

EE360: SIGNALS AND SYSTEMS I

CH1: SIGNALS AND SYSTEMS

CONTINUOUS-TIME AND DISCRETE-TIME SIGNALS

CHAPTER 1.0-1.1

INTRODUCTION

- Signals are **quantitative** descriptions of physical phenomena
 - Represent a pattern of variation

EXAMPLE SIGNALS I

- Circuit
- v_s - voltage signal
- v_c - voltage signal
- i - current signal
- These are continuous-time signals

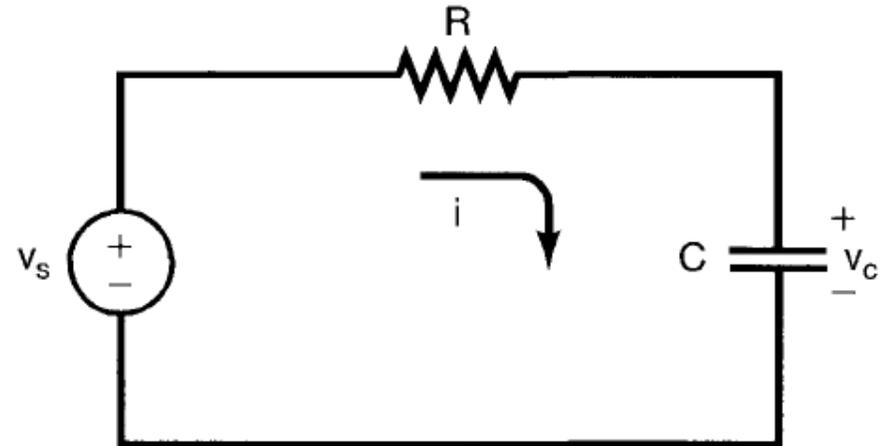


Figure 1.1 A simple RC circuit with source voltage v_s and capacitor voltage v_c .

EXAMPLE SIGNALS II

- Stock market price
- p – closing price signal
- Discrete time signal



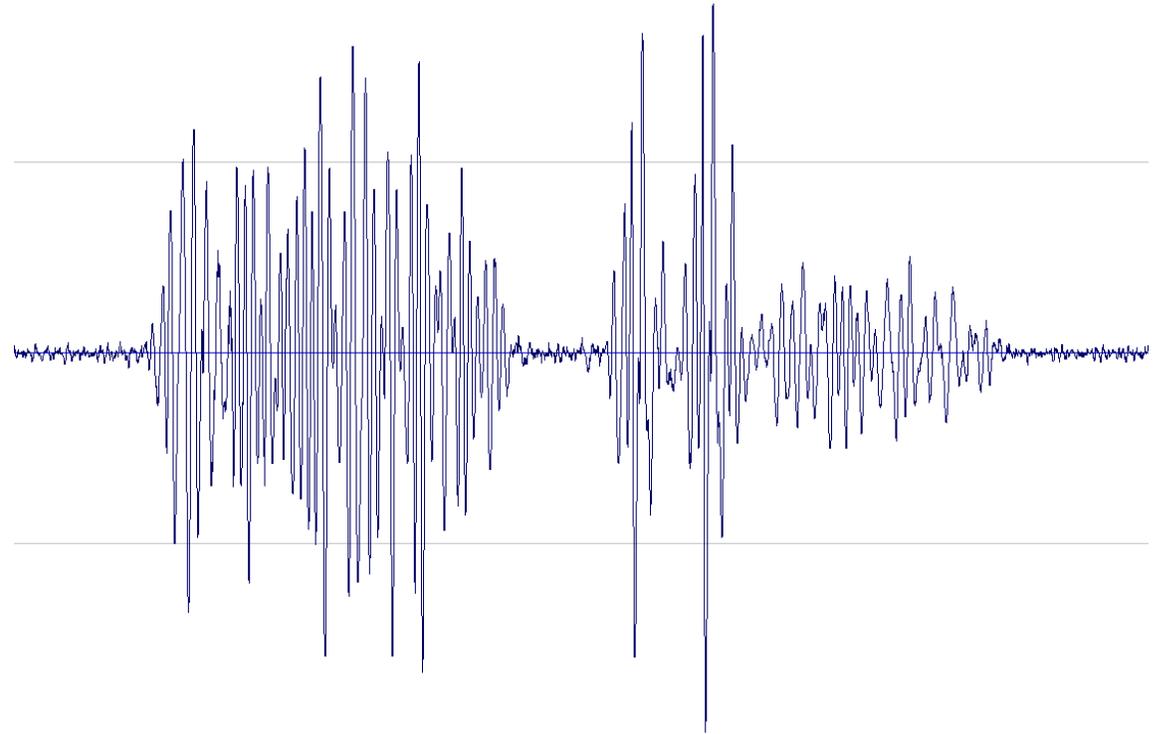
EXAMPLE SIGNALS II

- Stock market price
- p – closing price signal
- Discrete time signal
- Tesla stock for fun
 - Last 3 months
 - Last 5 years



EXAMPLE SIGNALS III

- Audio signal
- Continuous signal in “raw” form
- Discrete signal when store on a CD/computer



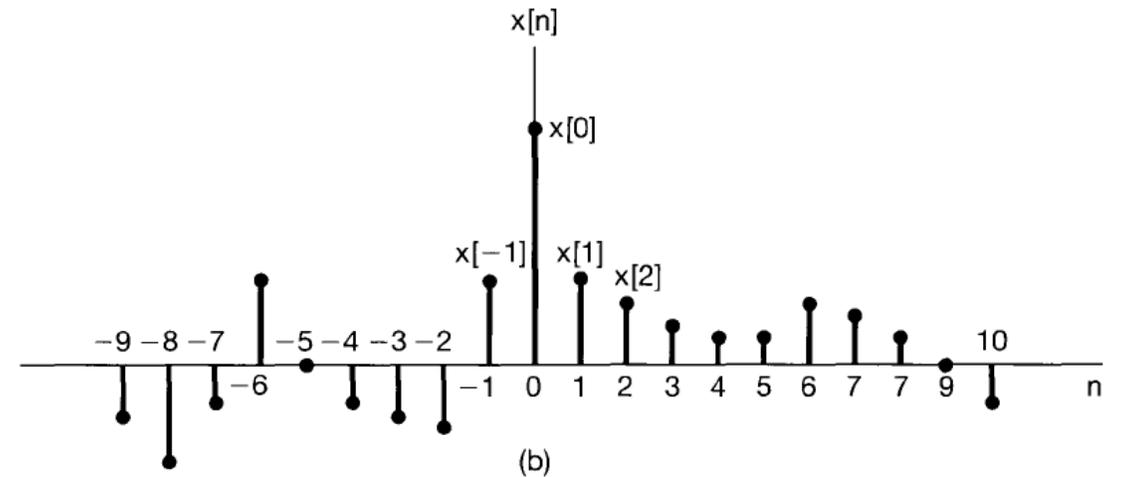
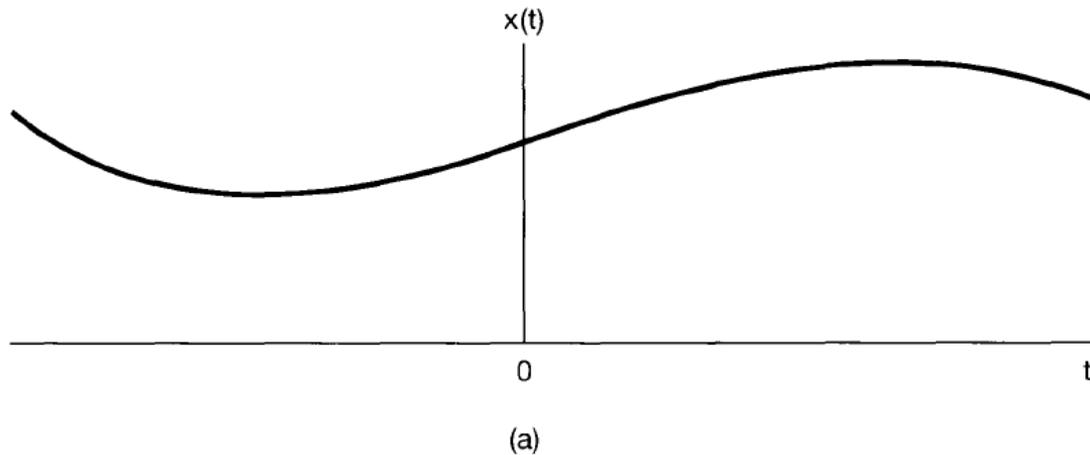
MATHEMATICAL FORMULATION

- In these examples, the signal is a function of one variable, **time**
 - $f(t)$ ← focus of the book
- More generally, a signal can be a function of multiple variables and not just time
 - E.g. an image $I(x, y)$

SIGNAL TYPES

- This course deals with two types of signals
- Continuous-time (CT) signals
 - $x(t)$ with $t \in \mathbb{R}$ a real-values variable, denoting continuous time
 - Notice the parenthesis is used to denote a CT signal
- Discrete-time (DT) signals
 - $x[n]$ with $n \in \mathbb{Z}$ an integer-valued variable, denoting discrete time
 - Notice the square brackets to denote a DT signal
 - $x[1]$ is defined but $x[1.5]$ is not defined

GRAPHICALLY



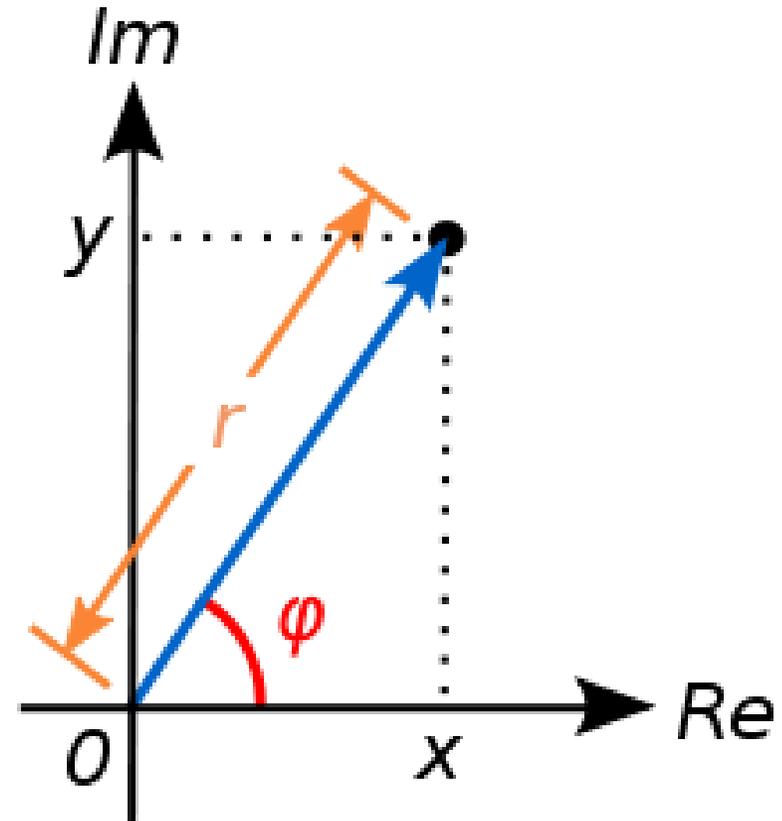
- Note: $x(t)$ could signify the full signal or a value of the signal at a specific time t
 - May see $x(t_0)$ for a specific value of signal $x(t)$ when $t = t_0$ for clarity

COMPLEX NUMBER REVIEW

- This course will often work with complex signals as they are mathematically convenient
 - $x(t) \in \mathbb{C}, \quad x[n] \in \mathbb{C}$
 - $\mathbb{C} = \{z | z = x + jy, \quad x, y \in \mathbb{R}, \quad j = \sqrt{-1}\}$
- Note the use of j for the imaginary number in electrical engineering rather than i

COMPLEX NUMBER REPRESENTATION

- Rectangular/Cartesian form
 - $z = x + jy$
 - $Re\{z\} = x$ real-part
 - $Im\{z\} = y$ imaginary-part
- Polar form
 - $z = r e^{j\theta}$
 - $r^2 = x^2 + y^2$
 - $\theta = \arctan\left(\frac{y}{x}\right)$
 - $x = r \cos \theta$
 - $y = r \sin \theta$



EULER'S FORMULA

- $e^{j\theta} = \cos \theta + j \sin \theta$

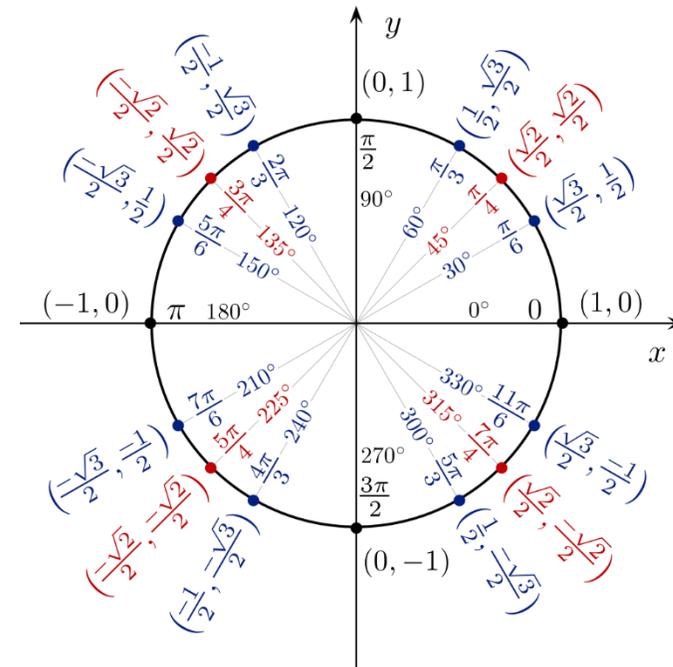
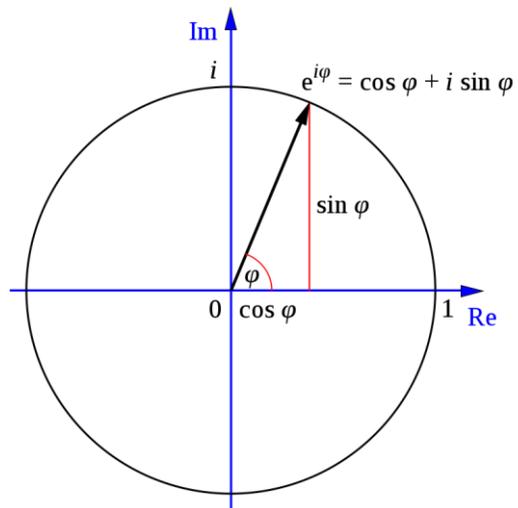
- Note:

- $j = e^{j\pi/2}$ $-1 = e^{j\pi}$

- $-j = e^{j3\pi/2}$ $1 = e^{j2\pi k}$

- Know trig functions for common angles

- For inverse trig function you must account for the quadrant



EXAMPLES: COMPLEX NUMBERS

- Express in polar form

- $1 - j$

- Express in polar form

- $(1 - j)^2$

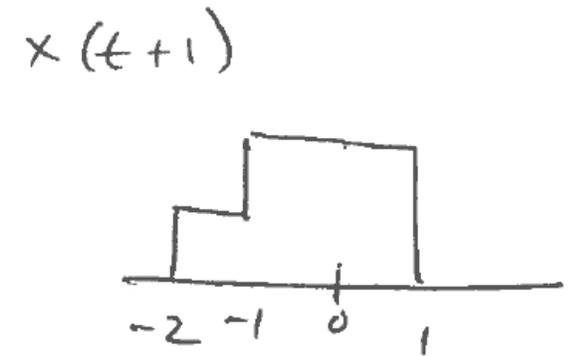
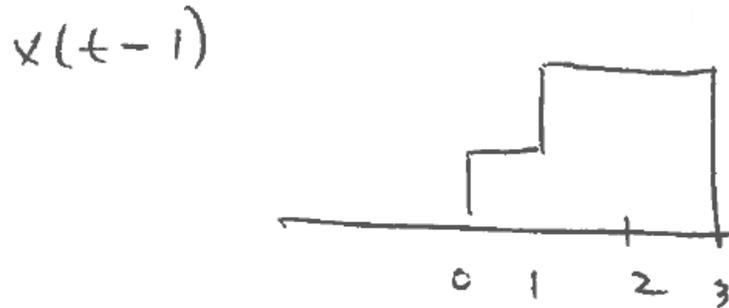
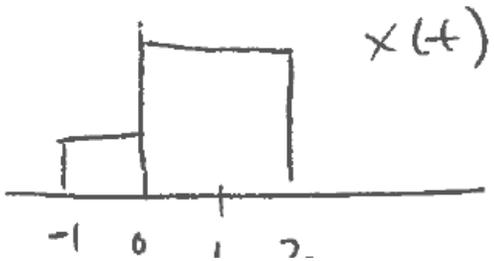
TRANSFORMATIONS OF THE INDEPENDENT VARIABLE

CHAPTER 1.2

TIME SHIFT

- $x(t) \rightarrow x(t - t_0)$ $x[n] \rightarrow x[n - n_0]$
- $t_0 > 0 \Rightarrow$ delay $t_0 < 0 \Rightarrow$ advance

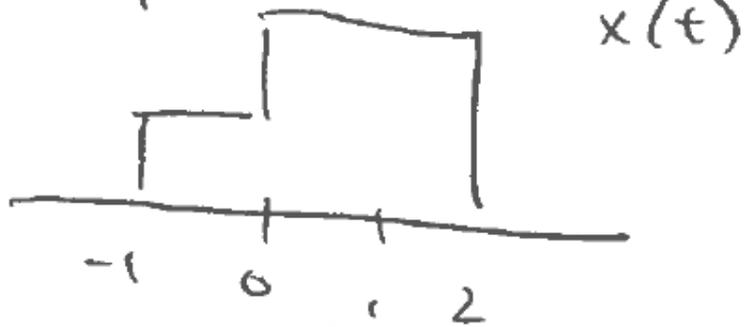
example



TIME REVERSAL

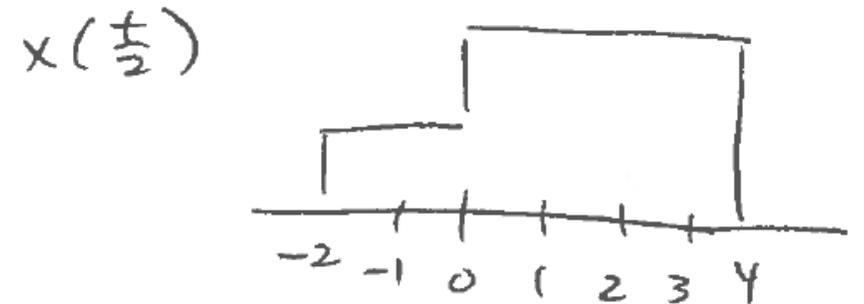
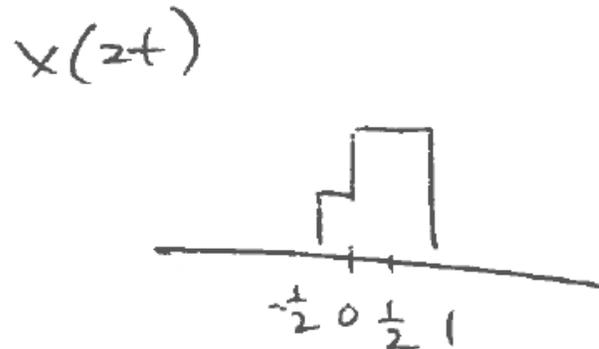
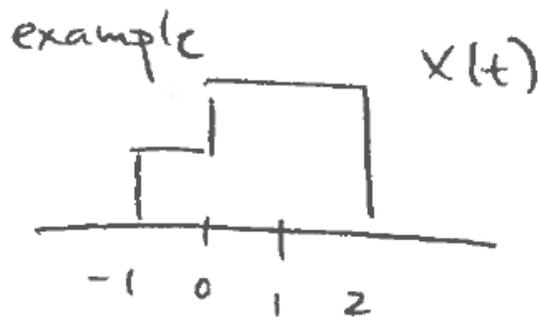
- $x(t) \rightarrow x(-t)$ $x[n] \rightarrow x[-n]$
 - Flip signal across y-axis ($t = 0$ axis)

example



TIME SCALING

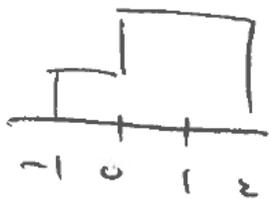
- $x(t) \rightarrow x(at) \quad a > 0$
 - $a > 1 \Rightarrow$ shrink time scale (“speed-up” or compress)
 - $0 < a < 1 \Rightarrow$ expand time scale (“slow-down” or stretch)
- $x[n] \rightarrow x[an] \quad a \in \mathbb{Z}^+$



GENERAL TRANSFORMATION

- $x(t) \rightarrow x(\alpha t - \beta)$ $\alpha < 0$ for time reversal
- General methodology – shift, then scale
 1. Shift: define $v(t) = x(t - \beta)$
 2. Scale: define $y(t) = v(\alpha t) = x(\alpha t - \beta)$
- Notice: scaling is only applied to time variable t

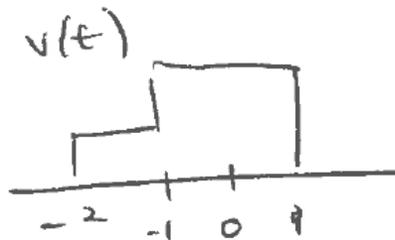
example $x(t)$



sketch $x(1-t)$

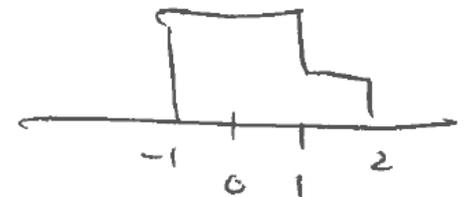
$$v(t) = x(t - (-1))$$

shift left 1 unit



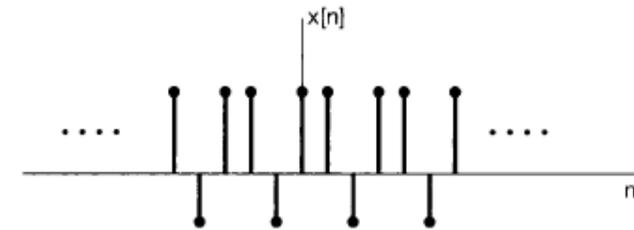
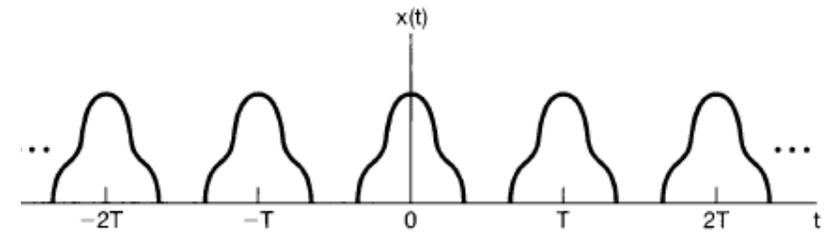
$$\begin{aligned} y(t) &= v(-t) \\ &= x(-t - (-1)) \\ &= x(1-t) \end{aligned}$$

flip across
 $t=0$



PERIODIC SIGNALS

- A signal is periodic if a shift of the signals leaves it unchanged
- Periodicity constraint
 - CT: there exists a $T > 0$ s.t.
 - $x(t) = x(t + T) \quad \forall t \in \mathbb{R}$
 - DT: there exists a $N > 0$ s.t.
 - $x[n] = x[n + N] \quad \forall n \in \mathbb{Z}$



FUNDAMENTAL PERIOD/FREQUENCY

- Note: $x(t) = x(t + T) = x(t + 2T) = x(t + 3T) = \dots$
 - Periodic with period T or kT
- Fundamental period
 - T_0 is the fundamental period of $x(t)$ if it is the smallest value of $T > 0$ to satisfy the periodicity constraint (N_0 for DT)
- Fundamental frequency – inverse relationship to time
 - $\omega_0 = \frac{2\pi}{T_0}$ occasionally, $\Omega_0 = \frac{2\pi}{N_0}$
- Aperiodic signal – signal with no T, N satisfying periodicity constraint

EXAMPLES: FIND PERIOD

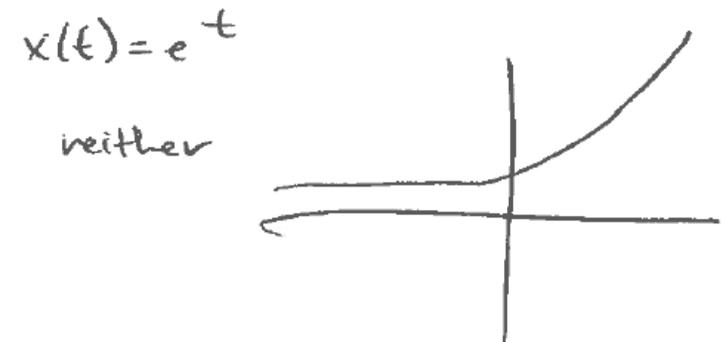
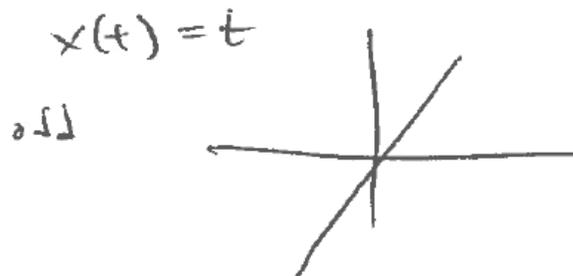
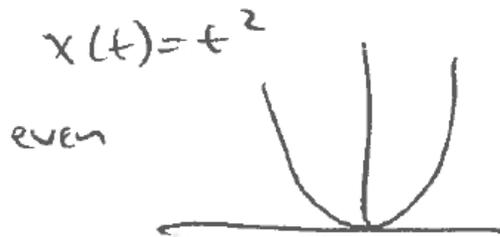
- $x(t) = e^{j\pi t/5}$

- $x[n] = e^{j\pi n/5}$

EVEN/ODD SIGNALS

- Even signal – same flipped across y-axis
 - $x[-n] = x[n]$
- Odd signal – upside-down when flipped
 - $x[-n] = -x[n]$
 - Note: must have $x[n] = 0$ at $n = 0$
- Decomposition theorem – any signal can be broken into sum of even and odd signals
 - $x(t) = y(t) + z(t)$, $y(t)$ even, $z(t)$ odd
 - $y(t) = \text{Ev}\{x(t)\} = \frac{1}{2}[x(t) + x(-t)]$
 - $z(t) = \text{Odd}\{x(t)\} = \frac{1}{2}[x(t) - x(-t)]$

examples



EXPONENTIAL AND SINUSOIDAL SIGNALS

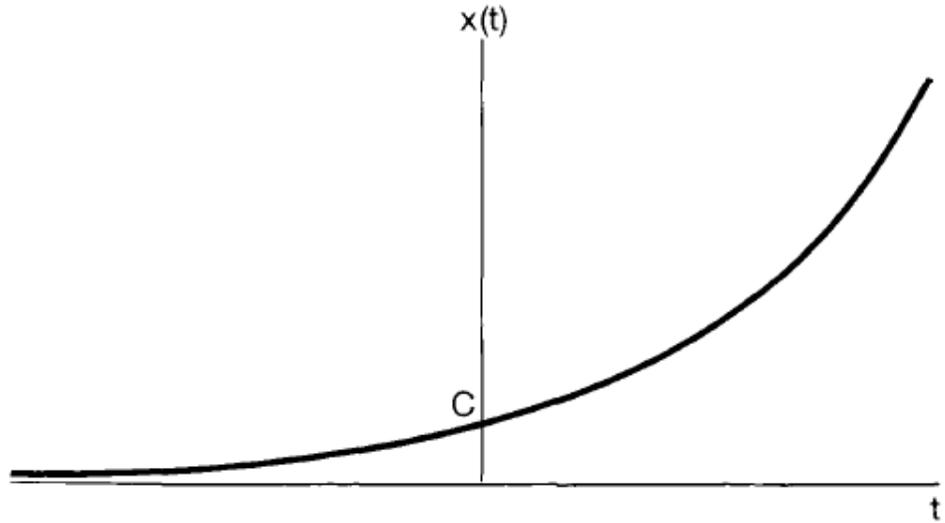
CHAPTER 1.3

IMPORTANT CLASSES OF SIGNALS

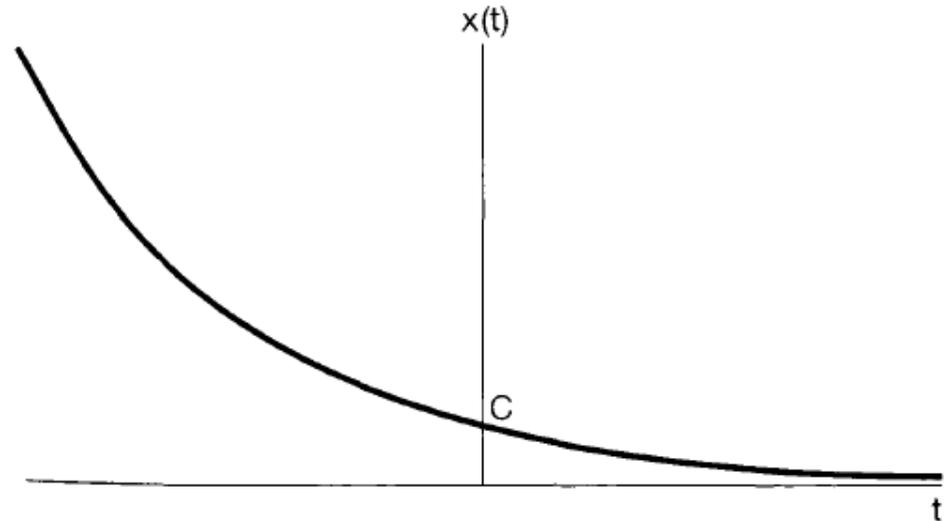
1. Complex exponential - Ce^{at}, Ce^{an} $C, a \in \mathbb{C}$
2. Impulse function - $\delta(t), \delta[n]$
 - Will want to represent general signals as linear combination of these special signals
 - The essence of linear system analysis
 - Typically,
 - Impulse functions \rightarrow time-domain analysis
 - Complex exponentials \rightarrow frequency/transform domain analysis

REAL EXPONENTIAL SIGNALS

- $x(t) = Ce^{at}$ $C, a \in \mathbb{R}$



$a > 0$ exponential growth



$a < 0$ exponential decay

- $a = 0, x(t) = C$: constant function

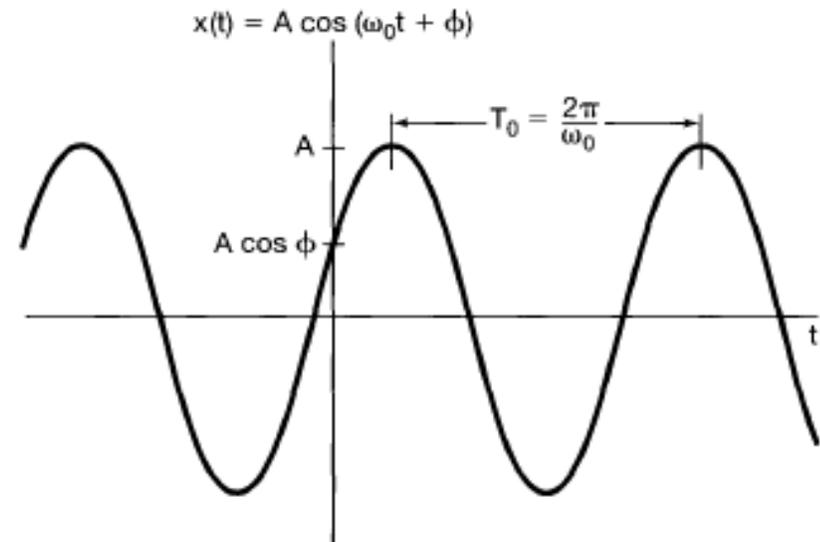
PERIODIC COMPLEX EXPONENTIAL

- $x(t) = C e^{at}$
 - $a = j\omega_0, C = A e^{j\theta}$
 - a is purely complex

$$\begin{aligned} x(t) &= A e^{j\theta} e^{j\omega_0 t} = A e^{j(\omega_0 t + \theta)} \\ &= \underbrace{A \cos(\omega_0 t + \theta)}_{\text{real}} + j \underbrace{A \sin(\omega_0 t + \theta)}_{\text{imaginary}} \end{aligned}$$

- $x(t)$ is a pair of sinusoidal signals with the same amplitude A , frequency ω_0 , and phase shift θ

- $\text{Re}\{x(t)\} = A \cos(\omega_0 t + \theta)$



- Proof
$$\begin{aligned} x(t) &= x(t + T_0) \\ &= C e^{j\omega(t + 2\pi/\omega_0)} \\ &= C e^{j\omega_0 t} \underbrace{e^{j2\pi}}_{=1} = C e^{j\omega_0 t} = x(t) \end{aligned}$$

GENERAL COMPLEX EXPONENTIAL

- $x(t) = Ce^{at}$

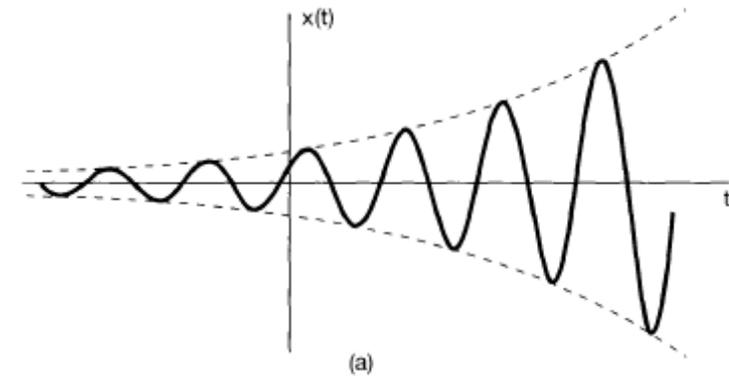
- $a = r + j\omega, C = Ae^{j\theta}$

$$\begin{aligned}
 x(t) &= Ce^{at} = Ae^{j\theta} e^{(r+j\omega_0)t} = Ae^{rt} e^{j(\omega_0 t + \theta)} \\
 &= \underbrace{Ae^{rt} \cos(\omega_0 t + \theta)}_{\text{real}} + j \underbrace{Ae^{rt} \sin(\omega_0 t + \theta)}_{\text{imaginary}}
 \end{aligned}$$

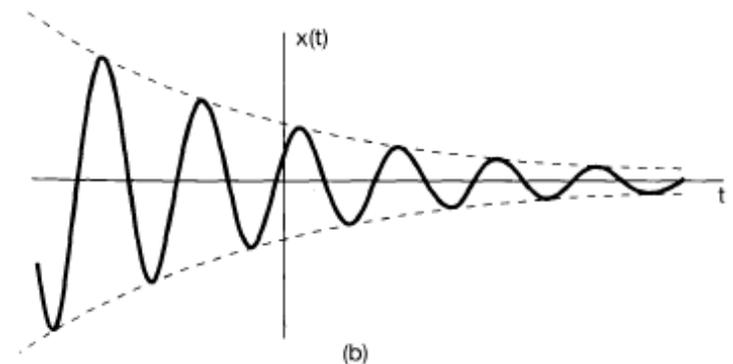
- Amplitude controlled sinusoid

- Ae^{rt} defines envelope

- $r > 0$



- $r < 0$

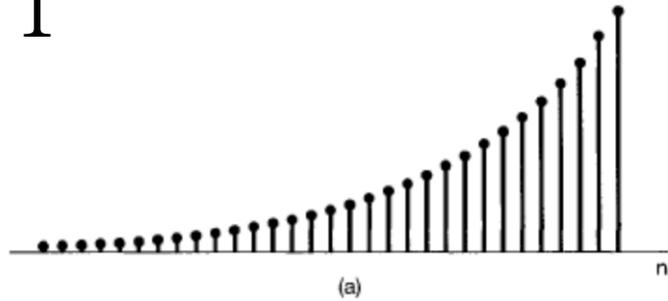


DT COMPLEX EXPONENTIAL - REAL

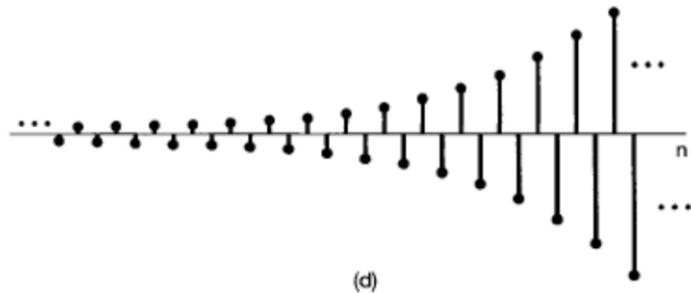
- $x[n] = Ce^{\beta n}$ or $x[n] = C\alpha^n$

- $\alpha = e^\beta, C, \beta \in \mathbb{C}$

- $\alpha > 1$



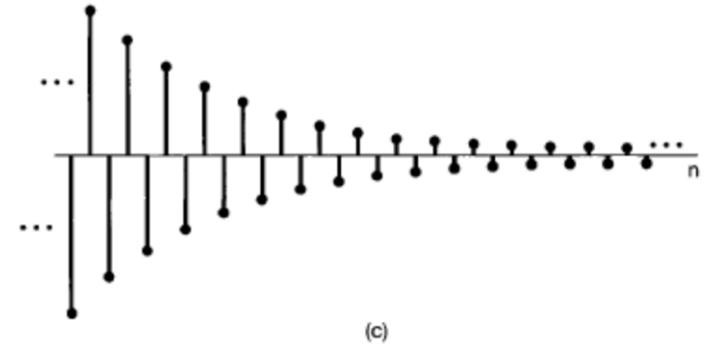
- $\alpha < -1$



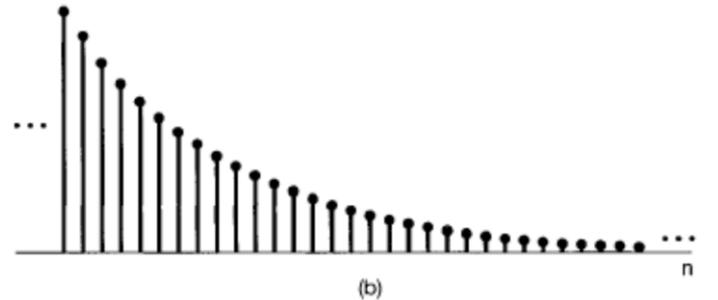
- Real exponential

- $C, \alpha \in \mathbb{R}$

- $-1 < \alpha < 0$



- $0 < \alpha < 1$



GENERAL DT COMPLEX EXPONENTIAL

- $x[n] = C\alpha^n, C, \alpha \in \mathbb{C}$

- $C = |C|e^{j\theta}, \alpha = |\alpha|e^{j\omega_0}$

$$x[n] = |C|e^{j\theta} (|\alpha|e^{j\omega_0})^n$$

$$= |C||\alpha|^n e^{j(\omega_0 n + \theta)}$$

$$= |C||\alpha|^n \cos(\omega_0 n + \theta) + j|C||\alpha|^n \sin(\omega_0 n + \theta)$$

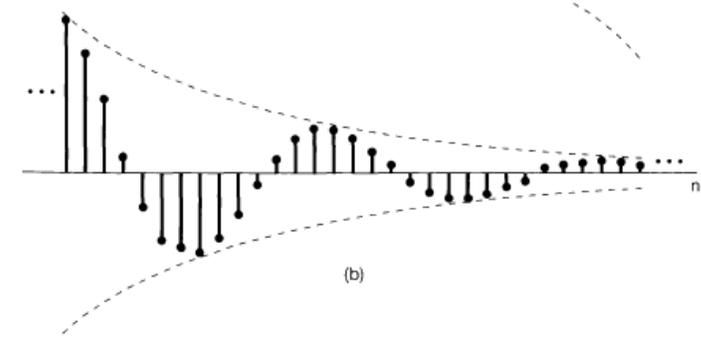
- Three cases for $|\alpha|$

- $|\alpha| = 1$

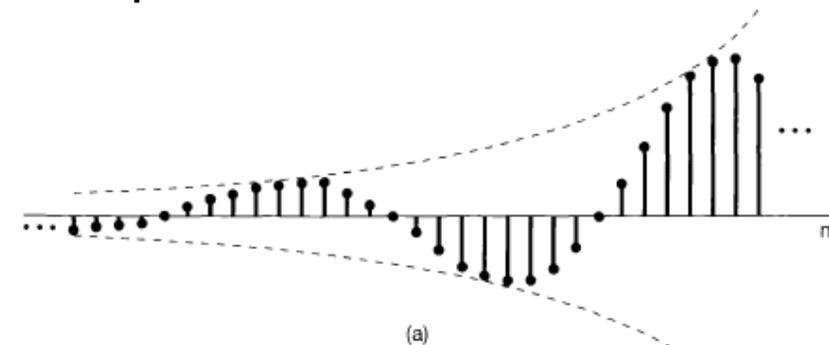
$$x[n] = |C| \cos(\omega_0 n + \theta) + j|C| \sin(\omega_0 n + \theta)$$

- Not necessarily periodic

- $|\alpha| < 1$ - decaying exponential envelope



- $|\alpha| > 1$ - Growing exponential envelope



PERIODICITY OF DT COMPLEX EXPONENTIALS

- Unlike CT, there are conditions for periodicity
- Consider frequency $\omega_0 + 2\pi$
- $e^{j(\omega_0+2\pi)n} = e^{j\omega_0n} e^{j2\pi n} = e^{j\omega_0n}$
 - Exponential with freq $\omega_0 + 2\pi$ is the same as exp. with freq ω_0
- \rightarrow Only need to consider a 2π interval for ω_0
 - $0 \leq \omega_0 \leq 2\pi$ or $-\pi \leq \omega_0 \leq \pi$
 - See Fig. 1.27 of book

DT FREQUENCY RANGE

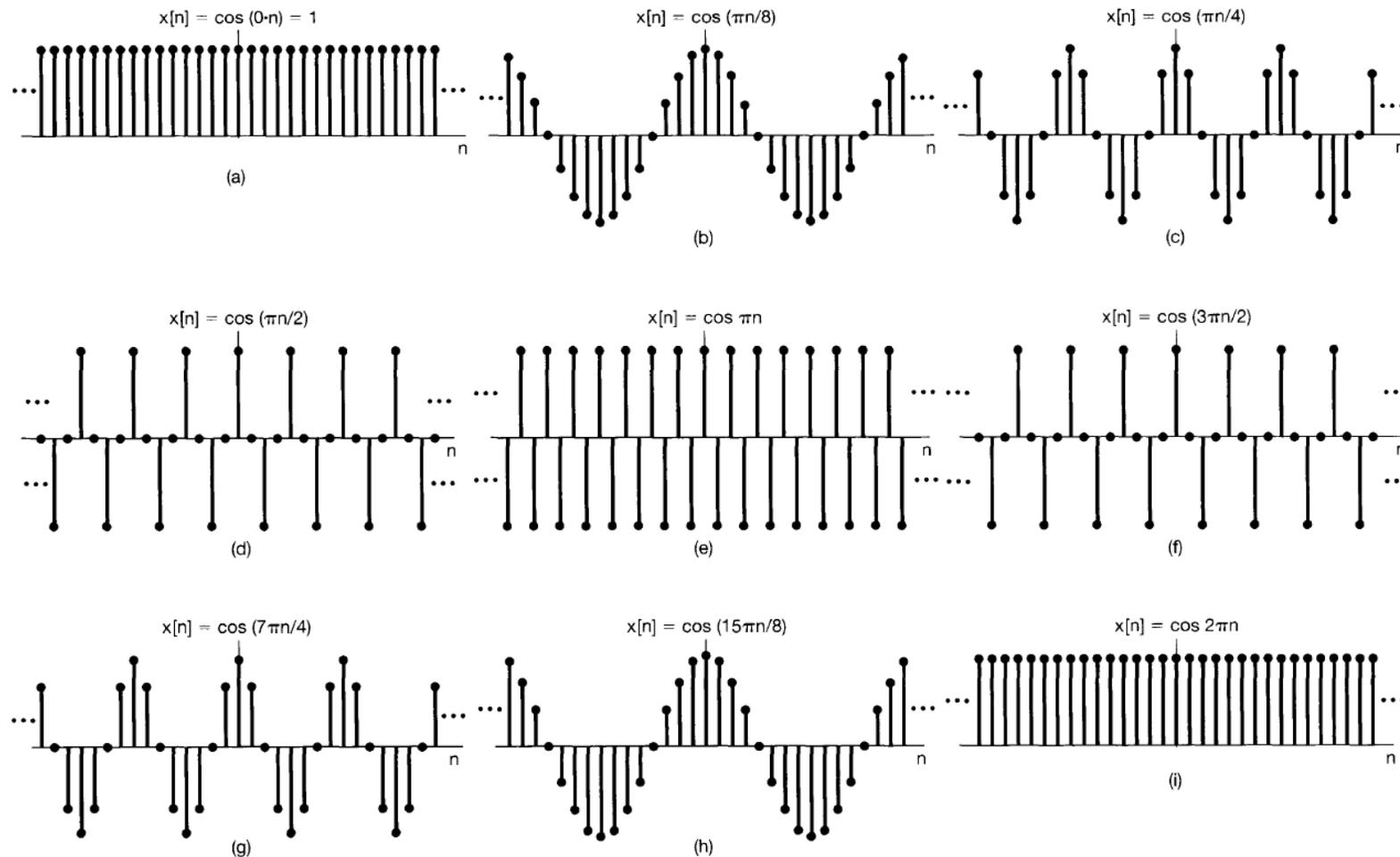


Figure 1.27 Discrete-time sinusoidal sequences for several different frequencies.

DT PERIODICITY CONSTRAINT

$$x[n] = x[n + N] \quad \forall n \in \mathbb{Z}$$

$$e^{j\Omega_0 n} = e^{j\Omega_0(n+N)} = e^{j\Omega_0 n} e^{j\Omega_0 N}$$

$$\Rightarrow e^{j\Omega_0 N} = 1 = e^{j2\pi m} \quad m \in \mathbb{Z}$$

$$\Rightarrow \Omega_0 N = 2\pi m$$

$$\Rightarrow \Omega_0 = \frac{2\pi m}{N}$$

- $e^{j\Omega_0 n}$ is periodic iff Ω_0 is a rational multiple of 2π
 - Fundamental period: $N = \frac{2\pi m}{\Omega_0}$
 - $\frac{m}{N}$ is in reduced form
 - $\text{gcd}(m, N) = 1 \leftarrow$ greatest common denominator

- Table 1.1 is good for highlighting the differences between DT and CT

TABLE 1.1 Comparison of the signals $e^{j\omega_0 t}$ and $e^{j\omega_0 n}$.

$e^{j\omega_0 t}$	$e^{j\omega_0 n}$
Distinct signals for distinct values of ω_0	Identical signals for values of ω_0 separated by multiples of 2π
Periodic for any choice of ω_0	Periodic only if $\omega_0 = 2\pi m/N$ for some integers $N > 0$ and m .
Fundamental frequency ω_0	Fundamental frequency* ω_0/m
Fundamental period $\omega_0 = 0$: undefined $\omega_0 \neq 0$: $\frac{2\pi}{\omega_0}$	Fundamental period* $\omega_0 = 0$: undefined $\omega_0 \neq 0$: $m \left(\frac{2\pi}{\omega_0} \right)$

*Assumes that m and N do not have any factors in common.

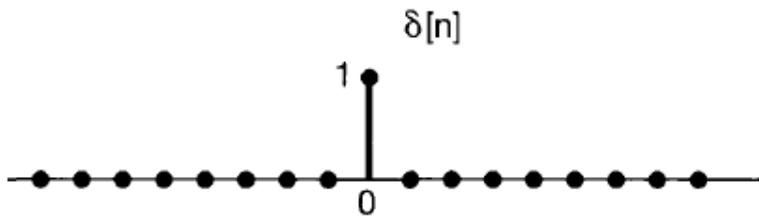
THE UNIT IMPULSE AND UNIT STEP FUNCTIONS

CHAPTER 1.4

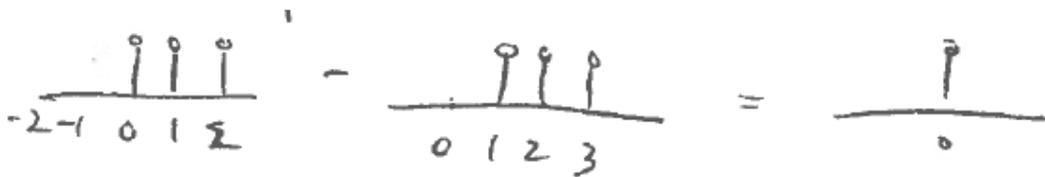
DT IMPULSE AND UNIT STEP FUNCTIONS

- Unit impulse (Kronecker delta)

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

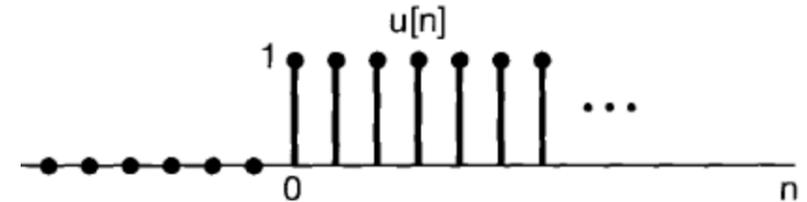


$$\delta[n] = u[n] - u[n - 1]$$



- Unit step

$$u[n] = \begin{cases} 1 & n \geq 0 \\ 0 & n < 0 \end{cases}$$

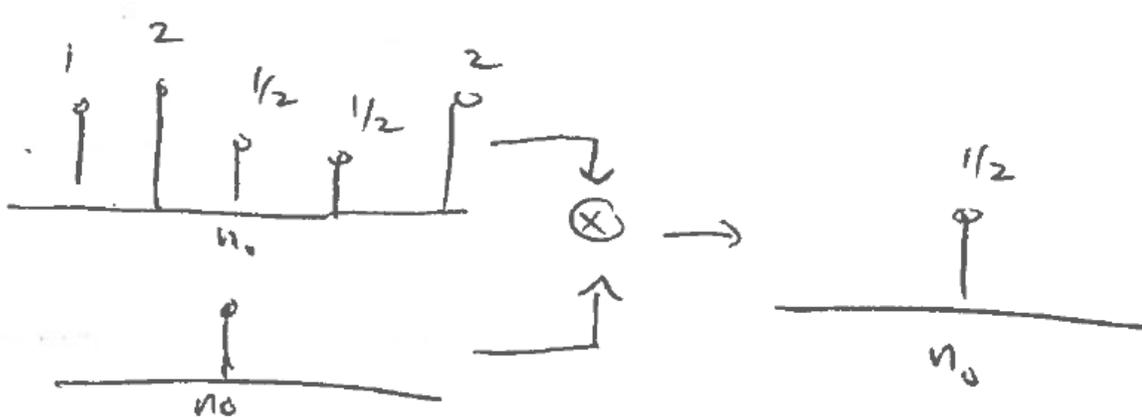


- $u[n] = \sum_{m=-\infty}^n \delta[m]$
 - Running (cumulative) sum
- $u[n] = \sum_{k=0}^{\infty} \delta[n - k]$
- $= \sum_{k=-\infty}^{\infty} u[k] \delta[n - k]$
 - Sum of delayed impulses

SAMPLING/SIFTING PROPERTIES

■ Sampling Property

- $x[n]\delta[n] = x[0]\delta[n]$
- $x[n]\delta[n - n_0] = x[n_0]\delta[n - n_0]$



■ Product of signals is a signal

- Multiply values at corresponding time

■ Sifting Property

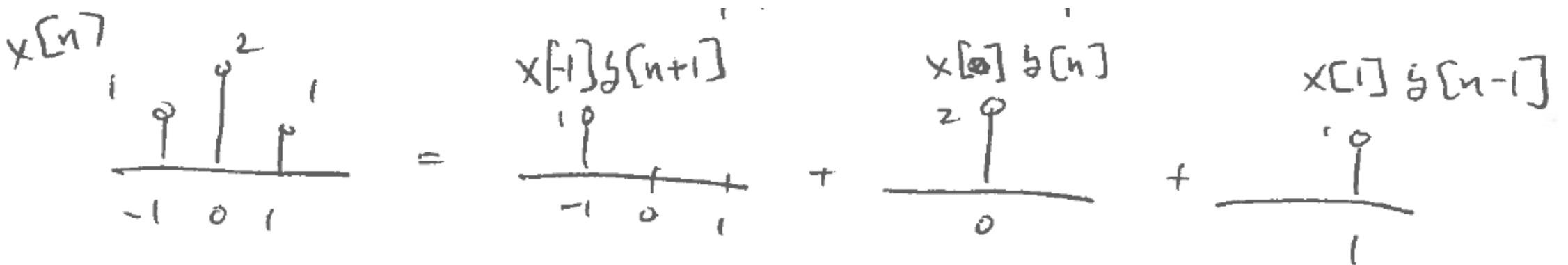
- $\sum_{m=-\infty}^{\infty} x[m]\delta[m] = x[0]$
- $\sum_{m=-\infty}^{\infty} x[m]\delta[m - n_0] = x[n_0]$

■ Notice above is summation of values in the sampled signal

■ More generally, this summation holds for any limits that contain the impulse

REPRESENTATION PROPERTY

- Every DT signal can be represented as a linear combination of shifted impulses
- $x[n] = \sum_{k=-\infty}^{\infty} x[k] \delta[n - k]$
 - $x[k]$ – value of signal at time k
 - A bit complicated but useful for study of LTI systems (Ch2)

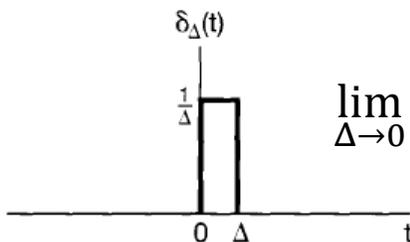
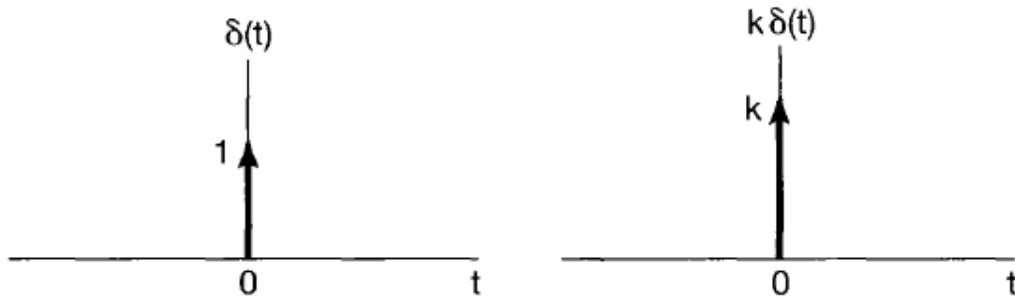


CT IMPULSE AND UNIT STEP FUNCTIONS

■ Unit impulse (dirac delta)

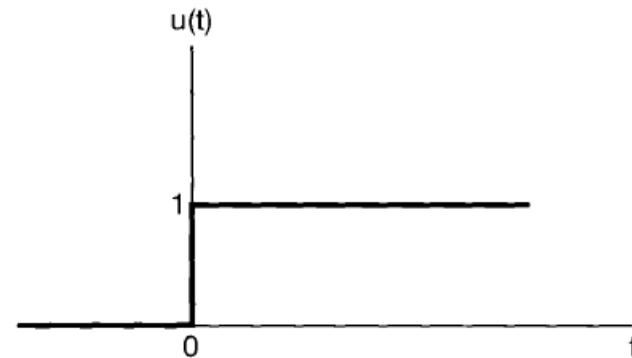
$$\delta(t) = \begin{cases} \infty & t = 0 \\ 0 & t \neq 0 \end{cases}$$

$$\text{With } \int_{-\infty}^{\infty} \delta(t) dt = 1$$



■ Unit step

$$u(t) = \begin{cases} 1 & t \geq 0 \\ 0 & t < 0 \end{cases}$$



■ Relationship

$$u(t) = \int_{-\infty}^t \delta(\tau) d\tau$$

$$\delta(t) = \frac{du(t)}{dt}$$

PROPERTIES

■ Sampling

- $x(t)\delta(t) = x(0)\delta(t)$
- $x(t)\delta(t - t_0) = x(t_0)\delta(t - t_0)$
- Product of two signals is a signal

■ Representation property

- $x(t) = \int_{-\infty}^{\infty} x(\tau)\delta(t - \tau)d\tau$

■ Example

- $u(t) = \int_{-\infty}^{\infty} u(\tau)\delta(t - \tau)d\tau$

■ Sifting

$$\begin{aligned} \int_{-\infty}^{\infty} x(t)\delta(t)dt &= \int_{-\infty}^{\infty} x(0)\delta(t)dt \\ &= x(0) \underbrace{\int_{-\infty}^{\infty} \delta(t)dt}_{=1} \\ &= x(0) \end{aligned}$$

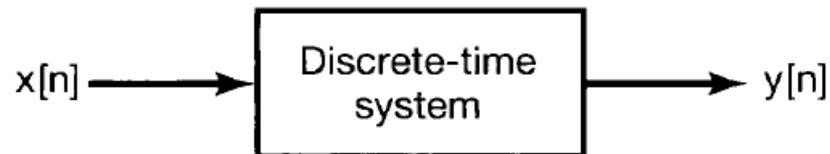
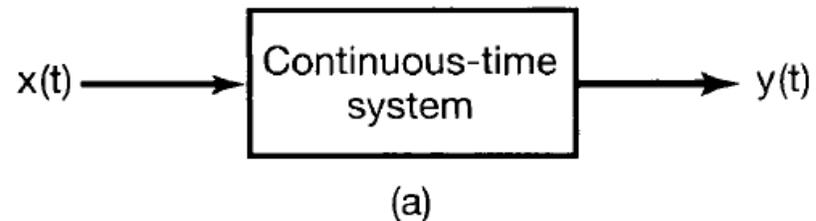
$$\int_{-\infty}^{\infty} x(t)\delta(t - t_0)dt = x(t_0)$$

CONTINUOUS-TIME AND DISCRETE-TIME SYSTEMS

CHAPTER 1.5

SYSTEMS

- A system is a quantitative description of a physical process to transform an input signal into an output signal
 - Systems are a black box – a mathematical abstraction



- Shorthand notation
 - $x(t) \rightarrow y(t)$
- More complex systems
 - Sampling

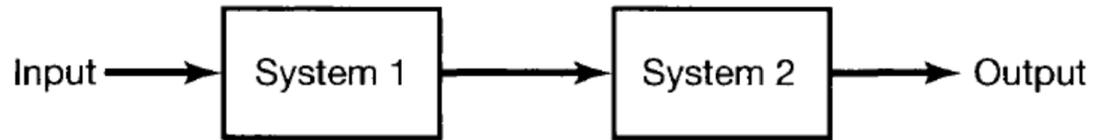


- MIMO (multi input/multi output)

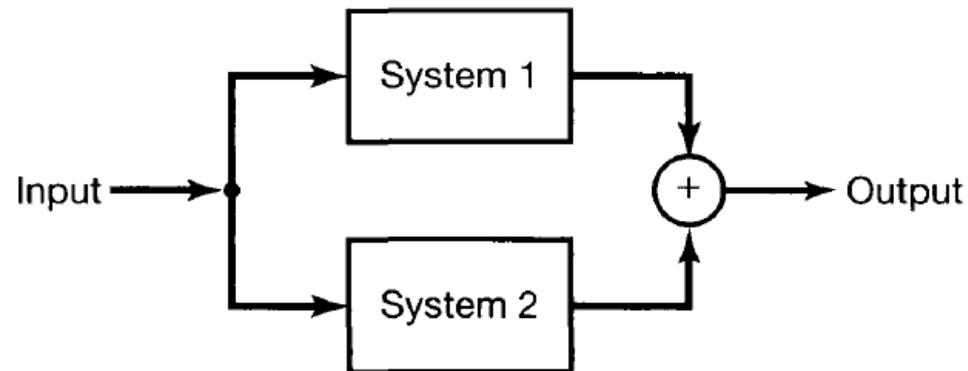


SYSTEM INTERCONNECTIONS

- Series/cascade connection

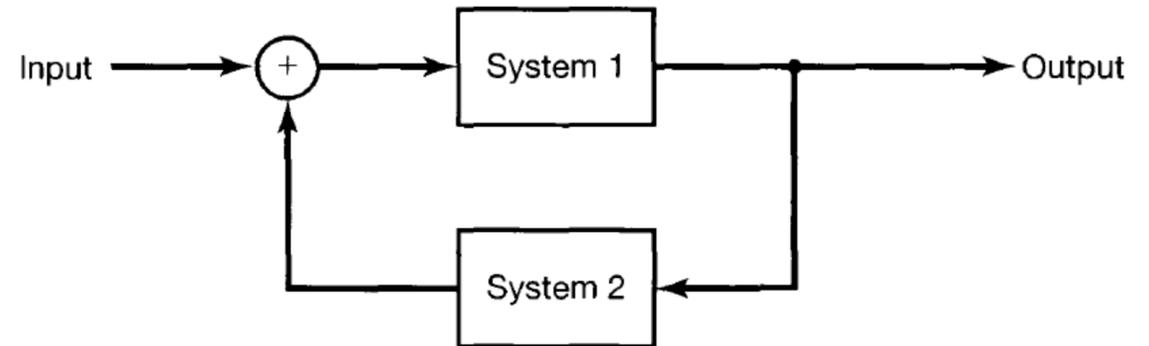


- Parallel interconnection



- Feedback connection

- Very important in controls



- More complex systems can be composed by various series/parallel interconnections

BASIC SYSTEM PROPERTIES

CHAPTER 1.6

BASIC SYSTEM PROPERTIES

- Memoryless
- Invertibility
- Causality
- Stability
- Linearity
- Time-invariance

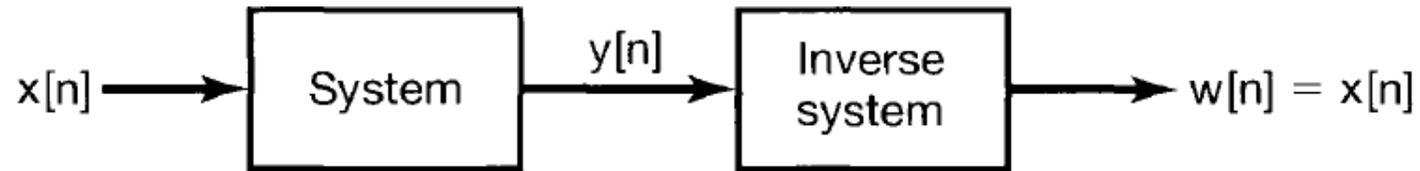
Define an important class of systems called LTI

MEMORYLESS SYSTEMS

- A system is memoryless if the output at a time t depends only on input at the same time t
- Examples
 - $y(t) = (2x(t) - x^2(t))^2$
 - $y[n] = x[n]$
 - $y[n] = x[n - 1]$
 - $y[n] = x[n] + y[n - 1]$

INVERTIBILITY

- A system is invertible if distinct inputs lead to distinct outputs



- Rules for proving invertible systems
 - Show invertible by given the inverse system expression/formula
 - Show non-invertible by any counter example

EXAMPLES: INVERSE SYSTEMS

- $y(t) = (\cos t + 2)x(t)$

- $x(t) = \frac{y(t)}{\cos t + 2}$

- Invertible (no divide by zero!)

- $y[n] = \sum_{k=-\infty}^n x[k]$

- $\Rightarrow y[n] = x[n] + y[n - 1]$

- $\Rightarrow x[n] = y[n] - y[n - 1]$

- Invertible

- $y(t) = x^2(t)$

- $x_1(t) = 1 \Rightarrow y_1(t) = 1$

- $x_2(t) = -1 \Rightarrow y_2(t) = 1$

- Need unique input \rightarrow distinct output

- Not invertible

CAUSALITY

- A system is causal if the output at any time t depends only on the input at same time t or past times $\tau < t$
- Real systems must be causal because we cannot know future values
 - Buffering gives the appearance of non-causality
- Examples
 - $y[n] = x[n]$
 - $y[n] = x[n] + x[n + 1]$
 - $y(t) = \int_{-\infty}^t x(\tau) d\tau$
 - $y[n] = x[-n]$
 - $y(t) = x(t)(\cos(t + 2))$

STABILITY

- A system is stable if a bounded input results in a bounded output signal \rightarrow BIBO stable
- A signal is bounded if there exists a constant B such that
 - $|x(t)| \leq B \quad \forall t$ and $B < \infty$
- BIBO condition
 - $|x(t)| \leq B \rightarrow |y(t)| < \infty$

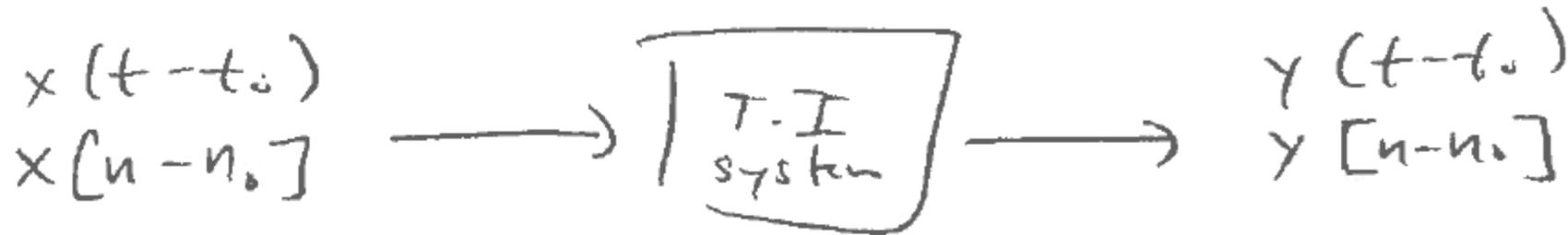
EXAMPLES: BIBO STABILITY

- $y(t) = 2x^2(t - 1) + x(3t)$

$$y[n] = \begin{cases} 0 & n < 0 \\ 1.01y[n - 1] + x[n] & n \geq 0 \end{cases}$$

TIME INVARIANCE

- A system is time-invariant if a time shift in input signal results in an identical time shift in the output signal



- Steps to check for TI
 - Assume $x(t) \rightarrow y(t)$
 1. Check $y_1(t) = y(t - t_0)$ time shift on output
 2. Check $y_2(t) = f(x(t - t_0))$ operate on time shifted input
 3. Verify $y_1(t) = y_2(t)$ for TI

EXAMPLES: TIME INVARIANT SYSTEMS

- $y(t) = \sin(x(t))$

- $y[n] = nx[n]$

LINEARITY

- A system is linear if it is additive and scalable
- If $x_1(t) \rightarrow y_1(t)$ and $x_2(t) \rightarrow y_2(t)$
 - Additive
 - $x_1(t) + x_2(t) \rightarrow y_1(t) + y_2(t)$
 - Scalable
 - $ax_1(t) \rightarrow ay_1(t) \quad a \in \mathbb{C}$
- Then, $ax_1(t) + bx_2(t) \rightarrow ay_1(t) + by_2(t)$

EXAMPLES: LINEAR SYSTEMS

- $y(t) = tx(t)$

- $y[n] = 2x^2[n]$