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

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Contents

3 Convolutions

		soids	1
	1.1	Continuous-Time Signals	1
		1.1.1 Frequency in Continuous-Time Signals	1
		1.1.2 Properties of Continuous-Time Sinusoidal Signals	1
	1.2	Discrete-Time Signals	2
		1.2.1 Frequency in Discrete-Time Signals	2
		1.2.2 Properties of Discrete-Time Sinusoidal Signals	2
		1.2.3 Frequency Aliases	2
	1.3	Sampling Rates and Sampling Frequency	2
		1.3.1 Nyquist Rate	3
		1.3.2 Nyquist Frequency	3
	1.4	Digital Signals	3
		1.4.1 Quantization	3
		1.4.1.1 Quantization Levels	3
		1.4.1.2 Quantization Error	3
		1.4.1.3 Bit Requirements	3
		1.4.1.4 Bit Rate	3
		1.4.1.4 Dit 1(a)(c	J
2	Disc	rete-Time Systems	3
	2.1	Representing Discrete-Time Systems	3
		2.1.1 Functional Representation	3
		2.1.2 Tabular Representation	3
		2.1.3 Sequence Representation	3
	2.2		
	Z.Z	Elementary Discrete-Time Signals	
	2.2	Elementary Discrete-Time Signals	4
	2.2	2.2.1 Unit Impulse Signal	4
	2.2	2.2.1 Unit Impulse Signal	4 4 4
	2.2	2.2.1 Unit Impulse Signal	4 4 4
		2.2.1 Unit Impulse Signal2.2.2 Unit Step Signal2.2.3 Unit Ramp Signal2.2.4 Exponential Signal	4 4 4 4
	2.3	2.2.1 Unit Impulse Signal	4 4 4 4 4
		2.2.1 Unit Impulse Signal 2.2.2 Unit Step Signal 2.2.3 Unit Ramp Signal 2.2.4 Exponential Signal Classification of Discrete-Time Signals 2.3.1 Energy Signal	4 4 4 4 4 4 5
		2.2.1 Unit Impulse Signal 2.2.2 Unit Step Signal 2.2.3 Unit Ramp Signal 2.2.4 Exponential Signal Classification of Discrete-Time Signals 2.3.1 Energy Signal 2.3.2 Power Signal	4 4 4 4 4 5 5
		2.2.1 Unit Impulse Signal 2.2.2 Unit Step Signal 2.2.3 Unit Ramp Signal 2.2.4 Exponential Signal Classification of Discrete-Time Signals 2.3.1 Energy Signal 2.3.2 Power Signal 2.3.3 Periodic and Aperiodic Signals	4 4 4 4 4 5 5 5
	2.3	2.2.1 Unit Impulse Signal 2.2.2 Unit Step Signal 2.2.3 Unit Ramp Signal 2.2.4 Exponential Signal Classification of Discrete-Time Signals 2.3.1 Energy Signal 2.3.2 Power Signal 2.3.3 Periodic and Aperiodic Signals 2.3.4 Symmetric and Antisymmetric Signals	4 4 4 4 4 4 5 5 5
		2.2.1 Unit Impulse Signal 2.2.2 Unit Step Signal 2.2.3 Unit Ramp Signal 2.2.4 Exponential Signal Classification of Discrete-Time Signals 2.3.1 Energy Signal 2.3.2 Power Signal 2.3.3 Periodic and Aperiodic Signals 2.3.4 Symmetric and Antisymmetric Signals Discrete-Time Signal Manipulations	4 4 4 4 4 5 5 5 5 5
	2.3	2.2.1 Unit Impulse Signal 2.2.2 Unit Step Signal 2.2.3 Unit Ramp Signal 2.2.4 Exponential Signal Classification of Discrete-Time Signals 2.3.1 Energy Signal 2.3.2 Power Signal 2.3.3 Periodic and Aperiodic Signals 2.3.4 Symmetric and Antisymmetric Signals Discrete-Time Signal Manipulations 2.4.1 Transformation of the Independent Variable (Time)	4 4 4 4 4 4 5 5 5
	2.3	2.2.1 Unit Impulse Signal 2.2.2 Unit Step Signal 2.2.3 Unit Ramp Signal 2.2.4 Exponential Signal 2.3.1 Energy Signal 2.3.2 Power Signal 2.3.3 Periodic and Aperiodic Signals 2.3.4 Symmetric and Antisymmetric Signals Discrete-Time Signal Manipulations 2.4.1 Transformation of the Independent Variable (Time) 2.4.1.1 Shifting in Time	4 4 4 4 4 5 5 5 5 5
	2.3	2.2.1 Unit Impulse Signal 2.2.2 Unit Step Signal 2.2.3 Unit Ramp Signal 2.2.4 Exponential Signal Classification of Discrete-Time Signals 2.3.1 Energy Signal 2.3.2 Power Signal 2.3.3 Periodic and Aperiodic Signals 2.3.4 Symmetric and Antisymmetric Signals Discrete-Time Signal Manipulations 2.4.1 Transformation of the Independent Variable (Time) 2.4.1.1 Shifting in Time 2.4.1.2 Folding	4 4 4 4 4 5 5 5 5 5 6
	2.3	2.2.1 Unit Impulse Signal 2.2.2 Unit Step Signal 2.2.3 Unit Ramp Signal 2.2.4 Exponential Signal 2.2.4 Exponential Signal Classification of Discrete-Time Signals 2.3.1 Energy Signal 2.3.2 Power Signal 2.3.3 Periodic and Aperiodic Signals 2.3.4 Symmetric and Antisymmetric Signals Discrete-Time Signal Manipulations 2.4.1 Transformation of the Independent Variable (Time) 2.4.1.1 Shifting in Time 2.4.1.2 Folding 2.4.2 Addition, Multiplication, and Scaling	4 4 4 4 4 5 5 5 5 5 6 6
	2.3	2.2.1 Unit Impulse Signal 2.2.2 Unit Step Signal 2.2.3 Unit Ramp Signal 2.2.4 Exponential Signal 2.2.5 Exponential Signal 2.2.6 Exponential Signal 2.2.7 Energy Signal 2.2.8 Power Signal 2.3.9 Periodic and Aperiodic Signals 2.3.1 Symmetric and Antisymmetric Signals 2.3.2 Power Signal 2.3.3 Periodic and Aperiodic Signals 2.3.4 Symmetric and Antisymmetric Signals 2.3.4 Symmetric and Antisymmetric Signals 2.3.5 Discrete-Time Signal Manipulations 2.4.1 Transformation of the Independent Variable (Time) 2.4.1.1 Shifting in Time 2.4.1.2 Folding 2.4.2 Addition, Multiplication, and Scaling 2.4.2.1 Addition	44 44 44 45 55 55 55 56 66 66
	2.3	2.2.1 Unit Impulse Signal 2.2.2 Unit Step Signal 2.2.3 Unit Ramp Signal 2.2.4 Exponential Signal 2.2.4 Exponential Signal Classification of Discrete-Time Signals 2.3.1 Energy Signal 2.3.2 Power Signal 2.3.3 Periodic and Aperiodic Signals 2.3.4 Symmetric and Antisymmetric Signals Discrete-Time Signal Manipulations 2.4.1 Transformation of the Independent Variable (Time) 2.4.1.1 Shifting in Time 2.4.1.2 Folding 2.4.2 Addition, Multiplication, and Scaling	4 4 4 4 4 5 5 5 5 5 6 6

6

4	The Z-Transform	6
	4.1 The Z-Transform	6
	4.1.1 The One-Sided Z -Transform	6
	4.2 The Inverse Z -Transform	7
	4.2.1 The Inverse Z-Transform by Contour Integration	7
	4.2.2 The Inverse Z-Transform by Power Series Expansion	7
	4.2.3 The Inverse Z-Transform by Partial-Fraction Expansion	7
	4.3 Properties of the Z-Transform	7
	4.3.1 Properties of the One-Sided Z-Transform	7
	4.4 Rational Z-Transforms	7
	4.4.1 Decomposition of Rational Z-Transforms	7
	4.5 Analysis of LTI Systems in the Z-Domain	7
\mathbf{A}	Trigonometry	8
	A.1 Trigonometric Formulas	8
	A.2 Euler Equivalents of Trigonometric Functions	8
	A.3 Angle Sum and Difference Identities	8
	A.4 Double-Angle Formulae	8
	A.5 Half-Angle Formulae	8
	A.6 Exponent Reduction Formulae	8
	A.7 Product-to-Sum Identities	8
	A.8 Sum-to-Product Identities	9
	A.9 Pythagorean Theorem for Trig	9
	A.10 Rectangular to Polar	9
	A.11 Polar to Rectangular	9
\mathbf{B}	Calculus	10
	B.1 Fundamental Theorems of Calculus	10
	B.2 Rules of Calculus	10
	B.2.1 Chain Rule	10
\mathbf{C}	Laplace Transform	11

11

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 ∞ . 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. Ω , the frequency in radians per second (rad/s)
- 3. θ , the *phase* in radians.

Instead of Ω , 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. ω , the angular frequency in radians per sample
- 4. θ , is the *phase*, in radians.

Instead of ω , 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 2π 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 2π 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 Ω 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) and is defined as

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

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

4.1.1 The One-Sided Z-Transform

Defn 15 (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.

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

$$\tag{4.2}$$

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.

$$x(n) = \frac{1}{2\pi j} \oint_C X(z)z^{n-1} dz \tag{4.3}$$

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 (4.3).
- 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 16 (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

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

$$\tag{4.4}$$

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

$$\frac{1}{2\pi j} \oint_C \frac{f(z)}{(z-z_0)^k} dz = \begin{cases} \frac{1}{(k-1)!} \frac{d^{k-1}f(z)}{dz^{k-1}} \Big|_{z=z_0}, & \text{if } z_0 \text{ is inside } C\\ 0, & \text{if } z_0 \text{ is outside } C \end{cases}$$
(4.5)

- 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)

4.3.1 Properties of the One-Sided Z-Transform

(i)

- 4.4 Rational Z-Transforms
- 4.4.1 Decomposition of Rational Z-Transforms
- 4.5 Analysis of LTI Systems in the Z-Domain

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)

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

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

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