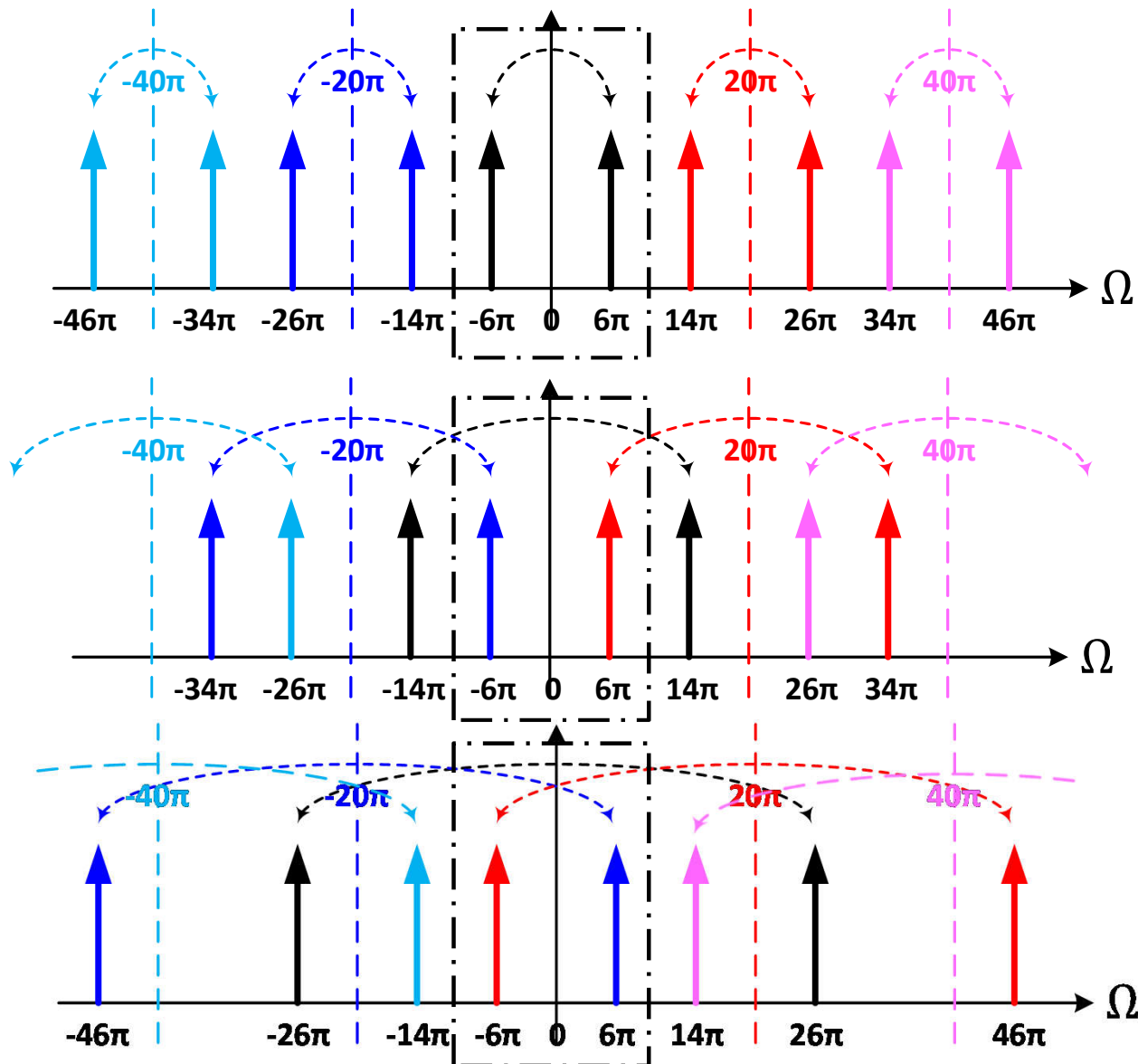
# An Example of Sampling

*In Frequency Domain*

- The angular sampling frequency is  $\Omega_s = 20\pi$



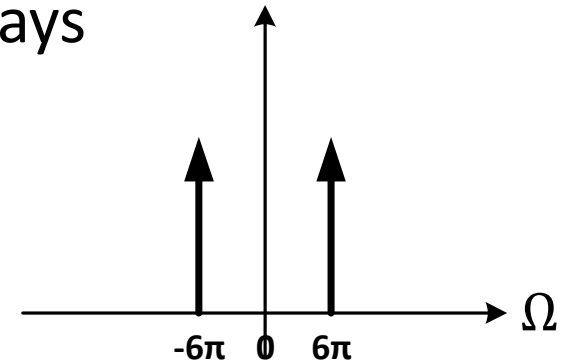
# To recover the CT signal



Use a lowpass filter with cutoff frequency at  

$$\Omega_c = \Omega_s/2 = 10\pi$$

The signal filtered out is always

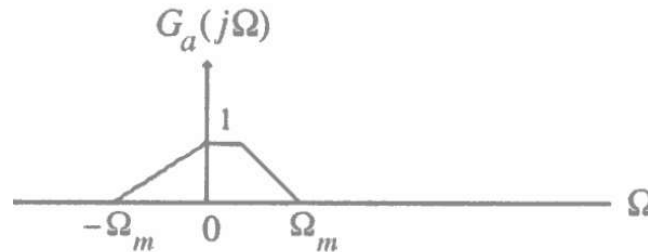


That is  $g_1(t) = \cos(6\pi t)$   
 in time domain



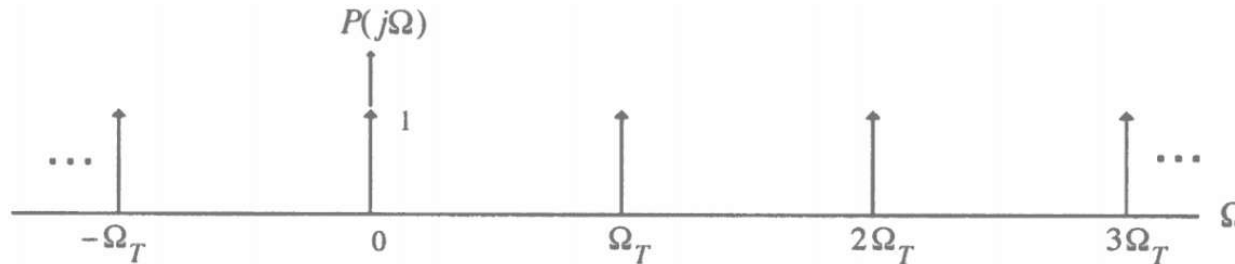
# Another Example of Sampling

- Assume  $g_a(t)$  is a band-limited signal with a CTFT  $G_a(j\Omega)$ :

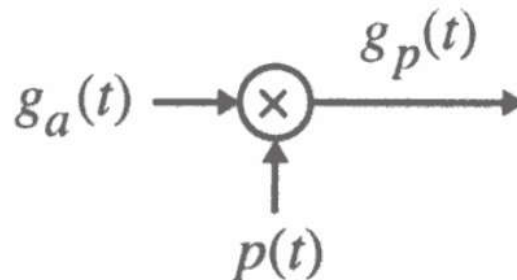


The highest frequency of  $g_a(t)$  is  $\Omega_m$

- The spectrum  $P(j\Omega)$  of  $p(t)$  having a sampling period  $T = \frac{2\pi}{\Omega_T}$

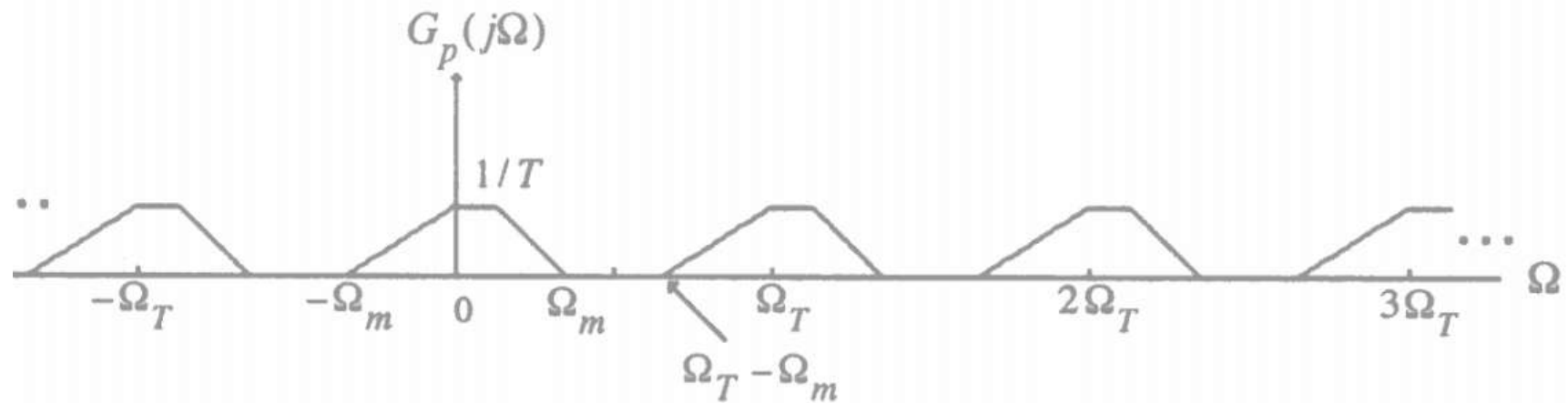


- Perform the sampling is to copy and paste  $G_a(j\Omega)$  at every  $\Omega_T$

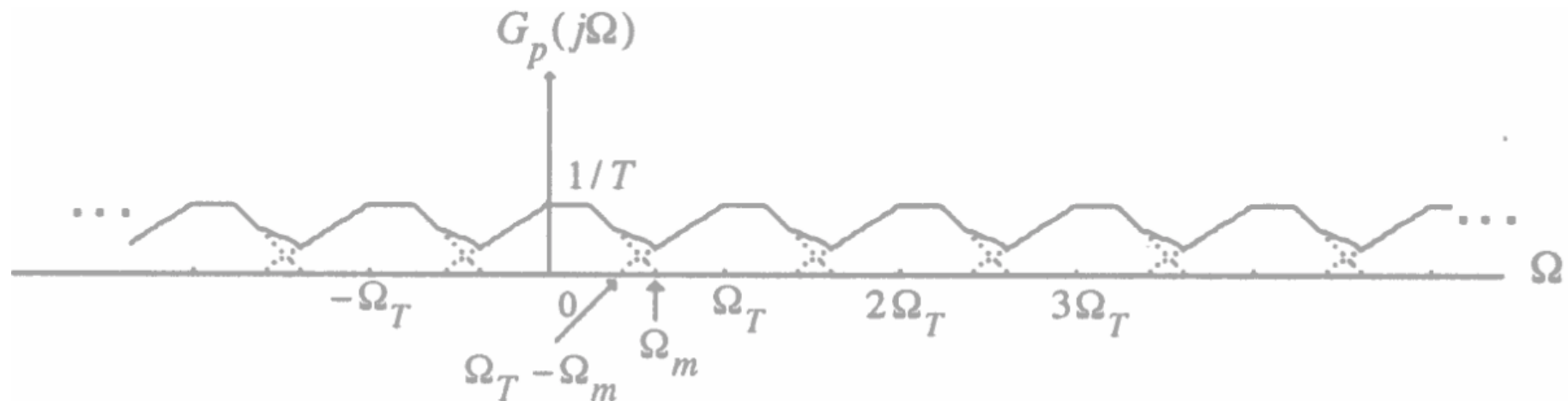


# Another Example of Sampling

- If  $\Omega_m \leq \Omega_T - \Omega_m$



- If  $\Omega_m \geq \Omega_T - \Omega_m$



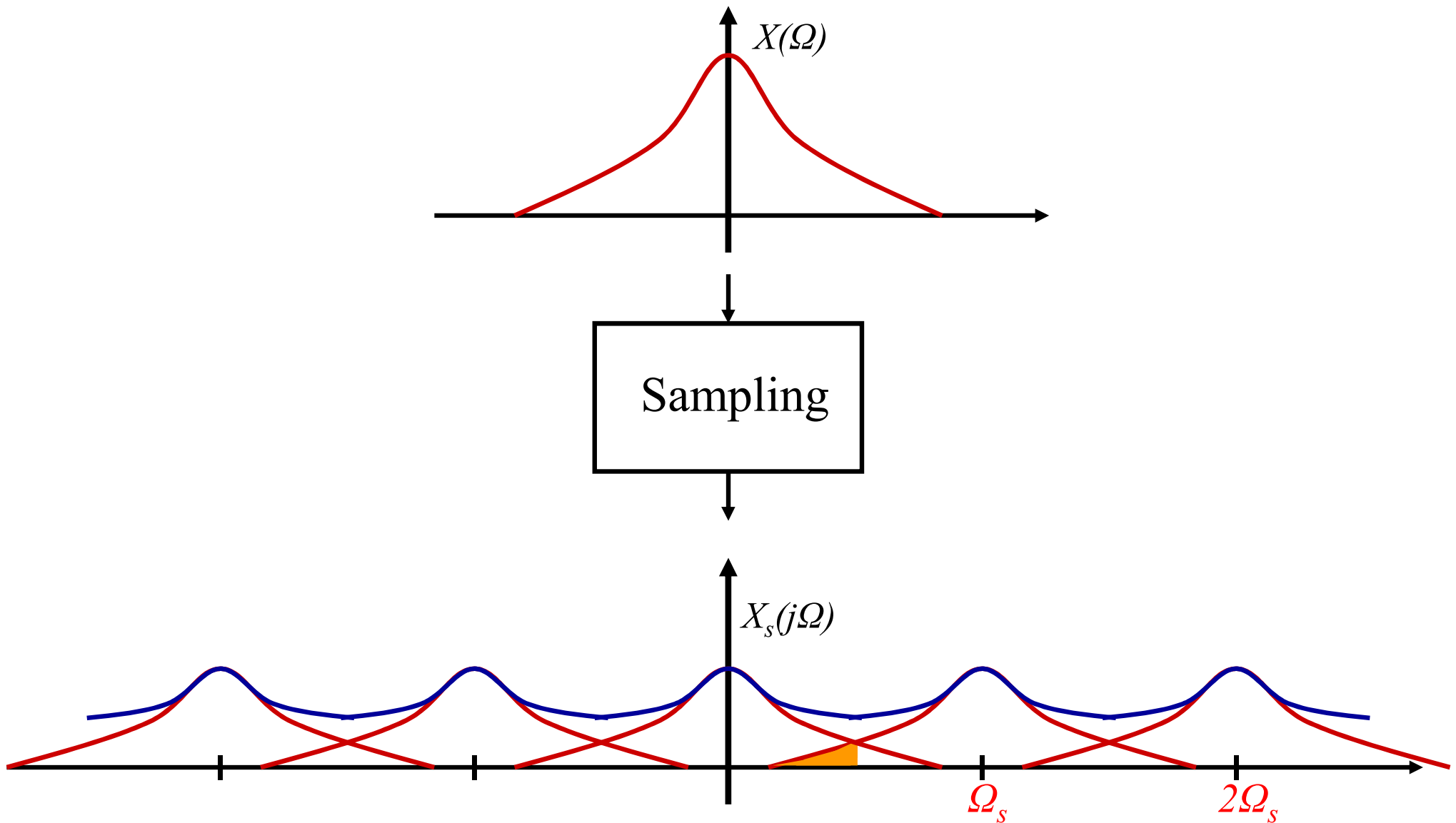
# Sampling Theorem

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- The highest frequency  $\Omega_m$  contained in the signal is called the Nyquist frequency since it determines the minimum sampling frequency  $\Omega_T = 2\Omega_m$
- The frequency  $\Omega_T / 2$  is referred to as the folding frequency
  - Critical sampling corresponds to  $\Omega_T = 2\Omega_m$
  - Oversampling corresponds to  $\Omega_T \gg 2\Omega_m$
  - Undersampling corresponds to  $\Omega_T < 2\Omega_m$

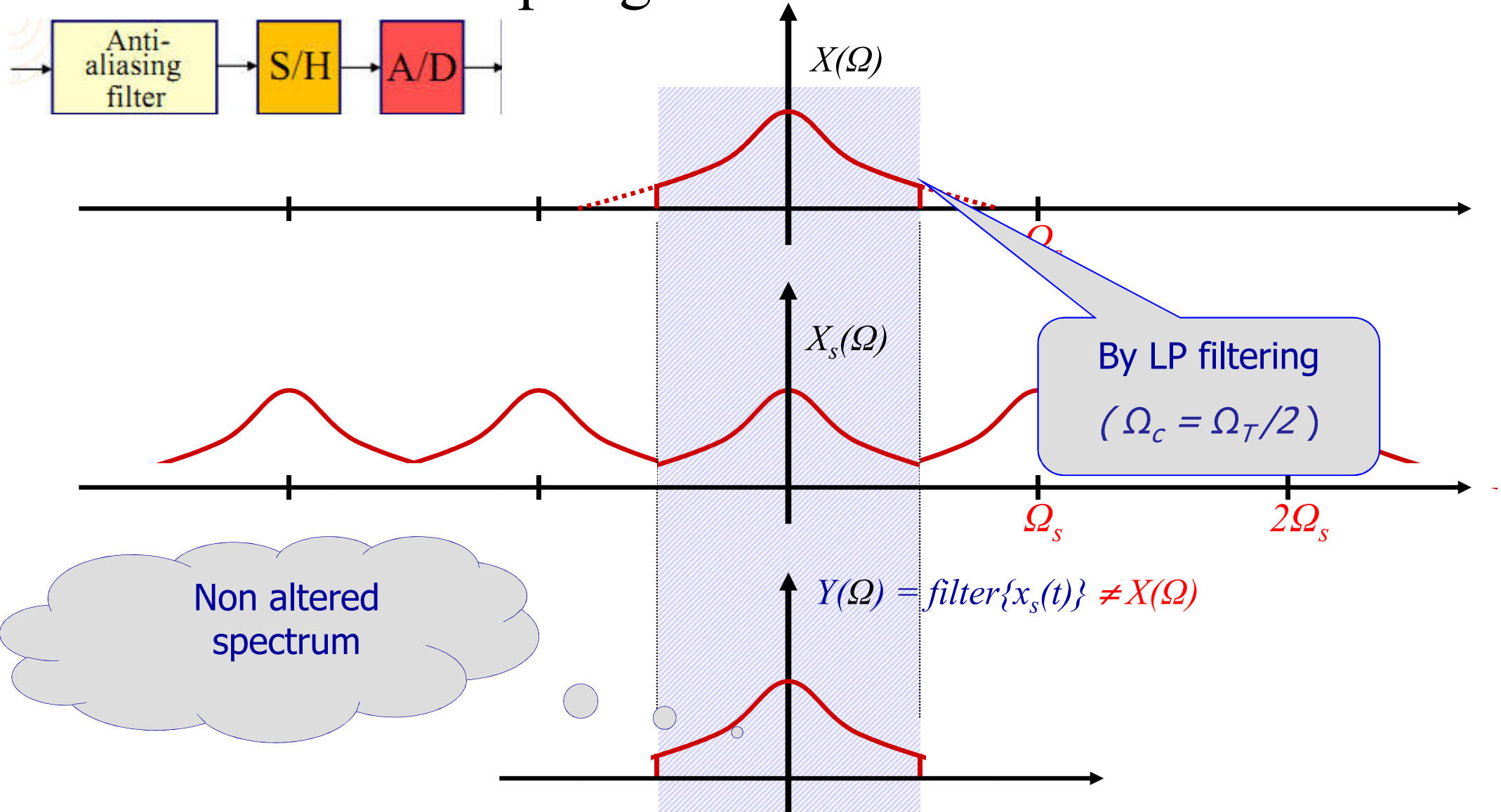
# Under-sampling

- Frequency folding error (aliasing)




# Anti-Aliasing Filter

- When the sampling rate is below the  $2B$

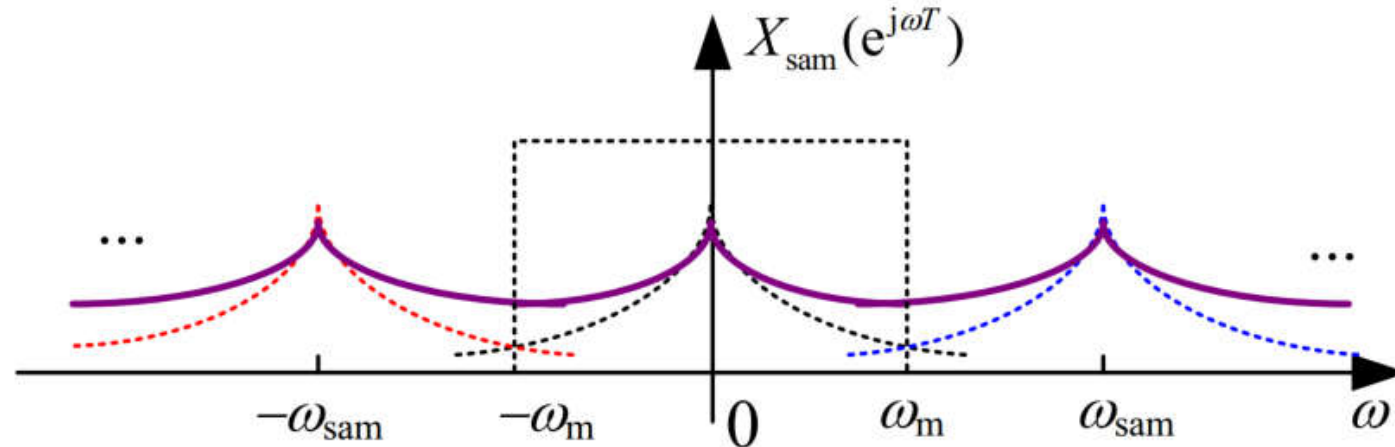
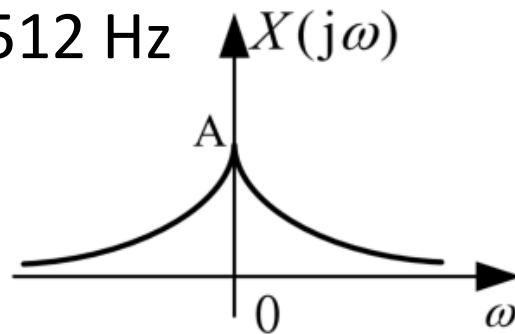


# Anti-Aliasing Filter

Original,  $F_s = 44100$  Hz 

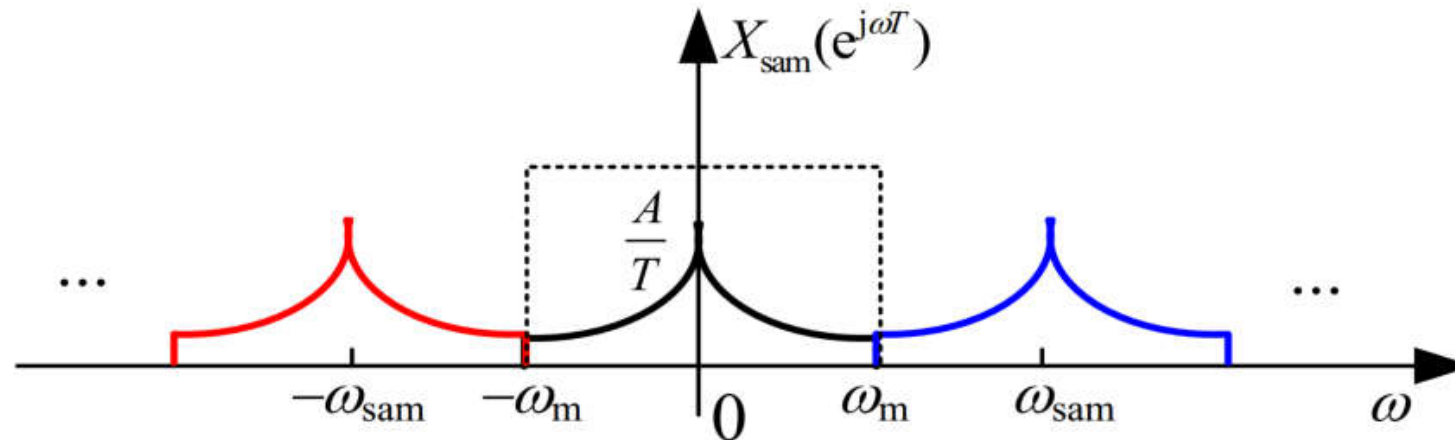
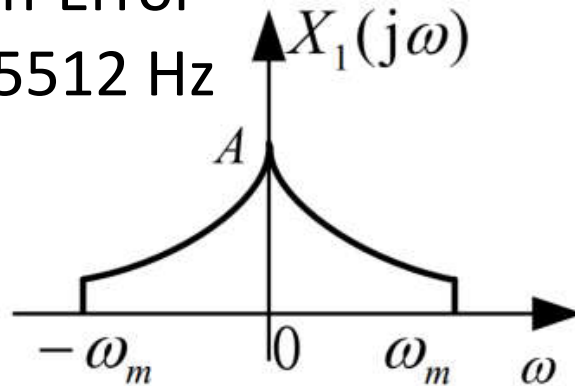
Folding Error

$F_s = 5512$  Hz



Cutoff Error

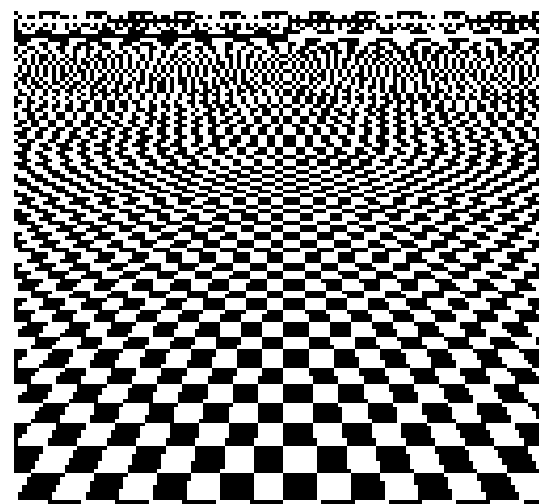
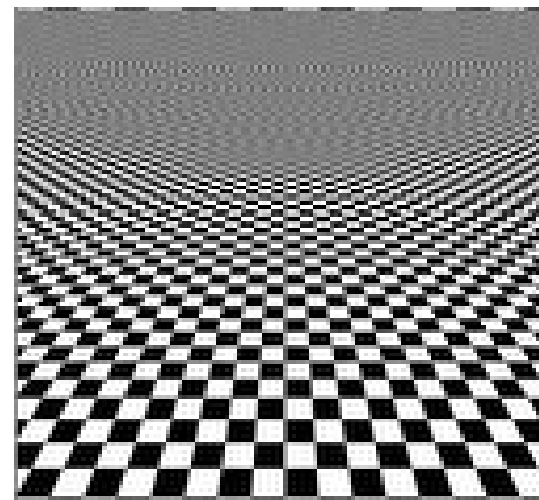
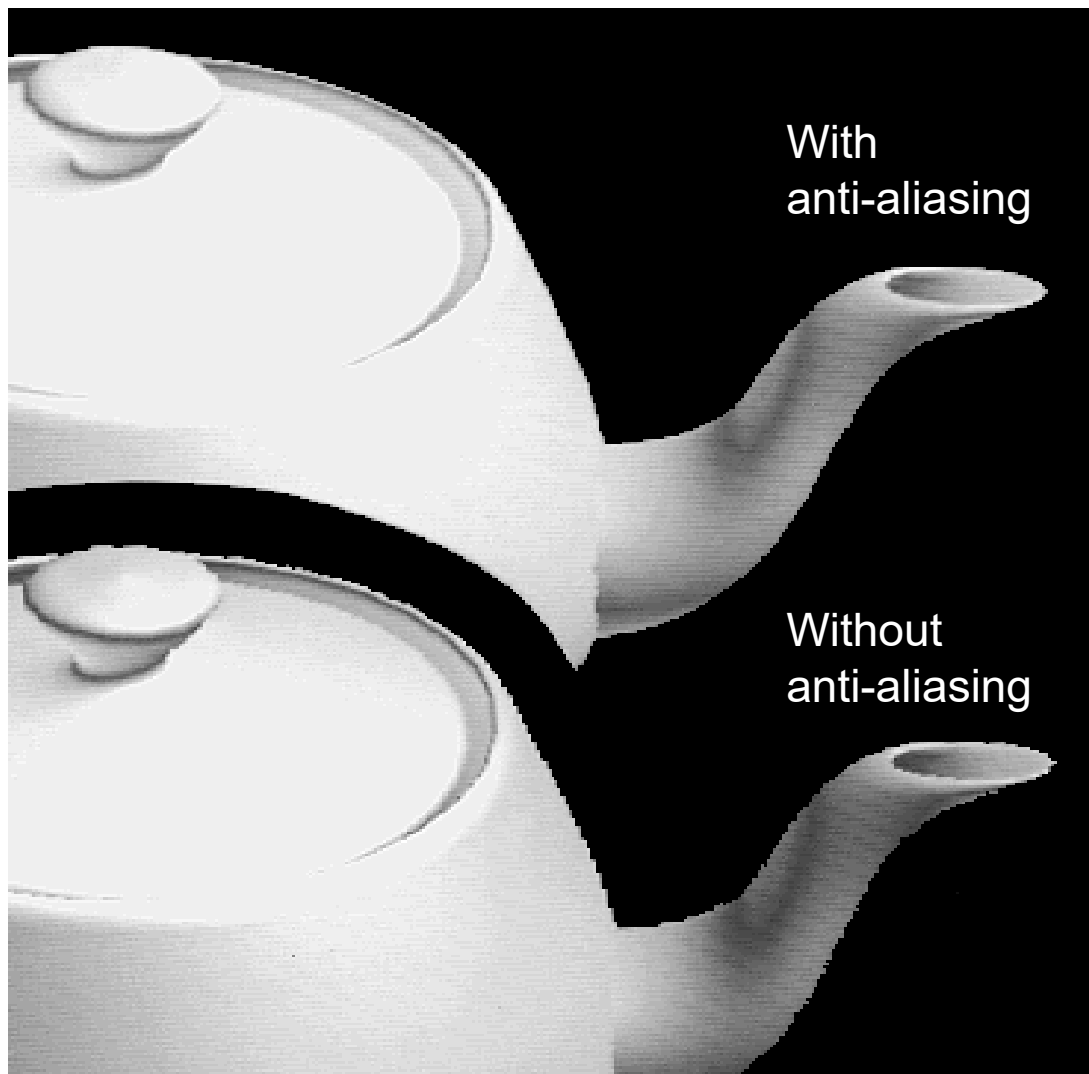
$F_s = 5512$  Hz





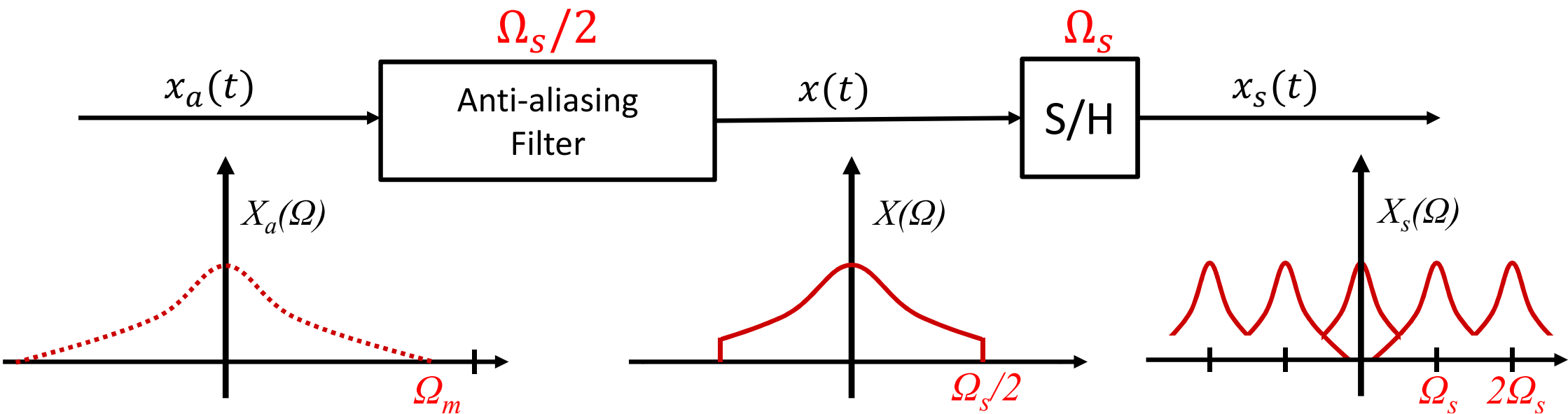
# Aliasing in Digital Images

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## 3\_2 Wrap up

- Nyquist theorem: to avoid aliasing, the sampling frequency
$$\Omega_s \geq 2\Omega_{max}$$
- Three types of sampling: over sampling, critical sampling and under sampling
- Aliasing: If Nyquist theorem was not satisfied, aliasing happens
  - To reduce the aliasing error, passing the CT signal through an “Anti-aliasing filter” before sampling it.



# EEE336 Signal Processing and Digital Filtering

## Lecture 3 Sampling and Reconstruction

### 3\_3 Interpolation / Reconstruction

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# Interpolation

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- What is interpolation?
  - In the mathematical field of numerical analysis, interpolation is a method of *constructing new data points* within the range of a discrete set of known data points.
  - In this module, interpolation is a procedure whereby we convert a discrete-time (DT) sequence  $x[n]$  to a continuous-time (CT) function  $x(t)$ .
  - Requirement: for the CT function  $x(t)$ , its values at multiples of  $T_s$  should be equal to the corresponding points of the DT sequence  $x[n]$ :

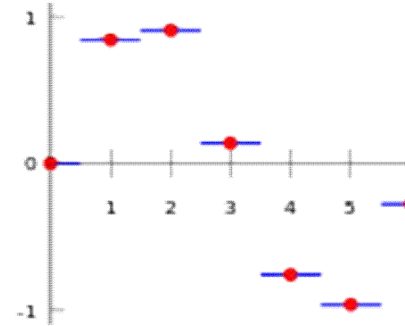
$$x(t) \Big|_{t=nT_s} = x[n]$$

The interpolation problem now reduces to “filling the gap” between these instants.

# Interpolation methods

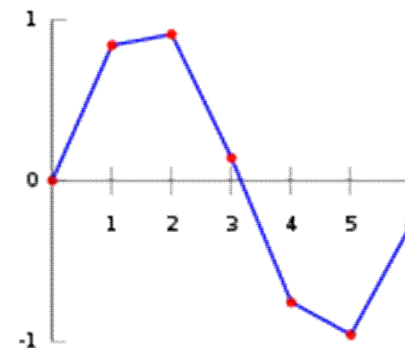
- 1. Zero-order/Local Interpolation

$$I_0(t) = \text{rect}(t)$$

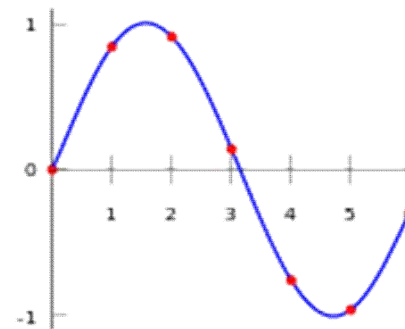


- 2. First-order/Linear Interpolation

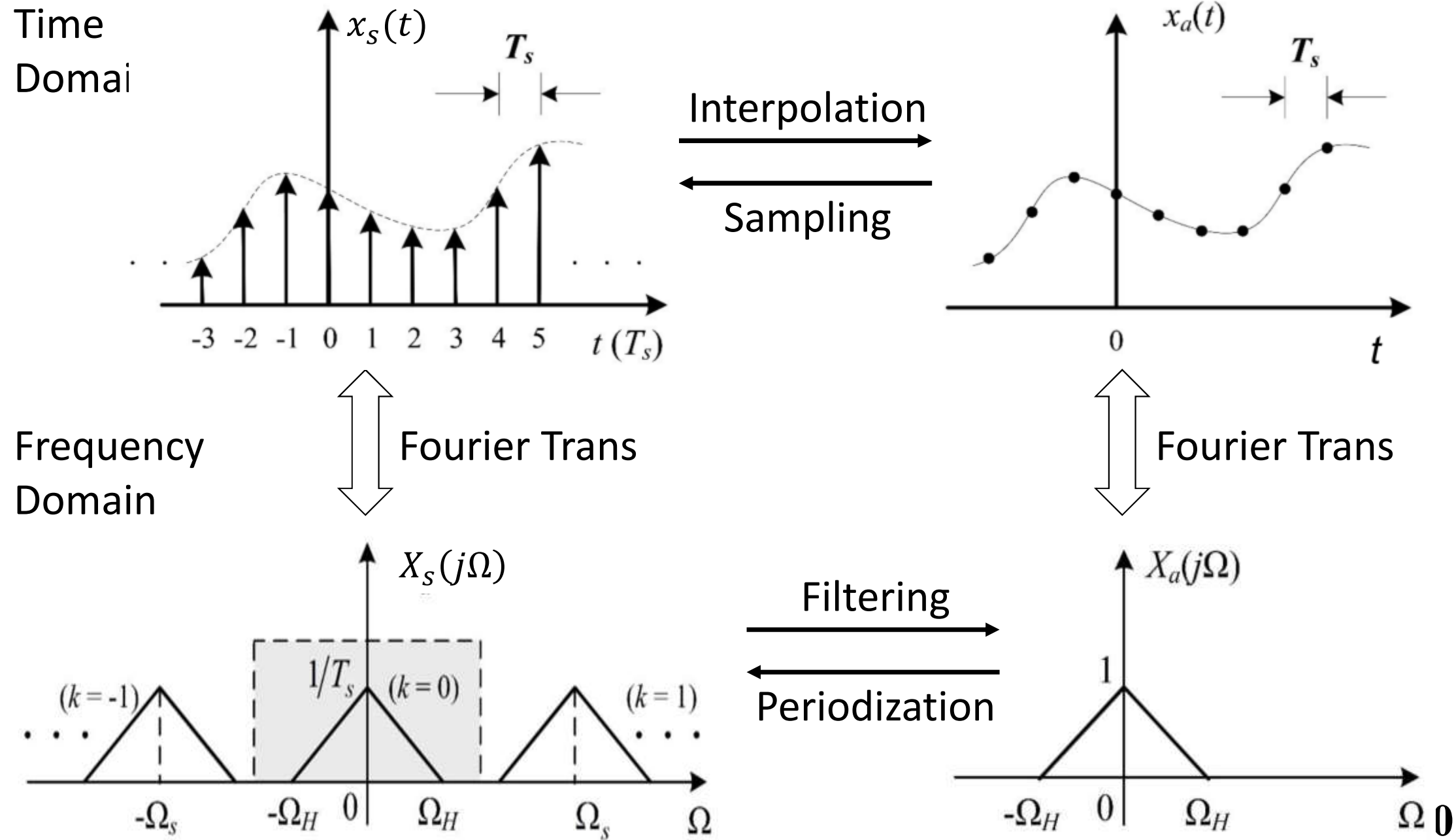
$$I_1(t) = \begin{cases} 1 - |t| & \text{if } |t| < 1 \\ 0 & \text{otherwise} \end{cases}$$



- 3. Higher-order/Polynomial Interpolation

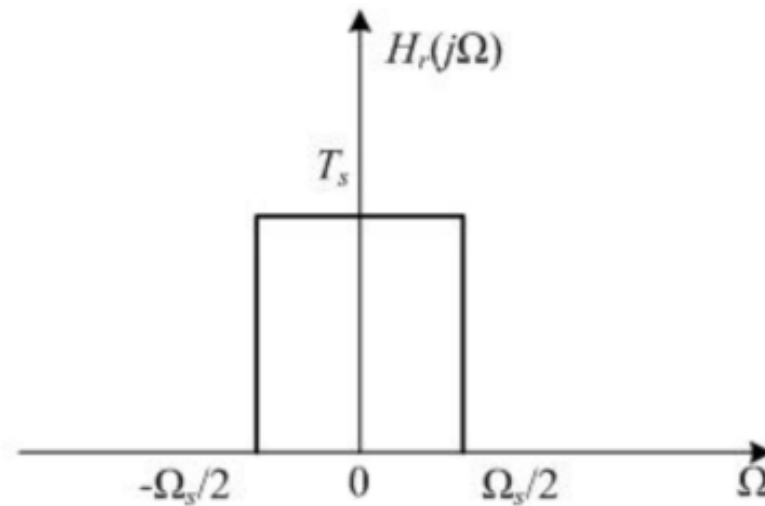


# Reconstruction Theory



# Reconstruction filter in Frequency domain

- Reconstruction or smoothing filter is used to eliminate all the replicas of the spectrum outside the baseband
- Ideal lowpass filter
  - Frequency domain  $H_r(j\Omega) = \begin{cases} T_s, & |\Omega| < \Omega_c \\ 0, & |\Omega| > \Omega_c \end{cases}$

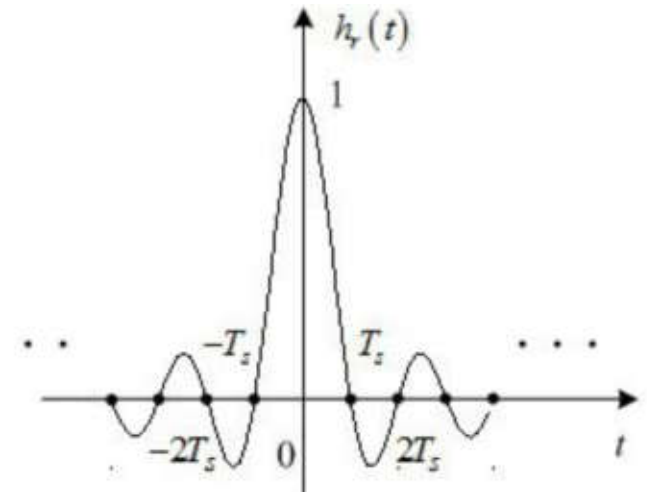


# Reconstruction filter in time domain

- This lowpass filter in time domain is a “sinc” function:

- Time domain

$$\begin{aligned} h_r(t) &= \frac{1}{2\pi} \int_{-\infty}^{\infty} H_r(j\Omega) e^{j\Omega t} d\Omega = \frac{T_s}{2\pi} \int_{-\Omega_c}^{\Omega_c} e^{j\Omega t} d\Omega \\ &= \frac{\sin(\Omega_s t/2)}{\Omega_s t/2} = \frac{\sin(\pi t/T_s)}{\pi t/T_s} = \text{sinc}\left(\frac{t}{T_s}\right) \end{aligned}$$



- Multiply with  $H_r(j\Omega)$  (in FD) is equivalent to convolve with  $h_r(t)$  (in TD), the recovered signal  $x_r(t) = x_s(t) * h_r(t)$

- Impulse train  $x_s(t)$ :  $x_s(t) = \sum_{n=-\infty}^{\infty} x_a(nT_s) \delta(t - nT_s)$

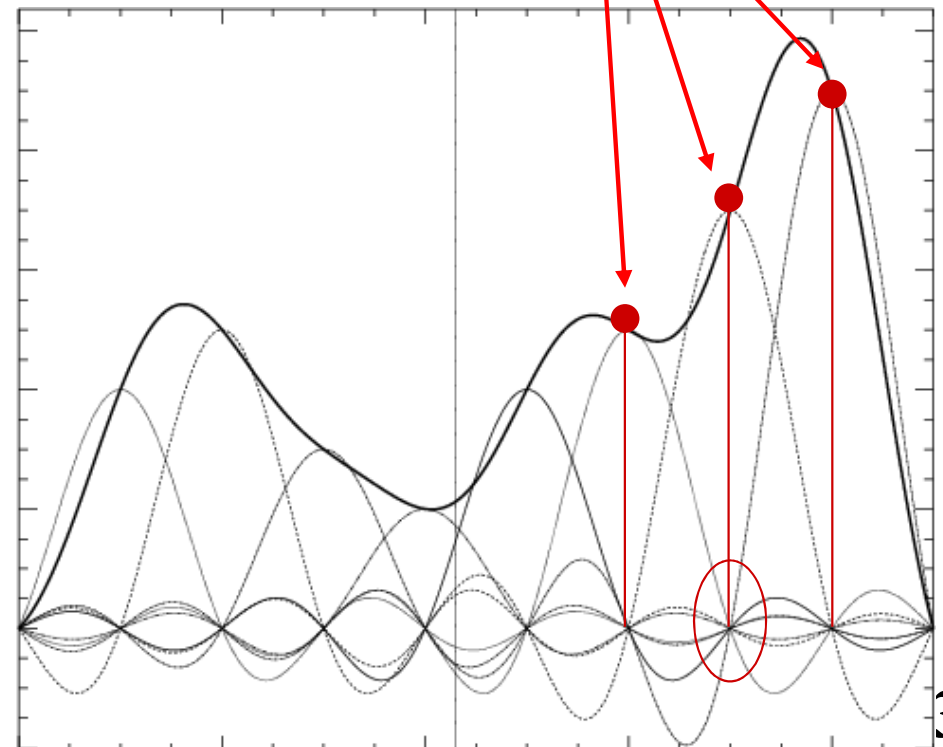


# Reconstruction filter in time domain

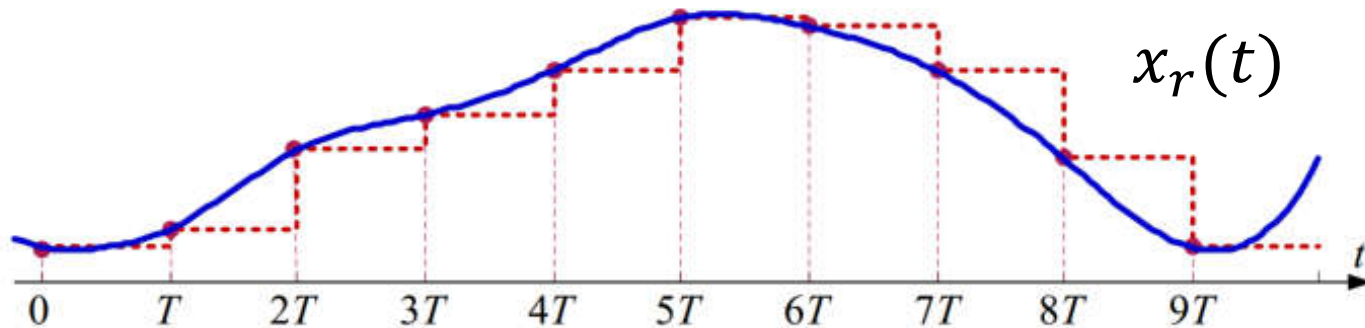
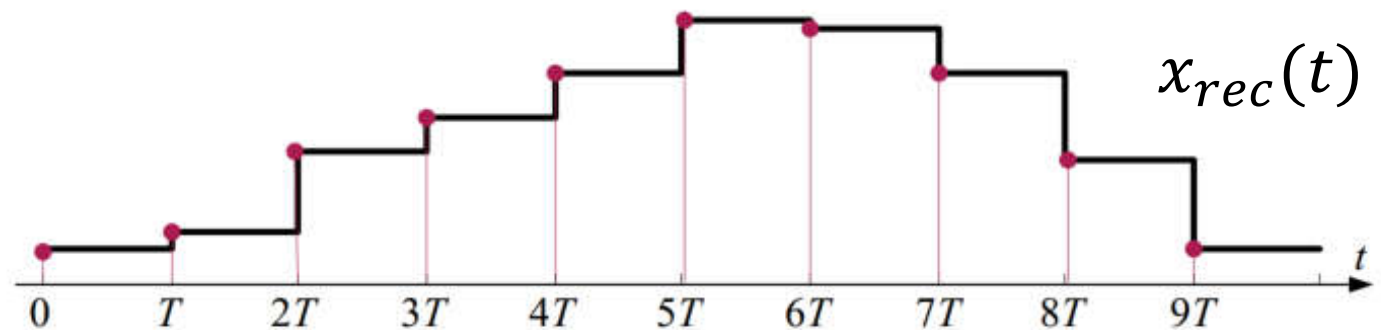
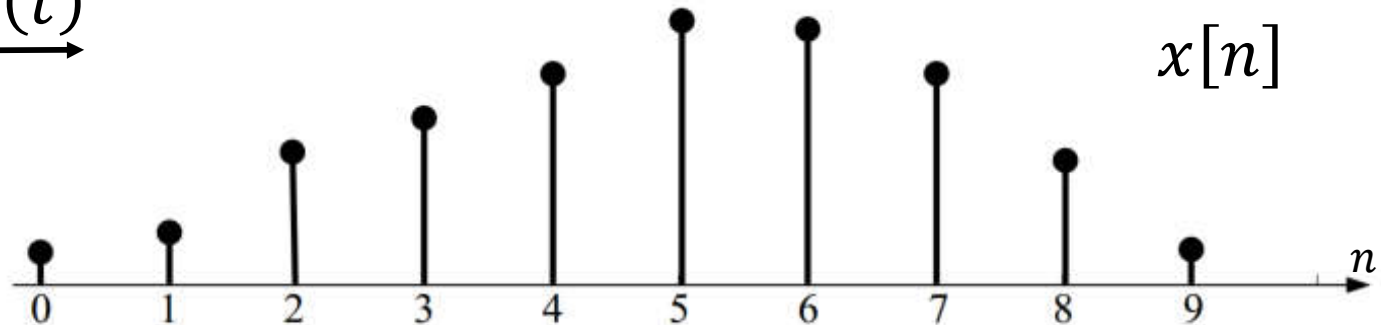
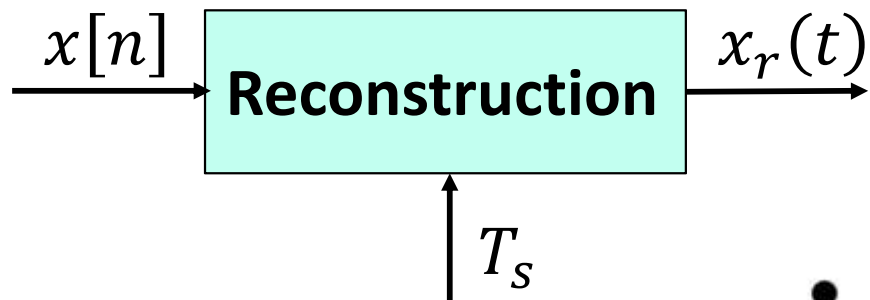
- Convolution between the discretized signal  $x_s(t)$  and the reconstruction lowpass filter  $h_r(t)$ :

$$x_r(t) = x_s(t) * h_r(t) = \sum_{n=-\infty}^{\infty} x_a(nT_s)h_r(t - nT_s) = \sum_{n=-\infty}^{\infty} x[n]\text{sinc}(t - nT_s)$$

- The values are interpolated as a linear combination of the time-shifted sinc functions
- The amplitudes are scaled according to the sample values at the center locations of the sinc (the interpolation functions)



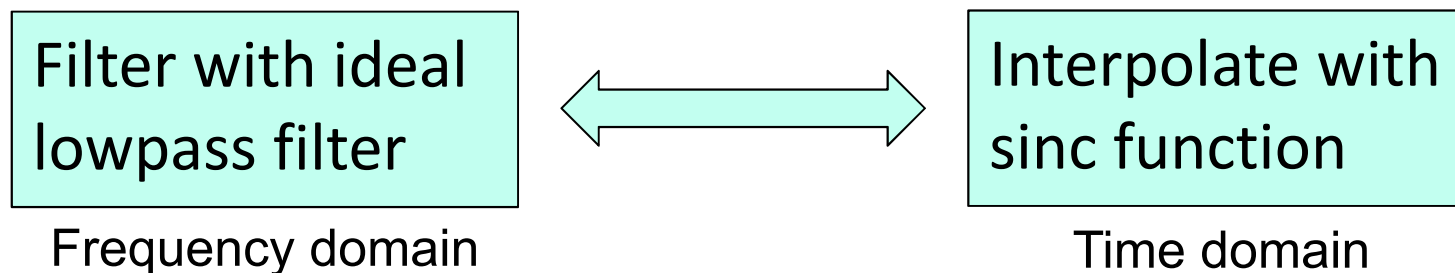
# Realization



## 3\_3 Wrap up

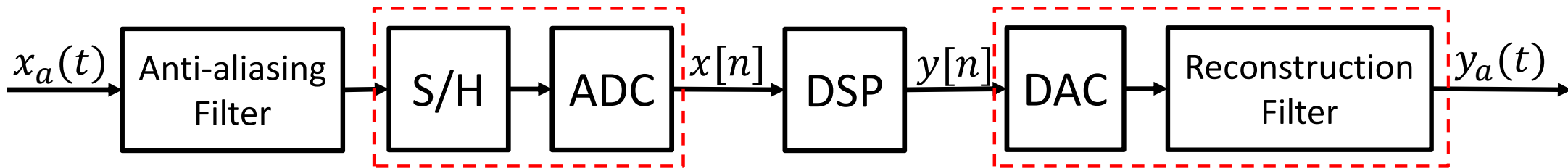
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- Continuous-time signal can be reconstructed from the discrete-time sequence;
- Reconstruction can be realized as
  - In time domain: interpolation;
  - In frequency domain: filtering.
- Ideal reconstruction:



# Chapter 3 Summary

- The whole process of A-D-A



- What is sampling? The time domain and frequency domain representation, including the equations and drawing;
- What is aliasing? What is the effect from aliasing and how to reduce it? Explain how the anti-aliasing filter works.
- What are interpolation and reconstruction? How do they work in time domain and frequency domain?

# Acronym

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- CT – Continuous time
- DT – Discrete time
- TD – Time domain
- FD – Frequency domain
- TransD – Transform domain