

* Simulations Using Simulink

Ex 1:

Given $\dot{x} = \begin{bmatrix} -7 & 3 & 3 \\ -6 & 1 & 4 \\ 0 & 1 & -2 \end{bmatrix} x + \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} u$

$$y = \begin{bmatrix} 0 & 1 & 1 \end{bmatrix} x$$

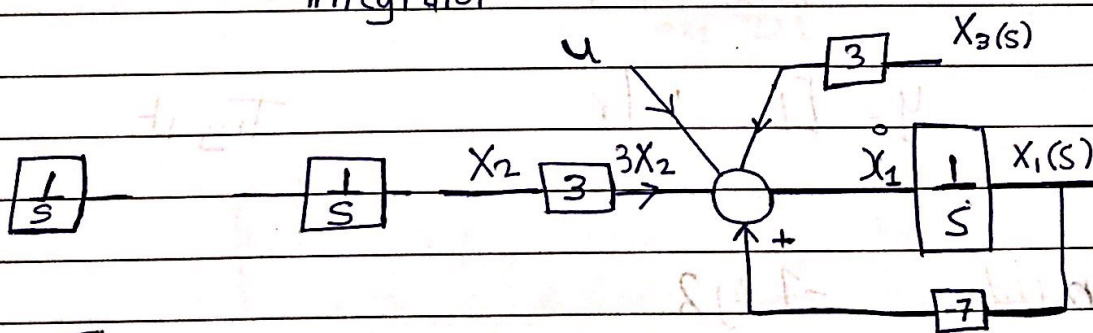
$$\dot{x}_1(t) = -7x_1 + 3x_2 + 3x_3 + u$$

$$sX_1(s) = -7X_1(s) + 3X_2(s) + 3X_3(s) + u(s)$$

$$X_1(s) = \frac{1}{s} [-7X_1(s) + 3X_2(s) + 3X_3(s) + u(s)]$$

↑

integrator

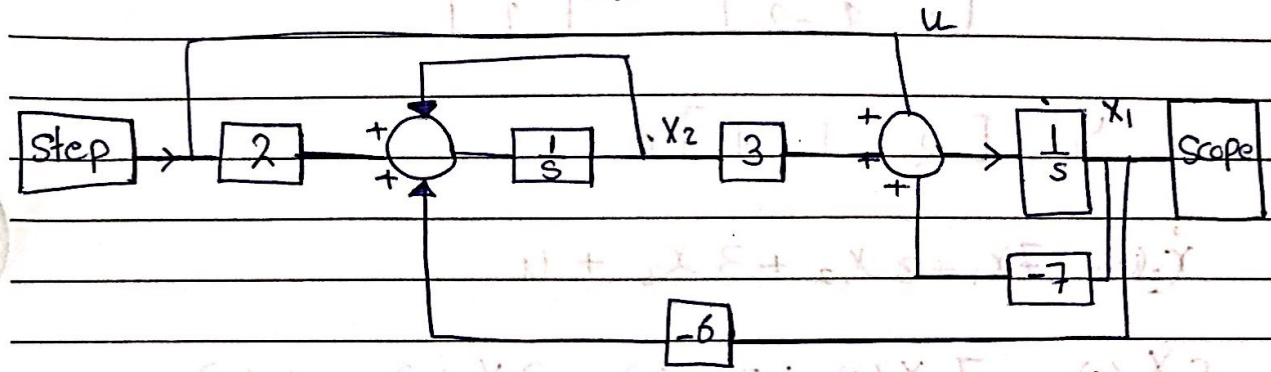
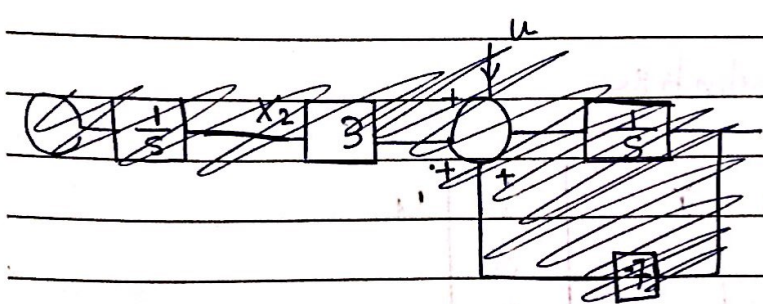


Finish the Simulation at
home.

Ex 2:

$$\dot{x} = \begin{bmatrix} -7 & 3 \\ -6 & 1 \end{bmatrix} x + \begin{bmatrix} 1 \\ 2 \end{bmatrix} u, y = \begin{bmatrix} 1 & 1 \end{bmatrix} x$$

2/2/2017



Ex:3

Given $\dot{x} = \begin{bmatrix} -1 & 2 \\ -2 & -1 \end{bmatrix} x + \begin{bmatrix} 1 \\ 2 \end{bmatrix} u$

$y = \begin{bmatrix} 1 & 1 \end{bmatrix} x$

Try it

eig value $-1 \pm j2$

Example: Simulate using Simulink.

$$\dot{X} = \begin{bmatrix} -7 & 3 & 3 \\ -6 & 1 & 4 \\ 0 & 1 & -2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} + \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}$$

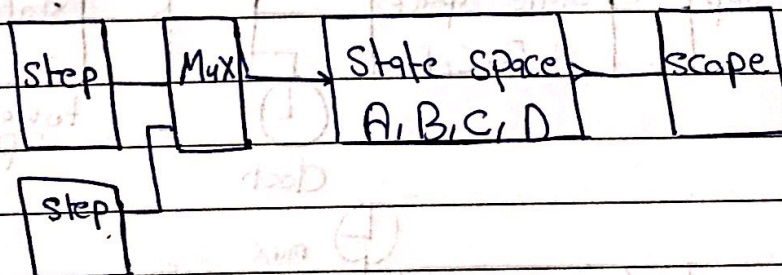
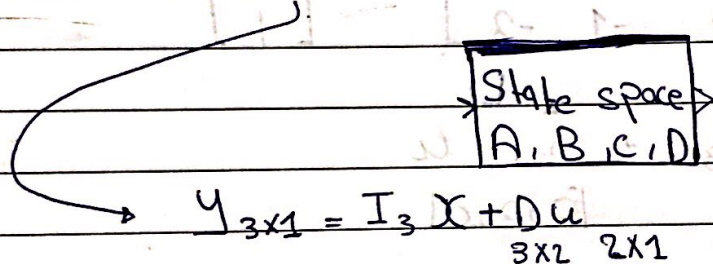
$$y = I_3 x \Rightarrow y_1 = x_1 \Rightarrow C = \text{eye}(3)$$

$$y_2 = x_2 \Rightarrow \text{eye}(\text{size}(A))$$

$$y_3 = x_3$$

$$\Rightarrow A = \begin{bmatrix} -7 & 3 & 3 \\ -6 & 1 & 4 \\ 0 & 1 & -2 \end{bmatrix}, B = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 1 \end{bmatrix}$$

$$\text{و } C = \text{eye}(\text{size}(A)); D = \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix}$$



$$\dot{X} = \begin{bmatrix} -7 & 3 & 3 \\ 0 & 1 & 4 \\ 0 & 1 & -2 \end{bmatrix} \Rightarrow \text{eigen values}$$

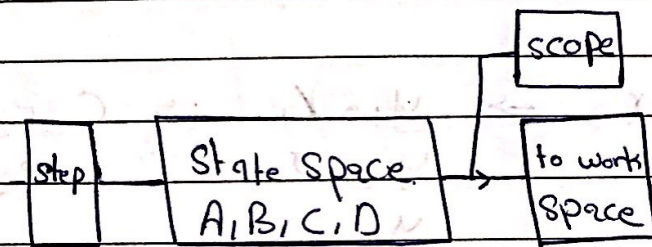
لو كانت صفها صفها
مصفاه الترتيب تكون
مصفاه

$$-7 \neq \lambda \begin{bmatrix} 1 & -4 \\ 1 & -2 \end{bmatrix}$$

check X_{ss}

$$\dot{X} = -A^{-1}B \rightarrow \text{steady state}$$

7/2/2017

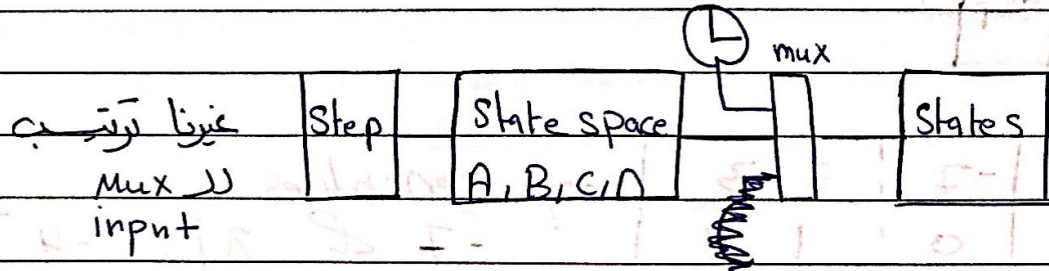
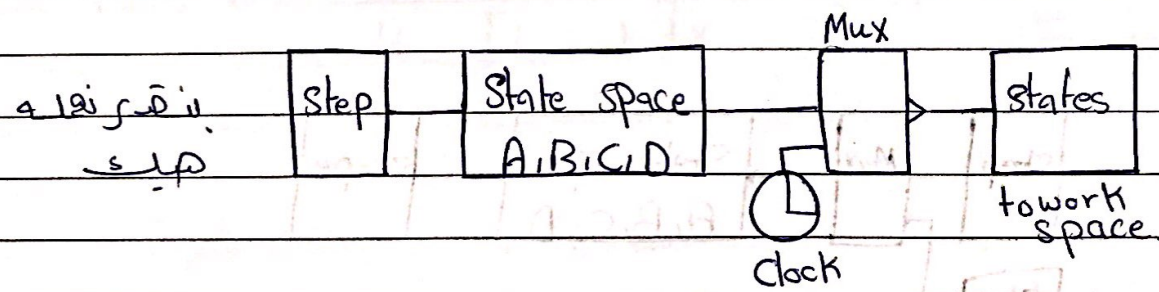


$$\dot{X} = AX + Bu$$

$$\dot{X} = \begin{bmatrix} -5 & 3 & 3 \\ 0 & -2 & 1 \\ 0 & -1 & -2 \end{bmatrix} X + \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} u$$

$$y = I_3 X + \begin{matrix} \times D \\ 0_{3 \times 1} \end{matrix} u$$

$[0 \ 0 \ 0]$



7/2/2017

to draw two signal against each other



$\text{plot}(X(:, 2), X(:, 3))$
 $X_2 \quad X_3$

$\text{plot}(X(:, 3), X(:, 2), X(:, 1))$ Three dimensional

⇒ Needed Review From Control I(441)

« Routh's Stability Criterion »

Having obtained the CE $1 + G(s)H(s) = 0$

⇒ Best illustrated by an Example &c

if the CE = $s^3 + 6s^2 + 11s + 6 + k = 0$

s^3	1	11
s^2	6	$6+k$
s^1	$\frac{60-k}{6}$	0
s^0	$6+k$	

* 6

إذا ضربنا بـ 6

ننقله من المقسوم



هذا المقوم الذي (بهمية)

عشان (Stability)

بهمية ما يكون في تغيير

بالإشارة

For stability, because $1 > 0$ we require all numbers in the first column to be positive.

$$60 - k > 0$$

$$6 + k > 0$$

$$-6 < k < 60$$

Check using matlab.

» $k = 10$; $nc = k$; $dc = [1 \ 6 \ 11 \ 6+k]$ و $sys = tf(nc, dc)$

$Cpoles = roots(dc)$, $step(sys)$

Example: let CE = $s^5 + 2s^4 + 2s^3 + 4s^2 + 3s + 6 = 0$

$$\begin{array}{r}
 s^5 \quad 1 \quad 2 \quad 3 \\
 s^4 \quad 2 \quad 4 \quad 6 \\
 s^3 \quad 0 \quad 0 \quad 0 \\
 s^2 \quad 2 \quad 6 \\
 s^1 \quad -16 \quad 0 \\
 s^0 \quad 6
 \end{array}$$

un
Stable

$$A(s) = 2s^4 + 4s^2 + 6 = 0$$

$$d(A(s)) = 8s^3 + 8s$$

→ system is unstable with two poles of positive real parts

N.B : 2) If the C.E has a difference in sign then the system is unstable $s^5 + 4s^4 + 3s^3 + 2s^2 + s - 1 = 0$

22) If the C.E has a missing power then the system is unstable $s^5 + 3s^3 + 2s^2 + s + 1 = 0$

222) A Second order system is stable whenever the signs of the C.E are the same.

$$\left. \begin{aligned} 10^{100} s^2 + 143s + 5117 &= 0 \\ -10^{100} s^2 - 143s - 5117 &= 0 \end{aligned} \right\} \text{Stable}$$

s^2	a	c
s^1	b	0
s^0	c	

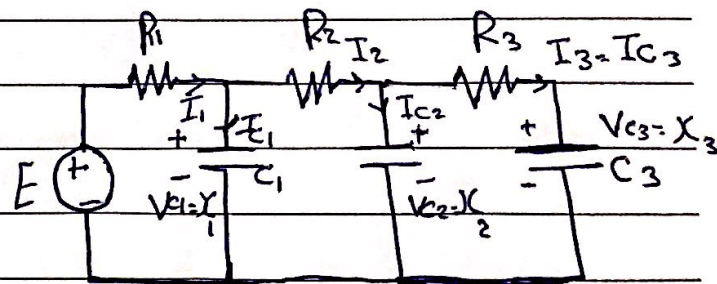
Modeling of systems

Consider the following circuit &.

① \Rightarrow obtain a detailed block diagram representation of the circuit using integrators only, with $E(s)$ as the set value and V_{c3} as output

② \Rightarrow obtain a state space model of the circuit with X_3 output, $E(t)$ input

③ \Rightarrow obtain the transfer function of the circuit using three different methods
↓
in concept



⇒ The Transfer function matrix

Given $\dot{X} = AX + Bu$ $X \in \mathbb{R}^n, u \in \mathbb{R}^m$ $m \leq n$

$y = CX + Du$ $y \in \mathbb{R}^2$ $1 \leq n$

It can be shown

$$G(s) = C [sI_n - A]^{-1} B + D$$

Using matlab so

» $A = [0 \ 1; -2 \ -3]$, $B = [0; 1]$; $C = [5 \ 1]$; $D = 0$

» $\text{syms } s$; $G = C * \text{inv}(s * \text{eye}(\text{size}(A)) - A) * B + D$,
 $\text{pretty}(G) \Rightarrow \frac{5s+1}{s^2+3s+2}$

» $q = \text{panel}(3)$; $\text{size}(q)$

» $G = \text{ss2tf}(A, B, C, D)$

No B given $S(A, B, C, D)$

It can be shown that the steady state value of x due to a unit step is

$$x_{ss} = -A^{-1}B = \lim_{t \rightarrow \infty} x(t)$$

provided the system is stable and $|A| \neq 0$

Example: $\dot{x} = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} x + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u$, $x(0) = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$

either use

$$x(t) = e^{At} x(0) + \int_0^t e^{A(t-\tau)} B u(\tau) d\tau$$

then find $\lim_{t \rightarrow \infty} x(t)$

or use $x_{ss} = -A^{-1}B$

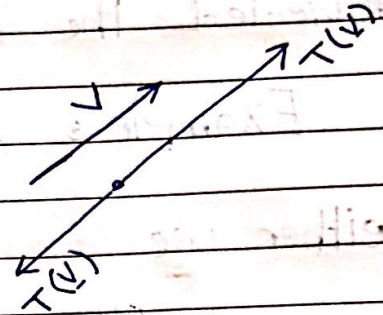
⇒ Review of certain linear Algebra Concepts

* Eigen value and Eigen vector &c

They arise out of the following question Given a transformation $T(\cdot)$ does there exist a non-zero vector \underline{v} such that $T(\underline{v})$ remains in the same direction i.e.

$$T(\underline{v}) = \lambda \underline{v}$$

↑
magnifying
or



⇒ Given the fact that linear transform can be represented by matrices (say A), then

$$T(\underline{v}) = A\underline{v} = \lambda \underline{v}$$

$$\text{or } (A - \lambda I)\underline{v} = 0 \rightarrow (A - \lambda I_n)\underline{v} = 0$$

↑
 $n \times n$
matrix

For \underline{v} not to be zero $A - \lambda I_n$ shouldn't have an inverse i.e. $A - \lambda I_n$ should be singular i.e.

$$|A - \lambda I_n| = 0$$

$$T(V) = \overset{?}{A} \overset{?}{V} = \overset{?}{\lambda} V$$

$|A - \lambda I_n| = 0$ results in an n th order polynomial
is solved to determine λ

To determine the eigenvectors for each λ_i $i=1, \dots, n$
solve

$$[A - \lambda_i I_n] \overset{?}{V}_i = 0 \quad V_i \neq 0$$

Example 80

$$A = \begin{bmatrix} -5 & 3 & 3 \\ -6 & 3 & 4 \\ 0 & 1 & 0 \end{bmatrix}$$

$$|A - \lambda I_n| = \begin{vmatrix} -5-\lambda & 3 & 3 \\ -6 & 3-\lambda & 4 \\ 0 & 1 & -\lambda \end{vmatrix}$$

$$(-1)[4(-5-\lambda) + 18] - \lambda[(-5-\lambda)(3-\lambda) + 18] = 0$$

→ Check it at Home

→ matlab

$$\gg \text{poly}(A) \Rightarrow S^3 + 2S^2 - S - 2 = 0$$

$$\lambda^3 + 2\lambda^2 - \lambda - 2 = 0$$

$$\gg \text{roots}([1 \ 2 \ -1 \ -2])$$

$$\text{given } \lambda_1 = 1, \lambda_2 = -1, \lambda_3 = -2$$

Example 8

$$\gg \text{let } A = \begin{bmatrix} -1 & 2 \\ 3 & 4 \end{bmatrix}$$

 $\gg \text{poly}(A)$
 $\gg \text{roots}([1 \quad -3 \quad -10])$

$$\text{ans} = 5, -2$$

$$A = \begin{bmatrix} -1-\lambda & 2 \\ 3 & 4-\lambda \end{bmatrix}$$

$$(-1-\lambda)(4-\lambda) - 5 = 0$$

$$-4 + \lambda - 4\lambda + \lambda^2 - 5 = 0$$

$$\lambda^2 - 3\lambda - 9 = 0$$

$$\lambda = 5, \lambda = -2$$

$$|A - (\lambda I)| = 0$$

eigen value

$$\begin{bmatrix} 1 & 2 \\ 3 & 6 \end{bmatrix} \begin{bmatrix} V_1 \\ V_2 \end{bmatrix} = 0$$

determinant = 0

$$\begin{bmatrix} 1 & 2 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} V_1 \\ V_2 \end{bmatrix} = 0$$

 V_2 arbitrary let $V_2 = 1, V_1 = -2$

$$\gg [r \quad e] = \text{eig}(A) \rightarrow \begin{matrix} r = \text{vector} \\ e = \text{value} \end{matrix}$$

» For mat rat → rational

- 5 diagonal

$$|A - \lambda I| \Rightarrow A = \begin{bmatrix} -6 & 2 \\ 3 & -1 \end{bmatrix} \times \frac{-1}{2} - r_2$$

$$-6 - 6 = 0$$

$$\begin{bmatrix} 0 & 0 \\ 3 & -1 \end{bmatrix} \begin{bmatrix} V_1 \\ V_2 \end{bmatrix} = 0$$

V_1 arbitrary $V_1 = 1$, $V_2 = 3$

Example 2

$$A = \begin{bmatrix} -5 & 3 & 3 \\ -6 & 3 & 4 \\ 0 & 1 & 0 \end{bmatrix}$$

$$[r \ e] = \text{eig}(A)$$

$$\text{eigen value} = 1, -2, -1$$

$(-5 + 3 + 0)$ diagonal

← as a check

diagonal = eigen values

Sum

« Trace »

» $\det(A)$

$$|A| = 3$$

→ eigen value

Zero \Leftarrow eigen value $0 = |A|$

Example 8. let $A = \begin{bmatrix} -3 & 3 & 3 \\ -6 & 5 & 4 \\ 0 & 1 & 2 \end{bmatrix}$ $\lambda = 3$ $V = \begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix}$

it can be shown that $\lambda_1 = -1$, $\lambda_2 = 3$, $\lambda_3 = 0$

Check: $\sum A_{ii} = \sum \lambda_i = 4$
Trace

$|A| = \prod_{i=1}^n \lambda_i = 0$

To determine eigenvector associated with $\lambda_2 = 3$

$$[A - (3)I_3]V = 0 \Rightarrow \begin{bmatrix} -6 & 3 & 3 \\ -6 & 2 & 4 \\ 0 & 1 & -1 \end{bmatrix} V = 0$$

must be singular

or $|A| = 0$

eigen vector $\neq 0$

(A) or has no inverse

$r_2 - r_1$ $\begin{bmatrix} -6 & 3 & 3 \\ 0 & -1 & 1 \\ 0 & 1 & -1 \end{bmatrix} V = 0$

$r_2 - r_3$ $\begin{bmatrix} -6 & 3 & 3 \\ 0 & -1 & 1 \\ 0 & 0 & 0 \end{bmatrix} V = 0 \Rightarrow \begin{bmatrix} -6 & 3 & 3 \\ 0 & -1 & 1 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} V_1 \\ V_2 \\ V_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$

$$\Rightarrow 0 \cdot V_1 + 0 \cdot V_2 + 0 \cdot V_3 = 0$$

let V_3 be arbitrary say 1

$$\Rightarrow -V_2 + V_3 = 0$$

$$V_2 = V_3 = 1$$

$$\Rightarrow -6V_1 + 3V_2 + 3V_3 = 0$$

$$-6V_1 + 3 \times 1 + 3 \times 1 = 0$$

$$V_1 = 1$$

$$\Rightarrow \text{eigen vector } V = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

associated with

$$\lambda_2 = 3$$

Ex: Determine the two eigen vectors associated with $\lambda_1 = 1, \lambda_3 = 0$

Check the answer using

$[r \ c] = \text{eig}(A)$ To find eigen vector

=> Determination of eigen vector using the Adjoint of a matrix

* Best illustrated by an example

Given $A = \begin{bmatrix} -3 & 3 & 3 \\ -6 & 5 & 4 \\ 0 & 1 & 2 \end{bmatrix}$

determine the eigen vector associated with $\lambda = 1$

$$V_1 = \text{adj}(A - (\lambda)I_n)$$

$$= \text{adj} \begin{pmatrix} + & - & + \\ -4 & 3 & 3 \\ -6 & 4 & 4 \\ 0 & 1 & 1 \end{pmatrix}$$

$$= \begin{bmatrix} 0 & 0 & 0 \\ 6 & -4 & -2 \\ -6 & 4 & +2 \end{bmatrix}$$



eigen vector

المتجه الخاص

بالمورد

العمود

المتجه الخاص

بالمورد

العمود

26/2/2017
⇒ properties of matrices :-

let $\lambda[\cdot]$ defines the eigenvalue of a matrix, $\mathcal{V}[\cdot]$ denote the eigenvectors of a matrix.

$$\textcircled{1} \sum_{i=1}^n \lambda_i[M] = \text{trace}(M) = \sum_{i=1}^n M_{ii}$$

$$\textcircled{2} \prod_{i=1}^n \lambda[M] = |A| \quad A = M$$

$$\textcircled{3} |A^T| = |A|, \quad |AA_2| = |A_1||A_2|$$

$$\textcircled{4} \lambda[A^T] = \lambda[A] \quad ; \quad \mathcal{V}[A] \neq \mathcal{V}[A^T] \text{ are not the same}$$

$$\textcircled{5} \lambda[A_1 A_2] = \lambda[A_2 A_1] \text{ generally } A_1 A_2 \neq A_2 A_1$$

$$\textcircled{6} \text{Trace}(A_1 A_2) = \text{Trace}(A_2 A_1)$$

$$\textcircled{7} \lambda[A + K I_n] = \lambda[A] + K \quad \text{i.e. eigenvalue are shifted by } K$$

$$\textcircled{8} \mathcal{V}[A + K I_n] \text{ are the same as } \mathcal{V}[A]$$

$$\textcircled{9} \text{ if } A = A^T \text{ i.e. } A \text{ is symmetric then } \lambda[A] \text{ are real (cannot be complex)}$$

$$\mathcal{V}[A] = \text{perpendicular i.e. } (x^T \cdot x = 0)$$

$$(10) \lambda[A^{-1}] = \frac{1}{\lambda[A]}$$

inverse $\leftarrow A$ إذا $\lambda \neq 0$
(فقط إذا $\lambda \neq 0$)

$$(11) \lambda[A_1 A_2] = \lambda[A_1] \lambda[A_2]$$

$$(12) \lambda[A_1 + A_2] \neq \lambda[A_1] + \lambda[A_2]$$

\hookrightarrow eigen Value
القيمة الذاتية

matlab Ex: given A_1, A_2 , Validate the previous properties at Home P

$$\Rightarrow A_1 = \begin{bmatrix} -5 & 3 & 3 \\ -6 & 3 & 4 \\ 0 & 1 & 0 \end{bmatrix}; A_2 = \begin{bmatrix} -3 & 3 & 3 \\ -6 & 5 & 4 \\ 0 & 1 & 2 \end{bmatrix}$$

$$\lambda \begin{bmatrix} A_{n \times n} & 0 \\ C & B_{m \times m} \end{bmatrix} = \lambda \begin{bmatrix} A_{n \times n} & D \\ 0 & B_{m \times m} \end{bmatrix} = \lambda[A] \cup \lambda[B],$$

\cup indicates union counting multiplicity

$$\lambda \begin{bmatrix} \text{diagonal} \\ \text{matrix} \end{bmatrix} \text{ or } \lambda \begin{bmatrix} \text{upper-triangular} \\ \text{matrix} \end{bmatrix} \text{ or } \lambda \begin{bmatrix} \text{lower-triangular} \\ \text{matrix} \end{bmatrix}$$

Example 6c are the elements of the diagonal

$$A = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 0 & 5 & 6 & 7 \\ 0 & 0 & 8 & 9 \\ 0 & 0 & 0 & 10 \end{bmatrix}$$

eigen values: 1, 5, 8, 10

$$\lambda[A] = \{1, 5, 8, 10\}$$

$$\lambda[A] = \begin{bmatrix} -1 & 2 & 0 \\ 3 & 4 & 0 \\ 10 & 17 & 34 & 5 \end{bmatrix} = \{5\} \cup \lambda \begin{bmatrix} -1 & 2 \\ 3 & 4 \end{bmatrix}$$

use $|B - \lambda I| = 0$

$$\rightarrow |A| = 5 \times -10 = -50 \quad \{5\} \cup \{-2, 5\}$$

$$= \{-2, 5, 5\}$$

* If $A = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ a_1 & a_2 & a_3 & a_4 \end{bmatrix}$

$$a_1 \neq 0$$

$$|A - \lambda I_n| = \lambda^4 - a_4 \lambda^3 - a_3 \lambda^2 - a_2 \lambda - a_1 = 0$$

Known as characteristic equation (polynomial) C.E

→ The calculation of the eigenvector for this Special Stair Case matrix are

$$\begin{bmatrix} 1 \\ \lambda \\ \lambda^2 \\ \lambda^3 \\ \vdots \\ \lambda^{n-1} \end{bmatrix}$$

→ if the eigen values are identical, say 3 eigen values then the three eigenvectors are

$$\begin{bmatrix} 1 \\ \lambda \\ \lambda^2 \\ \vdots \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 2\lambda \\ \vdots \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ -2 \\ \vdots \end{bmatrix}$$

अदिश

→ Example: calculate the eigen values and eigenvector of

$$\begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -4 & -8 & -5 \end{bmatrix}$$

- (1) using Gauss elimination method
- (2) adjoint method for two eigenvectors
- (3) using the Stair Case property

$\lambda \leq \|A\|$ For any matrix norm

Example: let $A = \begin{bmatrix} -2 & -1 & 0 \\ 1 & -2 & 0 \\ 3 & 4 & 5 \end{bmatrix}$

$$\lambda[A] = \lambda \left[\begin{bmatrix} -2 & -1 \\ 1 & -2 \end{bmatrix} \right] \cup \lambda[5]$$

$$= \{-2 \pm j, 5\} \rightarrow \text{Used as a check to show that } |\lambda| < \min\{7.5, 5\}$$

$$\|A\|_1 = \max\{6, 7, 5\} = 7 \Rightarrow \text{Column}$$

$$\|A\|_\infty = \max\{3, 3, 12\} = 12 \Rightarrow \text{Row}$$

$$\|A\|_2 = 7.25 = \max\{\lambda[A^T A]\}$$

$$|-2 \pm j| = \sqrt{(-2)^2 + (\pm 1)^2} = \sqrt{5}$$

$$|5| = 5$$

\Rightarrow Calculating e^{At} using the Eigen vectors

Given $\dot{x} = Ax + Bu$

Assuming A has distinct eigenvalues λ_i $i=1, 2, \dots, n$
then the n eigenvector w_i are independent

leading to $W = [w_1, w_2, \dots, w_n]$ being nonsingular i.e. w_i exists.

Using $Aw_i = \lambda_i w_i$

$$A[w_1 \ w_2 \ \dots \ w_n] = [w_1 \ w_2 \ \dots \ w_n] \begin{bmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \lambda_n \end{bmatrix}$$

$$AW = W\Lambda \quad ; \quad |w| \neq 0$$

$$A = W\Lambda W^{-1} \quad \text{order important}$$

$$A^t e = e$$

$$= W\Lambda^t W^{-1} e$$

It can be $A^t e = W\Lambda^t W^{-1} e$

Shown that \rightarrow

$$= W \begin{bmatrix} \lambda_1^t & 0 & \dots & 0 \\ 0 & \lambda_2^t & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \lambda_n^t \end{bmatrix} W^{-1} e$$

Example: Given $\dot{x} = \begin{bmatrix} 0 & 1 \\ -20 & -9 \end{bmatrix} x + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u$

Calculate e^{At}

$$\lambda^2 + 9\lambda + 20 = 0$$

$$(\lambda + 4)(\lambda + 5) = 0$$

eigen vector

$$\lambda_1 = -4 \Rightarrow$$

$$\lambda_2 = -5$$

$$w_1 = \begin{bmatrix} 1 \\ -4 \end{bmatrix} \text{ or } \begin{bmatrix} -\frac{1}{4} \\ 1 \end{bmatrix}$$

$$w_2 = \begin{bmatrix} 1 \\ -5 \end{bmatrix}$$

$$e^{At} = W e^{At} W^{-1}$$

$$= \begin{bmatrix} 1 & 1 \\ -4 & -5 \end{bmatrix} \begin{bmatrix} e^{-4t} & 0 \\ 0 & e^{-5t} \end{bmatrix} \begin{bmatrix} -5 & -1 \\ 4 & 1 \end{bmatrix}$$

* -1 ← determinant.

$$\begin{bmatrix} e^{-4t} & -5e^{-5t} \\ -4e^{-4t} & 5e^{-5t} \end{bmatrix}$$

$$\begin{bmatrix} -4e^{-4t} & -5e^{-5t} & -4e^{-4t} & -5e^{-5t} \\ 5e^{-4t} & 4e^{-5t} & e^{-4t} & -e^{-5t} \\ -20e^{-4t} + 20e^{-5t} & -4e^{-4t} + 5e^{-5t} \end{bmatrix}$$

t=0 case

In

...

...

* Ex: Confirm the result using 2) $[sI - A]^{-1}$

$$2) e^{At} = I + At + \frac{A^2 t^2}{2!} + \frac{A^3 t^3}{3!} + \dots$$

$$e^{At} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + \begin{bmatrix} 0 & t \\ -20t & -9t \end{bmatrix} + \begin{bmatrix} -20t^2 & -9t^2 \\ 180t^2 & 61t^2 \end{bmatrix} + \dots$$

power series expansion method

using Matlab

$$\gg A = \begin{bmatrix} 0 & 1 \\ -20 & -9 \end{bmatrix}, \text{expm}(A)$$

⇒ Steady State Value Due to a unit step

Given: $\dot{X} = AX + Bu$ Where the system is stable

$$X(t) = e^{At} X(0) + \int_0^t e^{A(t-\tau)} B u(\tau) d\tau, X(0) \neq 0$$

Compare with $\frac{dX}{dt} = aX + f(t)$

$$X_{ss} = \lim_{t \rightarrow \infty} X(t) = \lim_{t \rightarrow \infty} e^{At} X(0) + \lim_{t \rightarrow \infty} e^{At} \int_0^t e^{-A\tau} B u(\tau) d\tau$$

Zero

$$X_{ss} = e^{At} \left[\lim_{t \rightarrow \infty} \int_0^t e^{-A\tau} \times 1 \times d\tau \right] B$$

$$\text{Evaluate } \int_0^t e^{A\tau} d\tau = \int_0^t \left(I + At + \frac{A^2 t^2}{2!} + \dots \right) d\tau$$

7/3/2017

$$= \left[I\tau + \frac{A\tau^2}{2!} + \frac{A^2\tau^3}{3!} + \dots \right] \Big|_0^t$$

9/3/2017

$$\int_0^t e^{-A\tau} d\tau = \int_0^t \left[I_n - A\tau + \frac{A^2\tau^2}{2!} - \frac{A^3\tau^3}{3!} + \dots \right] d\tau$$

$$\begin{aligned} &= \left[I_n\tau - \frac{A\tau^2}{2!} + \frac{A^2\tau^3}{3!} - \dots \right] A^{-1} \Big|_0^t \\ &= \left[A\tau - \frac{A^2\tau^2}{2!} + \frac{A^3\tau^3}{3!} - \dots \right] A^{-1} \Big|_0^t \\ &= \left[I_n - e^{-A\tau} \right] A^{-1} \Big|_0^t \end{aligned}$$

$$= \left[I_n - e^{-At} - I_n + I_n \right] A^{-1} = \left[I_n - e^{-At} \right] A^{-1}$$

$$X_{ss} = \lim_{t \rightarrow \infty} \left[e^{At} \left[I_n - e^{-At} \right] A^{-1} B \right]$$

$$X(t) = \lim_{t \rightarrow \infty} \left[e^{At} - I_n \right] A^{-1} B$$

$$= -A^{-1}B$$

in stable system
 $e^{At} = 0$

Ex: Use the transfer function matrix to confirm this result

$$G(s) = C[sI - A]^{-1}B + D$$

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Five Apple

Hint: Consider suitable C & D together with any limiting theorems in the S-domain

Ex: Use $\dot{x} = Ax + Bu$ to confirm the previous result ; u is a unit system is stable A^{-1} exists.

Theorem 8 For linear time-invariant systems

IF $u(t) \rightarrow x(t)$ then

$$\frac{d}{dt} u(t) \rightarrow \frac{d}{dt} x(t)$$

$$\int u(t) dt \rightarrow \int x(t) dt$$

Ex: Given $\dot{x} = Ax + Bu$

Use $x(t)$ for a unit step (see notes and book)

Determine $x(t)$ for a unit ramp input (c.f. with book)

12/3/2017

Sunday

Diagonalization of a MatrixConsider $i = Ax + Bu$

$$\begin{bmatrix} -1 & 2 \\ 3 & 4 \end{bmatrix} x + \begin{bmatrix} 1 \\ 1 \end{bmatrix} u$$

$$\lambda_1 + \lambda_2 = -1 + 4 = 3$$

$$\lambda_1 \lambda_2 = -10$$

Solving $\lambda_1 = -2, \lambda_2 = 5$

$$w_1 = \text{adj} \begin{bmatrix} 1 & 2 \\ 3 & 6 \end{bmatrix} = \begin{bmatrix} 6 \\ -3 \end{bmatrix}$$

$$w_2 = \text{adj} \begin{bmatrix} -6 & 2 \\ 3 & -1 \end{bmatrix} = \begin{bmatrix} -1 \\ -3 \end{bmatrix} \text{ or } \begin{bmatrix} 1 \\ 3 \end{bmatrix}$$

$$\Lambda = W^{-1}AW$$

A in
caps.

$$= \begin{bmatrix} 3 & -1 \\ -1 & -2 \end{bmatrix} \begin{bmatrix} -1 & 2 \\ 2 & 4 \end{bmatrix} \begin{bmatrix} -2 & 1 \\ 1 & 3 \end{bmatrix} \times \frac{-1}{7} = \begin{bmatrix} -2 & 0 \\ 0 & 5 \end{bmatrix}$$

This is a diagonal
Matrix

* example: Diagonalize $A = \begin{bmatrix} -5 & 3 & 3 \\ -6 & 3 & 4 \\ 0 & 1 & 0 \end{bmatrix} \Rightarrow \text{Sol.}$

Q

$$\rightarrow \text{eigen}(A) \text{ ans. } \rightarrow [1 \quad -2 \quad 1]$$

write down the Matlab which Determine the eigen vector. $\rightarrow \text{eig}(A)$
commands.
the Trace

(28)

Five Apple

Sol. For previous example.

$$\Rightarrow [w \ e] = \text{eig}(A)$$

$$\lambda = \text{inv}(w) * A * w$$

$$\text{ans. } \begin{bmatrix} 1 & 0 & 0 \\ 0 & -2 & 0 \\ 0 & 0 & -1 \end{bmatrix}$$

→ check the answer By hand calculations.

* System invariant Under Similarity Transformation.

$$\text{given. } \begin{aligned} \dot{\hat{x}} &= Ax + Bu \\ y &= Cx + Du \end{aligned}$$

$$\text{let } \hat{x} = P\bar{x}$$
$$P\dot{\bar{x}} = AP\bar{x} + Bu$$

$$\dot{\bar{x}} = P^{-1}AP\bar{x} + P^{-1}Bu = \bar{A}\bar{x} + \bar{B}u$$

$$y = Cx + Du$$

$$y = C(P\bar{x}) + Du$$

$$x = p \cdot \bar{x} \quad \text{Identical}$$

Def: Given matrices A and \bar{A} they are similar iff there exists a transformation P such that

$$\bar{A} = P^{-1} A P$$

or

$$A = P \bar{A} P^{-1}$$

$$\begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} \text{ is similar to } \begin{bmatrix} -1 & 0 \\ 0 & -2 \end{bmatrix}$$

$$\text{Where } P = \begin{bmatrix} 1 & 1 \\ -1 & -2 \end{bmatrix}$$

↑ ↑
eigenvector
of A

* System invariants

$$\text{Trace}(P^{-1} A P) = \text{Trace}(A) = \text{Trace}(\bar{A})$$

$$|A| = |P^{-1} A P| = |A| |P^{-1} P| = |A|$$

Proof: Using $|A||B| = |BA| = |A||B|$

$$|CB| = |BC| = |C||B| = |B||C|$$

Note:

Similarity doesn't mean identically

$$\text{Let } C = P^{-1} A, B = P$$

$$|P^{-1} A P| = |P P^{-1} A| = |I_n A| = |A| \quad \#$$

$$\lambda[\bar{A}] = \lambda[A]$$

Proof: using $\lambda[CB] = \lambda[BC]$

$$\lambda[\bar{A}] = \lambda[P^{-1} A P]$$

$$\lambda[P P^{-1} A] = \lambda[I_n A] = \lambda[A]$$

N.B. Similarity Transformation (ST) doesn't change the eigen values It changes the eigenvectors (not invariant under ST)

- The TFM is invariant under ST

$$G(s) = \bar{C} [sI - \bar{A}]^{-1} \bar{B} + \bar{D}$$

$$= C P [sI - P^{-1} A P]^{-1} P B + D$$

$$= C [sI P^{-1} - P^{-1} A P P^{-1}]^{-1} P^{-1} B + D$$

Using $(CD)^{-1} = D^{-1} C$

$$= C [P sI P^{-1} - P^{-1} A P P^{-1}]^{-1} B + D$$

$$= C [sI - A]^{-1} B + D$$

$$= G(s)$$

Theorem: every matrix satisfies its C.E. $\downarrow C.E.$

$$\text{If } \lambda^n + a_1 \lambda^{n-1} + a_2 \lambda^{n-2} + \dots + a_n = 0$$

is the C.E. of A then

$$A^n + a_1 A^{n-1} + a_2 A^{n-2} + \dots + a_n I_n = 0$$

→ This is known as Cayley-Hamilton Theorem

Ex: let $A = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix}$

- i) determine G.E
- ii) verify C.H theorem
- iii) use the theorem to determine A^2 without sequence A
- iv) " " " " " A^{-1} .

→ Fifteen → 2.1

المسألة

* Jordan Blocks (Forms)

Result When there is insufficient number of independent eigenvectors associated with a repeated eigen value.

i.e. if a matrix A has 1, 2, 3, 3, 3 eigen values then possible Jordan block maybe

$$\begin{bmatrix} 3 & 1 & 1 & 0 & 0 \\ 0 & 3 & 1 & 0 & 0 \\ 0 & 0 & 3 & 0 & 0 \\ 0 & 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

J_1

or

$$\begin{bmatrix} 3 & 1 & 0 & 0 & 0 \\ 0 & 3 & 0 & 0 & 0 \\ 0 & 0 & 3 & 0 & 0 \\ 0 & 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

J_2

* Generalized Eigen Vectors

Consider a matrix A with 3 identical eigen values (λ) then at least an eigen vector w_1 exists

$$[A - \lambda I] w_1 = 0$$

To get the other two eigen vectors (called generalized eigen vector), w_2, w_3

Solve $[A - \lambda I] w_2 = w_1$; $[A - \lambda I] = 0$, no inverse

$$\neq [A - \lambda I] w_3 = w_2$$

Example: let $A = \begin{bmatrix} 0 & 1 \\ -4 & 4 \end{bmatrix}$ $\lambda^2 - 4\lambda + 4 = 0$

$$(\lambda - 2)^2 = 0$$

$$\lambda_1 = 2, \lambda_2 = 2$$

$$\begin{bmatrix} -2 & 1 \\ -4 & 2 \end{bmatrix} w_1 = 0$$

$$\begin{bmatrix} -2 & 1 \\ 0 & 0 \end{bmatrix} w_1 = 0 \Rightarrow w_1 = \begin{bmatrix} 1/2 \\ 1 \end{bmatrix}$$

Solve \Rightarrow

$$\begin{bmatrix} -2 & 1 \\ -4 & 2 \end{bmatrix} w_2 = \begin{bmatrix} 1/2 \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} -2 & 1 \\ 0 & 0 \end{bmatrix} w_2 = \begin{bmatrix} \frac{1}{2} \\ 0 \end{bmatrix}$$

$$w_{22} = 1$$

$$-2w_{11} + 1 = \frac{1}{2} \Rightarrow w_{11} = \frac{1}{4}$$

$$w_2 = \begin{bmatrix} \frac{1}{4} \\ 1 \end{bmatrix}$$

$\Rightarrow w_1, w_2$ are independent i.e. $|w_1 w_2| \neq 0$

To check answer

$$A_J = W^T A W$$

$$= \begin{bmatrix} \frac{1}{2} & \frac{1}{4} \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ -4 & 4 \end{bmatrix} \begin{bmatrix} \frac{1}{2} & \frac{1}{4} \\ 1 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} 2 & 1 \\ 0 & 2 \end{bmatrix}$$

→ another set of vector

$$w_1 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \quad w_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$A_J = \begin{bmatrix} 1 & 0 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ -4 & 4 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 2 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} 2 & 1 \\ 0 & 2 \end{bmatrix}$$

* The exponential matrix e^{Jt}

$$\text{let } J = \begin{bmatrix} \lambda & 1 & 0 \\ 0 & \lambda & 1 \\ 0 & 0 & \lambda \end{bmatrix} \quad \text{or} \quad \begin{bmatrix} \lambda & 1 & 0 \\ 0 & \lambda & 0 \\ 0 & 0 & \lambda \end{bmatrix}$$

Simple structure

it can be shown that « see book »

$$e^{Jt} = \begin{bmatrix} e^{\lambda t} & t e^{\lambda t} & t^2 e^{\lambda t} \\ 0 & e^{\lambda t} & t e^{\lambda t} \\ 0 & 0 & e^{\lambda t} \end{bmatrix}$$

Example 20

$$\text{let } A = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 2 & 1 \\ 0 & 0 & 2 \end{bmatrix}$$

$$e^{At} = \begin{bmatrix} e^{3t} & 0 & 0 \\ 0 & e^{2t} & t e^{2t} \\ 0 & 0 & e^{2t} \end{bmatrix}$$

to check it at $t=0$
must give us the
Identity

$$\Rightarrow [e^{At}]^{-1} = e^{-At}$$

$$[e^{At}]^{-1} = \begin{bmatrix} e^{-3t} & 0 & 0 \\ 0 & e^{-2t} & -t e^{-2t} \\ 0 & 0 & e^{-2t} \end{bmatrix}$$

A and e^{At} commute, i.e.

$$A e^{At} = e^{At} A$$

$$\frac{d}{dt} e^{At} = A e^{At} = e^{At} A$$

$$(a+b)t e^{At} = a e^{At} + b e^{At}$$

a, b scalar.

$$(A+B)t e^{At} = A e^{At} + B e^{At}$$

A, B matrix

A, B commute.

$$\rightarrow P = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 1 \\ 2 & 0 & 1 \end{bmatrix}$$

$$J = \begin{bmatrix} 3 & 1 & 0 \\ 0 & 3 & 1 \\ 0 & 0 & 3 \end{bmatrix}$$

$$\Rightarrow \text{trace}(J)$$

$$|J| = 27$$

$$A = P * J * \text{inv}(P)$$

$$\text{Ex: Given } A = \begin{bmatrix} 5 & 1 & -1 \\ -2 & 3 & 1 \\ 4 & 2 & 1 \end{bmatrix}$$

$$\Rightarrow \text{trace}(A)$$

$$|A| = 9$$

Obtain the Jordan blocks for A i.e. get the eigenvectors with associated eigenvalues and generalized eigenvectors, then compute

$$P^{-1} A P \rightarrow J, \text{ Check using matlab } [P, e] =$$

$$\text{eig}(A);$$

$$J = \text{inv}(P) * (A) * P$$

* Basis of Jordan Blocks

To diagonalize a matrix A , we use a similarity transformation W consisting of the eigenvectors as columns.

i.e. $A [w_1 \ w_2 \ \dots \ w_n] = [w_1 \ w_2 \ \dots \ w_n] \begin{bmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \lambda_n \end{bmatrix}$

$$\rightarrow AW = W\Delta$$

$$\rightarrow A = W\Delta W^{-1}$$

$$\rightarrow W^{-1}AW = \Delta$$

If w are the eigenvector of A corresponding to distinct eigenvalues then W is always nonsingular

hence $\Delta = W^{-1}AW$ where Δ is diagonal.

→ Suppose now, we have repeated eigenvalues, then a possible form resembling that of diagonal form is the following

$$A [w_1 \ w_2 \ w_3] = [w_1 \ w_2 \ w_3] \begin{bmatrix} \lambda & 1 & 0 \\ 0 & \lambda & 1 \\ 0 & 0 & \lambda \end{bmatrix}$$

$$Aw_1 = \lambda w_1 \Rightarrow [A - \lambda I]w_1 = 0$$

$$Aw_2 = w_1 + \lambda w_2 \Rightarrow [A - \lambda I]w_2 = w_1$$

$$Aw_3 = w_2 + \lambda w_3 \Rightarrow [A - \lambda I]w_3 = w_2$$

\Rightarrow eigen vector $\neq 0$

$$[A - \lambda I] w_1 = 0$$

$$[A - \lambda I] w_2 = w_1 \Rightarrow [A - \lambda I]^2 w_2 = 0$$

$$[A - \lambda I] w_3 = w_2 \Rightarrow [A - \lambda I]^3 w_3 = 0$$

Ex: $A = \begin{bmatrix} 0 & 1 \\ -4 & 4 \end{bmatrix}$, $\lambda_1 = 2, \lambda_2 = 2$

$$\begin{bmatrix} -2 & 1 \\ -4 & 2 \end{bmatrix} w_1 = 0 \Rightarrow w_1 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

$$\begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} w_2 = 0 \text{, many choices for } w_2 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \dots$$

Check: let $w_1 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$

$$Aw = w \Delta$$

$$\Delta = w^{-1} A w$$

$$\Delta = \begin{bmatrix} 5 & -1 \\ -2 & 1 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ -4 & 4 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 2 & 5 \end{bmatrix}$$

$$\Delta = \begin{bmatrix} 2 & 3 \\ 0 & 2 \end{bmatrix}$$

When we chose $w_2 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$

$$\Delta = W^{-1} A W$$

$$= \begin{bmatrix} 0 & 1 \\ -2 & -1 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ -4 & 4 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 2 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} 2 & -2 \\ 0 & 2 \end{bmatrix}$$

The importance of Jordan block is getting them as upper triangular matrices

* Independence of vectors so

Def: Vectors V_1, V_2, \dots, V_n are independent iff

$$C_1 V_1 + C_2 V_2 + \dots + C_n V_n = 0$$

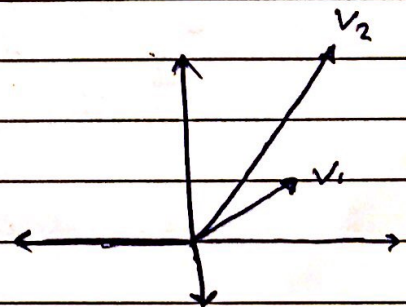
implies $C_1 = C_2 = \dots = C_n = 0$

Example: let $V_1 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$, $V_2 = \begin{bmatrix} 2 \\ 5 \end{bmatrix}$

$$\begin{bmatrix} 1 \\ 2 \end{bmatrix} C_1 + \begin{bmatrix} 2 \\ 5 \end{bmatrix} C_2 = 0$$

$$C_1 + 2C_2 = 0$$

$$2C_1 + 5C_2 = 0$$



$$0 + C_2 = 0 \Rightarrow C_2 = 0$$

$$C_1 = -2C_2 = 0$$

hence V_1 and V_2 are independent

Example: let $V_1 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$, $V_2 = \begin{bmatrix} -4 \\ -8 \end{bmatrix}$

$$\begin{bmatrix} 1 \\ 2 \end{bmatrix} C_1 + \begin{bmatrix} -4 \\ -8 \end{bmatrix} C_2 = 0$$

$$C_1 - 4C_2 = 0$$

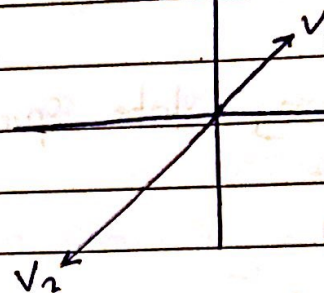
$$2C_1 - 8C_2 = 0$$

\Rightarrow

using Gauss-elimination

$$\begin{bmatrix} 1 & -4 \\ 2 & -8 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} 1 & -4 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$



$$0 \cdot c_2 = 0 \Rightarrow c_2 = r \neq 0 \text{ arbitrary}$$

$$c_1 = 4c_2 \Rightarrow c_1 = 4r$$

* Implications: Consider 4 3 dimensional vectors.

- Two 1D vectors span a plane.
- Three 1D vectors span a volume (space).
- If 3 vectors constitute a square matrix then the determinant is non-zero.

This can be used as a check of independency.

- Any fourth vector can't be independent of the other 3 1D vectors.
- The number of 1D vectors in a matrix (as columns of a matrix) is the rank of that matrix.

- If A is $m \times n$ matrix

then

$$\text{rank}(A) \leq \min(m, n)$$

- If A is square then if $\text{rank}(A) = n$ then $|A| \neq 0$

$\Rightarrow A$ is non-singular

* State Space, Transfer functions, and Block Diagram models of System.

Obtaining state space models from Differential equations

Given

$$y^{(n)} + a_1 y^{(n-1)} + a_2 y^{(n-2)} + \dots + a_n y = \underset{\substack{\uparrow \\ \text{constant}}}{b} u(t)$$

$$\text{let } y = x_1$$

$$y' = \dot{x}_1 = x_2$$

$$y'' = \dot{x}_2 = x_3$$

$$\vdots$$

$$y^{(n-1)} = \dot{x}_{n-1} = x_n$$

$$\begin{aligned} \dot{x}_n = y^{(n)} &= -a_1 y^{(n-1)} - a_2 y^{(n-2)} - \dots - a_n y + b u(t) \\ &= -a_1 x_n - a_2 x_{n-1} - \dots - a_n x_1 + b u(t) \end{aligned}$$

$$\dot{X} = \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \vdots & & & & \\ 0 & & & & 1 \\ -a_n & \dots & -a_2 & -a_1 \end{bmatrix} X + \begin{bmatrix} 0 \\ 0 \\ \vdots \\ b \end{bmatrix} u \quad \rightarrow b \neq 0$$

$$y = [1 \ 0 \ \dots \ 0] X + 0 \cdot u$$

This form is known as a controllable companion form.

→ $G(s)$ can be obtained either 1) Through the LT of the DE
2) Using

$$G(s) = C[sI - A]^{-1}B + D$$

→ To obtain block diagram presentation consider the 3rd order system for simplicity.

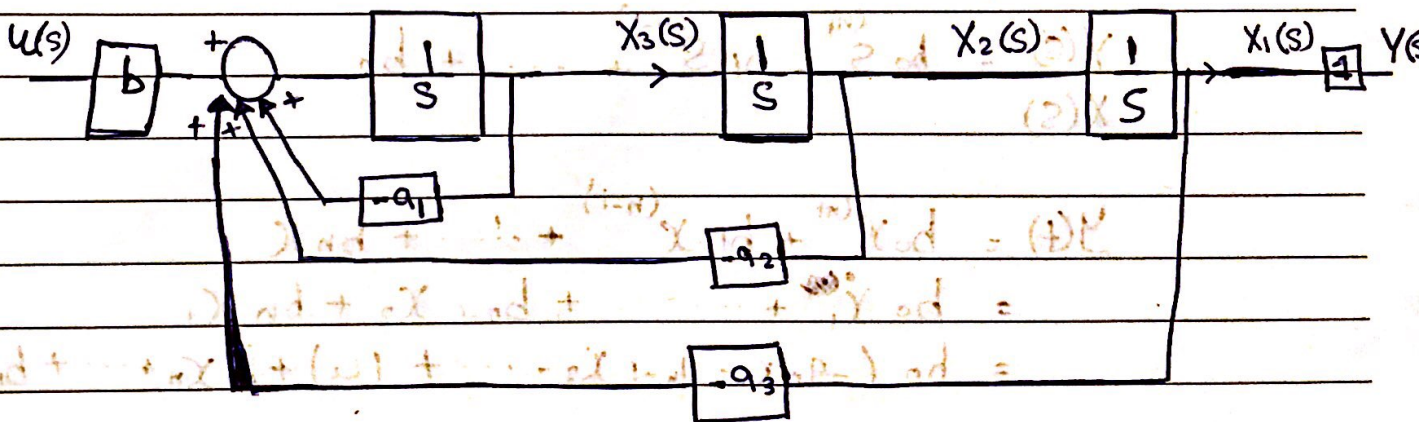
$$\dot{X} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -a_3 & -a_2 & -a_1 \end{bmatrix} X + \begin{bmatrix} 0 \\ 0 \\ b \end{bmatrix} u$$

$$y = [1 \ 0 \ 0] X$$

$$\Rightarrow \dot{X}_1 = X_2 \Rightarrow X_1(s) = \frac{1}{s} X_2(s)$$

$$\dot{X}_2 = X_3 \Rightarrow X_2(s) = \frac{1}{s} X_3(s)$$

integrator.



The case when derivatives of $u(t)$ are present

$$y^{(n)} + a_1 y^{(n-1)} + a_2 y^{(n-2)} + \dots + a_n y = b_0 u^{(n)} + b_1 u^{(n-1)} + \dots + b_n u$$

$$\frac{Y(s)}{U(s)} = \frac{b_0 s^n + b_1 s^{n-1} + \dots + b_n}{s^n + a_1 s^{n-1} + \dots + a_n}$$

$$\frac{Y(s)}{X(s)} \cdot \frac{X(s)}{U(s)} = (b_0 s^n + b_1 s^{n-1} + \dots + b_n) \times \frac{1}{s^n + a_1 s^{n-1} + \dots + a_n}$$

$$x^{(n)} + a_1 x^{(n-1)} + \dots + a_n x = u, \text{ let } x = x_1$$

$$\dot{x}_1 = x_2$$

$$x_2 = x_3$$

$$\Rightarrow A = \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ -a_n & -a_{n-1} & -a_{n-2} & \dots & -a_1 \end{bmatrix}; B = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$$

$$\frac{Y(s)}{X(s)} = \frac{b_0 s^n + b_1 s^{n-1} + \dots + b_n}{s^n + a_1 s^{n-1} + \dots + a_n}$$

$$\begin{aligned} Y(s) &= b_0 x^{(n)} + b_1 x^{(n-1)} + \dots + b_n x \\ &= b_0 \dot{x}_1 + \dots + b_{n-1} x_2 + b_n x_1 \\ &= b_0 (-a_n x_1 - a_{n-1} x_2 - \dots + u) + b_1 x_2 + \dots + b_n x_1 \end{aligned}$$

$$\Rightarrow C = [b_n - b_0 a_n \quad b_{n-1} - b_0 a_{n-1} \quad \dots] ; D = b_0$$

30/3/2017

Examples $G(s) = \frac{2s^2 + 3s + 4}{s^3 + 5s^2 + 6}$ $\Rightarrow \frac{Y(s)}{X(s)} \cdot \frac{X(s)}{U(s)} = \frac{Y(s)}{U(s)} \frac{\text{out}}{\text{input}}$

$$\frac{X(s)}{U(s)} = \frac{1}{1.s^3 + 5s^2 + 6}$$

$$A = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -6 & 0 & -5 \end{bmatrix}$$

(45)

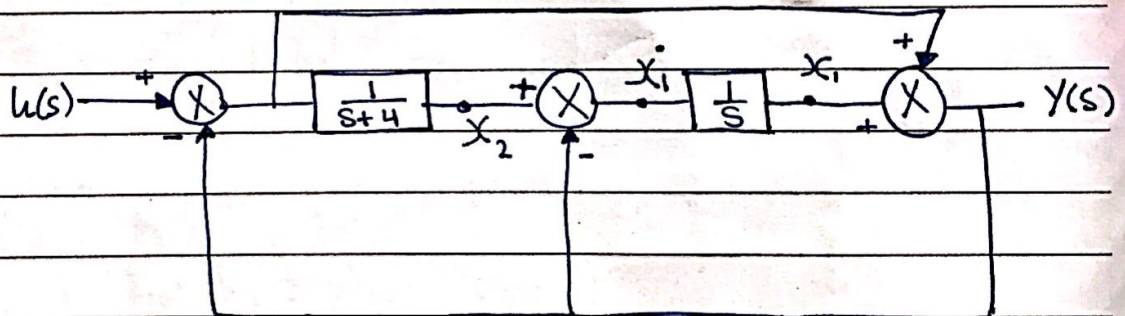
Poles in T.F \Rightarrow eigen value For the matrix A
 4 poles $\Rightarrow A \rightarrow 4 \times 4$

Ex: Convert $y''' + 4y'' - 5y' + 2y = 3u'' + 6u' - 8u$
 to a transfer function $G(s) = \frac{Y(s)}{U(s)}$

\rightarrow all initial conditions = 0

* Moving from Block Diagram to SS model

Introduce States at the output of integrators and Lags



$$\dot{x}_1 = x_2 - y$$

$$y = x_1 + u - y \Rightarrow y = \frac{1}{2}x_1 + \frac{1}{2}u$$

$$\text{hence } \dot{x}_1 = -\frac{1}{2}x_1 + x_2 - \frac{1}{2}u \dots \text{«1»}$$

$$x_2 = \frac{1}{s+4}(u - y)$$

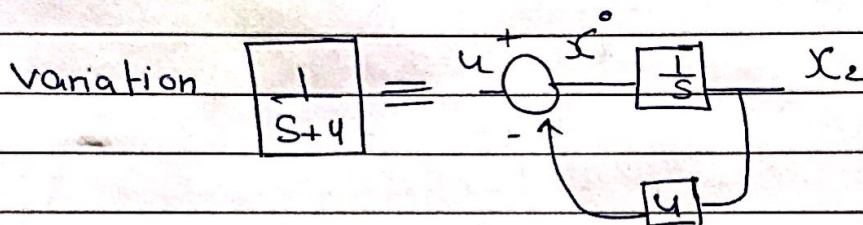
$$\dot{x}_2 + 4x_2 = u - \frac{1}{2}x_1 - \frac{1}{2}u$$

$$\dot{x}_2 = -\frac{1}{2}x_1 - 4x_2 + \frac{1}{2}u \quad \dots (2)$$

$$\dot{x} = \begin{bmatrix} -\frac{1}{2} & 1 \\ -\frac{1}{2} & -4 \end{bmatrix} x + \begin{bmatrix} -\frac{1}{2} \\ \frac{1}{2} \end{bmatrix} u$$

$$y = \begin{bmatrix} \frac{1}{2} & 0 \end{bmatrix} x + \frac{1}{2}u$$

Check: evaluate $G(s) = C[sI - A]^{-1}B + D$
do it at home



$$\frac{s+2}{s+5} = 1 + \frac{-3}{s+5}$$

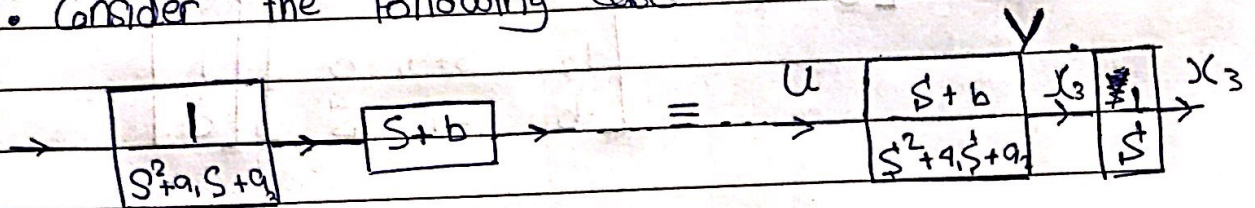
→ See additional

Example in the
book Ch9

$$\frac{s+10}{s} = 1 + \frac{10}{s}$$

4/4/2017

Example: Consider the following case



$$\frac{Y}{U} = \frac{X}{U} = \frac{(s+b)}{s^2+a_1s+a_2} \cdot \frac{1}{s}$$

$$\Rightarrow Y = X, \quad X_1 = X \Rightarrow Y = X_1 \Rightarrow \dot{X}_1 = X_2$$

$$\dot{X}_2 = -a_2X_1 - a_1X_2 + U$$

$$Y = \dot{X}_1 + bX_1 = bX_1 + X_2$$

$$\dot{X}_3 = Y = \dot{X}_1 + bX_1 = bX_1 + X_2$$

$$\dot{X} = \begin{bmatrix} 0 & 1 & 0 \\ -a_2 & -a_1 & 0 \\ b & 1 & 0 \end{bmatrix} X + \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} U$$

if the output z is that of the integrator

$$\text{then } \Rightarrow z = X_3$$

$$C = [0 \quad 0 \quad 1]$$

$$\text{eigen values } \Rightarrow -a_1, 0, 0$$

$$\text{check the T.F } \Rightarrow G(s) = C[sI - A]^{-1}B + D$$

$$G(s) = \begin{bmatrix} 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} s & -1 & 0 \\ a_2 & s+a_1 & 0 \\ -b & -1 & s \end{bmatrix} \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} s & -1 & 0 \\ a_2 & s+a_1 & 0 \\ -b & -1 & s \end{bmatrix} \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \xrightarrow{\Delta} \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

تدوير مصفوفة ب zero في اعمدة اليمين

$$\text{inverse the last column} \rightarrow \begin{bmatrix} 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} s & -1 & 0 \\ a_2 & s+a_1 & 0 \\ -b & -1 & s \end{bmatrix} \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

$$G(s) = \frac{s+b}{s(s(s+a_1)+a_2)}$$

$$G(s) = \frac{s+b}{s(s^2+a_1s+a_2)}$$

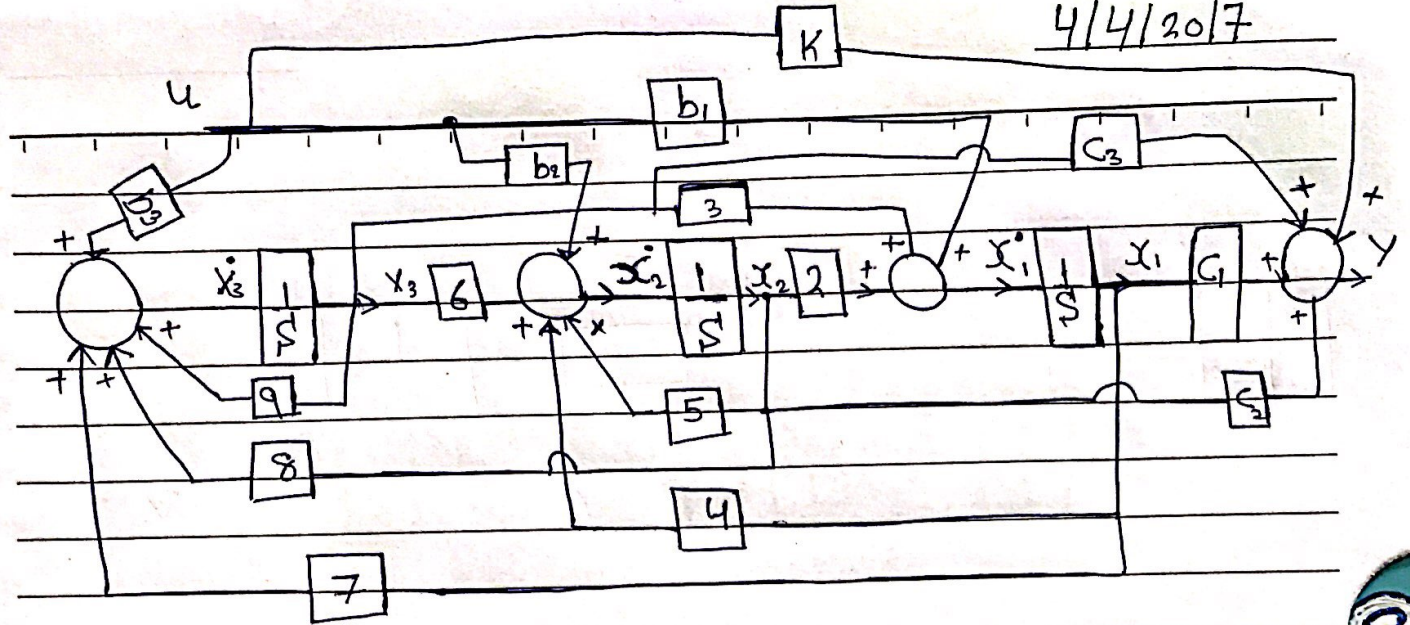
* Obtaining a BD using SS Representation

$$\text{Given } \dot{x} = \begin{bmatrix} 0 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix} x + \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} u$$

$$y = \begin{bmatrix} c_1 & c_2 & c_3 \end{bmatrix} x + \begin{bmatrix} k \end{bmatrix} u$$

⇒ Since system is Third order use 3 integrator

4/4/2017



9/4/2017

* Obtaining BD and SS model Using T.F

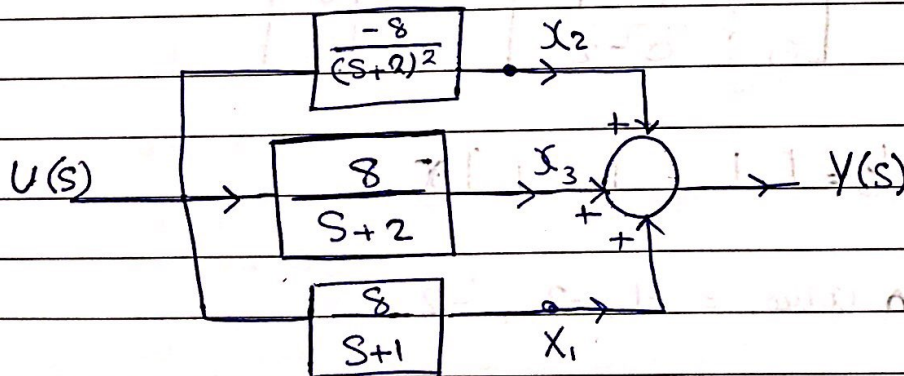
Example: let
$$\frac{Y(s)}{U(s)} = \frac{8}{(s+1)(s+2)^2}$$

$$= \frac{8}{s+1} + \frac{-8}{(s+2)^2} + \frac{C}{s+2}$$

$$C = \frac{d}{ds} \left(\frac{8}{s+1} \right) \Big|_{s=-2}$$

$$Y(s) = \frac{8U(s)}{s+1} + \frac{-8U(s)}{(s+2)^2} + \frac{-8U(s)}{s+2}$$

Giving the following B.D



9/4/2017

* Giving also the SS model

$$X_1(s) = \frac{8u(s)}{s+1} \Rightarrow s X_1(s) = -X_1(s) + 8u(s)$$

$$\dot{X}_1 = -X_1 + 8u(t)$$

$$X_2(s) = \frac{-1}{s+2} \cdot \frac{8u(s)}{(s+2)} \Rightarrow \frac{-1 * X_1(s)}{s+2}$$

$$\dot{X}_3 = -2X_3 + 8u(t)$$

$$\dot{X}_2 = -2X_2 - X_3$$

$$Y(s) = X_1(s) + X_2(s) + X_3(s)$$

$$y = x_1 + x_2 + x_3$$

$$X' = \begin{bmatrix} -1 & 0 & 0 \\ 0 & -2 & -1 \\ 0 & 0 & -2 \end{bmatrix} X + \begin{bmatrix} 8 \\ 0 \\ 8 \end{bmatrix} u$$

$$y = [1 \quad 1 \quad 1] X$$

eigen value = -1, -2, -2

Ex: obtain the B/D and SS models of

$$i) \frac{Y(s)}{U(s)} = \frac{640(s+5)}{(s+2)(s+4)^3}$$

59

Five Apple

$$22) \frac{Y(s)}{U(s)} = \frac{58}{(s+1)(s^2+4s+29)}$$

>> Syms s, g = C*Inv(s*eye(size(A))-A)*B+D,
 Simplify
 or, Simple(g)
 or, pretty(g)

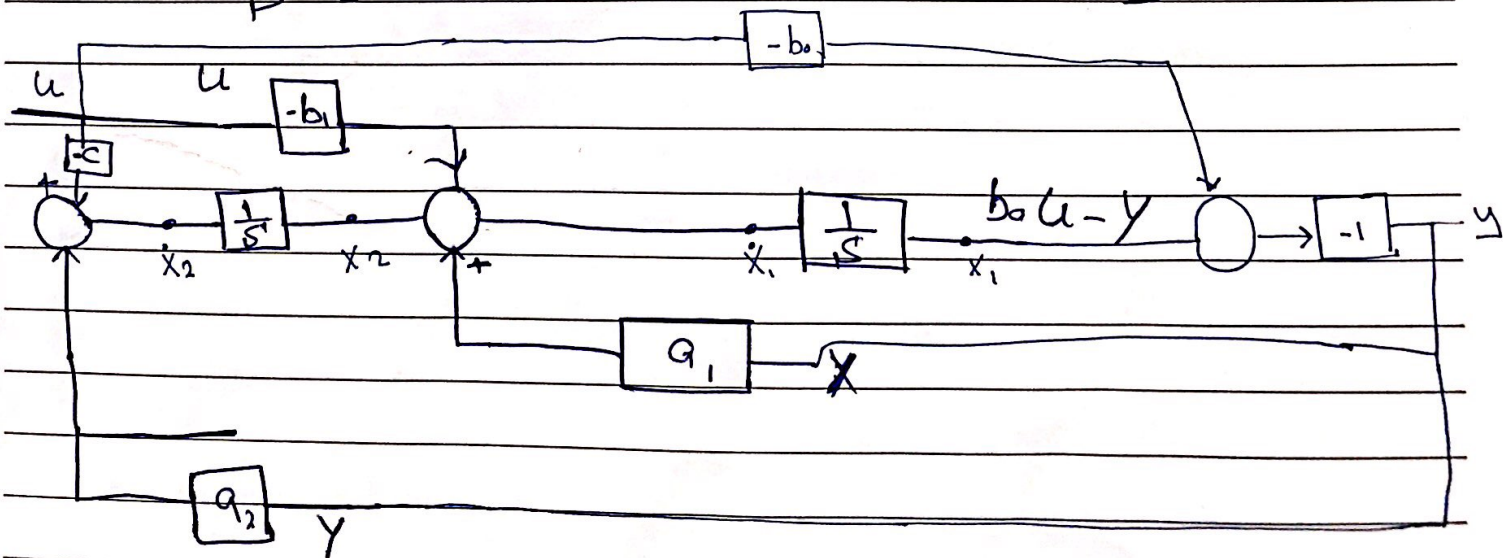
Giving also the SS model

$$\text{Example: let } \frac{Y(s)}{U(s)} = \frac{bs^2 + b_1s + c}{s^2 + a_1s + a_2}$$

$$Y(s) \left(\frac{1}{s^2 + a_1s + a_2} \right) = \left(b_0 + b_1 \frac{1}{s} + c \frac{1}{s^2} \right) U(s)$$

$$= (a_2y - c u) \frac{1}{s^2} + (a_1y - b_1u) \frac{1}{s} = b_0u - y$$

$$= \frac{1}{s} \left[(a_2y - c u) \frac{1}{s} + (a_1y - b_1u) \right] = b_0u - y$$



$$X_1 = b_0 u - y$$

$$y = b_0 u - X_1$$

$$\dot{X}_1 = X_2 + a_1(-X_1 + b_0 u) - b_1 u$$

$$-a_1 X_1 + X_2 + (a_1 b_0 - b_1) u$$

$$X_2 = a_2(-X_1 + b_0 u) - c v$$

$$= -a_2 X_1 + X(a_2 b_0 - c) u$$

$$\dot{X} = \begin{bmatrix} -a_1 & 1 \\ -a_2 & 0 \end{bmatrix} X + \begin{bmatrix} a_1 b_0 - b_1 \\ a_2 b_0 - c \end{bmatrix} u$$

$$cP - y = \begin{bmatrix} -1 & 0 \end{bmatrix} X + b_0 u$$

$$(2) \left(\frac{1}{s} + \frac{1}{s} + 0 \right) = \left(\frac{1}{s} + \frac{1}{s} + 0 \right) (2) y$$

$$V - U_d = \frac{1}{2} (U_d - y) + \frac{1}{2} (U_d - y) =$$

$$V - U_d = \frac{1}{2} (U_d - y) + \frac{1}{2} (U_d - y) =$$

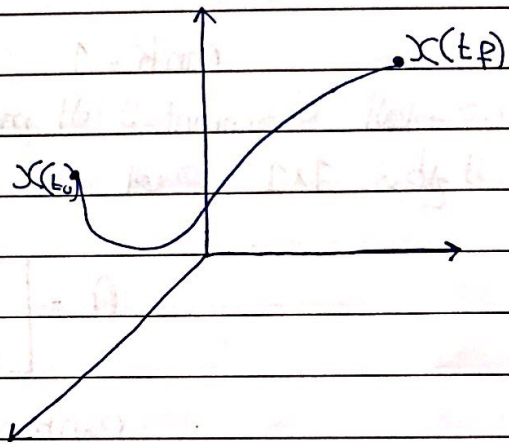
→ Structural properties of systems

They Give qualitative information about Systems

→ Controllability

Def:

The system is completely Controllable if it is possible to move it from an initial state $x(t_0)$ to any arbitrary final state $x(t_f)$ in a finite time using an unconstrained control law



* A Test of Controllability (CC)

→ A system $\dot{x} = Ax + Bu$ is completely CC iff $\text{rank}(M_c) = \text{rank}([B \quad AB \quad A^2B \quad \dots \quad A^{n-1}B]) = n$

Def: rank of a matrix is the number of independent columns

IF A is $m \times n$ matrix then $\text{rank}(A) \leq \min(m, n)$

$\text{rank} = 0 \rightarrow \text{matrix} = 0$

→ For a square matrix A if $|A| \neq 0$ then the rank is n .

Generally for a non square matrix the rank is the highest dimension of a sub matrix having a non zero determinant.

Example: let $A = \begin{bmatrix} 1 & 4 & -3 & 5 \\ 2 & 8 & -6 & 10 \end{bmatrix}_{2 \times 4} \leq 2$

rank = 1

دترمینان Sub matrix

Zero \rightarrow determinant of matrix \rightarrow rank = 0

1 = rank \rightarrow 1x1 matrix

$$A = \begin{bmatrix} 1 & 4 & -3 & 5 \\ 2 & 8 & 6 & 10 \end{bmatrix}$$

rank = 2

Sub matrix

$$|2 \times 2| \neq 0$$

$$A = \begin{bmatrix} 1 & 4 & -3 & 5 \\ 2 & 8 & 6 & 10 \\ 4 & 16 & 0 & 20 \end{bmatrix}$$

0 = $|3 \times 3|$ matrix \rightarrow rank = 0

0 = $|2 \times 2|$ matrix \rightarrow rank = 0

rank = 1

rank = 1

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Example: $\dot{X} = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} X + \begin{bmatrix} 1 \\ -1 \end{bmatrix} u$

$M_c = [B \ AB]$

$M_c = \begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix}$

$|M_c| \leq 2$

$\text{rank}(M_c) = 1 < 2 \Rightarrow \text{uncontrollable}$

Ex: $\dot{X} = \begin{bmatrix} -5 & 3 & 3 \\ -6 & 3 & 4 \\ 0 & 1 & 0 \end{bmatrix} X + \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} u$

\Rightarrow Uncontrollability results in pole zero cancellation in the T.F

Example: $\dot{X} = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} X + \begin{bmatrix} 1 \\ -1 \end{bmatrix} u$

$y = \begin{bmatrix} 1 & 2 \end{bmatrix} X$

Calculate $G(s)$ then judge Controllability Confirm using the rank test

→ The rank of an $m \times n$ matrix is the number of maximum number of independent columns (rows)

Column rank = row rank

Example e $\dot{x} = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} x + \begin{bmatrix} 1 \\ -1 \end{bmatrix} u$ → un controllable

$\begin{bmatrix} 0 \\ 1 \end{bmatrix}$ → any number
→ stair Controllable.

$y = \begin{bmatrix} 1 & 2 \end{bmatrix} x + \begin{bmatrix} 0 \end{bmatrix} u$ always controllable.

$G(s) = C[sI - A]^{-1}B + D$

$G(s) = \begin{bmatrix} 1 & 2 \end{bmatrix} \begin{bmatrix} s+3 & 1 \\ -2 & s \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix} \times \frac{1}{(s+1)(s+2)}$

$= \begin{bmatrix} s+1 & 1+2s \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix} \times \frac{1}{(s+1)(s+2)}$

$= \frac{-s-2}{(s+1)(s+2)} = \frac{-\cancel{(s+2)}}{(s+1)\cancel{(s+2)}} = \frac{-1}{s+1}$ Cancellation un controllable.

* Controllability Based on Diagonalization *

1) Distinct eigen values :

There exists a transformation according to $x = P \bar{x}$ giving

$$\dot{\bar{x}} = \bar{P}^{-1} A P \bar{x} + \bar{P}^{-1} B u$$

$$\dot{\bar{x}} = \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{bmatrix} \bar{x} + \begin{bmatrix} \beta_1 \\ \beta_2 \\ \beta_3 \end{bmatrix} u$$

→ The system is completely controllable iff all $\beta_i \neq 0$

Ex: $\dot{x} = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} x + \begin{bmatrix} 1 \\ -1 \end{bmatrix} u$ has eigen values $-1, -2$

$$P = \begin{bmatrix} 1 & 1 \\ -1 & -2 \end{bmatrix}, \quad P^{-1} = \begin{bmatrix} 2 & 1 \\ -1 & -1 \end{bmatrix}$$

$$\bar{P}^{-1} A P = \begin{bmatrix} -1 & 0 \\ 0 & -2 \end{bmatrix}, \quad \bar{B} = \bar{P}^{-1} B = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

⇒ eigen value -2 is uncontrollable

suppose $B = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$, $P^{-1} B = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$

⇒ system controllable

Ex: Given $A = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ x & x & x & x \end{bmatrix}$ & $B = \begin{bmatrix} 0 \\ 0 \\ 0 \\ x \end{bmatrix}$

Prove that the system is CC irrespective of x ?!

* Controllability of Jordan Form:

Given

$$\dot{X} = \begin{bmatrix} \lambda & 1 & 0 \\ 0 & \lambda & 0 \\ 0 & 0 & \lambda \end{bmatrix} X + \begin{bmatrix} B_1 \\ B_2 \\ B_3 \end{bmatrix} U$$

$$M_c = [B \quad AB \quad A^2B]$$

$$A^2B = A \cdot AB$$

$$M_c = \begin{bmatrix} B_1 & B_1\lambda + B_2 & B_1\lambda^2 + B_2\lambda + B_3 \\ B_2 & B_2\lambda & B_2\lambda^2 \\ B_3 & B_3\lambda & B_3\lambda^2 \end{bmatrix}$$

For Controllability $B_2 \neq 0$
 $B_3 \neq 0$

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eg $\dot{x} = \begin{bmatrix} 2 & 1 \\ 0 & 2 \end{bmatrix} x + \begin{bmatrix} 1 \\ 0 \end{bmatrix} u$

$\text{rank}(M_c) = \begin{pmatrix} \begin{bmatrix} 1 & 2 \\ 0 & 0 \end{bmatrix} \end{pmatrix} = 1 < n=2 \Rightarrow \text{un controllable}$

إذا اخرجت $0 = B$ في JB
un controllable $\leftarrow 0 = B$

$\Rightarrow A = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$

$\begin{bmatrix} 27 & -27 & 9 \end{bmatrix}$

$\text{eig}(A)$

$\Rightarrow B = \text{rand}(3,1) \Rightarrow \text{rank}([B \ A \times B \ A^2 \times B])$

$\Rightarrow M_c = \text{ctrb}(A, B), r = \text{rank}(M_c)$

* Observability

Def: A system is completely observable if it is possible to retrieve all the states from measurements of the output over a finite time period.

* A Test of observability

A system is completely observable iff $\text{rank} \begin{bmatrix} C \\ CA \\ \vdots \\ CA^{n-1} \end{bmatrix} = n$

* IF a system is unobservable, then we have pole-zero cancellation in the T.F.

Ex: $\dot{x} = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} x + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u$ Staircase and $B = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$
 $y = \begin{bmatrix} 1 & -1 \end{bmatrix} x$ → controllable by inspection

$$\text{rank} \begin{bmatrix} C \\ CA \end{bmatrix} = \text{rank} \begin{bmatrix} 1 & -1 \\ 2 & 4 \end{bmatrix} = 2 \text{ observable.}$$

IF $C = \begin{bmatrix} 1 & 1 \end{bmatrix}$

$$\text{rank} \begin{bmatrix} 1 & 1 \\ -2 & -2 \end{bmatrix} = 1 \Rightarrow \text{unobservable.}$$

IF $C = \begin{bmatrix} 1 & -2 \end{bmatrix}$

$$G(s) = [1 \quad d] \begin{bmatrix} s+3 & 1 \\ -2 & s \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} \times \frac{1}{(s+1)(s+2)}$$

$$G(s) = \frac{1+d s}{(s+1)(s+2)}$$

We have pole-zero cancellation when $d=1$

Given $G(s) = \frac{1}{s+2}$

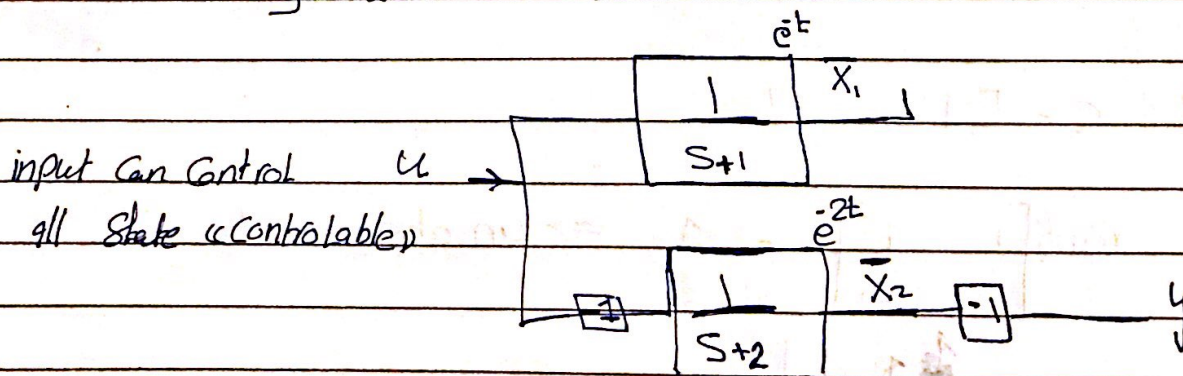
Diagonalize the system $\rho = \begin{bmatrix} 1 & 1 \\ -1 & -2 \end{bmatrix}$; $\rho^{-1} = \begin{bmatrix} 2 & 1 \\ -1 & -1 \end{bmatrix}$

$$\dot{\bar{x}} = \bar{P} \rho \bar{P}^{-1} \bar{x} + \bar{P}^{-1} B u$$

$$y = C \bar{P} \bar{x}$$

$$\dot{\bar{x}} = \begin{bmatrix} -1 & 0 \\ 0 & -2 \end{bmatrix} \bar{x} + \begin{bmatrix} 1 \\ 1 \end{bmatrix} u \quad \leadsto \text{Controllable}$$

$$y = [0 \quad -1] \bar{x} = -\bar{x}_2$$



by inspection unobservable, controllable.

i.e. e^{-t} cannot be obtained from y

N.B: Stability (s), Controllable (cc), and Observability (oo) don't imply each other i.e.

S cc oo

0 0 0

0 0 1

0 1 0

0 1 1

1 0 0

1 0 1

1 1 0

1 1 1

Theorem: For any uncontrollable system there exists a transformation P where $x = P\bar{x}$

if $\dot{x} = Ax + Bu$ and $y = Cx + Du$

then $\dot{\bar{x}} = \bar{P}AP\bar{x} + \bar{P}Bu \Rightarrow \bar{A}\bar{x} + \bar{B}u$

$\bar{y} = CP\bar{x} + Du \Rightarrow \bar{C}\bar{x} + Du$

$$\dot{\bar{x}} = \begin{bmatrix} A_c & A_2 \\ 0 & A_u \end{bmatrix} \bar{x} + \begin{bmatrix} B_c \\ 0 \end{bmatrix} u$$

The cc eigenvalues are those of A_c

The uc " " " " " " A_u

Example: $\dot{\bar{x}} = \begin{bmatrix} 0 & 1 \\ -20 & -9 \end{bmatrix} \bar{x} + \begin{bmatrix} 1 \\ -5 \end{bmatrix} u$

choose $P = [B \quad \begin{bmatrix} 0 \\ 1 \end{bmatrix}] = \begin{bmatrix} 1 & 0 \\ -5 & 1 \end{bmatrix}$

$$\dot{\bar{x}} = \begin{bmatrix} 1 & 0 \\ 5 & 1 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ -20 & -9 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ -5 & 1 \end{bmatrix} \bar{x} + \begin{bmatrix} 1 \\ 0 \end{bmatrix} u$$

$$= \begin{bmatrix} 0 & 1 \\ -20 & -4 \end{bmatrix} \bar{x} + \begin{bmatrix} 1 \\ 0 \end{bmatrix} u$$

$$= \begin{bmatrix} -5 & 1 \\ 0 & -4 \end{bmatrix} \bar{x} + \begin{bmatrix} 1 \\ 0 \end{bmatrix} u$$

-4 is uncontrollable eigen value

Check rank $[B \quad AB]$

$$\text{rank} \left(\begin{bmatrix} 1 & -5 \\ -5 & 25 \end{bmatrix} \right) = 1 < n=2 \Rightarrow \text{uncontrollable}$$

Ex: consider the previous system where

$$P = \begin{bmatrix} 1 & \alpha \\ -5 & 1 \end{bmatrix} \quad \text{prove uncontrollability for any value of } \alpha$$

Ex: given $\dot{x} = \begin{bmatrix} -5 & 3 & 3 \\ -6 & 3 & 4 \\ 0 & 1 & 0 \end{bmatrix} x + \begin{bmatrix} 1 \\ 1 \\ 6 \end{bmatrix} u$

AB
↓

choose $P = [B \quad \begin{bmatrix} \cdot \\ \cdot \\ \cdot \end{bmatrix}]$ must be invertible (non-singular).
Investigate the controllability.

* Duality of linear systems &

For a system of the form

$$\dot{x} = Ax + Bu$$

Original system (OS)

$$y = Cx + Du$$

Another system associated with it and called the dual system is of the form

$$\dot{z} = A^T z + C^T u$$

Dual system (DS)

$$w = B^T z + D^T u$$

* The OS and DS have the same stability properties
proof: eigen value (A) = eigen value (A^T)

$$\lambda[A] = \lambda[A^T]$$

* The two systems have the same C.E

$$CE = (\lambda - \lambda_1)(\lambda - \lambda_2) \dots (\lambda - \lambda_n) = 0 = |A^T - \lambda I_n|$$

→ Theorem: A system is $CC(oo)$ iff its dual is $oo(CC)$

» Ctrb → Test controllability

» obsv → Test observability

→ Theorem: An unobservable system can be decomposed as follows:

$$\dot{\bar{x}} = \begin{bmatrix} A_0 & 0 \\ A_1 & A_0 \end{bmatrix} \bar{x} + \begin{bmatrix} B_1 \\ B_2 \end{bmatrix} u$$

$$y = \begin{bmatrix} C & 0 \end{bmatrix} \bar{x}$$

proof: using Duality

* Another proof: using $OO = \begin{bmatrix} C_0 & 0 \\ C_0 A & 0 \\ C_0 A^2 & 0 \\ \vdots & 0 \\ C_0 A^{n-1} & 0 \end{bmatrix}$

$$\text{rank}(OO) < n \Rightarrow \text{unobservable}$$

* CC & OO by Inspection

↗ state space

There exists a transformation P which puts $S(A, B, C, D)$ in a jordan form. In which case CC & OO can be determined by inspection.

This is best illustrated by Example (P680)

$\bar{X} =$	1	0	0	0	0	0	0	0	0	$\bar{X} +$	NZ	IR	1 → CC
	0	2	0	0	0	0	0	0	0		IR	NZ	2 → CC
	0	0	3	1	0	0	0	0	0		0	0	JB
	0	0	0	3	6	0	0	0	0		NZ	IR	3 → CC
	0	0	0	0	4	1	6	0	0		NZ	NZ	
	0	0	0	0	0	4	1	0	0		NZ	NZ	
	0	0	0	0	0	0	4	0	0		0	0	4 is UC
	0	0	0	0	0	0	0	5	0		0	0	5 is UC
	0	0	0	0	0	0	0	0	6		IR	NZ	6 is CC

NZ: Non-Zero

IR: irrelevant

آخر الحركات

Non
Zero

$Y =$	0	IR	0	NZ	NZ	0	0	NZ	0	CC
	0	NZ	0	NZ	0	0	0	0	0	

التي اولي

Black

1 is UC 2 is UC 3 is UC 4 is UC 5 is UC 6 is UC

انها لا

un observable

For additional example See book

* pole (Eigenvalue) placement (assignment)

* Suppose a system $S(A, B, C, D)$ response is deemed unsatisfactory.
 * One way to change its behaviour is to employ state feedback.
 i.e.

$$u = r - Kx$$

r is reference point

$$\dot{x} = Ax + Bu$$

$$u = r - Kx$$

$y = Cx + Du \rightarrow$ output
 state \rightarrow feedback

$$\dot{x} = Ax + B(r - Kx)$$

$$\dot{x} = (A - BK)x + Br$$

* So choose a K to end up with a desired set of eigenvalues.

* Ackermann's Method

1. let $\lambda_1, \lambda_2, \dots, \lambda_n$ be the desired eigenvalues to be assigned.

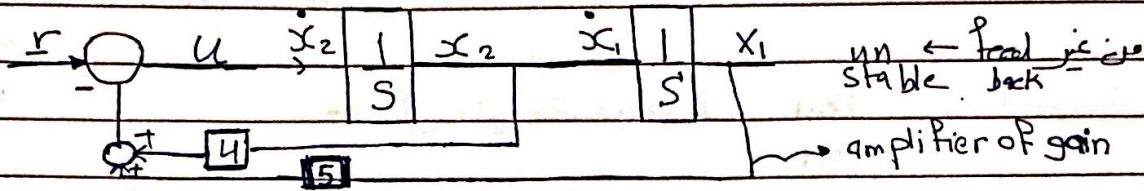
2. generate $p(\lambda) = (\lambda - \lambda_1) \dots (\lambda - \lambda_n) = \lambda^n + a_1 \lambda^{n-1} + \dots + a_n$

then

$$p(A) = A^n + a_1 A^{n-1} + \dots + a_n I_n$$

$$3. K = [0 \ 0 \ \dots \ 1] [B \ AB \ \dots \ A^{n-1}B]^{-1} p(A)$$

Examples Given $\dot{x} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} x + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u$



un stable so it needs stabilization
Assign $-2 \pm j$.

$$P(\lambda) = (\lambda + 2 - j)(\lambda + 2 + j) = \lambda^2 + 4\lambda + 5$$

$$P(A) = A^2 + 4A + 5I_2 = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} + 4 \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} 5 & 0 \\ 0 & 5 \end{bmatrix} = \begin{bmatrix} 5 & 4 \\ 0 & 5 \end{bmatrix}$$

$$K = \begin{bmatrix} 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}^{-1} \begin{bmatrix} 5 & 4 \\ 0 & 5 \end{bmatrix}$$

$$\rightarrow K = \begin{bmatrix} 5 & 4 \end{bmatrix}$$

Check $A - BK$

$$\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} - \begin{bmatrix} 0 & 0 \\ 5 & 4 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -5 & -4 \end{bmatrix}$$

$$\lambda^2 + 4\lambda + 5 = 0 \Rightarrow \lambda_1 = -2 + j, \lambda_2 = -2 - j$$

$$\Rightarrow q = \begin{bmatrix} 0 & 1 & \infty \end{bmatrix} ; b = \begin{bmatrix} 0 & 1 \end{bmatrix} ; K = \text{acker}(q, b, [-2 + j \ -2 - j])$$

N.B: K calculated for a single input system is unique.

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إذا كان في الترتيب input \leftarrow K is not unique.

في $a = [0 \ 1 \ 0 \ 0]$ و $b = [0 \ 1 \ 0 \ 0]$ و $K = \text{acker}(a, b, [-2+j-2j \ -2-j \ -2-j \ -2-j])$
 و Figure (1) و Step (a, b, eye(2), 0) و Figure (2) و Step(a-b*K, b, eye(2), 0)

كل ما زادته قيمة ال Real Part \leftarrow أصبح أسرع ال System

بمعنى gain بتزيد \leftarrow more power consumption

Ex: Given $\dot{x} = \begin{bmatrix} -5 & 3 & 3 \\ -6 & 3 & 4 \\ 0 & 1 & 0 \end{bmatrix} x + \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}$

$$K = [c \ \dots \ 1] [B \ AB \ A^2B]^T p(A)$$

N.B: One disadvantage of the Ackerman's method is that it doesn't apply to uncontrollable system.

Another disadvantage it doesn't work at multi input.

2/5/2017

* The entire - Eigenstructure Assignment Method ($E^N T$)

Also deals with multi-input systems and uncontrollable systems.

Given $\dot{x} = Ax + Bu$

Let $\lambda_1, \lambda_2, \dots, \lambda_n$ be the eigenvalues to be assigned.

1. Solve $[A - \lambda_i I_n] w_i = B z_i$, For w_i given a particular z_i for each λ_i

$$2. K = [z_1 \ z_2 \ \dots \ z_n] [w_1 \ w_2 \ \dots \ w_n]^{-1}$$

N.B: IF an λ_i is a re-assigned or uncontrollable eigenvalue - then

$$|A - \lambda_i I_n| = 0$$

Example: Given $\dot{x} = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} x + \begin{bmatrix} 1 \\ -1 \end{bmatrix} u$

assign -4 and -2

re-assigned because it is uncontrollable

For -4:

$$\begin{bmatrix} 4 & 1 \\ -2 & 1 \end{bmatrix} w_1 = \begin{bmatrix} 1 \\ -1 \end{bmatrix} * 6$$

z_i is a vector of size n

2/5/2017

$$W_1 = \begin{bmatrix} 1 & -1 \\ 2 & 4 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix} \cdot \frac{6 \times 1}{6} = \begin{bmatrix} 2 \\ -2 \end{bmatrix}$$

For -2 ;

$$\text{Inverse matrix} \rightarrow \begin{bmatrix} 2 & 1 \\ -2 & -1 \end{bmatrix} W_1 = \begin{bmatrix} 1 \\ -1 \end{bmatrix} * 1$$

Now Solve using gauss-elimination, one solution is

$$\begin{bmatrix} 2 & 1 \\ 0 & 6 \end{bmatrix} \begin{bmatrix} W_1' \\ W_2'' \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \Rightarrow W_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$K = \begin{bmatrix} 6 & 1 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ -2 & 1 \end{bmatrix}^{-1}$$

$$= \begin{bmatrix} 6 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ -2 & -2 \end{bmatrix} \cdot \frac{1}{2} = \begin{bmatrix} 4 & 1 \end{bmatrix}$$

Check $A - BK$

$$A = \begin{bmatrix} 1 \\ -1 \end{bmatrix} \begin{bmatrix} 4 & 1 \end{bmatrix}$$

$$\begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} - \begin{bmatrix} 4 & 1 \\ -4 & -1 \end{bmatrix} = \begin{bmatrix} -4 & 0 \\ 2 & -2 \end{bmatrix} \neq$$

$$\dot{x} = \begin{bmatrix} -5 & 3 & 3 \\ -6 & 3 & 4 \\ 0 & 1 & 0 \end{bmatrix} x + \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} z_i$$

This system needs a controller since it is unstable due to A having an eigenvalue of $+1$.

assign $-3, -4$ & -5 negative. System stable.

Solve $[A - \lambda_i I_n] w_i = B z_i$

for $\lambda_1 = 3, \lambda_2 = -4, \lambda_3 = -5$
 Choosing $z_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, z_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, z_3 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$

$\gg B = \begin{bmatrix} 1 & 0 & 1 \\ 1 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}; z_1 = \begin{bmatrix} 1 & 0 \end{bmatrix}; z_2 = \begin{bmatrix} 0 & 1 \end{bmatrix}; z_3 = \begin{bmatrix} 1 & 1 \end{bmatrix}$

$\gg \lambda_m = -3; A1 = A - \lambda_m * \text{eye}(3); w_1 = A1 \setminus B * z_1$

$$w_1 = \begin{bmatrix} 1 \\ 1.5 \\ -0.5 \end{bmatrix}$$

$\gg \lambda_m = -4; A1 = A - \lambda_m * \text{eye}(3); w_2 = A1 \setminus B * z_2$

$$w_2 = \begin{bmatrix} -0.3 \\ -0.1\bar{3} \\ 0.0\bar{3} \end{bmatrix}$$

$\gg \lambda_m = -5; A1 = A - \lambda_m * \text{eye}(3); w_3 = A1 \setminus B * z_3$

$$w_3 = \begin{bmatrix} 0.1\bar{6} \\ 0.41\bar{6} \\ -0.08\bar{3} \end{bmatrix}$$

$$\Rightarrow Z = [z_1 \ z_2 \ z_3]; W = [w_1 \ w_2 \ w_3];$$

$$K = Z * \text{inv}(W), \text{eig}(A - B * K)$$

$$K = \begin{bmatrix} -1 & 5 & 11 \\ -5.1429 & 6 & 7.7143 \end{bmatrix}$$

$$\text{Ans} = -5$$

$$-4$$

$$-3$$

Ex: Given $\dot{x} = \begin{bmatrix} 0 & 1 & 0 \\ 4 & 2 & 1 \\ 0 & 0 & 1 \end{bmatrix} x + \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} u$

unstable ↘

Positive eigen value.

Assign -1, -2, -3 using the two method

N.B: K is unique for single input systems irrespective of the method used.

* Observers

Used to estimate the states of a system which are unmeasurable (either inaccessible or sensors are unavailable or they are expensive)

- consider a system described by

$$\dot{x} = Ax + Bu$$

$$y = Cx$$

Where A , B , and C are known.

and an observer described by

$$K_e = \begin{bmatrix} x \\ x \\ x \end{bmatrix}$$

$$\dot{\hat{x}} = A\hat{x} + Bu + K_e(y - C\hat{x})$$

let error e be defined as $e = \overset{?}{x} - \overset{\check{}}{\hat{x}}$

$$\begin{aligned} \Rightarrow \dot{e} &= \dot{x} - \dot{\hat{x}} = Ax + Bu - A\hat{x} - Bu - K_e Cx + K_e C\hat{x} \\ &= Ae - K_e Ce \\ \dot{e} &= (A - K_e C)e \end{aligned}$$

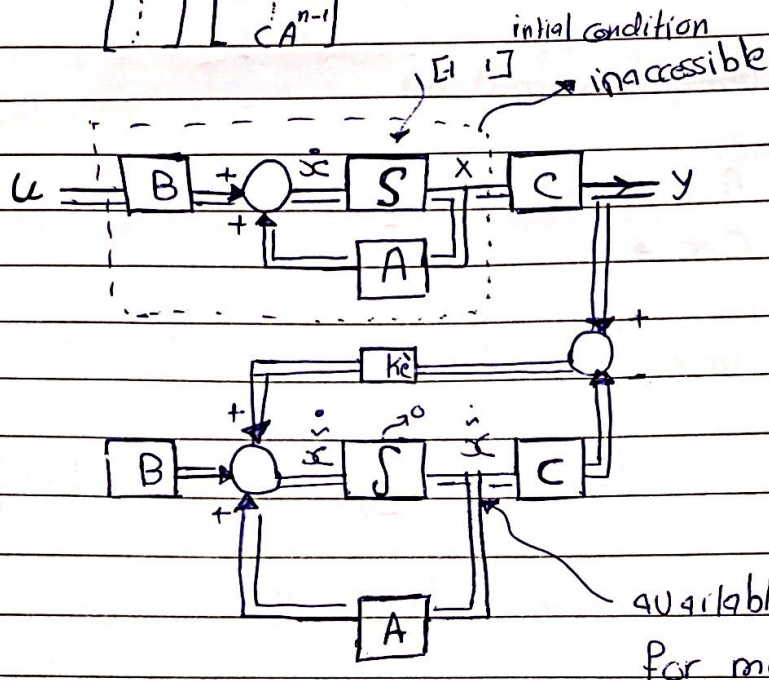
to get $\lim_{t \rightarrow \infty} e(t) = 0$, the eigenvalues of $A - K_e C$ should have negative real parts.

So K_e is chosen to fulfill this requirement

resulting $\lim_{t \rightarrow \infty} x = \lim_{t \rightarrow \infty} \hat{x}$

* Using Ackermann's method, it can be shown (prove that) that

$$K_e = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \begin{bmatrix} C \\ CA \\ \vdots \\ CA^{n-1} \end{bmatrix}^{-1} p(A)$$



>> $A = \begin{bmatrix} 0 & 1 & -2 & -2 \end{bmatrix}$; $B = \begin{bmatrix} 0 & 2 \end{bmatrix}$; $C = \begin{bmatrix} 1 & 2 \end{bmatrix}$; $K_e = \begin{bmatrix} -8 & 14 \end{bmatrix}$
 $K_e = \text{acker}(A', C', [-4 \quad -4])$

* The Separation Principle is

Suppose $u = r - k\tilde{x}$ where $k = \begin{bmatrix} \quad \end{bmatrix}$ while $K_e = \begin{bmatrix} \quad \\ \quad \end{bmatrix}$

So $\dot{\tilde{x}} = A\tilde{x} + B(r - k\tilde{x})$

$= A\tilde{x} - BK\tilde{x} + Br$

Recall

$e = x - \tilde{x}$ & $\dot{e} = (A - K_e C)e$
 $\tilde{x} = x - e$

$$\therefore \dot{x} = Ax - BKx + BKe + B\bar{r}$$

$$\dot{x} = (A-BK)x + BKe + B\bar{r} \quad \dots (2)$$

i.e.

$$\begin{bmatrix} \dot{x} \\ \dot{e} \end{bmatrix} = \begin{bmatrix} A-BK & BK \\ 0 & A-KeC \end{bmatrix} \begin{bmatrix} x \\ e \end{bmatrix} + \begin{bmatrix} B \\ 0 \end{bmatrix} r$$

adu \rightarrow eigen values ايجن والوز
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Ex

$$Ke = \begin{bmatrix} -0.8 \\ 10.8 \end{bmatrix} \quad \text{Choose } k_1 \text{ to get eigenvalue } -2 \text{ f } -2$$

using $A \neq B$

$$\gg K = \text{acker}(A, B, [-2 \ -2]) \Rightarrow K = [1 \ 3]$$

Choice of K & Ke } See book.

* Determination of System Zeros etc

Given $\dot{x} = Ax + Bu$ $u \in \mathbb{R}^m$

$y = Cx$ Where $|CB| \neq 0$

\rightarrow One method to determine the System Zeros is to calculate $G(s)$

\rightarrow Another method is as follows

- i) Determine the eigen values of $A = [I_n - B(CB)^{-1}C]A$
- ii) The zeros are the eigenvalue of A_z excluding m eigen value of value 0.

i.e. { zeros of system } = $\{\lambda[A_z] - \{0, 0, \dots\}\}$

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Example: $\dot{x} = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} x + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u$

$y = \begin{bmatrix} 1 & 1 \end{bmatrix} x$

$Az = [I - B(CB)^{-1}C]A$

$= \begin{bmatrix} 0 & 1 \\ 0 & -1 \end{bmatrix}$

$\lambda[Az] = \{0, -1\}$

Zero of system = $\{0, -1\} - \{0\} = \{-1\}$

Check answer determining $G(s)$

Ex: $\dot{x} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -6 & 11 & -6 \end{bmatrix} x + \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} u$

$y = \begin{bmatrix} 20 & 9 & 1 \end{bmatrix} x$