EENG 479 : Digital Signal Processing (DSP)

Lecture # 3

2.2 Typical sequences and sequence representation 2.3 The sampling Process

Prof. Mohab A. Mangoud

Professor of Wireless Communications (Networks, IoT and AI) University of Bahrain, College of Engineering Department of Electrical and Electronics Engineering P.O.Box 32038- Kingdom of Bahrain <u>mmangoud@uob.edu.bh</u> <u>http://mangoud.com</u>

2.2 Typical sequences and sequence representation

2.2.1 Some Basic Sequences

The most common basic sequences are the unit sample sequence, the unit step sequence, the sinusoidal sequence, and the exponential sequence. These sequences are defined next.

Unit Sample Sequence

The simplest and one of the most useful sequences is the unit sample sequence, often called the discrete-time impulse or the unit impulse, as shown in Figure 2.14(a). It is denoted by $\delta[n]$ and defined by

$$\delta[n] = \begin{cases} 1, & n = 0, \\ 0, & n \neq 0. \end{cases}$$
(2.39)

The unit sample sequence shifted by k samples is thus given by

$$\delta[n-k] = \begin{cases} 1, & n=k, \\ 0, & n\neq k. \end{cases}$$

Figure 2.14(b) shows $\delta[n-2]$. We shall show later in this section that any arbitrary sequence can be represented as a sum of weighted time-shifted unit sample sequences. In Section 2.6.1, we demonstrate that a certain class of discrete-time systems is completely characterized in the time-domain by its output response to a unit impulse input. Furthermore, knowing this particular response of the system, we can compute its response to any arbitrary input sequence.



Figure 2.14: (a) The unit sample sequence $\{\delta[n]\}$ and (b) the shifted unit sample sequence $\{\delta[n-2]\}$.



Figure 2.15: (a) The unit step sequence $\{\mu[n]\}\$ and the shifted unit step sequence $\{\mu[n+2]\}\$.

Unit Step Sequence

A second basic sequence is the *unit step sequence*, shown in Figure 2.15(a). It is denoted by $\mu[n]$ and is defined by

$$\mu[n] = \begin{cases} 1, & n \ge 0, \\ 0, & n < 0. \end{cases}$$

The unit step sequence shifted by k samples is thus given by

Figure 2.15(b) shows $\mu[n+2]$.

The unit sample and the unit step sequences are related as follows (Problem 2.3):

$$\mu[n] = \sum_{m=0}^{\infty} \delta[n-m] = \sum_{k=-\infty}^{n} \delta[k], \qquad (2.41a)$$

$$\delta[n] = \mu[n] - \mu[n-1].$$
(2.41b)

(2.40)

Sinusoidal and Exponential Sequences

A commonly encountered sequence is the real sinusoidal sequence with constant amplitude of the form

$$x[n] = A\cos(\omega_o n + \phi), \qquad -\infty < n < \infty, \tag{2.42}$$

where A, ω_o , and ϕ are real numbers. The parameters A, ω_o , and ϕ are called, respectively, the *amplitude*, the *angular frequency*, and the *phase* of the sinusoidal sequence x[n].

Figure 2.16 shows different types of sinusoidal sequences. The real sinusoidal sequence of Eq. (2.42) can be written alternatively as

$$x[n] = x_i[n] + x_q[n], (2.43)$$

where $x_i[n]$ and $x_q[n]$ are, respectively, the *in-phase* and the quadrature components of x[n], and are given by

$$x_i[n] = A\cos\phi\cos(\omega_o n), \qquad x_q[n] = -A\sin\phi\sin(\omega_o n). \tag{2.44}$$



Figure 2.16: A family of sinusoidal sequences given by $x[n] = 1.5 \cos \omega_0 n$: (a) $\omega_0 = 0$, (b) $\omega_0 = 0.1\pi$, (c) $\omega_0 = 0.2\pi$, (d) $\omega_0 = 0.8\pi$, (e) $\omega_0 = 0.9\pi$, (f) $\omega_0 = \pi$, (g) $\omega_0 = 1.1\pi$, and (h) $\omega_0 = 1.2\pi$.

Another set of basic sequences is formed by taking the *n*th sample value to be the *n*th power of a real or complex constant. Such sequences are termed *exponential sequences*, and their most general form is given by

$$x[n] = A\alpha^n, \qquad -\infty < n < \infty, \qquad (2.45)$$

where A and α are real or complex numbers. By expressing

$$\alpha = e^{(\sigma_o + j\omega_o)}, \qquad A = |A| e^{j\phi},$$

we can rewrite Eq. (2.45) as

$$x[n] = Ae^{(\sigma_o + j\omega_o)n} = |A| e^{\sigma_o n} e^{j(\omega_o n + \phi)}$$

$$= |A| e^{\sigma_o n} \cos(\omega n + \phi) + i |A| e^{\sigma_o n} \sin(\omega n + \phi)$$
(2.46a)
(2.46b)

to arrive at an alternative general form of a complex exponential sequence where σ_o , ϕ , and ω_o are now real numbers. If we write $x[n] = x_{re}[n] + jx_{im}[n]$, then from Eq. (2.46b),

 $x_{\rm re}[n] = |A| e^{\sigma_o n} \cos(\omega_o n + \phi),$ $x_{\rm im}[n] = |A| e^{\sigma_o n} \sin(\omega_o n + \phi).$

Thus, the real and imaginary parts of a complex exponential sequence are real sinusoidal sequences with constant ($\sigma_o = 0$), growing ($\sigma_o > 0$), or decaying ($\sigma_o < 0$) amplitudes for n > 0. Figure 2.17 depicts a complex exponential sequence with a decaying amplitude. Note that in the display of a complex exponential sequence, its real and imaginary parts are shown separately.



Figure 2.17: A complex exponential sequence $x[n] = e^{(-1/12 + j\pi/6)n}$. (a) Real part and (b) imaginary part.



Figure 2.18: Examples of real exponential sequences: (a) $x[n] = 0.2(1.2)^n$, (b) $x[n] = 20(0.9)^n$.

Assignment (1)

Matlab Exercises

Example 2.8: Determination of the period of sinusoidal sequence

Example 2.9: Generation of a sequence wave sequence Assignment (2)

page 115, M2.1,3,4

Due date next week

2.2.2 Sequence Generation Using MATLAB

MATLAB includes a number of functions that can be used for signal generation. Some of these functions of interest are

exp, sin, cos, square, sawtooth

For example, the code fragments to generate a length-N complex exponential sequence with an exponent a + jb and of the form shown in Figure 2.17 is given by

n = 1:N;x = K*exp((a + b*i)*n);

The complete code is given in Program 2.2. Likewise, the code fragments to generate a length-N + 1 real exponential sequence with an exponent a and of the form shown in Figure 2.18 is given by

n = 0:N; x = K*a.^n;

The complete code is given in Program 2.3. Another type of sequence generation using MATLAB can be found earlier in Example 2.1.

```
% Program 2 2
% Generation of complex exponential sequence
%
a = input('Type in real exponent = ');
b = input('Type in imaginary exponent = ');
c = a + b^*i;
K = input('Type in the gain constant = ');
N = input ('Type in length of sequence = ');
n = 1:N;
x = K^* exp(c^*n);
                                        %Generate the sequence
```

%Plot the real part

stem(n,real(x));

xlabel('Time index n'); ylabel('Amplitude'); title('Real part');

disp('PRESS RETURN for imaginary part');

pause

stem(n,imag(x));%Plot the imaginary part xlabel('Time index n');ylabel('Amplitude'); title('Imaginary part');

- % Program 2_3
- % Generation of real exponential sequence
- %
- a = input('Type in argument = ');
- K = input('Type in the gain constant = ');
- N = input ('Type in length of sequence = ');
- n = 0:N;
- x = K*a.^n;
- stem(n,x);
- xlabel('Time index n');ylabel('Amplitude');
- title(['\alpha = ',num2str(a)]);

2.2.3 Representation of an Arbitrary Sequence

An arbitrary sequence can be represented in the time-domain as a weighted sum of a basic sequence and its delayed versions. A commonly used basic sequence in the representation of an arbitrary sequence is the unit sample sequence. For example, the sequence x[n] of Figure 2.21 can be expressed as

$$x[n] = 0.5\,\delta[n+2] + 1.5\,\delta[n-1] - \delta[n-2] + \delta[n-4] + 0.75\,\delta[n-6]. \tag{2.52}$$

An implication of this type of representation is considered later in Section 2.5.1, where we develop the general expression for calculating the output sequence of certain types of discrete-time systems for an arbitrary input sequence.

Since the unit step sequence and the unit sample sequence are simply related through Eq. (2.41a), it is also possible to represent an arbitrary sequence as a weighted combination of delayed unit step sequences (Problem 2.6)



2.3 The Sampling Process

We indicated earlier that often the discrete-time sequence is developed by uniformly sampling a continuoustime signal $x_a(t)$, as illustrated in Figure 2.2. The relation between the two signals is given by Eq. (2.2), where the time variable t of the continuous-time signal is related to the time variable n of the discrete-time signal only at discrete-time instants t_n given by

$$t_n = nT = \frac{n}{F_T} = \frac{2\pi n}{\Omega_T},\tag{2.53}$$

with $F_T = 1/T$ denoting the sampling frequency and $\Omega_T = 2\pi F_T$ denoting the sampling angular frequency. For example, if the continuous-time signal is

$$x_a(t) = A\cos(2\pi f_o t + \phi) = A\cos(\Omega_o t + \phi), \qquad (2.54)$$

the corresponding discrete-time signal is given by

$$x[n] = A\cos(\Omega_o nT + \phi)$$

= $A\cos\left(\frac{2\pi\Omega_o}{\Omega_T}n + \phi\right) = A\cos(\omega_o n + \phi),$ (2.55)

where

$$\omega_o = \frac{2\pi \,\Omega_o}{\Omega_T} = \Omega_o T \tag{2.56}$$

is the normalized digital angular frequency of the discrete-time signal x[n]. The unit of the normalized digital angular frequency ω_o is radians per sample, while the unit of the normalized analog angular frequency Ω_o is radians per second and the unit of the analog frequency f_o is hertz if the unit of the sampling period T is in seconds.

See Example 2.11



the solid line describes a 0.5Hz continuous-time sinusoidal signal and the dash-dot line describes a 1.5 Hz continuous time sinusoidal signal.

When both signals are sampled at the rate of *Fs equals two* samples/sec, their samples coincide, as indicated by the circles in the figures. This means that *x* one of *n* (*T* sub *s*) is equal to *x* two of *n* (*T* sub *s*) and there is no way to distinguish the two signals apart from their sampled versions.

This phenomenon, known as aliasing, occurs whenever F2 plus or minus F1 is a multiple of the sampling rate



Figure 2.22: Ambiguity in the discrete-time representation of continuous-time signals. $g_1(t)$ is shown with the solid line, $g_2(t)$ is shown with the dashed line, $g_3(t)$ is shown with the dashed-dot line, and the sequence obtained by sampling is shown with circles.

Frequency = 3, 7, 13 Hz, sampling rate = 10 Hz with T=0.1 sec g1[n]=cos(0.6 .pi. n), g2[n]=cos(1.4 .pi. n), g3[n]=cos(2.6 .pi. n)

As a result all three sequences above are identical and it is difficult to Associate a unique continuous time function with any one of these sequences In the general case, the family of continuous-time sinusoids

 Ω_T

$$x_{a,k}(t) = A\cos(\pm(\Omega_o t + \phi) + k\Omega_T t), \qquad k = 0, \pm 1, \pm 2, \dots$$

(2.57)

(2.58)

leads to identical sampled signals:

$$x_{a,k}(nT) = A\cos\left((\Omega_o + k\Omega_T)nT + \phi\right) = A\cos\left(\frac{2\pi(\Omega_o + k\Omega_T)n}{\Omega_T} + \phi\right)$$
$$= A\cos\left(\frac{2\pi\Omega_o n}{\Omega_T} + \phi\right) = A\cos(\omega_o n + \phi) = x[n],$$

The above phenomenon of a continuous-time sinusoidal signal of higher frequency acquiring the identity of a sinusoidal sequence of lower frequency after sampling is called *aliasing*. Since there are an infinite number of continuous-time functions that can lead to a given sequence when sampled periodically, additional conditions need to be imposed so that the sequence $\{x[n]\} = \{x_a(nT)\}$ can uniquely represent the parent continuous-time function $x_a(t)$. In which case, $x_a(t)$ can be fully recovered from a knowledge of $\{x[n]\}$.

EXAMPLE 2.12 Illustration of Aliasing

Determine the discrete-time signal v[n] obtained by uniformly sampling a continuous-time signal $v_a(t)$ composed of a weighted sum of five sinusoidal signals of frequencies 30 Hz, 150 Hz, 170 Hz, 250 Hz, and 330 Hz, at a sampling rate of 200 Hz, as given below:



Aliasing Demo (1)

The phenomenon of aliasing happens when the sampling frequency is less than twice the highest frequency of band-limited input signal.

In this example, the input signal is a sinusoidal signal of frequency 1.8KHz.

Three output sound signals are generated in sampling rates 8KHz, 4KHz and 2.6667KHz respectively.

Among these three outputs, we can observe that the aliasing arises only at sampling frequency of 2.6667KHz, which is less than twice of the highest input frequency 3.6KHz.

MATLAB file: aliasing.m