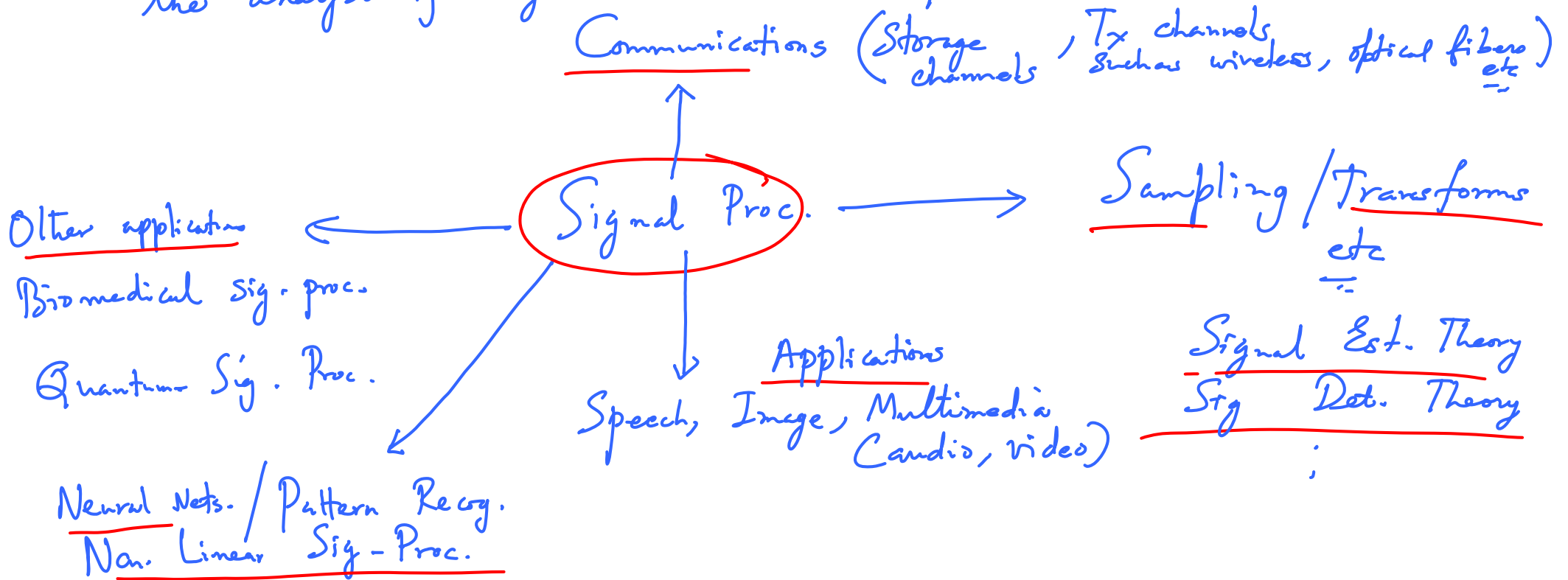


# Mathematical Methods and Techniques in Sig. Proc.

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- Pre requisites:
- 1) U. G. Course in signals & systems  
or a basic DSP course.
  - 2) Familiarity in linear algebra / probability &  
random processes  
etc  
..

Signal Proc is an area of applied math dealing with the analysis of signals in discrete / cont. time.



## Mathematical Tools / Techniques used in S-P.

- 1) Transform Theory
- 2) Prob. & Stochastic Processes
- 3) Calculus / Analysis / Functional Analysis
- 4) Linear Algebra
- 5) Numerical methods / Approx. Theory.
- 6) Optimization
- 7) Stat. Decision Theory (req. A-6)

# 1. D signals & systems

## Basic Sequences

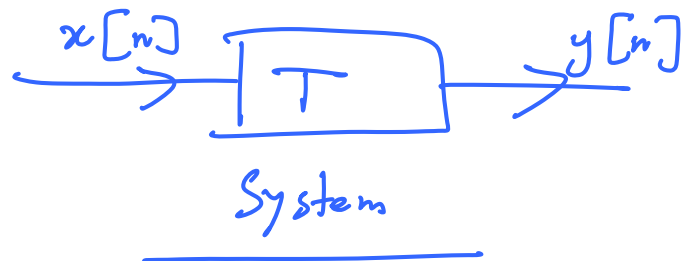
$$\delta[n] = \begin{cases} 1 & n = 0 \\ 0 & \text{else} \end{cases}$$

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

$$x(n) = a^n u(n) \quad |a| < 1$$

exp. decaying seq.

# Systems



## Examples

1) Delay system:

$x[n]$   $\xrightarrow{\text{Delay by } k \text{ units}}$

$x[n-k]$

2) M. A (Moving Average)

$$y[n] = \frac{1}{M_1 + M_2 + 1} \sum_{k=-M_1}^{M_2} x[n-k]$$

## Memory less / Memory Systems

O/p  $y[n]$  depends only on the i/p  $x[n]$

Ex:  $y[n] = x^3[n]$

$y[n] = x^2[n] + 2x[n]$  & so on

} Memory less

$y[n] = x^3[n] + x^2[n-1] + x[n-2]$

## Linear / Non linear Systems

All linear systems abide by "superposition" principle

$$T(x_1(n) + x_2(n)) = \underbrace{T(x_1(n)) + T(x_2(n))}_{y_1(n) + y_2(n)} \left. \vphantom{T(x_1(n) + x_2(n))} \right\} \text{Additivity} \quad \textcircled{P_1}$$

$$T(ax(n)) = a T(x(n)) \quad \left( \text{Scaling / Homogeneity} \right) \quad \textcircled{P_2}$$

$\textcircled{P_1}$  &  $\textcircled{P_2}$  imply

$$T(a_1 x_1(n) + a_2 x_2(n)) = a_1 T(x_1(n)) + a_2 T(x_2(n))$$

This is the 'superposition' rule

## Time Invariance

Suppose an i/p sequence  $x[n]$  is delayed by  $n_0$

$$x_1[n] = x[n - n_0]$$

If the o/p sequence  $y[n]$  is also delayed by  $n_0$

i.e.)  $y_1[n] = y[n - n_0]$

$$x[n] \leftrightarrow y[n]$$

then it is called 'shift invariance'

Exercise: Suppose  $y[n] = x[Ln]$

Is this shift invariant?

$L$  is any <sup>+</sup>ve integer



## Causality

A system is causal if for every choice of  $n_0$ , the o/p sequence @ time  $n = n_0$  depends only on the i/p sequence for  $n \leq n_0$

Examples :

$$y(n) = \underbrace{x(n+1)}_{\text{sample in the future}} - x(n) \quad (\text{ANTI CAUSAL!})$$
$$y(n) = x(n) - x(n-1) \quad (\text{CAUSAL!})$$

'Causality' is a powerful idea!

## Stability

A system is (BIBO) bounded i/p bounded o/p stable iff every bounded i/p sequence produces a bounded o/p sequence

I/p is bounded if  $|x(n)| \leq B_x < \infty \quad \forall n$

BIBO stability requires  $|y(n)| \leq B_y < \infty \quad \forall n$

Exercise:  $y(n) = \sum_{k=-\infty}^n u(k) = \begin{cases} 0 & n < 0 \\ (n+1) & n \geq 0 \end{cases}$

Examine if  $y(n)$  is bounded?

Answer: Unbounded!

## Linear & time invariant systems

Suppose  $h_k(n)$  denotes the response of a system to an impulse @  $n=k$  i.e.,  $\delta(n-k)$

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

Since the system is linear, by superposition,

$$y(n) = T(x(n)) = T\left(\sum_{k=-\infty}^{\infty} x(k) \delta(n-k)\right)$$
$$= \sum_{k=-\infty}^{\infty} x(k) T\{\delta(n-k)\}$$

Linearity  
↓

SHIFT INVARIANCE

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

By shift invariance

Response to  $\delta(n-k)$

$$\longrightarrow h(n-k)$$

i.e.,  $h_k(n) = h(n-k)$

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

CONVOLUTION  
OPERATOR

Reflection  $\longrightarrow$  Shift  $\longrightarrow$  Multiply  $\longrightarrow$  Add

## Modes in a linear system

Often, given a sequence of o/p data from a system, one is interested in modeling the signal as the o/p of a linear time invariant system and analyzing the spectral content. 'Spectral content' is linked to "system mode".

Example: Suppose we have a difference eqn given by  $y(t+2) + a_1 y(t+1) + a_2 y(t) = 0$  ①

This is a homogeneous 2<sup>nd</sup> order system b. s. of ①, we get a characteristic eq<sup>n</sup>

Taking z-transform on  $(z^2 + a_1 z + a_2) = 0$   $(\because (z^2 + a_1 z + a_2) y(z) = 0)$

$$z = \frac{-a_1 \pm \sqrt{a_1^2 - 4a_2}}{2}$$

$$p_1 = \frac{-a_1 + \sqrt{a_1^2 - 4a_2}}{2}$$

$$p_2 = \frac{-a_1 - \sqrt{a_1^2 - 4a_2}}{2}$$

$$z^2 + a_1 z + a_2 = (z - p_1)(z - p_2)$$

Case A:  $p_1 \neq p_2$

$$y(t) = c_1 (p_1)^t + c_2 (p_2)^t \quad t \geq 0$$

$c_1$  &  $c_2$  can be determined from the initial conditions

CASE B:  $p_1 = p_2 = p$

$$y(t) = (c_1 + c_2 t) p^t$$

Example : Suppose we have a mixture of 2 sinusoids  
& observations are "noise free".

$$y(t) = a_1 \cos(\omega_1 t) + a_2 \cos(\omega_2 t)$$

We need to determine the mode frequencies.

$$\cos(\omega_i t) = \frac{e^{j\omega_i t} + e^{-j\omega_i t}}{2}$$



Each  $\cos(\omega_i t)$  has 2 modes!

$\Rightarrow$   $y(t)$  is governed by a 4<sup>th</sup> order difference eqn.

Let us set up the recursive eqn

$$y(t) + \sum_{i=1}^4 c_i y(t-i) = 0$$

$$\begin{bmatrix} -y(3) & -y(2) & -y(1) & -y(0) \\ & & \ddots & \\ & & & \\ -y(6) & -y(5) & -y(4) & -y(3) \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ c_3 \\ c_4 \end{bmatrix} = \begin{bmatrix} y(4) \\ y(5) \\ y(6) \\ y(7) \end{bmatrix}$$


Can solve for  $[c_1 \ c_2 \ c_3 \ c_4]^T$

We need 8 measurements for 4 modes!

$$\begin{cases} \because y(4) + c_1 y(3) + c_2 y(2) \\ \quad + c_3 y(1) + c_4 y(0) = 0 \\ \vdots \end{cases}$$



## Home Work

- 1) What are the modes for the linear ramp (noise free)?  
 $\{ 0, 1, 2, \dots \}$  
- 2) Repeat Problem 1 for the signal  
 $\{ 1, \frac{3}{4}, \frac{1}{2}, \frac{5}{16}, \dots \}$
- 3) Develop a simple Matlab model for the generalized mixture of sinusoids. Experiment with your choice of frequencies. Can you determine the amplitudes & the phases from the initial conditions?  
$$y(t) = \sum_{i=1}^n a_i \cos(\omega_i t + \theta_i)$$

## Linear discrete time models

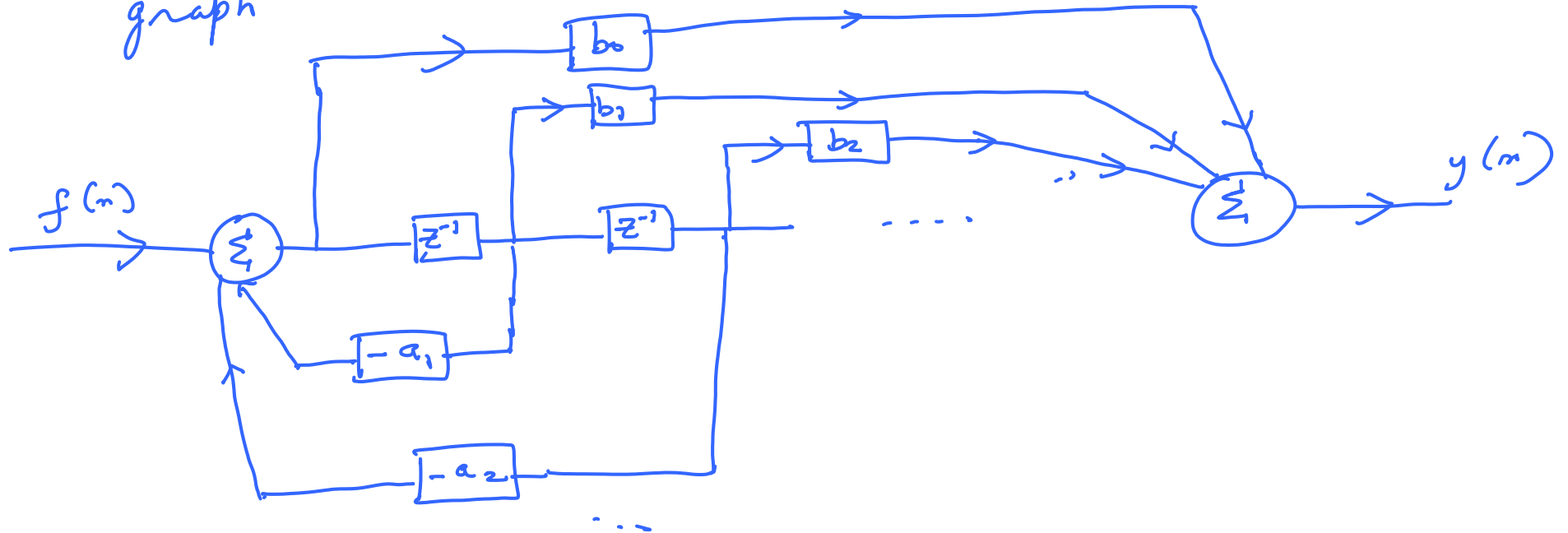
The general equation for a linear discrete time model is (ARMA) given by

$$\sum_{k=0}^{\phi} a_k y(n-k) = \sum_{k=0}^q b_k f(n-k) \quad \text{--- (1)}$$

When  $\phi = 0$ , eqn (1) is a moving average signal since we scale the i/p over a  $(q+1)$  window

When  $q = 0$ , with  $a_0 = 1$ ,  $\phi$   
 $y(n) = b_0 f(n) - \sum_{k=1}^{\phi} a_k y(n-k)$  (AUTO REGRESSIVE MODEL of order ' $\phi$ ')

Let us try to realize the system as a signal flow graph



The above set up is useful for state space representation

## Generalized state space model Linear discrete time models

Consider the linear discrete time model with transfer function for (p=q) case.

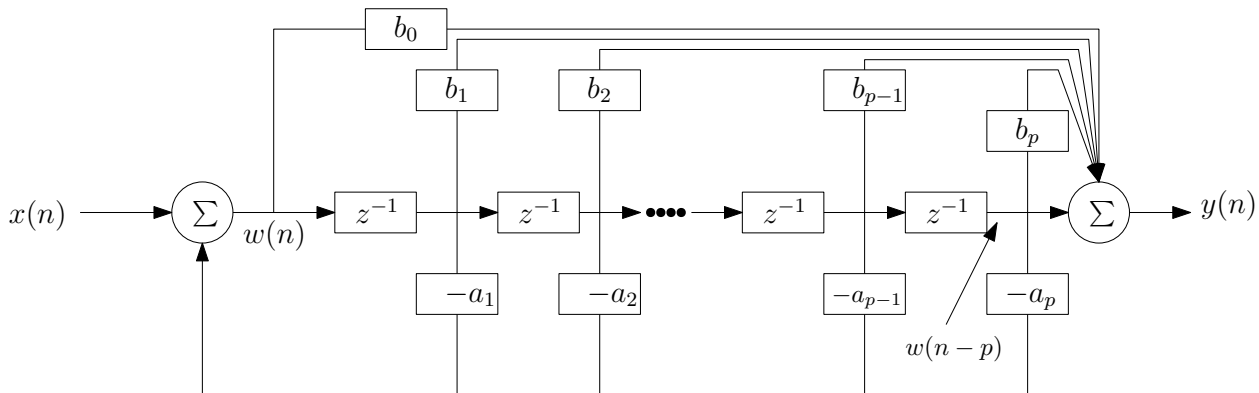
$$H(z) = \frac{\sum_{k=0}^p b_k z^{-k}}{1 + \sum_{k=1}^p a_k z^{-k}} = \frac{Y(z)}{X(z)}$$

Let us define two related transfer functions as follows

$$\frac{Y(z)}{W(z)} = \sum_{k=0}^p b_k z^{-k}$$

$$\frac{W(z)}{X(z)} = \frac{1}{1 + \sum_{k=1}^p a_k z^{-k}}$$

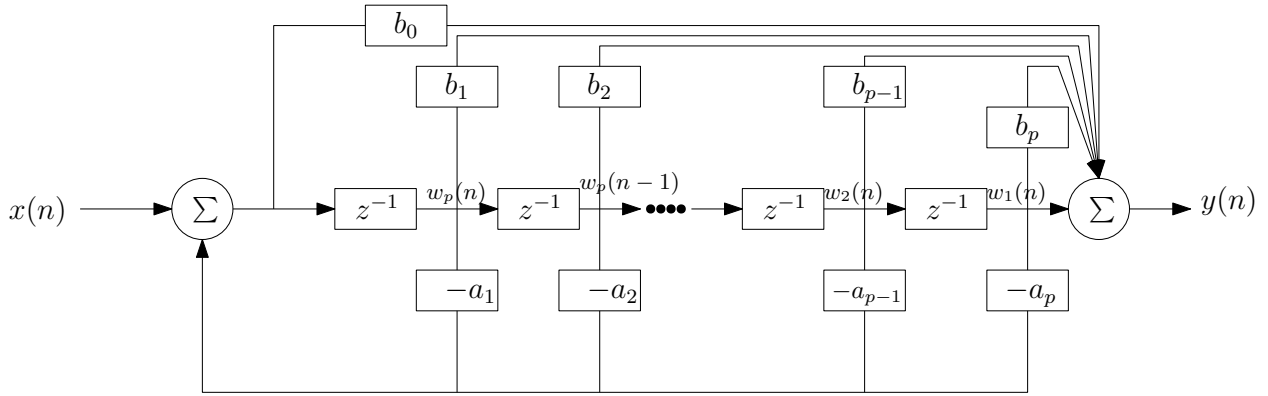
Let us form the signal flow graph for representing transfer functions above.



Define the state variables as follows:

$$\begin{aligned} w_p(n) &= w(n-1) \\ w_{p-1}(n) &= w(n-2) \\ &\vdots \\ w_1(n) &= w(n-p) \end{aligned}$$

As the signal  $w(n)$  passes through the delay line, the state variables  $[w_1(n), \dots, w_p(n)]$  form a vector. The time to space mapping dictates that the signal in time can be transformed to a vector in space. The signal



dynamics can be visualized as a **trajectory** as below.

$$\begin{aligned}
 w_0(n+1) &= w_1(n) \\
 w_1(n+1) &= w_2(n) \\
 &\vdots \\
 w_{p-1}(n+1) &= w_p(n) \\
 w_p(n+1) &= x(n) - a_1 w_p(n) - a_2 w_{p-1}(n) - \dots - a_p w_1(n)
 \end{aligned}$$

Let us form a state vector  $\underline{W}(n) = [w_1(n), \dots, w_p(n)]^T$ . Using this and the above expressions, we have

$$\underline{W}(n+1) = \mathbf{A}\underline{W}(n) + \mathbf{b}x(n)$$

where

$$\mathbf{A} = \begin{bmatrix} 0 & 1 & 0 & 0 & \dots & 0 & 0 \\ 0 & 0 & 1 & 0 & \dots & 0 & 0 \\ \vdots & & & & & & \\ -a_p & -a_{p-1} & -a_{p-2} & -a_{p-3} & \dots & -a_2 & -a_1 \end{bmatrix},$$

$$\mathbf{b} = \underbrace{[0, 0, \dots, 0, 1]^T}_{p \text{ elements}}$$

Similarly, one can do the math for expressing the output  $y(n)$  through a sequence of equations below:

$$\begin{aligned}
 y(n) &= b_0 w(n) + \sum_{k=1}^p b_k w_{p+1-k}(n) \\
 y(n) &= b_0 w_p(n+1) + \sum_{k=1}^p b_k w_{p+1-k}(n) \\
 y(n) &= b_0 [x(n) - a_1 w_p(n) - a_2 w_{p-1}(n) - \dots - a_p w_1(n)] + b_1 w_p(n) + b_2 w_{p-1}(n) + \dots + b_p w_1(n) \\
 y(n) &= \sum_{k=1}^p [b_k - b_0 a_k] w + b_0 x(n) \\
 y(n) &= \mathbf{c}^T \underline{W}(n) + \mathbf{d}x(n)
 \end{aligned}$$

where

$$\mathbf{c} = \begin{bmatrix} b_p - b_0 a_p \\ \vdots \\ b_1 - b_0 a_1 \end{bmatrix},$$
$$\mathbf{d} = b_0.$$

In the time domain,

$$\underline{w}(n) = A^n \underline{w}(0) + \sum_{k=0}^{n-1} A^k \underline{b} X(n-1-k)$$

The o/p is

$$y(n) = \underline{c}^T \underline{w}(n) + d X(n)$$

Exercise: Suppose  $H(z) = \frac{1 + 2z^{-1} + z^{-2}}{1 - 0.75z^{-1} + 0.125z^{-2}}$

and  $x[n] = \left(\frac{1}{2}\right)^n u(n)$

- 1) Obtain the state variable representation & study the o/p response  
(a) mathematically (b) via simulations
- 2) Verify these results by any of your favourite undergrad. methods studied in DSP

## Derivation of the transfer function from state variable representation

$$\left. \begin{aligned} \underline{W}(n+1) &= \underline{A} \underline{W}(n) + \underline{b} X(n) - (a) \\ y(n) &= \underline{c}^T \underline{W}(n) + d X(n) - (b) \end{aligned} \right\}$$

$$z \underline{W}(z) = \underline{A} \underline{W}(z) + \underline{b} X(z) \quad \text{--- (1)}$$

$$Y(z) = \underline{c}^T \underline{W}(z) + d X(z) \quad \text{--- (2)}$$

Rewriting (1),

$$(zI - A) \underline{W}(z) = \underline{b} X(z) \quad \text{--- (3)}$$

Plug (3) in (2)

$$Y(z) = \left( \underline{c}^T (zI - A)^{-1} \underline{b} + d \right) X(z) \quad \text{--- (4)}$$

$$\begin{aligned} \frac{Y(z)}{X(z)} &= H(z) \\ &= \underline{c}^T (zI - A)^{-1} \underline{b} \\ &\quad + d \end{aligned}$$



## Non-unique state representations

For any invertible  $p \times p$  matrix  $T$ , we can form a different state representation

$$\underline{w}(n) = T \underline{z}(n) ; \quad w = Tz$$

State variable representation is 'not' unique.

$$T \underline{z}(n+1) = A T \underline{z}(n) + \underline{b} X(n) \quad \text{--- (i)}$$

$$y(n) = \underline{c}^T T \underline{z}(n) + d X(n) \quad \text{--- (ii)}$$

Re writing (i) in a slightly different way,

$$\underline{z}(n+1) = T^{-1} A T \underline{z}(n) + T^{-1} \underline{b} X(n) \quad \text{--- (iii)}$$

$$y(n) = \underline{c}^T T \underline{z}(n) + d X(n) \quad \text{--- (iv)}$$

$$(\tilde{A}, \tilde{b}, \tilde{c}, \tilde{d}) = (T^{-1} A T, T^{-1} \underline{b}, T^T \underline{c}, d)$$

$$(a, b, c) \longrightarrow (b, c, a)$$

$$\begin{bmatrix} b \\ c \\ a \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix}$$

If we 'permute' the state variables, we get a non unique representation "trivial"

What characterizes the "Similarity" between various state representations?

Eigen values

1. State variable formulation & defn. of a state vector
2. Derive the transfer function.
3. State space model (Analyze the dynamics for various forcing functions ( $i/p/s$ ))
4. State space models need not be unique

## Vector Spaces

A finite dimensional vector may be written as  
 $\underline{x} = [x_1 \ x_2 \ \dots \ x_n]^T$ . The elements are  $x_1 \ x_2 \ \dots \ x_n$

Each of the elements  $\in$  some set such as  $\mathbb{R}$  i.e.,  $x_i \in \mathbb{R}$   
or  $x_i \in \mathbb{F}_2$ . They are the scalars of the vector space.

Definition : A linear vector space  $S$  over a set of scalars  
 $R$  is a collection of objects known as "vectors" together with  
an additive (+) operation and scalar multiplication ( $\cdot$ )  
Satisfying the following properties :

VS1:  $S$  forms a group under addition

(a) For any  $\underline{x}$  and  $\underline{y} \in S$ ,  $\underline{x} + \underline{y} \in S$

(b) There is an identity element in  $S$  denoted by  $\underline{0}$ .

$$\underline{x} + \underline{0} = \underline{0} + \underline{x} = \underline{x}$$

(c) For every element  $\underline{x} \in S$ , there is another element

$$\underline{y} \in S \quad / \quad \underline{x} + \underline{y} = \underline{0}$$

$\underline{y} = -\underline{x}$  is the "additive" inverse

(d) Addition operation is associative. For any  $\underline{x}, \underline{y}, \underline{z} \in S$

$$(\underline{x} + \underline{y}) + \underline{z} = \underline{x} + (\underline{y} + \underline{z})$$

VS<sub>2</sub>: For any  $a$  and  $b \in \mathbb{R}$   $\underline{x}, \underline{y} \in S$

$$a \underline{x} \in S$$

$$a(b \underline{x}) = (ab) \underline{x}$$

$$(a+b) \underline{x} = a \underline{x} + b \underline{x}$$

$$a(\underline{x} + \underline{y}) = a \underline{x} + a \underline{y}$$

VS<sub>3</sub>: There is a multiplicative element (identity)  $1 \in \mathbb{R}$   
Such that  $1 \cdot \underline{x} = \underline{x}$ ,  $0 \in \mathbb{R} / 0 \cdot \underline{x} = \underline{0}$ .

Examples : A most familiar vector space is  $\mathbb{R}^n$ ; set of all  $n$  tuples

$$\underline{x}_1 = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \in \mathbb{R}^3 \quad \underline{x}_2 = \begin{bmatrix} 0 \\ 1 \\ 8 \end{bmatrix} \in \mathbb{R}^3 \quad \dots$$

Other examples :

- 1) Set of  $m \times n$  matrices with real entries.
- 2) Set of all polynomials up to degree  $n$  with real coeffs.

## Infinite dimensional vector spaces

Examples:

- 1) Sequence spaces: Set of all  $\infty$ -long sequences  $\{x_n\}$
- 2) Set of continuous functions defined over the interval  $[a, b]$  etc.



Defn: Let  $S$  be a vector space. If  $V \subset S$  is a subset /  $V$  itself is a vector space, then  $V$  is called a subspace of  $S$ .

Examples: (From codes)

Let  $S = \left\{ \begin{array}{l} (00000) \\ (01001) \\ (10001) \\ (11000) \end{array} \right\}$

+ operation is modulo 2.

$V = \left\{ (00000), (01001) \right\}$  Is  $V$  a subspace of  $S$ ?

'Signals' can be thought of vectors in a signal space.  
The notion of V.S. can be naturally extended to signals.