

Solutions – Chapter 4
Sampling of Continuous-Time Signals

4.1.

$$\begin{aligned} x[n] &= x_c(nT) \\ &= \sin\left(2\pi(100)n\frac{1}{400}\right) \\ &= \sin\left(\frac{\pi}{2}n\right) \end{aligned}$$

4.2. The discrete-time sequence

$$x[n] = \cos\left(\frac{\pi n}{4}\right)$$

results by sampling the continuous-time signal

$$x_c(t) = \cos(\Omega_o t).$$

Since $\omega = \Omega T$ and $T = 1/1000$ seconds, the signal frequency could be:

$$\Omega_o = \frac{\pi}{4} \cdot 1000 = 250\pi$$

or possibly:

$$\Omega_o = \left(2\pi + \frac{\pi}{4}\right) \cdot 1000 = 2250\pi.$$

4.3. (a) Since $x[n] = x_c(nT)$,

$$\begin{aligned} \frac{\pi n}{3} &= 4000\pi nT \\ T &= \frac{1}{12000} \end{aligned}$$

(b) No. For example, since

$$\cos\left(\frac{\pi}{3}n\right) = \cos\left(\frac{7\pi}{3}n\right),$$

T can be $7/12000$.

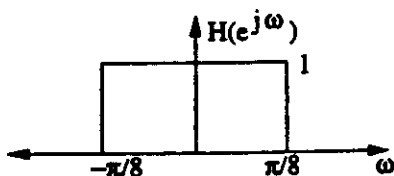
4.4. (a) Letting $T = 1/100$ gives

$$\begin{aligned} x[n] &= x_c(nT) \\ &= \sin\left(20\pi n\frac{1}{100}\right) + \cos\left(40\pi n\frac{1}{100}\right) \\ &= \sin\left(\frac{\pi n}{5}\right) + \cos\left(\frac{2\pi n}{5}\right) \end{aligned}$$

(b) No, another choice is $T = 11/100$:

$$\begin{aligned} x[n] &= x_c(nT) \\ &= \sin\left(20\pi n\frac{11}{100}\right) + \cos\left(40\pi n\frac{11}{100}\right) \\ &= \sin\left(\frac{11\pi n}{5}\right) + \cos\left(\frac{22\pi n}{5}\right) \\ &= \sin\left(\frac{\pi n}{5}\right) + \cos\left(\frac{2\pi n}{5}\right) \end{aligned}$$

4.5. A plot of $H(e^{j\omega})$ appears below.



(a)

$$x_c(t) = 0, \quad |\Omega| \geq 2\pi \cdot 5000$$

The Nyquist rate is 2 times the highest frequency. $\Rightarrow T = \frac{1}{10,000}$ sec. This avoids all aliasing in the C/D converter.

(b)

$$\begin{aligned} \frac{1}{T} &= 10 \text{ kHz} \\ \omega &= T\Omega \\ \frac{\pi}{8} &= \frac{1}{10,000} \Omega_c \\ \Omega_c &= 2\pi \cdot 625 \text{ rad/sec} \\ f_c &= 625 \text{ Hz} \end{aligned}$$

(c)

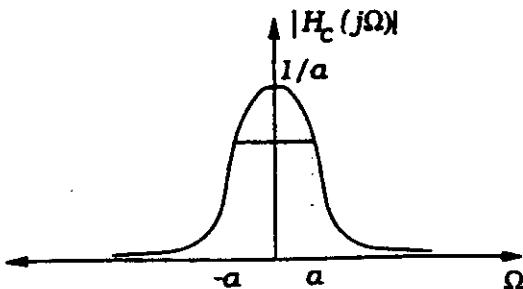
$$\begin{aligned} \frac{1}{T} &= 20 \text{ kHz} \\ \omega &= T\Omega \\ \frac{\pi}{8} &= \frac{1}{20,000} \Omega_c \\ \Omega_c &= 2\pi \cdot 1250 \text{ rad/sec} \\ f_c &= 1250 \text{ Hz} \end{aligned}$$

4.6. (a) The Fourier transform of the filter impulse response

$$\begin{aligned} H_c(j\Omega) &= \int_{-\infty}^{\infty} h_c(t) e^{-j\Omega t} dt \\ &= \int_0^{\infty} a^{-at} e^{-j\Omega t} dt \\ &= \frac{1}{a + j\Omega} \end{aligned}$$

So, we take the magnitude

$$|H_c(j\Omega)| = \left(\frac{1}{a^2 + \Omega^2} \right)^{\frac{1}{2}}$$



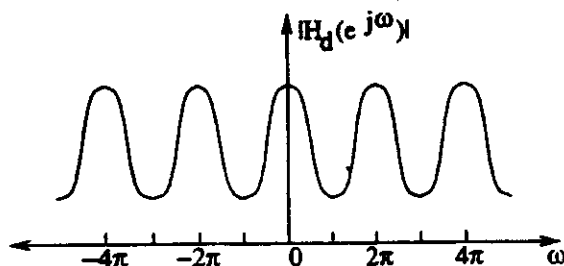
(b) Sampling the filter impulse response in (a), the discrete-time filter is described by

$$\begin{aligned} h_d[n] &= T e^{-anT} u[n] \\ H_d(e^{j\omega}) &= \sum_{n=0}^{\infty} T e^{-anT} e^{-j\omega n} \\ &= \frac{T}{1 - e^{-aT} e^{-j\omega}} \end{aligned}$$

Taking the magnitude of this response

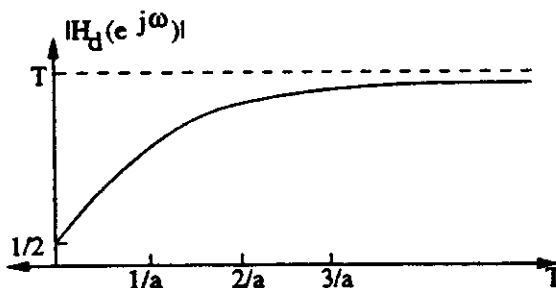
$$|H_d(e^{j\omega})| = \frac{T}{(1 - 2e^{-aT} \cos(\omega) + e^{-2aT})^{\frac{1}{2}}}$$

Note that the frequency response of the discrete-time filter is periodic, with period 2π .



(c) The minimum occurs at $\omega = \pi$. The corresponding value of the frequency response magnitude is

$$\begin{aligned} |H_d(e^{j\pi})| &= \frac{T}{(1 + 2e^{-aT} + e^{-2aT})^{\frac{1}{2}}} \\ &= \frac{T}{1 + e^{-aT}} \end{aligned}$$



4.7. The continuous-time signal contains an attenuated replica of the original signal with a delay of τ_d .

$$x_c(t) = s_c(t) + \alpha s_c(t - \tau_d)$$

(a) Taking the Fourier transform of the analog signal:

$$X_c(j\Omega) = S_c(j\Omega) \cdot (1 + \alpha e^{-j\tau_d \Omega})$$

Note that $X_c(j\Omega)$ is zero for $|\Omega| > \pi/T$. Sampling the continuous-time signal yields the discrete-time sequence, $x[n]$. The Fourier transform of the sequence is

$$\begin{aligned} X(e^{j\omega}) &= \frac{1}{T} \sum_{r=-\infty}^{\infty} S_c\left(\frac{j\omega}{T} + j\frac{2\pi r}{T}\right) \\ &\quad + \frac{\alpha}{T} \sum_{r=-\infty}^{\infty} S_c\left(\frac{j\omega}{T} + j\frac{2\pi r}{T}\right) e^{-j\tau_d\left(\frac{\omega}{T} + \frac{2\pi r}{T}\right)} \end{aligned}$$

(b) The desired response:

$$H(j\Omega) = \begin{cases} 1 + \alpha e^{-j\tau_d\Omega}, & \text{for } |\Omega| \leq \frac{\pi}{T} \\ 0, & \text{otherwise} \end{cases}$$

Using $\omega = \Omega T$, we obtain a discrete-time system which simulates the above response:

$$H(e^{j\omega}) = 1 + \alpha e^{-j\frac{\tau_d\omega}{T}}$$

(c) We need to take the inverse Fourier transform of the discrete-time impulse response of part (b).

$$\begin{aligned} h[n] &= \frac{1}{2\pi} \int_{-\pi}^{\pi} H(e^{j\omega}) e^{j\omega n} d\omega \\ &= \frac{1}{2\pi} \int_{-\pi}^{\pi} (1 + \alpha e^{-j\frac{\tau_d\omega}{T}}) e^{j\omega n} d\omega \end{aligned}$$

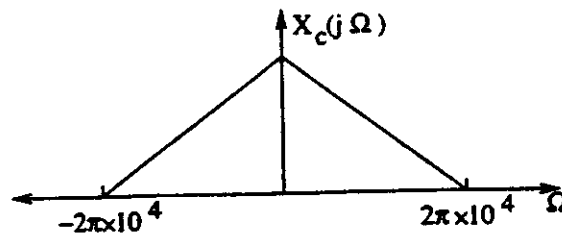
(i) Consider the case when $\tau_d = T$:

$$\begin{aligned} h[n] &= \frac{1}{2\pi} \int_{-\pi}^{\pi} (e^{j\omega n} + \alpha e^{j\omega(n-1)}) d\omega \\ &= \frac{\sin(\pi n)}{\pi n} + \frac{\alpha \sin[\pi(n-1)]}{\pi(n-1)} \\ &= \delta[n] + \alpha \delta[n-1] \end{aligned}$$

(ii) For $\tau_d = T/2$:

$$\begin{aligned} h[n] &= \frac{1}{2\pi} \int_{-\pi}^{\pi} (e^{j\omega n} + \alpha e^{j\omega(n-\frac{1}{2})}) d\omega \\ &= \frac{\sin(\pi n)}{\pi n} + \frac{\alpha \sin[\pi(n-\frac{1}{2})]}{\pi(n-\frac{1}{2})} \\ &= \delta[n] + \frac{\alpha \sin[\pi(n-\frac{1}{2})]}{\pi(n-\frac{1}{2})} \end{aligned}$$

4.8. A plot of $X_c(j\Omega)$ appears below.



(a) For $x_c(t)$ to be recoverable from $x[n]$, the transform of the discrete signal must have no aliasing. When sampling, the radian frequency is related to the analog frequency by

$$\omega = \Omega T.$$

No aliasing will occur if the sampling interval satisfies the Nyquist Criterion. Thus, for the band-limited signal, $x_c(t)$, we should select T as:

$$T \leq \frac{1}{2 \times 10^4}.$$

- (b) Assuming that the system is linear and time-invariant, the convolution sum describes the input-output relationship.

$$y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k]$$

We are given

$$\begin{aligned} y[n] &= T \sum_{k=-\infty}^n x[k] \\ &= T \sum_{k=-\infty}^{\infty} x[k]u[n-k] \end{aligned}$$

Hence, we may infer that the impulse response of the system

$$h[n] = T \cdot u[n].$$

- (c) We use the expression for $y[n]$ as given and examine the limit

$$\begin{aligned} \lim_{n \rightarrow \infty} y[n] &= \lim_{n \rightarrow \infty} T \cdot \sum_{k=-\infty}^n x[k] \\ &= T \cdot \sum_{k=-\infty}^{\infty} x[k] \end{aligned}$$

Recall the analysis equation for the Fourier transform:

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

Hence,

$$\lim_{n \rightarrow \infty} y[n] = T \cdot X(e^{j\omega})|_{\omega=0}$$

- (d) We use the result from part (c). Noting that

$$X(e^{j\omega}) = \frac{1}{T} \sum_{r=-\infty}^{\infty} X_c\left(\frac{j\omega}{T} + \frac{j2\pi r}{T}\right).$$

Thus, we have

$$T \cdot X(e^{j\omega})|_{\omega=0} = \sum_{r=-\infty}^{\infty} X_c\left(\frac{j2\pi r}{T}\right)$$

From the given information, we seek a value of T such that:

$$\begin{aligned} \sum_{r=-\infty}^{\infty} X_c\left(\frac{j2\pi r}{T}\right) &= \int_{-\infty}^{\infty} x_c(t) dt \\ &= X_c(j\Omega)|_{\Omega=0} \end{aligned}$$

For the final equality to be true, there must be no contribution from the terms for which $r \neq 0$. That is, we require no aliasing at $\Omega = 0$. Since we are only interested in preserving the spectral component at $\Omega = 0$, we may sample at a rate which is lower than the Nyquist rate. The maximum value of T to satisfy these conditions is

$$T \leq \frac{1}{1 \times 10^4}.$$

- 4.9. (a) Since $X(e^{j\omega}) = X(e^{j(\omega-\pi)})$, $X(e^{j\omega})$ is periodic with period π .
 (b) Using the inverse DTFT,

$$\begin{aligned}
 x[n] &= \frac{1}{2\pi} \int_{(2\pi)} X(e^{j\omega}) e^{j\omega n} d\omega \\
 &= \frac{1}{2\pi} \int_{(2\pi)} X(e^{j(\omega-\pi)}) e^{j\omega n} d\omega \\
 &= \frac{1}{2\pi} \int_{(2\pi)} X(e^{j\omega}) e^{j(\omega+\pi)n} d\omega \\
 &= \frac{1}{2\pi} e^{j\pi n} \int_{(2\pi)} X(e^{j\omega}) e^{j\omega n} d\omega \\
 &= (-1)^n x[n].
 \end{aligned}$$

All odd samples of $x[n] = 0$, because $x[n] = -x[n]$. Hence $x[3] = 0$.

- (c) Yes, $y[n]$ contains all even samples of $x[n]$, and all odd samples of $x[n]$ are 0.

$$x[n] = \begin{cases} y[n/2], & n \text{ even} \\ 0, & \text{otherwise} \end{cases}$$

- 4.10. Use $x[n] = x_c(nT)$, and simplify:

- (a) $x[n] = \cos(2\pi n/3)$.
 (b) $x[n] = \sin(4\pi n/3) = -\sin(2\pi n/3)$
 (c) $x[n] = \frac{\sin(2\pi n/5)}{\pi n/5000}$

- 4.11. (a) Pick T such that

$$x[n] = x_c(nT) = \sin(10\pi nT) = \sin(\pi n/4) \implies T = 1/40$$

There are other choices. For example, by realizing that $\sin(\pi n/4) = \sin(9\pi n/4)$, we find $T = 9/40$.

- (b) Choose $T = 1/20$ to make $x[n] = x_c(nT)$. This is unique.

- 4.12. (a) Notice first that $H(e^{j\omega}) = 10j\omega$, $-\pi \leq \omega < \pi$.

- (i) After sampling,

$$\begin{aligned}
 x[n] &= \cos\left(\frac{3\pi}{5}n\right), \\
 y[n] &= |H(e^{j\frac{3\pi}{5}})| \cos\left(\frac{3\pi}{5}n + \angle H(e^{j\frac{3\pi}{5}})\right) \\
 &= 6\pi \cos\left(\frac{3\pi}{5}n + \frac{\pi}{2}\right) \\
 &= -6\pi \sin\left(\frac{3\pi}{5}n\right) \\
 y_c(t) &= -6\pi \sin(6\pi t).
 \end{aligned}$$

- (ii) After sampling, $x[n] = \cos(\frac{7\pi}{5}n) = \cos(\frac{3\pi}{5}n)$, so again, $y_c(t) = -6\pi \sin(6\pi t)$.

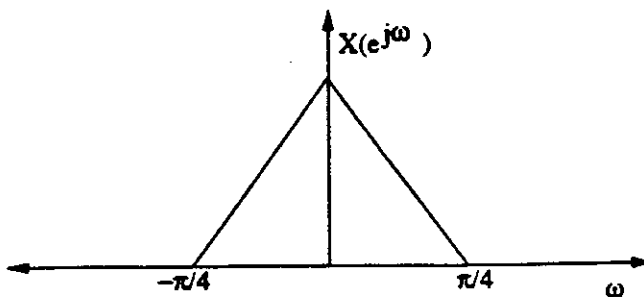
- (b) $y_c(t)$ is what you would expect from a differentiator in the first case but not in the second case. This is because aliasing has occurred in the second case.

4.13. (a)

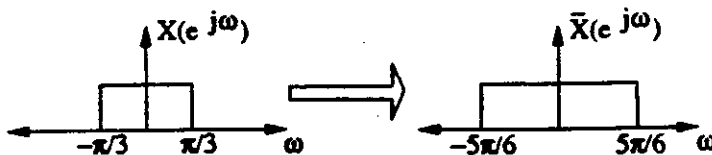
$$\begin{aligned}
 x_c(t) &= \sin\left(\frac{\pi}{20}t\right) \\
 y_c(t) &= \sin\left(\frac{\pi}{20}(t-5)\right) \\
 &= \sin\left(\frac{\pi}{20}t - \frac{\pi}{4}\right) \\
 y[n] &= \sin\left(\frac{\pi n}{2} - \frac{\pi}{4}\right)
 \end{aligned}$$

(b) We get the same result as before:

$$\begin{aligned}
 x_c(t) &= \sin\left(\frac{\pi}{10}t\right) \\
 y_c(t) &= \sin\left(\frac{\pi}{10}(t-2.5)\right) \\
 &= \sin\left(\frac{\pi}{10}t - \frac{\pi}{4}\right) \\
 y[n] &= \sin\left(\frac{\pi n}{2} - \frac{\pi}{4}\right)
 \end{aligned}$$

(c) The sampling period T is not limited by the continuous time system $h_c(t)$.4.14. There is no loss of information if $X(e^{j\omega/2})$ and $X(e^{j(\omega/2-\pi)})$ do not overlap. This is true for (b), (d), (e).4.15. The output $x_r[n] = x[n]$ if no aliasing occurs as result of downsampling. That is, $X(e^{j\omega}) = 0$ for $\pi/3 \leq |\omega| \leq \pi$.(a) $x[n] = \cos(\pi n/4)$. $X(e^{j\omega})$ has impulses at $\omega = \pm\pi/4$, so there is no aliasing. $x_r[n] = x[n]$.(b) $x[n] = \cos(\pi n/2)$. $X(e^{j\omega})$ has impulses at $\omega = \pm\pi/2$, so there is aliasing. $x_r[n] \neq x[n]$.(c) A sketch of $X(e^{j\omega})$ is shown below. Clearly there will be no aliasing and $x_r[n] = x[n]$.

4.16. (a) In the frequency domain, we have



$$\frac{M}{L} = \frac{5\pi/6}{\pi/3} = \frac{5}{2}$$

This is unique.

(b) One choice is

$$\frac{M}{L} = \frac{\pi/2}{3\pi/4} = \frac{2}{3}$$

However, this is not unique. We can also write $\tilde{x}_d[n] = \cos(\frac{5\pi}{2}n)$, so another choice is

$$\frac{M}{L} = \frac{5\pi/2}{3\pi/4} = \frac{10}{3}$$

4.17. (a) In the frequency domain,

$$X(e^{j\omega}) = \begin{cases} 1, & |\omega| < 2\pi/3 \\ 0, & 2\pi/3 < |\omega| < \pi \end{cases}$$

After the sampling rate change,

$$\tilde{X}_d(e^{j\omega}) = \begin{cases} 4/3, & |\omega| < \pi/2 \\ 0, & \pi/2 < |\omega| < \pi \end{cases}$$

which leads to

$$x[n] = \frac{4 \sin(\pi n/2)}{\pi n}$$

(b) Upsampling by 3 and low-pass filtering $x[n] = \sin(3\pi n/4)$ results in $\sin(\pi n/4)$. Downsampling by 5 gives us $\tilde{x}_d[n] = \sin(5\pi n/4) = -\sin(3\pi n/4)$.

4.18. For the condition to be satisfied, we have to ensure that $\omega_0/L \leq \min(\pi/L, \pi/M)$, so that the lowpass filtering does not cut out part of the spectrum.

(a) $\omega_0/2 \leq \pi/3 \Rightarrow \omega_{0,max} = 2\pi/3$.

(b) $\omega_0/3 \leq \pi/5 \Rightarrow \omega_{0,max} = 3\pi/5$.

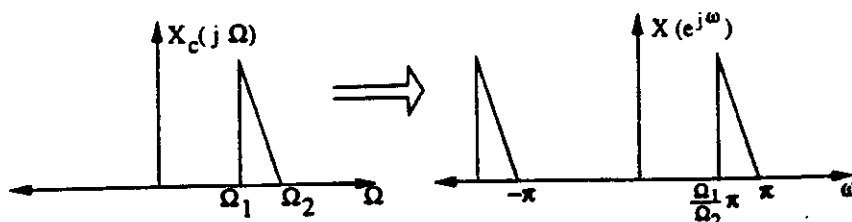
(c) Since $L > M$, there is no chance of aliasing. Hence $\omega_{0,max} = \pi$.

4.19. The nyquist sampling property must be satisfied: $T \leq \pi/\Omega_0$.

4.20. (a) The Nyquist sampling property must be satisfied: $T \leq \pi/\Omega_0 \Rightarrow F_s \geq 2000$.

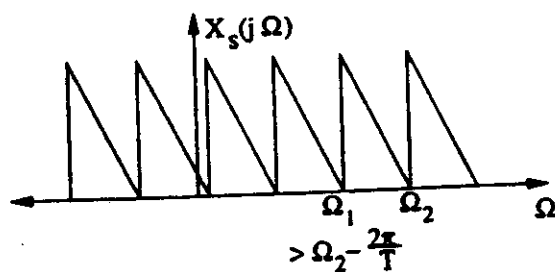
(b) We'd have to sample so that $X(e^{j\omega})$ lies between $|\omega| < \pi/2$. So $F_s \geq 4000$.

4.21. (a) Keeping in mind that after sampling, $\omega = \Omega T$, the Fourier transform of $x[n]$ is

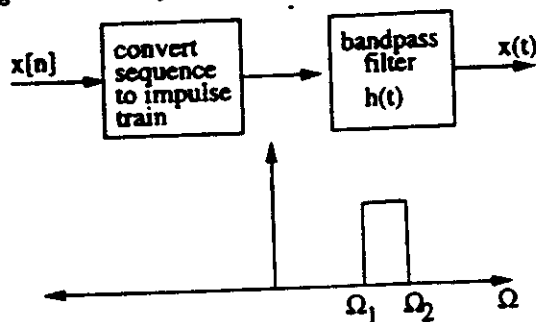


(b) A straight-forward application of the Nyquist criterion would lead to an incorrect conclusion that the sampling rate is at least twice the maximum frequency of $x_c(t)$, or $2\Omega_2$. However, since the spectrum is bandpass, we only need to ensure that the replications in frequency which occur as a result of sampling do not overlap with the original. (See the following figure of $X_s(j\Omega)$.) Therefore, we only need to ensure

$$\Omega_2 - \frac{2\pi}{T} < \Omega_1 \Rightarrow T < \frac{2\pi}{\Delta\Omega}$$

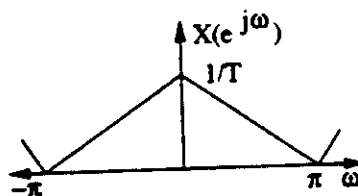


(c) The block diagram along with the frequency response of $h(t)$ is shown here:

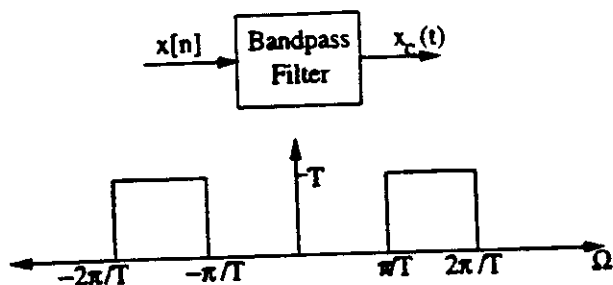


4.22. (a)

$$\omega = \Omega T, \quad T = \frac{2\pi}{\Omega_0}$$



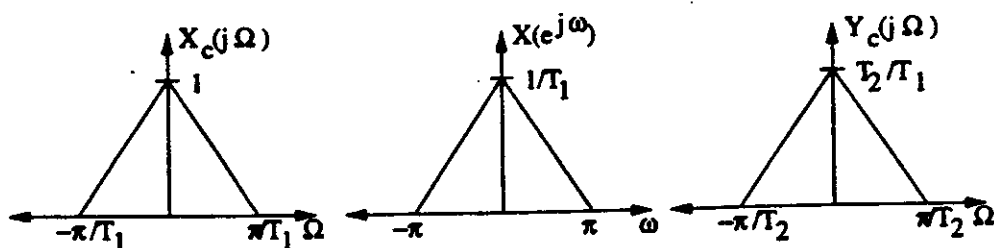
(b) To recover simply filter out the undesired parts of $X(e^{j\omega})$.



(c)

$$T \leq \frac{2\pi}{\Omega_0}$$

4.23. In the frequency domain, we have



$$x_c(t) = 0, \quad |\Omega| \geq \frac{\pi}{T_1}$$

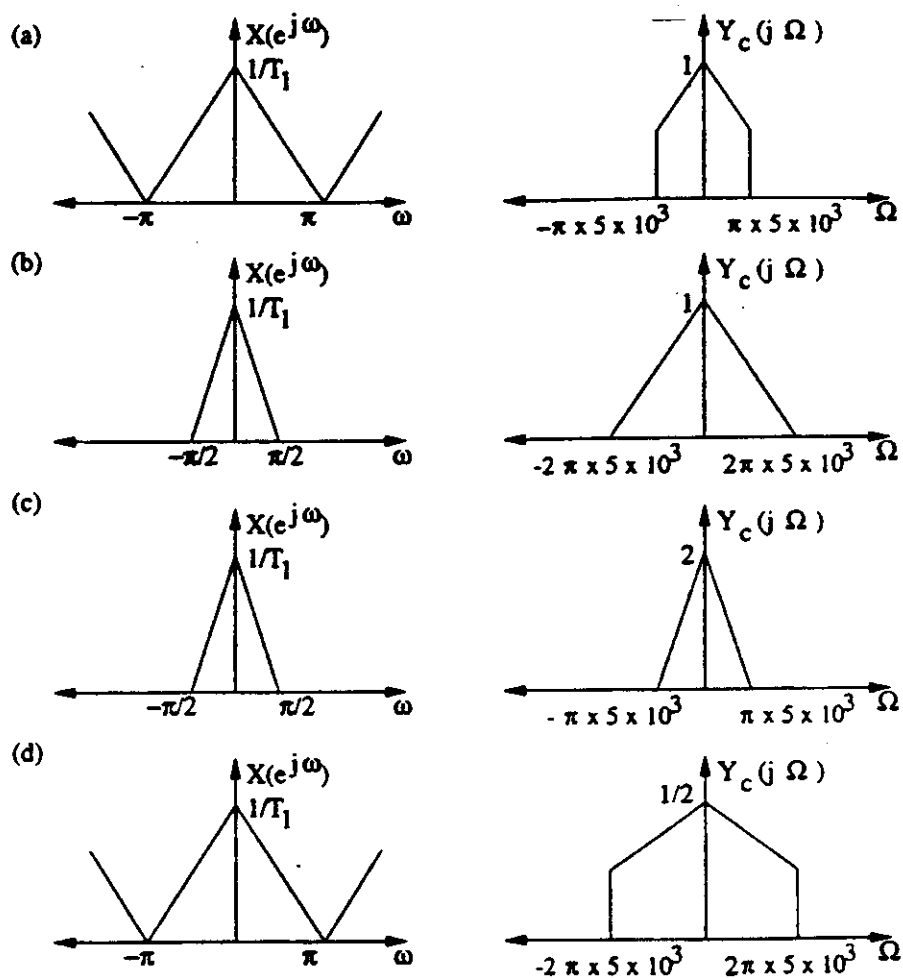
Therefore, since we are sampling this $x_c(t)$ at the Nyquist frequency $x[n]$ will be full band and unaliased.

$$x[n] = x_c(nT_1)$$

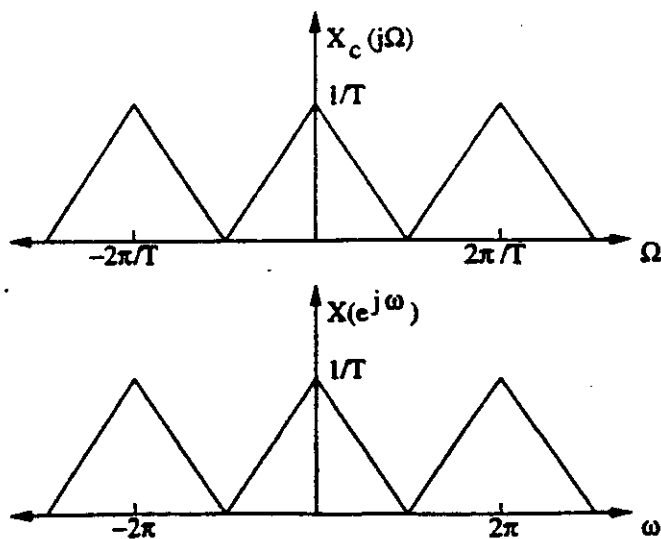
$y_c(t)$ is a band-limited interpolation of $x[n]$ at a different period. Since no aliasing occurs at $x[n]$, the spectrum of $y_c(t)$ will be a frequency axis scaling of the spectrum of $x_c(t)$ for $T_1 > T_2$ or $T_1 < T_2$. As we show in the figure,

$$y_c(t) = \frac{T_2}{T_1} x_c\left(\frac{T_2}{T_1}t\right)$$

4.24. The Fourier transform of $y_c(t)$ is sketched below for each case.



4.25. (a) $x_s(t) = x_c(t)s(t) \Rightarrow X_s(j\Omega) = X_c(j\Omega) * s(j\Omega)$



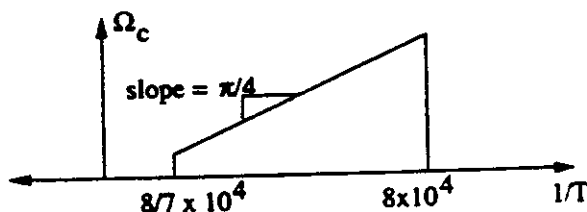
- (b) Since $H_d(e^{j\omega})$ is an ideal lowpass filter with $\omega_c = \frac{\pi}{4}$, we don't care about any signal aliasing that occurs in the region $\frac{\pi}{4} \leq \omega \leq \pi$. We require:

$$\begin{aligned}\frac{2\pi}{T} - 2\pi \cdot 10000 &\geq \frac{\pi}{4T} \\ \frac{1}{T} &\geq \frac{8}{7} \cdot 10000 \\ T &\leq \frac{7}{8} \times 10^{-4} \text{ sec}\end{aligned}$$

Also, once all of the signal lies in the range $|\omega| \leq \frac{\pi}{4}$, the filter will be ineffective, i.e., $\frac{\pi}{4} \leq T(2\pi \times 10^4)$. So, $T \geq 12.5 \mu\text{sec}$.

(c)

$$\Omega = \frac{\omega}{T} \Rightarrow \Omega_c = \frac{\pi}{4T}$$



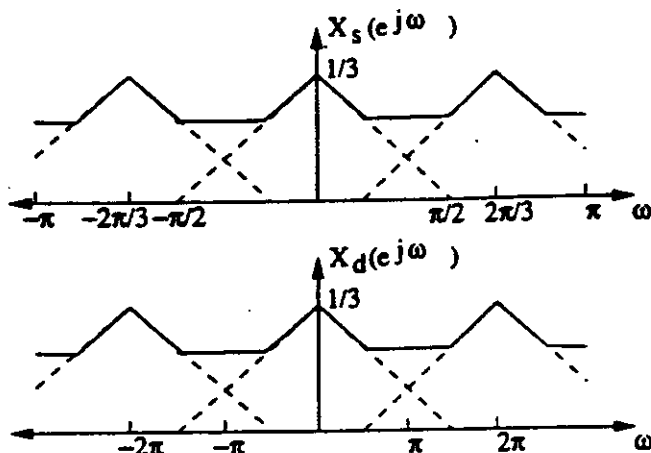
4.26. First we show that $X_s(e^{j\omega})$ is just a sum of shifted versions of $X(e^{j\omega})$:

$$\begin{aligned}x_s[n] &= \begin{cases} x[n], & n = Mk, \quad k = 0, \pm 1, \pm 2 \\ 0, & \text{otherwise} \end{cases} \\ &= \left(\frac{1}{M} \sum_{k=0}^{M-1} e^{j(2\pi kn/M)} \right) x[n] \\ X_s(e^{j\omega}) &= \sum_{n=-\infty}^{\infty} x_s[n] e^{-j\omega n} \\ &= \sum_{n=-\infty}^{\infty} \frac{1}{M} \sum_{k=0}^{M-1} x[n] e^{j(2\pi kn/M)} e^{-j\omega n} \\ &= \frac{1}{M} \sum_{k=0}^{M-1} \sum_{n=-\infty}^{\infty} x[n] e^{-j[\omega - (2\pi k/M)]n} \\ &= \frac{1}{M} \sum_{k=0}^{M-1} X(e^{j[\omega - (2\pi k/M)]})\end{aligned}$$

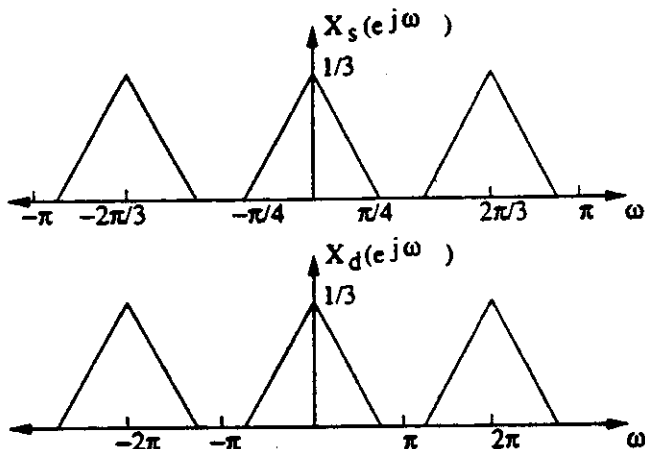
Additionally, $X_d(e^{j\omega})$ is simply $X_s(e^{j\omega})$ with the frequency axis expanded by a factor of M :

$$\begin{aligned}X_d(e^{j\omega}) &= \sum_{n=-\infty}^{\infty} X_s[Mn] e^{-j\omega n} \\ &= \sum_{l=-\infty}^{\infty} x_s[l] e^{-j(\omega/M)l} \\ &= X_s(e^{j(\omega/M)})\end{aligned}$$

(a) (i) $X_s(e^{j\omega})$ and $X_d(e^{j\omega})$ are sketched below for $M = 3$, $\omega_H = \pi/2$.



(ii) $X_s(e^{j\omega})$ and $X_d(e^{j\omega})$ are sketched below for $M = 3$, $\omega_H = \pi/4$.



(b) From the definition of $X_s(e^{j\omega})$, we see that there will be no aliasing if the signal is bandlimited to π/M . In this problem, $M = 3$. Thus the maximum value of ω_H is $\pi/3$.

4.27. Parseval's Theorem:

$$\sum_{n=-\infty}^{\infty} |x[n]|^2 = \frac{1}{2\pi} \int_{-\pi}^{\pi} |X(e^{j\omega})|^2 d\omega$$

When we upsample, the added samples are zeros, so the upsampled signal $x_u[n]$ has the same energy as the original $x[n]$:

$$\sum_{n=-\infty}^{\infty} |x[n]|^2 = \sum_{n=-\infty}^{\infty} |x_u[n]|^2,$$

and by Parseval's theorem:

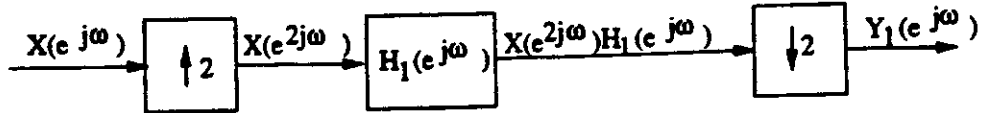
$$\frac{1}{2\pi} \int_{-\pi}^{\pi} |X(e^{j\omega})|^2 d\omega = \frac{1}{2\pi} \int_{-\pi}^{\pi} |X_u(e^{j\omega})|^2 d\omega.$$

Hence the amplitude of the Fourier transform does not change.

When we downsample, the downsampled signal $x_d[n]$ has less energy than the original $x[n]$ because some samples are discarded. Hence the amplitude of the Fourier transform will change after downsampling.

- 4.28. (a) Yes, the system is linear because each of the subblocks is linear. The C/D step is defined by $x[n] = x_c(nT)$, which is clearly linear. The DT system is an LTI system. The D/C step consists of converting the sequence to impulses and of CT LTI filtering, both of which are linear.
- (b) No, the system is not time-invariant.
For example, suppose that $h[n] = \delta[n]$, $T = 5$ and $x_c(t) = 1$ for $-1 \leq t \leq 1$. Such a system would result in $x[n] = \delta[n]$ and $y_c(t) = \text{sinc}(\pi/5)$. Now suppose we delay the input to be $x_c(t-2)$. Now $x[n] = 0$ and $y_c(t) = 0$.

4.29. We can analyze the system in the frequency domain:



$Y_1(e^{j\omega})$ is $X(e^{j2\omega})H_1(e^{j\omega})$ downsampled by 2:

$$\begin{aligned} Y_1(e^{j\omega}) &= \frac{1}{2} \left\{ X(e^{j2\omega/2})H_1(e^{j\omega/2}) + X(e^{j(2\omega-2\pi)/2})H_1(e^{j(\omega-\pi)}) \right\} \\ &= \frac{1}{2} \left\{ X(e^{j\omega})H_1(e^{j\omega/2}) + X(e^{j(\omega-2\pi)})H_1(e^{j(\omega/2-\pi)}) \right\} \\ &= \frac{1}{2} \left\{ H_1(e^{j\omega/2}) + H_1(e^{j(\omega/2-\pi)}) \right\} X(e^{j\omega}) \\ &= H_2(e^{j\omega})X(e^{j\omega}) \\ H_2(e^{j\omega}) &= \frac{1}{2} \left\{ H_1(e^{j\omega/2}) + H_1(e^{j(\omega/2-\pi)}) \right\} \end{aligned}$$

4.30.

$$\begin{aligned} X_c(j\Omega) &= 0, & |\Omega| &\geq 4000\pi \\ Y(j\Omega) &= |\Omega|X_c(j\Omega), & 1000\pi &\leq |\Omega| \leq 2000\pi \end{aligned}$$

Since only half the frequency band of $X_c(j\Omega)$ is needed, we can alias everything past $\Omega = 2000\pi$. Hence, $T = 1/3000$ s.

Now that T is set, figure out $H(e^{j\omega})$ band edges.

$$\begin{aligned} \omega_1 = \Omega_1 T &\Rightarrow \omega_1 = 2\pi \cdot 500 \cdot \frac{1}{3000} \Rightarrow \omega_1 = \frac{\pi}{3} \\ \omega_2 = \Omega_2 T &\Rightarrow \omega_2 = 2\pi \cdot 1000 \cdot \frac{1}{3000} \Rightarrow \omega_2 = \frac{2\pi}{3} \end{aligned}$$

$$H(e^{j\omega}) = \begin{cases} |\omega| & \frac{\pi}{3} \leq |\omega| \leq \frac{2\pi}{3} \\ 0 & 0 \leq |\omega| < \frac{\pi}{3}, \frac{2\pi}{3} < |\omega| \leq \pi \end{cases}$$

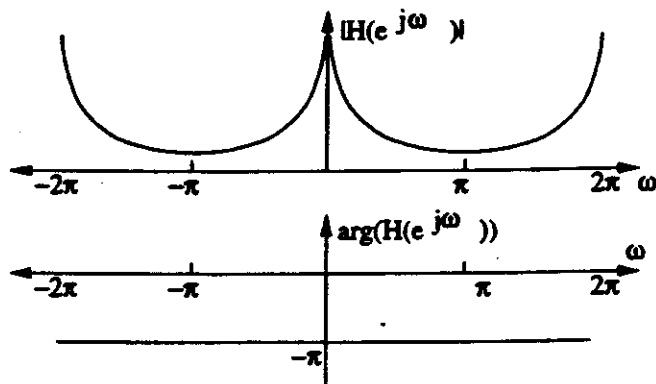
4.31.

$$X_c(j\Omega) = 0, \quad |\Omega| > \frac{\pi}{T}$$

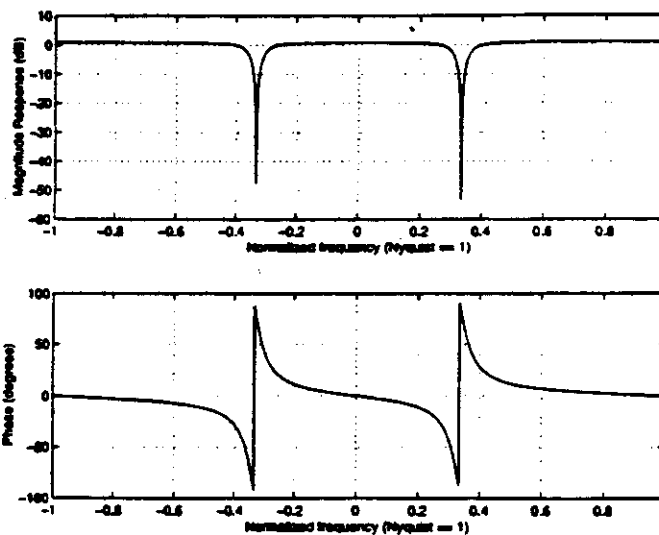
$$y_r(t) = \int_{-\infty}^t x_c(\tau) d\tau \Rightarrow H_c(j\Omega) = \frac{1}{j\Omega}$$

In discrete-time, we want

$$H(e^{j\omega}) = \begin{cases} \frac{1}{j\omega}, & -\pi \leq \omega \leq \pi \\ 0, & \text{otherwise} \end{cases}$$



4.32. (a) The highest frequency is $\pi/T = \pi \times 10000$.



(b) To filter the 60Hz out,

$$\omega_0 = T\Omega = \frac{1}{10,000} \cdot 2\pi \cdot 60 = \frac{3\pi}{250}$$

4.33.

$$\begin{aligned} y[n] &= x^2[n] \\ Y(e^{j\omega}) &= X(e^{j\omega}) * X(e^{j\omega}) \end{aligned}$$

therefore, $Y(e^{j\omega})$ will occupy twice the frequency band that $X(e^{j\omega})$ does if no aliasing occurs.

If $Y(e^{j\omega}) \neq 0$, $-\pi < \omega < \pi$, then $X(e^{j\omega}) \neq 0$, $-\frac{\pi}{2} < \omega < \frac{\pi}{2}$ and so $X(j\Omega) = 0$, $|\Omega| \geq 2\pi(1000)$.

Since $\omega = \Omega T$,

$$\frac{\pi}{2} \geq T \cdot 2\pi(1000)$$

$$T \leq \frac{1}{4000}$$

- 4.34. (a) Since there is no aliasing involved in this process, we may choose T to be any value. Choose $T = 1$ for simplicity. $X_c(j\Omega) = 0, |\Omega| \geq \pi/T$. Since $Y_c(j\Omega) = H_c(j\Omega)X_c(j\Omega)$, $Y_c(j\Omega) = 0, |\Omega| \geq \pi/T$. Therefore, there will be no aliasing problems in going from $y_c(t)$ to $y[n]$. Recall the relationship $\omega = \Omega T$. We can simply use this in our system conversion:

$$\begin{aligned} H(e^{j\omega}) &= e^{-j\omega/2} \\ H(j\Omega) &= e^{-j\Omega T/2} \\ &= e^{-j\Omega/2}, \quad T = 1 \end{aligned}$$

Note that the choice of T and therefore $H(j\Omega)$ is not unique.

(b)

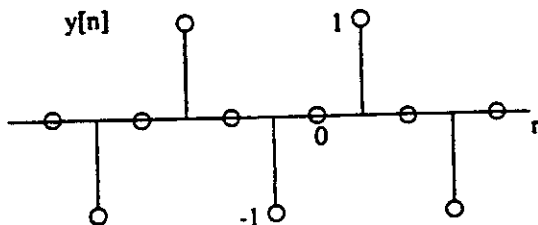
$$\begin{aligned} \cos\left(\frac{5\pi}{2}n - \frac{\pi}{4}\right) &= \frac{1}{2} \left[e^{j(\frac{5\pi}{2}n - \frac{\pi}{4})} + e^{-j(\frac{5\pi}{2}n - \frac{\pi}{4})} \right] \\ &= \frac{1}{2} e^{-j(\pi/4)} e^{j(5\pi/2)n} + \frac{1}{2} e^{j(\pi/4)} e^{-j(5\pi/2)n} \end{aligned}$$

Since $H(e^{j\omega})$ is an LTI system, we can find the response to each of the two eigenfunctions separately.

$$y[n] = \frac{1}{2} e^{-j(\pi/4)} H(e^{j(5\pi/2)}) e^{j(5\pi/2)n} + \frac{1}{2} e^{j(\pi/4)} H(e^{-j(5\pi/2)}) e^{-j(5\pi/2)n}$$

Since $H(e^{j\omega})$ is defined for $0 \leq |\omega| \leq \pi$ we must evaluate the frequency at the baseband, i.e., $5\pi/2 \Rightarrow 5\pi/2 - 2\pi = \pi/2$. Therefore,

$$\begin{aligned} y[n] &= \frac{1}{2} e^{-j(\pi/4)} H(e^{j(5\pi/2)}) e^{j(5\pi/2)n} + \frac{1}{2} e^{j(\pi/4)} H(e^{-j(5\pi/2)}) e^{-j(5\pi/2)n} \\ &= \frac{1}{2} \left(e^{j[(5\pi/2)n - (\pi/2)]} + e^{-j[(5\pi/2)n - (\pi/2)]} \right) \\ &= \cos\left(\frac{5\pi}{2}n - \frac{\pi}{2}\right) \end{aligned}$$



- 4.35. The frequency response $H(e^{j\omega}) = H_c(j\Omega/T)$. Finding that

$$H_c(j\Omega) = \frac{1}{(j\Omega)^2 + 4(j\Omega) + 3}$$

$$\begin{aligned} H(e^{j\omega}) &= \frac{1}{(10j\omega)^2 + 4(10j\omega) + 3} \\ &= \frac{1}{-100\omega^2 + 3 + 40j\omega} \end{aligned}$$

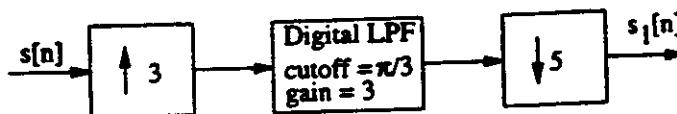
- 4.36. (a) Since $\Omega T = \omega$, $(2\pi \cdot 100)T = \frac{\pi}{2} \Rightarrow T = \frac{1}{400}$

- (b) The downsampler has $M = 2$. Since $x[n]$ is bandlimited to $\frac{\pi}{M}$, there will be no aliasing. The frequency axis simply expands by a factor of 2.

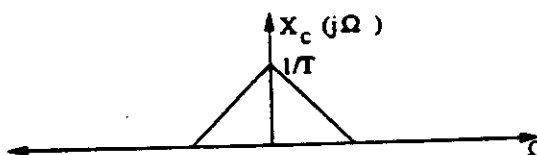
For $y_c(t) = x_c(t) \Leftrightarrow Y_c(j\Omega) = X_c(j\Omega)$.

Therefore $\Omega T' \Rightarrow 2\pi \cdot 100T' \Rightarrow T' = \frac{1}{200}$.

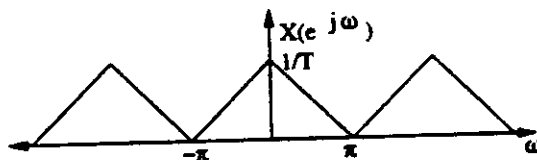
- 4.37. In both systems, the speech was filtered first so that the subsequent sampling results in no aliasing. Therefore, going $s[n]$ to $s_1[n]$ basically requires changing the sampling rate by a factor of $3\text{kHz}/5\text{kHz} = 3/5$. This is done with the following system:



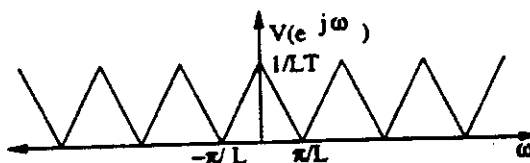
- 4.38. $X_c(j\Omega)$ is drawn below.



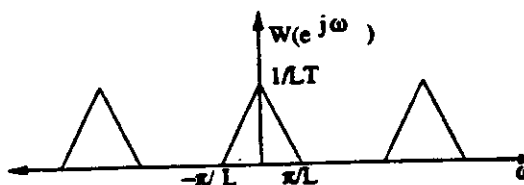
$x_c(t)$ is sampled at sampling period T , so there is no aliasing in $x[n]$.



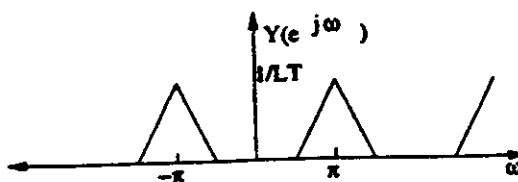
Inserting $L - 1$ zeros between samples compresses the frequency axis.



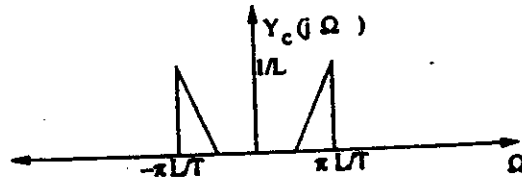
The filter $H(e^{j\omega})$ removes frequency components between π/L and π .



The multiplication by $(-1)^n$ shifts the center of the frequency band from 0 to π .



The D/C conversion maps the range $-\pi$ to π to the range $-\pi/T$ to π/T .



4.39. (a)

$$h[n] = 0, \quad |n| > (RL - 1)$$

Therefore, for causal system delay by $RL - 1$ samples.

(b) General interpolator condition:

$$\begin{aligned} h[0] &= 1 \\ h[kL] &= 0, \quad k = \pm 1, \pm 2, \dots \end{aligned}$$

(c)

$$y[n] = \sum_{k=-(RL-1)}^{(RL-1)} h[k]v[n-k] = h[0]v[n] + \sum_{k=1}^{RL-1} h[n](v[n-k] + v[n+k])$$

This requires only $RL-1$ multiplies, (assuming $h[0] = 1$.)

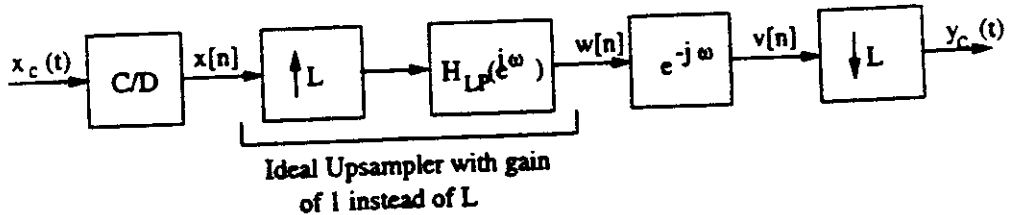
(d)

$$y[n] = \sum_{k=n-(RL-1)}^{n+(RL-1)} v[k]h[n-k]$$

If $n = mL$ (m an integer), then we don't have any multiplications since $h[0] = 1$ and the other non-zero samples of $v[k]$ hit at the zeros $h[n]$. Otherwise the impulse response spans $2RL - 1$ samples of $v[n]$, but only $2R$ of these are non-zero. Therefore, there are $2R$ multiplies.

4.40. Split $H(e^{j\omega})$ into a lowpass and a delay.

$$\begin{aligned} H(e^{j\omega}) &= H_{LP}(e^{j\omega})e^{-j\omega} \\ H_{LP}(e^{j\omega}) &= \begin{cases} 1, & |\omega| < \frac{\pi}{L} \\ 0, & \frac{\pi}{L} < |\omega| \leq \pi \end{cases} \end{aligned}$$



Then we analyze the system as follows:

$$x[n] = x_c(nT) \quad \text{no aliasing assumed}$$

$$w[n] = \frac{1}{L}x_c\left(n\frac{T}{L}\right) \quad \text{rate change}$$

$$v[n] = w[n-1] = \frac{1}{L}x_c\left(n\frac{T}{L} - \frac{T}{L}\right), \quad \text{delay at higher rate}$$

$$y[n] = v[nL] = \frac{1}{L}x_c\left(nT - \frac{T}{L}\right)$$

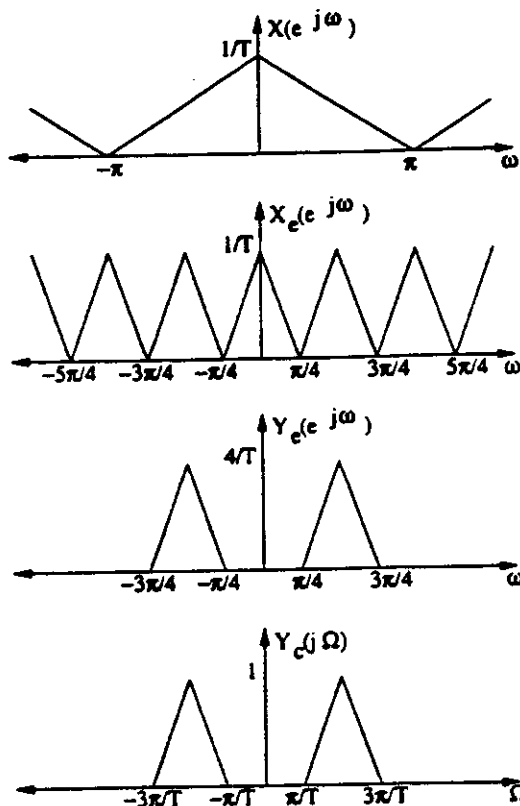
4.41. (a) See figures below.

(b) From part(a), we see that

$$Y_c(j\Omega) = X_c(j(\Omega - \frac{2\pi}{T})) + X_c(j(\Omega + \frac{2\pi}{T}))$$

Therefore,

$$y_c(t) = 2x_c(t) \cos(\frac{2\pi}{T}t)$$



4.42. (a) The Nyquist criterion states that $x_c(t)$ can be recovered as long as

$$\frac{2\pi}{T} \geq 2 \times 2\pi(250) \Rightarrow T \leq \frac{1}{500}.$$

In this case, $T = 1/500$, so the Nyquist criterion is satisfied, and $x_c(t)$ can be recovered.

(b) Yes. A delay in time does not change the bandwidth of the signal. Hence, $y_c(t)$ has the same bandwidth and same Nyquist sampling rate as $x_c(t)$.

(c) Consider first the following expressions for $X(e^{j\omega})$ and $Y(e^{j\omega})$:

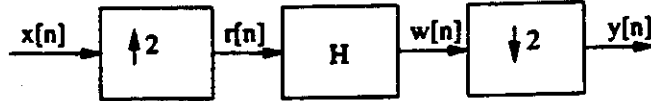
$$\begin{aligned} X(e^{j\omega}) &= \frac{1}{T} X_c(j\Omega) \big|_{\Omega=\frac{\omega}{T}} = \frac{1}{500} X_c(j500\omega) \\ Y(e^{j\omega}) &= \frac{1}{T} Y_c(j\Omega) \big|_{\Omega=\frac{\omega}{T}} = \frac{1}{T} e^{-j\Omega/1000} X_c(j\Omega) \big|_{\Omega=\frac{\omega}{T}} \\ &= \frac{1}{500} e^{-j\omega/2} X_c(j500\omega) \\ &= e^{-j\omega/2} X(e^{j\omega}) \end{aligned}$$

Hence, we let

$$H(e^{j\omega}) = \begin{cases} 2e^{-j\omega}, & |\omega| < \frac{\pi}{2} \\ 0, & \text{otherwise} \end{cases}$$

Then, in the following figure,

$$\begin{aligned} R(e^{j\omega}) &= X(e^{j2\omega}) \\ W(e^{j\omega}) &= \begin{cases} 2e^{-j\omega} X(e^{j2\omega}), & |\omega| < \frac{\pi}{2} \\ 0, & \text{otherwise} \end{cases} \\ Y(e^{j\omega}) &= e^{-j\omega/2} X(e^{j\omega}) \end{aligned}$$



(d) Yes, from our analysis above,

$$H_2(e^{j\omega}) = e^{-j\omega/2}$$

4.43. (a) Notice first that

$$X_c(j\Omega) = \begin{cases} F_c(j\Omega)|H_{aa}(j\Omega)|e^{-j\Omega^3}, & |\Omega| \leq 400\pi \\ E_c(j\Omega)|H_{aa}(j\Omega)|e^{-j\Omega^3}, & 400\pi \leq |\Omega| \leq 800\pi \\ 0, & \text{otherwise} \end{cases}$$

For the given $T = 1/800$, there is no aliasing from the C/D conversion. Hence, the equivalent CT transfer function $H_c(j\Omega)$ can be written as

$$H_c(j\Omega) = \begin{cases} H(e^{j\omega})|_{\omega=\Omega T}, & |\Omega| \leq \pi/T \\ 0, & \text{otherwise} \end{cases}$$

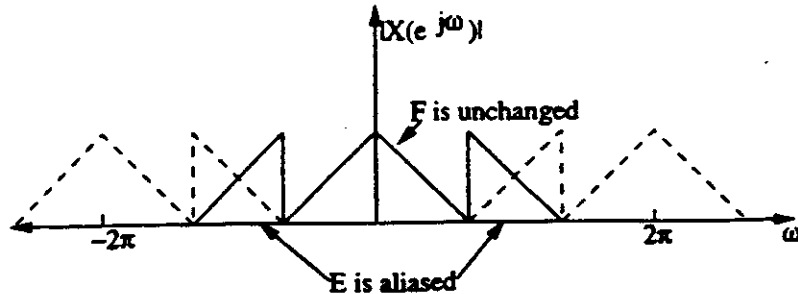
Furthermore, since $Y_c(j\Omega) = H_c(j\Omega)X_c(j\Omega)$, the desired transfer function is

$$H_c(j\Omega) = \begin{cases} e^{j\Omega^3}, & |\Omega| \leq 400\pi \\ 0, & \text{otherwise} \end{cases}$$

Combining the two previous equations, we find

$$H(e^{j\omega}) = \begin{cases} e^{j(800\omega)^3}, & |\omega| \leq \pi/2 \\ 0, & \pi/2 \leq |\omega| \leq \pi \end{cases}$$

(b) Some aliasing will occur if $2\pi/T < 1600\pi$. However, this is fine as long as the aliasing affects only $E_c(j\Omega)$ and not $F_c(j\Omega)$, as we show below:



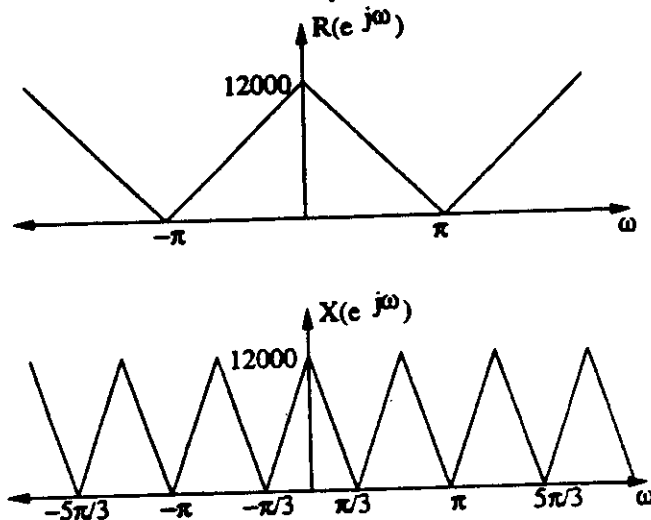
In order for the aliasing to not affect $F_c(j\Omega)$, we require

$$\frac{2\pi}{T} - 800\pi \geq 400\pi \Rightarrow \frac{2\pi}{T} \geq 1200\pi$$

The minimum $\frac{2\pi}{T}$ is 1200π . For this choice, we get

$$H(e^{j\omega}) = \begin{cases} e^{j(400\omega)^2}, & |\omega| \leq 2\pi/3 \\ 0, & 2\pi/3 \leq |\omega| \leq \pi \end{cases}$$

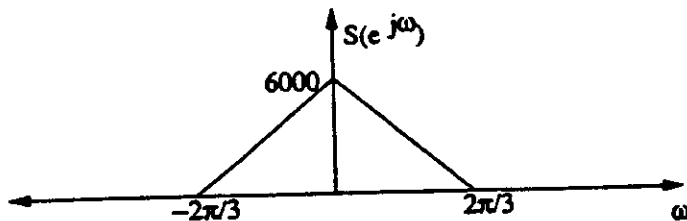
4.44. (a) See the following figure:



(b) For this to be true, $H(e^{j\omega})$ needs to filter out $X(e^{j\omega})$ for $\pi/3 \leq |\omega| \leq \pi$. Hence let $\omega_0 = \pi/3$. Furthermore, we want

$$\frac{\pi/2}{T_2} = 2\pi(1000) \Rightarrow T_2 = 1/6000$$

(c) Matching the following figure of $S(e^{j\omega})$ with the figure for $R_c(j\Omega)$, and remembering that $\Omega = \omega/T$, we get $T_3 = (2\pi/3)/(2000\pi) = 1/3000$.



4.45. Notice first that since $x_c(t)$ is time-limited,

$$A = \int_0^{10} x_c(t) dt = \int_{-\infty}^{\infty} x_c(t) dt = X_c(j\Omega)|_{\Omega=0}.$$

To estimate $X_c(j \cdot 0)$ by DT processing, we need to sample only fast enough so that $X_c(j \cdot 0)$ is not aliased. Hence, we pick

$$2\pi/T = 2\pi \times 10^4 \Rightarrow T = 10^{-4}.$$

The resulting spectrum satisfies

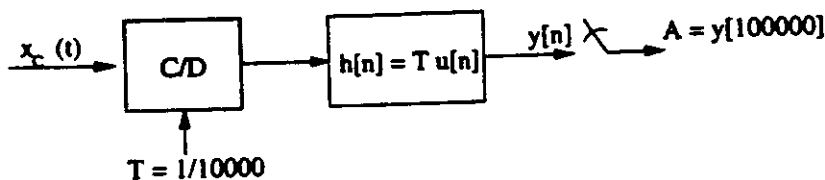
$$X(e^{j\cdot 0}) = \frac{1}{T} X_c(j \cdot 0)$$

Further,

$$X(e^{j\cdot 0}) = \sum_{n=-\infty}^{\infty} x[n].$$

Therefore, we pick $h[n] = Tu[n]$, which makes the system an accumulator. Our estimate \hat{A} is the output $y[n]$ at $n = 10/(10^{-4}) = 10^5$, when all of the non-zero samples of $x[n]$ have been added-up. This is an exact estimate given our assumption of both band- and time-limitedness. Since the assumption can never be exactly satisfied, however, this method only gives an approximate estimate for actual signals.

The overall system is as follows:



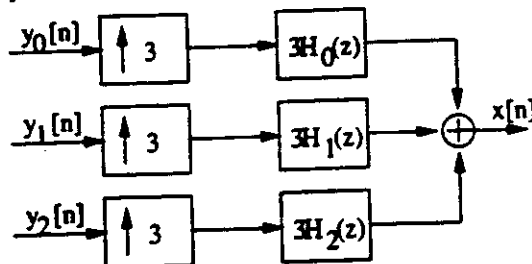
4.46. (a) Notice that

$$\begin{aligned} y_0[n] &= x[3n] \\ y_1[n] &= x[3n+1] \\ y_2[n] &= x[3n+2], \end{aligned}$$

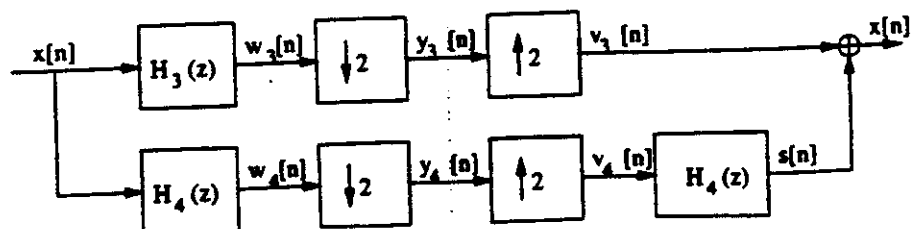
and therefore,

$$x[n] = \begin{cases} y_0[n/3], & n = 3k \\ y_1[(n-1)/3], & n = 3k+1 \\ y_2[(n-2)/3], & n = 3k+2 \end{cases}$$

(b) Yes. Since the bandwidth of the filters are $2\pi/3$, there is no aliasing introduced by downsampling. Hence to reconstruct $x[n]$, we need the system shown in the following figure:



(c) Yes, $x[n]$ can be reconstructed from $y_3[n]$ and $y_4[n]$ as demonstrated by the following figure:



In the following discussion, let $x_e[n]$ denote the even samples of $x[n]$, and $x_o[n]$ denote the odd samples of $x[n]$:

$$\begin{aligned} x_e[n] &= \begin{cases} x[n], & n \text{ even} \\ 0, & n \text{ odd} \end{cases} \\ x_o[n] &= \begin{cases} 0, & n \text{ even} \\ x[n], & n \text{ odd} \end{cases} \end{aligned}$$

In the figure, $y_3[n] = x[2n]$, and hence,

$$\begin{aligned} v_3[n] &= \begin{cases} x[n], & n \text{ even} \\ 0, & n \text{ odd} \end{cases} \\ &= x_e[n] \end{aligned}$$

Furthermore, it can be verified using the IDFT that the impulse response $h_4[n]$ corresponding to $H_4(e^{j\omega})$ is

$$h_4[n] = \begin{cases} -2/(j\pi n), & n \text{ odd} \\ 0, & \text{otherwise} \end{cases}$$

Notice in particular that every other sample of the impulse response $h_4[n]$ is zero. Also, from the form of $H_4(e^{j\omega})$, it is clear that $H_4(e^{j\omega})H_4(e^{j\omega}) = 1$ and hence $h_4[n] * h_4[n] = \delta[n]$.

Therefore,

$$\begin{aligned} v_4[n] &= \begin{cases} y_4[n/2], & n \text{ even} \\ 0, & n \text{ odd} \end{cases} \\ &= \begin{cases} w_4[n], & n \text{ even} \\ 0, & n \text{ odd} \end{cases} \\ &= \begin{cases} (x * h_4)[n], & n \text{ even} \\ 0, & n \text{ odd} \end{cases} \\ &= x_o[n] * h_4[n] \end{aligned}$$

where the last equality follows from the fact that $h_4[n]$ is non-zero only in the odd samples.

Now, $s[n] = v_4[n] * h_4[n] = x_o[n] * h_4[n] * h_4[n] = x_e[n]$, and since $x[n] = x_e[n] + x_o[n]$, $s[n] + v_3[n] = x[n]$.

4.47. Sampling random processes

$$\phi_{x,x_c}(\tau) = E(x_c(t)x_c^*(t+\tau)) \Leftrightarrow P_{x,x_c}(\Omega) = \int_{-\infty}^{\infty} \phi_{x,x_c}(\tau)e^{-j\Omega\tau} d\tau$$

(a)

$$\begin{aligned} \phi_{xx}[m] &= E(x[n]x^*[n+m]) = E(x_c(nT)x_c^*(nT+mT)) \\ &= \phi_{x,x_c}(mT), \quad \text{i.e., sampled autocovariance} \end{aligned}$$

(b) Since $\phi_{xx}[m]$ is a sampled $\phi_{x,x_c}(\tau)$

$$P_{xx}(\omega) = \frac{1}{T} \sum_{K=-\infty}^{\infty} P_{x,x_c} \left(\frac{\omega}{T} + \frac{2\pi k}{T} \right)$$

(c) If

$$P_{x,x_c} = 0, \text{ for } |\omega| \geq \pi$$

then

$$P_{xx}(\omega) = \frac{1}{T} P_{x,x_c} \left(\frac{\omega}{T} \right), \quad |\omega| \leq \pi$$

4.48. (a)

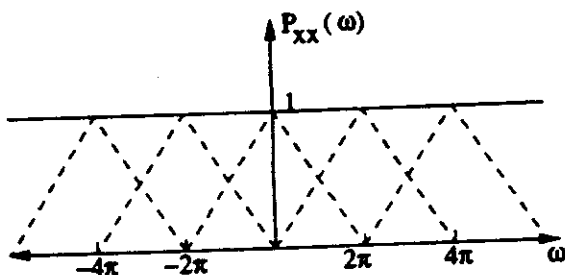
$$\begin{aligned}
 \phi_{x,x_c}(\tau) &= E(x_c(t)x_c(t+\tau)) \\
 \phi_{xx}[m] &= E(x[n]x[n+m]) = E(x_c(nT)x_c(nT+mT)) \\
 &= \phi_{x,x_c}(mT)
 \end{aligned}$$

(b)

$$P_{xx}(\omega) = \frac{1}{T} \sum_{r=-\infty}^{\infty} P_{x,x_c} \left(\frac{\omega}{T} + \frac{2\pi r}{T} \right)$$

Therefore, we require that $\frac{\pi}{T} \geq \Omega_0$.

- (c) For the spectrum of Fig P3.8-2 it is clear that if $T = \frac{2\pi}{\Omega_0}$ then the discrete-time power spectrum will be white, as shown in the figure above.

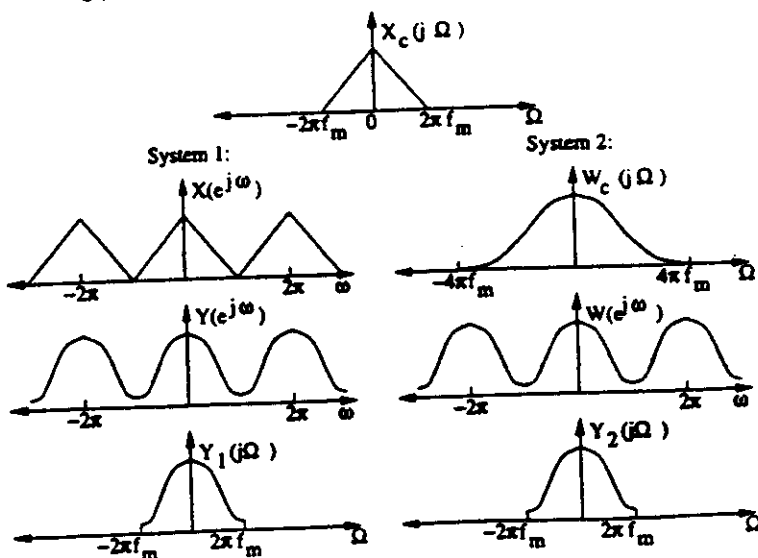


- (d) For white discrete-time signal $\Rightarrow \phi_{xx}[m] = 0$, $m \neq 0$ but $\phi_{xx}[m] = \phi_{x,x_c}(mT)$. Therefore, any analog signal whose autocorrelation function has zeros equally spaced at intervals of T will yield a white discrete-time sequence is sampled with sampling period T . For example, for Fig P3.8-1:

$$\phi_{x,x_c}(\tau) = \frac{\sin \Omega_0 T}{T\pi} \Rightarrow \phi_{xx}[m] = \frac{\sin \Omega_0 mT}{\pi mT}$$

$$\text{if } T = \frac{\pi}{\Omega_0} \quad \phi_{xx}[m] = \frac{\sin \pi m}{\pi^2 m / \Omega_0} = 0, \quad m \neq 0$$

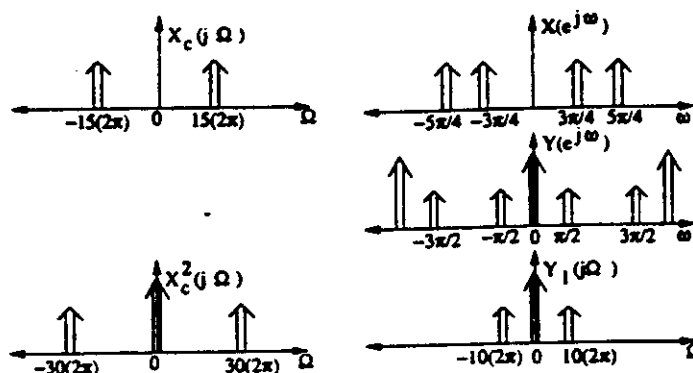
4.49. (a) Consider the following plots.



$y_1(t) = y_2(t)$: Convolution is a linear process. Aliasing is a linear process. Periodic convolution is equivalent to convolution followed by aliasing.

$y_1(t) \neq x^2(t)$: System 2 at Step 1 shows $X_c^2(j\Omega)$. This is clearly not $Y_1(j\Omega)$. $Y_1(j\Omega)$ is an aliased version of $X_c(j\Omega)$

(b) Now,



(c)

$$\begin{aligned} x(t) &= A \cos(30\pi t) \\ x^3(t) &= \frac{3}{4}A \cos(30\pi t) + \frac{1}{4}A \cos(3 \cdot 30\pi t), \\ v[n] &= \frac{3}{4}A \cos\left(\frac{3}{4}\pi n\right) + \frac{1}{4}A \cos\left(\frac{1}{4}\pi n\right) \\ v[n] &= x^3[n] \\ y[n] &= x[n] \end{aligned}$$

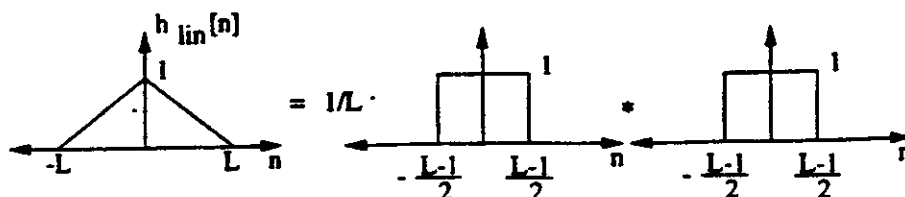
We can see here that sometimes aliasing won't be destructive. When aliased sections do not overlap they can be reconstructed.

- (d) This is the inverse to part (c). Since multiplication in time corresponds to convolution in frequency, a signal $x^2(t)$ has at most two times the bandwidth of $x(t)$. Therefore, $x^{1/2}$ will have at least $\frac{1}{2}$ the bandwidth of $x(t)$. If we run our signal through a box that will raise it to the $1/M$ power, then the sampling rate can be decreased by a factor of M .

4.50. (a)

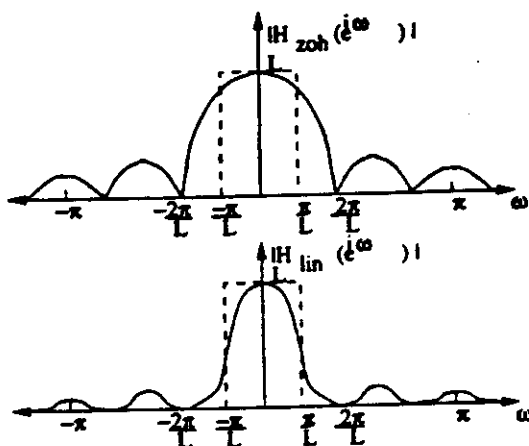
$$\begin{aligned} x_i[n] &= x_u[n] * h_{zoh}[n] \\ h_{zoh}[n] &= \begin{cases} 1, & 0 \leq n \leq L-1 \\ 0, & \text{else} \end{cases} \\ H_{zoh}(e^{j\omega}) &= \frac{\sin(\omega L/2)}{\sin(\omega/2)} e^{-j(L-1)\omega/2} \end{aligned}$$

- (b) The impulse response $h_{lin}[n]$ corresponds to the convolution of two rectangular sequences, as shown below.



$$H_{lin}(e^{j\omega}) = \frac{1}{L} \left(\frac{\sin(\omega L/2)}{\sin(\omega/2)} \right)^2$$

- (c) The frequency response of zero-order-hold is flatter in the region $[-\pi/L, \pi/L]$, but achieves less out-of-band attenuation.



4.51.

$$\begin{aligned}\phi_{xx}[n] &= x[n] * x[-n] \\ \Phi_{xx}(e^{j\omega}) &= X(e^{j\omega}) * X^*(e^{j\omega})\end{aligned}$$

The bandwidth of $\Phi_{xx}(e^{j\omega})$ is no larger than the bandwidth of $X(e^{j\omega})$. Therefore, the outputs of the systems will be the same if $H_2(e^{j\omega})$ is an ideal lowpass filter with a cutoff of π/L .

- 4.52. The idea here is to exploit the fact that every other sample supplied to $h[n]$ in Fig 3.27-1 is zero. That is,

$$\begin{aligned}y_1[n] &= h[n] * w[n] = \sum_{k=-\infty}^{\infty} w[n-k]h[k] \\ &= aw[n] + bw[n-1] + cw[n-2] + dw[n-3] + ew[n-4] \\ &= \begin{cases} ax[n/2] + cx[(n/2)-1] + e[(n/2)-2], & n \text{ even} \\ bx[(n/2)-(1/2)] + dx[(n/2)-(3/2)], & n \text{ odd} \end{cases}\end{aligned}$$

$$\begin{aligned}w_1[n] &= \begin{cases} h_1[n/2] * x[n/2], & n \text{ even} \\ 0, & n \text{ odd} \end{cases} \\ &= \begin{cases} h_1[0]x[n/2] + h_1[1]x[(n/2)-1] + h_1[2]x[(n/2)-2], & n \text{ even} \\ 0, & n \text{ odd} \end{cases}\end{aligned}$$

$$\begin{aligned}w_2[n] &= \begin{cases} h_2[n/2] * x[n/2], & n \text{ even} \\ 0, & n \text{ odd} \end{cases} \\ &= \begin{cases} h_2[0]x[n/2] + h_2[1]x[(n/2)-1] + h_2[2]x[(n/2)-2], & n \text{ even} \\ 0, & n \text{ odd} \end{cases}\end{aligned}$$

Comparing $w_1[n]$, $w_2[n]$ with $y_1[n]$ above:

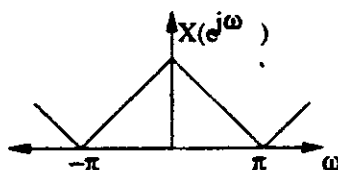
$w[n]$ can give even samples if $h_1[0] = a$, $h_1[1] = c$, $h_2[2] = e$. Similarly, $w_2[n]$ can give the odd samples if $h_3[n]$ delays $w_2[n]$ by one sample, i.e., $h_3[0] = 0$, $h_3[1] = 0$, $h_3[2] = 0$. Thus

$$w_3[n] = \begin{cases} h_2[0]x[(n-1)/2] + h_2[1]x[(n-1)/2-1] + h_2[2]x[(n-1)/2-2], & n \text{ even} \\ 0, & n \text{ odd} \end{cases}$$

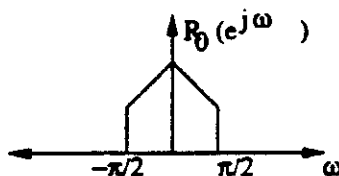
$$h_2[0] = b, \quad h_2[1] = d, \quad h_2[2] = 0$$

4.53. Sketches appear below.

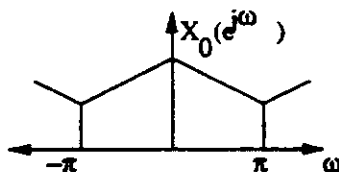
(a) First, $X(e^{j\omega})$ is plotted.



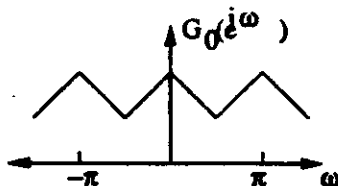
The lowpass filter cuts off at $\frac{\pi}{2}$.



The downsampler expands the frequency axis. Since $R_0(e^{j\omega})$ is bandlimited to $\frac{\pi}{M}$, no aliasing occurs.



The upsampler compresses the frequency axis by a factor of 2.



The lowpass filter cuts off at $\frac{\pi}{2} \Rightarrow Y_0(e^{j\omega}) = R_0(e^{j\omega})$ as sketched above.

(b) $G_0(e^{j\omega}) = \frac{1}{2} (X(e^{j\omega})H_0(e^{j\omega}) + X(e^{j(\omega+\pi)})H_0(e^{j(\omega+\pi)}))$

$$\begin{aligned}
 (c) \quad Y_0(e^{j\omega}) &= \frac{1}{2} H_0(e^{j\omega}) \left(X(e^{j\omega}) H_0(e^{j\omega}) + X(e^{j(\omega+\pi)}) H_0(e^{j(\omega+\pi)}) \right) \\
 Y_1(e^{j\omega}) &= \frac{1}{2} H_1(e^{j\omega}) \left(X(e^{j\omega}) H_1(e^{j\omega}) + X(e^{j(\omega+\pi)}) H_1(e^{j(\omega+\pi)}) \right) \\
 Y(e^{j\omega}) &= Y_0(e^{j\omega}) - Y_1(e^{j\omega}) \\
 &= \frac{1}{2} X(e^{j\omega}) [H_0^2(e^{j\omega}) - H_1^2(e^{j\omega})] \\
 &\quad + \underbrace{\frac{1}{2} X(e^{j(\omega+\pi)}) [H_0(e^{j\omega}) H_0(e^{j(\omega+\pi)}) - H_1(e^{j\omega}) H_1(e^{j(\omega+\pi)})]}_{=0}
 \end{aligned}$$

The aliasing terms always cancel. $Y(e^{j\omega})$ is proportional to $X(e^{j\omega})$ if $[H_0^2(e^{j\omega}) - H_1^2(e^{j\omega})]$ is a constant.

$X(e^{j\omega}) = 0, \pi/3 \leq |\omega| \leq \pi$. $x[n]$ can be thought of as an oversampled signal. The approach is to determine whether n_0 is odd or even, then sample so that n_0 is avoided, upsampled and lowpass filter. This recovers $x[n_0]$.

4.54. (a) In the case where n_0 is not known, we determine whether it is even or odd as follows:

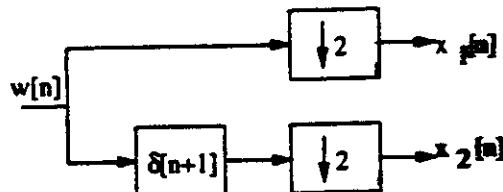
$$\begin{aligned}
 \hat{x}[n] &= x[n] - A\delta[n - n_0] \\
 \hat{X}(e^{j\omega}) &= X(e^{j\omega}) - Ae^{-j\omega n_0} \\
 \hat{X}(e^{j\omega})|_{\omega=\pi/2} &= \sum_n x[n](-j)^n \\
 \hat{X}(e^{j(\pi/2)}) &= -A(-j)^{n_0}
 \end{aligned}$$

If the result is real, n_0 is even. If the result is imaginary, n_0 is odd.

- (b) If n_0 is even, sample $\hat{x}[n]$ so that the even-numbered sequence values are set to zero. If n_0 is odd, sample so the odd-numbered samples are set to zero.
- (c) Filter the sampled sequence with a lowpass filter with cutoff frequency $\pi/3$, and gain 2. This is an exact procedure if ideal filters are used.

4.55. (a)

$$\begin{aligned}
 w[n] &= \begin{cases} x_1[n/2], & n \text{ even} \\ x_2[(n-1)/2], & n \text{ odd} \end{cases} \\
 x_1[n] &= w[2n] \\
 x_2[n] &= w[2n+1]
 \end{aligned}$$



The system is linear, time-varying (due to downsampling), non-causal (due to $\delta[n+1]$), and stable.

(b)

$$T = \frac{\pi}{\Omega_N} = \frac{\pi}{2\pi \times 5000} = 10^{-4} \text{ sec}, \quad \frac{L\omega_1}{T} = 2\pi \times 10^5$$

To avoid aliasing in $y_c(t)$:

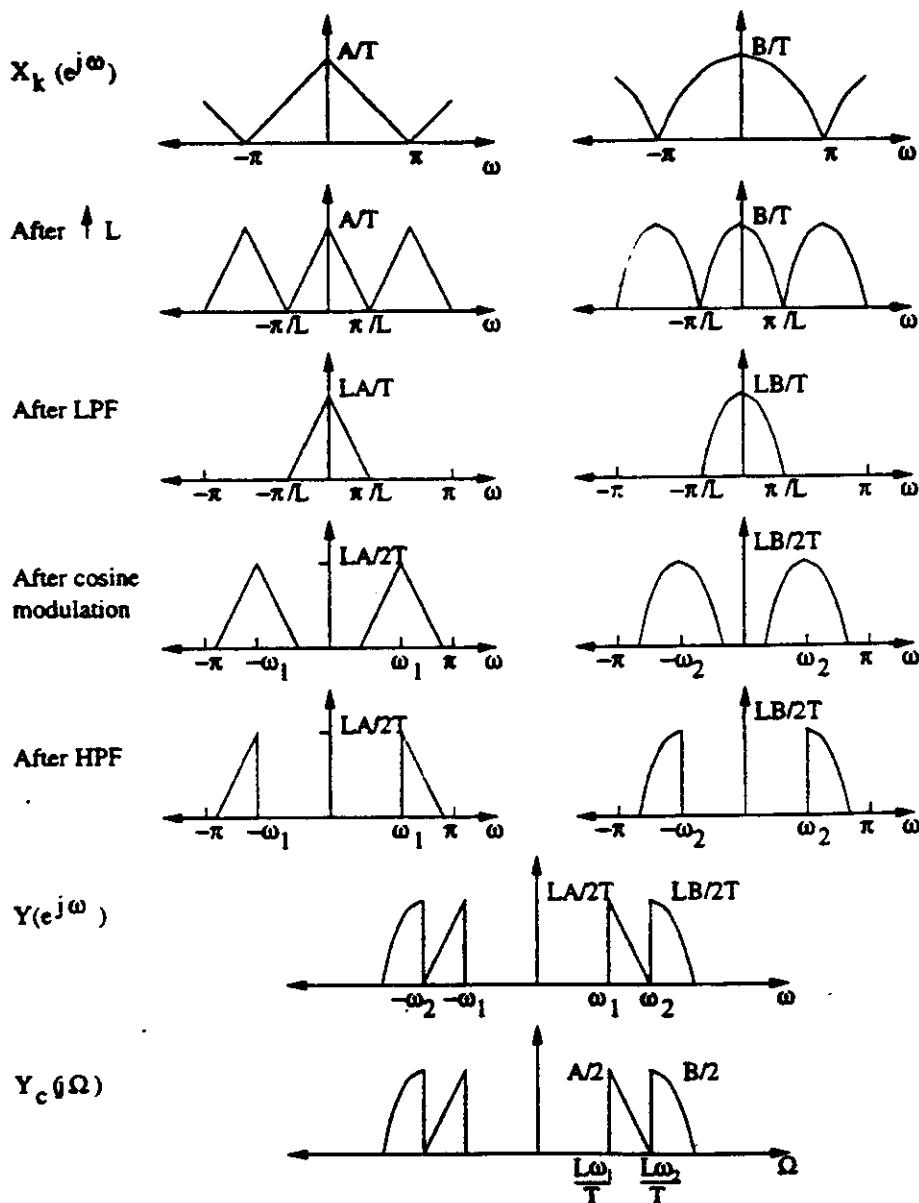
$$\frac{L\omega_1}{T} + \frac{2\pi}{T} \leq \frac{L\pi}{T}$$

$$\omega_1 = \frac{20\pi}{L}$$

$$20\pi + 2\pi = L\pi$$

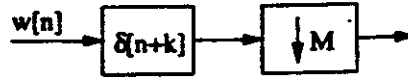
$$L = 22, \quad \omega_1 = 2\pi\left(\frac{10}{22}\right)$$

(c) The Fourier transforms are sketched below.

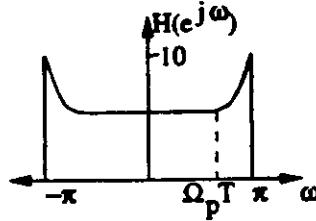


(d) To generalize for M channels, we would use the same modulators, but we would choose a larger value of L to make room for additional spectra above the lower frequency bound. If the lower

bound remained $2\pi \cdot 10^5$, L would become $L = 20 + M$ for M channels.
A branch of the TDM demultiplexing system would be:



4.56. Since we want $W(e^{j\omega})$ to equal $X(e^{j\omega})$, then $H(e^{j\omega})$ must compensate for the drop offs in $H_{aa}(j\Omega)$.



4.57. (a)

$$E(e) = \int e p(e) de = \frac{1}{\Delta} \int_{-\Delta/2}^{\Delta/2} e de = \frac{e^2}{2\Delta} \Big|_{-\Delta/2}^{\Delta/2} = 0$$

$$\sigma_e^2 = E(e^2 - 0) = \frac{1}{\Delta} \int_{-\Delta/2}^{\Delta/2} e^2 de = \frac{e^3}{3\Delta} \Big|_{-\Delta/2}^{\Delta/2} = \frac{\Delta^2}{12}$$

$$r[m, n] = E(e[m]e[n]) = \begin{cases} E(e[m])E(e[n]), & m \neq n \\ E(e^2[n]), & m = n \end{cases}$$

$$r[n, m] = r[n - m] = \frac{\Delta^2}{12} \delta[n - m]$$

(b)

$$\text{SNR} = \frac{\sigma_z^2}{\sigma_e^2} = \frac{12\sigma_z^2}{\Delta^2}$$

(c) Let $e_y[n]$ be the output noise.

$$e_y[n] = \sum_k h[k]e[n - k]$$

$$E(e_y^2[n]) = E\left(\sum_k h[k]e[n - k] \sum_l h[l]e[n - l]\right) = \sum_k \sum_l h[k]h[l] \underbrace{E(e[n - k]e[n - l])}_{\sigma_e^2 \delta[k - l]}$$

$$\begin{aligned} \sigma_{e_y}^2 &= \sigma_e^2 \sum_k h^2[k] \\ &= \sigma_e^2 \sum_{k=0}^{\infty} \frac{1}{4} (a^k + (-a)^k)^2 = \frac{\sigma_e^2}{4} \sum_{k=0}^{\infty} (a^{2k} + 2a^k(-a)^k + (-a)^{2k}) \\ &= \frac{\sigma_e^2}{2} \left(\sum_{k=0}^{\infty} a^{2k} + \sum_{k=0}^{\infty} (-a^2)^k \right) = \frac{\sigma_e^2}{2} \left(\frac{1}{1 - a^2} + \frac{1}{1 + a^2} \right) \\ &= \sigma_e^2 \left(\frac{1}{1 - a^4} \right) = \frac{\Delta^2}{12(1 - a^4)} \end{aligned}$$

The variance of $x[n]$ is weighted similarly so the SNR does not change. $\text{SNR}_{\text{out}} = 12 \frac{\sigma_z^2}{\Delta^2}$.

(d) $f[n] = x[n]e[n]$

$$E(f[n]) = E(x[n]e[n]) = E(x[n])E(e[n]) = 0$$

$$\sigma_f^2 = E(f^2[n]) = E(x^2[n]e^2[n]) = E(x^2[n])E(e^2[n]) = \sigma_x^2\sigma_e^2$$

$$r_f[n, m] = E(x[n]x[m]e[n]e[m]) = \underbrace{E(x[n]x[m])}_{\sigma_x^2\delta[n-m]} \cdot \underbrace{E(e[n]e[m])}_{\sigma_e^2\delta[n-m]}$$

(e)

$$\text{SNR} = \frac{\sigma_x^2}{\sigma_f^2} = \frac{\sigma_x^2}{\sigma_x^2\sigma_e^2} = \frac{1}{\sigma_e^2} = \frac{12}{\Delta^2}$$

(f) Using the results of part (c).

$$\sigma_{e_r}^2 = \sigma_f^2 \left(\frac{1}{1 - \alpha^4} \right) = \frac{\sigma_x^2\sigma_e^2}{1 - \alpha^4}$$

Again, the variance of $x[n]$ is weighted by the same factor, so the SNR does not change.

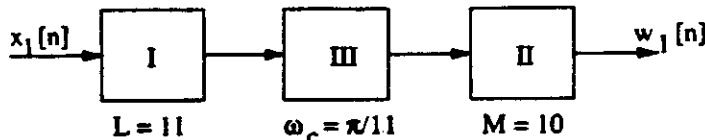
$$\text{SNR}_{\text{out}} = \frac{12}{\Delta^2}.$$

4.58. First, notice that since $y_c(t) = x_1(t)x_2(t)$, $Y_c(j\Omega) = \frac{1}{2\pi}(X_1(j\Omega) * X_2(j\Omega))$, and so $Y_c(j\Omega) = 0$ for $|\Omega| \geq 11\pi/2 \times 10^4$. Hence the Nyquist rate $T = 1/55000$ s.

Choose System A and B such that $w_1[n] = \alpha x_1(nT)$ and $w_2[n] = \beta x_2(nT)$.

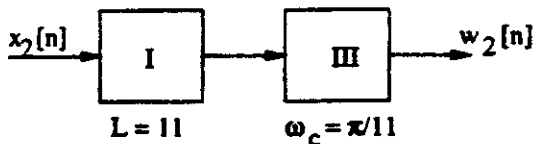
For System A, we need to resample such that

$$\frac{M}{L} = \frac{T}{T_1} = \frac{2 \times 10^{-5}}{1/55000} = \frac{10}{11}$$



For System B, we need to resample such that

$$\frac{M}{L} = \frac{T}{T_1} = \frac{2 \times 10^{-4}}{1/55000} = \frac{1}{11}$$



System C is simply the identity system.

4.59. The speech is first sampled at 44.1 kHz, and we wish to resample it so that the sampling rate is at 8 kHz. There are no aliasing effects anywhere in the system. Hence

$$\frac{L}{M} = \frac{44.1}{8} = \frac{441}{80}$$

We simply make $L = 441$, $M = 80$, and $\omega_c = \pi/441$.

4.60. Ω_p , and Ω_s has to be chosen such that

(a) The region $|\Omega| \leq \Omega_p$ maps to $|\omega| \leq \pi/4$:

$$\Omega_p T = \frac{\pi}{4} \Rightarrow \Omega_p = 44\pi$$

(b) No aliasing occurs in the region $|\Omega| \leq \Omega_p$ during sampling:

$$\frac{2\pi}{T} - \Omega_s = \Omega_p \Rightarrow \Omega_s = 2\pi(4 \cdot 44) - 44\pi = 308\pi$$

4.61. (a)

$$\begin{aligned} V(z) &= H_1(z)(X(z) - Y(z)) \\ U(z) &= H_2(z)(V(z) - Y(z)) \\ Y(z) &= U(z) + E(z) \\ &= \frac{H_1(z)H_2(z)}{1 + H_2(z)(1 + H_1(z))} X(z) + \frac{1}{1 + H_2(z)(1 + H_1(z))} E(z) \end{aligned}$$

Substituting $H_1(z) = 1/(1 - z^{-1})$ and $H_2(z) = z^{-1}/(1 - z^{-1})$, we find

$$\begin{aligned} H_{xy}(z) &= z^{-1} \\ H_{ey}(z) &= (1 - z^{-1})^2 \end{aligned}$$

Hence the difference equation is $y[n] = x[n - 1] + f[n]$, where

$$f[n] = e[n] - 2e[n - 1] + e[n - 2].$$

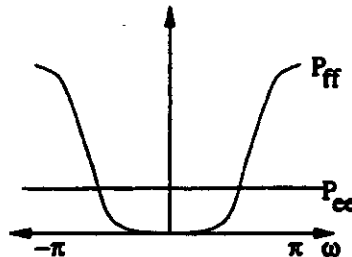
(b)

$$\begin{aligned} P_{ff}(e^{j\omega}) &= \sigma_e^2 |H_{ey}(e^{j\omega})|^2 \\ &= \sigma_e^2 |1 - e^{-j\omega}|^2 \\ &= \sigma_e^2 (1 - e^{-j\omega})^2 (1 - e^{j\omega})^2 \\ &= \sigma_e^2 (2 - 2\cos(\omega))^2 \\ &= \sigma_e^2 (4\sin^2(\omega/2))^2 \\ &= 16\sigma_e^2 \sin^4(\omega/2) \end{aligned}$$

The total noise power σ_f^2 is the autocorrelation of $f[n]$ evaluated at 0:

$$\begin{aligned} \sigma_f^2 &= E[(e[n] - 2e[n - 1] + e[n - 2])^2] \\ &= E[e^2[n]] + E[-2e^2[n - 1]] + E[e^2[n - 2]] \\ &= 6\sigma_e^2, \end{aligned}$$

where we have used linearity of expectations, and the fact that since $e[n]$ is white, $E[e[n]e[n - k]] = 0$ for $k \neq 0$.



(c) Since $X(e^{j\omega})$ is bandlimited, $x[n] * h_3[n] = x[n]$. Hence,

$$w[n] = y[n] * h_3[n] = (x[n-1] + f[n]) * h_3[n] = x[n-1] + g[n],$$

where $g[n]$ is the quantization noise in the region $|\omega| < \pi/M$.

(d) For a small angle x , $\sin x \approx x$. Therefore,

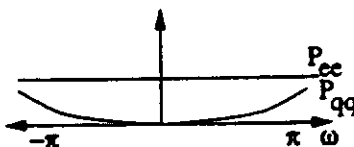
$$\begin{aligned}\sigma_g^2 &= \frac{1}{2\pi} \int_{-\pi/M}^{\pi/M} \sigma_e^2 (2 \sin \omega/2)^4 d\omega \\ &\approx \frac{1}{2\pi} \int_{-\pi/M}^{\pi/M} \sigma_e^2 (2\omega/2)^4 d\omega \\ &= \frac{\sigma_e^2}{2\pi} \frac{\omega^5}{5} \Big|_{-\pi/M}^{\pi/M} \\ &= \frac{\sigma_e^2 \pi^4}{5M^5}\end{aligned}$$

(e) $X_c(j\Omega)$ must be sufficiently bandlimited that $X(e^{j\omega}) = X_c(j\Omega T)$ is zero for $|\omega| > \pi/M$. Hence $X_c(j\Omega) = 0$ for $|\Omega| > \pi/MT$.

Assuming that is satisfied, $v_x[n] = x[Mn-1] = x_c(MTn-T)$.

Downsampling does not change the variance of the noise, and hence $\sigma_q^2 = \sigma_g^2$.

$$\begin{aligned}P_{qq}(e^{j\omega}) &= P_{gg}(e^{j\omega/M}) \\ &= 16\sigma_e^2 \sin^4(\omega/2M)\end{aligned}$$



4.62. (a) (i) The transfer function from $x[n]$ to $y_x[n]$ is

$$H_{xy}(z) = \frac{\frac{z^{-1}}{1-z^{-1}}}{1 + \frac{z^{-1}}{1-z^{-1}}} = z^{-1}$$

Hence $y_x[n] = x[n-1]$.

(ii) The transfer function from $e[n]$ to $y_e[n]$ is

$$H_{ey}(z) = \frac{1}{1 + \frac{z^{-1}}{1-z^{-1}}} = 1 - z^{-1}$$

So

$$\begin{aligned}P_{y_x}(\omega) &= P_e(\omega) H_{ey}(e^{j\omega}) H_{ey}^* e^{-j\omega} \\ &= \sigma_e^2 (1 - e^{-j\omega})(1 - e^{j\omega}) \\ &= \sigma_e^2 (2 - 2 \cos(\omega))\end{aligned}$$

(b) (i) $x[n]$ contributes only to $y_1[n]$, but not $y_2[n]$. Therefore

$$\begin{aligned}y_{1x}[n] &= x[n-1] \\ r_x[n] &= x[n-2]\end{aligned}$$

(ii) In part(a), the difference equation describing the sigma-delta noise-shaper is

$$y[n] = x[n - 1] + e[n] - e[n - 1].$$

So here we apply the difference equation to both sigma-delta modulators:

$$\begin{aligned} y_{1e}[n] &= e_1[n] - e_1[n - 1] \\ y_{2e}[n] &= e_1[n - 1] + e_2[n] - e_2[n - 1] \\ r_e[n] &= y_{1e}[n - 1] - (y_{2e}[n] - y_{2e}[n - 1]) \\ &= -e_2[n] + 2e_2[n - 1] - e_2[n - 2] \\ H_{e2r}(z) &= -(1 - z^{-1})^2 \\ P_{r_e}(\omega) &= \sigma_e^2(2 - 2\cos\omega)^2 \end{aligned}$$