

# EI331 Signals and Systems

## Lecture 17

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# Contents

1. Sampling Theorem

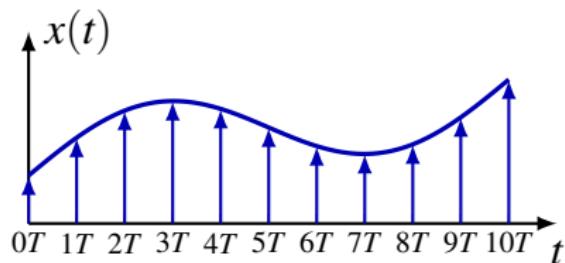
2. Zero-order Hold and Linear Interpolation

3. Aliasing

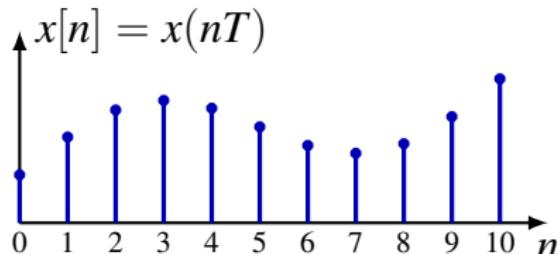
# Sampling CT Signals

Sampling converts CT signals to DT signals

CT signal



DT signal



$T$  = sampling period, uniform sampling most common

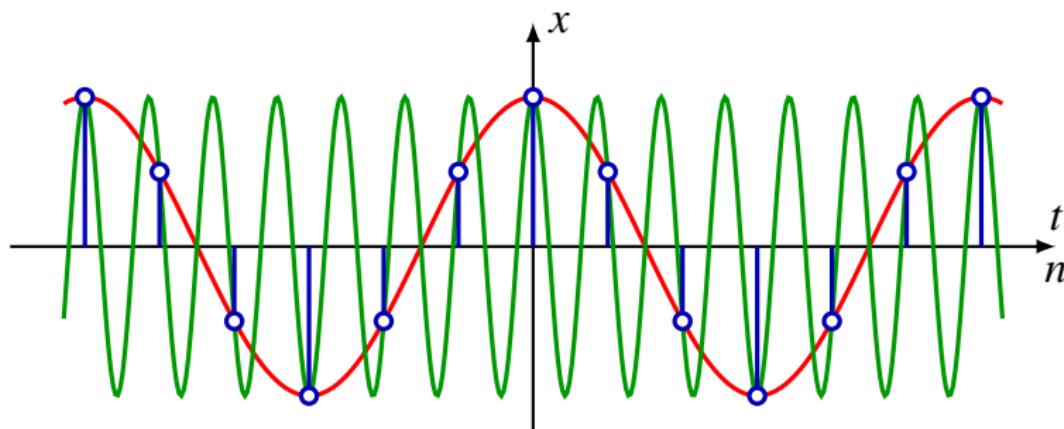
Allows use of digital electronics to process, record, transmit, store, and retrieve CT signals

- MP3, digital camera, printer

# Sampling CT Signals

Sampling loses information, different signals may have same samples

- $x_1(t) = \cos(\frac{\pi}{3}t)$ ,  $x_2(t) = \cos(\frac{7\pi}{3}t)$ , **different**
- $x_1[n] = \cos(\frac{\pi}{3}n) = x_2[n] = \cos(\frac{7\pi}{3}n)$ , **identical**

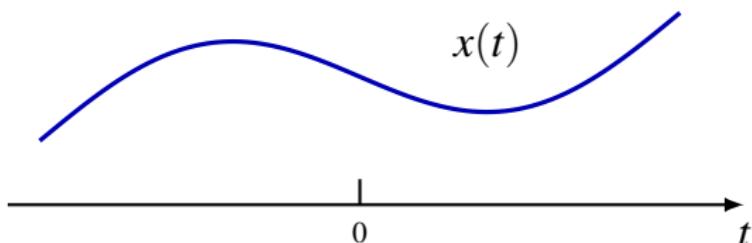


Under what conditions can we recover signal from samples?

# Impulse-train Sampling

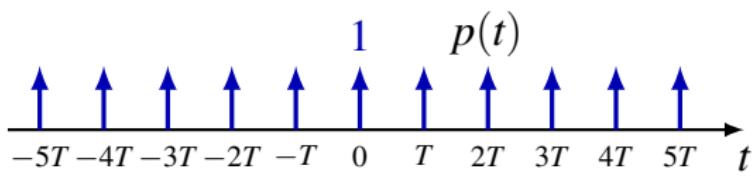
Time domain

CT signal  $x(t)$



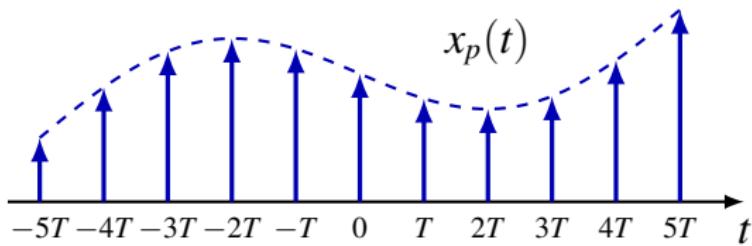
impulse train

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



$$x_p(t) = x(t)p(t)$$

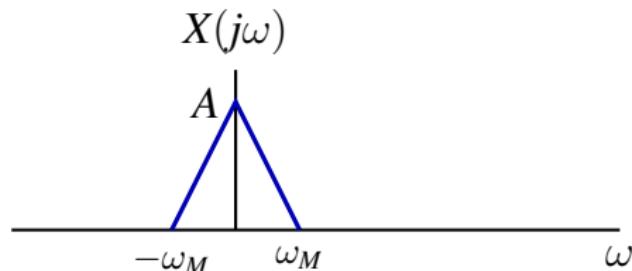
$$= \sum_{n=-\infty}^{\infty} x(nT) \delta(t - nT)$$



# Impulse-train Sampling

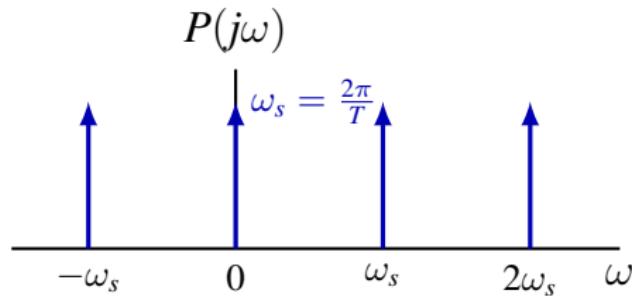
Frequency domain

CT signal  $X(j\omega)$



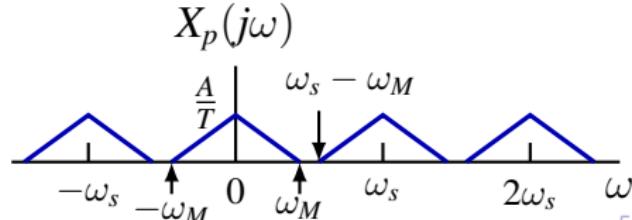
impulse train

$$P(j\omega) = \omega_s \sum_{k=-\infty}^{\infty} \delta(\omega - k\omega_s)$$



$$X_p(j\omega) = \frac{1}{2\pi} (X * P)(\omega)$$

$$= \frac{1}{T} \sum_{k=-\infty}^{\infty} X(j(\omega - k\omega_s))$$



# Impulse-train Sampling

Frequency domain

band-limited CT signal

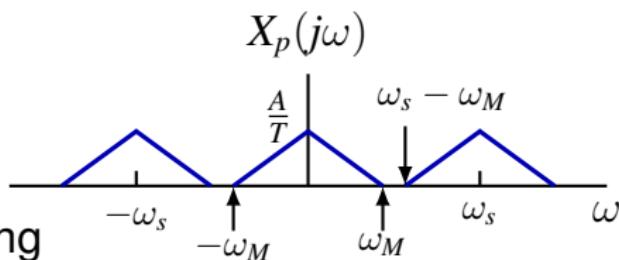
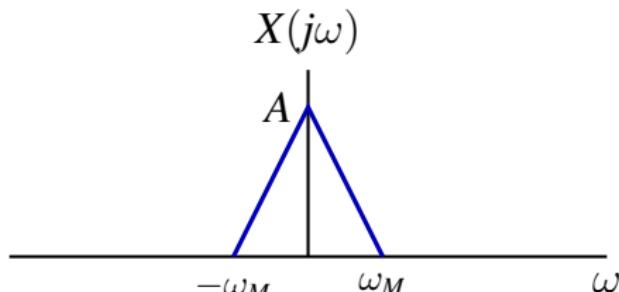
$$X(j\omega) = 0 \text{ for } |\omega| > \omega_M$$

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

Case 1:  $\omega_s > 2\omega_M$

**no overlap** between replicas

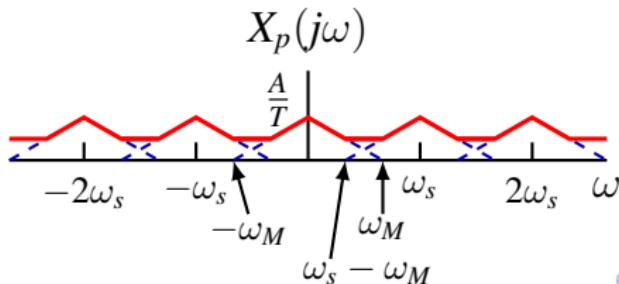
**can** recover  $X$  by lowpass filtering



Case 2:  $\omega_s < 2\omega_M$

replicas **overlap**

**cannot** recover  $X$



# Sampling Theorem

Band-limited CT signal  $x(t)$  whose spectrum  $X(j\omega) = 0$  for  $|\omega| > \omega_M$  is uniquely determined by its samples  $x(nT), n \in \mathbb{Z}$  if

$$\omega_s \triangleq \frac{2\pi}{T} > 2\omega_M, \quad 2\omega_M \text{ called Nyquist rate}$$

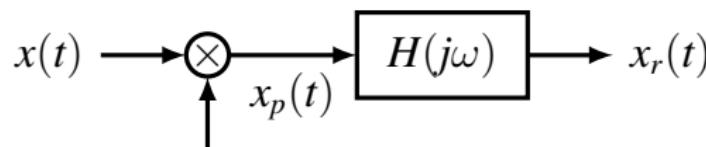
Given  $\{x(nT) : n \in \mathbb{Z}\}$ ,  $x(t)$  can be reconstructed as follows

1. construct  $x_p(t) = \sum_{n=-\infty}^{\infty} x(nT)\delta(t - nT)$
2. send  $x_p$  through lowpass filter with gain  $T$  and cutoff frequency  $\omega_c \in (\omega_M, \omega_s - \omega_M)$ , i.e.

$$H(j\omega) = T[u(\omega + \omega_c) - u(\omega - \omega_c)]$$

3. filter output  $x_r(t)$  with  $X_r(j\omega) = X_p(j\omega)H(j\omega)$  is same as  $x(t)$

# Reconstruction in Frequency Domain



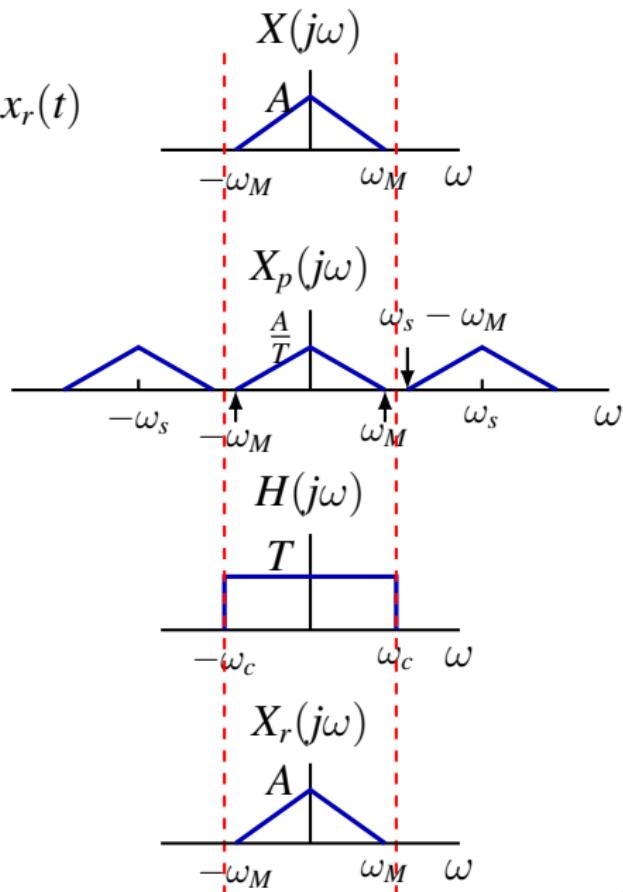
$$p(t) = \sum_{n=-\infty}^{\infty} \delta(t - nT)$$

- Nyquist frequency  $\omega_M$
- Nyquist rate  $2\omega_M$

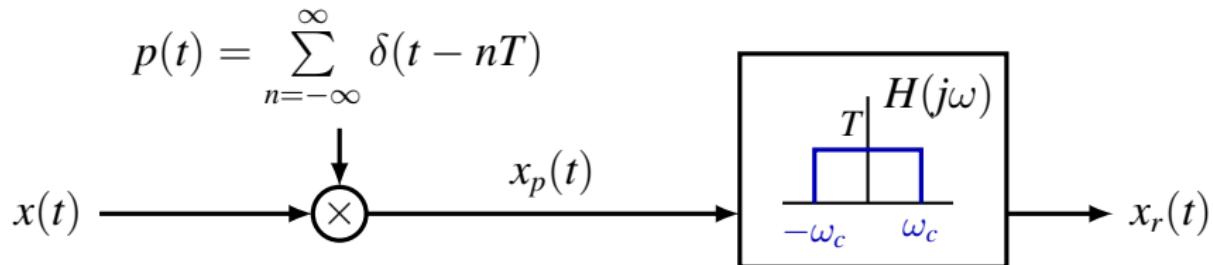
Lowpass filter

- gain  $T$
- cutoff frequency

$$\omega_M < \omega_c < \omega_s - \omega_M$$



# Reconstruction in Time Domain



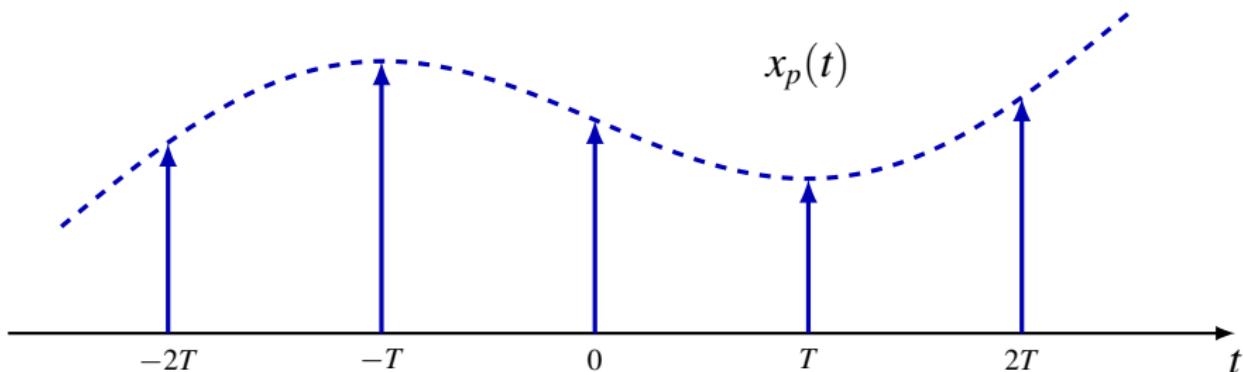
Impulse response of lowpass filter

$$h(t) = \frac{T \sin(\omega_c t)}{\pi t}$$

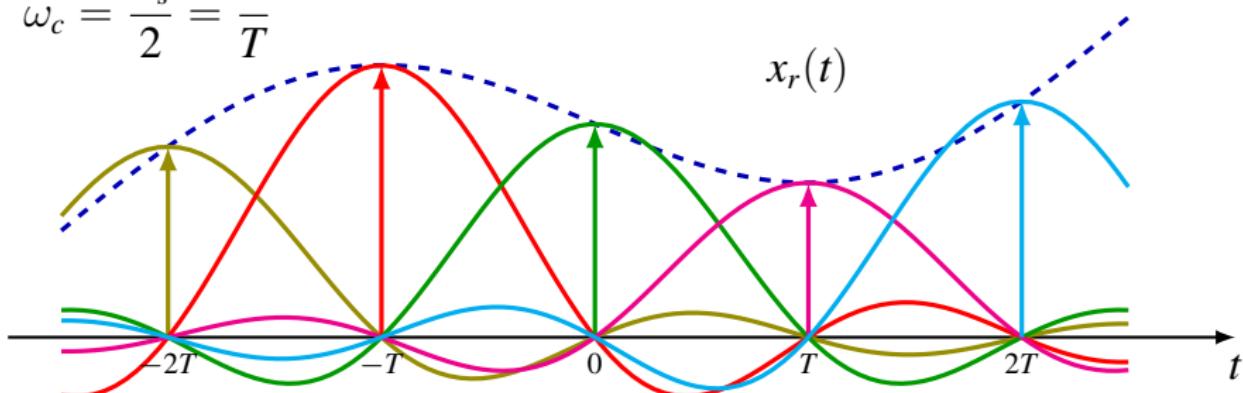
$x$  recovered by **band-limited interpolation** using sinc function

$$x(t) = x_r(t) = (x_p * h)(t) = \sum_{n=-\infty}^{\infty} x(nT) \frac{T \sin(\omega_c(t - nT))}{\pi(t - nT)}$$

# Reconstruction in Time Domain



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



# Reconstruction in Time Domain

Setting  $\omega_c = \frac{\omega_s}{2}$  yields Whittaker-Shannon interpolation formula

$$x_r(t) = \sum_{n=-\infty}^{\infty} x(nT) \operatorname{sinc}\left(\frac{t-nT}{T}\right), \quad \text{where } \operatorname{sinc}(t) = \frac{\sin(\pi t)}{\pi t}$$

Since

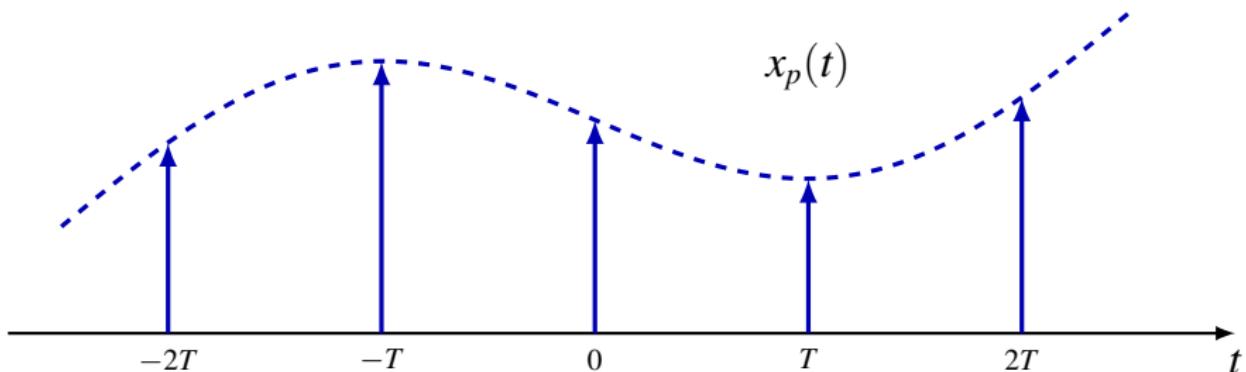
$$\operatorname{sinc}\left(\frac{t-nT}{T}\right) \xleftrightarrow{\mathcal{F}} Te^{-jnT\omega} [u(\omega + \frac{\pi}{T}) - u(\omega - \frac{\pi}{T})]$$

Parseval's identity (or multiplication property) implies

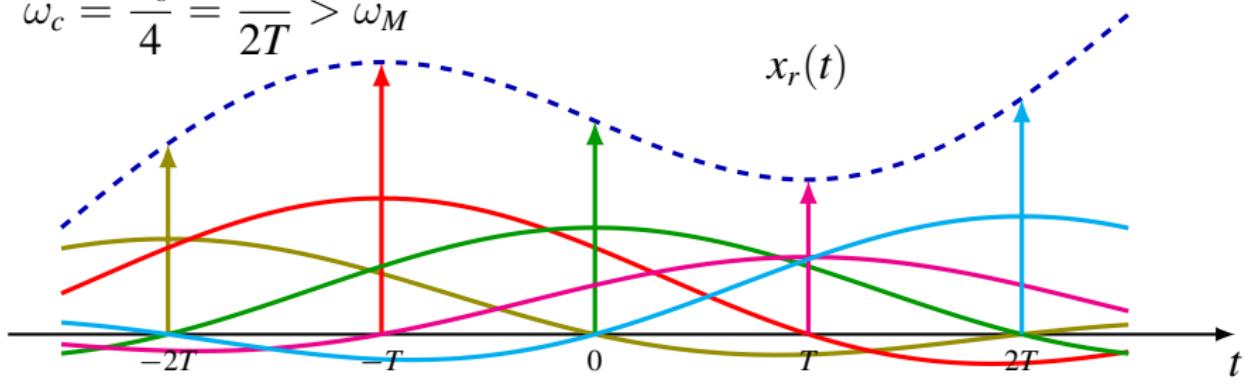
$$\int_{\mathbb{R}} \operatorname{sinc}\left(\frac{t-nT}{T}\right) \operatorname{sinc}\left(\frac{t-mT}{T}\right) dt = \frac{T^2}{2\pi} \int_{-\frac{\pi}{T}}^{\frac{\pi}{T}} e^{j(m-n)T\omega} d\omega = T\delta[n-m]$$

Whittaker-Shannon formula is orthogonal expansion

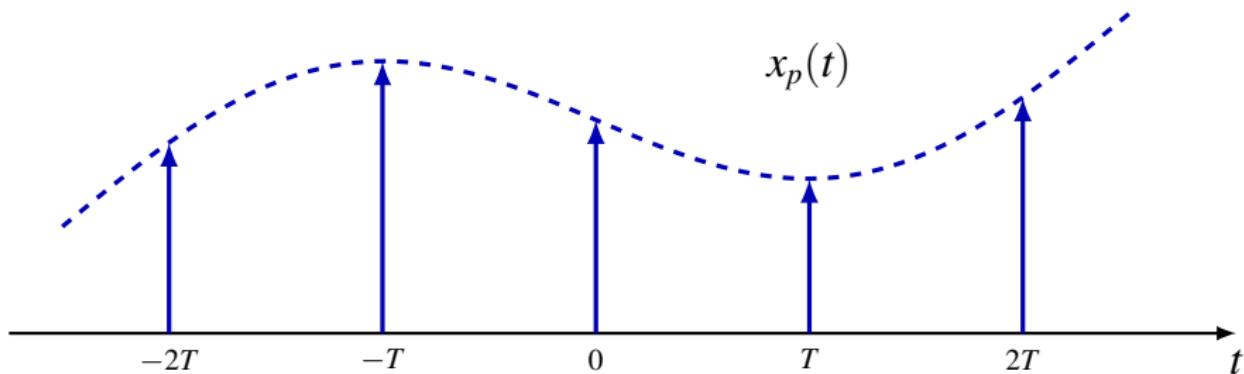
# Reconstruction in Time Domain



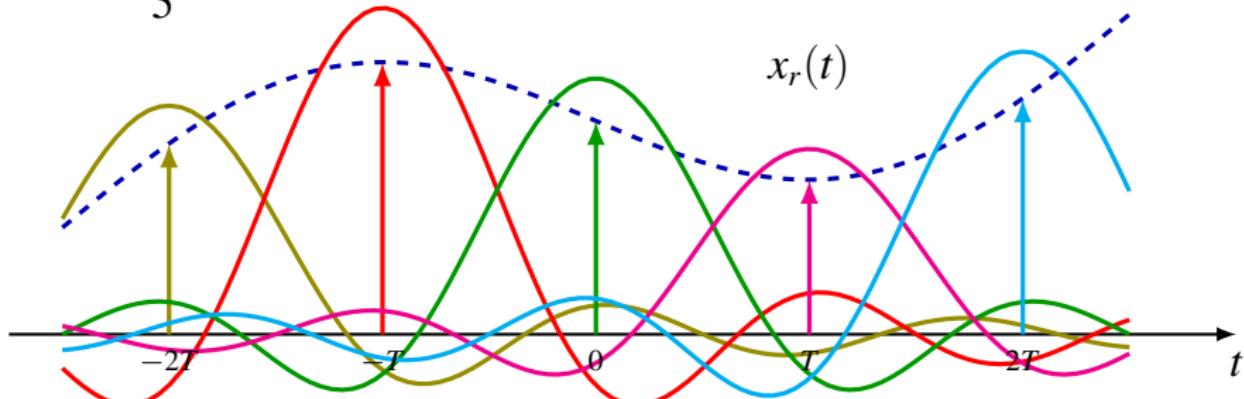
$$\omega_c = \frac{\omega_s}{4} = \frac{\pi}{2T} > \omega_M$$



# Reconstruction in Time Domain



$$\omega_c = \frac{3\omega_s}{5} < \omega_s - \omega_M$$



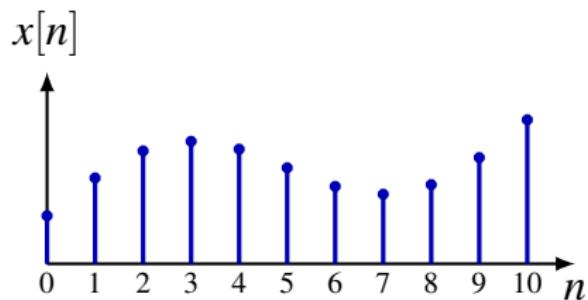
# Contents

1. Sampling Theorem

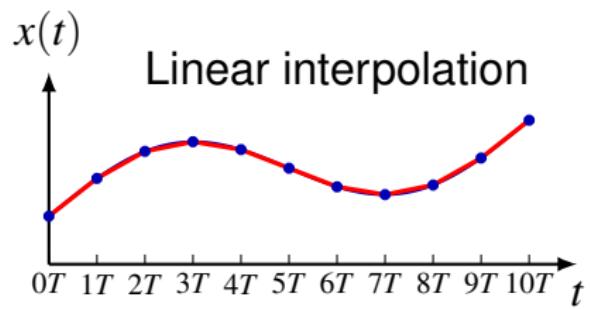
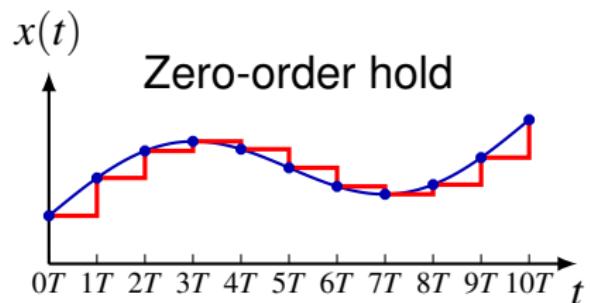
2. Zero-order Hold and Linear Interpolation

3. Aliasing

# Other Interpolation Methods

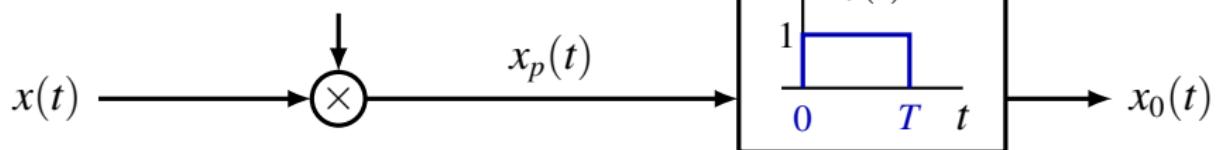


$T$  = sampling period



# Zero-order Hold

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

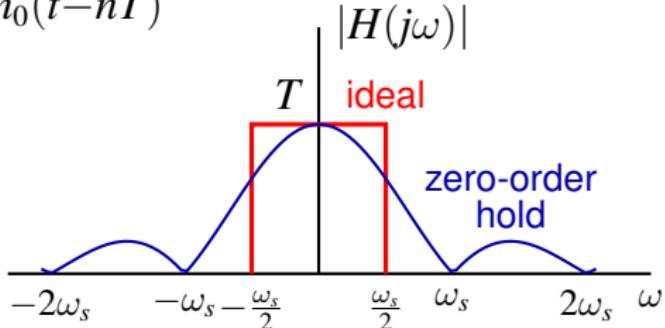


Reconstructed signal

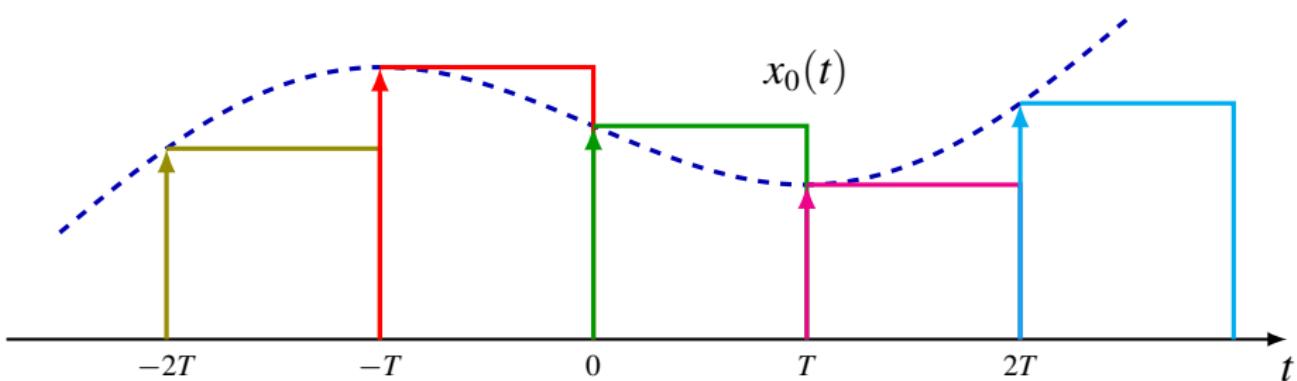
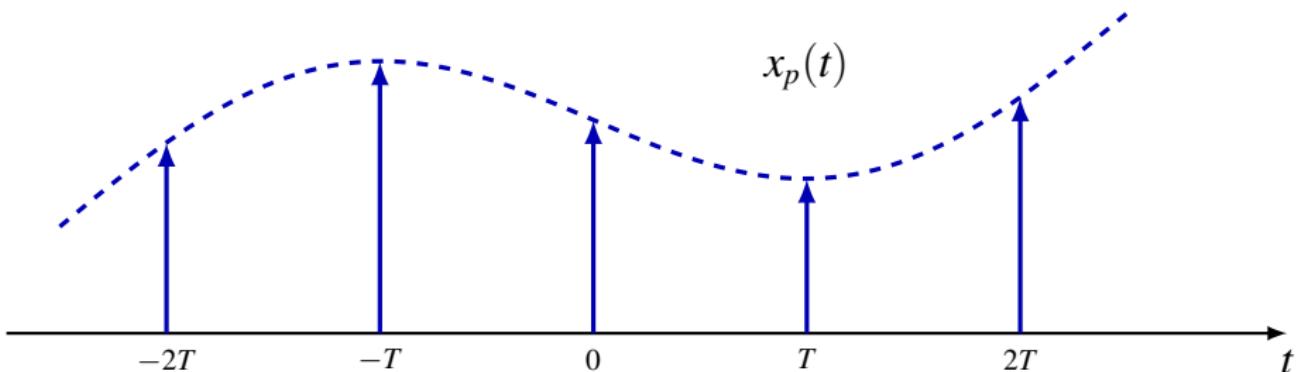
$$x_0(t) = (x_p * h_0)(t) = \sum_{n=-\infty}^{\infty} x(nT)h_0(t-nT)$$

Zero-order hold filter

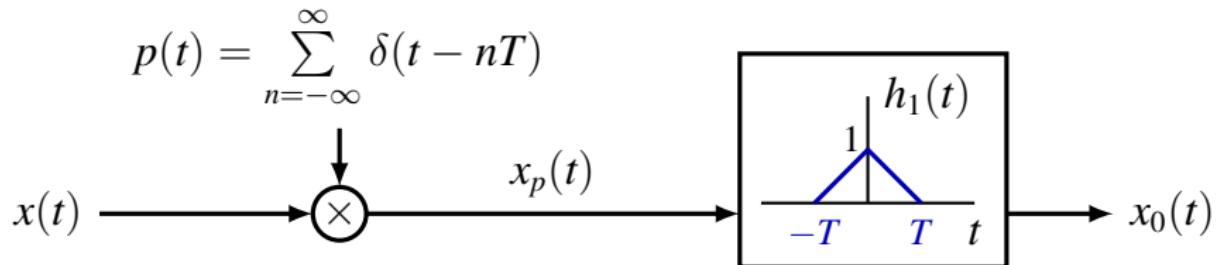
$$H_0(j\omega) = e^{-j\omega T/2} \frac{2 \sin(\omega T/2)}{\omega}$$



# Zero-order Hold



# Linear Interpolation (First-order Hold)

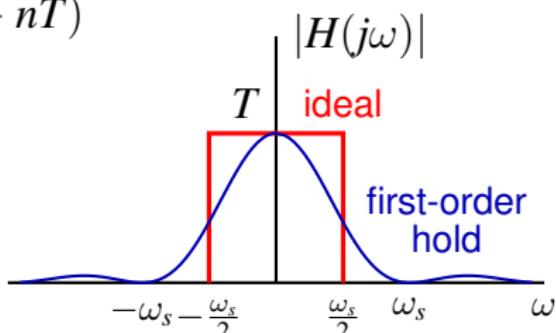


Reconstructed signal

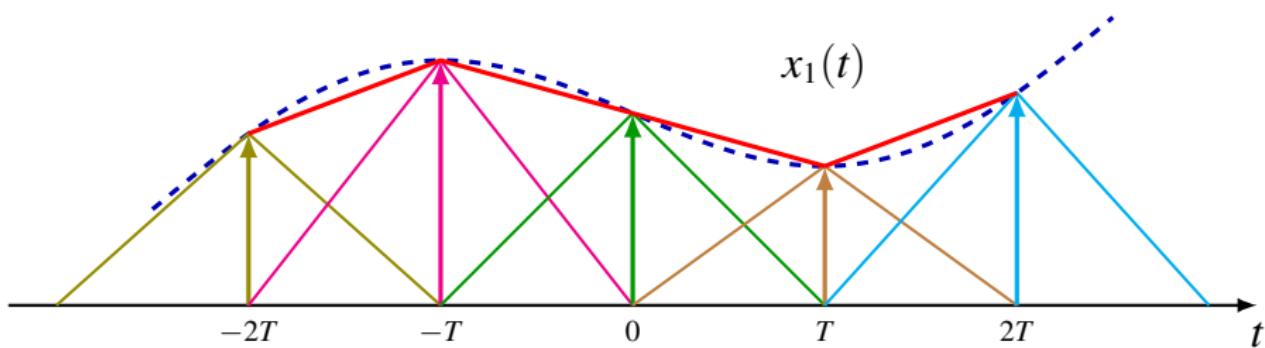
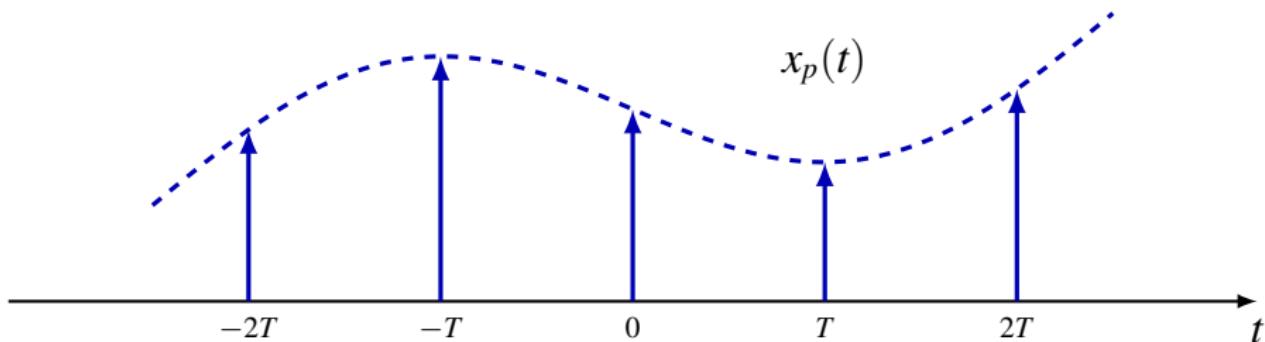
$$x_1(t) = (x_p * h_1)(t) = \sum_{n=-\infty}^{\infty} x(nT)h_1(t - nT)$$

First-order hold filter

$$H_1(j\omega) = \frac{1}{T} \left[ \frac{\sin(\omega T/2)}{\omega/2} \right]^2$$



# Linear Interpolation



# Contents

1. Sampling Theorem

2. Zero-order Hold and Linear Interpolation

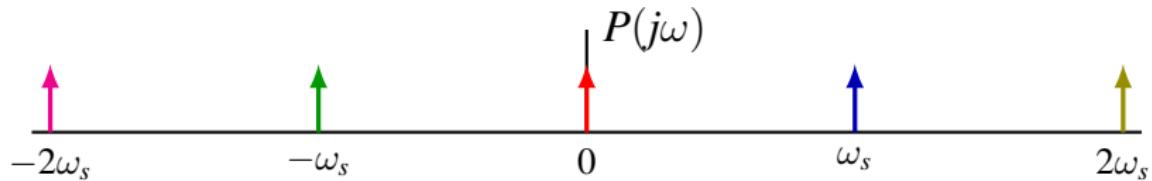
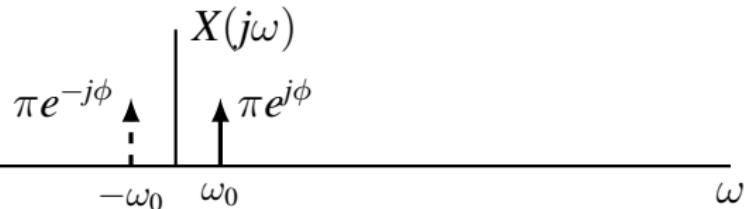
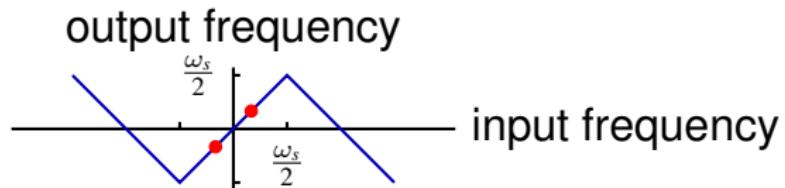
3. Aliasing

# Aliasing

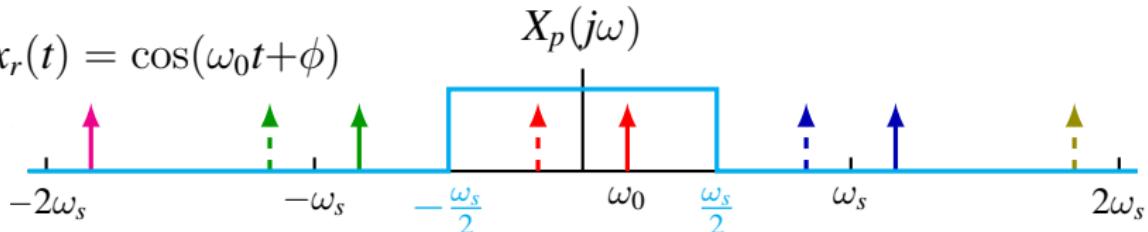
Aliasing wraps frequencies

$$x(t) = \cos(\omega_0 t + \phi)$$

$$\omega_0 < \frac{1}{2}\omega_s$$



$$x_r(t) = \cos(\omega_0 t + \phi)$$

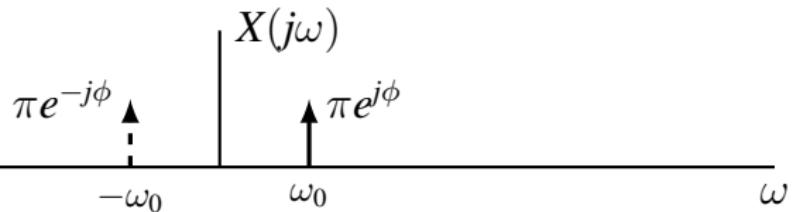
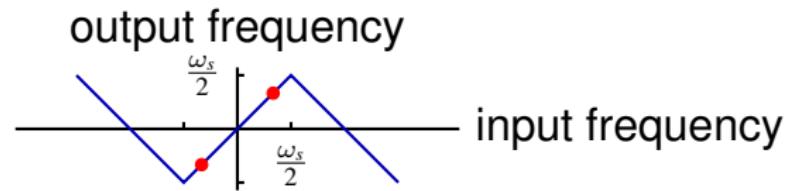


# Aliasing

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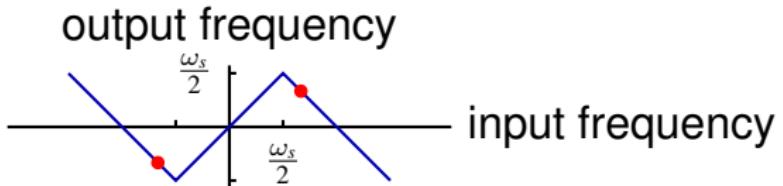


$$x_r(t) = \cos(\omega_0 t + \phi)$$



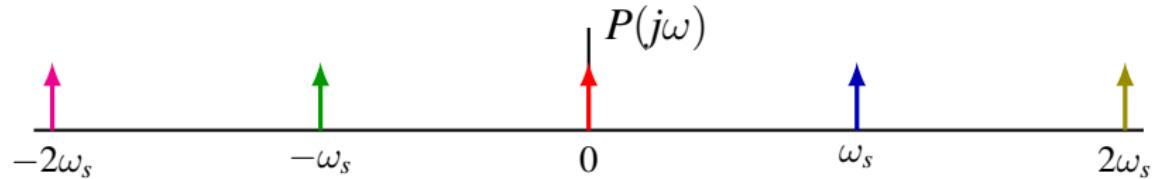
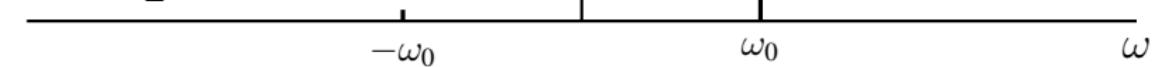
# Aliasing

Aliasing wraps frequencies

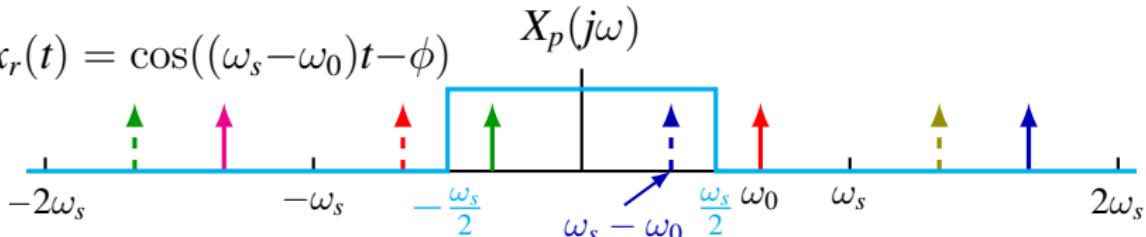


$$x(t) = \cos(\omega_0 t + \phi)$$

$$\omega_0 > \frac{1}{2}\omega_s$$



$$x_r(t) = \cos((\omega_s - \omega_0)t - \phi)$$

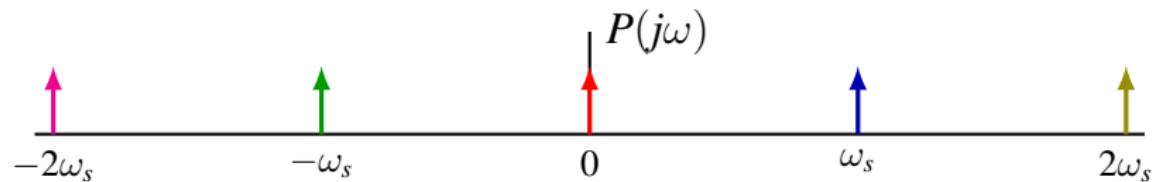
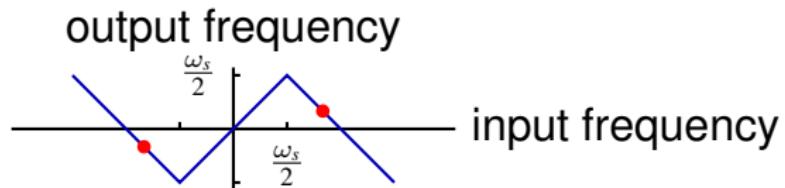


# Aliasing

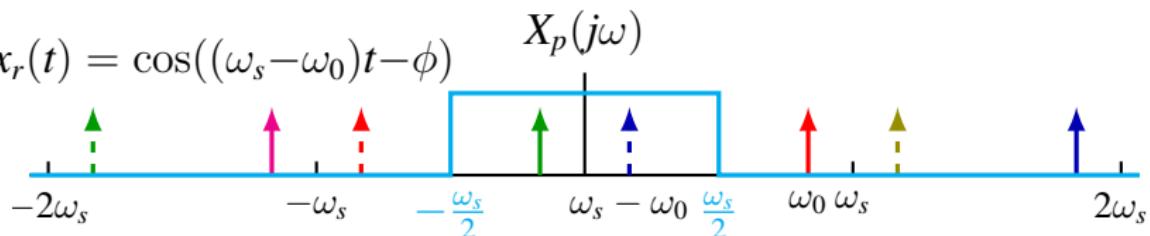
Aliasing wraps frequencies

$$x(t) = \cos(\omega_0 t + \phi)$$

$$\omega_0 > \frac{1}{2}\omega_s \quad \pi e^{-j\phi}$$

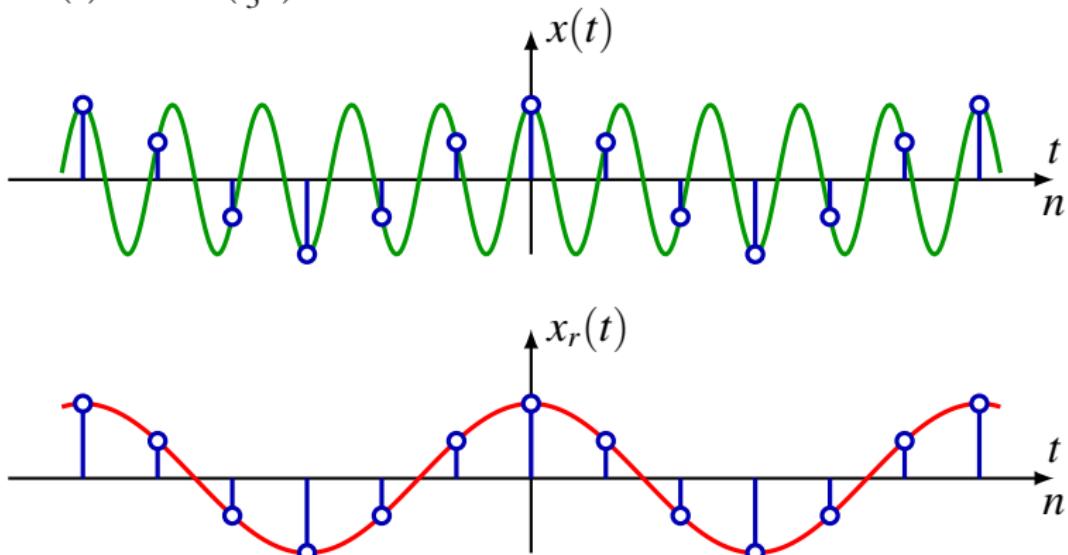


$$x_r(t) = \cos((\omega_s - \omega_0)t - \phi)$$



# Aliasing

- $x(t) = \cos(\frac{5\pi}{3}t)$ ,  $x[n] = \cos(\frac{5\pi}{3}n) = \cos(\frac{\pi}{3}n)$
- $x_r(t) = \cos(\frac{\pi}{3}t)$



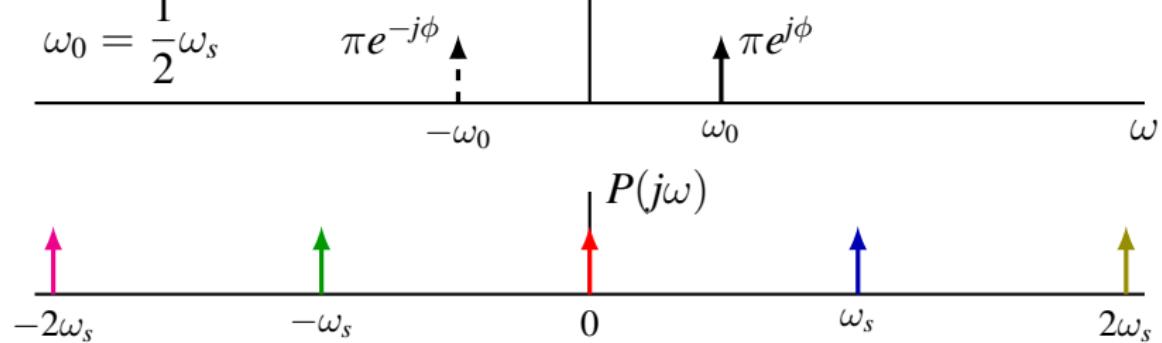
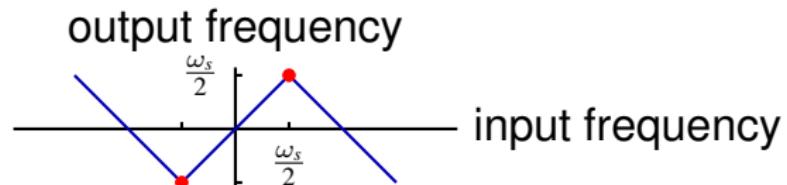
**Example.** In movies, wheels often appear to rotate more slowly than they actually do and even in wrong direction

# Aliasing

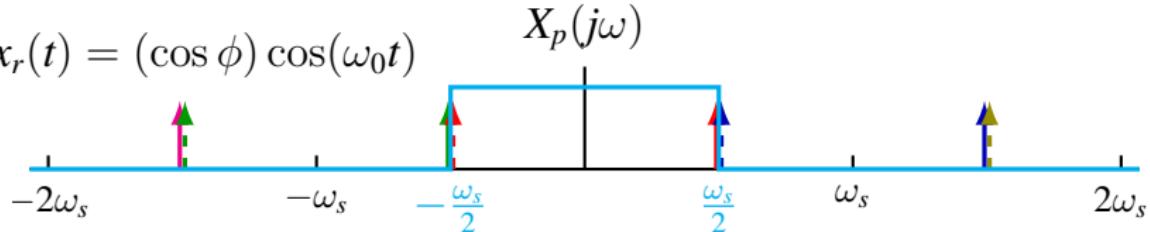
Aliasing wraps frequencies

$$x(t) = \cos(\omega_0 t + \phi)$$

$$\omega_0 = \frac{1}{2}\omega_s$$

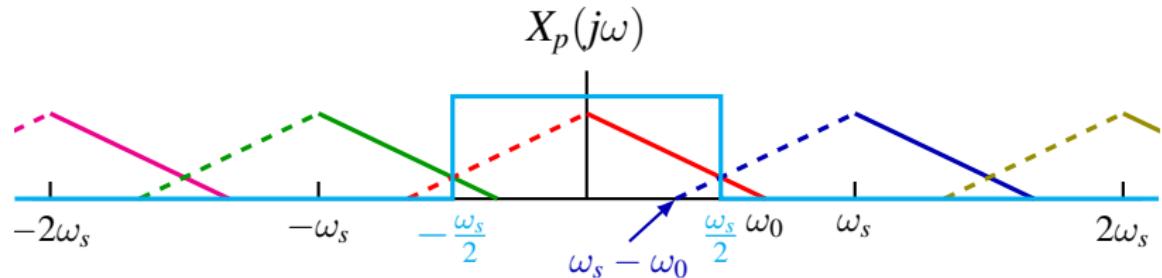
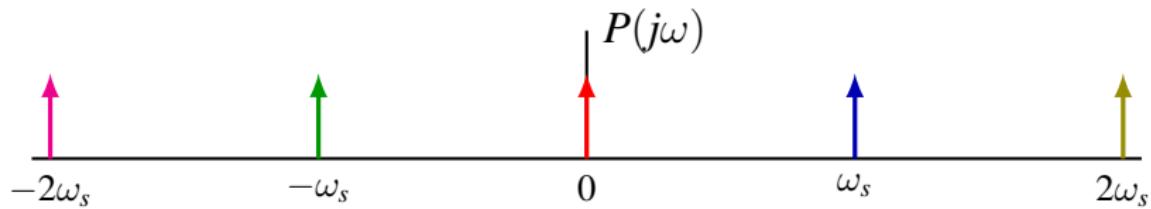
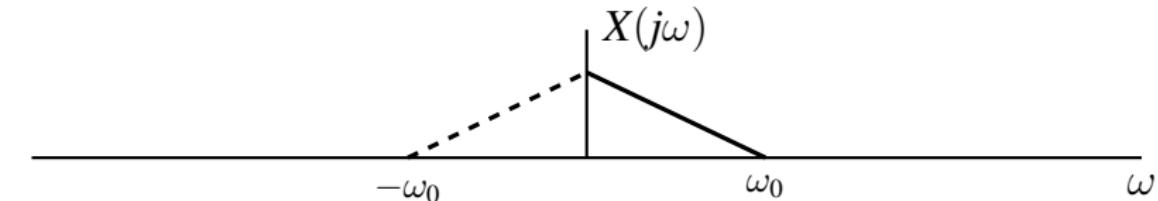


$$x_r(t) = (\cos \phi) \cos(\omega_0 t)$$



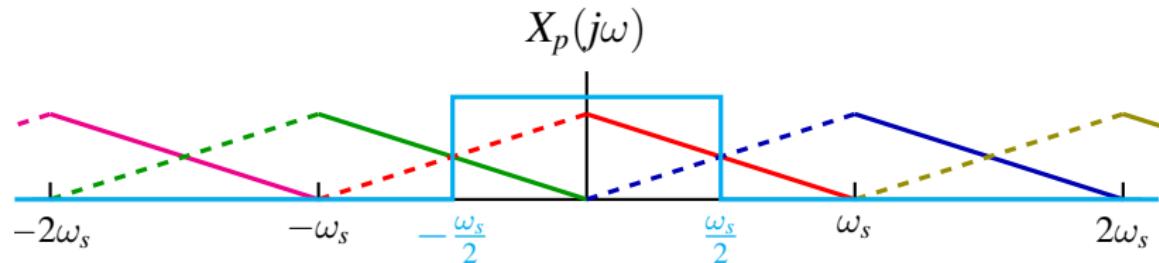
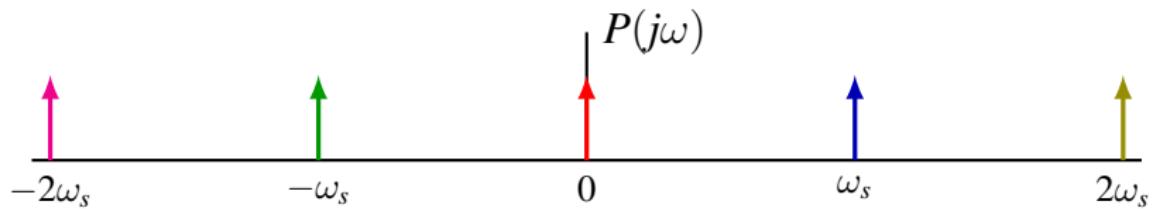
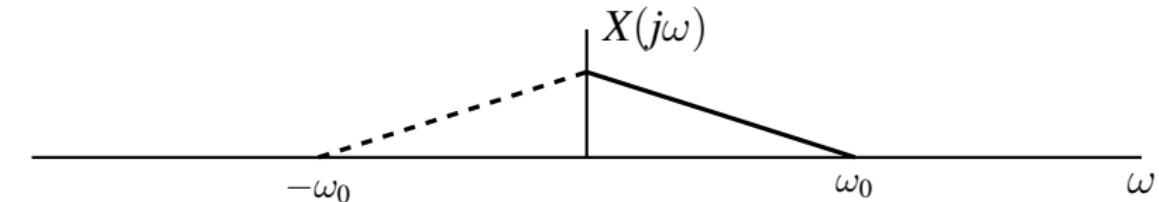
# Aliasing

Aliasing for more complex signals also wraps frequencies



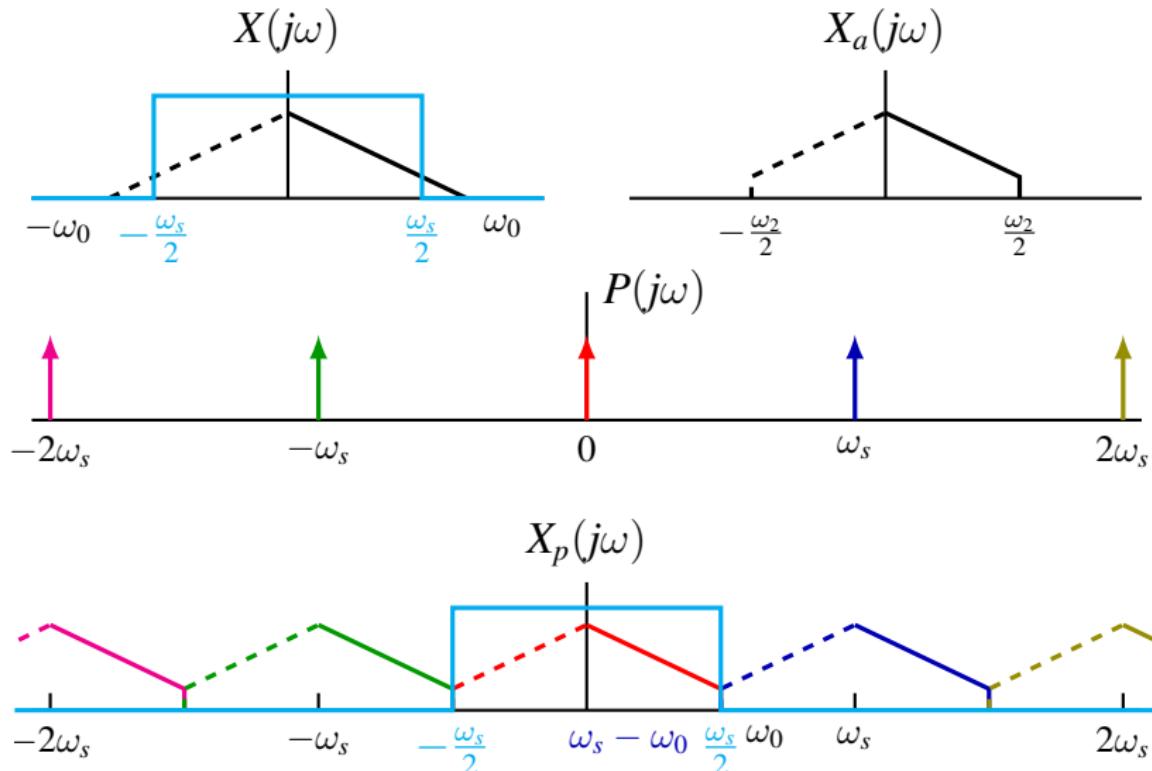
# Aliasing

Aliasing increases as sampling rate decreases



# Anti-aliasing Filter

Filter out frequencies above  $\frac{\omega_s}{2}$  before sampling



# Anti-aliasing Filter

Filter out frequencies above  $\frac{\omega_s}{2}$  before sampling

