

# Administrative

- MATLAB1 due @ midnight tonight
  - Turnitin problems? Check 'howto' guide in Lecture4 folder on Trunk. Most importantly, use the shibboleth website
- MATLAB2 is open on Trunk, due midnight 10/2 (related to today's lecture)
- See HW\_Sampling.pdf in Lecture 5 folder – 2 small problems due next Monday

# Updated schedule, from Trunk

Date		Unit	Topic	book section # (P&M)	OUT-MATLAB	IN-MATLAB
<b>6-Sep</b>	1	review	Course overview; LTI systems	2.1-2.2	MATLAB1	
11-Sep	2	review	Convolution, start Z transform	2.3-2.4		
<b>13-Sep</b>	3	review	Z, Fourier transform	3.1-3.2; 4.1, 4.4		
18-Sep	4	review	Sampling	6.1, 6.2		
<b>20-Sep</b>	5	"review"	Reconstruction	"	MATLAB2	MATLAB1
25-Sep	6	LTI systems	LTI system analysis using Z-transform; Rational systems	3.5, some of 3.6		
<b>27-Sep</b>	7	LTI systems	LTI systems analysis using the Fourier transform	5.1,5.2		
2 Oct	8	LTI systems	Phase and group delay, geometric interpretation	5.4		MATLAB2
<b>4-Oct</b>	9	LTI systems	<b>15 min quiz, lectures 1-6.</b> Filter design by pole-zero placement, common simple filters	5.4	MATLAB 3	

# **EE-125: Digital Signal Processing**

## **Lecture 5: Review of Sampling (A/D) and Reconstruction (D/A)**

**Professor Tracey**

**Tufts**

# Lecture 5: Outline

- Finish up sampling from (A/D)
- Signal reconstruction (Digital-to-analog, or D/A)
  - Idealized vs. practical reconstruction
- Aliasing

## Link to book:

- Idealized sampling and reconstruction is P&M 6.1
- Practical A/D and D/A is in 6.3
- **(Bonus topic)** Band-limited sampling (P&M 6.4.1)  
Worth knowing about, but not on HW/tests

# Questions (from last lecture)



Continuous time (CT)

$x(t)$ ,  $X(f)$

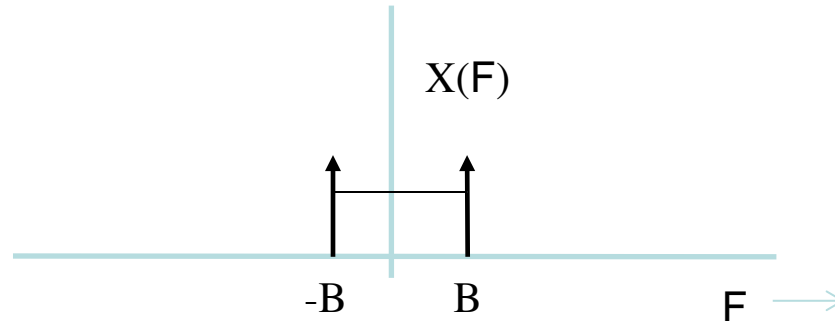
Discrete time (DT)

$x(n)$ ,  $X(\omega)$

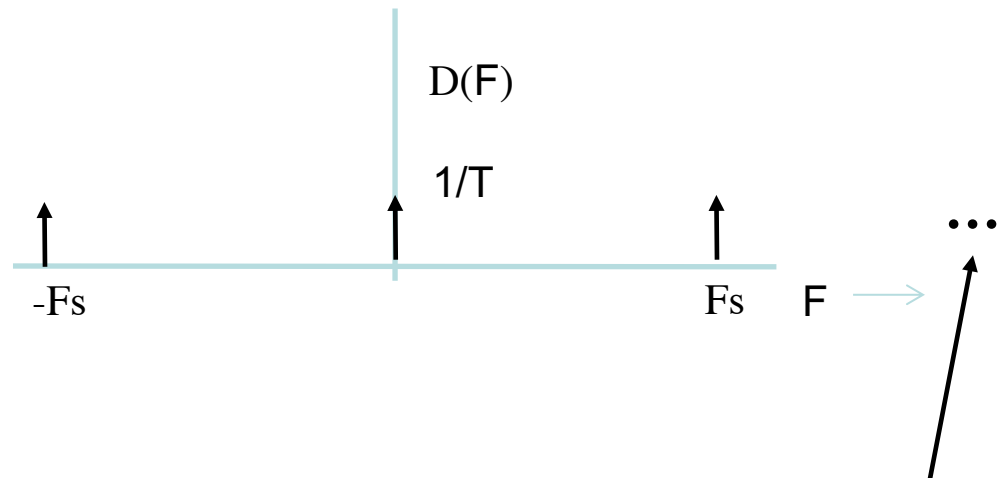
- 1) How are CT quantities (time  $t$ , frequency  $f$ ) related to DT quantities (sample  $n$ , radial frequency  $\omega$ )?
- 2) How does the process of sampling affect the frequency response?

# Sampling, in frequency domain

1) We start with the true CT signal, in continuous frequency. Here the signal is a cosine (peaks) plus other content



2) We take samples at  $t = nT$ , (time domain pulse train) where  $T = 1/F_s$ , the sampling rate. Mathematically, we multiply  $x(t)$  by  $D(t)$ , a pulse train.  $D(F)$  is shown here.

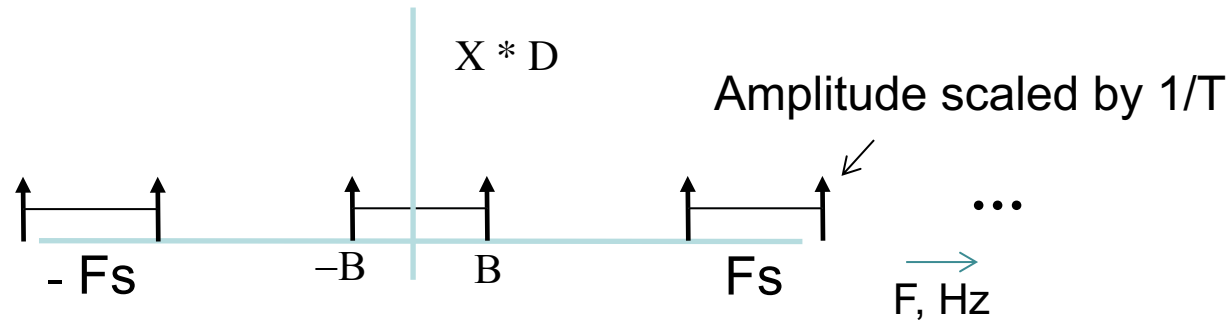


Infinite # of copies at  $\pm 2 F_s, \pm 3 F_s$ , etc

# Sampling - continued

3) The sampled signal's spectrum is  $X * D$ .

- Periodic w/ period  $F_s$
- Get overlap if  $B > F_s/2$

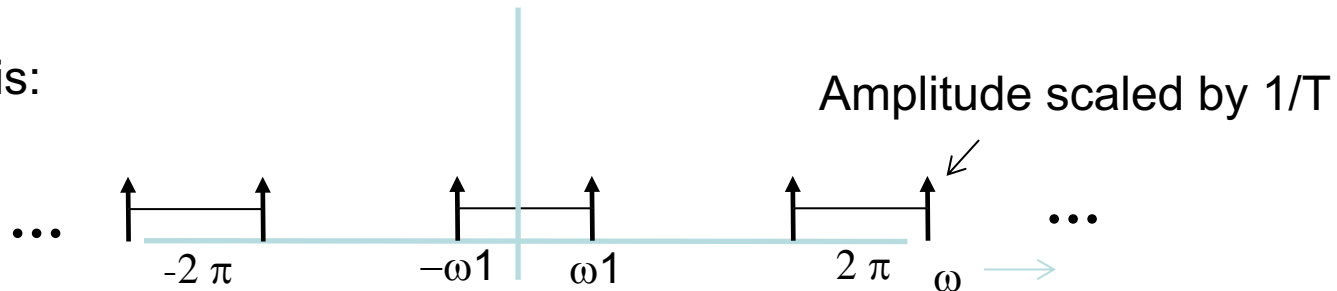


4) Mapping from  $F$  to  $\omega$  is:

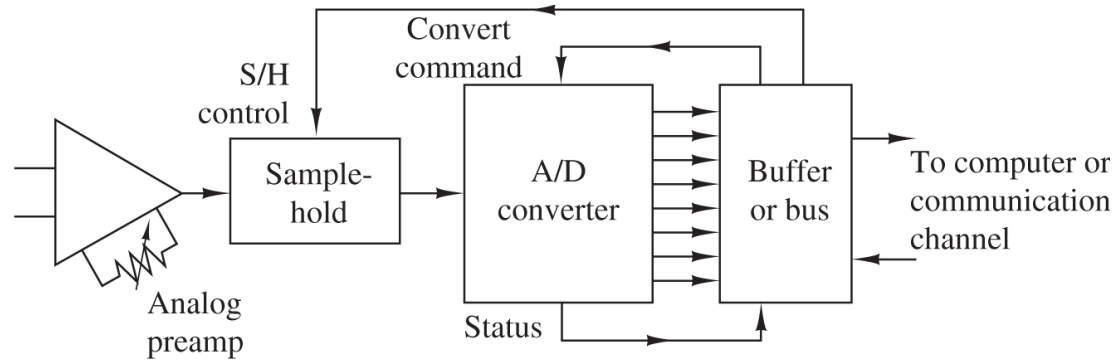
$$F_s \rightarrow 2\pi$$

$$F_s/2 \rightarrow \pi$$

$$B \rightarrow 2\pi B/F_s = \omega_1$$

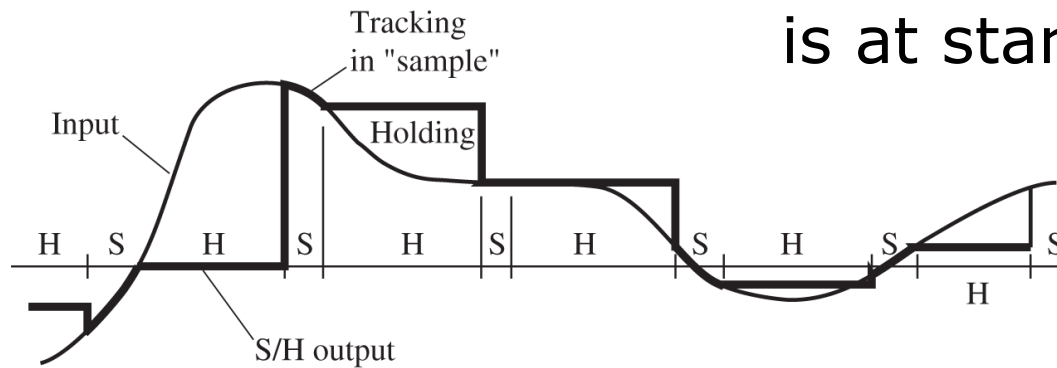


# Practical A/D: Sample and Hold



(a)

Idealized  $\delta(t)$  sample is at start of 'hold'



(b)

**Figure 6.3.1** (a) Block diagram of basic elements of an A/D converter;  
(b) time-domain response of an ideal S/H circuit.



# Practical A/D: quantization

- A/D output is in integer “counts”, or levels
- Spacing of levels is the quantization step:  
 $\Delta = (X_{\max} - X_{\min}) / (L - 1)$   
for  $L$  levels (figure shows  $L=8$ )

Lots more detail in P&M 6.3

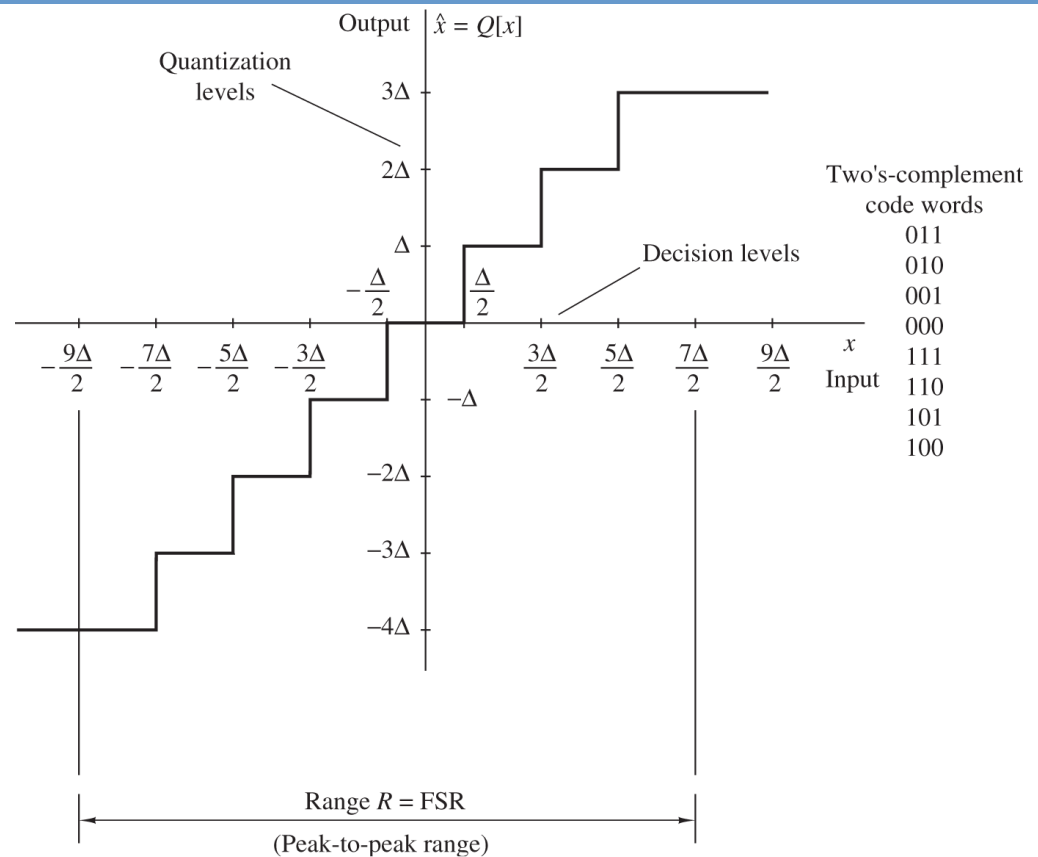
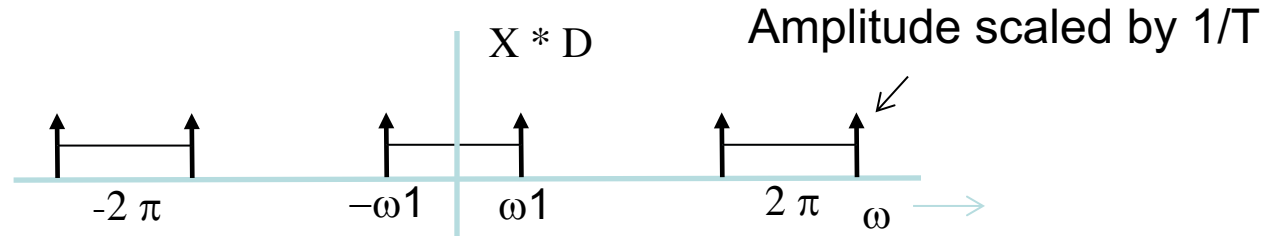


Figure 6.3.3 Example of a midtread quantizer.

# Reconstruction (D/A)

## Really simple in pictures (in freq)....

The sampled signal's spectrum is  $X * D$ .

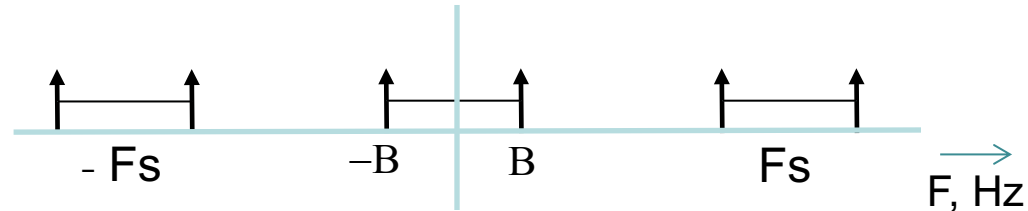


Mapping from  $\omega$  to  $F$ :

$$F_s \rightarrow 2\pi$$

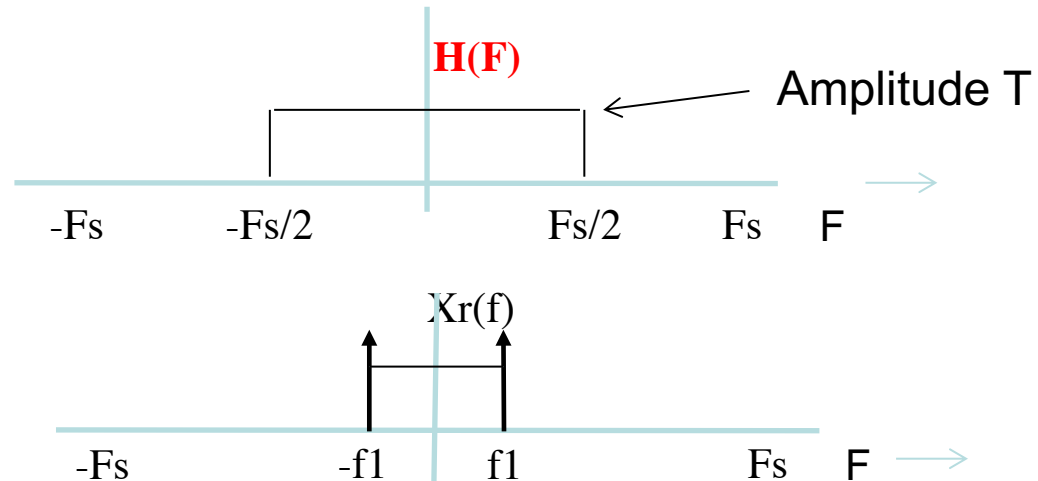
$$F_s/2 \rightarrow \pi$$

$$f_1 \rightarrow 2\pi f_1/F_s = \omega 1$$

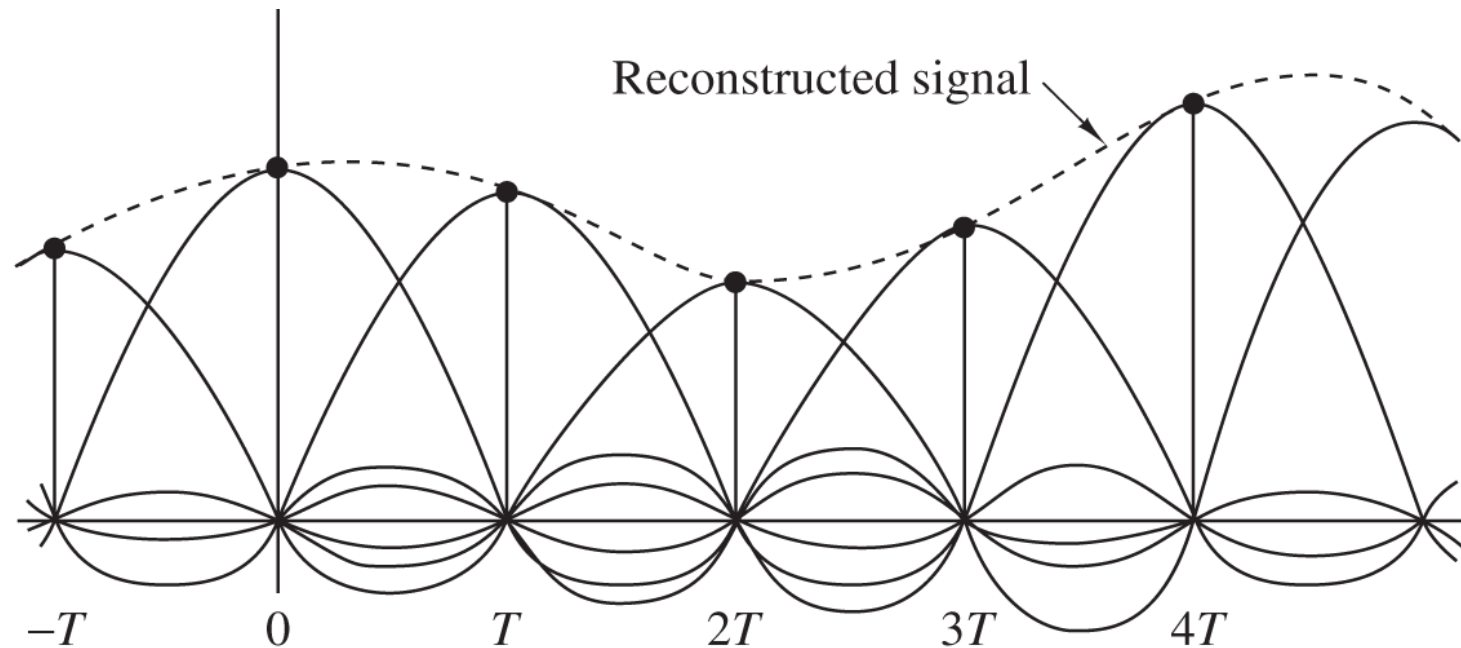


For idealized reconstruction, we use a perfect **ANALOG** low pass filter (LPF)...

... thus recovering the original spectrum



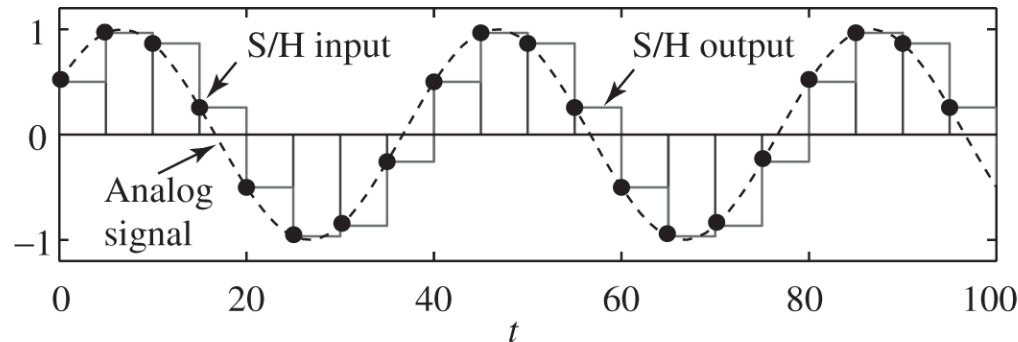
# Sinc-based interpolation during reconstruction



**Figure 6.1.2** Reconstruction of a continuous-time signal using ideal interpolation.

See 'recon.pdf' on Trunk for derivation

# Practical D/A: Sample & Hold + LPF



**Figure 6.3.8** Response of an S/H interpolator to a discrete-time sinusoidal signal.

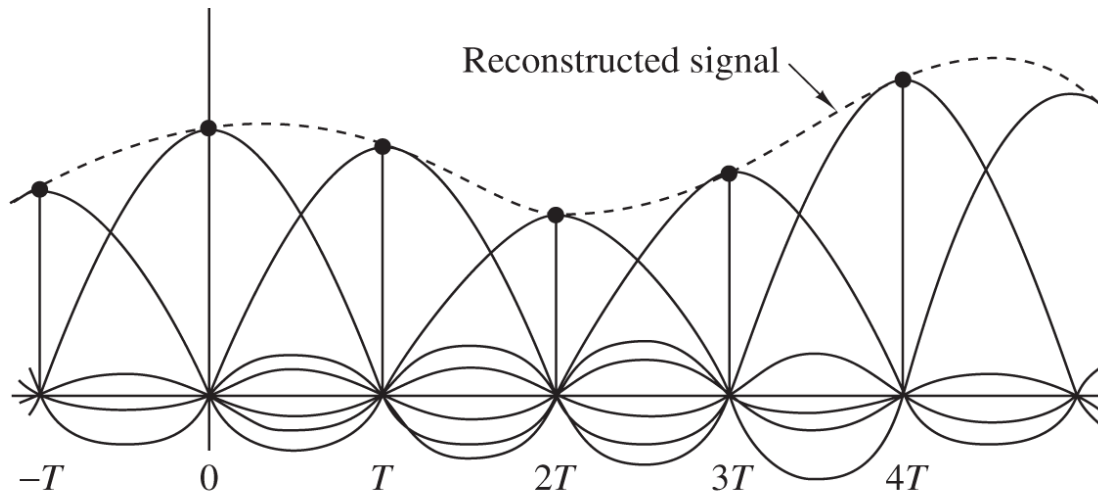
Then, pass S/H output through a low-pass filter to smooth out the rough edges

This is a fairly simple system that can be analyzed exactly....

# Setup for Matlab2, 2<sup>nd</sup> part

- D/A converters (this lecture): convert  $x(n)$  to  $x(t)$  by interpolating the original samples  $x(n)$  (at times  $t = nT$ ) to other times  $t$
- Matlab2, 2<sup>nd</sup> part: goal is to study reconstruction without having you build hardware!
  - Write code that interpolates the original samples  $x(n)$  (at times  $t = nT$ ) to other discrete time points between the original samples
  - A picture should help...

# Sinc-based interpolation during reconstruction



**Figure 6.1.2** Reconstruction of a continuous-time signal using ideal interpolation.

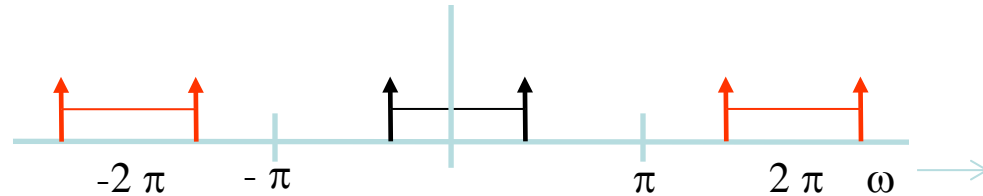
Question: how would linear interpolation in time look?

# Aliasing (related to part 1 of MATLAB2)

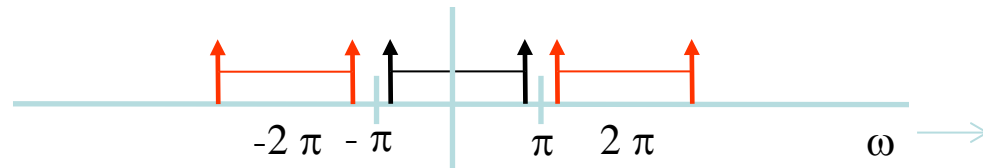
- Aliasing of sinusoidal components
- Aliasing of non-bandlimited signals
  - Definition of bandlimited
- Need for prefiltering before A/D

# Sampling refresher – Aliasing in the frequency domain (DTFT)

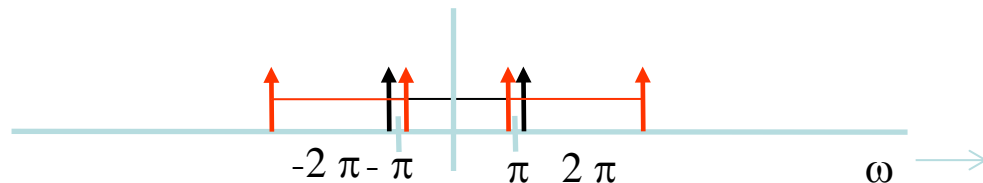
1) Here we are well sampled – high  $F_s$



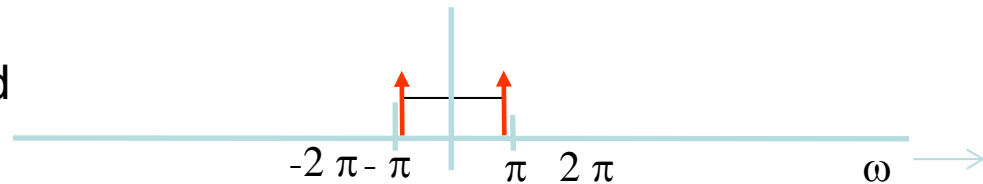
2) Here we've lowered sampling rate so  $F_s/2$  ( $\omega = \pi$ ) is just a little larger than the max frequency



3) Here we've lowered sampling rate so  $F_s/2$  ( $\omega = \pi$ ) is  $<$  signal frequency; signal is **aliased**



4) After low-pass filtering and reconstruction, **sinusoid appears at lower frequency**



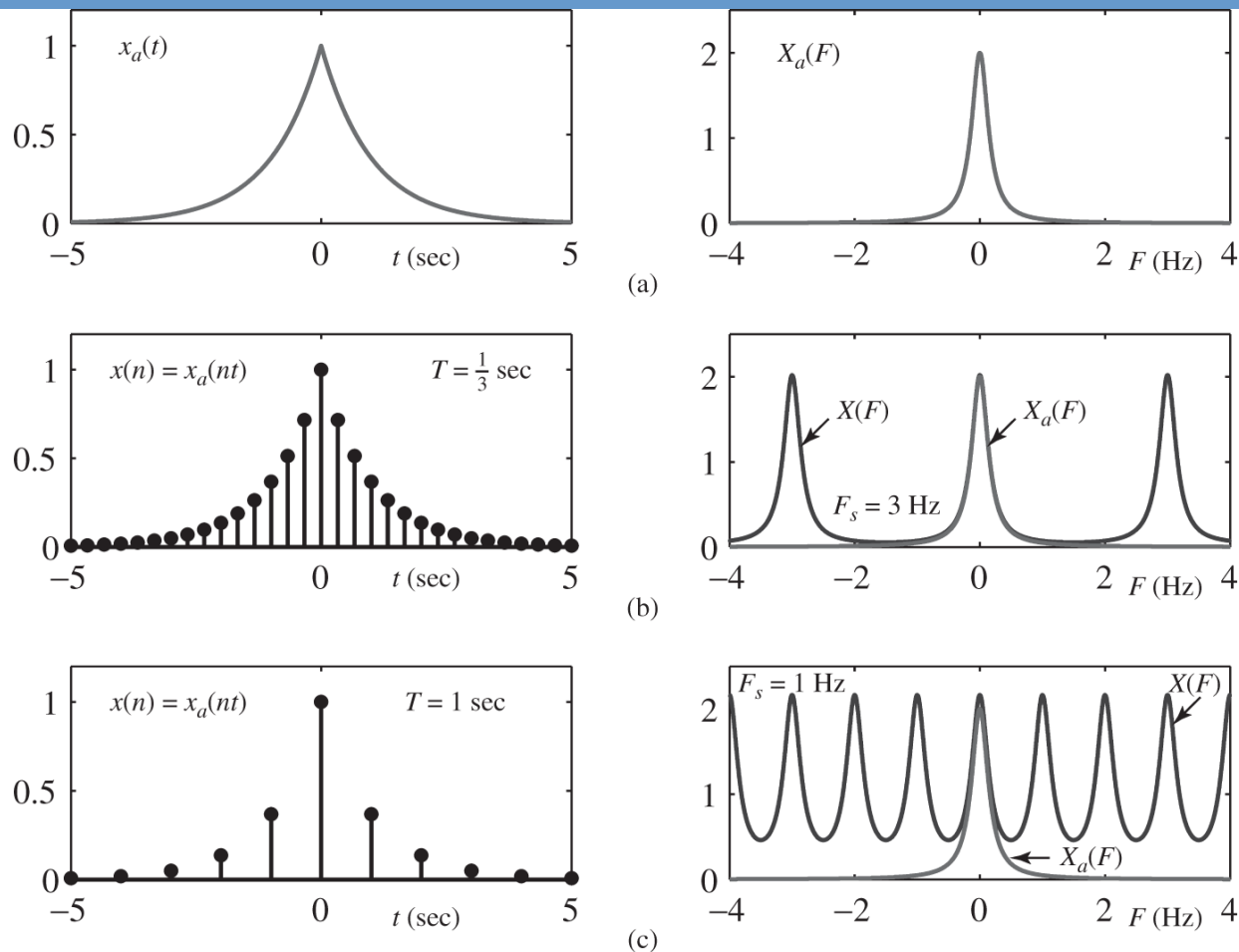
[Video example of aliasing](https://www.youtube.com/watch?v=lEOF9VDE_kk)

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Tufts



# Aliasing of non-bandlimited signals (see example in book)



**Figure 6.1.7** (a) Analog signal  $x_a(t)$  and its spectrum  $X_a(F)$ ; (b)  $x(n) = x_a(nT)$  and its spectrum for  $F_s = 3$  Hz; and (c)  $x(n) = x_a(nT)$  and its spectrum for  $F_s = 1$  Hz.

# Lecture 5: Outline

- Review of sampling from last week (A/D)
- Signal reconstruction (Digital-to-analog, or D/A)
- **Band-limited sampling (P&M 6.4.1)**
  - For reading only...

# A quick question

- True or false:

The Nyquist criterion says that we should sample at twice the highest frequency in the data

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- True or **false**:

The Nyquist criterion says that we should sample at twice the highest frequency in the data

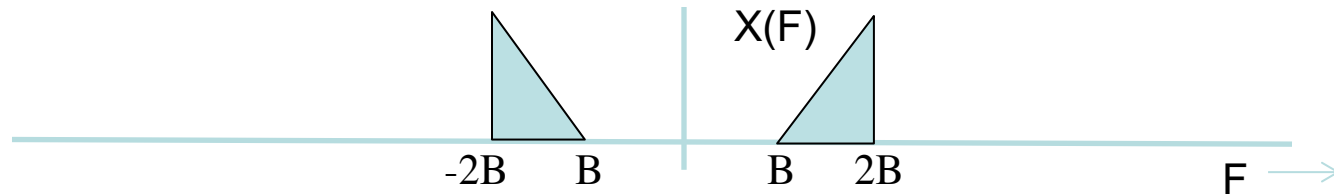
- It says to sample at  $F_s > 2B$

where  $B$  is bandwidth – not necessarily the highest frequency

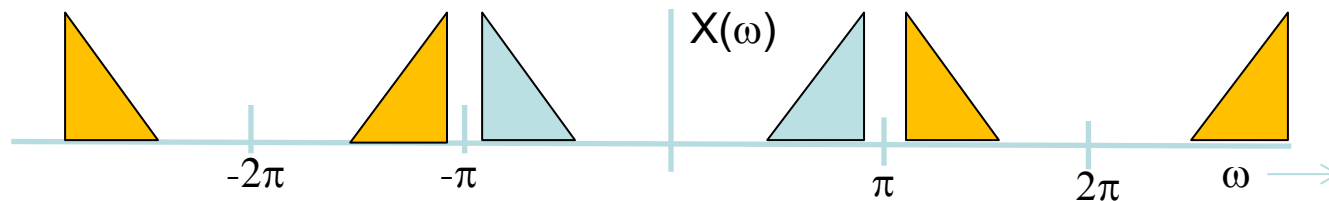
# Sampling/reconstruction of bandpass signals

- We often think of the sampling theorem as stating “sample at twice the highest frequency of the signal”
- Really, it states “sample at twice the bandwidth”.
  - For most signals, the bandwidth is from 0 Hz to the highest frequency of interest
  - However, when signals are *bandpass* (0 outside some range  $F_1 < F < F_2$ ) we can exploit this
- See section 6.4 for discussion of bandpass sampling
  - Integer band positioning (6.4.1) will be explored as part of MATLAB1.
  - Arbitrary band positioning is much more involved but same general idea

# Integer band sampling, graphically

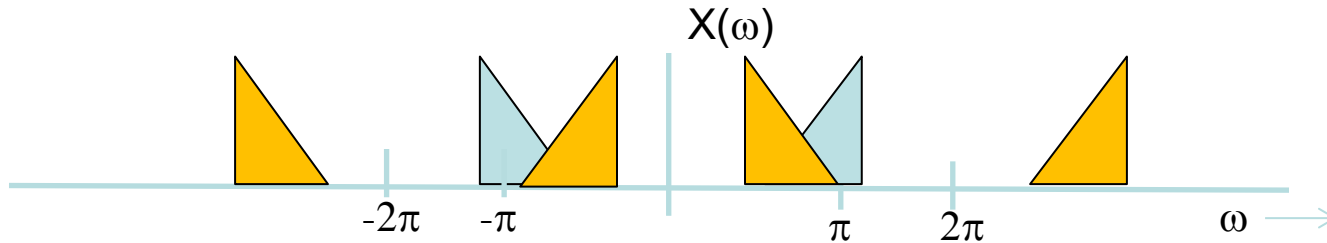


This bandpass signal is an example of integer band positioning, with highest frequency = integer multiple of the bandwidth;  $F_h = mB$  (here  $m=2$ )

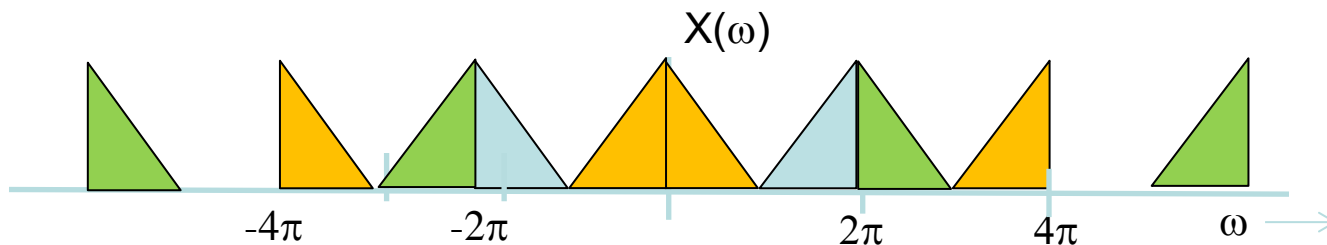


Sampled signal,  $F_s > 2 \times 2B$  (>twice highest frequency). Here, the periodic repetitions of the signal (orange) don't encroach into the  $[-\pi, \pi]$  region.

# Integer band sampling, graphically



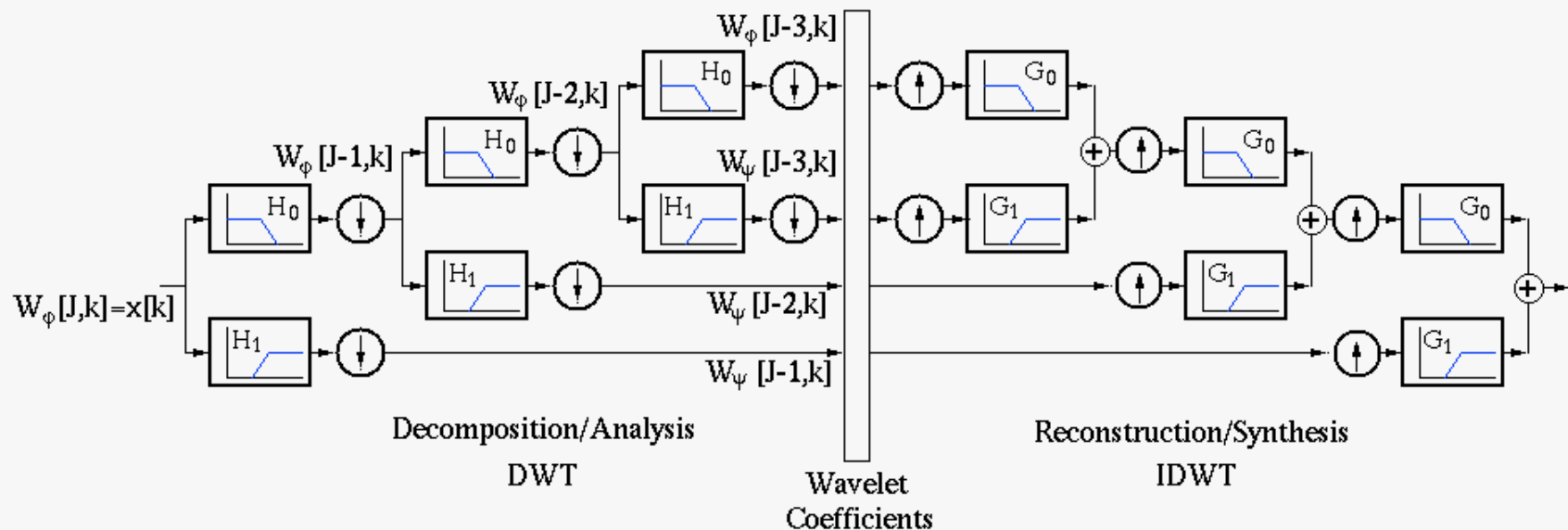
Sampled signal, lower sample rate:  $2B < F_s < 2 \cdot 2B$ . Aliasing is clear. Note, plot doesn't show signals centered around  $\pm 4\pi$ , etc.



Sampled signal, even lower sample rate:  $F_s = 2B$ . No overlap, so signal can be recovered!

Here, showing signals centered around  $\pm 4\pi$  in green, but not  $\pm 6\pi$  – those would fill in the other gaps

# Wavelet decomposition – one use of bandpass sampling



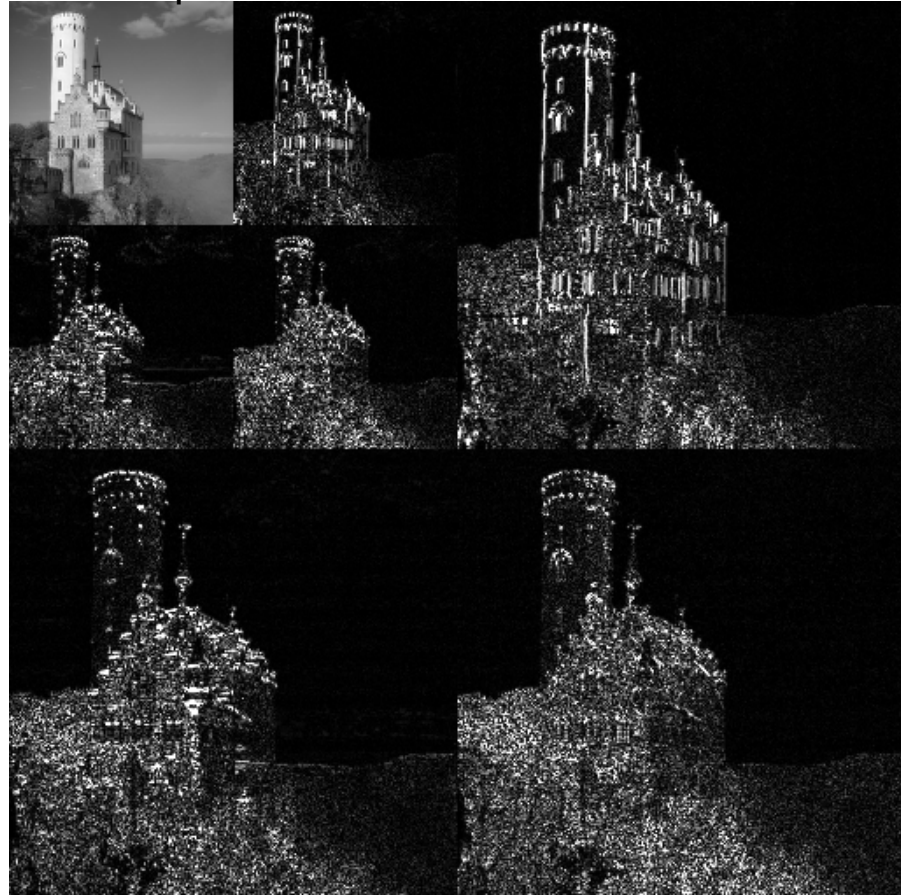
- With wavelets, we split a signal iteratively into low-pass and high-pass components. If the filters are designed correctly (special filters needed) then we can reconstruct the signal
- Bandpass sampling says that the highpass and lowpass signals at each stage can be sampled at half the rate of the previous stage (as each contains half the bandwidth).
- This limits the data storage needed



# Wavelet compression (JPEG2000)

- Example from image processing.
- The image can be split iteratively into bandpass versions w/o increasing storage
- Because many wavelet coefficients are zero, we don't need to store them – this leads to compression

Low freq



High freq