

COMP ENG 4TL4:

# Digital Signal Processing

Notes for Lecture #10

Tuesday, September 30, 2003

## 3.3 More on Sampling Theory

### Relating the DTFT to the CTFT:

Recall from Lecture #2 that the impulse-train approximation  $x_s(t)$  of a sampled continuous-time signal  $x_c(t)$  has the Fourier transform:

$$X_s(j\Omega) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c(j(\Omega - k\Omega_s)),$$

where  $T$  is the sampling period,  $\Omega$  is the continuous-time frequency in radians/s, and  $\Omega_s (= 2\pi f_s)$  is the angular sampling frequency in radians/s.

Consequently,  $X_s(j\Omega)$  consists of copies of  $X_c(j\Omega)$  scaled by  $1/T$  and shifted by  $k\Omega_s$ .

Recall from the derivation of the DTFT in Lecture #8:

$$X_s(j\Omega) = \sum_{n=-\infty}^{\infty} x[n] e^{-j\Omega nT} \quad \text{and}$$

$$X(e^{j\omega}) = \sum_{n=-\infty}^{\infty} x[n] e^{-j\omega n} \quad (\Leftarrow \text{DTFT})$$

It follows that:

$$X_s(j\Omega) = X(e^{j\omega}) \Big|_{\omega=\Omega T} = X(e^{j\Omega T}).$$

Consequently:

$$X(e^{j\Omega T}) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c(j(\Omega - k\Omega_s)),$$

or equivalently:

$$X(e^{j\omega}) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c\left(j\left(\frac{\omega}{T} - \frac{2\pi k}{T}\right)\right).$$

From these equations we can see that the DTFT  $X(e^{j\omega})$  is simply a frequency-scaled version of  $X_s(j\Omega)$ , with the frequency scaling specified by  $\omega = \Omega T$ .

This scaling can alternatively be viewed as a normalization of the frequency axis so that  $\Omega = \Omega_s$  in  $X_s(j\Omega)$  is normalized to  $\omega = 2\pi$  radians in  $X(e^{j\omega})$ .

## Resampling of discrete-time sequences:

Consider a discrete-time sequence  $x[n]$  obtained by sampling a continuous time sequence  $x_c(t)$  with sampling period  $T$ , i.e.:

$$x[n] = x_c(nT) .$$

It is often necessary to change the sampling rate of a discrete-time signal, such that:

$$x'[n] = x_c(nT') ,$$

where  $T' \neq T$ .

One approach: reconstruct the continuous-time signal and resample it with period  $T'$ .

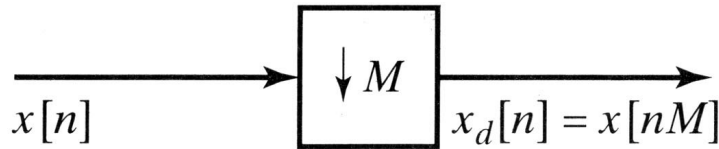
Problem: nonideal reconstruction filters, D/A converters and A/D converters.

Thus: need methods of changing the sampling rate that involve only discrete-time operations.

## Sampling rate reduction by an integer factor:

$$x_d[n] = x[nM] = x_c(nMT)$$

decimation or  
downsampling



(Oppenheim  
and Schaffer)

Sampling  
period  $T$

Sampling  
period  $T' = MT$

**Figure 4.20** Representation of a  
compressor or discrete-time sampler.

To avoid aliasing, the signal  $x[n]$  should be bandlimited to  $\Omega_N < \pi/T$  radians/s ( $\equiv \omega_N < \pi$  radians)  $\Rightarrow$

For the decimated signal, an  $M$ -times lower cutoff frequency  $\Omega_{d,N} < \pi/MT$  radians/s ( $\equiv \omega_{d,N} < \pi/M$  radians) is required.

That is, aliasing can be avoided if:

- the original sampling rate was  $\geq M$  times the Nyquist rate, or
- the bandwidth of the sequence is reduced by a factor of  $M$  by a discrete-time filter before downsampling.

The DTFT of  $x_d[n] = x[nM] = x_c(nT')$  is:

$$X_d(e^{j\omega}) = \frac{1}{T'} \sum_{r=-\infty}^{\infty} X_c\left(j\left(\frac{\omega}{T'} - \frac{2\pi r}{T'}\right)\right).$$

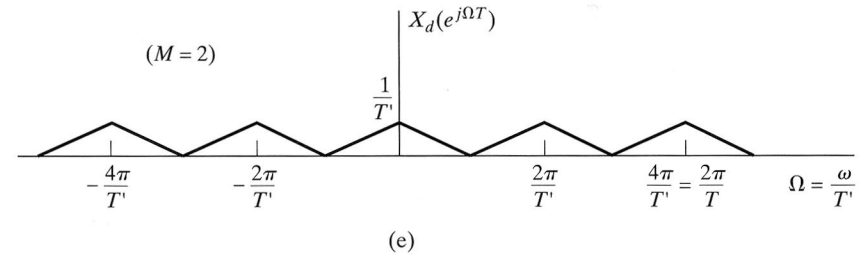
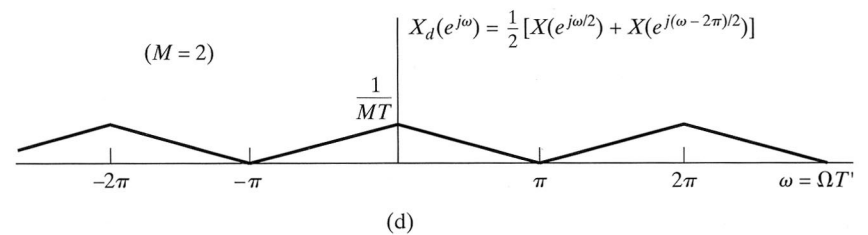
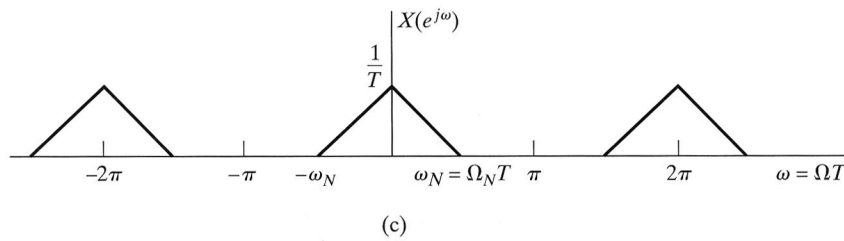
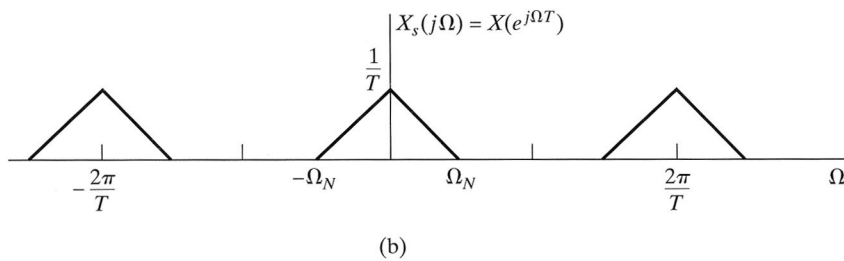
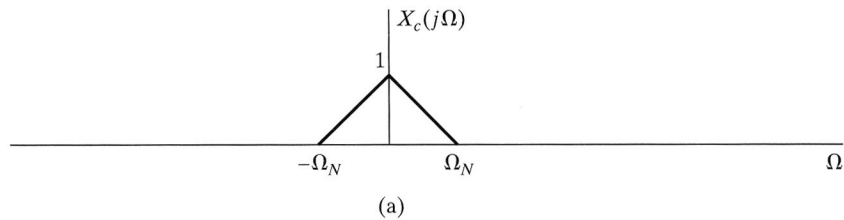
Since  $T' = MT$ :

$$\begin{aligned} X_d(e^{j\omega}) &= \frac{1}{MT} \sum_{r=-\infty}^{\infty} X_c\left(j\left(\frac{\omega}{MT} - \frac{2\pi r}{MT}\right)\right) \\ &= \frac{1}{M} \sum_{i=0}^{M-1} \left[ \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c\left(j\left(\frac{\omega - 2\pi i}{MT} - \frac{2\pi k}{T}\right)\right) \right] \\ &= \frac{1}{M} \sum_{i=0}^{M-1} X\left(e^{j(\omega/M - 2\pi i/M)}\right). \end{aligned}$$

That is:

- $X(e^{j\omega})$  consists of copies of  $X_c(j\Omega)$  scaled by  $1/T$ , and frequency scaled by  $1/T$  and shifted by  $2\pi k$ , and
- $X_d(e^{j\omega})$  consists of copies of  $X(e^{j\omega})$  scaled by  $1/M$ , and frequency scaled by  $1/M$  and shifted by  $2\pi i$ .

# Downsampling example #1: ( $M = 2$ )

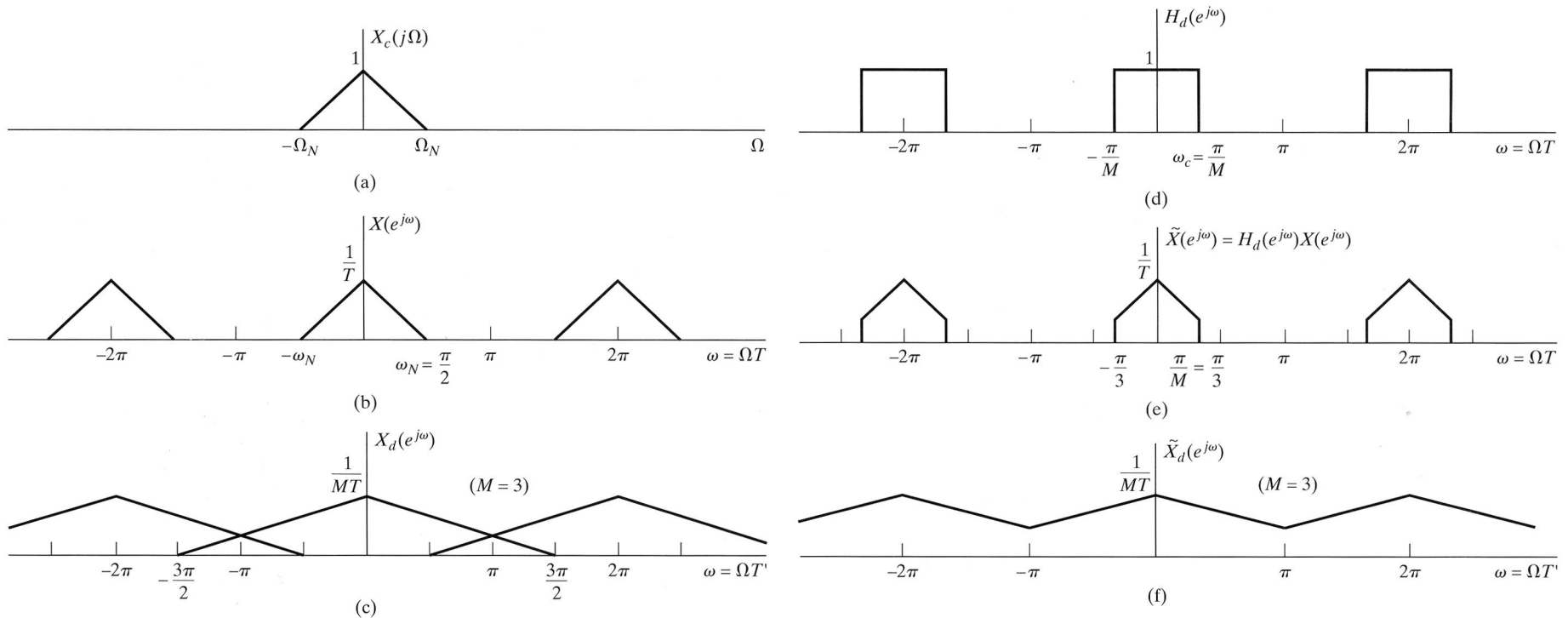


**Figure 4.21** Frequency-domain illustration of downsampling.

*(Oppenheim  
and Schaffer)*



# Downsampling example #2: ( $M = 3$ )



**Figure 4.22** (a)–(c) Downsampling with aliasing. (d)–(f) Downsampling with prefiltering to avoid aliasing.

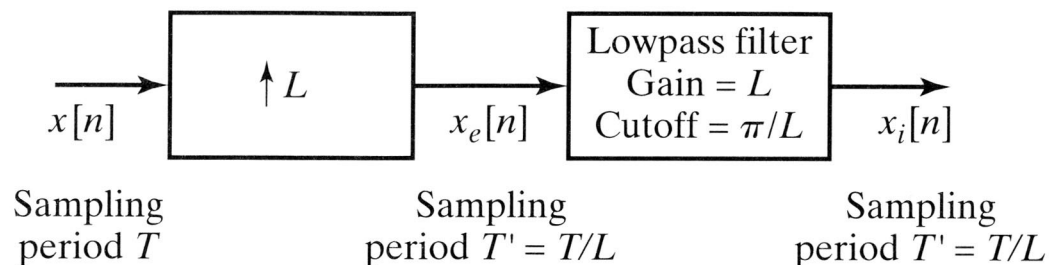
*(Oppenheim  
and Schaffer)*

## Increasing the sampling rate by an integer factor:

$$x_i[n] = x[n/L] = x_c(nT/L)$$

interpolation  
or upsampling

where  $n = 0, \pm L, \pm 2L, \dots$



(Oppenheim  
and Schaffer)

**Figure 4.24** General system for sampling rate increase by  $L$ .

$$x_e[n] = \begin{cases} x[n/L], & n = 0, \pm L, \pm 2L, \dots, \\ 0, & \text{otherwise,} \end{cases} \quad \text{expander}$$

or equivalently:

$$x_e[n] = \sum_{k=-\infty}^{\infty} x[k] \delta[n - kL].$$

Viewing the upsampling operation in the frequency domain:

$$\begin{aligned} X_e(e^{j\omega}) &= \sum_{n=-\infty}^{\infty} \left( \sum_{k=-\infty}^{\infty} x[k] \delta[n - kL] \right) e^{-j\omega n} \\ &= \sum_{k=-\infty}^{\infty} x[k] e^{-j\omega Lk} = X(e^{j\omega L}), \end{aligned}$$

we observe that the Fourier transform at the output of the expander is a frequency-scaled version of the input, i.e.,  $\omega$  is replaced by  $\omega L$  so that  $\omega$  is now normalized by  $\omega = \Omega T'$ .

As illustrated on the next slide,  $X_i(e^{j\omega})$  can be obtained from  $X_e(e^{j\omega})$  by correcting the amplitude scale from  $1/T$  to  $1/T'$  and by removing all the frequency-scaled images of  $X_c(j\Omega)$  except at integer multiples of  $2\pi$ , via a discrete-time lowpass filter.

# Upsampling example: ( $L = 2$ )

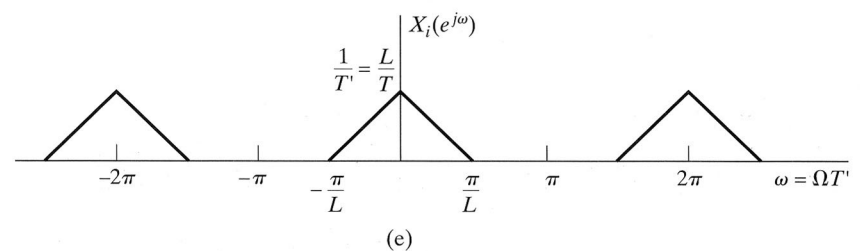
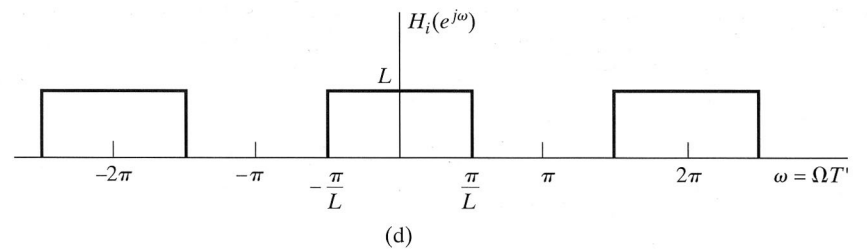
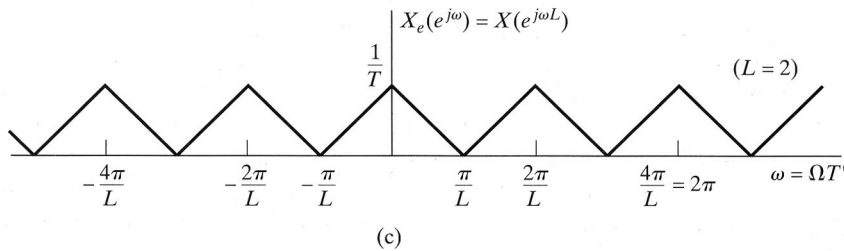
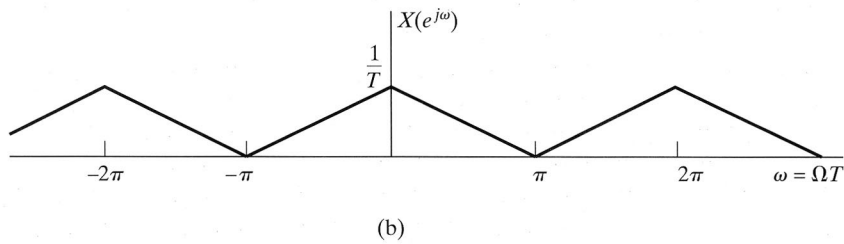
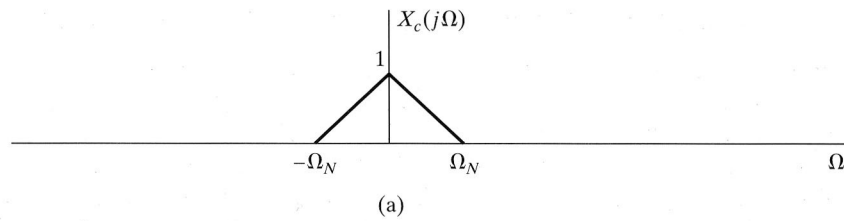


Figure 4.25 Frequency-domain illustration of interpolation.

(Oppenheim  
and Schaffer)

As was the case for the lowpass reconstruction filter in D/A conversion, this discrete-time lowpass filter can be viewed as an *interpolator* in the time domain, with the impulse response:

$$h_i[n] = \frac{\sin(\pi n/L)}{\pi n/L}.$$

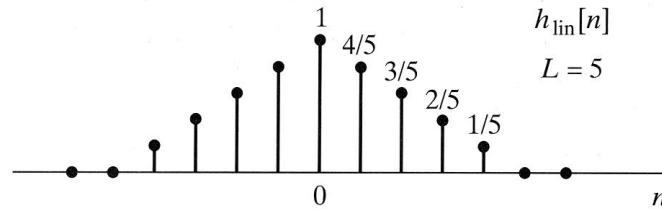
In practice, an ideal lowpass filter cannot be implemented exactly, but very good approximations can be designed.

In some cases, very simple interpolation processes are adequate, such as *linear interpolation*, which has the impulse response:

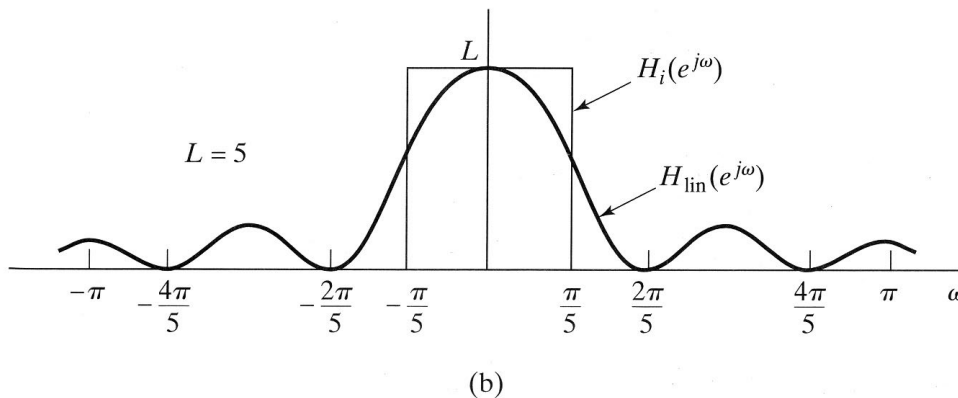
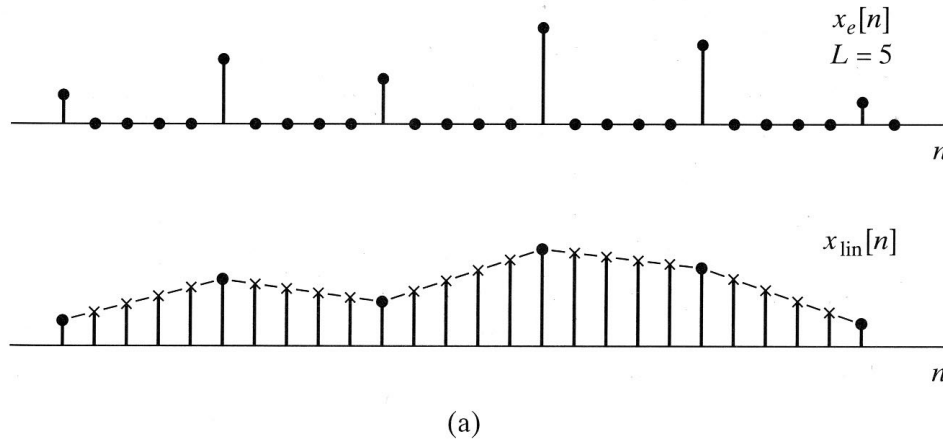
$$h_{\text{lin}}[n] = \begin{cases} 1 - |n|/L, & |n| \leq L, \\ 0, & \text{otherwise.} \end{cases}$$

Linear interpolation is illustrated on the next slide.

# Linear interpolation example: ( $L = 5$ )



**Figure 4.26** Impulse response for linear interpolation.

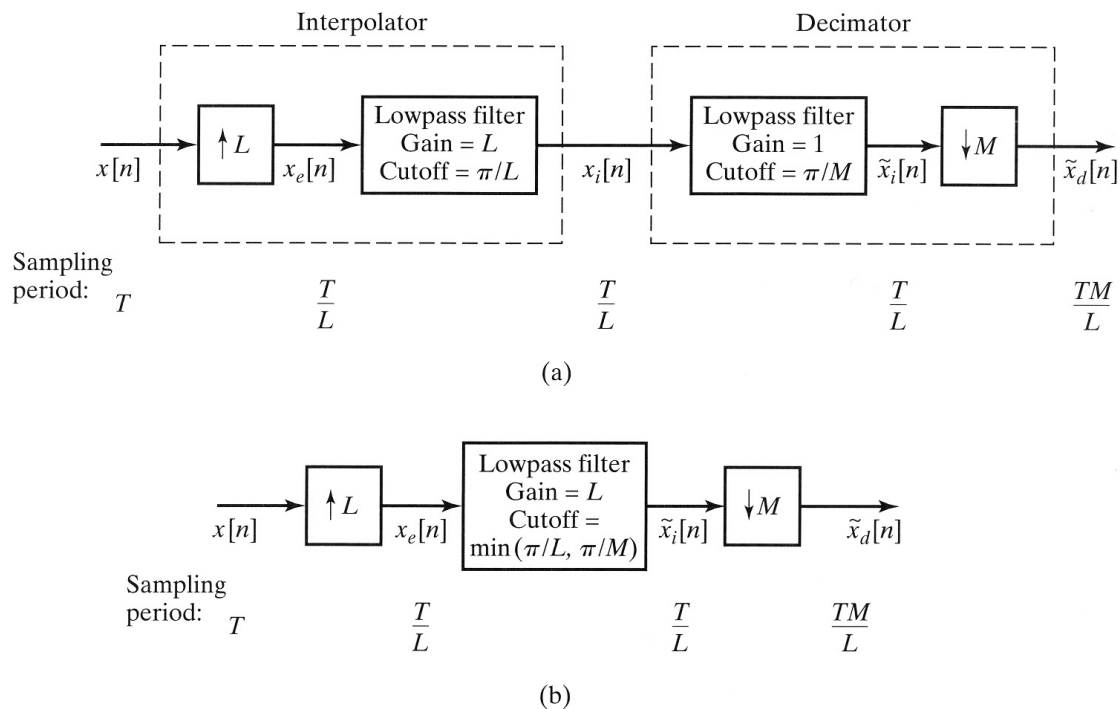


*(Oppenheim  
and Schaffer)*

**Figure 4.27** (a) Illustration of linear interpolation by filtering. (b) Frequency response of linear interpolator compared with ideal lowpass interpolation filter.

## Changing the sampling rate by a noninteger factor:

It is possible to combine decimation and interpolation to change the sampling rate by a noninteger, rational factor  $L/M$ .



**Figure 4.28** (a) System for changing the sampling rate by a noninteger factor. (b) Simplified system in which the decimation and interpolation filters are combined. *(Oppenheim and Schaffer)*

For example, if  $L = 101$  and  $M = 100$ , then the sampling rate will be increased by a factor of 1.01.