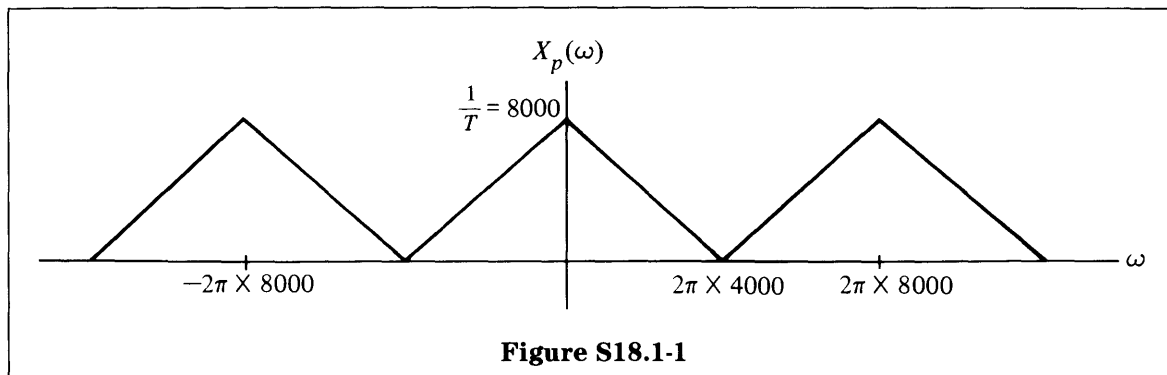


# 18 Discrete-Time Processing of Continuous-Time Signals

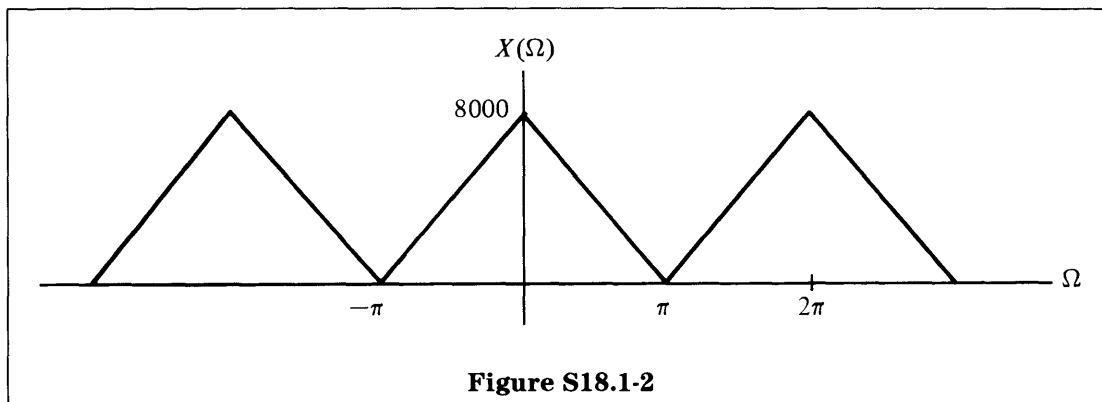
## Solutions to Recommended Problems

S18.1

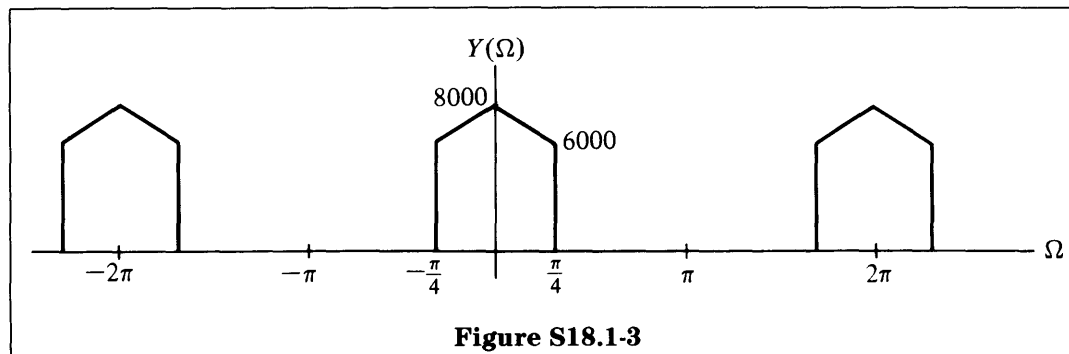
- (a) Since  $x_p(t) = x_c(t)p(t)$ , then  $X_p(\omega)$  is just a replication of  $X_c(\omega)$  centered at multiples of the sampling frequency, namely 8 kHz or  $2\pi \times 8 \times 10^3$  rad/s. The sampling period is  $T = 1/8000$ .



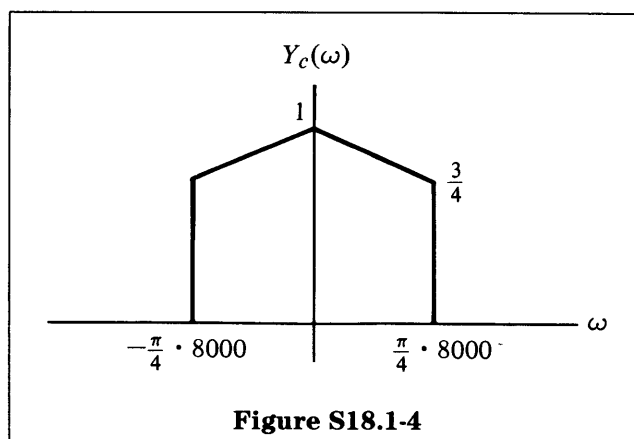
- (b)  $X(\Omega)$  is just a rescaling of the frequency axis, where  $2\pi \times 10^3$  becomes  $2\pi$ .  $X(\Omega)$  is shown in Figure S18.1-2.



- (c)  $Y(\Omega)$  is the product  $G(\Omega)X(\Omega)$ . Therefore,  $Y(\Omega)$  appears as in Figure S18.1-3.



- (d)  $Y_c(\omega)$  is a frequency-scaled version of  $Y(\omega)$  but only in the range  $\Omega = -\pi$  to  $\pi$ , as shown in Figure S18.1-4. Also note the gain of  $T$ .

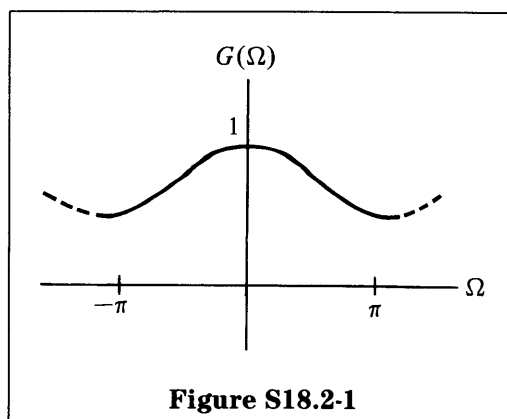


## S18.2

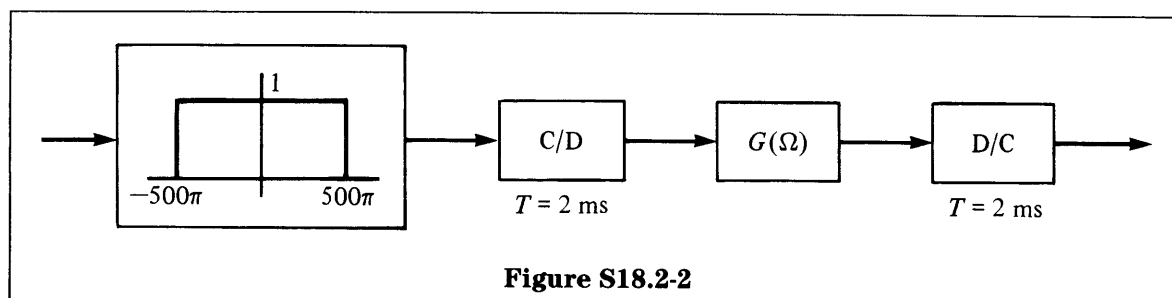
- (a) The maximum nonzero frequency component of  $H(\omega)$  is  $500\pi$ . Therefore, this frequency can correspond to, at most, the maximum digital frequency before folding, i.e.,  $\Omega = \pi$ . From the relation  $\omega T = \Omega$ , we get

$$T_{\max} = \frac{\pi}{500\pi} = 2 \text{ ms}$$

- (b) Since  $\omega = 500\pi$  maps to  $\Omega = \pi$ , the discrete-time filter  $G(\Omega)$  is as shown in Figure S18.2-1.

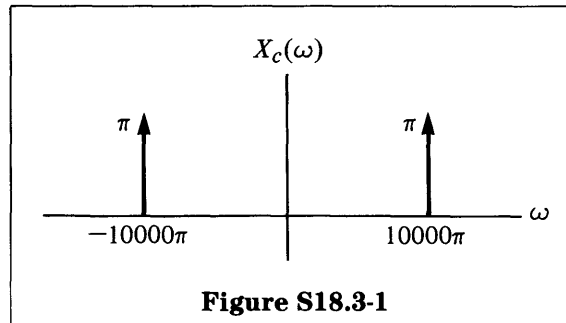


- (c) The complete system is given by Figure S18.2-2. Note the need for an anti-aliasing filter.



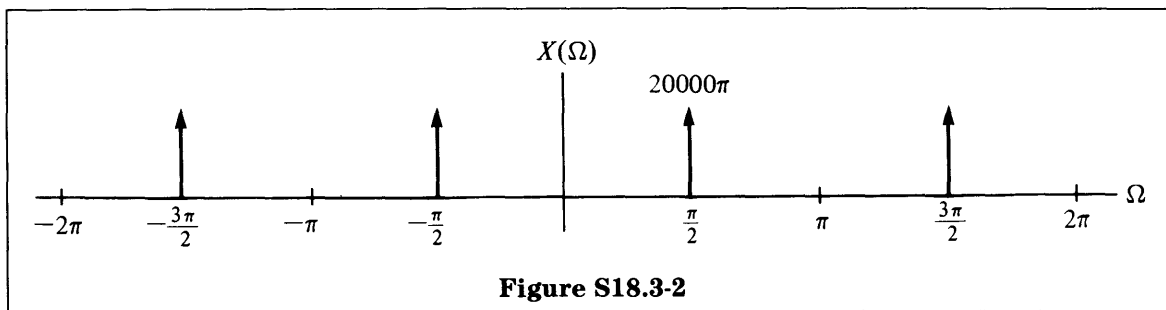
**S18.3**

(a) Recall that  $X_c(\omega)$  is as given by Figure S18.3-1.


**Figure S18.3-1**

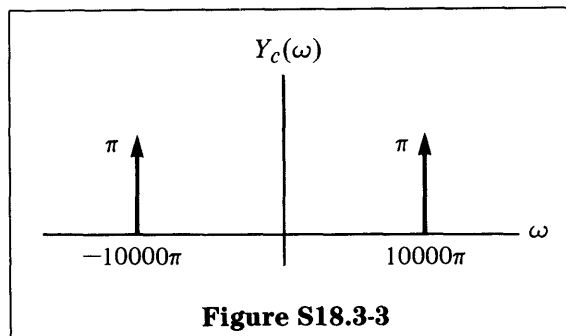
$X(\Omega)$  is given by eq. (S18.3-1) and Figure S18.3-2.

$$X(\Omega) = \frac{1}{T} \sum_{n=-\infty}^{\infty} X_c\left(\frac{\Omega}{T} - \frac{2\pi n}{T}\right) = 20000 \sum_{n=-\infty}^{\infty} X_c[20000(\Omega - 2\pi n)] \quad (\text{S18.3-1})$$


**Figure S18.3-2**

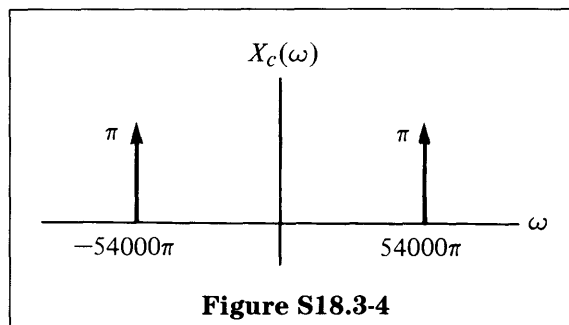
$Y_c(\omega)$  is given by eq. (S18.3-2) and Figure S18.3-3.

$$Y_c(\omega) = \begin{cases} TX(\omega T), & |\omega| < \frac{\pi}{T} \\ 0, & \text{elsewhere} \end{cases} \quad (\text{S18.3-2})$$

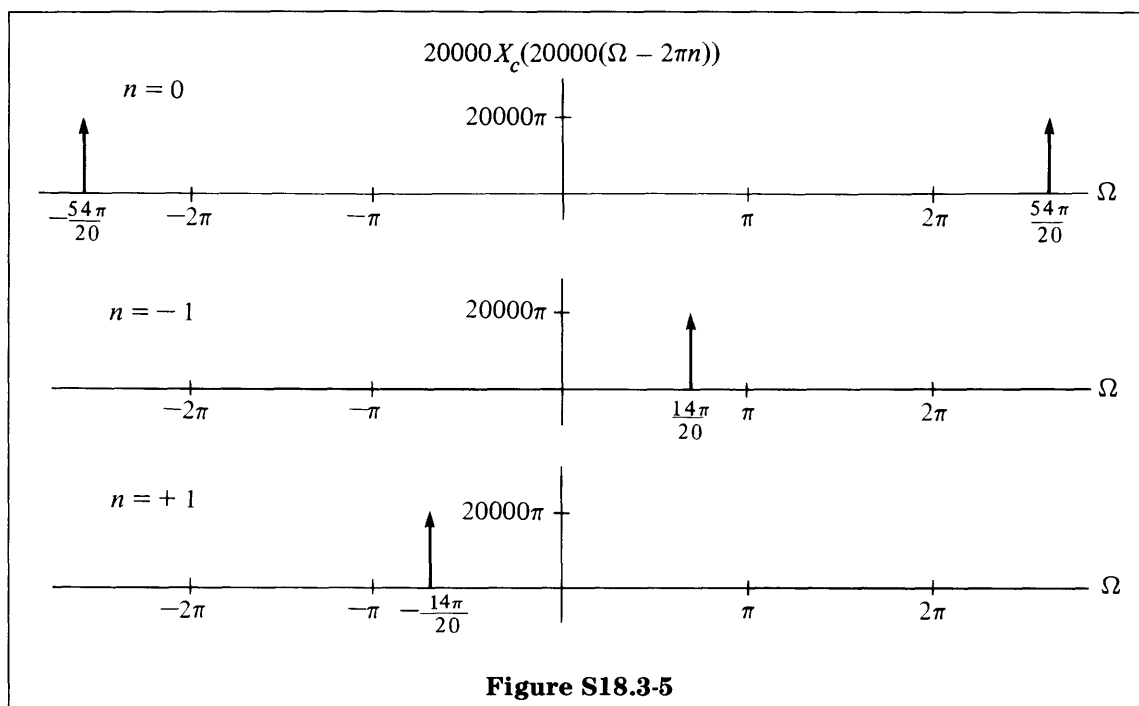

**Figure S18.3-3**

Thus  $x(t) = y(t)$  in this case.

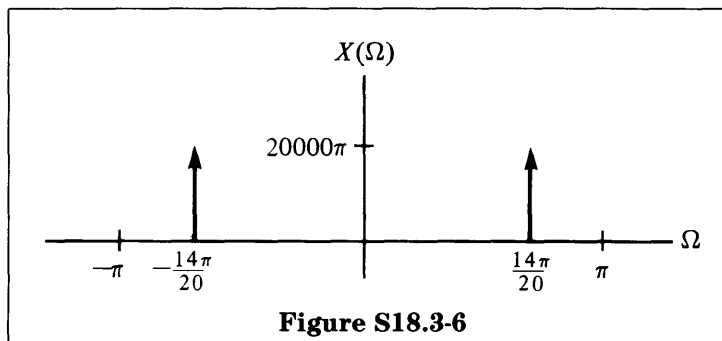
(b)  $X_c(\omega)$  is as given in Figure S18.3-4.



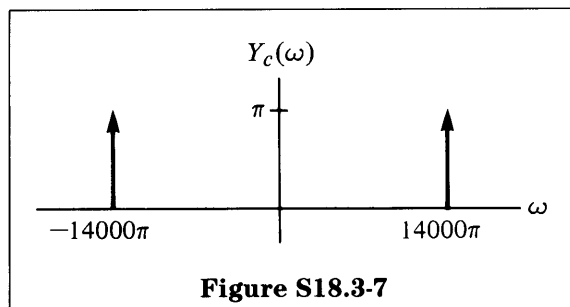
We now use eq. (S18.3-1), shown in Figure S18.3-5.



Thus, in the range  $\pm\pi$ ,  $X(\Omega) = 20000 \sum_{n=-\infty}^{\infty} X_c[20000(\Omega - 2\pi n)]$  is given as in Figure S18.3-6.

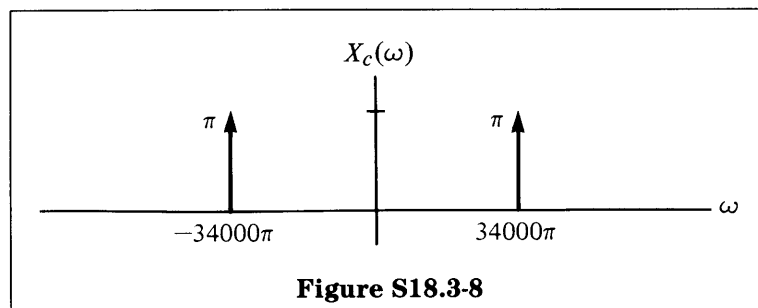


Using eq. (S18.3-2), we find  $Y_c(\omega)$  as in Figure S18.3-7.

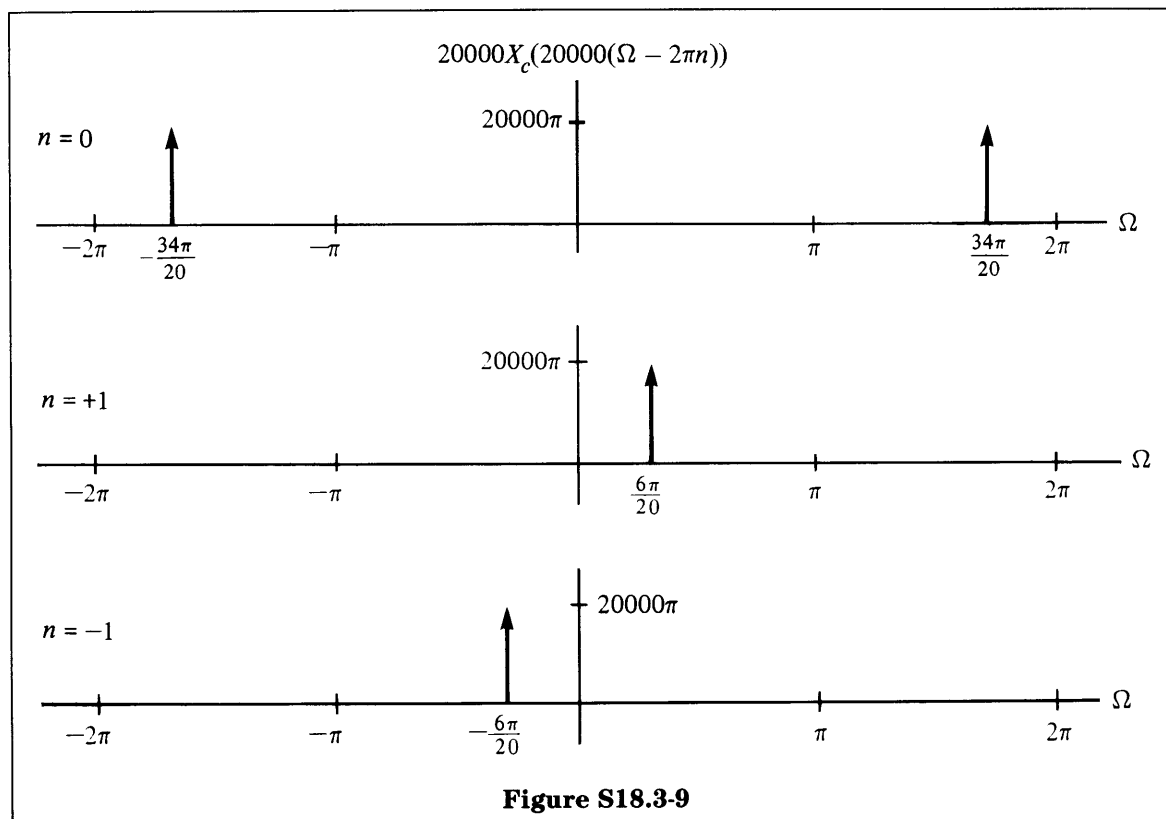


Note aliasing since 27000 Hz is above half the sampling rate of 20000 Hz.

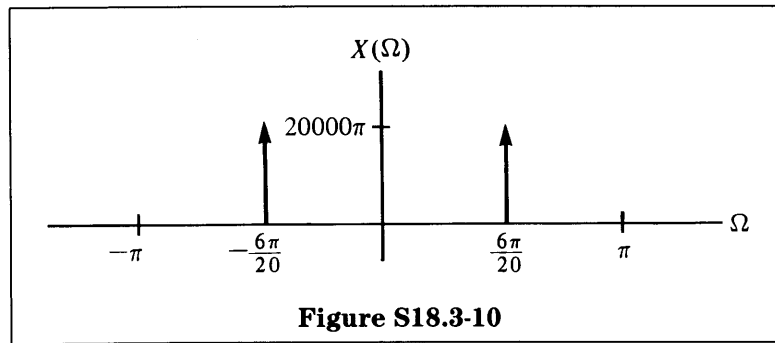
(c)  $X_c(\omega)$  is as given in Figure S18.3-8.



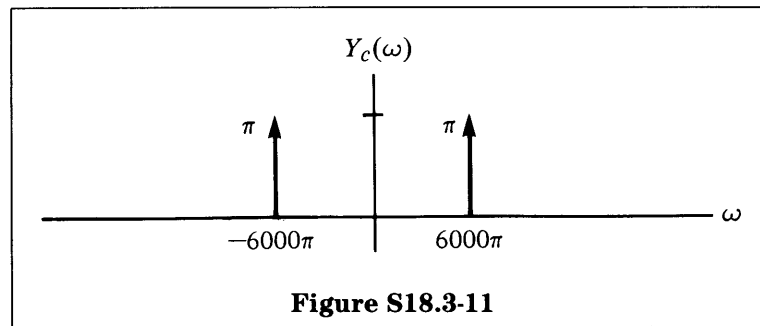
Again we use eq. (S18.3-1), shown in Figure S18.3-9.



Thus  $X(\Omega)$  is given as in Figure S18.3-10.



Finally, from eq. (S18.3-2) we have  $Y_c(\omega)$  shown in Figure S18.3-11.



#### S18.4

It is required that we sample at a rate such that the discrete-time frequency  $\pi/2$  will correspond to  $\omega_c$ . The relation between  $\Omega_c$  and  $\omega_c$  is  $\Omega_c = \omega_c T_0$ . Thus, we require

$$\frac{\pi/2}{\omega_c} = T_0$$

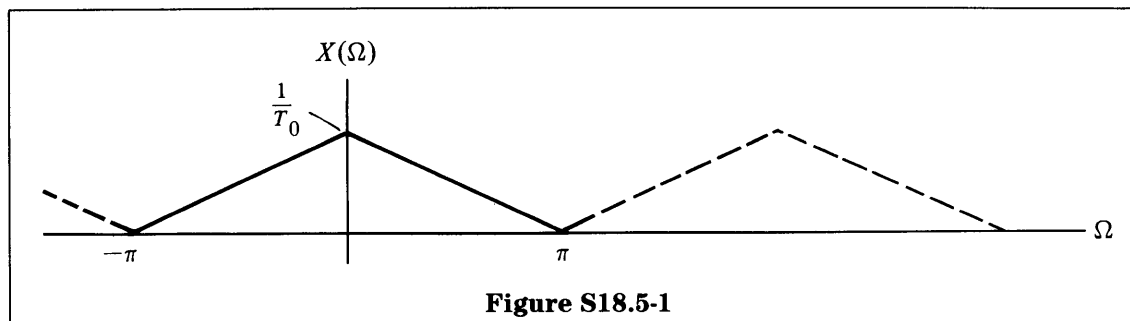
As  $\omega_c$  increases, demanding a wider filter,  $T_0$  decreases, and consequently the sampling frequency must be increased. There are two ways to calculate  $\omega_a$ . First, since we are sampling at a rate of

$$\frac{2\pi}{T_0} \quad \text{or} \quad \frac{2\pi}{(\pi/2)/\omega_c} = 4\omega_c,$$

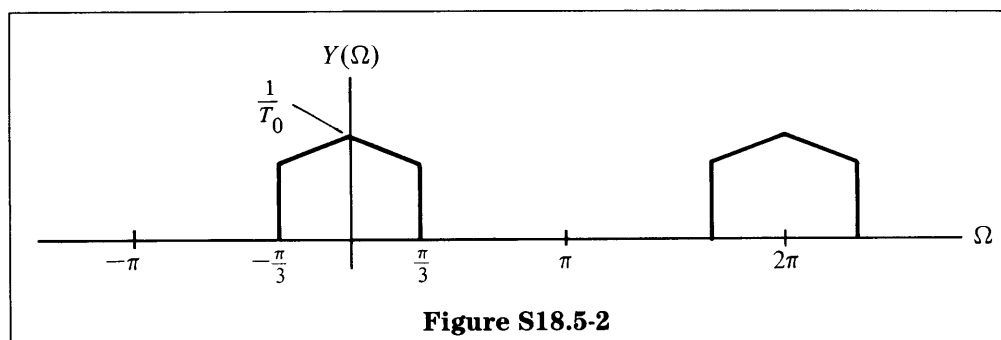
we need an anti-aliasing filter that will remove power at frequencies higher than half the sampling rate; therefore  $\omega_a = 2\omega_c$ . Alternatively, we note that the “folding frequency,” or the frequency at which aliasing begins, is  $\Omega = \pi$ . Since  $\Omega = \pi/2$  corresponds to  $\omega_c$ , then  $\pi$  must correspond to  $2\omega_c$ .

**S18.5**

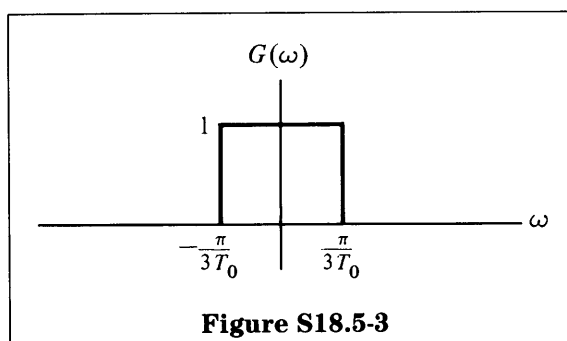
- (a) We sketch  $X(\Omega)$  by stretching the frequency axis so that  $2\pi$  corresponds to the sampling frequency with a gain of  $1/T_0$ . We then repeat the spectrum, as shown in Figure S18.5-1.



After filtering,  $Y(\Omega)$  is given as in Figure S18.5-2.



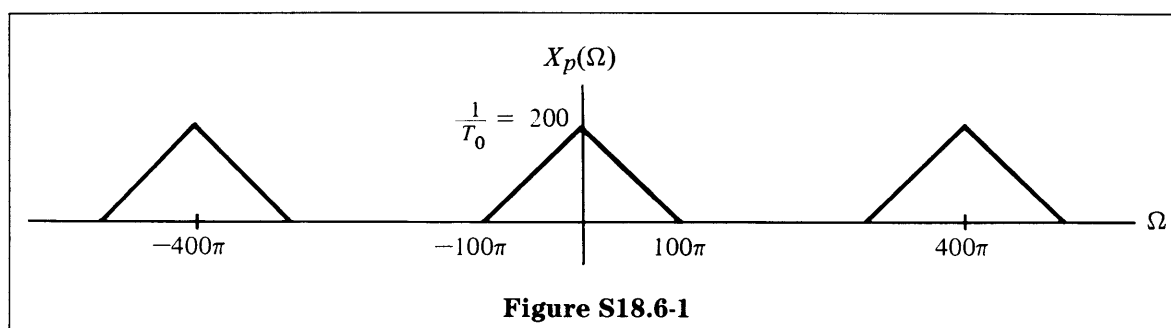
- (b) We see that  $Y(\Omega)$  looks like  $X(\omega)$  filtered and then sampled. The discrete-time frequency is  $\pi/3$ . Again,  $2\pi$  corresponds to  $2\pi/T_0$ , so  $\pi/3$  corresponds to  $\pi/3T_0$ . Thus, if  $x(t)$  is filtered by  $G(\omega)$  as given in Figure S18.5-3, then  $y[n] = z[n]$ .



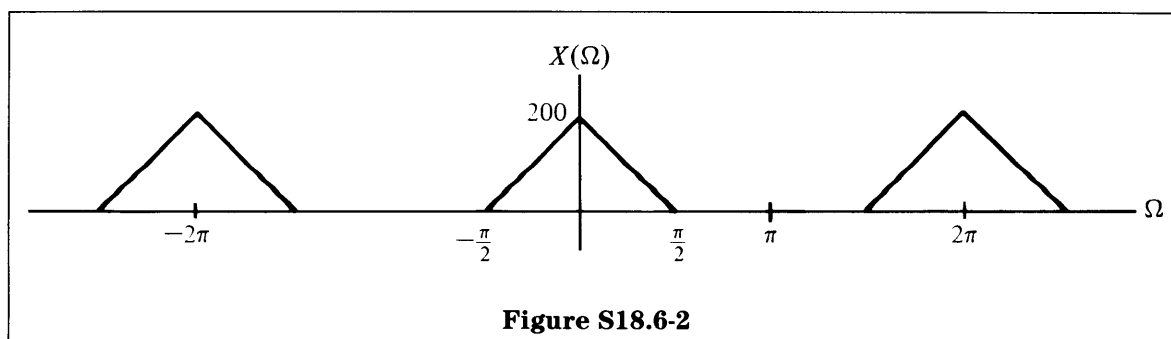
## Solutions to Optional Problems

### S18.6

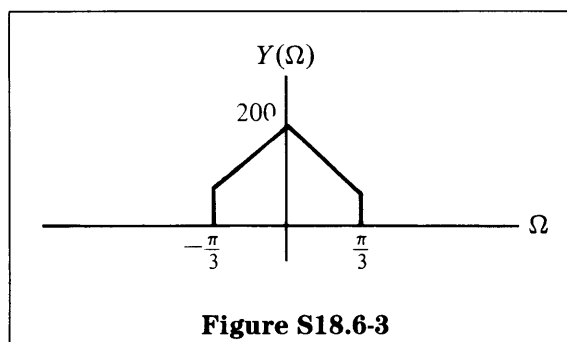
- (a) Since we are allowing all frequencies less than  $100\pi$  through the anti-aliasing filter, we need to sample at least twice  $100\pi$ , or  $200\pi$ . Thus,  $200\pi = 2\pi/T_0$  or  $T_0 = 10$  ms. To find  $K$ , recall that impulse sampling introduces a gain of  $1/T_0$ . To account for this,  $K$  must equal  $T_0$ , or  $K = 0.01$ .
- (b) (i) Since  $X(\omega)$  is bandlimited to  $100\pi$ , the anti-aliasing filter has no effect. The Fourier transform of  $x_p(t)$ , the modulated pulse train, is given in Figure S18.6-1.



Since  $T_0 = 0.005$ , the sampling frequency is  $400\pi$ . After conversion to a discrete-time signal,  $X(\Omega)$  appears as in Figure S18.6-2.



After filtering,  $Y(\Omega)$  is given by Figure S18.6-3.





- (ii) There are three effects to note in D/C conversion: (1) a gain of  $T_0$ , (2) a frequency scaling by a factor of  $T_0$ , and (3) the removal of repeated spectra. Thus,  $Y(\omega)$  is as shown in Figure S18.6-4.

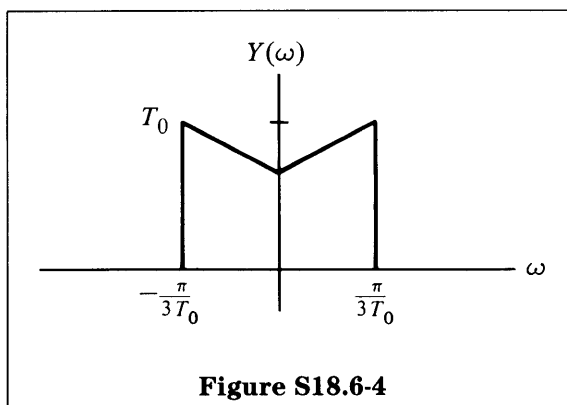


Figure S18.6-4

**S18.7**

After the initial shock, you should realize that this problem is not as difficult as it seems. If instead of  $h[n]$  we had been given the frequency response  $H(\Omega)$ , then  $H_c(\omega)$  would be just a scaled version of  $H(\Omega)$  bandlimited to  $\pi/T$ . Let us find, then,  $H(\Omega)$ . Using properties of the Fourier transform, we have

$$Y(\Omega) = \frac{1}{2}e^{-j\Omega}Y(\Omega) + X(\Omega)$$

$$H(\Omega) = \frac{Y(\Omega)}{X(\Omega)} = \frac{1}{1 - \frac{1}{2}e^{-j\Omega}}$$

Thus,

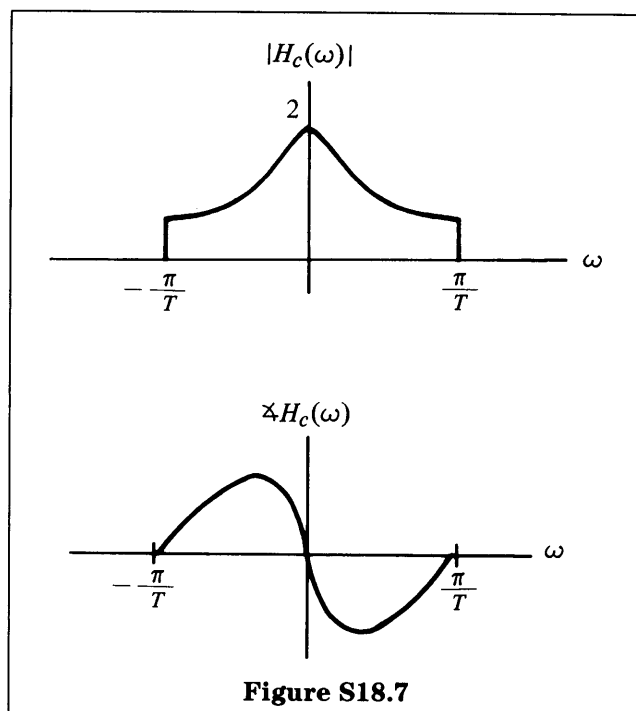
$$|H(\Omega)| = \frac{1}{\sqrt{\frac{5}{4} - \cos \Omega}}$$

$$\angle H(\Omega) = -\tan^{-1} \left( \frac{\frac{1}{2} \sin \Omega}{1 - \frac{1}{2} \cos \Omega} \right)$$

Therefore, the magnitude and phase of  $H_c(\omega)$  are as shown in Figure S18.7.

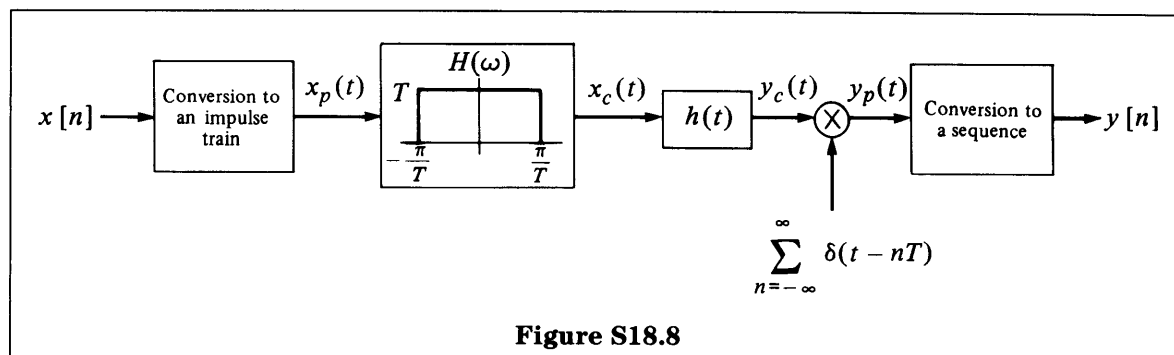
$$|H_c(\omega)| = \begin{cases} \frac{1}{\sqrt{\frac{5}{4} - \cos \omega T}}, & |\omega| < \frac{\pi}{T}, \\ 0, & \text{elsewhere} \end{cases}$$

$$\angle H_c(\omega) = \begin{cases} -\tan^{-1} \left( \frac{\frac{1}{2} \sin \omega T}{1 - \frac{1}{2} \cos \omega T} \right), & |\omega| < \frac{\pi}{T}, \\ 0, & \text{elsewhere} \end{cases}$$



### S18.8

The system under study is shown in Figure S18.8.



From our previous study, we know that  $X_c(\omega)$  in the range  $\pm\pi/T$  looks just like  $X(\Omega)$  in the range  $\pm\pi$ . Similarly,  $Y_c(\omega)$  between  $-\pi/T$  and  $+\pi/T$  looks like  $Y(\Omega)$  in the range  $-\pi$  to  $\pi$ . Although there is a factor of  $T$ , we can disregard it in analyzing this system because it is accounted for in the  $H(\omega)$  filter. The transformation of  $x_c(t)$  to  $y_c(t)$  will correspond to filtering  $x[n]$ , yielding  $y[n]$ . In fact, the equivalent system will have a system function  $H(\Omega)$  given by

$$H(\Omega) = H_c\left(\frac{\Omega}{T}\right), \quad |\Omega| < \pi,$$

where  $H_c(\omega)$  is the Fourier transform of  $h(t)$ . Thus, we need to find  $H_c(\omega)$ . The relation between  $y_c(t)$  and  $x_c(t)$  is governed by the following differential equation:

$$\frac{d^2 y_c(t)}{dt^2} + 4 \frac{dy_c(t)}{dt} + 3y_c(t) = x_c(t)$$

Using the properties of the Fourier transform, we have

$$(j\omega)^2 Y_c(\omega) + 4(j\omega)Y_c(\omega) + 3Y_c(\omega) = X_c(\omega),$$

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

Therefore,

$$H(\Omega) = \frac{1}{\left(j\frac{\Omega}{T}\right)^2 + 4j\frac{\Omega}{T} + 3}, \quad |\Omega| < \pi$$

### S18.9

(a) It is instructive to sketch a typical  $y_p(t)$ , which we have done in Figure S18.9-1.

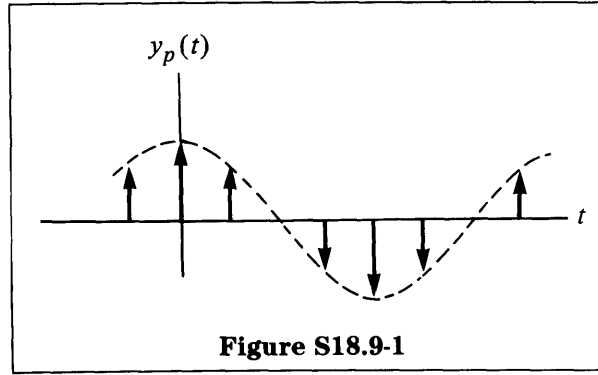


Figure S18.9-1

Let us suppose that  $T$  is changed by being reduced. Then the envelope of  $y_p(t)$  seems to correspond to a higher-frequency cosine. At time  $kt$ ,

$$y_p(t) = \cos \frac{2\pi k}{N} \delta(t - kT) = \cos \frac{2\pi(kT)}{NT} \delta(t - kT) = \cos \frac{2\pi t}{NT} \delta(t - kT),$$

where we use the sampling property of the impulse function. Thus,

$$y_p(t) = \sum_{k=-\infty}^{\infty} \cos \frac{2\pi k}{N} \delta(t - kT) = \cos \omega_0 t \sum_{k=-\infty}^{\infty} \delta(t - kT),$$

where  $\omega_0 = 2\pi/NT$ .

If the minimum  $\omega_0$  is  $\omega_1$ , and since  $T = 2\pi/N\omega_0$ ,

$$T_{\max} = \frac{2\pi}{N\omega_1}$$

Similarly,

$$T_{\min} = \frac{2\pi}{N\omega_2}$$

- (b) Recall that sampling with an impulse train repeats the spectrum with a period of  $2\pi/T$  and a gain factor of  $1/T$ . Since  $\mathcal{F}[\cos(2\pi t/NT)]$  is as given by Figure S18.9-2,  $Y_p(\omega)$  is then given by Figure S18.9-3.

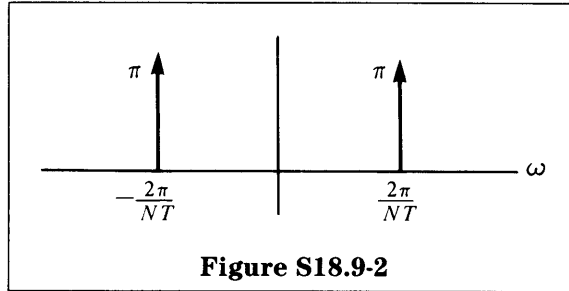


Figure S18.9-2

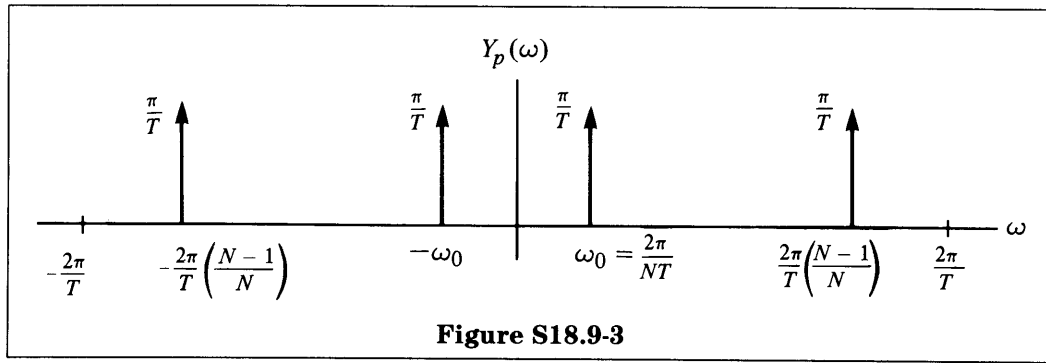


Figure S18.9-3

- (c) The minimum value of  $N$  is 2, corresponding to the impulses at  $\omega_0$  and  $(2\pi/T - \omega_0)$  being superimposed at  $\pi/T$ . The lowpass filter cutoff frequency must be such that the (superimposed) impulses at  $\pi/T$  are in the passband and those at  $3\pi/T$  are outside the passband. Consequently,

$$\frac{\pi}{T} < \omega_c < \frac{3\pi}{T}$$

- (d) Comparing  $Y(\omega)$  and  $Y_p(\omega)$  in Figures S18.9-2 and S18.9-3 respectively, we see that for  $N > 2$  the cosine output will have an amplitude of  $1/T = \omega/2\pi$ . If  $N = 2$ , then the output amplitude will be  $2/T = \omega/\pi$ .

### S18.10

- (a) By sampling  $s_c(t)$ , we get

$$s[n] = s_c(nT) = x(nT) + \alpha x(nT - T_0) = x(nT) + \alpha x[(n-1)T]$$

since  $T = T_0$ . Let  $x[n] = x(nT)$ . Then

$$s[n] = x[n] + \alpha x[n-1]$$

Therefore

$$x[n] = -\alpha x[n-1] + s[n]$$

This is a first-order difference equation, so given  $s[n]$ , we can find  $x[n]$ . Since  $x(t)$  is appropriately bandlimited, we can then set

$$y[n] = -\alpha y[n-1] + s[n]$$

which will make

$$y_c(t) = \frac{A}{T} x(t)$$

**(b)** From part (a) we see that  $T = A$  will make  $y(t) = x(t)$ .

**(c)** Since we do not want to alias, we still need  $T < \pi/\omega_M$ . Now

$$s(t) = x(t) + \alpha x(t - T_0)$$

Taking the *continuous* Fourier transform, we see that

$$S(\omega) = X(\omega) + \alpha e^{-j\omega T_0} X(\omega)$$

Thus, the continuous-time inverse system has frequency response

$$H_c(\omega) = \frac{1}{1 + \alpha e^{-j\omega T_0}}$$

We want to implement this in discrete time. Therefore, using the relation, we obtain

$$H(\Omega) = H_c\left(\frac{\Omega}{T}\right) = \frac{1}{1 + \alpha e^{-j\Omega(T_0/T)}}, \quad \frac{-\pi}{\omega_M} < \Omega < \frac{\pi}{\omega_M}$$

Again, the filter should be  $A = T$ .

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