

# **ELEC 221 Lecture 18**

## **CT/DT conversion and sampling DT signals**

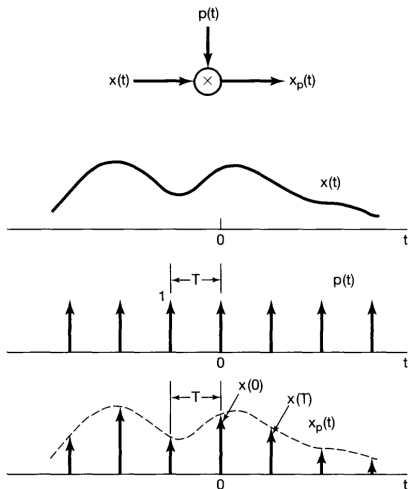
Thursday 14 November 2024

# Announcements

- Quiz 8 on Tuesday (L17 and L18)
- Assignment 4 due Saturday 23 Nov at 23:59 (do 4.2, 4.3, 4.4 after today; can try 4.5)
- Tutorial assignment 4 in Monday's tutorial (image processing)

## Last time

We modeled **sampling** of CT signals as multiplication of a (band-limited) signal with a periodic impulse train:



## Last time

We went to the frequency domain to get a better understanding:

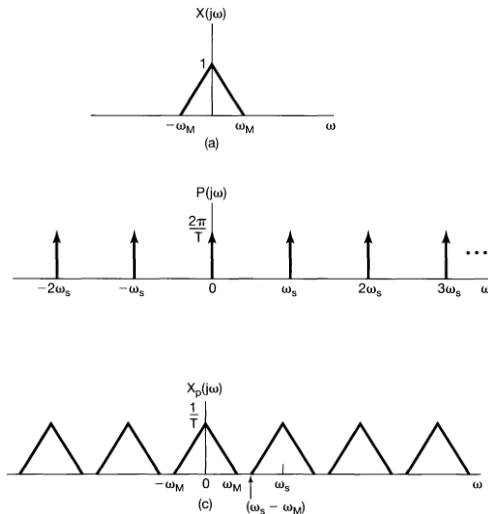


Image credit: Oppenheim 7.1

## Last time

We recovered the original signal by applying a low pass filter

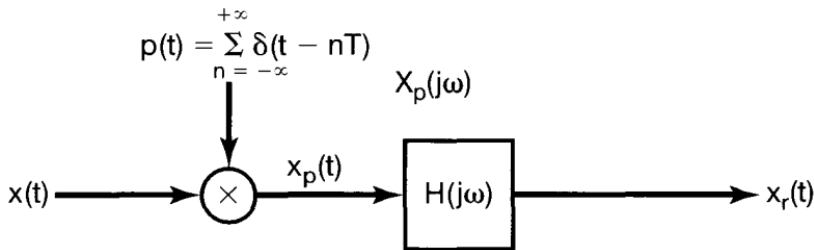
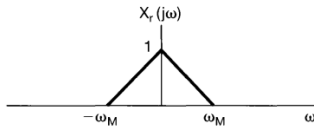
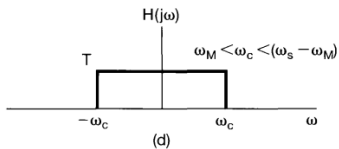
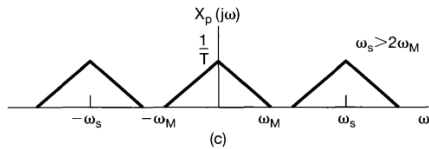


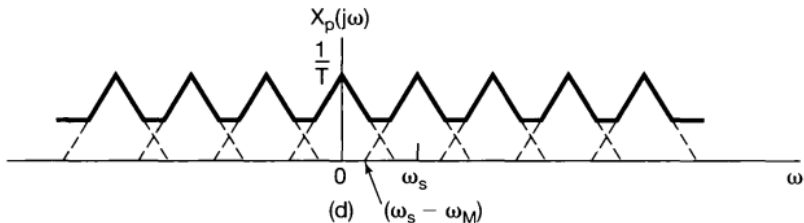
Image credit: Oppenheim 7.1

## Last time



## Last time

This only works if the sampling rate is higher than the **Nyquist rate**, i.e.,  $\omega_s > 2\omega_m$



# Last time

If the frequency isn't high enough, **aliasing** occurs.

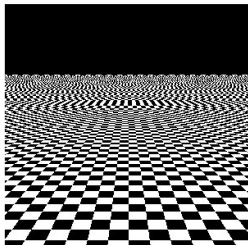
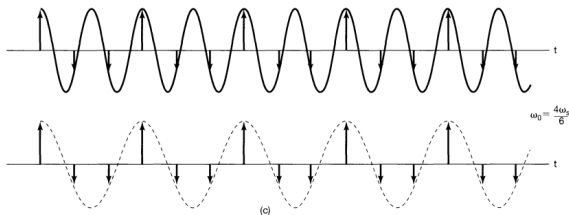


Image credit: Oppenheim 7.3, <https://textureingraphics.wordpress.com/what-is-texture-mapping/anti-aliasing-problem-and-mipmapping/>

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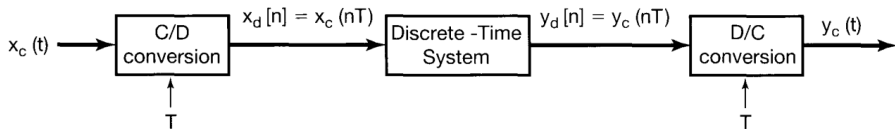


## Learning outcomes:

- describe the frequency domain effects of sampling from a discrete signal
- decimate and interpolate a DT signal
- determine how decimation and interpolation affect the spectrum of a DT signal

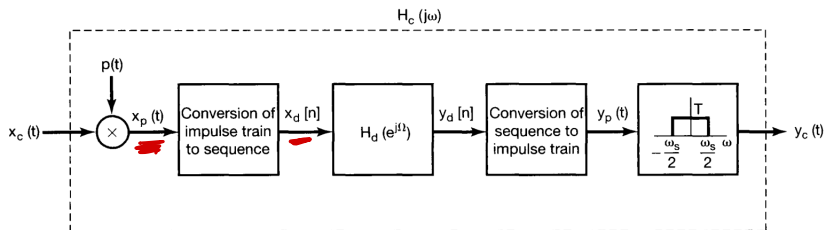
## Converting between DT $\leftrightarrow$ CT

Often convenient to process CT signals by first converting to DT, processing, then converting back.



What is the theory that makes this possible?

# Converting between DT $\leftrightarrow$ CT



Let's explore what happens at the level of the spectra

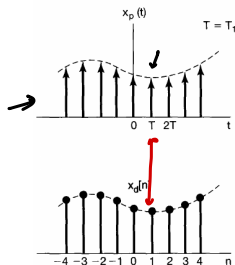
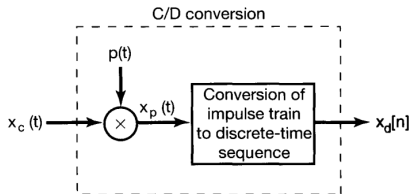
Note: we have *two frequencies*, one in CT, one in DT. Write:

$$\begin{aligned} X_c(j\omega), & \quad Y_c(j\omega) \\ X_d(e^{j\Omega}), & \quad Y_d(e^{j\Omega}) \end{aligned}$$

Image credit: Oppenheim 7.24

# Converting between DT $\leftrightarrow$ CT

First: how are  $X_p(j\omega)$  and  $X_d(e^{j\Omega})$  related?



$$x_p(t) = x(t)p(t) = \sum_{n=-\infty}^{\infty} x_c(nT) \delta(t - nT) \Rightarrow X_p(j\omega) = \sum_{n=-\infty}^{\infty} x_c(nT) e^{-j\omega nT}$$

$$X_d(e^{j\Omega}) = \sum_{n=-\infty}^{\infty} x_d[n] e^{-j\Omega n} = \sum_{n=-\infty}^{\infty} x_c(nT) e^{-j\Omega n} = X_p(j\frac{\Omega}{T})$$

## Converting between DT $\leftrightarrow$ CT

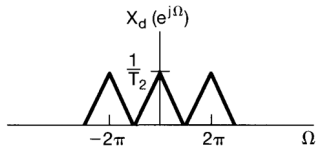
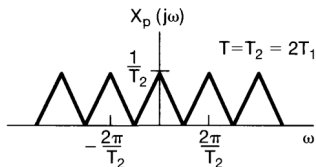
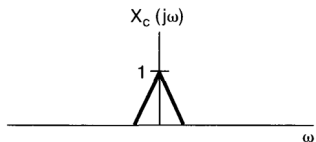
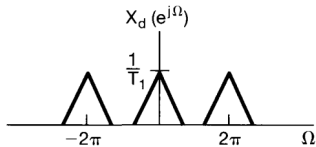
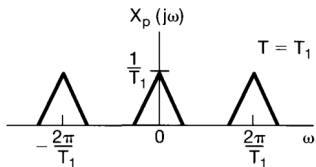
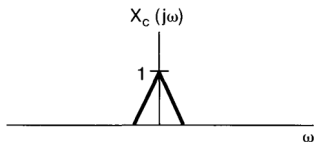
Relate  $X_d(e^{j\Omega})$  back to the original spectrum  $X_c(j\omega)$

$$X_p(j\omega) = \sum_{n=-\infty}^{\infty} x_c(nT) e^{-j\omega nT} = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c(j(\omega - k\omega_s))$$

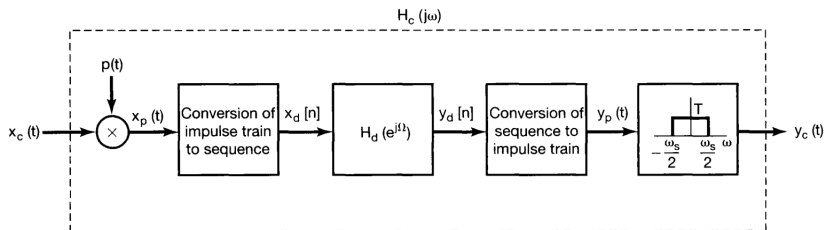
$$\begin{aligned} X_d(e^{j\Omega}) &= X_p(j\frac{\Omega}{T}) \\ &= \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c(j(\frac{\Omega}{T} - k\omega_s)) \\ &= \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c(j\frac{\Omega - 2\pi k}{T}) \end{aligned}$$

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

# Converting between DT $\leftrightarrow$ CT



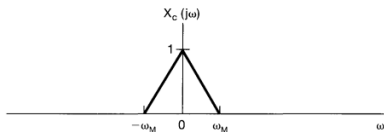
# Converting between DT $\leftrightarrow$ CT



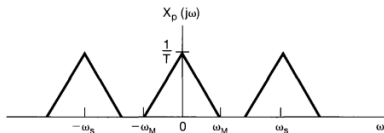
The converted signal  $x_d[n]$  enters a DT system:

$$\begin{aligned}
 Y_d(e^{j\Omega}) &= X_d(e^{j\Omega}) H_d(e^{j\Omega}) \\
 &= \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c(j\frac{\Omega - 2\pi k}{T}) \cdot H_d(e^{j\Omega})
 \end{aligned}$$

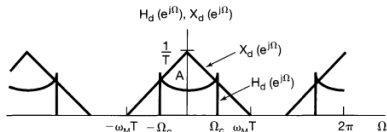
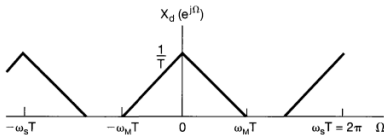
# Converting between DT $\leftrightarrow$ CT



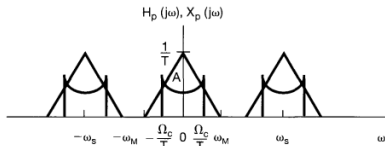
(a)



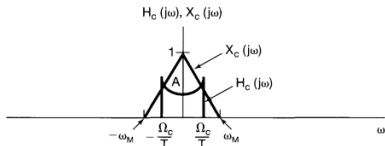
(b)



(d)

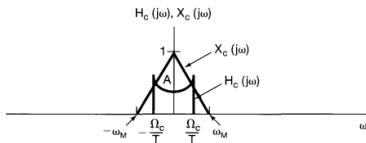
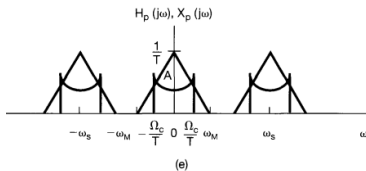


(e)





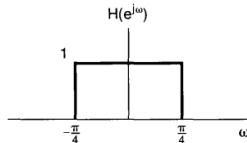
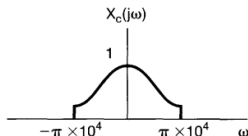
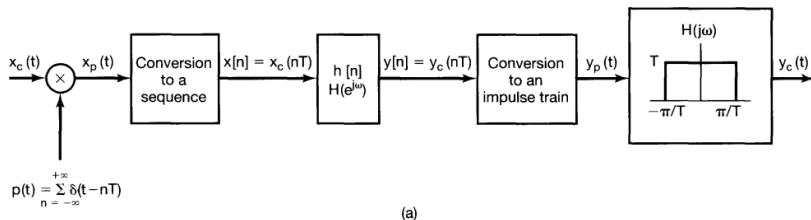
# Sampling of DT signals



$$Y_c(j\omega) = H_c(j\omega) X_c(j\omega)$$

$$H_c(j\omega) = \begin{cases} H_d(e^{j\omega T}) & |\omega| < \omega_s/2 \\ 0 & |\omega| > \omega_s/2 \end{cases}$$

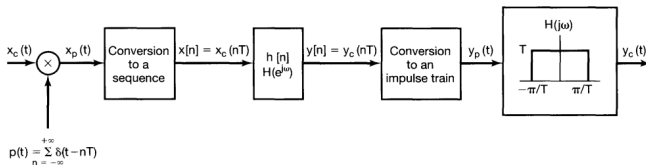
# Example



Sketch:  $X_p(j\omega)$ ,  $X_d(e^{j\omega})$ ,  $Y_d(e^{j\omega})$ ,  $Y_p(j\omega)$ ,  $Y_c(j\omega)$  if  $1/T = 20\text{kHz}$ .

$$\omega_s = \frac{2\pi}{T} = 4\pi \cdot 10^4 \text{ Hz}$$

# Example



(a)

$$\omega_s = 4\pi \cdot 10^4$$

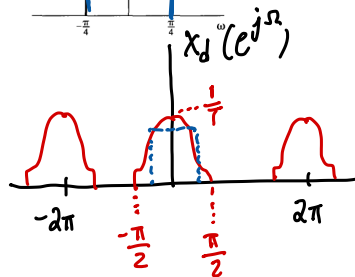
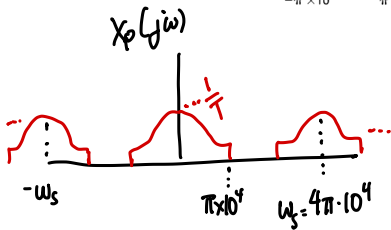
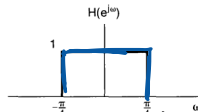
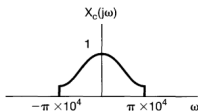
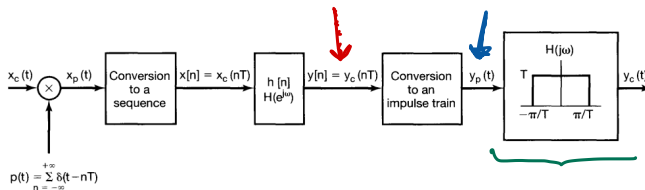


Image credit: Oppenheim Problem 7.29

# Example



(a)

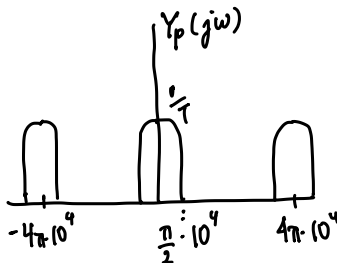
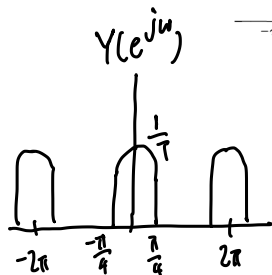
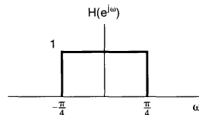
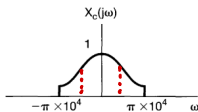
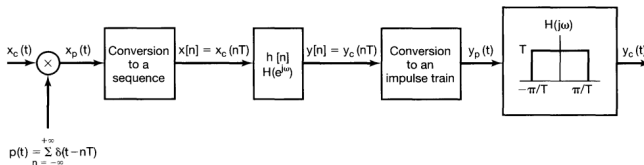


Image credit: Oppenheim Problem 7.29

# Example



(a)

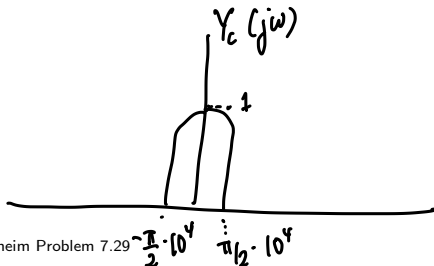
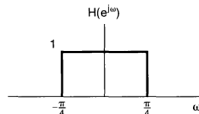
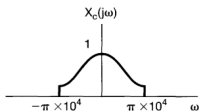
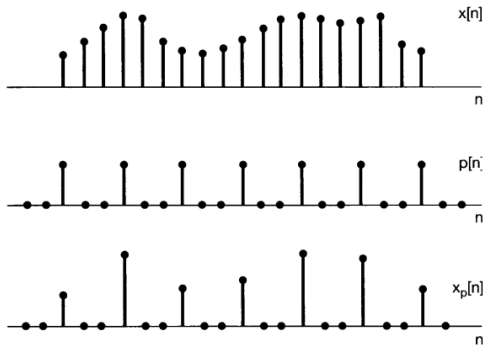


Image credit: Oppenheim Problem 7.29

# Sampling of discrete-time signals

Sample with DT impulse train of period  $N$ :

$$p[n] = \sum_{k=-\infty}^{\infty} \delta[n - kN] \quad x_p[n] = \begin{cases} x[n], & n \text{ integer mult of } N \\ 0 & \text{otherwise} \end{cases}$$



## Sampling of discrete-time signals

Take similar approach as we did in CT:

$$p[n] = \sum_{m=-\infty}^{\infty} \delta[n-mN] = \sum_{m=\langle N \rangle} C_m e^{jm \frac{2\pi n}{N}}$$

$\vdots$

$$C_m = \frac{1}{N}$$

Exercise: derive  $P(e^{j\omega})$

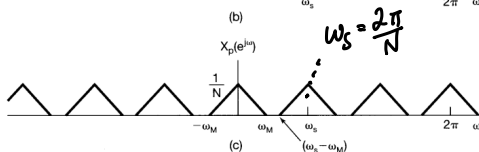
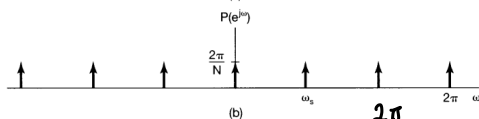
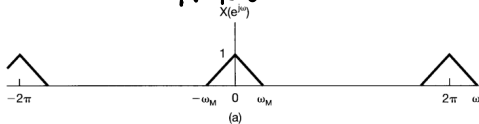
$$P(e^{j\omega}) = \frac{2\pi}{N} \sum_{k=-\infty}^{\infty} \delta(\omega - k\omega_s)$$

$$X[n] \cdot p[n] \Rightarrow \frac{1}{2\pi} X(e^{j\omega}) * P(e^{j\omega})$$

# Sampling of discrete-time signals

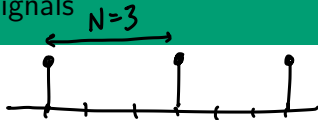
$$X_p(e^{j\omega}) = \frac{1}{2\pi} X(e^{j\omega}) * P(e^{j\omega})$$

$$= \frac{1}{N} \sum_{k=0}^{N-1} X(e^{j(\omega - k\omega_s)})$$



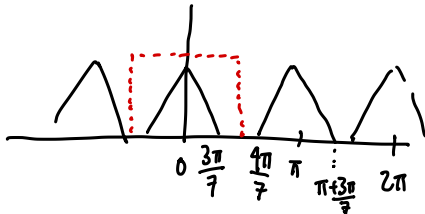
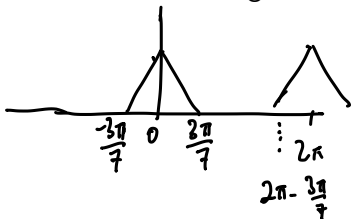


# Sampling of discrete-time signals



Aliasing can happen in DT; some differences due to DT frequency range ( $\pi$  is the highest frequency).

Exercise: suppose  $x[n]$  has  $X(e^{j\omega})$  that is 0 for  $3\pi/7 \leq |\omega| \leq \pi$ . What is the largest sampling period  $N$  we can use without aliasing?



$$\omega_s = \frac{2\pi}{N} > \frac{6\pi}{7} \Rightarrow N_{\max} = 2 \Rightarrow \omega_s = \pi$$

# Decimation

Sampling and then transmitting a DT signal in this way is inefficient

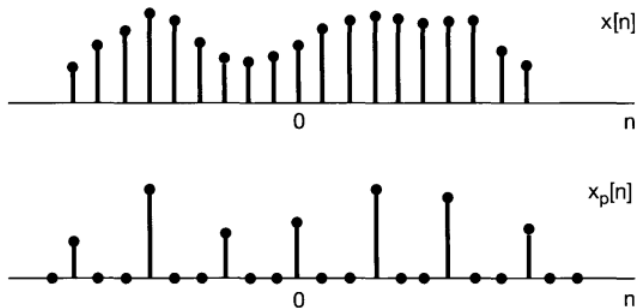
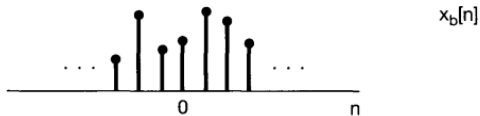
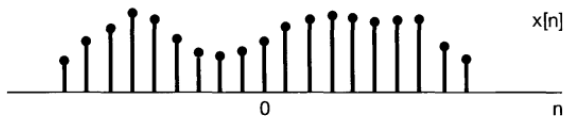


Image credit: Oppenheim 7.31

# Decimation

We can *compress* the representation:



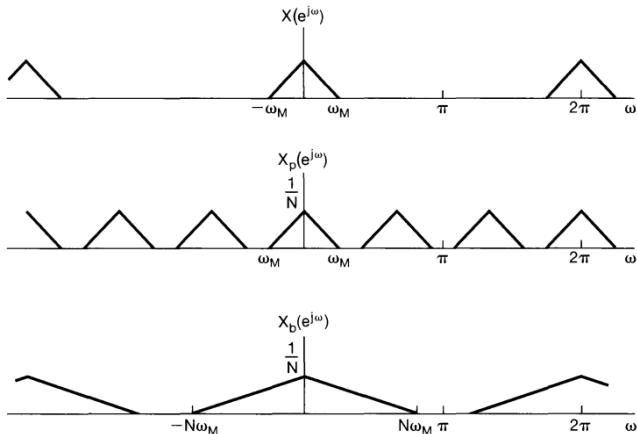
$$x_b[n] = x[nN]$$

Frequency domain effect:

$$X_b(e^{j\omega}) = X_p(e^{j\frac{\omega}{N}})$$

# Decimation

Decimation spreads out the spectrum



If original signal was CT, say that decimation has *downsampled* it.

# Interpolation (upsampling)

Opposite of decimation: add  $N - 1$  points between.

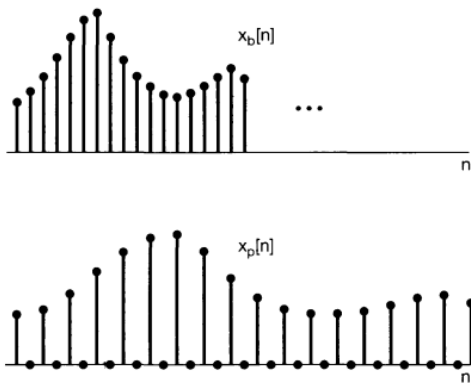


Image credit: Oppenheim 7.5

# Interpolation (upsampling)

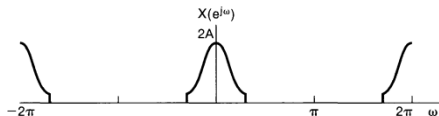
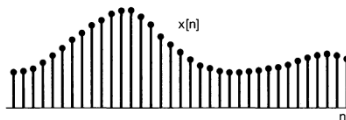
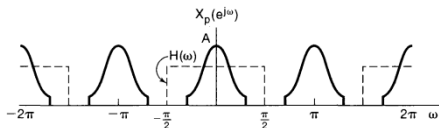
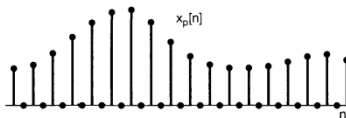
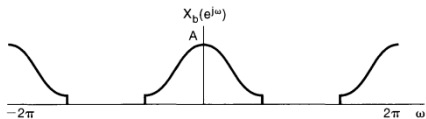
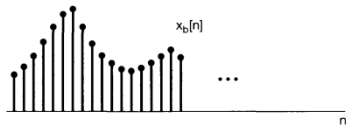


Image credit: Oppenheim 7.5

## Example: down/upsampling

Oppenheim problem 7.19, *★ try yourself!*

**7.19.** Consider the system shown in Figure P7.19, with input  $x[n]$  and the corresponding output  $y[n]$ . The zero-insertion system inserts two points with zero amplitude between each of the sequence values in  $x[n]$ . The decimation is defined by

$$y[n] = w[5n],$$

where  $w[n]$  is the input sequence for the decimation system. If the input is of the form

$$x[n] = \frac{\sin \omega_1 n}{\pi n},$$

determine the output  $y[n]$  for the following values of  $\omega_1$ :

- (a)  $\omega_1 \leq \frac{3\pi}{5}$
- (b)  $\omega_1 > \frac{3\pi}{5}$

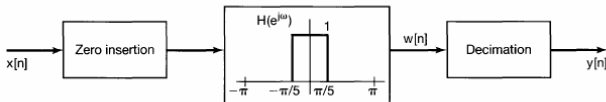


Figure P7.19



## Example: down/upsampling

First case:  $\omega_1 \leq \frac{3\pi}{5}$

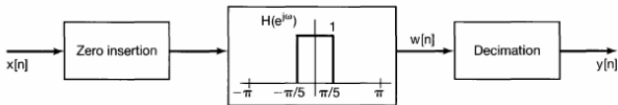


Figure P7.19

## Example: down/upsampling

Second case:  $\omega_1 > \frac{3\pi}{5}$

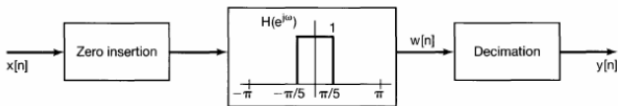


Figure P7.19

## For next time

### Content:

- moving into topic of modulation / communication systems

### Action items:

1. Work on assignment 4
2. Prepare for quiz 8 on Tuesday (L17 and L18 material)
3. Tutorial Assignment 4 Monday

### Recommended reading:

- From this class: Oppenheim 7.4-7.6
- Suggested problems: 7.17, 7.18, 7.20, 7.30, 7.32
- For next class: Oppenheim 8.0-8.4