

Signals and Systems

#11: Sampling and reconstruction

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Some of the images in this presentation are from "Signals and Systems", A. V. Oppenheim, A. S. Willsky, and S. H. Nawab, 2nd ed. Pearson.

Outline

- ...
- #06 Basic properties of the Fourier transform
- #07 Convolution property
- #08 Multiplication property
- #09 First-order systems
- #10 Second-order systems
- **#11 Sampling and reconstruction**
- #12 Communication systems

Sampling

We want to represent a signal $x(t)$ through its samples

T is the sampling interval

$\omega_s = \frac{2\pi}{T}$ is the sampling frequency

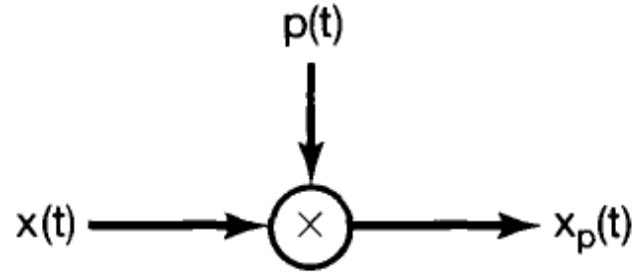
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$$p(t) = \sum_{n=-\infty}^{\infty} \delta(t - nT)$$

$$\begin{aligned} x_p(t) &= x(t) \cdot p(t) = \sum_{n=-\infty}^{\infty} x(t) \delta(t - nT) \\ &= \sum_{n=-\infty}^{\infty} x(nT) \delta(t - nT) \end{aligned}$$



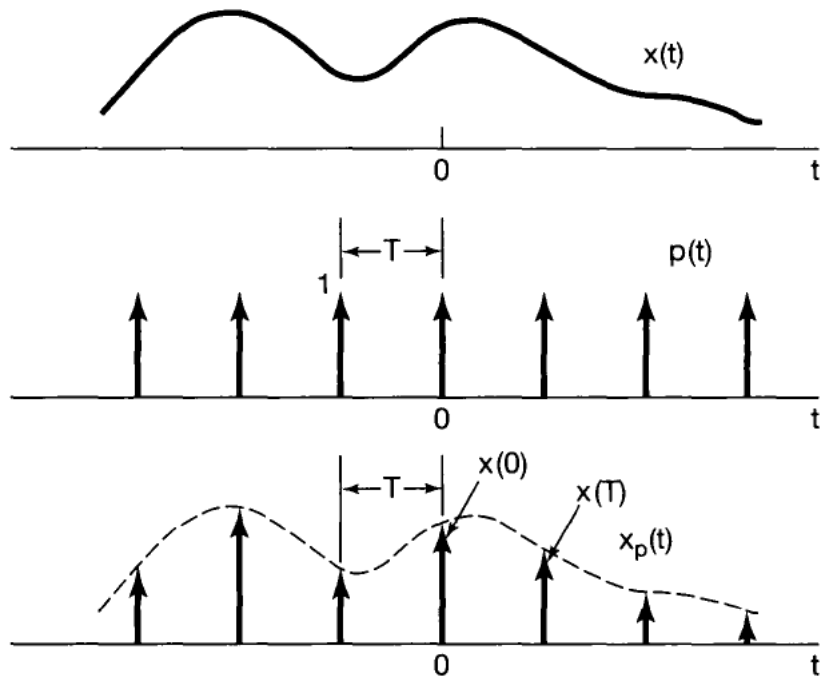
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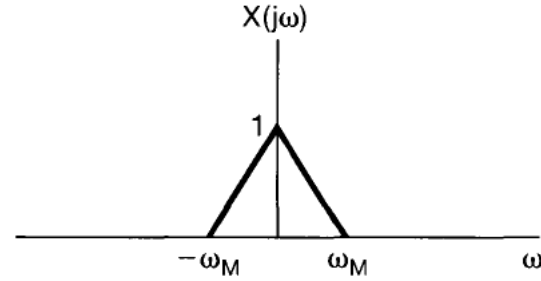
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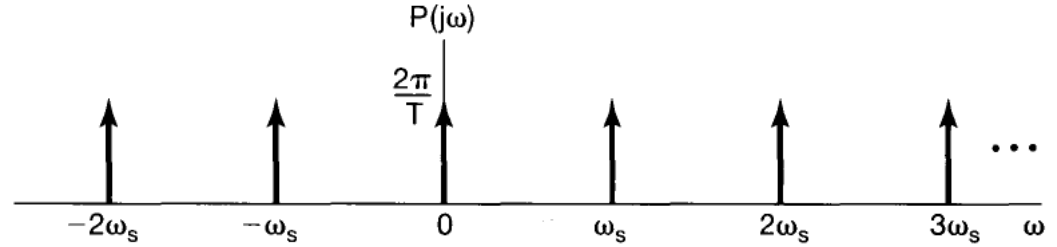
In the frequency domain:

$$X_p(j\omega) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(j\theta)P(j(\omega - \theta))d\theta = \dots$$



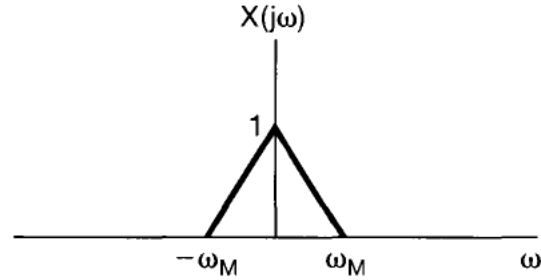
recall

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



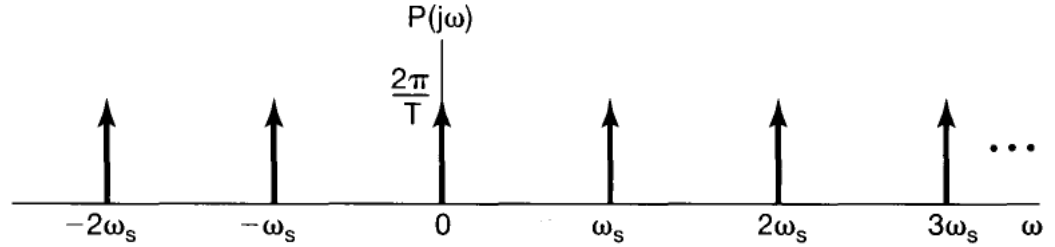
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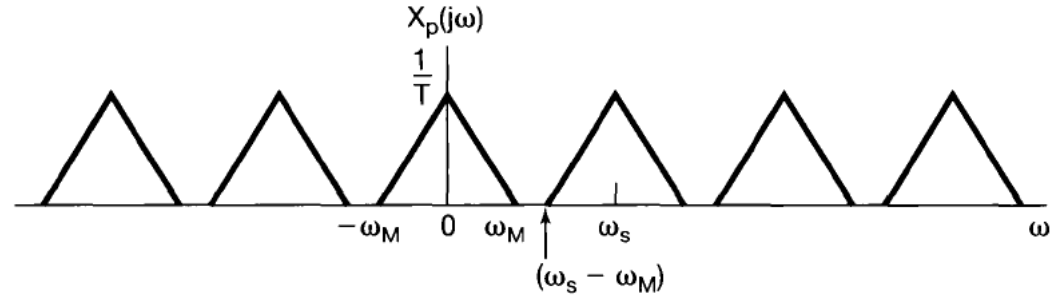


recall

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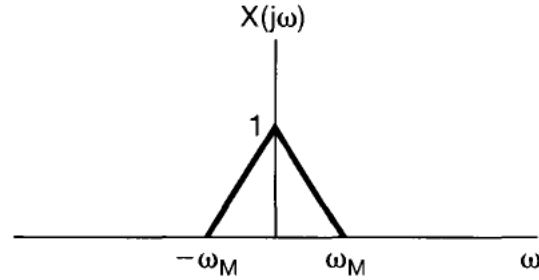


$$\dots = \frac{1}{T} \sum_{k=-\infty}^{\infty} X(j(\omega - k\omega_s))$$



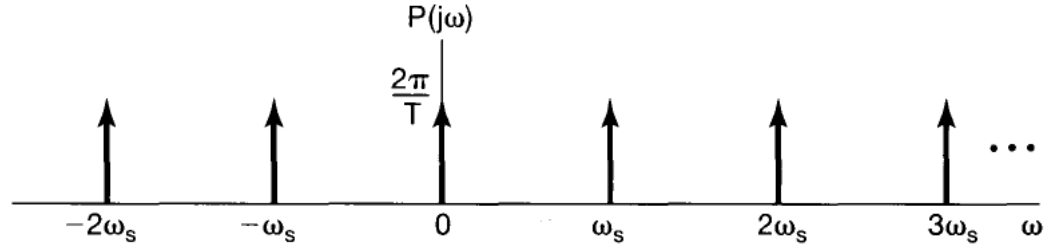
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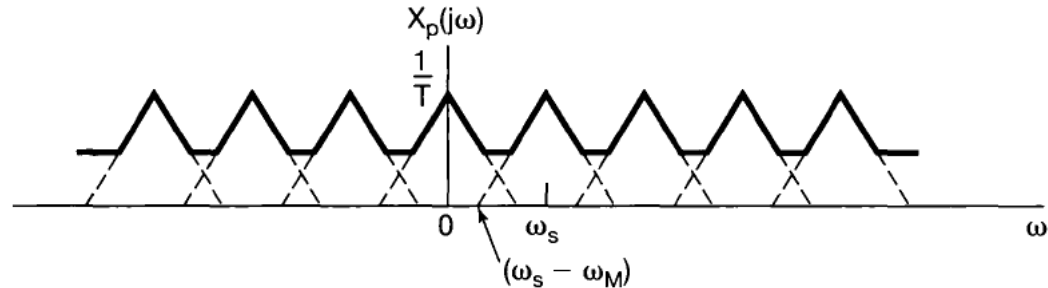


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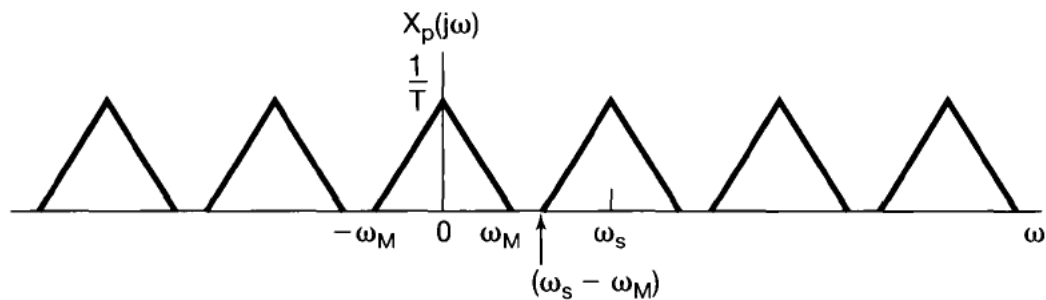


Sampling theorem:

if $\omega_s > 2\omega_M$

no overlap

perfect reconstruction possible

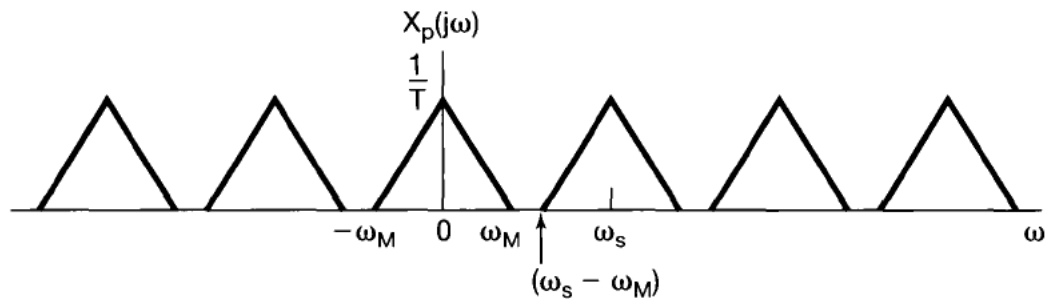


Sampling theorem:

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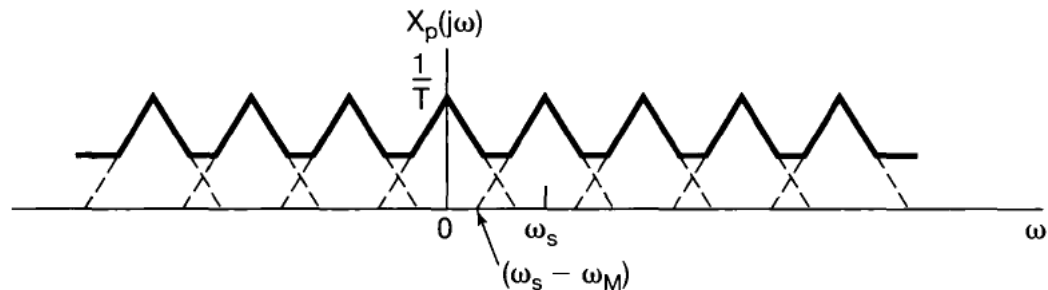
perfect reconstruction possible



if $\omega_s < 2\omega_M$

overlap (*aliasing*)

perfect reconstruction not possible



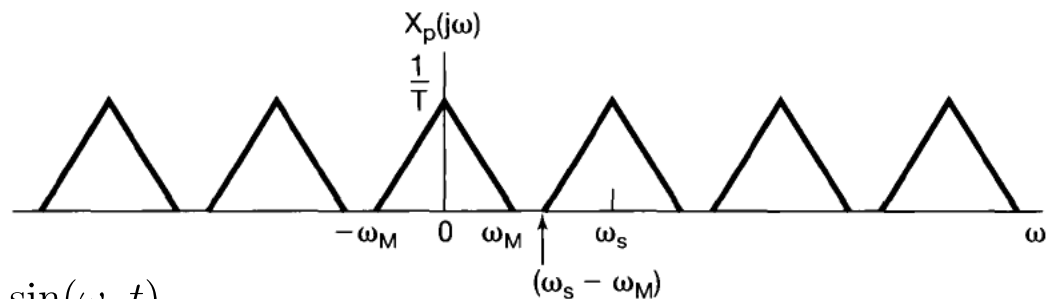
the frequency $2\omega_M$ is denoted the **Nyquist rate**

Reconstruction

Band-limited interpolation

Interpolate the sampled signal

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

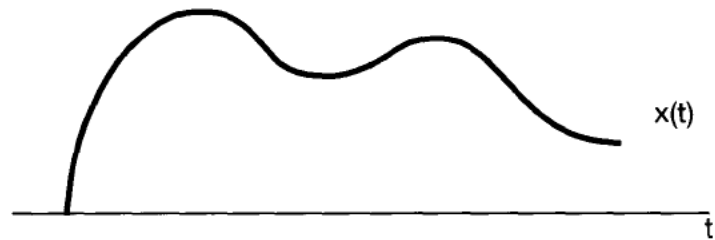


through lowpass filtering $h(t) = \frac{\omega_c T \sin(\omega_c t)}{\pi \omega_c t}$

$$x_r(t) = x_p(t) * h(t)$$

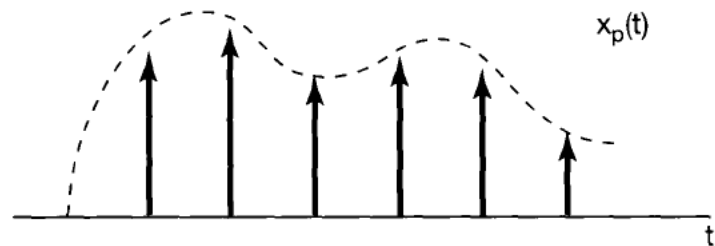
$$= \sum_{n=-\infty}^{\infty} x(nT) \frac{\omega_c T \sin(\omega_c(t - nT))}{\pi \omega_c(t - nT)}$$

Band-limited interpolation



Interpolate the sampled signal

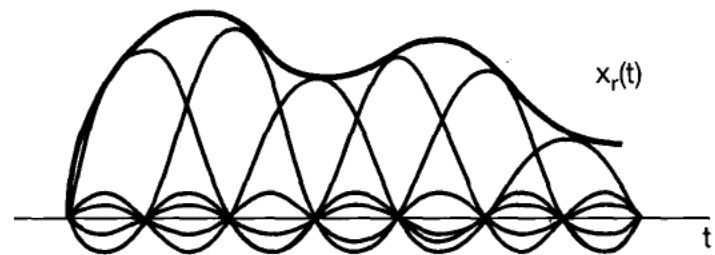
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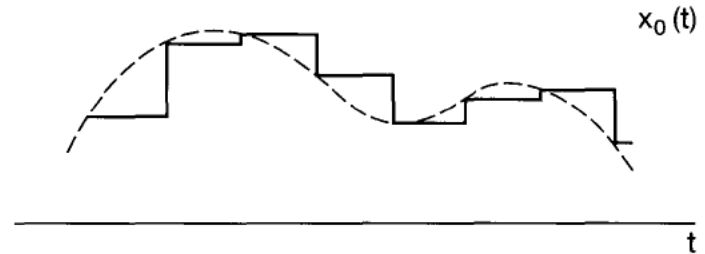
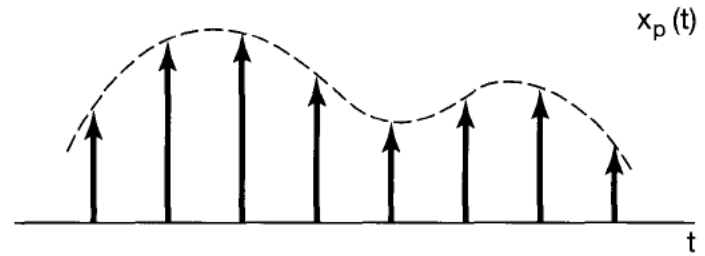
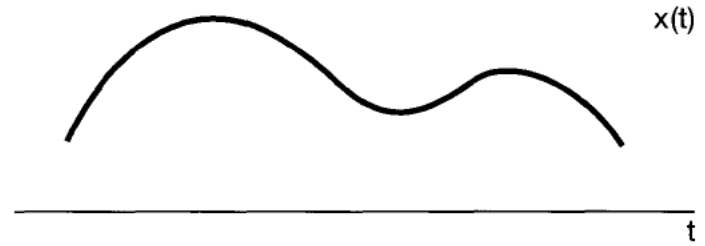
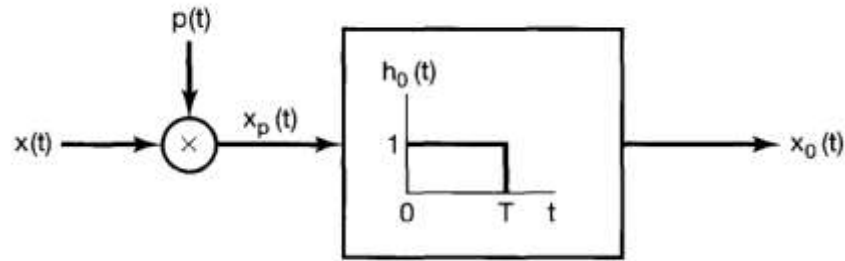
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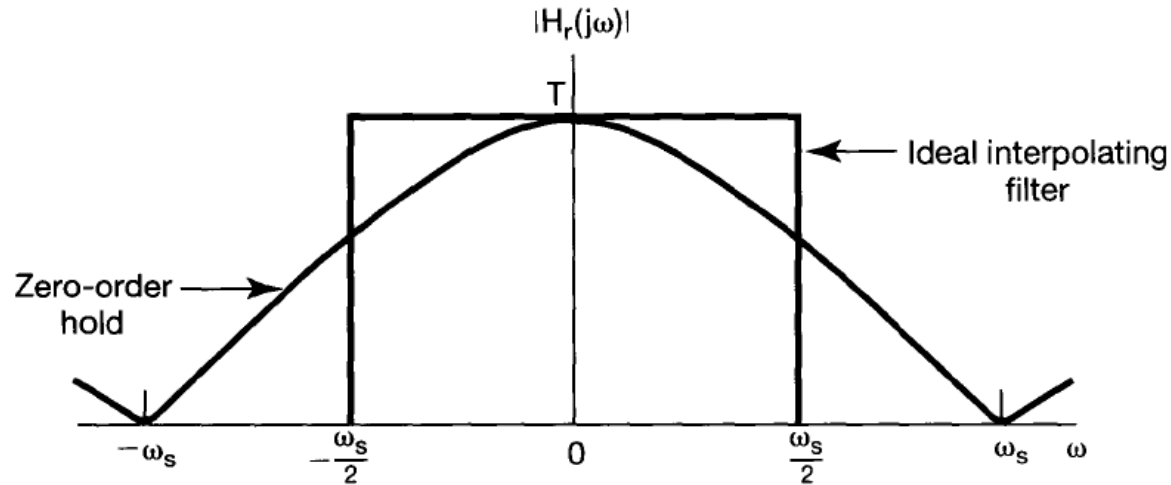
Zero-order hold

in time domain:



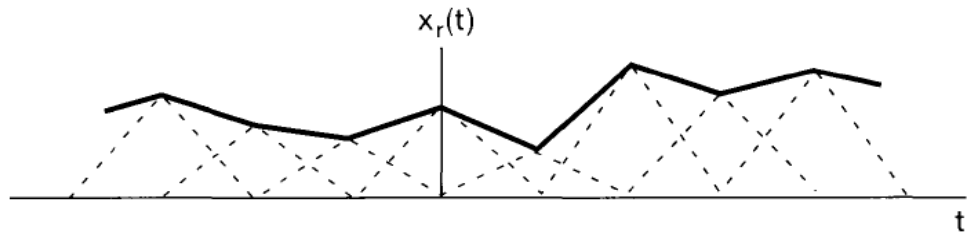
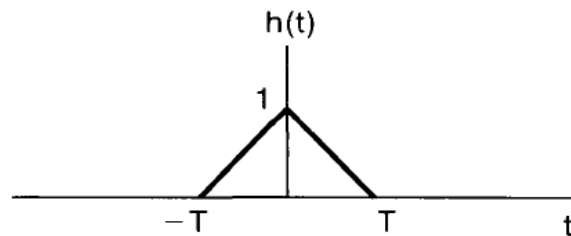
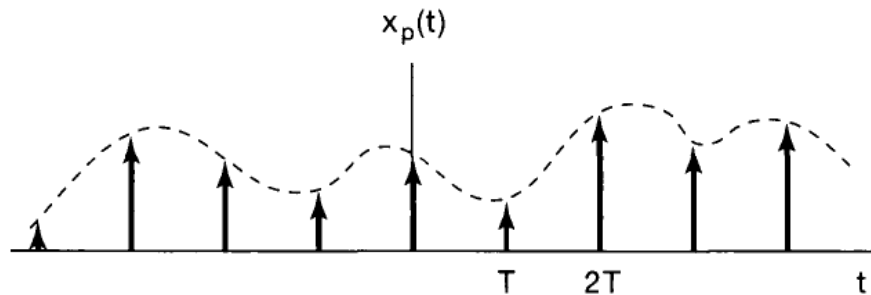
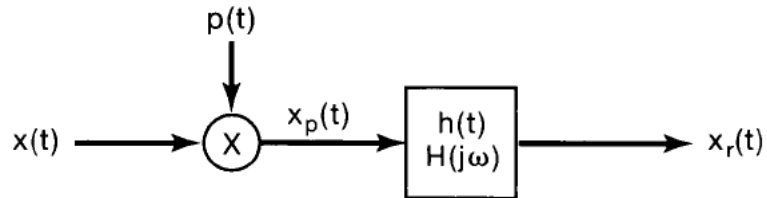
Zero-order hold

in frequency domain:
$$H_0(j\omega) = e^{-j\omega T/2} \left[\frac{2 \sin(\omega T/2)}{\omega} \right]$$



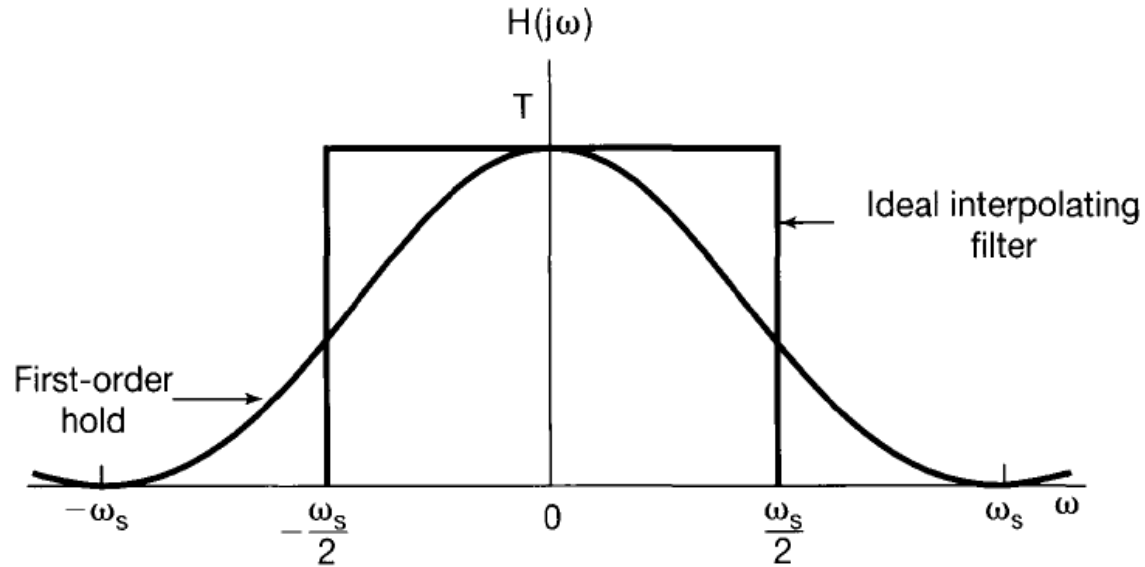
First-order hold (linear interpolation)

in time domain:



First-order hold (linear interpolation)

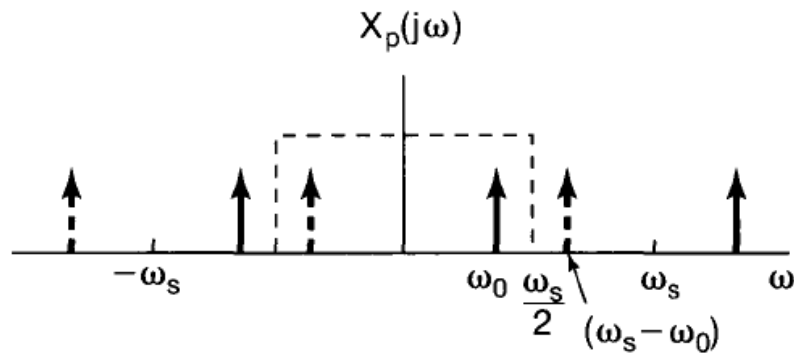
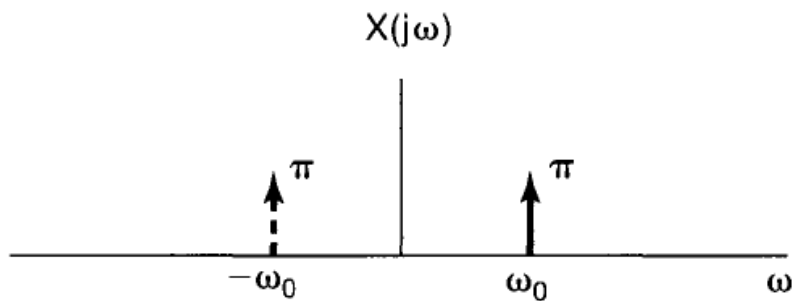
in frequency domain:
$$H(j\omega) = \frac{1}{T} \left[\frac{\sin(\omega T/2)}{\omega/2} \right]^2$$



Example

Example: $x(t) = \cos \omega_0 t \xleftrightarrow{\mathcal{F}} X(j\omega) = \pi \delta(\omega - \omega_0) + \pi \delta(\omega + \omega_0)$

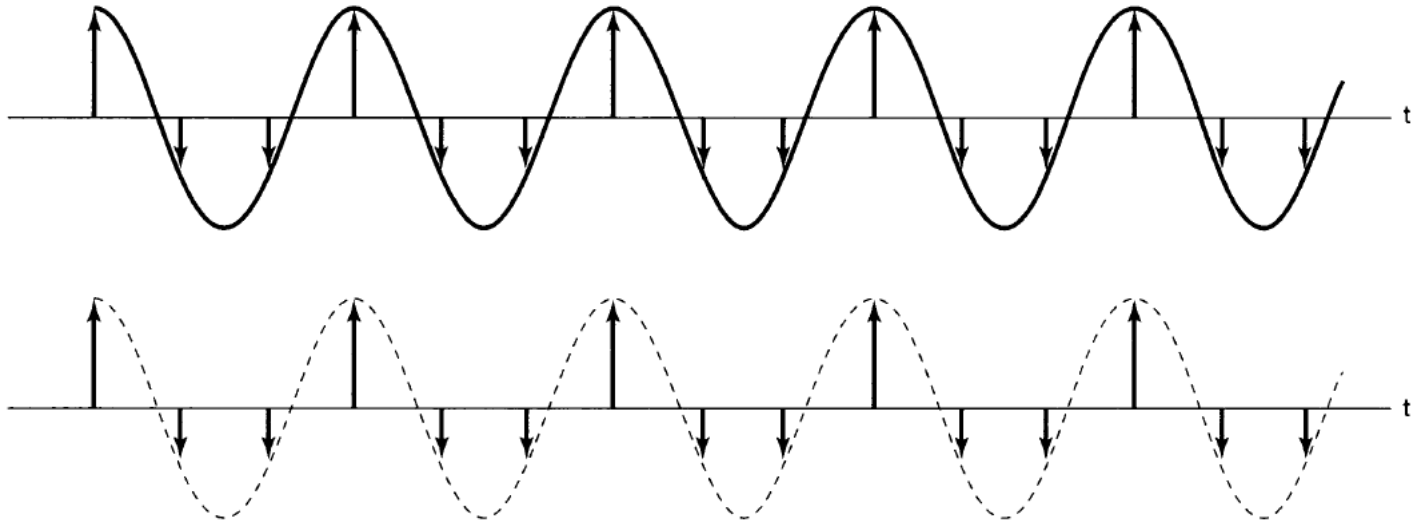
$\omega_s = 3\omega_0$ (frequency domain)



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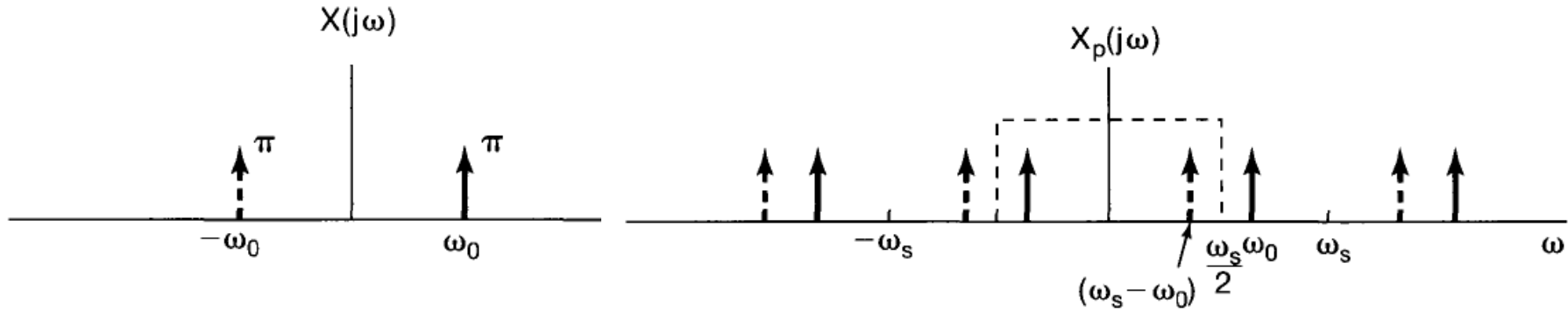
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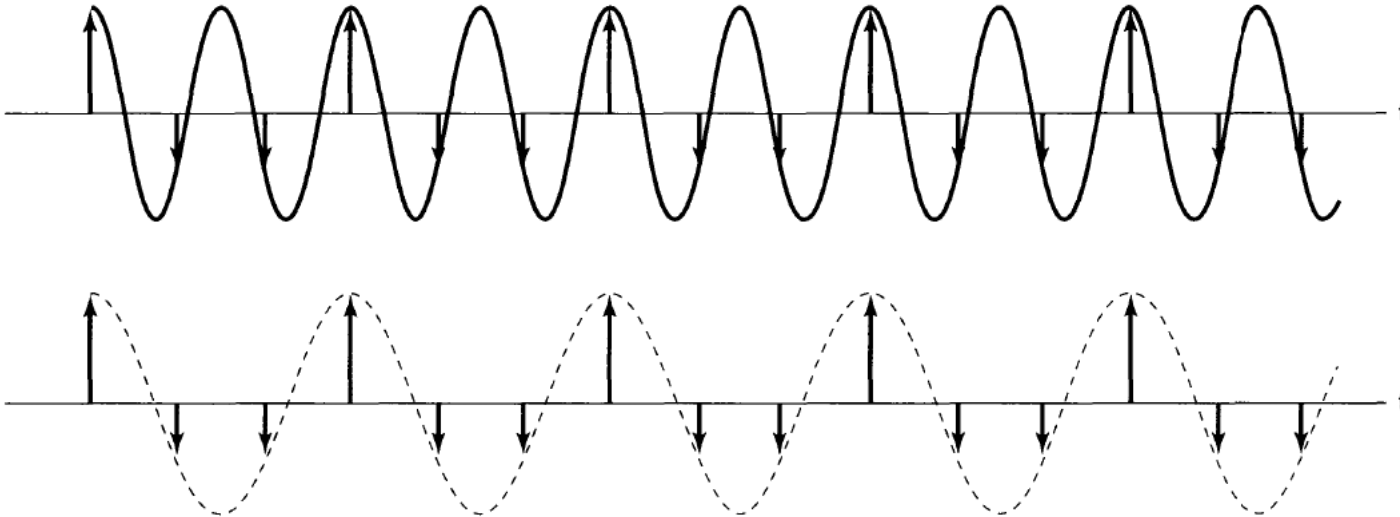
$\omega_s = 3\omega_0 / 2$ (frequency domain)



$$x_r(t) = \cos \frac{\omega_0}{2} t \neq x(t)$$

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