

(Review of Last Time).

19-1

Impulse-Train Sampling

$x(t)$ = an arbitrary CT signal

$p(t) = \sum_n \delta(t - nT)$ = the sampling function

T = sampling interval

$\omega_s = 2\pi/T$ = (radian) sampling frequency

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

$$= x(t) \sum_n \delta(t - nT) = \sum_n x(nT) \delta(t - nT)$$

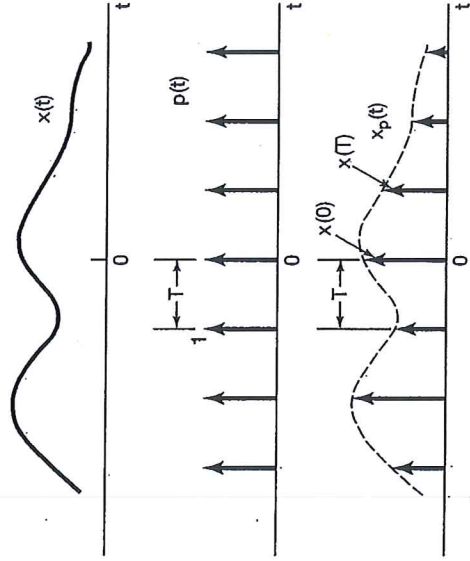
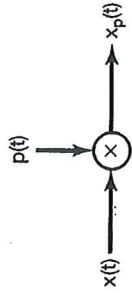


Figure 7.2 Impulse-train sampling.

From the multiplication-in-time/convolution-in-freq. prop. of CTFT we know

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

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Also have

$$\sum_n \delta(t - nT) \leftrightarrow \frac{2\pi}{T} \sum_k \delta(\omega - k\omega_s) \quad \omega_s = \frac{2\pi}{T}$$

$$\therefore X_p(j\omega) = \frac{1}{T} \sum_k X(j(\omega - k\omega_s))$$

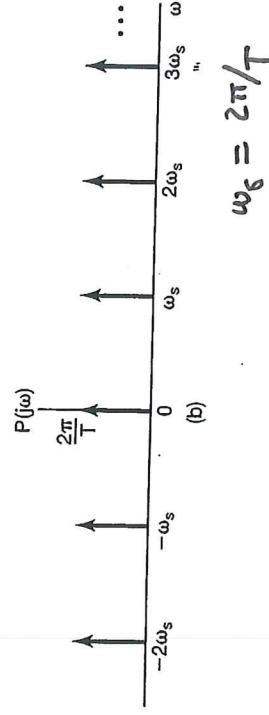
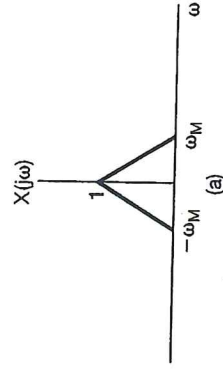


Figure 7.3 Effect in the frequency domain of sampling in the time domain: (a) spectrum of original signal; (b) spectrum of sampling function;

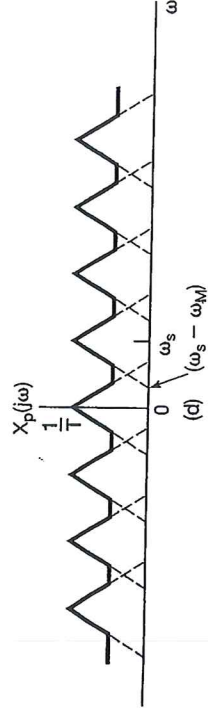
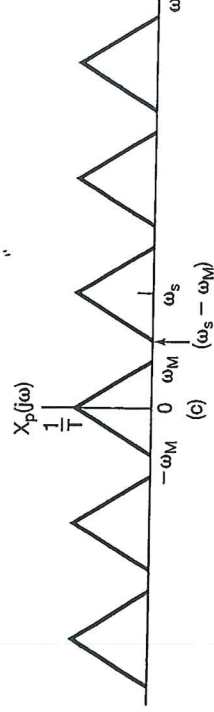


Figure 7.3 Continued (c) spectrum of sampled signal with $\omega_s > 2\omega_M$; (d) spectrum of sampled signal with $\omega_s < 2\omega_M$.

Case: $\omega_s - \omega_M > \omega_M \Leftrightarrow \omega_s > 2\omega_M$ no overlap

Case: $\omega_s - \omega_M < \omega_M \Leftrightarrow \omega_s < 2\omega_M$ overlap or aliasing.

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Fact If $\omega_s > 2\omega_M$ (i.e. Nyquist rate) then $x(t)$ can be recovered from its samples by a lowpass filter.

(This is called the Sampling Theorem. Some names associated with it are

H. Nyquist, J.M. Whittaker, C. Shannon)

The proof is "by picture"

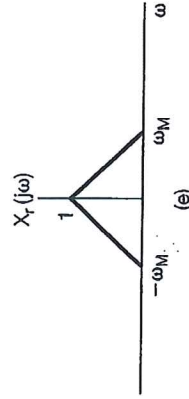
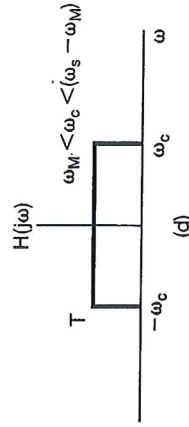
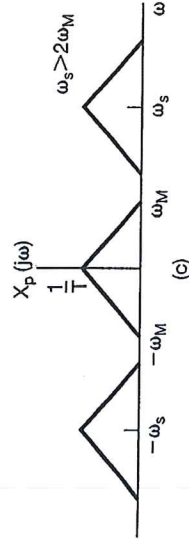
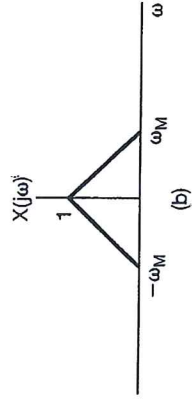
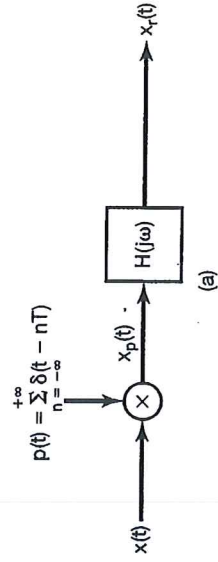


Figure 7.4 Exact recovery of a continuous-time signal from its samples using an ideal lowpass filter: (a) system for sampling and reconstruction; (b) representative spectrum for $x(t)$; (c) corresponding spectrum for $x_p(t)$; (d) ideal lowpass filter to recover $X(j\omega)$ from $X_p(j\omega)$; (e) spectrum of $x_r(t)$.

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The Reconstruction or Interpolation Formula

Consider the reconstruction filter of the previous diagram. Say

$$h_r(t) \longleftrightarrow H_r(j\omega) = \begin{cases} T & |\omega| < \omega_c \\ 0 & \text{else} \end{cases}$$

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

Clearly

$$x_r(t) = x_p * h_r(t) = \sum_n x(nT) \delta(t - nT) * h_r(t)$$

$$= \sum_n x(nT) h_r(t - nT)$$

↑
called an interpolation formula because it shows how to fit a continuous curve between sample points.

$$= \sum_n x(nT) \frac{\omega_c T}{\pi} \frac{\sin(\omega_c(t - nT))}{\omega_c(t - nT)}.$$

Notice that if $\omega_c = \omega_m = \frac{1}{2}\omega_s$ then

① $h_r(0) = 1$

② $h_r(nT) = 0$ if $n \neq 0$

(Draw a picture of interleaving).

$\Rightarrow x_r(nT) = x(nT)$ so it indeed interpolates.

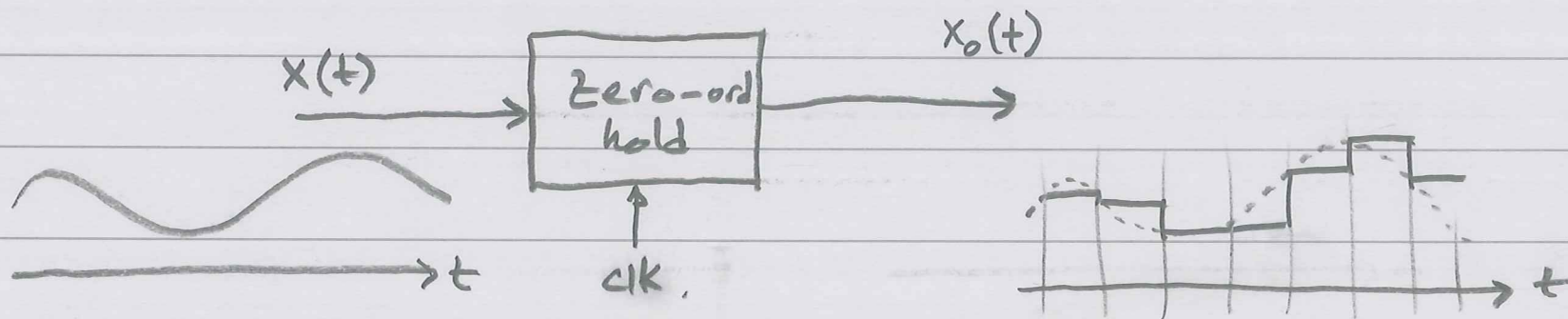
~~What if $\omega_m < \omega_c < \omega_s - \omega_m$?~~

~~$$h_r(t) = \frac{\sin \omega_c t}{\omega_c t} \quad t \neq 0$$~~

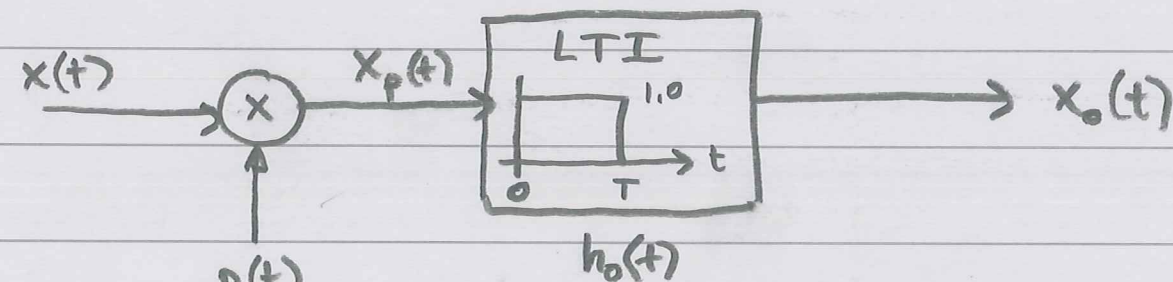
~~$$t = \frac{\pi}{\omega_c}$$~~

Today: Investigate "More Practical" Schemes for Sampling

Zero-order hold sampling



To make a model (using 301 things)



impulse train
sampling

$$x_p(t) = \sum_n x(nT) \delta(t - nT)$$

$$\begin{aligned} \Rightarrow x_0(t) &= x_p * h_0(t) = \sum_n x(nT) \delta(t - nT) * h_0(t) \\ &= \sum_n x(nT) h_0(t - nT) \end{aligned} \quad \rightarrow (*)$$

To see: ① Effect of ZOH sampling in freq. domain
 == ② How we might recover $x(t)$ from $x_o(t)$

$$h_o(t) = \begin{array}{c} 1.0 \\ \text{---} \\ 0 \end{array} \quad \begin{array}{c} \text{---} \\ 0 \end{array} \quad \begin{array}{c} \text{---} \\ T \end{array} \quad \begin{array}{c} \text{---} \\ t \end{array} \quad \longleftrightarrow \quad H_o(j\omega) = T e^{-j\omega T/2} \quad \frac{2 \sin(\omega T/2)}{T \omega}$$

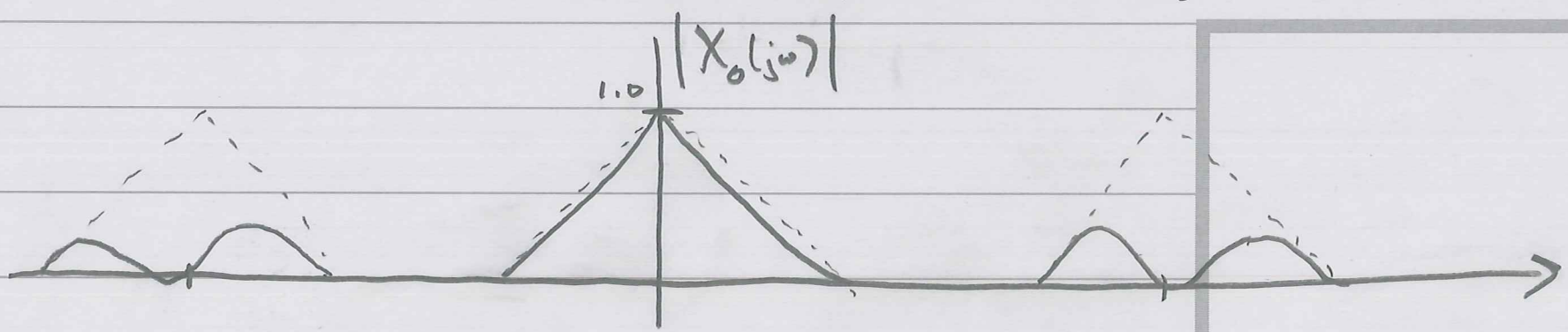
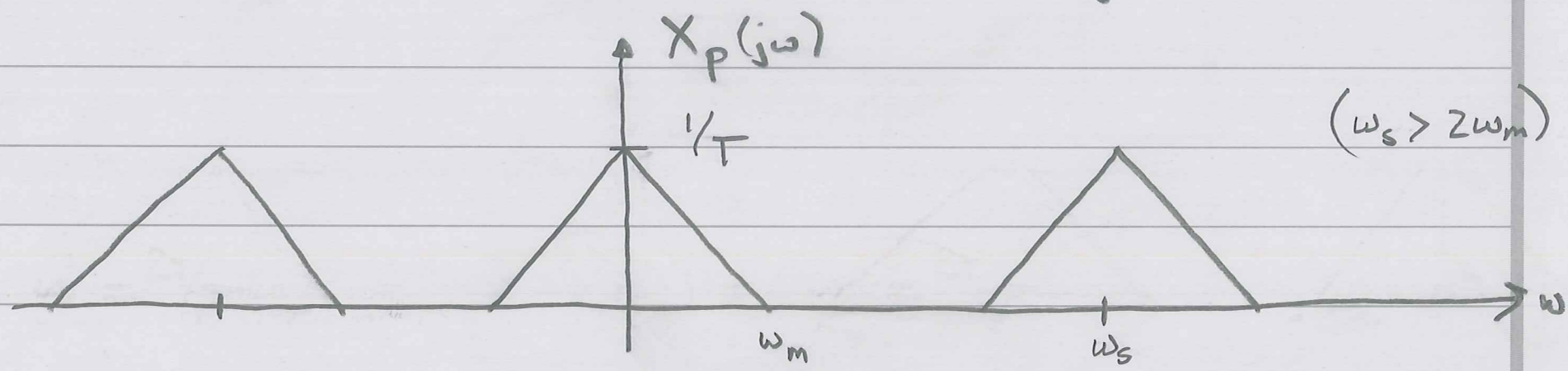
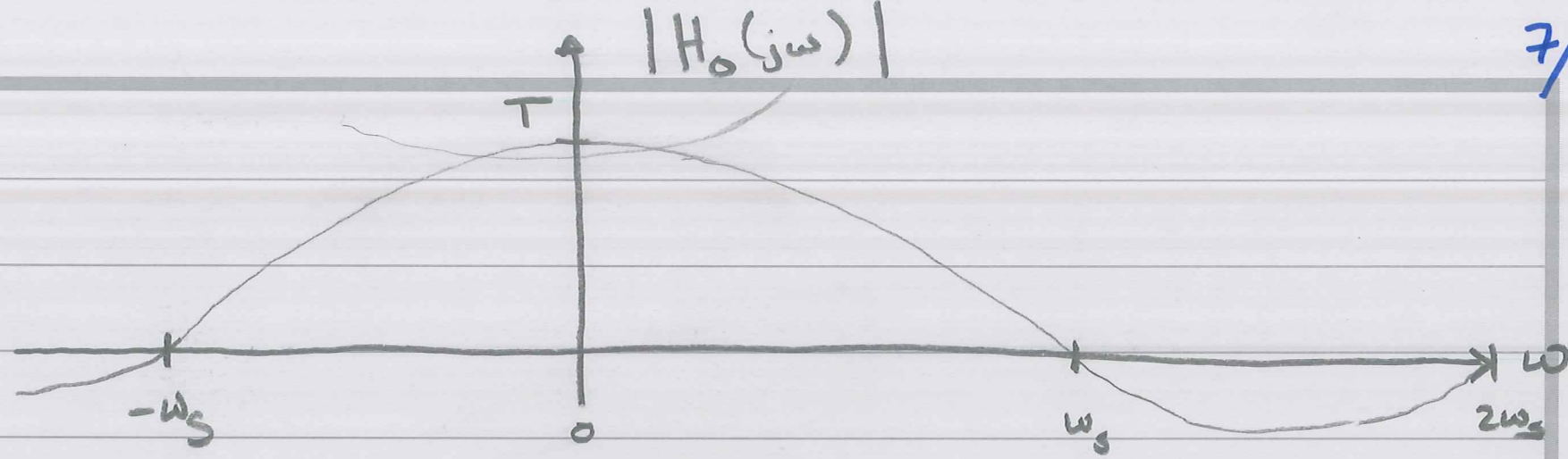
From (*) we know that

$$\begin{aligned} X_o(j\omega) &= H_o(j\omega) X_p(j\omega) \\ &= H_o(j\omega) \underbrace{\frac{1}{T} \sum_k X(j(\omega - k\omega_s))}_{X_p(j\omega)} \quad \omega_s = 2\pi/T \end{aligned}$$

Need to know how this looks.

$$H_o(j\omega) = T e^{-j\omega T/2} \underbrace{\left[\frac{\sin(\omega T/2)}{\omega T/2} \right]}_{??} \cdot e^{-j\omega T/2}$$

$$\begin{aligned} &= T @ \omega = 0 \\ &= 0 \quad \frac{\omega T}{2} = \pi k \\ &\Leftrightarrow \omega = \frac{2\pi k}{T} = k\omega_s \end{aligned}$$

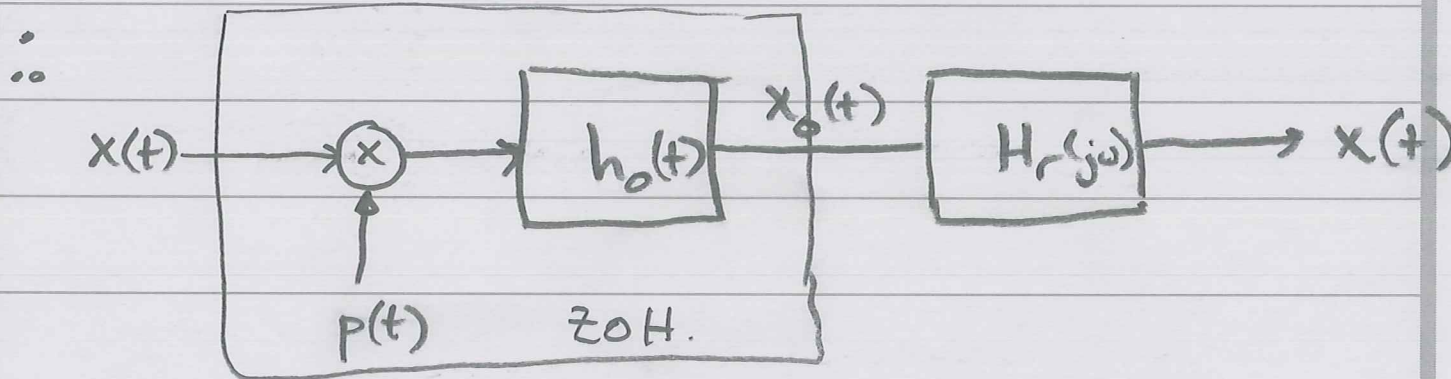
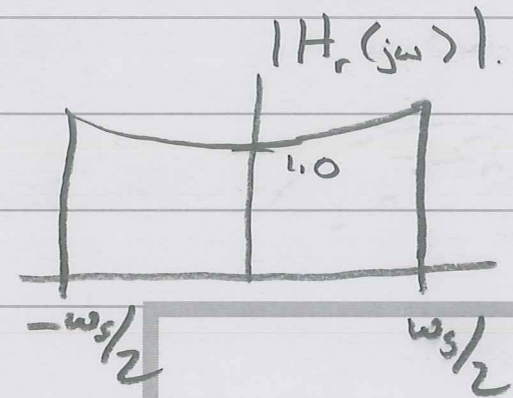


- No zeros in $|w| < w_m$ provided that we sample @ Nyquist or above.

Can use a LP reconstruction filter

$$H_r(j\omega) = \underbrace{\frac{1}{H_o(j\omega)}}_{|w| < \omega_s/2} \underbrace{H(j\omega)}_{\text{the ideal LP filter from impulse sampling.}}$$

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



Example of Aliasing

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

Fix ω_s sampling rate
Vary ω_0

