

EDISP (LTIlect)  
(English) Digital Signal Processing  
DT systems, LTI

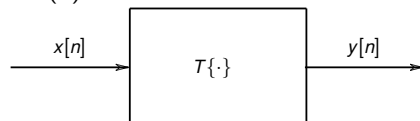
November 26, 2015

# DT systems

A DT system: an operator mapping an input sequence  $x[n]$  into an output sequence  $y[n]$ .

$$y[n] = T\{x[n]\}$$

→ A rule (formula) for computing  $y(n)$  from  $x(n)$



## Implementations:

- PC program
- matlab m-file
- custom VLSI or FPGA
- programmable digital signal processor

Examples:

$$y(n) = 3 \cdot x(n)$$

$$y(n) = \frac{x(n) + x(n-1)}{2}$$

$$y(n) = \frac{1}{M} \sum_{k=0}^{M-1} x(n-k)$$

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

$$y(n) = x(n)^2$$

## Linearity property

$$T\{\alpha_1 x_1[n] + \alpha_2 x_2[n]\} = \alpha_1 T\{x_1[n]\} + \alpha_2 T\{x_2[n]\}$$

in other words:

if

$$x_1[n] \longrightarrow y_1[n]$$

$$x_2[n] \longrightarrow y_2[n]$$

then

$$\alpha x_1[n] \longrightarrow \alpha y_1[n] \quad (\text{scaling, homogeneity})$$

$$x_1[n] + x_2[n] \longrightarrow y_1[n] + y_2[n] \quad (\text{additivity})$$

## Time invariance (shift invariance)

If

$$T\{x[n]\} = y[n]$$

then

$$\forall n_0, T\{x[n - n_0]\} = y[n - n_0]$$

Shift does not modify result  $\leftrightarrow$  System properties do not change

- $y(n) = 3 \cdot x(n)$  – is linear; it is also *memoryless*
- $y(n) = \frac{x(n)+x(n-1)}{2}$  (not memoryless):

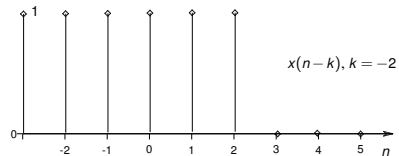
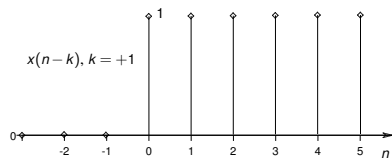
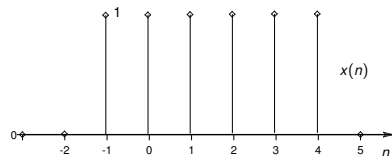
$$\begin{aligned} T\{\alpha_1 x_1(n) + \alpha_2 x_2(n)\} &= \frac{[\alpha_1 x_1(n) + \alpha_2 x_2(n)] + [\alpha_1 x_1(n-1) + \alpha_2 x_2(n-1)]}{2} = \\ &= \frac{\alpha_1 x_1(n) + \alpha_1 x_1(n-1)}{2} + \frac{\alpha_2 x_2(n) + \alpha_2 x_2(n-1)}{2} = \\ &= \alpha_1 \frac{x_1(n) + x_1(n-1)}{2} + \alpha_2 \frac{x_2(n) + x_2(n-1)}{2} = \\ &= \alpha_1 y_1(n) + \alpha_2 y_2(n) \quad \text{cnd} \end{aligned}$$

**not L**  $y(n) = (x(n))^2$  because

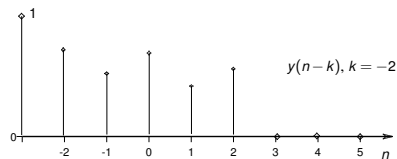
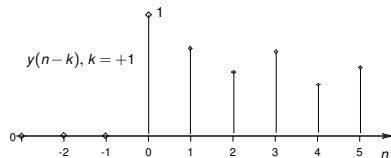
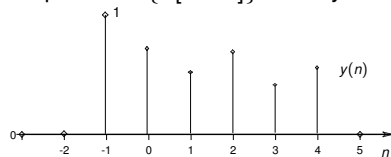
$$T\{x_1(n) + x_2(n)\} = (x_1(n) + x_2(n))^2 = (x_1(n))^2 + (x_2(n))^2 + [2 \cdot x_1(n)x_2(n)]$$

# shift example

Input signals  $x[n-k]$ .

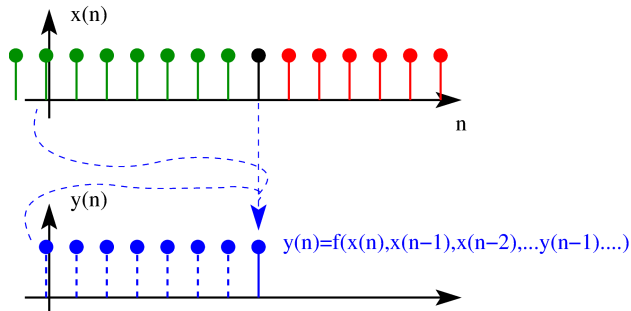


Responses  $T\{x[n-k]\}$  of TI system  $T\{.\}$



# Other properties: **causality**, stability

## causality

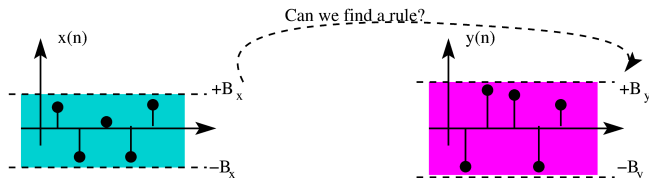


→  $y(n_0)$  depends only on  $x(n)$ ,  $n \leq n_0$  ( *important in real-time implementations, unimportant for off-line processing* )

## stability

→ bounded input causes bounded output [BIBO]

bounded →  $\exists B_x : \forall n |x(n)| \leq B_x < \infty$



## Decimator (compressor)

$$y(n) = x(Mn)$$

→ L, but not TI (*prove it!*)

## 1-st order difference

forward:  $y(n) = x(n+1) - x(n)$  → noncausal

backward:  $y(n) = x(n) - x(n-1)$  → causal

## Accumulator

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

→ unstable; (*hint: feed it with  $u[n]$* )



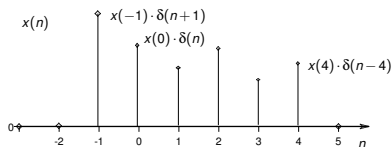
# LTI systems: impulse response

$h[n] = T\{\delta[n]\} \longrightarrow$  impulse response of  $T\{.\}$

$h[n]$  characterizes completely system  $T\{.\}$  – we may compute its response for any input  $x[n]$ .

- Decompose  $x[n]$  into weighted sum of impulses  $\delta[n-k]$

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

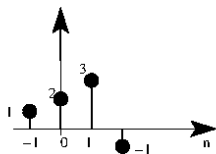


- Superpose responses (use LTI properties)

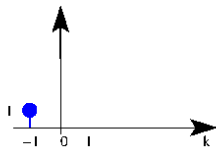
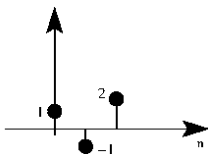
$$y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k]$$

$\longrightarrow$  this is a **convolution sum**

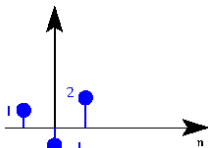
# Convolution example



\*

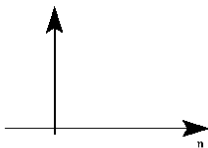


$k=-1$

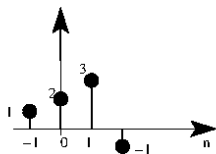


1  
2  
3  
-1

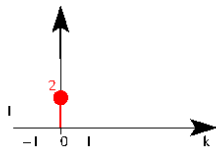
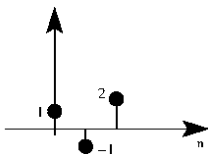
1	-1	2
---	----	---



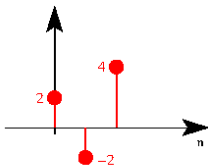
# Convolution example



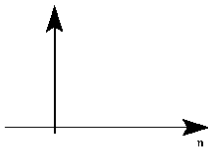
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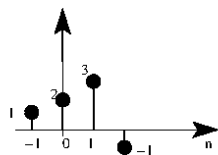
$k=0$



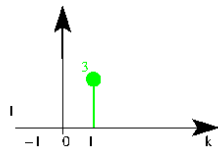
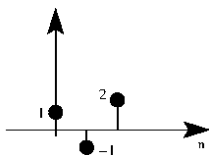
2 -2 4



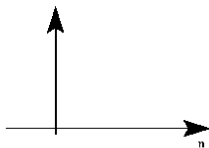
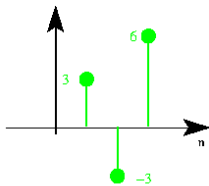
# Convolution example



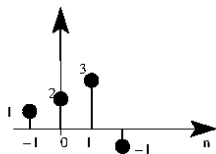
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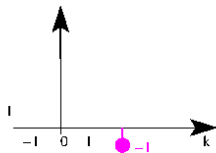
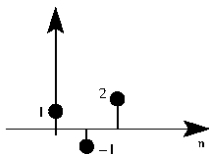
$k=1$



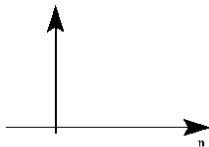
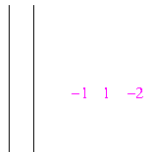
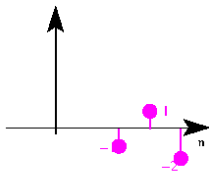
# Convolution example



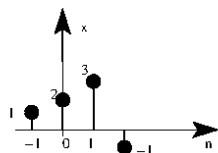
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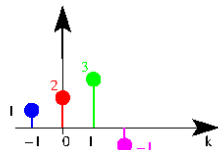
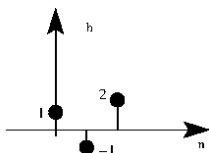
$k=2$



# Convolution example



\*



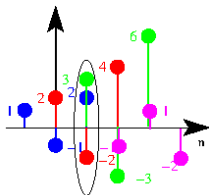
$k=-1$

$k=0$

$k=1$

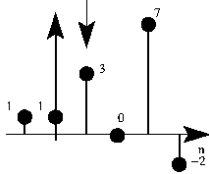
$k=2$

$x(k)h(n-k)$



1	1	-1	2			
2		2	-2	4		
3			3	-3	6	
-1					-1	1
						-2
	1	1	3	0	7	-2

$y(n) = x(n) * h(n)$



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

we denote as

$$y[n] = x[n] * h[n]$$

## Properties of “\*”

“\*” is commutative:  $x[n] * h[n] = h[n] * x[n]$

“\*” distributes over addition  $x[n] * (h_1[n] + h_2[n]) = x[n] * h_1[n] + x[n] * (h_2[n])$

→ Convolution of a signal  $x[n]$  with given fixed  $h[n]$  is linear!

## System and $h[n]$

- causality  $\Leftrightarrow h[n] = 0, n < 0$ . A hint:  $y(n) = \sum_{k=-\infty}^{\infty} h(k)x(n-k)$
- stability  $\Leftrightarrow S = \sum_{k=-\infty}^{\infty} |h(k)| < \infty$

# Linear difference equations

... describe an important class of LTI systems.

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

or

$$\begin{aligned} y(n) &= -a_1 \cdot y(n-1) - a_2 \cdot y(n-2) - \dots - a_n \cdot y(n-N) + \\ &+ b_0 \cdot x(n) + b_1 \cdot x(n-1) + b_2 \cdot x(n-2) + \dots + b_n \cdot x(n-M) \end{aligned}$$

**Note:** if, for a given input  $x_p[n]$ , an output sequence  $y_p[n]$  satisfies given difference equation,

$$y[n] = y_p[n] + y_h[n]$$

will also satisfy the equation, if  $y_h[n]$  is a solution to  $\sum_{k=0}^N a_k y(n-k) = 0$  (homogenous equation).



# Difference equation – example

An equation:  $y(n) = a \cdot y(n-1) + x(n)$

with input

$$x(n) = 0, n < 0$$

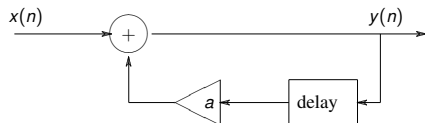
$$x(n) \neq 0, n > 0.$$

$$y(0) = a \cdot y(-1) + x(0)$$

$$y(1) = a \cdot y(0) + x(1)$$

$$y(2) = a \cdot y(1) + x(2)$$

...



Initial condition:  $y(-1) = \alpha$

Let  $x[n] = \delta[n]$

$$y(0) = a \cdot \alpha + 1$$

$$y(1) = a(a \cdot \alpha + 1) = a^2 \alpha + a$$

$$y(2) = a^3 \alpha + a^2$$

...

$$y(n) = a^{n+1} \alpha + a^n$$

# Difference equation – impulse response (example continued)

$$y(n] = a \cdot y[n-1] + x[n]$$

Initial condition:  $y[-1] = \alpha$

$$x[n] = \delta[n]$$

Solution:  $y[n] = a^{n+1} \alpha + a^n$

Find a homogenous part!

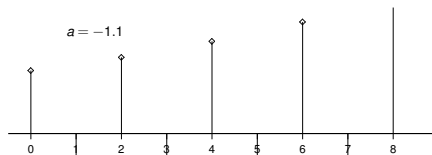
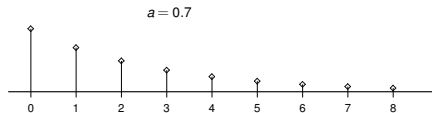
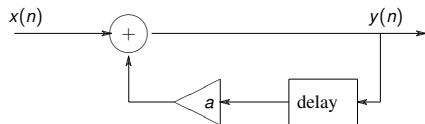
Stability:

$$1 < a: a^n \rightarrow \infty$$

$$0 < a < 1: a^n \rightarrow 0$$

$$-1 < a < 0: a^n \rightarrow 0$$

$$a < -1: a^n \rightarrow ???$$



# Z-transform – what and why

- DTFT – a transform based on periodic decomposition (basis:  $e^{jn\theta}$  sequences)
- responses of most LTI systems – made of short sequences (FIR) and decaying complex exponentials  $r^n e^{jn\theta}$  (IIR)
- $r^n e^{jn\theta} = (re^{j\theta})^n = z^n$  can be a good basis

Z-transform is a tool for analyzing transient signals, such as an impulse response of a system.

# Z-transform definition

$\mathcal{Z}$  – a generalization of DTFT, similar to  $\mathcal{L}$  as a generalization of CTFT

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

→ DTFT is equal to  $X(z)$  at unit circle  $z = e^{j\theta}$

**Convergence:** same as for DTFT of  $x[n] \cdot r^{-n}$  (substitute  $z = r \cdot e^{j\theta}$ )

$$\sum_{n=-\infty}^{\infty} |x(n)r^{-n}| < \infty$$

example:  $u[n]$  is not absolutely summable;  $u(n) \cdot r^{-n}$  can be, if  $|r^{-1}| < 1$

→  $\mathcal{Z}(u[n])$  is convergent for  $r > 1$ .

# Properties:

- Linearity (*each student will recite the formula at 4 a.m.*),
- shift  $x(n - n_0) \xleftrightarrow{Z} z^{-n_0} \cdot X(z)$ ,
- multiplication  $z_0^n \cdot x(n) \xleftrightarrow{Z} X(z/z_0)$ ,
- transform differentiation  $nx(n) \xleftrightarrow{Z} -z dX(z)/dz$ ,
- conjugation  $x^*(n) \xleftrightarrow{Z} X^*(z^*)$ ,
- time reversal  $x(-n) \xleftrightarrow{Z} X(1/z)$ ,
- initial value  $x(0) = \lim_{z \rightarrow \infty} X(z)$  if  $x(n) = 0$  for  $n < 0$  (*hint: limit of each term ...*)
- multiplication  $x_1(n) \cdot x_2(n) \xleftrightarrow{Z} 1/(2\pi j) \oint_C X_1(v) X_2(z/v) v^{-1} dv$  **complex!**  
convolution

What happens to ROC with above transformations?

# Examples

- $x(n) = a^n u(n)$  (causal)

$$X(z) = \sum_{n=-\infty}^{\infty} a^n u(n) z^{-n} = \sum_{n=0}^{\infty} (az^{-1})^n = \frac{1}{1 - az^{-1}} = \frac{z}{z - a}, \quad |z| > |a|$$

- $x(n) = -a^n u(-n-1)$  (non-causal)

$$X(z) = - \sum_{n=-\infty}^{-1} (az^{-1})^n = 1 - \sum_{n=0}^{\infty} (a^{-1}z)^n = \frac{1}{1 - az^{-1}} = \frac{z}{z - a}, \quad |z| < |a|$$

- $x(n) = \begin{cases} a^n & n = 0, 1, \dots, N-1 \\ 0 & \text{otherwise} \end{cases}$  (finite)

$$X(z) = \sum_{n=0}^{N-1} a^n z^{-n} = \sum_{n=0}^{N-1} (az^{-1})^n = \frac{1 - (az^{-1})^N}{1 - (az^{-1})} = \frac{1}{z^{(N-1)}} \frac{z^N - a^N}{z - a}$$

# Z-transform pairs

Remember about ROC (Region of Convergence):

- causal signal (non-zero only when  $n \geq 0$ ) – outside unit circle
- anticausal signal (non-zero only when  $n \leq 0$ ) – inside unit circle

In the table only causal prototypes are shown.

$$\begin{aligned}\delta(n) & \quad \text{---} \quad 1 \\ u(n) & \quad \text{---} \quad \frac{1}{1-z^{-1}} \\ a^n u(n) & \quad \text{---} \quad \frac{1}{1-az^{-1}} \\ n \cdot a^n u(n) & \quad \text{---} \quad \frac{az^{-1}}{(1-az^{-1})^2}\end{aligned}$$

# Inverse $\mathcal{Z}$ - transform

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

- Power series expansion (e.g for finite series)  $\rightarrow$  “a series of deltas”
- Partial fraction expansion:  $X(z)$  a rational function with  $M$  zeros and  $N$  distinct poles  $\rightarrow X(z) = \text{num}(z)/\text{den}(z) = \text{quot}(z) + \text{rem}(z)/\text{den}(z)$

$$X(z) = \sum_{r=0}^{M-N} B_r \cdot z^{-r} + \sum_{k=1}^N \frac{A_k}{1 - d_k z^{-1}}, \quad A_k = (1 - d_k z^{-1}) \cdot X(z) \Big|_{z=d_k}$$

Are we looking for a

- causal (or maybe only right-sided)
- anticausal (... left-sided)
- noncausal (... two-sided?)

solution?

Remark: *these terms are overloaded – understand them as a short for “signal that may be an imp. response of a causal filter” etc.*



# $H(z)$ to $h(n)$ (or how to find $\mathcal{Z}^{-1}$ )

$$H(z) = \frac{Y(z)}{X(z)} = \frac{\sum_{k=0}^2 b_k z^{-k}}{\sum_{m=0}^2 a_m z^{-m}} = A \frac{\prod_{k=0}^2 (1 - c_k z^{-1})}{\prod_{m=0}^2 (1 - d_m z^{-1})}$$

Zeros at  $z = c_k \rightarrow (1 - c_k z^{-1}) = \frac{z - c_k}{z - 0}$  (plus pole at  $z = 0$ ).

Poles at  $z = d_m \rightarrow \frac{1}{(1 - d_m z^{-1})} = \frac{z - 0}{z - d_m}$  (plus a zero at  $z = 0$ ).

Knowing poles  $\rightarrow$  decompose into a polynomial + partial fractions:

$$H(z) = \sum_{r=0}^{M-N} B_r \cdot z^{-r} + \sum_{m=1}^N \frac{A_m}{1 - d_m z^{-1}}$$
$$h(n) = \sum_{r=0}^{M-N} B_r \cdot \delta(n - r) + \sum_{m=1}^N A_m u(n) (d_m)^n$$

So, two conjugate poles (at  $d_m$  and  $d_m^*$ ) give a decaying cosine!

# Z-transform of a convolution

$$y[n] = x[n] * h[n] \longrightarrow y(n) = \sum_{k=-\infty}^{+\infty} x(k) \cdot h(n-k)$$

$$Y(z) = \sum_{n=-\infty}^{+\infty} \left[ \sum_{k=-\infty}^{+\infty} x(k) \cdot h(n-k) \right] z^{-n} =$$

$$= \sum_{k=-\infty}^{+\infty} \left[ x(k) \sum_{n=-\infty}^{+\infty} h(n-k) z^{-n} \right] =$$

(we substitute  $m = n - k$  so  $n = k - m$ )

$$= \sum_{k=-\infty}^{+\infty} \left[ x(k) \sum_{m=-\infty}^{+\infty} h(m) z^{-k-m} \right] =$$

$$= \sum_{k=-\infty}^{+\infty} x(k) z^{-k} \sum_{m=-\infty}^{+\infty} h(m) z^{-m}$$

$$Y(z) = X(z) \cdot H(z)$$

And this is the main application of z-transform.

# Z-transform and difference equations (1)

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

$a_0 = 1$  traditionally

Simpler case of  $N = 0$  (FIR, no recursion)

$$\begin{aligned} y(n) &= \sum_{k=0}^M b_k x(n-k) \\ Y(z) &= \sum_{k=0}^M b_k \mathcal{Z}[x(n-k)] = \\ &= \sum_{k=0}^M b_k X(z) z^{-k} = \\ &= X(z) \cdot \sum_{k=0}^M b_k z^{-k} = \\ &= X(z) \cdot H(z) \end{aligned}$$

## Z-transform and difference equations (2)

Now the general case:

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

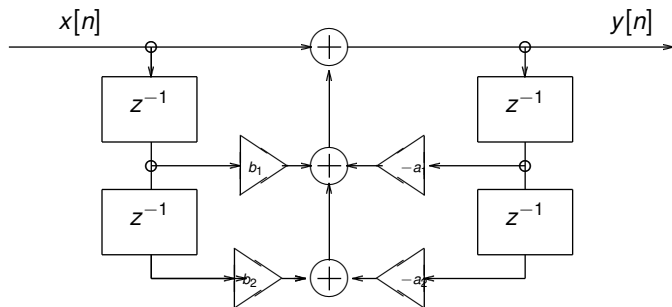
$$\sum_{k=0}^N a_k Y(z) z^{-k} = \sum_{k=0}^M b_k X(z) z^{-k}$$

$$Y(z) \sum_{k=0}^N a_k z^{-k} = X(z) \sum_{k=0}^M b_k z^{-k}$$

$$Y(z) = X(z) \cdot \frac{\sum_{k=0}^M b_k z^{-k}}{\sum_{k=0}^N a_k z^{-k}}$$

Recall that the transform is linear, and time shift by  $k$  is represented by  $z^{-k}$  operator.

# System and its difference equation



$$y(n) = x(n) + b_1x(n-1) + b_2x(n-2) - a_1y(n-1) - a_2y(n-2)$$

$$y(n) + a_1y(n-1) + a_2y(n-2) = x(n) + b_1x(n-1) + b_2x(n-2)$$

$$\sum_{m=0}^2 a_m y(n-m) = \sum_{k=0}^2 b_k x(n-k)$$

# Difference equation and $H(z)$

$z^{-1}$  - time shift (delay) operator

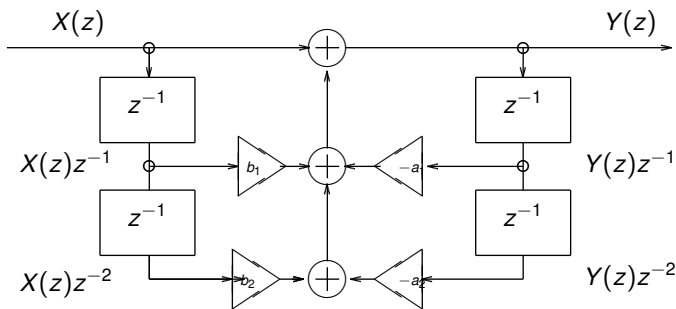
$$\sum_{m=0}^2 a_m y(n-m) = \sum_{k=0}^2 b_k x(n-k)$$

$$\sum_{m=0}^2 a_m Y(z) z^{-m} = \sum_{k=0}^2 b_k X(z) z^{-k}$$

$$Y(z) \sum_{m=0}^2 a_m z^{-m} = X(z) \sum_{k=0}^2 b_k z^{-k}$$

$$H(z) = \frac{Y(z)}{X(z)} = \frac{\sum_{k=0}^2 b_k z^{-k}}{\sum_{m=0}^2 a_m z^{-m}}$$

# System and its $H(z)$



$$Y(z) = X(z) + b_1 X(z)z^{-1} + b_2 X(z)z^{-2} - a_1 Y(z)z^{-1} - a_2 Y(z)z^{-2}$$
$$Y(z) + a_1 Y(z)z^{-1} + a_2 Y(z)z^{-2} = X(z) + b_1 X(z)z^{-1} + b_2 X(z)z^{-2}$$

$$\sum_{m=0}^2 a_m Y(z)z^{-m} = \sum_{k=0}^2 b_k X(z)z^{-k}$$

$$H(z) = \frac{Y(z)}{X(z)} = \frac{\sum_{k=0}^2 b_k z^{-k}}{\sum_{m=0}^2 a_m z^{-m}}$$

# System defined by $H(z)$ + complex sinusoid

$$x(n) = e^{jn\theta} \longrightarrow \boxed{h(n)} \longrightarrow y(n) = ?$$

$$\begin{aligned} y(n) &= \sum_k h(k) \cdot e^{j(n-k)\theta} = \\ &= \sum_k h(k) \cdot e^{j(-k)\theta} \cdot e^{jn\theta} = \\ &= e^{jn\theta} \sum_k h(k) \cdot e^{j(-k)\theta} = \\ &= e^{jn\theta} H(e^{j\theta}) \\ H(e^{j\theta}) &= A(\theta)e^{j\phi(\theta)} \end{aligned}$$

$A(\theta)$  - magnitude,  $\phi(\theta)$  - phase of  $H(e^{j\theta})$

If  $x(n)$  is periodic - we can decompose it into harmonics (linearity).

As you can see, when we deal with periodic signals, we use  $H(e^{j\theta}) = H(z)|_{z=e^{j\theta}}$



# System defined by $H(z)$ + sine/cosine signal

We say  $H(z)$  meaning “transfer function”, but we immediately substitute  $z = e^{j\theta}$  ....

$$x(n) = e^{jn\theta} \longrightarrow \boxed{h(n)} \longrightarrow y(n) = e^{jn\theta} H(e^{j\theta})$$

so if  $x(n) = \cos(n\theta) = 1/2 \cdot (e^{jn\theta} + e^{-jn\theta})$   
then  $y(n) = 1/2 \cdot (H(e^{j\theta})e^{jn\theta} + H(e^{-j\theta})e^{-jn\theta})$

$$H(e^{j\theta}) = A(\theta)e^{j\phi(\theta)}$$

and  $y(n) = A(\theta) \cdot 1/2 \cdot (e^{jn\theta} \cdot e^{j\phi(\theta)} + e^{-jn\theta} \cdot e^{-j\phi(\theta)})$

(for a real  $h(n)$   $\phi(\theta)$  is odd:  $\phi(-\theta) = -\phi(\theta)$ )

and  $y(n) = A(\theta) \cdot 1/2 \cdot (e^{j(n\theta+\phi(\theta))} + e^{-j(n\theta+\phi(\theta))})$

$$y(n) = A(\theta) \cdot \cos(n\theta + \phi(\theta))$$

Repeat the same with  $\sin()$   $\longrightarrow$  at home.

Example:  $x(n) = 3 + 5\sin(0.1\pi n) \longrightarrow$  a DC component and a  $0.1\pi$  sinusoidal signal.

So  $y(n) = A(0) \cdot 3 + A(0.1\pi) \cdot 5\sin(0.1\pi n + \phi(0.1\pi))$ .

Note: With periodic signals use Fourier and **NOT** Z-transform !!!