

## Chapter 4. Digital Processing of Continuous-Time Signals

Even though this book is concerned primarily with the processing of discrete-time signals, most signals we encounter in the real world are continuous in time, such as speech, music and images. The interface circuit performing the conversion of a continuous-time signal into a digital form is called the analog-to-digital (A/D) converter. Likewise, the reverse operation of converting a digital signal into a continuous-time signal is implemented by the interface circuit called the digital-to-analog (D/A) converter.

## 4.2 Sampling of Continuous-Time Signals

### 4.2.1 Effect of Sampling in the Frequency Domain

Let  $g_a(t)$  be a continuous-time signal that is sampled uniformly at  $t = nT$ , generating a sequence  $g[n]$  where

$$g[n] = g_a(nT), \quad -\infty < n < \infty$$

with  $T$  being the sampling period, and  $F_T = \frac{1}{T}$  being the sampling frequency.

The frequency-domain representation of  $g_a(t)$  is given by its continuous-time Fourier transform (CTFT):

$$G_a(j\Omega) = \int_{-\infty}^{\infty} g_a(t) e^{-j\Omega t} dt$$

On the other hand, the frequency-representation of  $g[n]$  is given by the discrete-time Fourier transform:

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

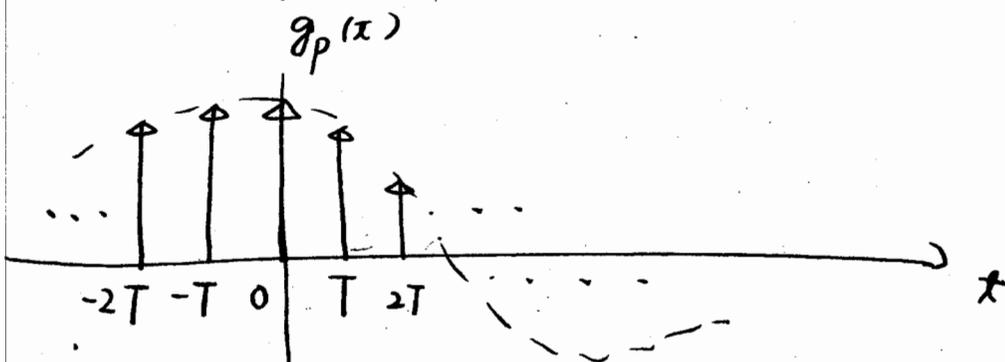
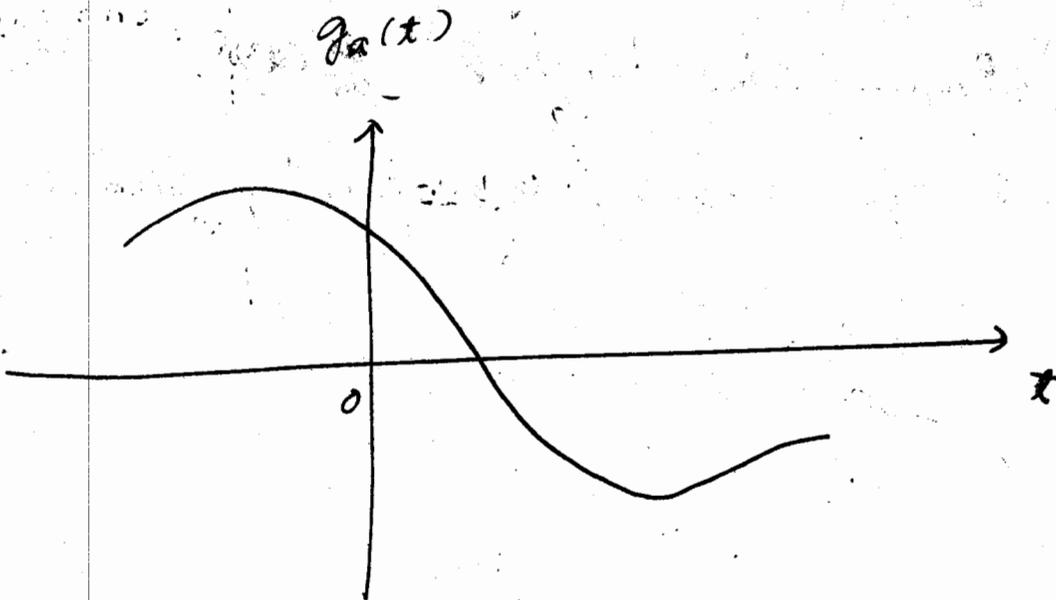
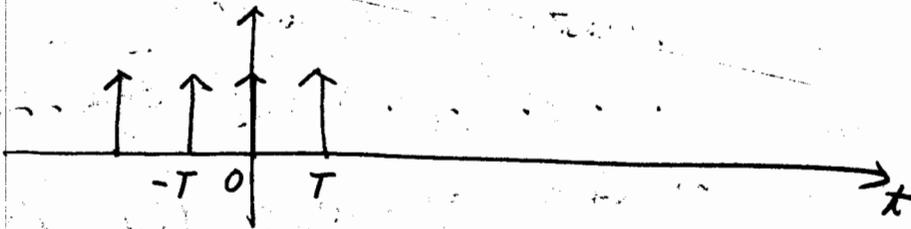
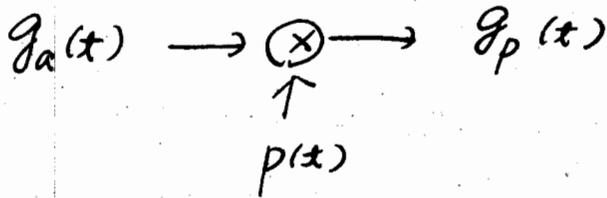
Now we would like to establish the relationship between the CTFT  $G_a(j\Omega)$  and the DTFT  $G(e^{j\omega})$ . Since the discrete-time sequence  $g[n]$  is the sampled version of the continuous-time signal  $g_a(t)$ , we introduce the sampling operation mathematically as a multiplication of  $g_a(t)$  by a periodic impulse train  $p(t)$ :

$$p(t) = \sum_{n=-\infty}^{\infty} \delta(t - nT)$$

Such a multiplication yields an impulse train  $g_p(t)$ :

$$g_p(t) = g_a(t) p(t)$$

$$= \sum_{n=-\infty}^{\infty} g_a(nT) \delta(t-nT)$$



The CTFT  $G_p(j\Omega)$  of  $g_p(t)$  is given by

$$\begin{aligned}
 G_p(j\Omega) &= \sum_{n=-\infty}^{\infty} g_a(nT) \int_{-\infty}^{\infty} \delta(t-nT) e^{-j\Omega t} dt \\
 &= \sum_{n=-\infty}^{\infty} g_a(nT) e^{-j\Omega nT} \\
 &= \frac{1}{T} \sum_{k=-\infty}^{\infty} G_a \left( j \left( \Omega + \frac{2\pi k}{T} \right) \right)
 \end{aligned}$$

Proof:

According to the Poisson's sum formula,

$$\sum_{n=-\infty}^{\infty} \phi(x+nT) = \frac{1}{T} \sum_{k=-\infty}^{\infty} \Phi \left( j \frac{2\pi k}{T} \right) e^{j \frac{2\pi k}{T} x},$$

where  $\Phi(j\Omega)$  is the CTFT of  $\phi(x)$ .

Let  $\phi(x) = g_a(x) e^{-j\Omega x}$ ;

$$\sum_{n=-\infty}^{\infty} g_a(x+nT) e^{-j\Omega(x+nT)}$$

$$= \frac{1}{T} \sum_{k=-\infty}^{\infty} G_a \left( j \left( \frac{2\pi k}{T} + \Omega \right) \right) e^{j \frac{2\pi k}{T} x}$$

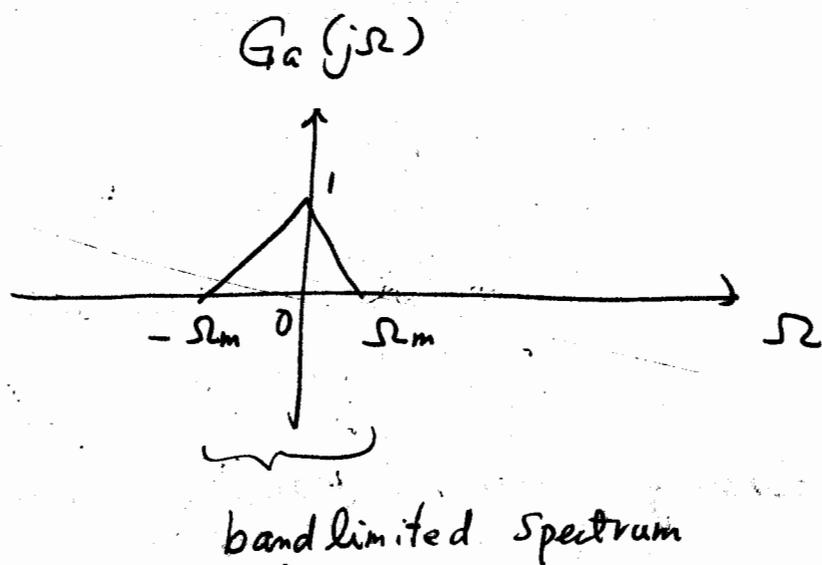
$x=0 \Rightarrow$

$$\sum_{n=-\infty}^{\infty} g_a(nT) e^{-j\Omega nT} = \frac{1}{T} \sum_{k=-\infty}^{\infty} G_a \left( j \left( \Omega + \frac{2\pi k}{T} \right) \right)$$

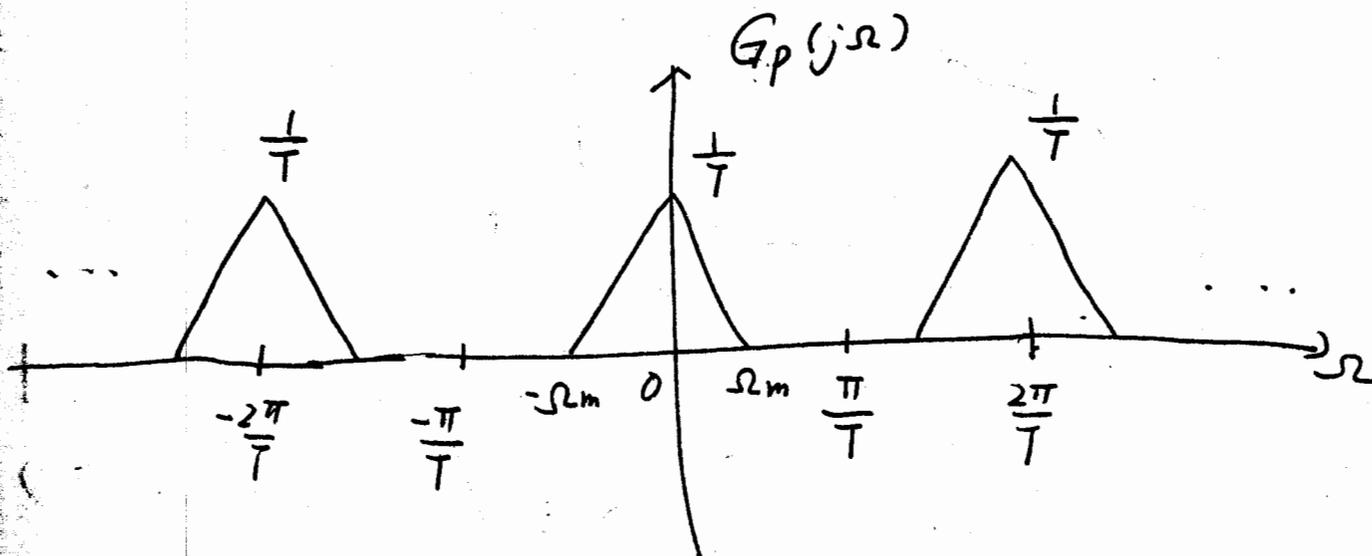
Q.E.D.

# Illustration of the Frequency-domain Effects of time-domain Sampling:

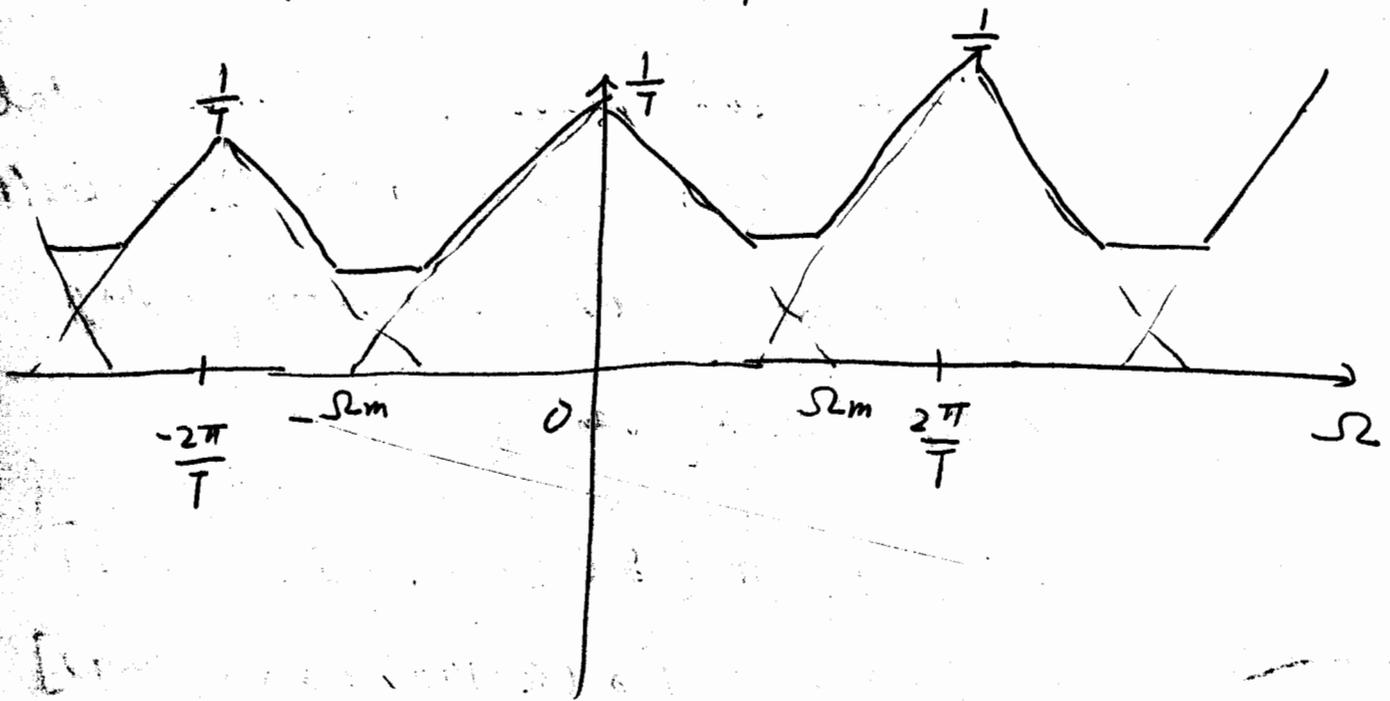
(a) Spectrum of Original continuous-time signal  $g_a(t)$



(b) The spectrum of the sampled signal  $g_p(t)$  when  $\frac{2\pi}{T} > 2\Omega_m$



(c) The spectrum of the sampled signal  $g_p(\omega)$  when  $\frac{2\pi}{T} < 2\Omega_m$  (aliasing)



As illustrated by the figures above, when  $\frac{2\pi}{T} < 2\Omega_m$ , the aliasing or the spectral distortion is observed. The frequency  $\frac{\pi}{T}$  is often referred to as the folding frequency or Nyquist frequency.

### \* Sampling Theorem

Let  $g_a(t)$  be a bandlimited signal with  $G_a(j\Omega) = 0$ , for  $|\Omega| > \Omega_m$ . Then  $g_a(t)$  is

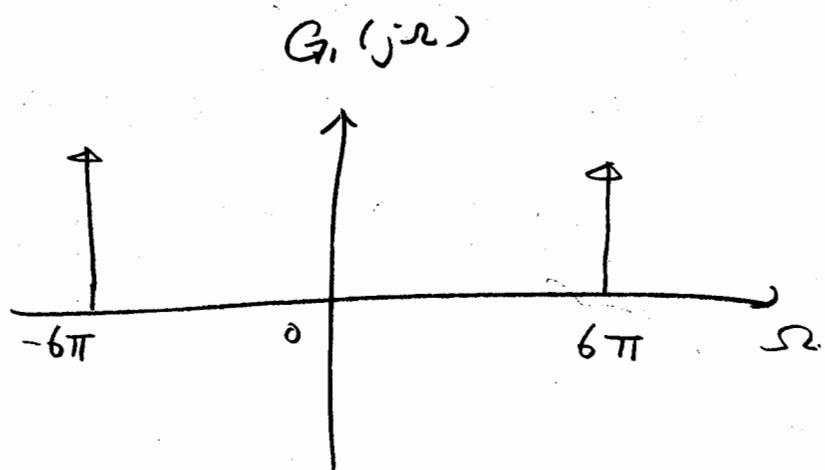
Uniquely determined by its samples  $g_a(nT)$ ,  
 $-\infty < n < \infty$ , if  $\frac{2\pi}{T} \geq 2\Omega_m$ , or  $\frac{\pi}{T} \geq \Omega_m$ .

Example: Consider the three pure sinusoidal signals:  $g_1(t) = \cos(6\pi t)$ ,  $g_2(t) = \cos(14\pi t)$ ,  $g_3(t) = \cos(26\pi t)$ . The corresponding CTFTs are given by

$$G_1(j\Omega) = \pi [\delta(\Omega - 6\pi) + \delta(\Omega + 6\pi)]$$

$$G_2(j\Omega) = \pi [\delta(\Omega - 14\pi) + \delta(\Omega + 14\pi)]$$

$$G_3(j\Omega) = \pi [\delta(\Omega - 26\pi) + \delta(\Omega + 26\pi)]$$



The Nyquist frequency is  $\frac{\pi}{T} = 6\pi$

$$\Rightarrow T \leq \frac{1}{6}$$

Similarly, the Nyquist frequency for  $g_2(t)$

$$\text{is } \frac{\pi}{T} = 14\pi \quad (T \leq \frac{1}{14}) \quad \text{and that for}$$

$$g_3(t) \text{ is } \frac{\pi}{T} = 26\pi \quad (T \leq \frac{1}{26}).$$

We now establish the relation between the discrete-time Fourier transform  $G(e^{j\omega})$  of the sequence  $g[n]$  and the continuous-time Fourier transform  $G_a(j\Omega)$  of the analog signal  $g_a(t)$ . Compare Eqs. (4.3) and (4.6) with (4.1), and we have

$$G(e^{j\omega}) = G_p(j\Omega) \Big|_{\Omega = \frac{\omega}{T}}$$

$$\Rightarrow G_p(j\Omega) = G(e^{j\omega}) \Big|_{\omega = \Omega T}.$$

According to Eq. (4.10), we have

$$\begin{aligned} G(e^{j\omega}) &= \frac{1}{T} \sum_{k=-\infty}^{\infty} G_a(j\Omega - j\frac{2\pi k}{T}) \Big|_{\Omega = \frac{\omega}{T}} \\ &= \frac{1}{T} \sum_{k=-\infty}^{\infty} G_a(j\frac{\omega}{T} - j\frac{2\pi k}{T}) \end{aligned}$$

## 4.2.2 Recovery of the Analog Signal

Now the impulse response  $h_r(t)$  of the ideal lowpass filter is obtained by taking the inverse CTFT of the corresponding frequency response  $H_r(j\Omega)$ :

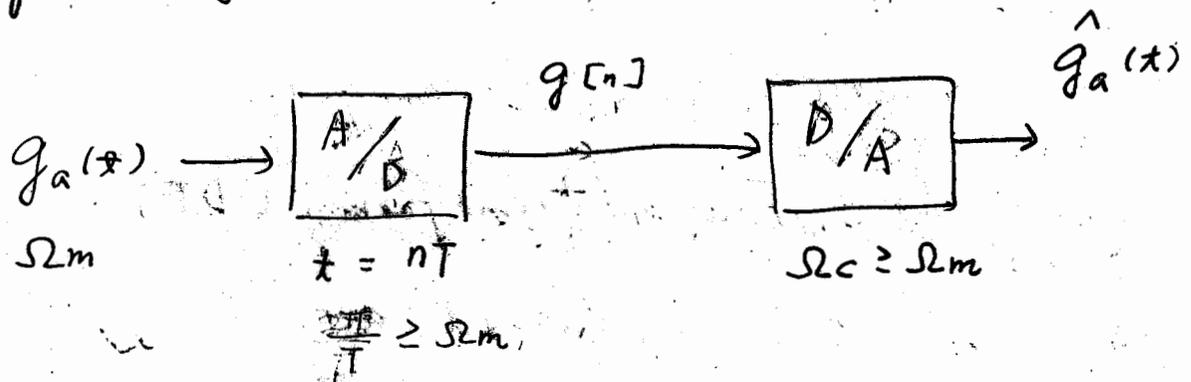
$$H_r(j\Omega) = \begin{cases} T, & |\Omega| \leq \Omega_c \\ 0, & |\Omega| > \Omega_c \end{cases}$$

and is given by

$$\begin{aligned} h_r(t) &= \frac{1}{2\pi} \int_{-\infty}^{\infty} H_r(j\Omega) e^{j\Omega t} d\Omega \\ &= \frac{T}{2\pi} \int_{-\Omega_c}^{\Omega_c} e^{j\Omega t} d\Omega \\ &= \frac{\sin(\Omega_c t)}{\frac{\pi t}{T}}, \quad -\infty < t < \infty \end{aligned}$$

The digital-to-analog converter (D/A) operates like a lowpass filter, which will have the cut-off frequency  $\Omega_c$  larger than

the highest frequency  $\Omega_m$  for the original signal  $g_a(t)$ .



$$g[n] = g_a(t) \Big|_{t=nT}, \quad -\infty < n < \infty$$

$$\begin{aligned} \hat{g}_a(t) &= \sum_{n=-\infty}^{\infty} g[n] h_r(t-nT) \\ &= \sum_{n=-\infty}^{\infty} g[n] \frac{\sin\left[\frac{\pi(t-nT)}{T}\right]}{\frac{\pi(t-nT)}{T}} \end{aligned}$$

It is noted that  $\hat{g}_a(t) = g_a(t)$  only when  $t = nT$ ,  $-\infty < n < \infty$ .  $\hat{g}_a(t)$  may not be equivalent to  $g_a(t)$  away from those time instants  $t = nT$ ,  $-\infty < n < \infty$ .