# ${\rm EITF75:}$ Systems and Signals - Reference Sheet

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### 1 Sinusoids

There are several ways to characterize Sinusoids. The first is by dimension:

- 1. Multidimensional/Multichannel Signals
- 2. Monodimensional/Monochannel Signals

You can also classify sinusoids by their independent variable (usually time) and the values they take.

- 1. Continuous-Time Signals or Analog Signals
- 2. Discrete-Time Signals
- 3. There is a third way to classify sinusoids and their signals: Digital Signals

**Defn 1** (Continuous-Time Signals). Continuous-time signals or Analog signals are defined for every value of time and they take on values in the continuous interval (a, b), where a can be  $-\infty$  and b can be  $\infty$ . Mathematically, these signals can be described by functions of a continuous variable.

For example,

$$x_1(t) = \cos \pi t, \ x_2(t) = e^{-|t|}, \ -\infty < t < \infty$$

**Defn 2** (Discrete-Time Signals). *Discrete-time signals* are defined only at certain specified values of time. These time instants *need not* be equidistant, but in practice, they are usually taken at equally speced intervals for computation convenience and mathematical tractability.

For example,

$$x(t_n) = e^{-|t_n|}, n = 0, \pm 1, \pm 2, \dots$$

A Discrete-Time Signals can be represented mathematically by a sequence of real or complex numbers.

Remark 2.1. To emphasize the discrete-time nature of the signal, we shall denote the signal as x(n), rather than x(t).

Remark 2.2. If the time instants  $t_n$  are equally spaced (i.e.,  $t_n = nT$ ), the notation x(nT) is also used.

### 1.1 Continuous-Time Signals

#### 1.1.1 Frequency in Continuous-Time Signals

A simple harmonic oscillation is mathematically described by Equation (1.1).

$$x_a(t) = A\cos(\Omega t + \theta), -\infty < t < \infty$$
 (1.1)

Remark. The subscript a is used with x(t) to denote an analog signal.

This signal is completely characterized by three parameters:

- 1. A, the amplitude of the sinusoid
- 2.  $\Omega$ , the frequency in radians per second (rad/s)
- 3.  $\theta$ , the *phase* in radians.

Instead of  $\Omega$ , the frequency F in cycles per second or hertz (Hz) is used.

$$\Omega = 2\pi F \tag{1.2}$$

Plugging (1.2) into (1.1), yields

$$x_a(t) = A\cos(2\pi F t + \theta), -\infty < t < \infty$$
(1.3)

#### 1.1.2 Properties of Continuous-Time Sinusoidal Signals

The analog sinusoidal signal in equation (1.3) is characterized by the following properties:

(i) For every fixed value of the frequency F,  $x_a(t)$  is periodic.

$$x_a(t+T_p) = x_a(t)$$

where  $T_p = \frac{1}{F}$  is the fundamental period.

- (ii) Continuous-time sinusoidal signals with distinct (different) frequencies are themselves distinct.
- (iii) Increasing the frequency F results in an increase in the rate of oscillation of the signal, in the sense that more periods are included in the given time interval.

### 1.2 Discrete-Time Signals

### 1.2.1 Frequency in Discrete-Time Signals

A discrete-time sinusoidal signal may be expressed as

$$x(n) = A\cos(\omega n + \theta), \ n \in \mathbb{Z}, \ -\infty < n < \infty$$
(1.4)

The signal is characterized by these parameters:

- 1. n, the sample number. MUST be an integer.
- 2. A, the amplitude of the sinusoid
- 3.  $\omega$ , the angular frequency in radians per sample
- 4.  $\theta$ , is the *phase*, in radians.

Instead of  $\omega$ , we use the frequency variable f defined by

$$\omega \equiv 2\pi f \tag{1.5}$$

Using (1.4) and (1.5) yields

$$x(n) = A\cos(2\pi f n + \theta), n \in \mathbb{Z}, -\infty < n < \infty$$
(1.6)

### 1.2.2 Properties of Discrete-Time Sinusoidal Signals

- (i) A discrete-time sinusoid is periodic *ONLY* if its frequency is a rational number.
- (ii) Discrete-time sinusoids whose frequencies are separated by an integer multiple of  $2\pi$  are identical. This leads us to the idea of a Frequency Alias.
- (iii) The highest rate of oscillation in a discrete-time sinusoid is attained when  $\omega = \pm \pi$  or, equivalently,  $f = \pm \frac{1}{2}$ .

#### 1.2.3 Frequency Aliases

The concept of a Frequency Alias is drawn from the idea that discrete-time sinusoids whose frequencies are separated by an integer multiple of  $2\pi$  are identical and that frequencies  $|f| > \frac{1}{2}$  are identical. (Properties (ii) and (iii))

**Defn 3** (Frequency Alias). A frequency alias is a sinusoid having a frequency  $|\omega| > \pi$  or  $|f| > \frac{1}{2}$ . This is because this sinusoid is indistinguishable (identical) to one with frequency  $|\omega| < \pi$  or  $|f| < \frac{1}{2}$ .

A frequency alias is a sequence resulting from the following assertion based on the sinusoid  $\cos(\omega_0 n + \theta)$ .

It follows that

$$\cos \left[ (\omega_0 + 2\pi) \, n + \theta \right] = \cos \left( \omega_0 n + 2\pi n + \theta \right) = \cos(\omega_0 n + \theta)$$

As a result, all sinusoidal sequences

$$x_k(n) = A\cos(\omega_k n + \theta), k = 0, 1, 2, ...$$

where

$$\omega k = \omega_0 + 2k\pi, \ -\pi \le \omega_0 \le \pi$$

are indistinguishable (i.e., identical).

Because of this, we regard frequencies in the range of  $-\pi \le \omega \le \pi$  or  $-\frac{1}{2} \le f \le \frac{1}{2}$  as unique, and all frequencies that fall outside of these ranges as aliases.

Remark 3.1. It should be noted that there is a difference between discrete-time sinusoids and continuous-time sinusoids have distinct signals for  $\Omega$  or F in the entire range  $-\infty < \Omega < \infty$  or  $-\infty < F < \infty$ .

### 1.3 Sampling Rates and Sampling Frequency

Most signals of interest are analog. To process these signals, they must be collected and converted to a digital form, that is, to convert them to a sequence of numbers having finite precision. This is called analog-to-digital (A/D) conversion. Conceptually, we view this conversion as a 3-step process.

- 1. Sampling
- 2. Quantization
- 3. Coding

#### 1.3.1 Nyquist Rate

#### 1.3.2 Nyquist Frequency

### 1.4 Digital Signals

**Defn 4** (Digital Signals). *Digital signals* are a subset of Discrete-Time Signals. In this case, not only are the values being measured occurring at fixed points in time, the values themselves can only take certain, fixed values.

#### 1.4.1 Quantization

**Defn 5** (Quantization). This is the conversion of a discrete-time continuous-valued signal into a discrete-time, discrete-value (digital) signal. The value of each signal sample is represented by a value selected from a finite set of possible values. The difference between the unquantized sample x(n) and the quantized output  $x_q(n)$  is called the Quantization Error.

#### 1.4.1.1 Quantization Levels

#### 1.4.1.2 Quantization Error

**Defn 6** (Quantization Error). The quantization error of something.

### 1.4.1.3 Bit Requirements

#### 1.4.1.4 Bit Rate

### 2 Discrete-Time Systems

As discussed in Section 1.2, x(n) is a function of an independent variable that is an integer. It is important to note that a discrete-time signal is not defined at instants between the samples. Also, if n is not an integer, x(n) is not defined.

Besides graphical representation of a discrete-time system, there are 3 ways to represent a discrete-time signal.

- 1. Functional Representation
- 2. Tabular Representation
- 3. Sequence Representation

### 2.1 Representing Discrete-Time Systems

### 2.1.1 Functional Representation

This representation of a discrete-time system is done as a mathematical function.

$$x(n) = \begin{cases} 1, & \text{for } n = 1, 3\\ 4, & \text{for } n = 2\\ 0, & \text{elsewhere} \end{cases}$$
 (2.1)

#### 2.1.2 Tabular Representation

This representation of a discrete-time sysem is done as a table of corresponding values.

#### 2.1.3 Sequence Representation

There are 2 methods of representation for this. The first includes all values for  $-\infty < n < \infty$ . In all cases, n = 0 is marked in the sequence, somehow. I will do this with an underline.

$$x(n) = \{\dots, 0, 0, 1, 4, 1, 0, 0, \dots\}$$
(2.2)

The second only works if all x(n) values for n < 0 are 0.

$$x(n) = \{ 0, 1, 4, 1, 0, 0, \dots \}$$
 (2.3)

A finite-duration sequence can be represented as

$$x(n) = \{3, -1, \underline{-2}, 5, 0, 4, -1\}$$

$$(2.4)$$

This is identified as a seven-point sequence.

A finite-duration sequence where x(n) = 0 for all n < 0 is represented as

$$x(n) = \{\underline{0}, 1, 4, 1\} \tag{2.5}$$

This is identified as a four-point sequence.

### 2.2 Elementary Discrete-Time Signals

The following signals are basic signals that appear often and play an important role in signal processing.

### 2.2.1 Unit Impulse Signal

**Defn 7** (Unit Impulse Signal). The unit impulse signal or unit sample sequence is denoted as  $\delta(n)$  and is defined as

$$\delta(n) \equiv \begin{cases} 1, & \text{for } n = 0\\ 0, & \text{for } n \neq 0 \end{cases}$$
 (2.6)

This function is a signal that is zero everywhere, except at n = 0, where its value is 1.

Remark 7.1. This signal is different that the analog signal  $\delta(t)$ , which is also called a unit impulse, and is defined to be 0 everywhere except t = 0. The discrete unit impulse sequence is much less mathematically complicated.

#### 2.2.2 Unit Step Signal

**Defn 8** (Unit Step Signal). The unit step signal is denoted as u(n) or as  $\mathcal{U}(n)$  and is defined as

$$\mathcal{U}(n) \equiv \begin{cases} 1, & \text{for } n \ge 0\\ 0, & \text{for } n < 0 \end{cases}$$
 (2.7)

#### 2.2.3 Unit Ramp Signal

**Defn 9** (Unit Ramp Signal). The unit ramp signal is denoted as  $u_r(n)$  and is defined as

$$u_r(n) \equiv \begin{cases} n, & \text{for } n \ge 0\\ 0, & \text{for } n < 0 \end{cases}$$
 (2.8)

#### 2.2.4 Exponential Signal

**Defn 10** (Exponential Signal). The exponential signal is a sequence of the form

$$x(n) = a^n \text{ for all } n \tag{2.9}$$

If a is real, then x(n) is a real signal. When a is complex valued  $(a \equiv b \pm cj)$ , it can be expressed as

$$x(n) = r^n e^{j\theta n}$$
  
=  $r^n (\cos \theta n + j \sin \theta n)$  (2.10)

This can be expressed by graphing the real and imaginary parts

$$x_R(n) \equiv r^n \cos \theta n$$
  

$$x_I(n) \equiv r^n j \sin \theta n$$
(2.11)

or by graphing the amplitude function and phase function.

$$|x(n)| = A(n) \equiv r^n$$

$$\angle x(n) = \phi(n) \equiv \theta n$$
(2.12)

### 2.3 Classification of Discrete-Time Signals

In order to apply some mathematical methods to discrete-time signals, we must characterize these signals.

#### 2.3.1 Energy Signal

**Defn 11** (Energy Signal). The energy E of a signal x(n) is defined as

$$E \equiv \sum_{n=-\infty}^{\infty} |x(n)|^2 \tag{2.13}$$

The energy of a signal can be finite or infinite. If E is finite  $(0 < E < \infty)$ , then x(n) is called an energy signal.

#### 2.3.2 Power Signal

**Defn 12** (Power Signal). The average power of a discrete time signal x(n) is defined as

$$P = \lim_{N \to \infty} \frac{1}{2N+1} \sum_{n=-N}^{N} |x(n)|^2$$
 (2.14)

This means that there are 2 potential outcomes:

- 1. If E is finite, P=0
- 2. If E is infinite, P may be either finite or infinite

If P is finite and nonzero, the signal is called a *power signal*.

#### 2.3.3 Periodic and Aperiodic Signals

A signal x(n) is periodic with period N (N > 0) if and only if

$$x(n+N) = x(n)$$
for all  $n$  (2.15)

The smallest value of N for which (2.15) holds is called the fundamental period. If there is no value of N that satisfies (2.15), the signal is called *nonperiodic* or *aperiodic*.

#### 2.3.4 Symmetric and Antisymmetric Signals

A real-valued signal x(n) is called *symmetric* or *even* if

$$x(n) = x(-n) \tag{2.16}$$

On the other hand, a signal x(n) is called *antisymmetric* or odd if

$$x(n) = -x(-n) \tag{2.17}$$

### 2.4 Discrete-Time Signal Manipulations

#### 2.4.1 Transformation of the Independent Variable (Time)

It is important to note that Shifting in Time and Folding are not commutative. For example,

$$TD_{k}{FD[x(n)]} = TD_{k}[x(-n)] = x(-n+k)$$
 (2.18)

whereas

$$FD\{TD_{k}[x(n)]\} = FD[x(n-k)] = x(-n-k)$$
 (2.19)

**2.4.1.1 Shifting in Time** A signal x(n) may be shifted in time by replacing the independent variable n by n-k, where k is an integer. If k is a positive integer, the time shift results in a delay of the signal by k units of time (moves left). If k is a negative integer, the time shift results in an advance of the signal by |k| units of time (moves right).

This could be denoted by

$$TD_{k}[x(n)] = x(n-k) \tag{2.20}$$

You cannot advance a signal that is being generated in real-time. Because that would involve signal samples that haven't been generated yet. So, you can only advance a signal that is stored on something. However, you can always introduce a delay to a signal.

**2.4.1.2** Folding Another useful modification of the time base is to replace n with -n. The result is a folding or reflection of the original signal around n = 0.

This could be denoted by

$$FD[x(n)] = x(-n) \tag{2.21}$$

### 2.4.2 Addition, Multiplication, and Scaling

Amplitude modifications include Addition, Multiplication, and Amplitude Scaling.

**2.4.2.1** Addition The sum of 2 signals  $x_1(n)$  and  $x_2(n)$  is a signal y(n) whose value at any instant is equal to the sum of the values of these two signals at that instant.

$$y(n) = x_1(n) + x_2(n), -\infty < n < \infty$$
(2.22)

**2.4.2.2** Multiplication The *product* of two signals  $x_1(n)$  and  $x_2(n)$  is a signal y(n) whose value at any instant is equal to the product of the values of these two signals at that instant.

$$y(n) = x_1(n)x_2(n), -\infty < n < \infty$$
 (2.23)

**2.4.2.3** Amplitude Scaling Amplitude scaling of a signal by a constant A is accomplished by multiplying every signal smaple by A. Consequently, we obtain

$$y(n) = Ax(n), -\infty < n < \infty \tag{2.24}$$

### 3 Convolutions

**Defn 13** (Convolution). The convolution operator.

$$y(t) = \sum_{k=-\infty}^{\infty} x(k) * h(n-k)$$
(3.1)

### 4 The Z-Transform

The Z-Transform plays the same role in the analysis of discrete-time signals and LTI systems as the Laplace Transform does in the analysis of continuous time-signals and LTI systems.

#### 4.1 The Z-Transform

**Defn 14** (Z-Transform). The z-transform is defined as the power series

$$X(z) \equiv \sum_{n=-\infty}^{\infty} x(n)z^{-n}$$
(4.1)

Remark 14.1. For convenience, the z-transform of a signal x(n) is denoted by

$$X(z) \equiv \mathbb{Z}\{x(n)\}\tag{4.2}$$

and the relationship between x(n) and X(z) is indicated by

$$x(n) \stackrel{\mathbf{z}}{\longleftrightarrow} X(z)$$
 (4.3)

### 4.1.1 Region of Convergence

**Defn 15** (ROC). The *ROC* or region of convergence is the region for which the infinite power series in the z-transform has a convergent solution.

Remark 15.1. Any time we cite a z-transform, we should also indicate its ROC

### Example 4.1: Simple Z-Transform.

Determine the z-transform of the signal

$$x(n) = \left(\frac{1}{2}\right)^n \mathcal{U}(n)$$

The z-transform is the infinite power series

$$X(z) = 1 + \frac{1}{2}z^{-1} + \left(\frac{1}{2}\right)^{-2} + \dots + \left(\frac{1}{2}\right)^{n}z^{-n} + \dots$$
$$= \sum_{n=0}^{\infty} \left(\frac{1}{2}\right)^{n}z^{-n} = \sum_{n=0}^{\infty} \left(\frac{1}{2}z^{-1}\right)^{n}$$

Because this is an infinite geometric series, we can solve with with our equivalency:

$$1 + A + A^2 + \dots + A^n + \dots = \frac{1}{1 - A}$$
 if  $|A| < 1$ 

Thus, X(z) converges to

$$X(z) = \frac{1}{1 - \frac{1}{2}z^{-1}}, \quad \text{ROC}: |z| > \frac{1}{2}$$

Signal	ROC
Causal Anticausal Two-Sided	Finite-Duration Signals Entire z-plane except $z=0$ Entire z-plane except $z=\infty$ Entire z-plane except $z=0$ and $z=\infty$
Causal Anticausal Two-Sided	Infinite-Duration Signals $  z  > r_2 $ $  z  < r_1 $ $ r_2 <  z  < r_1 $

Table 4.1: Characteristic Familes of Signals with Their Corresponding ROCs

#### 4.1.2 The One-Sided Z-Transform

**Defn 16** (One-Sided Z-Transform). The *one-sided z-transform* is the same as the Z-Transform, but is only defined at n values greater than or equal to 0.

$$X(z) \equiv \sum_{n=0}^{\infty} x(n)z^{-n}$$
(4.4)

The One-Sided Z-Transform is generally used when there are initial conditions on a causal signal. This captures the normal causal portion of the signal, while also showing the effect of the initial condition.

### 4.2 The Inverse Z-Transform

This is the formal definition of The Inverse Z-Transform.

$$x(n) = \frac{1}{2\pi i} \oint_C X(z) z^{n-1} dz \tag{4.5}$$

where the integrals is a contour integral over a closed path C that encloses the origin and lies within the region of convergence of X(z).

There are 3 methods that are often used for the evaluation of the inverse z-transform in practice:

- 1. Direct evaluation of (4.5).
- 2. Expansion into a series of terms, in the variable sz and  $z^{-1}$ .
- 3. Partial-fraction expansion and table lookup.

#### 4.2.1 The Inverse Z-Transform by Contour Integration

**Defn 17** (Cauchy's Integral Theorem). Let f(z) be a function of the complex variable z and C be a closed path in the z-plane. If the derivative  $\frac{\mathrm{d}f(z)}{\mathrm{d}z}$  exists on and inside the contour C and if f(z) has no poles at  $z=z_0$ , then

$$\frac{1}{2\pi j} \oint_C \frac{f(z)}{z - z_0} dz = \begin{cases} f(z_0), & \text{if } z_0 \text{ is inside } C\\ 0, & \text{if } z_0 \text{ is outside } C \end{cases}$$

$$\tag{4.6}$$

More generally, if the (k+1)-order derivative of f(z) exists and f(z) has no poles at  $z=z_0$ , then

$$\frac{1}{2\pi j} \oint_C \frac{f(z)}{(z-z_0)^k} dz = \begin{cases} \frac{1}{(k-1)!} \frac{d^{k-1}f(z)}{dz^{k-1}} \Big|_{z=z_0}, & \text{if } z_0 \text{ is inside } C\\ 0, & \text{if } z_0 \text{ is outside } C \end{cases}$$
(4.7)

### 4.2.2 The Inverse Z-Transform by Power Series Expansion

### 4.2.3 The Inverse Z-Transform by Partial-Fraction Expansion

### 4.3 Properties of the Z-Transform

- (i) Z-Transform Linearity
- (ii) Z-Transform Time Shifting
- (iii) Z-Domain Scaling
- (iv) Z-Transform Time Reversal
- (v) Z-Domain Differentiation
- (vi) Z-Domain Convolutions
- (vii) Z-Transform 2 Sequence Correlation
- (viii) Z-Transform 2 Sequence Multiplication
  - (ix) Parsevals Relation for Z-Transform
  - ( $\mathbf{x}$ ) Initial Value Theorem for Z-Transform

Property	Time Domain	z-Domain	ROC
Notation	x(n)	X(z)	$ROC: r_2 <  z  < r_1$
	$x_1(n)$	$X_1(z)$	$ROC_1$
	$x_2(n)$	$X_2(z)$	$ROC_2$
Z-Transform Linearity	$a_1x_1(n) + a_2x_2(n)$	$a_1 X_1(z) + a_2 X_2(z)$	At least the intersection of $ROC_1$ and $ROC_2$
Z-Transform Time Shift-	x(n-k)	$z^{-k}X(z)$	That of $X(z)$ , except $z = 0$ if
ing			$k > 0$ and $z = \infty$ if $k < 0$
Z-Domain Scaling	$a^n x(n)$	$X(a^{-1}z)$	$ a r_2 <  z  <  a r_1$
$Z ext{-}\operatorname{Transform}$ Time Rever-	x(-n)	$X(z^{-1})$	$\frac{1}{r_1} <  z  < \frac{1}{r_2}$
sal			, 1
Conjugation	$x^*(n)$	$X^*(z^*)$	ROC
Real Part	$\operatorname{Re}\{x(n)\}$	$rac{1}{2} \left[ X(z) + X^*(z^*)  ight] \ rac{1}{2} j \left[ X(z) - X^*(z^*)  ight]$	Includes ROC
p Imaginary Part	$\operatorname{Im}\{x(n)\}$	$\frac{1}{2}j\left[X(z) - X^*(z^*)\right]$	Includes ROC
Z-Domain Differentiation	nx(n)	$-z\frac{dX(z)}{dz}$	$r_2 <  z r_1$
Z-Domain Convolutions	$x_1 * x_2$	$X_1(z) \overset{az}{X_2}(z)$	At least, the intersection of ROC <sub>1</sub> and ROC <sub>2</sub>
Z-Transform 2 Sequence Correlation	$r_{x_1x_2}(l) = x_1(l) * x_2(-l)$	$R_{x_1 x_2}(z) = X_1(z) x_2(z^{-1})$	At least, the intersection of ROC of $X_1(z)$ and $X_2(z^{-1})$
Initial Value Theorem for $Z$ -Transform	If $x(n)$ causal	$x(0) = \lim_{z \to \infty} X(z)$	- ( ) - ( )
Z-Transform 2 Sequence Multiplication	$x_1(n)x_2(n)$	$\frac{1}{2\pi j} \oint_C X_1(v) X_2(\frac{z}{v}) v^{-1} dv$	At least, $r_{1l}r_{2l} <  a  < r_{1u}r_{2u}$
Parsevals Relation for $Z$ -Transform	$\sum_{n=-\infty}^{\infty} x_1(n) x_2^*(n)$	$= \frac{1}{2\pi j} \oint_C X_1(v) X_2^*(\frac{1}{v^*}) v^{-1} dv$	

Table 4.2: Properties of the Z-Transform

#### 4.3.1 Z-Transform Linearity

If

$$x_1(n) \stackrel{\mathbf{z}}{\longleftrightarrow} X_1(z)$$

$$x_2(n) \stackrel{\mathbf{z}}{\longleftrightarrow} X_2(z)$$

then

$$x(n) = a_1 x_1(n) + a_2 x_2(n) \stackrel{z}{\longleftrightarrow} X(z) = a_1 X_1(z) + a_2 X_2(z)$$
(4.8)

for any constants  $a_1$  and  $a_2$ .

The linearity property can be generalized to an arbitrary number of signals.

### Example 4.2: Simple Z-Transform Linearity Problem. Example 3.2.1

Deermine the z-transform and the ROC of the signal

$$x(n) = [3(2^n) - 4(3^n)] \mathcal{U}(n)$$

Solution on Page 158.

### Example 4.3: Z-Transform Linearity on Trig Functions. Example 3.2.2

Determine the z-transform of the signals

- (a)  $x(n) = (\cos \omega_0 n) \mathcal{U}(n)$
- **(b)**  $x(n) = (\sin \omega_0 n) \mathcal{U}(n)$

Solution on Pages 158-159.

### 4.3.2 Z-Transform Time Shifting

If

$$x(n) \stackrel{\mathbf{z}}{\longleftrightarrow} X(z)$$

then

$$x(n-k) \stackrel{\simeq}{\longleftrightarrow} z^{-k} X(z) \tag{4.9}$$

The ROC of  $z^{-k}X(z)$  is the same as that of X(z) except for z=0 if k>0 and  $z=\infty$  if k<0.

### Example 4.4: Z-Transform Time Shifting. Example 3.2.3

By applying the time-shifting property, determine the z-transform of the signals

$$x_1(n) = \{1, 2, \underline{5}, 7, 0, 1\}$$
  
 $x_2(n) = \{\underline{0}, 0, 1, 2, 5, 7, 0, 1\}$ 

from the z-transform of

$$x_0(n) = \{\underline{1},2,5,7,0,1\}$$
 
$$X_0(z) = 1 + 2z^{-1} + 5z^{-2} + 7z^{-3} + z^{-5}, \text{ROC}: \text{entire $z$-plane except } z = 0$$

Solution on Page 160.

#### 4.3.3 Z-Domain Scaling

If

$$x(n) \stackrel{\mathbf{z}}{\longleftrightarrow} X(z), \text{ ROC} : r_1 < |z| < r_2$$

then

$$a^n x(n) \stackrel{\mathbf{z}}{\longleftrightarrow} X\left(a^{-1}z\right), \text{ ROC}: |a|r_1 < |z| < |a|r_2$$
 (4.10)

#### 4.3.4 Z-Transform Time Reversal

If

$$x(n) \stackrel{\mathbf{z}}{\longleftrightarrow} X(z)$$
, ROC:  $r_1 < |z| < r_2$ 

then

$$x(-n) \stackrel{\mathbf{z}}{\longleftrightarrow} X(z^{-1}), \text{ROC} : \frac{1}{r_2} < |z| < \frac{1}{r_1}$$
 (4.11)

### Example 4.5: Z-Transform Time Reversal. Example 3.2.6

Determine the z-transform of the signal

$$x(n) = \mathcal{U}(-n)$$

The transform for  $\mathcal{U}(n)$  is given in Table 4.3.

$$\mathcal{U}(n) \stackrel{\mathbf{z}}{\longleftrightarrow} \frac{1}{1 - x^{-1}}, \text{ ROC} : |z| > 1$$

By using (4.11), we obtain

$$\mathcal{U}(-n) \stackrel{\mathbf{z}}{\longleftrightarrow} \frac{1}{1-z}, \text{ROC} : |z| < 1$$

#### 4.3.5 Z-Domain Differentiation

If

$$x(n) \stackrel{\mathbf{z}}{\longleftrightarrow} X(z)$$

then

$$nx(n) \stackrel{\mathbf{z}}{\longleftrightarrow} -z \frac{dX(z)}{dz}$$
 (4.12)

#### Example 4.6: Z-Domain Differentiation. Example 3.2.7

Determine the z-transform of the signal

$$x(n) = na^n \mathcal{U}(n)$$

The signal x(n) can be expressed as  $nx_1(n)$ , where  $x_1(n) = a^n \mathcal{U}(n)$ . By passing this through the z-transform, we have

$$x_1(n) = a^n \mathcal{U}(n) \stackrel{\mathbf{z}}{\longleftrightarrow} X_1(z) = \frac{1}{1 - az^{-1}}, \text{ ROC} : |z| > |a|$$

Then by using (4.12), we obtain

$$na^n \mathcal{U}(n) \stackrel{\mathbf{z}}{\longleftrightarrow} X(z) = -z \frac{dX_1(z)}{dz} = \frac{az^{-1}}{(1 - az^{-1})^2}$$

### 4.3.6 Z-Domain Convolutions

If

$$x_1(n) \stackrel{\mathbf{z}}{\longleftrightarrow} X_1(z)$$
  
 $x_2(n) \stackrel{\mathbf{z}}{\longleftrightarrow} X_2(z)$ 

then

$$x(n) = x_1(n) * x_2(n) \stackrel{\mathbf{z}}{\longleftrightarrow} X(z) = X_1(z)X_2(z)$$

$$\tag{4.13}$$

The ROC of X(z) is, at least, the intersection of that for  $X_1(z)$  and  $X_2(z)$ .

### Example 4.7: Z-Domain Convolutions. Example 3.2.9

Compute the convolution x(n) of the signals

$$x_1(n) = \{1, -2, 1\}$$
  
 $x_2(n) = \begin{cases} 1, & 0 \le n \le 6 \\ 0, & \text{elsewhere} \end{cases}$ 

When

$$X_1(z) = 1 - 2z^{-1} + z^{-2}$$
  
 $X_2(z) = 1 + z^{-1} + z^{-2} + z^{-3} + z^{-4} + z^{-5}$ 

According to (4.13) we carry out the multiplication of  $X_1(z)$  and  $X_2(z)$ . Thus

$$X(z) = X_1(z)X_2(z) = 1 - z^{-1} - z^{-6} + z^{-7}$$

Hence

$$x(n) = \{\underline{1}, -1, 0, 0, 0, 0, -1, 1\}$$

#### 4.3.7 Z-Transform 2 Sequence Correlation

If

$$x_1(n) \stackrel{\mathbf{z}}{\longleftrightarrow} X_1(z)$$
  
 $x_2(n) \stackrel{\mathbf{z}}{\longleftrightarrow} X_2(z)$ 

then

$$r_{x_1x_2}(l) = \sum_{n = -\infty}^{\infty} x_1(n)x_2(n - l) \stackrel{\mathbf{z}}{\longleftrightarrow} R_{x_1x_2}(z) = X_1(z)X_2(z^{-1})$$
(4.14)

### Example 4.8: Z-Transform 2 Sequence Correlation. Example 3.2.10

Determine the autocorrelation of the signal

$$x(n) = a^n \mathcal{U}(n), -1 < a < 1$$

Solution on Page 166.

#### 4.3.8 Z-Transform 2 Sequence Multiplication

If

$$x_1(n) \stackrel{\mathbf{z}}{\longleftrightarrow} X_1(z)$$
  
 $x_2(n) \stackrel{\mathbf{z}}{\longleftrightarrow} X_2(z)$ 

then

$$x(n) = x_1(n)x_2(n) \stackrel{\mathbf{z}}{\longleftrightarrow} X_z = \frac{1}{2\pi i} \oint_C X_1(v)X_2\left(\frac{z}{v}\right) v^{-1} dv \tag{4.15}$$

where C is a closed contour that encloses the origin and lies within the region of convergence common to both  $X_1(v)$  and  $X_2(\frac{1}{v})$ .

#### 4.3.9 Parsevals Relation for Z-Transform

If  $x_1(n)$  and  $x_2(n)$  are complex-valued sequences, then

$$\sum_{n=-\infty}^{\infty} x_1(n) x_2^*(n) = \frac{1}{2\pi j} \oint_C X_1(v) X_2^* \left(\frac{1}{v^*}\right) v^{-1} dv$$
(4.16)

### 4.3.10 Initial Value Theorem for Z-Transform

If 
$$x(n)$$
 is causal [i.e.,  $x(n) = 0$  for  $n < 0$ ], then 
$$x(0) = \lim_{z \to \infty} X(z) \tag{4.17}$$

### 4.4 Properties of the One-Sided Z-Transform

(i)

### 4.5 Rational Z-Transforms

An important family of z-transforms are those for which X(z) is a rational function, a ratio of two polynomials in  $z^{-1}$  (or z).

### 4.5.1 Poles and Zeros of a Z-Transform

### 4.5.2 Decomposition of Rational Z-Transforms

### 4.6 Analysis of LTI Systems in the Z-Domain

### 4.7 Common Z-Transforms

Signal, $x(n)$	z-Transform, $X(z)$	ROC
$\delta(n)$	1	All $z$
$\mathcal{U}(n)$	$\frac{1}{1-z^{-1}}$	z  > 1
$a^n \mathcal{U}(n)$	$\frac{1}{1-az^{-1}}$	z  >  a
$na^n \mathcal{U}(n)$	$\frac{az^{-1}}{(1-az^{-1})^2}$	z  >  a
$-a^n \mathcal{U}(-n-1)$	$\frac{1}{1-az^{-1}}$	z  <  a
$-na^n \mathcal{U}(-n-1)$	$\frac{az^{-z}}{(1-az^{-1})^2}$	z  <  a
$(\cos \omega_0 n) \mathcal{U}(n)$	$\frac{1 - z^{-1}\cos\omega_0}{1 - 2z^{-1}\cos\omega_0 + z^{-2}}$	z  > 1
$(\sin \omega_0 n) \mathcal{U}(n)$	$\frac{z^{-1}\sin\omega_0}{1 - 2z^{-1}\cos\omega_0 + z^{-2}}$	z  > 1
$(a^n\cos\omega_0 n)\mathcal{U}(n)$	$\frac{1 - az^{-1}\cos\omega_0}{1 - 2az^{-1}\cos\omega_0 + a^2z^{-2}}$	z  >  a
$(a^n \sin \omega_0 n) \mathcal{U}(n)$	$\frac{az^{-1}\sin\omega_0}{1 - 2az^{-1}\cos\omega_0 + a^2z^{-2}}$	z  >  a

Table 4.3: Common Z-Transforms

### A Trigonometry

### A.1 Trigonometric Formulas

$$\sin(\alpha) + \sin(\beta) = 2\sin\left(\frac{\alpha+\beta}{2}\right)\cos\left(\frac{\alpha-\beta}{2}\right)$$
 (A.1)

$$\cos(\theta)\sin(\theta) = \frac{1}{2}\sin(2\theta) \tag{A.2}$$

### A.2 Euler Equivalents of Trigonometric Functions

$$e^{\pm i\alpha} = \cos(\alpha) \pm i\sin(\alpha)$$
 (A.3)

$$\sin\left(x\right) = \frac{e^{ix} + e^{-ix}}{2} \tag{A.4}$$

$$\cos\left(x\right) = \frac{e^{\imath x} - e^{-\imath x}}{2\imath} \tag{A.5}$$

$$\sinh(x) = \frac{e^x - e^{-x}}{2}$$
 (A.6)

$$\cosh\left(x\right) = \frac{e^x + e^{-x}}{2} \tag{A.7}$$

### A.3 Angle Sum and Difference Identities

$$\sin(\alpha \pm \beta) = \sin(\alpha)\cos(\beta) \pm \cos(\alpha)\sin(\beta) \tag{A.8}$$

$$\cos(\alpha \pm \beta) = \cos(\alpha)\cos(\beta) \mp \sin(\alpha)\sin(\beta) \tag{A.9}$$

### A.4 Double-Angle Formulae

$$\sin(2\alpha) = 2\sin(\alpha)\cos(\alpha) \tag{A.10}$$

$$\cos(2\alpha) = \cos^2(\alpha) - \sin^2(\alpha) \tag{A.11}$$

### A.5 Half-Angle Formulae

$$\sin\left(\frac{\alpha}{2}\right) = \sqrt{\frac{1 - \cos\left(\alpha\right)}{2}}\tag{A.12}$$

$$\cos\left(\frac{\alpha}{2}\right) = \sqrt{\frac{1 + \cos\left(\alpha\right)}{2}}\tag{A.13}$$

### A.6 Exponent Reduction Formulae

$$\sin^2(\alpha) = \frac{1 - \cos(2\alpha)}{2} \tag{A.14}$$

$$\cos^2(\alpha) = \frac{1 + \cos(2\alpha)}{2} \tag{A.15}$$

### A.7 Product-to-Sum Identities

$$2\cos(\alpha)\cos(\beta) = \cos(\alpha - \beta) + \cos(\alpha + \beta) \tag{A.16}$$

$$2\sin(\alpha)\sin(\beta) = \cos(\alpha - \beta) - \cos(\alpha + \beta) \tag{A.17}$$

$$2\sin(\alpha)\cos(\beta) = \sin(\alpha + \beta) + \sin(\alpha - \beta) \tag{A.18}$$

$$2\cos(\alpha)\sin(\beta) = \sin(\alpha + \beta) - \sin(\alpha - \beta) \tag{A.19}$$

### A.8 Sum-to-Product Identities

$$\sin(\alpha) \pm \sin(\beta) = 2\sin\left(\frac{\alpha \pm \beta}{2}\right)\cos\left(\frac{\alpha \mp \beta}{2}\right)$$
 (A.20)

$$\cos(\alpha) + \cos(\alpha) = 2\cos\left(\frac{\alpha + \beta}{2}\right)\cos\left(\frac{\alpha - \beta}{2}\right) \tag{A.21}$$

$$\cos(\alpha) - \cos(\beta) = -2\sin\left(\frac{\alpha + \beta}{2}\right)\sin\left(\frac{\alpha - \beta}{2}\right)$$
(A.22)

### A.9 Pythagorean Theorem for Trig

$$\cos^2(\alpha) + \sin^2(\alpha) = 1^2 \tag{A.23}$$

### A.10 Rectangular to Polar

$$a + ib = \sqrt{a^2 + b^2}e^{i\theta} = re^{i\theta} \tag{A.24}$$

$$\theta = \begin{cases} \arctan\left(\frac{b}{a}\right) & a > 0\\ \pi - \arctan\left(\frac{b}{a}\right) & a < 0 \end{cases}$$
(A.25)

### A.11 Polar to Rectangular

$$re^{i\theta} = r\cos(\theta) + ir\sin(\theta) \tag{A.26}$$

### B Calculus

### **B.1** Fundamental Theorems of Calculus

**Defn B.1.1** (First Fundamental Theorem of Calculus). The first fundamental theorem of calculus states that, if f is continuous on the closed interval [a, b] and F is the indefinite integral of f on [a, b], then

$$\int_{a}^{b} f(x) dx = F(b) - F(a)$$
(B.1)

**Defn B.1.2** (Second Fundamental Theorem of Calculus). The second fundamental theorem of calculus holds for f a continuous function on an open interval I and a any point in I, and states that if F is defined by

$$F(x) = \int_{a}^{x} f(t) dt,$$

then

$$\frac{d}{dx} \int_{a}^{x} f(t) dt = f(x)$$

$$F'(x) = f(x)$$
(B.2)

**Defn B.1.3** (argmax). The arguments to the *argmax* function are to be maximized by using their derivatives. You must take the derivative of the function, find critical points, then determine if that critical point is a global maxima. This is denoted as

 $\operatorname*{argmax}_{r}$ 

#### B.2 Rules of Calculus

#### B.2.1 Chain Rule

**Defn B.2.1** (Chain Rule). The *chain rule* is a way to differentiate a function that has 2 functions multiplied together.

$$f(x) = g(x) \cdot h(x)$$

then,

$$f'(x) = g'(x) \cdot h(x) + g(x) \cdot h'(x)$$

$$\frac{df(x)}{dx} = \frac{dg(x)}{dx} \cdot g(x) + g(x) \cdot \frac{dh(x)}{dx}$$
(B.3)

## C Laplace Transform

 $\bf Defn~C.0.1$  (Laplace Transform). The  $\it laplace~transformation$  is ...