

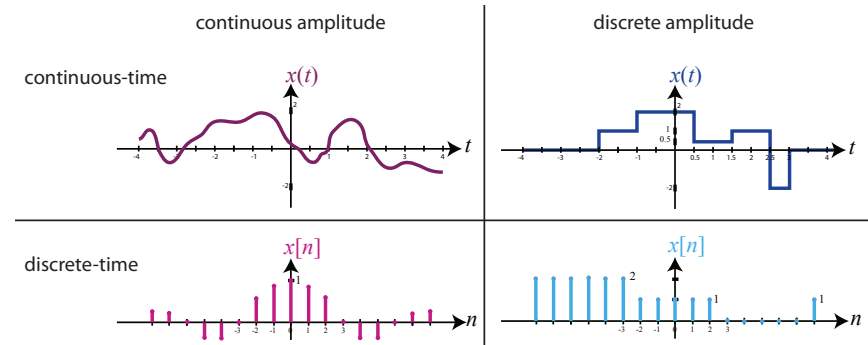
Introduction to Digital Signal Processing

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Analog and Digital Signals

- ▶ analog signal = continuous-time + continuous amplitude
- ▶ digital signal = discrete-time + discrete amplitude



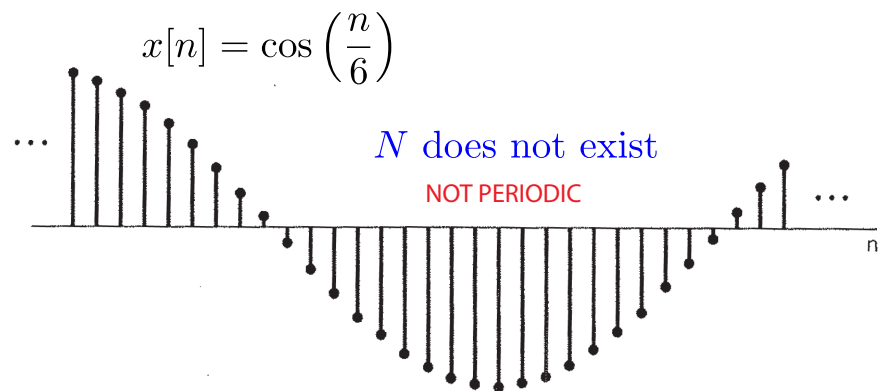
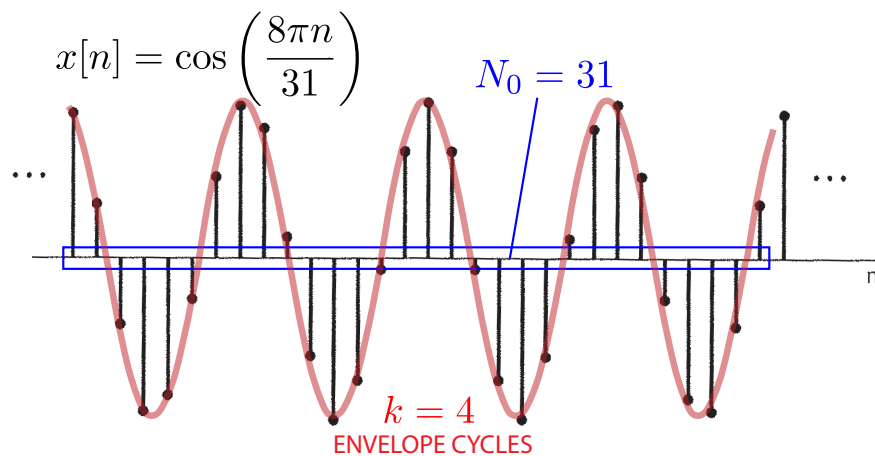
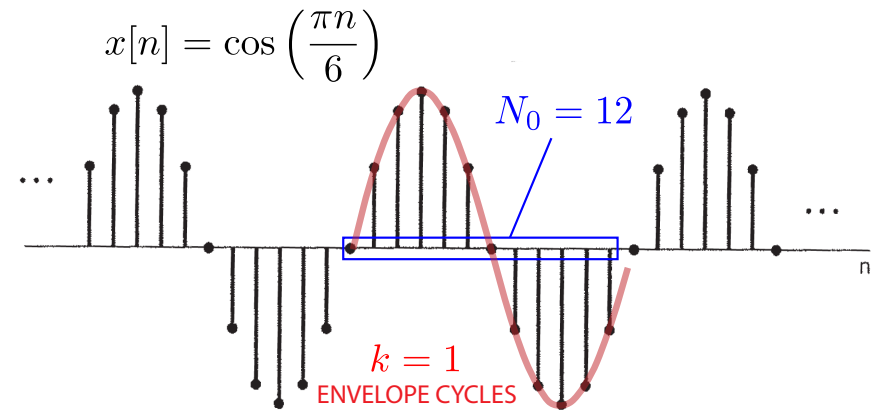
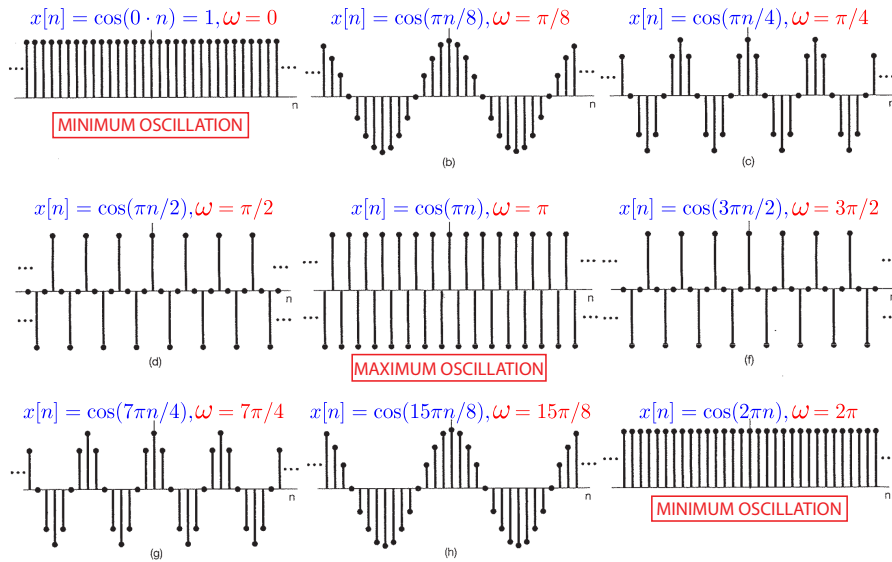
Analog and Digital Signals

- ▶ analog system = analog input + analog output
- ▶ digital system = digital input + digital output

Discrete-time Sinusoids

$$x(n) = A \cos(\omega n + \theta) = A \cos(2\pi f n + \theta), \quad n \in \mathbb{Z}$$

- ▶ discrete-time signal (not digital), $\because -A \leq x_a(t) \leq A$ and $n \in \mathbb{Z}$
- ▶ A = amplitude
- ▶ ω = frequency in rad/sample
- ▶ f = frequency in cycles/sample; note: $\omega = 2\pi f$
- ▶ θ = phase in rad



Uniqueness: Continuous-time

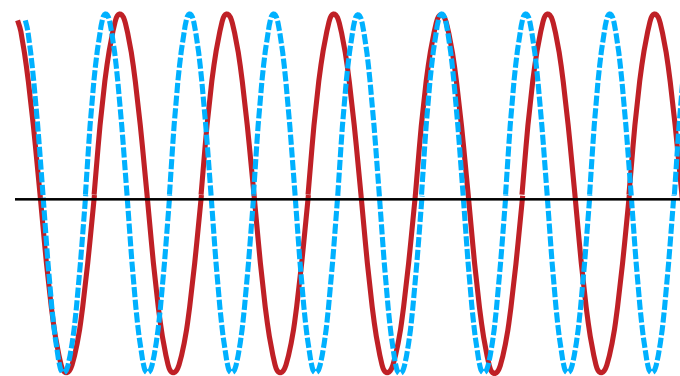
For $F_1 \neq F_2$,

$$A \cos(2\pi F_1 t + \theta) \neq A \cos(2\pi F_2 t + \theta)$$

except at discrete points in time.

Uniqueness: Continuous-time

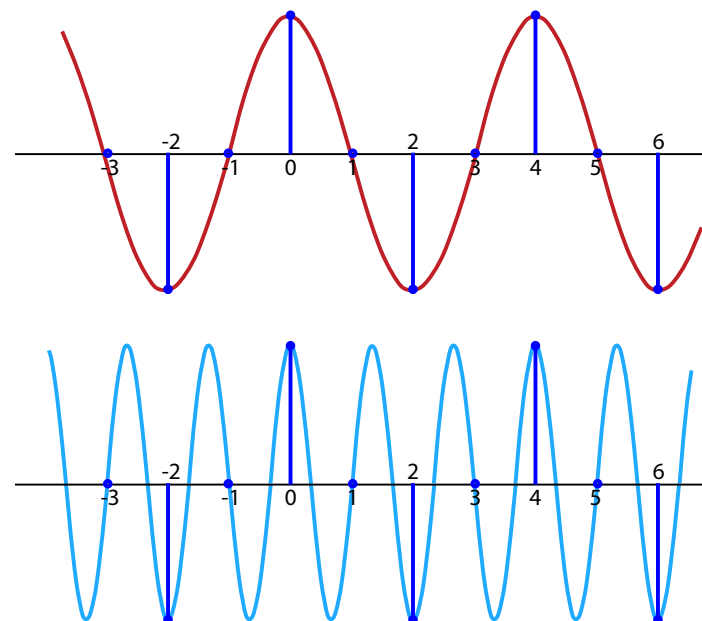
$F_1 \neq F_2$:



Uniqueness: Discrete-time

Let $f_1 = f_0 + k$ where $k \in \mathbb{Z}$,

$$\begin{aligned} x_1(n) &= A e^{j(2\pi f_1 n + \theta)} \\ &= A e^{j(2\pi(f_0 + k)n + \theta)} \\ &= A e^{j(2\pi f_0 n + \theta)} \cdot e^{j(2\pi k n)} \\ &= x_0(n) \cdot 1 = x_0(n) \end{aligned}$$



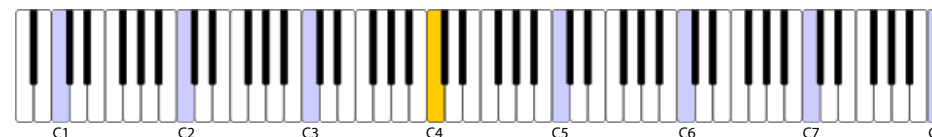
Harmonically Related Complex Exponentials

Harmonically related (cts-time) $s_k(t) = e^{jk\Omega_0 t} = e^{j2\pi k F_0 t}$,
 $k = 0, \pm 1, \pm 2, \dots$

Scientific Designation	Frequency (Hz)	k for $F_0 = 8.176$
C-1	8.176	1
C0	16.352	2
C1	32.703	4
C2	65.406	8
C3	130.813	16
C4	261.626	32
⋮	⋮	
C9	8372.018	1024

Harmonically Related Complex Exponentials

Scientific Designation	Frequency (Hz)	k for $F_0 = 8.176$
C1	32.703	4
C2	65.406	8
C3	130.813	16
C4 (middle C)	261.626	32
C5	523.251	64
C6	1046.502	128
C7	2093.005	256
C8	4186.009	512



Harmonically Related Complex Exponentials

What does the family of harmonically related sinusoids $s_k(t)$ have in common?

Harmonically related (cts-time) $s_k(t) = e^{jk\Omega_0 t} = e^{j2\pi(kF_0)t}$,
 $k = 0, \pm 1, \pm 2, \dots$

fund. period: $T_{0,k} = \frac{1}{\text{cyclic frequency}} = \frac{1}{kF_0}$

period: $T_k = \text{any integer multiple of } T_{0,k}$

common period: $T = k \cdot T_{0,k} = \frac{1}{F_0}$

Harmonically Related Complex Exponentials

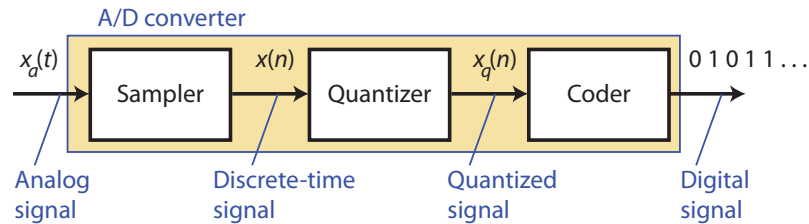
Discrete-time Case:

For periodicity, select $f_0 = \frac{1}{N}$ where $N \in \mathbb{Z}$:

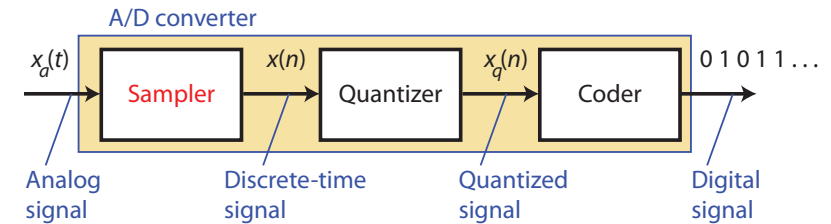
Harmonically related (dts-time) $s_k(n) = e^{j2\pi k f_0 n} = e^{j2\pi k n / N}$,
 $k = 0, \pm 1, \pm 2, \dots$

- There are only N distinct dst-time harmonics:
 $s_k(n)$, $k = 0, 1, 2, \dots, N-1$.

Analog-to-Digital Conversion



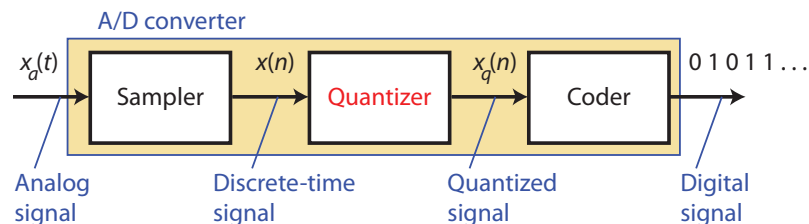
Analog-to-Digital Conversion



Sampling:

- ▶ conversion from cts-time to dst-time by taking “samples” at discrete time instants
- ▶ E.g., uniform sampling: $x(n) = x_a(nT)$ where T is the sampling period and $n \in \mathbb{Z}$

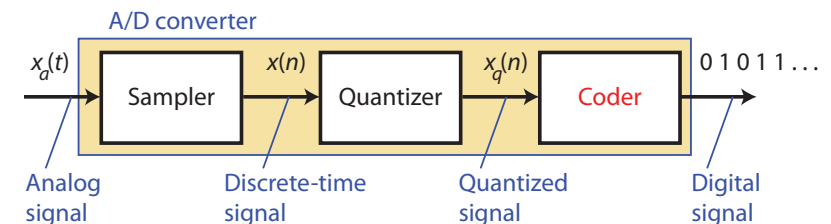
Analog-to-Digital Conversion



Quantization:

- ▶ conversion from dst-time cts-valued signal to a dst-time dst-valued signal
- ▶ quantization error: $e_q(n) = x_q(n) - x(n)$ for all $n \in \mathbb{Z}$

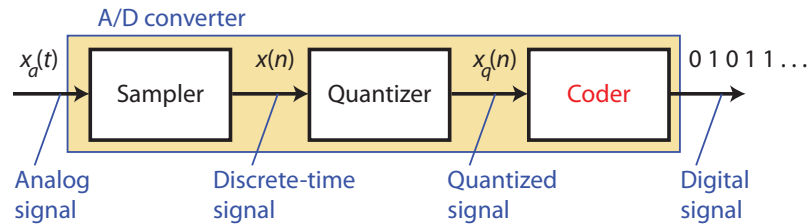
Analog-to-Digital Conversion



Coding:

- ▶ representation of each dst-value $x_q(n)$ by a **b -bit binary sequence**
- ▶ e.g., if for any n , $x_q(n) \in \{0, 1, \dots, 6, 7\}$, then the coder may use the following mapping to code the quantized amplitude:

Analog-to-Digital Conversion



Example coder:

0	000		4	100
1	001		5	101
2	010		6	110
3	011		7	111

Sampling Theorem

If the **highest frequency** contained in an analog signal $x_a(t)$ is $F_{max} = B$ and the signal is sampled at a rate

$$F_s > 2F_{max} = 2B$$

then $x_a(t)$ can be exactly recovered from its sample values using the interpolation function

$$g(t) = \frac{\sin(2\pi Bt)}{2\pi Bt}$$

Note: $F_N = 2B = 2F_{max}$ is called the **Nyquist rate**.

Sampling Theorem

$$\text{Sampling Period} = T = \frac{1}{F_s} = \frac{1}{\text{Sampling Frequency}}$$

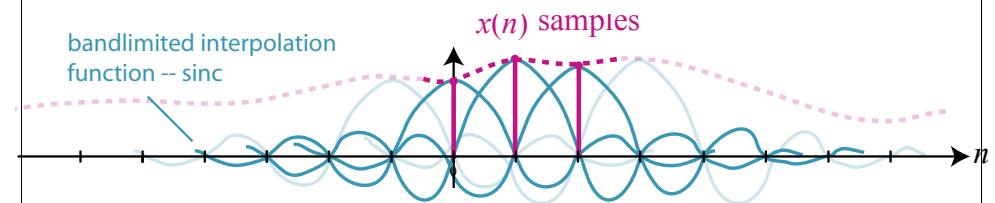
Therefore, given the interpolation relation, $x_a(t)$ can be written as

$$x_a(t) = \sum_{n=-\infty}^{\infty} x_a(nT)g(t - nT)$$

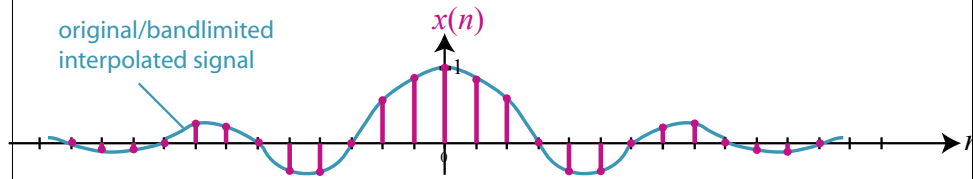
$$x_a(t) = \sum_{n=-\infty}^{\infty} x(n)g(t - nT)$$

where $x_a(nT) = x(n)$; called **bandlimited interpolation**.

Bandlimited Interpolation

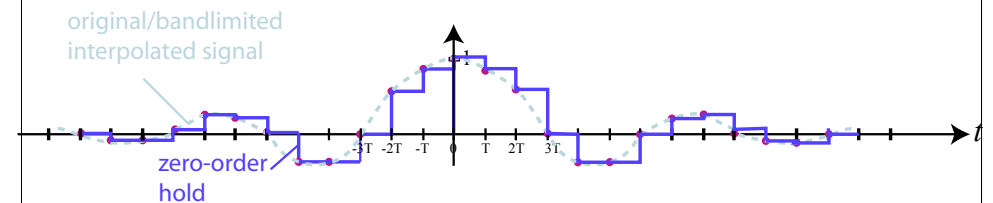


Digital-to-Analog Conversion



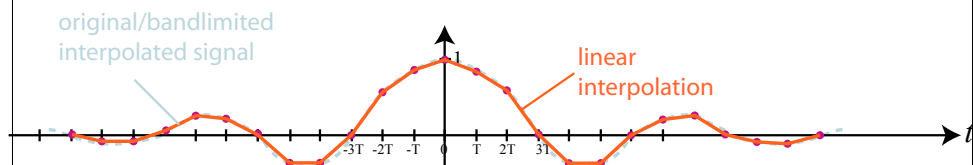
- ▶ Common interpolation approaches: bandlimited interpolation, zero-order hold, linear interpolation, higher-order interpolation techniques, e.g., using splines
- ▶ In practice, “cheap” interpolation along with a smoothing filter is employed.

Digital-to-Analog Conversion



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Digital-to-Analog Conversion



- ▶ Common interpolation approaches: bandlimited interpolation, zero-order hold, linear interpolation, higher-order interpolation techniques, e.g., using splines
- ▶ In practice, “cheap” interpolation along with a smoothing filter is employed.

Elementary Discrete-Time Signals

1. unit sample sequence (a.k.a. Kronecker delta function):

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

2. unit step signal:

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

3. unit ramp signal:

$$u_r(n) = \begin{cases} n, & \text{for } n \geq 0 \\ 0, & \text{for } n < 0 \end{cases}$$

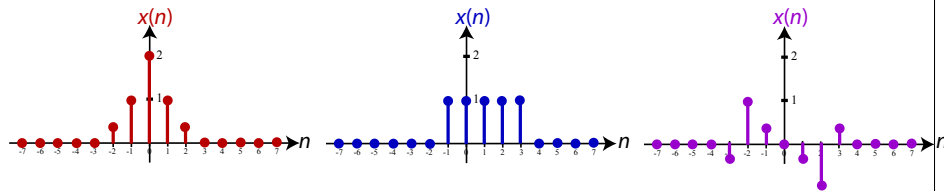
Note:

$$\begin{aligned} \delta(n) &= u(n) - u(n-1) = u_r(n+1) - 2u_r(n) + u_r(n-1) \\ u(n) &= u_r(n+1) - u_r(n) \end{aligned}$$

Signal Symmetry

Even signal: $x(-n) = x(n)$

Odd signal: $x(-n) = -x(n)$



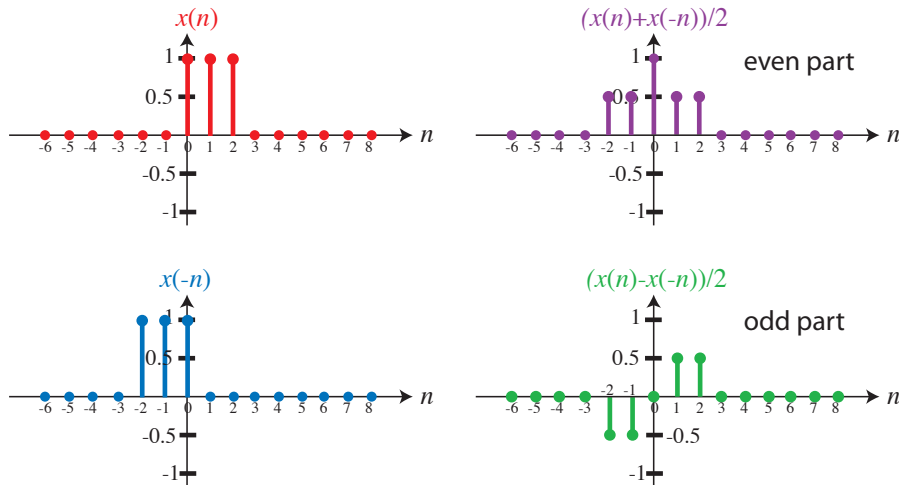
Signal Symmetry

Even signal component: $x_e(n) = \frac{1}{2} [x(n) + x(-n)]$

Odd signal component: $x_o(n) = \frac{1}{2} [x(n) - x(-n)]$

Note: $x(n) = x_e(n) + x_o(n)$

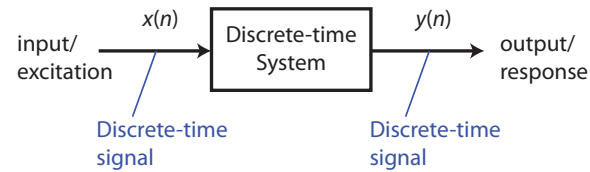
Signal Symmetry



Simple Manipulation of Discrete-Time Signals

- ▶ Transformation of independent variable:
 - ▶ time shift: $n \rightarrow n - k$, $k \in \mathbb{Z}$
 - ▶ Question: what if $k \notin \mathbb{Z}$?
 - ▶ time scale: $n \rightarrow \alpha n$, $\alpha \in \mathbb{Z}$
 - ▶ Question: what if $\alpha \notin \mathbb{Z}$?
- ▶ Additional, multiplication and scaling:
 - ▶ amplitude scaling: $y(n) = Ax(n)$, $-\infty < n < \infty$
 - ▶ sum: $y(n) = x_1(n) + x_2(n)$, $-\infty < n < \infty$
 - ▶ product: $y(n) = x_1(n)x_2(n)$, $-\infty < n < \infty$

Input-Output Description of Dst-Time Systems



- ▶ Input-output description (exact structure of system is unknown or ignored):

$$y(n) = \mathcal{T}[x(n)]$$

- ▶ “black box” representation:

$$x(n) \xrightarrow{\mathcal{T}} y(n)$$

Classification of Discrete-Time Systems

Common System Properties:

static vs. dynamic

time-invariant vs. time-variant

linear vs. nonlinear

causal vs. non-causal

stable vs. unstable systems

⋮

⋮

Static vs. Dynamic

- ▶ **Static system** (a.k.a. memoryless): the output at time n depends only on the input sample at time n ; otherwise the system is said to be **dynamic**

- ▶ a system is static iff (if and only if)

$$y(n) = \mathcal{T}[x(n), n]$$

for every time instant n .

Static vs. Dynamic

- ▶ Consider the general system:

$$y(n) = \mathcal{T}[x(n-N), x(n-N+1), \dots, x(n-1), x(n), x(n+1), \dots, x(n+M-1), x(n+M)], \quad N, M > 0$$

- ▶ For $N = M = 0$, $y(n) = \mathcal{T}[x(n)]$, the system is **static**.
- ▶ For $0 < N, M < \infty$, the system is said to be **dynamic** with finite memory of duration $N + M + 1$.
- ▶ For either N and/or M equal to infinite, the system is said to have infinite memory.

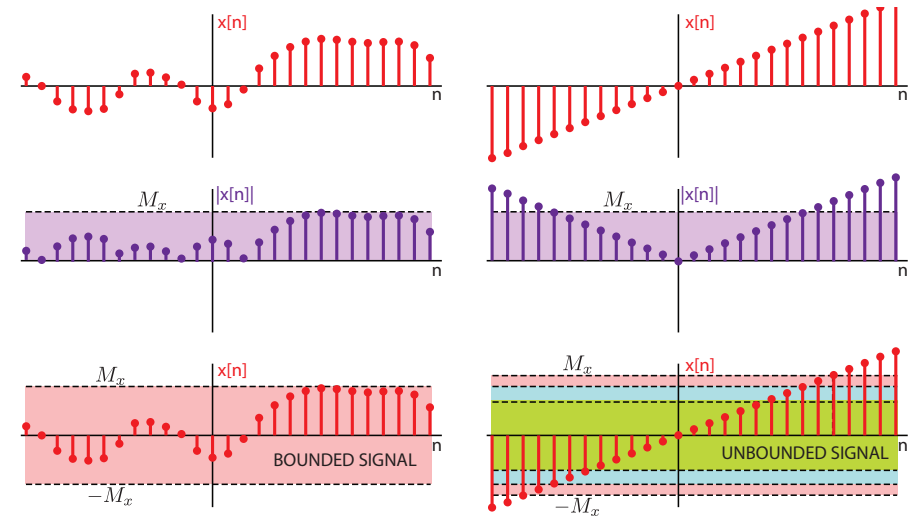
Static vs. Dynamic

Examples: memoryless or not?

- ▶ $y(n) = A x(n)$, $A \neq 0$
- ▶ $y(n) = A x(n) + B$, $A, B, \neq 0$
- ▶ $y(n) = x(n) \cos(\frac{\pi}{25}(n - 5))$
- ▶ $y(n) = x(-n)$
- ▶ $y(n) = x(n + 1)$
- ▶ $y(n) = \frac{1}{1-x(n+2)}$
- ▶ $y(n) = e^{3x(n)}$
- ▶ $y(n) = \sum_{k=-\infty}^n x(k)$

Ans: Y, Y, Y, N, N, N, Y, N

Discrete-Time Bounded Signals



The Convolution Sum

Recall:

$$x(n) = \sum_{k=-\infty}^{\infty} x(k) \delta(n - k)$$

The Convolution Sum

Let the response of a linear time-invariant (LTI) system to the unit sample input $\delta(n)$ be $h(n)$.

$$\begin{aligned} \delta(n) &\xrightarrow{\mathcal{T}} h(n) \\ \delta(n - k) &\xrightarrow{\mathcal{T}} h(n - k) \\ \alpha \delta(n - k) &\xrightarrow{\mathcal{T}} \alpha h(n - k) \\ x(k) \delta(n - k) &\xrightarrow{\mathcal{T}} x(k) h(n - k) \\ \sum_{k=-\infty}^{\infty} x(k) \delta(n - k) &\xrightarrow{\mathcal{T}} \sum_{k=-\infty}^{\infty} x(k) h(n - k) \\ x(n) &\xrightarrow{\mathcal{T}} y(n) \end{aligned}$$

The Convolution Sum

Therefore,

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

for any LTI system.

Finite vs. Infinite Impulse Response

Implementation: **Two classes**

Finite impulse response (FIR):

$$y(n) = \sum_{k=0}^{M-1} h(k)x(n-k) \quad \left. \vphantom{\sum_{k=0}^{M-1}} \right\} \text{nonrecursive systems}$$

Infinite impulse response (IIR):

$$y(n) = \sum_{k=0}^{\infty} h(k)x(n-k) \quad \left. \vphantom{\sum_{k=0}^{\infty}} \right\} \text{recursive systems}$$

System Realization

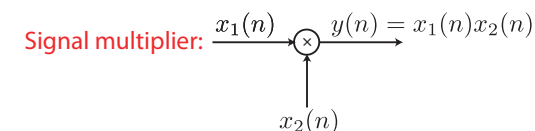
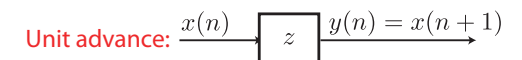
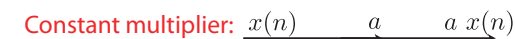
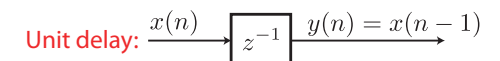
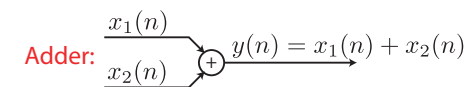
General expression for M th-order LCCDE:

$$\sum_{k=0}^N a_k y(n-k) = \sum_{k=0}^M b_k x(n-k) \quad a_0 \triangleq 1$$

Initial conditions: $y(-1), y(-2), y(-3), \dots, y(-N)$.

Need: (1) **constant scale**, (2) **addition**, (3) **delay** elements.

Building Block Elements



Direct Form I vs. Direct Form II Realizations

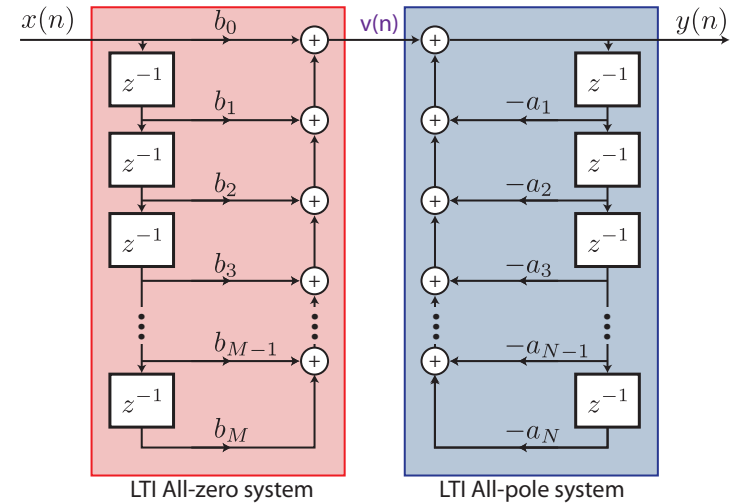
$$y(n) = - \sum_{k=1}^N a_k y(n-k) + \sum_{k=0}^M b_k x(n-k)$$

is equivalent to the **cascade** of the following systems:

$$\underbrace{v(n)}_{\text{output 1}} = \sum_{k=0}^M b_k \underbrace{x(n-k)}_{\text{input 1}} \quad \text{nonrecursive}$$

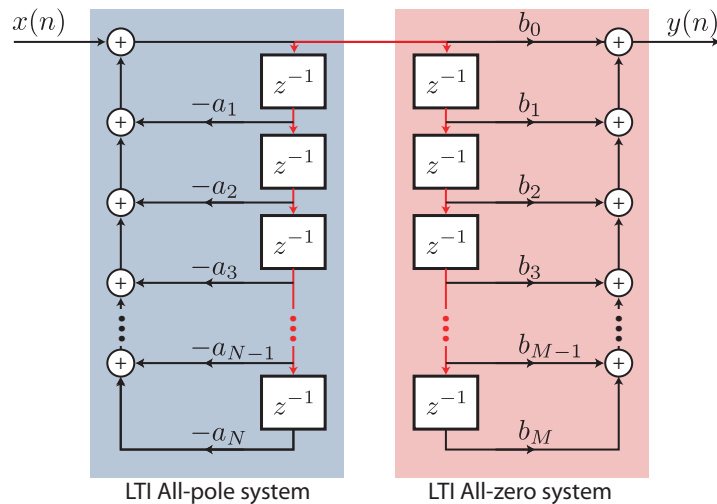
$$\underbrace{y(n)}_{\text{output 2}} = - \sum_{k=1}^N a_k y(n-k) + \underbrace{v(n)}_{\text{input 2}} \quad \text{recursive}$$

Direct Form I IIR Filter Implementation



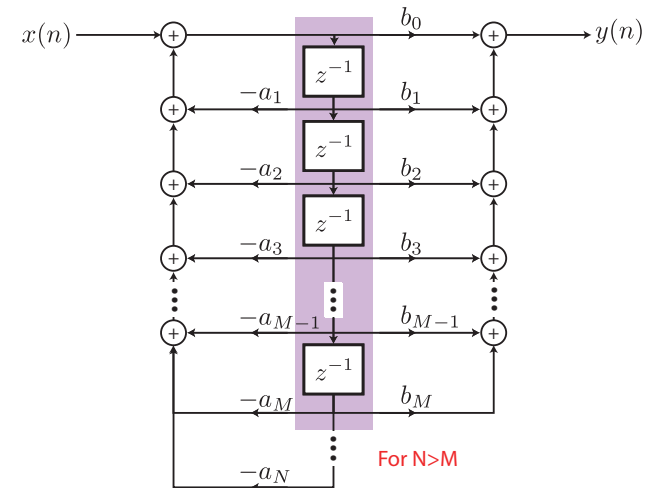
Requires: $M + N + 1$ multiplications, $M + N$ additions, $M + N$ memory locations

Direct Form II IIR Filter Implementation



Requires: $M + N + 1$ multiplications, $M + N$ additions, $M + N$ memory locations

Direct Form II IIR Filter Implementation



Requires: $M + N + 1$ multiplications, $M + N$ additions, $\max(M, N)$ memory locations

The Direct z-Transform

- ▶ Direct z-Transform:

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

- ▶ Notation:

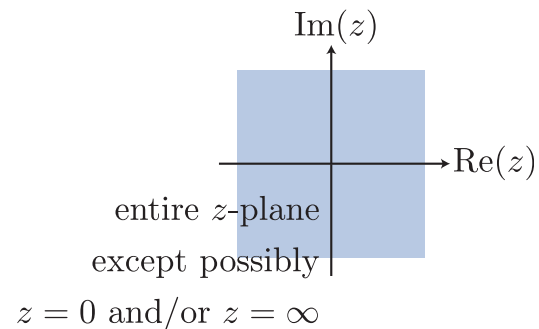
$$X(z) \equiv \mathcal{Z}\{x(n)\}$$

$$x(n) \xleftarrow{\mathcal{Z}} X(z)$$

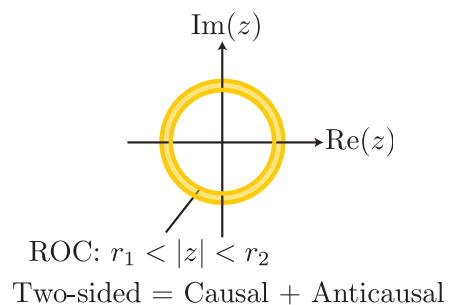
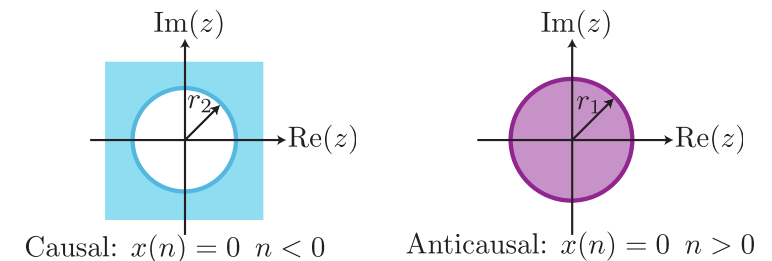
Region of Convergence

- ▶ the region of convergence (ROC) of $X(z)$ is the set of all values of z for which $X(z)$ attains a finite value
- ▶ The z-Transform is, therefore, uniquely characterized by:
 1. expression for $X(z)$
 2. ROC of $X(z)$

ROC Families: Finite Duration Signals



ROC Families: Infinite Duration Signals



z-Transform Properties

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
Linearity:	$a_1x_1(n) + a_2x_2(n)$	$a_1X_1(z) + a_2X_2(z)$	At least ROC ₁ ∩ ROC ₂
Time shifting:	$x(n - k)$	$z^{-k}X(z)$	At least ROC, except $z = 0$ (if $k > 0$) and $z = \infty$ (if $k < 0$)
z-Scaling:	$a^n x(n)$	$X(a^{-1}z)$	$ a r_2 < z < a r_1$
Time reversal	$x(-n)$	$X(z^{-1})$	$\frac{1}{r_1} < z < \frac{1}{r_2}$
Conjugation:	$x^*(n)$	$X^*(z^*)$	ROC
z-Differentiation:	$n x(n)$	$-z \frac{dX(z)}{dz}$	$r_2 < z < r_1$
Convolution:	$x_1(n) * x_2(n)$	$X_1(z)X_2(z)$	At least ROC ₁ ∩ ROC ₂
			among others ...

Common Transform Pairs

	Signal, $x(n)$	z-Transform, $X(z)$	ROC
1	$\delta(n)$	1	All z
2	$u(n)$	$\frac{1}{1-z^{-1}}$	$ z > 1$
3	$a^n u(n)$	$\frac{1}{1-az^{-1}}$	$ z > a $
4	$na^n u(n)$	$\frac{az^{-1}}{(1-az^{-1})^2}$	$ z > a $
5	$-a^n u(-n-1)$	$\frac{1}{1-az^{-1}}$	$ z < a $
6	$-na^n u(-n-1)$	$\frac{az^{-1}}{(1-az^{-1})^2}$	$ z < a $
7	$\cos(\omega_0 n) u(n)$	$\frac{1-z^{-1} \cos \omega_0}{1-2z^{-1} \cos \omega_0 + z^{-2}}$	$ z > 1$
8	$\sin(\omega_0 n) u(n)$	$\frac{z^{-1} \sin \omega_0}{1-2z^{-1} \cos \omega_0 + z^{-2}}$	$ z > 1$
9	$a^n \cos(\omega_0 n) u(n)$	$\frac{1-az^{-1} \cos \omega_0}{1-2az^{-1} \cos \omega_0 + a^2 z^{-2}}$	$ z > a $
10	$a^n \sin(\omega_0 n) u(n)$	$\frac{1-az^{-1} \sin \omega_0}{1-2az^{-1} \cos \omega_0 + a^2 z^{-2}}$	$ z > a $

