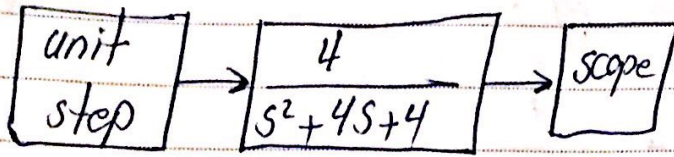
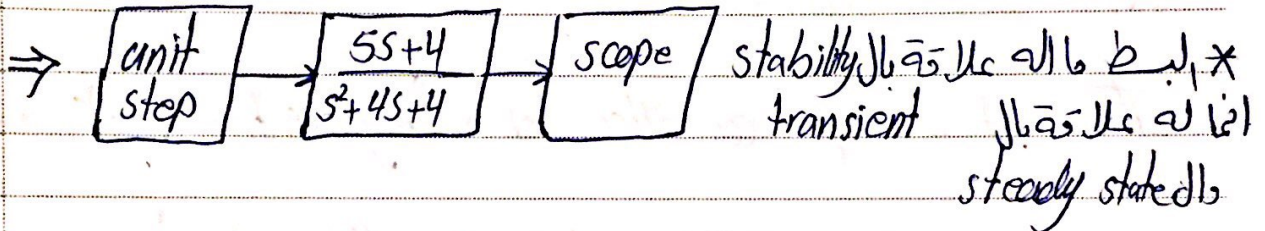


*



⇒ For second or first order system the coefficient of the denominator are all the same then the system is stable, further more for stable system the steady state value due to unit step is $\lim_{s \rightarrow 0} G(s) = T.F$



⇒ Routh's stability criteria:-

Ex:- given the characteristics equ. $s^3 + 4s^2 + 4s + K = 0$

s^3	1	4	0
s^2	4	K	0
s^1	$\frac{16-K}{4}$	0	
s^0	K		

* the system is stable when all the coeff charges are the same

- ① $K > 0$
- ② $16 - K > 0$
- ③ $0 < K < 16$

* Given the following system

Tuesday 1/10/2019

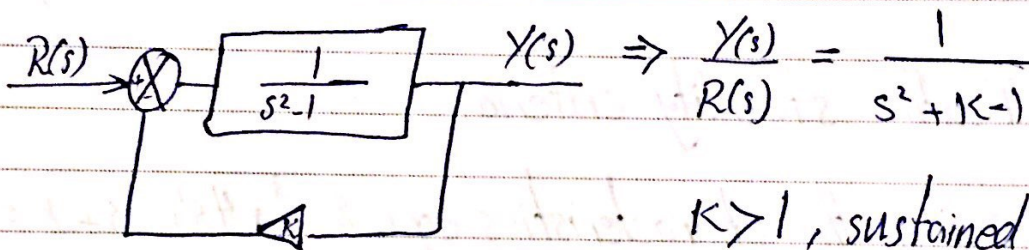
$$G(s) = \frac{1}{s^2 - 1}$$

T.F after reduction

i) Study this system :- the system will be always unstable irrespective of the nature of excitation

so, it's need stabilization

Think of a simple controller such as "negative feedback"



$$\Rightarrow \frac{Y(s)}{R(s)} = \frac{1}{s^2 + K - 1}$$

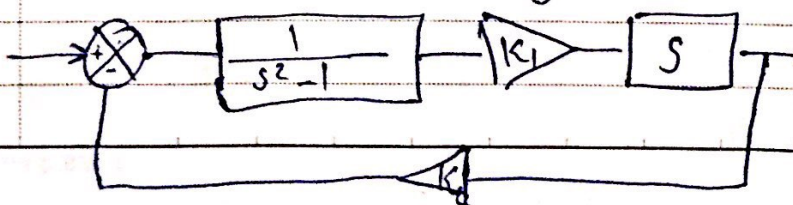
$K > 1$, sustained oscillation

still doesn't work

Conclusion :- Think of diminishing oscillation by damping

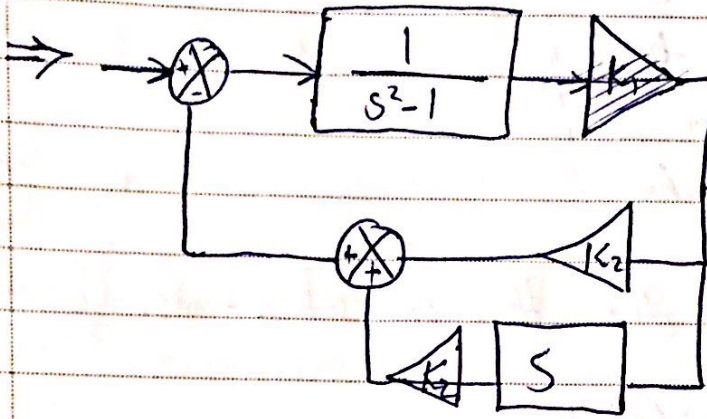
⇒ if one tries a design like this

دamping جابو سبب سبب سبب



$$\Rightarrow \text{so, } \frac{Y(s)}{R(s)} = \frac{K_1 S}{s^2 - 1 + K_1 K_2 S} = \frac{K_1 S}{s^2 + K_1 K_2 S} \text{ (1)}$$

we will cancel this sign to make it stable



\Rightarrow If there is a change in sign or a missing power Thursday 3/10 in the characteristic equation \rightarrow unstable system

\Rightarrow methods based on calculation are not welcomed.

* Symbolic matlab programming :-

\Rightarrow Syms t s a b c d

\Rightarrow A s [a b; c d]

\Rightarrow dA = det(A), invA = inv(A)

$$A_s \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

$$dA_s = a \times d - b \times c$$

$$\text{inv}A_s \begin{bmatrix} \frac{d}{a \times d - b \times c} & \frac{-b}{a \times d - b \times c} \\ \frac{-c}{a \times d - b \times c} & \frac{a}{a \times d - b \times c} \end{bmatrix}$$

\Rightarrow Simplify (invA) % the same answer

⇒ pretty (invA) or ⇒ pretty (simplify (invA))

$$\begin{bmatrix} \frac{d}{ad-bc} & \frac{-b}{ad-bc} \\ \frac{c}{ad-bc} & \frac{a}{ad-bc} \end{bmatrix}$$

⇒ $AT = A'$ % gives the conjugate transpose

$$\begin{bmatrix} \text{conj}(a) & \text{conj}(c) \end{bmatrix}$$

$$\begin{bmatrix} \text{conj}(b) & \text{conj}(d) \end{bmatrix}$$

⇒ transpose(A)

$$\begin{bmatrix} a & c \\ b & d \end{bmatrix}$$

*other examples:-

⇒ $A_s = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix}$, $B = \begin{bmatrix} 0 & 1 \end{bmatrix}$; $C_s = \begin{bmatrix} 5 & 1 \end{bmatrix}$

⇒ $G = C * \text{inv}(s * \text{eye}(2) - A) * B$

⇒ collect(G)

$$\text{ans} = (s+5) / (s^2 + 3s + 2)$$

Sunday 6/10

⇒ Syms $s + \lambda_1 \lambda_2$

⇒ $A = \begin{bmatrix} 0 & 1 \\ -\lambda_1 \lambda_2 & \lambda_1 + \lambda_2 \end{bmatrix}$,

⇒ $[r \ e] = \text{eig}(A)$

$\begin{cases} \text{eigen value} & \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix} \\ \text{eigen vector} & \begin{bmatrix} 1/\lambda_1 & 1/\lambda_2 \\ 1 & 1 \end{bmatrix} \end{cases}$

⇒ $A = \begin{bmatrix} 0 & 1 \\ -\lambda_1 \lambda_2 & \lambda_1 + \lambda_2 \end{bmatrix}$, $[r \ e] = \text{eig}(A)$, $\text{trace}(A)$
 $\text{det}(A)$, $\text{inv}(A)$

diagonal element مجموع

Exo-

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

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

$$G(s) = C [sI_3 - A]^{-1} B + D$$

⇒ $A = \begin{bmatrix} -7 & 3 & 3 \\ -6 & 1 & 4 \\ 0 & 1 & -2 \end{bmatrix}$, $B = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$, $C = \begin{bmatrix} 1 & 0 & 1 \end{bmatrix}$
 $D = 0$, $G = C \text{inv}(s * \text{eye}(3) - A) * B + D$

⇒ collect(G) % the same as simplify(G)

FIVE APPLE

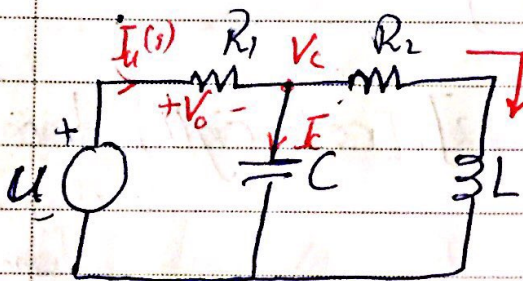
⇒ to ~~obtain~~ obtain time response



* System modeling :-

Tuesday 8/10

i) Input-Output Approach (s-domain) is best illustrated by an electrical example



$$\Rightarrow I_u(s) = \frac{U(s) - V_c(s)}{R_1}$$

$$\Rightarrow I_c = sC V_c(s)$$

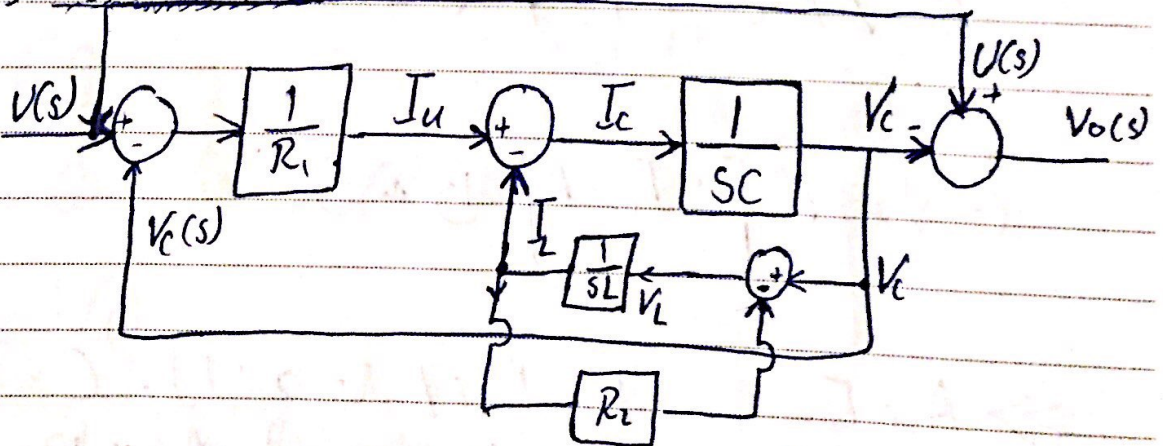
$$\Rightarrow \therefore V_c(s) = \frac{1}{sC} I_c(s)$$

$$\Rightarrow I_L(s) = I_u(s) - I_c(s)$$

$$\Rightarrow V_c(s) = I_L(s) R_2 + V_L(s)$$

$$\Rightarrow V_L(s) = sL I_L(s)$$

$$\Rightarrow V_o(s) = \frac{U(s) - V_c(s)}{R_1}$$



Exo: reduce the above block diagram; obtain $V_o(s)$

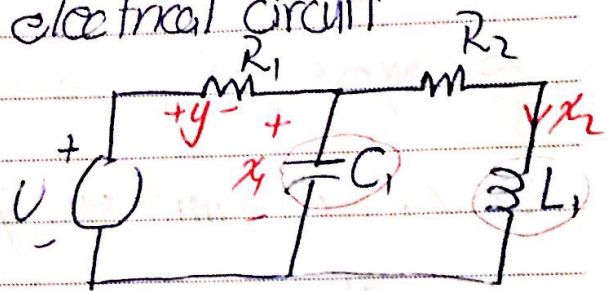
⇒ Simulate the system using $R_1 = 0.5 \Omega$, $R_2 = 4 \Omega$
 $C = 0.2 F$, $L = 0.125 H$

Thursday 10/10

ii) A time domain approach ~~to~~ modeling: the state space approach we write, obtain or get first order differential equation.

best illustrated considering an electrical circuit

let $x_1 = v_C$, $x_2 = i_L$



$$C \dot{x}_1 = \frac{U - x_1}{R_1} - x_2$$

$$L \dot{x}_2 = x_1 - R_2 x_2$$

$$y = U - x_1$$

$$\Rightarrow \dot{x}_1 = 10u - 10x_1 - 5x_2$$

i) substitute $R_1 = 0.5$, $R_2 = 4$

$C = 0.2 F$, $L = 0.125 H$

$$\dot{x}_1 = -10x_1 - 5x_2 + 10u \quad \text{--- (1)}$$

$$\dot{x}_2 = 8x_1 - 32x_2 \quad \text{--- (2)}$$

$$y = -x_1 + u \quad \text{--- (3)}$$

A NO. B

$$\Rightarrow \dot{x} = \begin{bmatrix} -10 & -5 \\ 8 & -32 \end{bmatrix} x + \begin{bmatrix} 10 \\ 0 \end{bmatrix} u$$

C D

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

$$\Rightarrow A = \begin{bmatrix} & \\ & \end{bmatrix}, B = \begin{bmatrix} \\ \end{bmatrix}, C = \begin{bmatrix} & \end{bmatrix}, D = \begin{bmatrix} \end{bmatrix}$$

$$\Rightarrow \text{syms } s$$

$$\Rightarrow G = C * \text{inv}(s * \text{eye}(2) - A) * B + D$$

by using

$$\Rightarrow \text{pretty}(\text{collect}(G))$$

$$\frac{s^2 + 32s + 40}{s^2 + 42s + 360} = G(s)$$

matlab

* if D not equal zero so the order of numerator is the same as the order of denominator

* the poles are -30, -12 $\Rightarrow \text{roots}([2, 42, 360])$

$\Rightarrow \text{poles}(G)$

*

* calculate the eigen values of A

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

$\rightarrow e$: eigen values

(* Obtaining state-space models using diff. equ and T.F)

⇒ N.B.: The state variables are not unique. However the Outputs are.

Conclusion: A system can have many state-space representations (in fact an infinite number).

⇒ First Case: The D.E's do not include derivation of the $u(t)$

Given the following D.E:

$$u(t) = y^{(n)} + a_1 y^{(n-1)} + a_2 y^{(n-2)} + \dots + a_n y, \quad y^{(n)} = \frac{dy^{(n)}}{dt^{(n)}}$$

Since the system is n^{th} order a minimum of n states are needed.

↓

$$\text{let } x_1 = y \quad \longrightarrow \quad \dot{x}_1 = \dot{y} = x_2$$

$$x_2 = \dot{y} \quad \longrightarrow \quad \dot{x}_2 = \ddot{y} = x_3$$

$$x_3 = \ddot{y}$$

$$\vdots$$

$$x_n = y^{(n-1)}$$

$$\dot{x}_n = y^{(n)} = -a_1 y^{(n-1)} - a_2 y^{(n-2)} - \dots - a_n y + u(t)$$

$$= -a_1 x_n - a_2 x_{n-1} - \dots - a_n x_1 + u$$

FIVE APPLE

NO.
stair-case form

$$\Rightarrow \dot{x} = \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \vdots \\ \dot{x}_n \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 1 \\ -a_n & \dots & -a_3 & -a_2 & -a_1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix} u$$

$$y = [1 \ 0 \ 0 \ \dots \ 0] x + 0u$$

→ This known as the controllable form companion phase-variable representation

→ stair-case name relates to matrix A controllable form relate to the special structure of

Tuesday
15/10

A & B

$$\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} \rightarrow \text{non-zero}$$

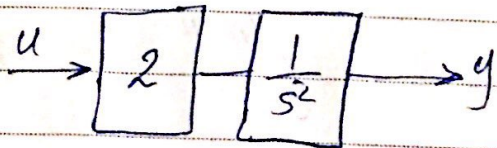
* Example: $y^{(2)} = \frac{dy^2}{dt^2} = 2u$

Soln: $x_1 = y$

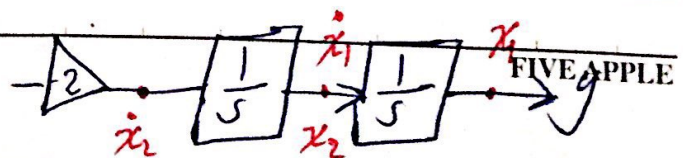
$$x_2 = \dot{y} = \dot{x}_1$$

$$\dot{x}_2 = \ddot{y} = 2u$$

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



$$y = [1 \ 0] x + 0u$$



Example: Given a system described by:

$$y''' + 5y'' + 9y' + 5y = 4u$$

y is the output of the system

$$\Rightarrow x_1 = y$$

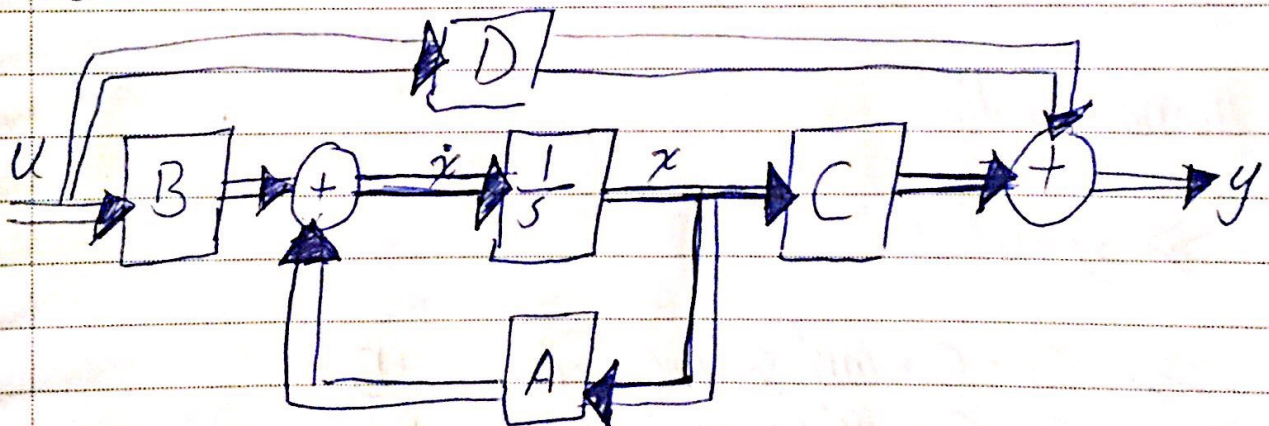
$$x_2 = \dot{y} = \dot{x}_1$$

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

$$x_4 = \ddot{\dot{y}} = \dot{x}_3 = -5x_1 - 9x_2 - 5x_3 + 4u$$

$$\dot{x} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -5 & -9 & -5 \end{bmatrix} x + \begin{bmatrix} 0 \\ 0 \\ 4 \end{bmatrix} u$$

$$y = [1 \ 0 \ 0] x + 0u$$



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

$$y = Cx + Du$$

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

$$y = \begin{bmatrix} 5 & 1 \end{bmatrix} x$$

Obtain the differential equ. describing the system

ii) $\dot{x} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -6 & 11 & -6 \end{bmatrix} x + \begin{bmatrix} 0 \\ 0 \\ -2 \end{bmatrix} u$

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

⇒ The transfer function matrix is given by

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

using matlab

⇒ `syms s`

⇒ `G = C * inv(s * eye(2) - A) * B + D`

or ⇒ `G = C * inv(s * eye(size(A)) - A) * B + D`

$$G(s) = \frac{s+5}{s^2+3s+2}$$

or ⇒ `[n, d] = ss2tf(A, B, C, D)`, ~~tf~~ `g2 = tf(n, d)`

$$g2 = \frac{s+5}{s^2+3s+2}$$

$$\Rightarrow n_s [5 \ 1]; d = [1 \ -3 \ 2]; [A, B, C, D] = \begin{bmatrix} \dots \\ \dots \\ \dots \\ \dots \\ \dots \end{bmatrix}$$

20/10 Sunday

* state space analysis :- (ch 9)

⇒ Give a system described by

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

taking the laplace transform (all initial conditions are zero)

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

Some of

⇒ Convenient forms are the following :-

i) Controllable companion form

stair-case

$$\dot{x} = \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 1 \\ -a_n & -a_{n-1} & -a_{n-2} & \dots & -a_1 \end{bmatrix} x + \begin{bmatrix} 0 \\ \vdots \\ 0 \\ 1 \end{bmatrix} u$$

$$y = [b_n - a_n b_0 \quad b_{n-1} - a_{n-1} b_0 \quad \dots \quad b_1 - a_1 b_0] x + b_0 u$$

ii) Observable Companion Form

$$\dot{x} = \begin{bmatrix} 0 & 0 & 0 & \dots & 0 & -a_n \\ 1 & 0 & 0 & \dots & 0 & -a_{n-1} \\ 0 & 1 & 0 & \dots & 0 & -a_{n-2} \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & \dots & 1 & -a_1 \end{bmatrix} x + \begin{bmatrix} b_n - a_n b_0 \\ \vdots \\ b_1 - a_1 b_0 \end{bmatrix} u$$

$$y = \begin{bmatrix} 0 & \dots & \dots & \dots & 0 \end{bmatrix} x + b_0 u$$

N.B. For both forms the T.F. $G(s) = C(sI_n - A)^{-1}B$ is identically the same

40

$$\Rightarrow A = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -6 & 11 & -6 \end{bmatrix}, \text{Poly}(A)$$

$$\Rightarrow A_2 = \begin{bmatrix} -7 & 3 & 3 \\ -6 & 1 & 4 \\ 0 & 1 & -2 \end{bmatrix}, \text{Poly}(A_2)$$

Model Form (Digital and/or Block diagonal Form): Tuesday 22/10
 \Rightarrow Best illustrated by an example

Given the system $G(s) = \frac{2s^2 + 3s + 4}{s^2 + 9s + 20} = \frac{Y(s)}{U(s)}$

$$\therefore G(s) = 2 + \frac{-15s - 36}{(s+4)(s+5)}$$

$$= \frac{24}{s+4} - \frac{39}{s+5} + 2$$

$$\begin{array}{r} 2 \\ s^2 + 9s + 20 \overline{) 2s^2 + 3s + 4} \\ \underline{-2s^2 - 18s + 40} \\ -15s - 36 \end{array}$$

$$\Rightarrow Y(s) = \frac{24 u(s)}{s+4} + \frac{-39 u(s)}{s+5} + 2 u(s)$$

$x_1(s)$ $x_2(s)$

$$\textcircled{1} x_1(s) = \frac{24}{s+4} u(s) \Rightarrow s x_1(s) + 4 x_1(s) = 24 u(s)$$

Take lablace inverse $\mathcal{L}^{-1}(s x_1(s) + 4 x_1(s) = 24 u(s))$

$$\Rightarrow \dot{x}_1 + 4 x_1 = 24 u$$

$$\sim \boxed{\dot{x}_1 = -4 x_1 + 24 u}$$

$$\textcircled{2} x_2(s) = \frac{-39}{s+5} u(s) \Rightarrow s x_2(s) + 5 x_2(s) = -39 u(s)$$

$$\mathcal{L}^{-1}(s x_2(s) + 5 x_2(s) = -39 u(s)) = \boxed{\dot{x}_2 = -5 x_2 - 39 u}$$

$$\therefore Y(s) = X_1(s) + X_2(s) + 2 U(s)$$

$$y(t) = x_1(t) + x_2(t) + 2 u(t)$$

$$\dot{x} = \begin{bmatrix} -4 & 0 \\ 0 & -5 \end{bmatrix} x + \begin{bmatrix} 24 \\ -39 \end{bmatrix} u$$

eigen values \rightarrow if this is zero uncontrollability

* To check validity of representation

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

12] The Jordan Form δ - (needed when the poles are identical)

Suppose δ :-
$$Y(s) = \frac{n_1}{U(s)} + \frac{n_2}{(s+P_1)^2} + \frac{n_3}{s+P_1}$$

$$Y(s) = \frac{n_1 U}{(s+P_1)^3} + \frac{n_2 U}{(s+P_1)^2} + \frac{n_3 U}{(s+P_1)}$$

$X_1 \quad X_2 \quad X_3$

Let $Y(s) = n_1 X_1 + n_2 X_2 + n_3 X_3$

① where $X_3 = \frac{U}{s+P_1} \Rightarrow \dot{X}_3 = -P_1 X_3 + U \quad \dots \textcircled{1}$

② $X_2 = \frac{U}{(s+P_1)^2} = \frac{1}{(s+P_1)} * \frac{U}{(s+P_1)} = \frac{1}{s+P_1} X_3(s)$

$\mathcal{L}^{-1}((s+P_1)X_2(s) = X_3(s)) \Rightarrow \dot{X}_2 = -P_1 X_2 + X_3 \quad \dots \textcircled{2}$

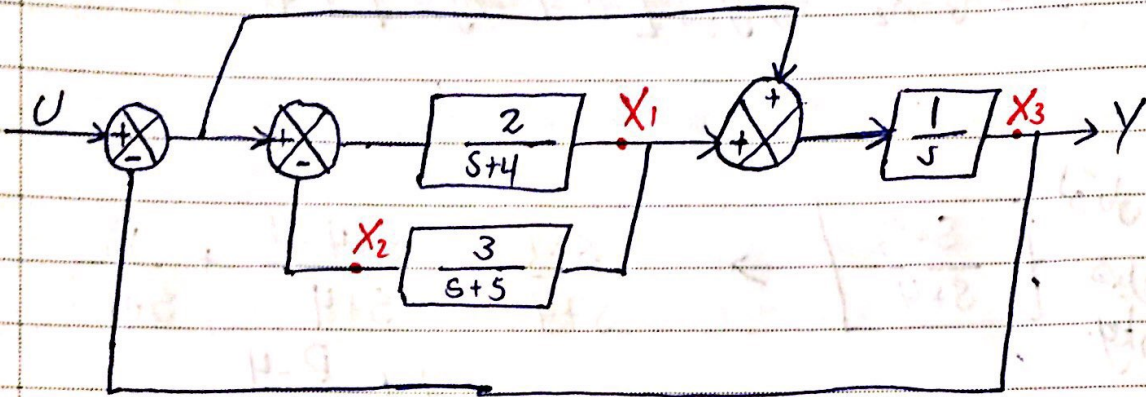
③ $X_1 = \frac{U}{(s+P_1)^3} = \frac{1}{(s+P_1)} * \frac{U}{(s+P_1)^2} = \frac{1}{(s+P_1)} X_2(s)$

$\Rightarrow \dot{X}_1 = -P_1 X_1 + X_2 \quad \dots \textcircled{3}$

$$\dot{x} = \begin{bmatrix} -P_1 & 1 & 0 \\ 0 & -P_1 & 1 \\ 0 & 0 & -P_1 \end{bmatrix} x + \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} U$$

$y = [n_1 \ n_2 \ n_3] x + 0 U$

⇒ Obtaining a state space model from a block diagram - don't attempt to reduce the diagram. Best illustrated by an example:-



$$\Rightarrow X_3 = \frac{1}{s} (X_1 + (U - X_3))$$

$$sX_3 = X_1 + U - X_3 \Rightarrow \dot{x}_3 = x_1 - x_3 + u$$

$$\Rightarrow X_1 = \frac{2}{s+4} (U - X_3 - X_2)$$

$$sX_1 + 4X_1 = -2X_2 - 2X_3 + 2U \Rightarrow \dot{x}_1 = -4x_1 - 2x_2 - 2x_3 + 2u$$

$$\Rightarrow X_2 = \frac{3}{s+5} (X_1) \Rightarrow \dot{x}_2 = 3x_1 - 5x_2$$

$$\Rightarrow Y = X_3$$

$$\therefore \dot{x} = \begin{bmatrix} -4 & -2 & -2 \\ 3 & -5 & 0 \\ 1 & 0 & -1 \end{bmatrix} x + \begin{bmatrix} 2 \\ 0 \\ 1 \end{bmatrix} u$$

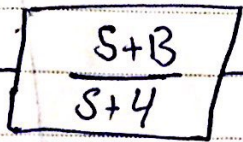
* هذا يختلف بفرض ال s
والرقم لازم ال diagonal
فرض ال s ونفس ال abt

$$y = [0 \ 0 \ 1] x + 0u$$

⇒ For the

$$y_2 = u - x_3 - x_2 \Rightarrow y_2 s [0 \ -1 \ -1] x + 1 \cdot u$$

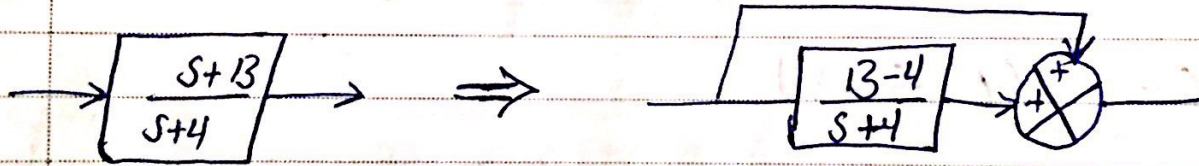
⇒ block division
 لول اول اول اول
 Integ. $\frac{1}{s}$



$$\begin{aligned} \frac{s+3}{s+4} &= \frac{s+4-1}{s+4} + \frac{3}{s+4} \\ &= 1 + \frac{3-4}{s+4} \end{aligned}$$

⇒ Variation :-

1) if having $\frac{s+3}{s+4} = 1 + \frac{3-4}{s+4}$



Exercises :- det state b as shown. check your answer using trans det and T.F In diagram.

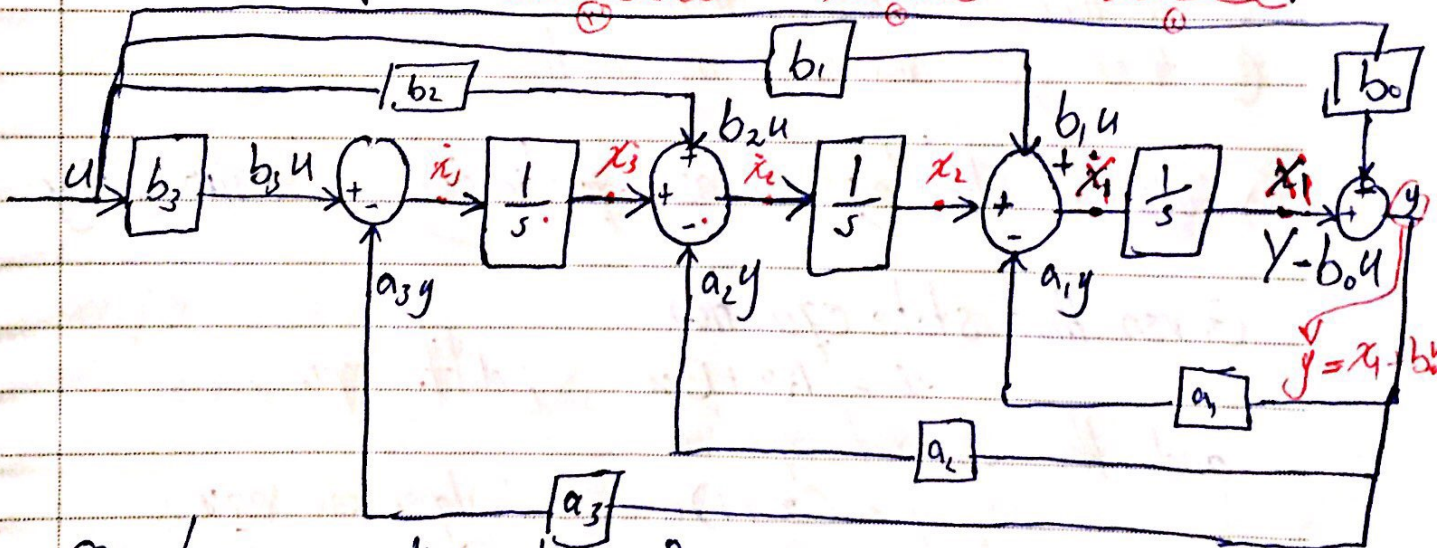
⇒ Obtaining an S.S model using T.F. :-

best illustrated by an example

$$G(s) = \frac{Y(s)}{U(s)} = \frac{b_3 s^3 + b_2 s^2 + b_1 s + b_0}{s^3 + a_1 s^2 + a_2 s + a_3}$$

It can be shown that :- $(Y(s) - b_0 U(s))$

$$\begin{aligned} (Y(s) - b_0 U(s)) &= \frac{1}{s^3} ((b_2 u - a_1 y) s^2 + (b_1 u - a_2 y) s + (b_3 u - a_3 y)) \\ &= \frac{1}{s} ((b_1 u - a_1 y) + (b_2 u - a_2 y) \frac{1}{s} + (b_3 u - a_3 y) \frac{1}{s^2}) \\ &= \frac{1}{s} \left(\frac{1}{s} [(b_3 u - a_3 y) \frac{1}{s} + (b_2 u - a_2 y)] + (b_1 u - a_1 y) \right) \end{aligned}$$



⇒ Questions may ask you of exam :-

① given a transfer function obtain a block diagram only

② and use it to find state space representation

$$\Rightarrow \dot{x} = \begin{bmatrix} -a_1 & 1 & 0 \\ -a_2 & 0 & 1 \\ -a_3 & 0 & 0 \end{bmatrix} x + \begin{bmatrix} b_1 - a_1 b_0 \\ b_2 - a_2 b_0 \\ b_3 - a_3 b_0 \end{bmatrix} u$$

$$y = [1 \ 0 \ 0] x + b_0 u$$

→ Check the answer using matlab

→ syms a1 a2 a3 b0 b1 b2 b3 s

→ A = [], B = [], C = [], D = []

→ G = C * inv(s * eye(size(A)) - A) * B + D

→ pretty(collect(G))

Ex: given $G(s) = \frac{2s^2 + 3s + 4}{s^2 + 5s + 6}$

① obtain a SS representation

② check your answer using matlab

* Solution of state-space Equations *

29/10 Tuesday

Given that state equation

$$\dot{x} = Ax + Bu \quad ; \text{ diff. equ.}$$

and the Output equation

$$y = Cx + Du \quad ; \text{ algebraic equ}$$

it can be shown that $x(t) = e^{A(t-t_0)} x(t_0) + \int_{t_0}^t e^{A(t-\tau)} B u(\tau) d\tau$

where e^{At} is known as the exponential matrix behind

it

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

not suitable numerically

$$\text{or } \boxed{e^{At} = \mathcal{L}^{-1} (sI - A)^{-1}} ; \text{ given a closed form solution}$$

*Exo: det e^{At} using as many methods as you know, given $\dot{x} = Ax + Bu$ where $A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$; $A = \begin{bmatrix} 0 & 1 \\ 0 & 1 \end{bmatrix}$; $A = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix}$ 31/10 Thursday

check your answer using matlab command `expm()`

$$\Rightarrow A = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix}$$

$$\Rightarrow g = \text{expm}(A*t)$$

*Properties of the exponential matrix e^{At}

$$1) e^{At} \Big|_{t=0} = I_n, \text{ used to check regarding the validity of } e^{At}$$

$$2) |e^{At}| > 0$$

$$3) e^{A(t_1+t_2)} = e^{At_1} \cdot e^{At_2}, \quad t = \text{scaler}$$

$$4) e^{(A+M)t} = e^{At} \cdot e^{Mt}, \quad \text{Provided } AM = MA \text{ of } B, A \text{ matrix}$$

$$5) [e^{At}]^{-1} = e^{A(-t)}$$

$$6) \frac{d}{dt} (e^{At}) = A e^{At} = e^{At} \cdot A$$

$$7) \int e^{At} dt = [e^{At} - I_n] \cdot A^{-1} ; \text{ inverse of } A \text{ الـعكس } A^{-1}$$

$$8) \text{trace } e^{At} > 0$$

$$9) e^{(At)^n} = e^{A(tn)}$$

$$10) e^{A(t-t_2)} \cdot e^{A(t_2-t_1)} = e^{A(t-t_1)}$$

in the book $\phi(t) = e^{At}$, $\phi(t)$ is known as transition matrix

(* Time Solution due to step input *)

3/11 Sunday

$$\Rightarrow \text{let } \overset{dp}{u(t)} = K u(t)$$

$$\text{then } \boxed{x(t) = e^{At} x(0) + e^{At} \int_0^t e^{-A\tau} B u(\tau) d\tau} \quad \underline{\underline{\text{في}}}$$

\Rightarrow to prove the above equ.:

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

$$e^{-At} \dot{x} = e^{-At} Ax + e^{-At} Bu$$

$$\underline{\underline{e^{-At}(\dot{x}) - e^{-At} Ax = e^{-At} Bu}}$$

$$\frac{d}{dt} (e^{-At} x) = e^{-At} B \cdot u$$

Integrate from $0 \rightarrow t$

$$e^{-At} x(t) \Big|_0^t = \int_0^t e^{-A\tau} B u(\tau) d\tau$$

→ It can be shown that (see book)

$$x(t) = e^{At} x(0) + [e^{At} - I_n] A^{-1} B K \rightarrow \text{const.}$$

→ If the system is asymptotically stable and matrix A is invertible then:

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

For step input

→ It can be also shown that the response due to ramp input of value $t u(t) \rightarrow \text{const.}$

$$x(t) = e^{At} x(0) + [e^{At} A^{-2} - A^{-2} - t A^{-1}] V$$

(N.B.) For a time-invariant system (the ones we study)

$$if \quad u(t) \longrightarrow x(t)$$

$$then \quad \frac{d}{dt} u(t) \longrightarrow \dot{x}(t)$$

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

Ex: given $\dot{x} = \begin{bmatrix} 0 & 1 \\ -2 & 3 \end{bmatrix} x + \begin{bmatrix} 0 \\ 6 \end{bmatrix} u$, calculate $x(t)$ due to a unit step excitation using as many methods as you

known. check x_{ss} your including matlab, Plot relevant graphs.

5/11 Tuesday

Example^o - $\dot{x} = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} x + \begin{bmatrix} 0 \\ 4 \end{bmatrix} u$, a asymptotically stable

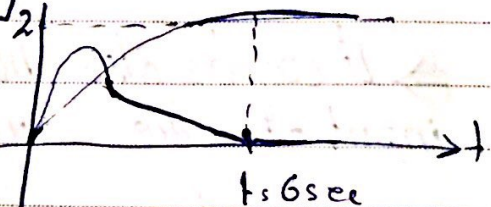
since $\lambda[A] = \{-1, -2\} \Rightarrow$ a steady state value due to input exist ^{step}

Simulate using state space block with $C = I_2$ and $D = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$



by matlab

\Rightarrow you will get something like that



by cal

$$\Rightarrow x_{ss} = -A^{-1} B R = \begin{bmatrix} 2 \\ 0 \end{bmatrix} \Rightarrow x_{ss1} = 2 \\ x_{ss2} = 0$$

(x) Determination of eigenvalues of a square matrix^o

\Rightarrow Given a transformation $T(\cdot)$ & a non-zero vector x

\Rightarrow what is x such that $T(x) \propto x \stackrel{i.e.}{\Rightarrow} T(x) = \lambda x$

\Rightarrow If the transformation is linear then it can be represented by a matrix leading to $Ax = \lambda x$

$$\Rightarrow Ax = \lambda x = x \lambda$$

$$[A - \lambda I_n] x = 0$$

$$\text{inverse} \quad |A - \lambda I_n| = 0$$

\rightarrow gives a monic n^{th} order polynomial
solved by eigen value λ

⇒ to determine the eigen vectors solve the singular system of equations:

$$[A - \lambda_i I_n] x_i = 0$$

Use Gauss-Elimination method

Ex^o - let $A = \begin{bmatrix} -5 & 3 & 3 \\ -6 & 3 & 4 \\ 0 & 1 & 0 \end{bmatrix} \Rightarrow |A - \lambda I_3| = \begin{vmatrix} -5-\lambda & 3 & 3 \\ -6 & 3-\lambda & 4 \\ 0 & 1 & 0-\lambda \end{vmatrix} = 0$

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

$$(\lambda+1)(\lambda-1)(\lambda+2) = 0$$

$$\lambda_1 = -1, \lambda_2 = 1, \lambda_3 = -2 \quad ; \text{eigen values}$$

N.B.:- \sum eigen values = negative coefficient of second highest order only if the coefficient of highest power is 1

$$\lambda_1 + \lambda_2 + \lambda_3 = -2 \quad \checkmark$$

⇒ to determine the eigen vector associated with $\lambda = -2$

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

$m = \frac{-6}{-3} = 2$

Pivot 1 $\begin{bmatrix} -3 & 3 & 3 \\ 0 & -1 & -2 \\ 0 & 1 & 2 \end{bmatrix} x = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \Rightarrow \begin{bmatrix} -3 & 3 & 3 \\ 0 & -1 & -2 \\ 0 & 0 & 0 \end{bmatrix} x = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$

$m = \frac{1}{-1} = -1$

diagonal elements

stop ← zero

$$\Rightarrow \text{let } \boxed{x_3 = r}$$

$$-x_2 - 2x_3 = 0 \Rightarrow \boxed{x_2 = -2r}$$

$$-3x_1 + 3x_2 + 3x_3 = 0 \Rightarrow \boxed{x_1 = -r}$$

$$\therefore x = \begin{bmatrix} -r \\ -2r \\ r \end{bmatrix}, x \neq 0 \Rightarrow r \neq 0$$

$$\text{let } r = -1 \Rightarrow x = \begin{bmatrix} 1 \\ 2 \\ -1 \end{bmatrix}$$

Exo: Calculate the remaining two eigen vector

* Determination of eigen vectors using the adjoint method

\Rightarrow best illustrated by an example \odot

$$\text{Suppose } \text{adj}(A - \lambda_3 I_3) = \text{adj} \begin{bmatrix} -3 & 3 & 3 \\ -6 & 5 & 4 \\ 0 & 1 & 2 \end{bmatrix} = \begin{bmatrix} 6 & -3 & -3 \\ 12 & -6 & -6 \\ -6 & 3 & 3 \end{bmatrix}$$

$$= \begin{bmatrix} 1 \\ 2 \\ -1 \end{bmatrix}$$

N.B: Revise calculating the adjoint

⇒ Calculation e^{At} using eigen vectors example :-

let V be an $n \times n$ matrix representing the eigen vector of A ; it can be shown that :-

$$e^{At} = V e^{\Delta t} V^{-1} = V * \begin{bmatrix} e^{\lambda_1 t} & & 0 \\ & \ddots & \\ 0 & & e^{\lambda_n t} \end{bmatrix} * V^{-1}$$

⇒ only valid when the eigen values are distinct (different)

⇒ $A = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix}$

⇒ $[r, e] = \text{eig}(A)$

⇒ syms t

⇒ $\exp A = r * \exp m(e * t) * \text{inv}(r)$

Example:- let $A = \begin{bmatrix} 0 & 1 \\ 2 & -3 \end{bmatrix}$, it can be shown that

$$\Delta = \begin{bmatrix} -1 & 0 \\ 0 & -2 \end{bmatrix} \text{ or } \begin{bmatrix} -2 & 0 \\ 0 & -1 \end{bmatrix}, \text{ in which case}$$

$$V = \begin{bmatrix} 1 & 1 \\ -1 & -2 \end{bmatrix} \text{ or } \begin{bmatrix} +1 & +1 \\ -2 & -1 \end{bmatrix}$$

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

* Generalized eigen vectors :-

motivation: let $A = \begin{bmatrix} 0 & 1 \\ -4 & 4 \end{bmatrix} \Rightarrow \lambda_1 = \lambda_2 = 2$

eigen vector associated with $\lambda_1 = 2$

To check

λ جمع = diagonal
 $\lambda \times \lambda = \det A$

$$V_1 = \text{adj} \left(\begin{bmatrix} -2 & 1 \\ -4 & 2 \end{bmatrix} \right) = \begin{bmatrix} 2 & -1 \\ 4 & -2 \end{bmatrix} \Rightarrow V_1 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

For the second eigen value we calculate a generalized eigen vector as follows :-

$$[A - \lambda_1 I] w_2 = w_1$$

$$[A - \lambda_2 I]^2 w_2 = 0 \quad ; \text{ remember } |[A - \lambda I]| = 0$$

inverse $\hat{=}$ Lib use

$$\text{So, } \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} w_2 = 0$$

w_2 any vector independent of V_1 ; so, w_2 may be ~~any number~~ $w_2 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$

Example: let $A = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 27 & -27 & 9 \end{bmatrix}$, $\lambda[A] = \{3, 3, 3\}$

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

$$r = \begin{bmatrix} 0.1048 & 0.1048 & 0.1048 \\ 0.3145 & 0.3145 & 0.3145 \\ 0.9435 & 0.9435 & 0.9435 \end{bmatrix}$$

$$V_1 = V_2 = V_3$$

so, it doesn't work

$$e = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 3 \end{bmatrix}$$

$\Rightarrow [r, e] = \text{Jordan}(A)$

$$r = \begin{bmatrix} 9 & -3 & 1 \\ 27 & 0 & 0 \\ 81 & 27 & 0 \end{bmatrix}$$

eigen vector

generalized eigen vector

\Rightarrow if vectors can be constitutes square matrix then they are independent only if the determine of that matrix \neq zero

$$e = \begin{bmatrix} 3 & 1 & 0 \\ 0 & 3 & 1 \\ 0 & 0 & 3 \end{bmatrix}, \text{ called Jordan Form}$$

e_i

* Independent of eigen vectors. or vectors in general :-

⇒ independence of vectors easily judged algebraically as follows

- 1) generate a matrix contains those vectors
- 2) look for a square sub matrix having non-zero determinant
- 3) If the determinant not zero then the vectors are independent

Ex: $M_1 = [v_1 \ v_2] = \begin{bmatrix} -1 & -5 \\ -2 & -10 \end{bmatrix}$

$|-5| \neq 0 \Rightarrow$ at least One independent eigen vector (IV)

$\begin{vmatrix} +1 & -5 \\ -2 & -10 \end{vmatrix} = 0 \Rightarrow$ no 2 IV

$M_2 = [v_1 \ v_2] = \begin{bmatrix} -1 & -2 \\ -4 & 8 \\ -2 & 4 \end{bmatrix}$

⇒ no 2×2 submatrix of non zero determinant

v_1 & v_2 are ~~in~~ dependent

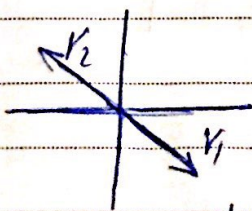
$M_3 = [v_1 \ v_2] = \begin{bmatrix} +1 & -2 \\ -4 & 8 \\ 2 & 4 \end{bmatrix}$

⇒ there exist a 2×2 sub matrix of a non-zero determinant

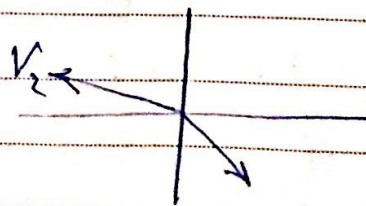
v_1 & v_2 are independent

* Theorem For a square matrix to have an inverse, the columns should be independent.

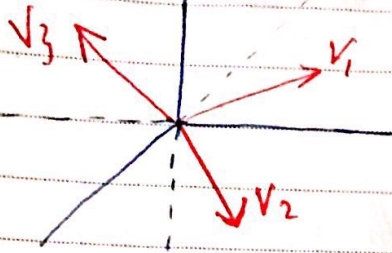
⇒ Geometrically :-



"dependent"



"independent"



\Rightarrow if all v_1, v_2, v_3 are in plane \Rightarrow dependent

\Rightarrow if at least one of them is not on a plane containing the other \Rightarrow independent

* Similarity transformation: a method to get a special form of matrices (\bar{A}) of matrix (A)

Ex: what are the eigen values and the determinants of $A = \begin{bmatrix} -7 & 3 & 3 \\ -6 & 1 & 4 \\ 0 & 1 & -2 \end{bmatrix}$ and $A^{-1} = P^{-1}AP$ where $P = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 2 & 1 \\ 1 & -1 & -1 \end{bmatrix}$

by similarity $\bar{A} = \begin{bmatrix} -1 & 0 & 0 \\ 0 & -4 & 0 \\ 0 & 0 & -3 \end{bmatrix}$ $\det(A) = \det(\bar{A}) = -1 \times -4 \times -3 = -12$
 $\lambda(A) = \{-1, -4, -3\}$

* Jordan Forms:

Thursday

12/11/2019

certain matrices of repeated eigen values can have the following forms after similarity transformation

Ex: A matrix 3×3 may have three identical eigen values say $-2, -2, -2$, when transformed of the following form may result:

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

where P contain eigen vectors and generalized eigen vectors as columns

⇒ \bar{A} may look like $\begin{bmatrix} -2 & 1 & 0 \\ 0 & -2 & 1 \\ 0 & 0 & -2 \end{bmatrix}$ ✓ not diagonal matrix

or like this $\begin{bmatrix} -2 & 1 & 0 \\ 0 & -2 & 0 \\ 0 & 0 & -2 \end{bmatrix}$ or like this $\begin{bmatrix} -2 & 0 & 0 \\ 0 & -2 & 1 \\ 0 & 0 & -2 \end{bmatrix}$ ✗

or this extreme case $\begin{bmatrix} -2 & 0 & 0 \\ 0 & -2 & 0 \\ 0 & 0 & -2 \end{bmatrix}$

* Exponential matrix of diagonal and Jordan forms:

⇒ If $A = \begin{bmatrix} \lambda_1 & & 0 \\ & \lambda_2 & \\ 0 & & \lambda_3 \end{bmatrix}$ then $e^{At} = \begin{bmatrix} e^{\lambda_1 t} & 0 & 0 \\ 0 & e^{\lambda_2 t} & 0 \\ 0 & 0 & e^{\lambda_3 t} \end{bmatrix}$

⇒ If $A = \begin{bmatrix} \lambda_1 & 1 \\ 0 & \lambda_1 \end{bmatrix}$ then $e^{At} = \begin{bmatrix} e^{\lambda_1 t} & t e^{\lambda_1 t} \\ 0 & e^{\lambda_1 t} \end{bmatrix}$

⇒ If $A = \begin{bmatrix} \lambda & 1 & 0 \\ 0 & \lambda & 1 \\ 0 & 0 & \lambda \end{bmatrix}$ then $e^{At} = \begin{bmatrix} e^{\lambda t} & t e^{\lambda t} & \frac{t^2}{2} e^{\lambda t} \\ 0 & e^{\lambda t} & t e^{\lambda t} \\ 0 & 0 & e^{\lambda t} \end{bmatrix}$

Exo- Use properties of matrices to evaluate e^{At} when A is:

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

$$\textcircled{2} A = \begin{bmatrix} 4 & 0 & 0 \\ 0 & 3 & 1 \\ 0 & 0 & 3 \end{bmatrix}$$

$$\textcircled{3} A = \begin{bmatrix} -3 & 1 & 0 \\ 0 & -3 & 1 \\ 0 & 0 & -3 \end{bmatrix}$$

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

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

* The Cayley-Hamilton Theorem (CH theorem) :- Very Useful

⇒ given the characteristic polynomial $p(\lambda)$ of a matrix

$$\begin{aligned} \text{i.e. } p(\lambda) &= |A - \lambda I_n| = |\lambda I_n - A| = 0 \\ &= \lambda^n + a_1 \lambda^{n-1} + a_2 \lambda^{n-2} + \dots + a_n = 0 \end{aligned}$$

then $P(A) = 0$

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

i.e. every square matrix satisfies its characteristic equation polynomial

$$\Rightarrow \text{in MATLAB} \circ \begin{bmatrix} 2 & 3 \\ 4 & 5 \end{bmatrix}^{-1} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \quad \text{no error in matlab}$$

⇒ in linear algebra :- cannot be done undefined operation

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

$$|A - \lambda I| = \lambda^2 - 3\lambda - 10 = 0 \Rightarrow 1, -3, -10$$

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

Hence $A^2 - 3A - 10I_2 = O_{2 \times 2}$

$\Rightarrow A^2 - 3A - 10 \times \text{eye}(2)$

ans = $\begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$

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

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

ans =

$18 \quad 19 \quad 12 \quad ; \therefore \text{CE} = \lambda^3 + 8\lambda^2 + 19\lambda + 12 = 0$

$\Rightarrow A^3 + 8A^2 + 19A + 12$

ans $\begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$

* If we want to calculate A^3 :-

1) $A^3 = -8A^2 - 19A - 12$

2) by matlab to check your answer A^3 & A^3

⇒ if you want $A^4 = -8A^3 - 19A^2 - 12A$

$$A^4 = -8A^3 - 19A^2 - 12A$$

$$= -8(-8A^2 - 19A - 12I_3) - 19A^2 - 12A$$

$$A^4 = 45A^2 + 140A + 96I_3 \quad \checkmark$$

Exercises: Use the CH theorem to calculate A^{-1}

solⁿ:

$$A^{-1} \times (A^3 + 8A^2 + 19A + 12I_3) = 0$$

$$A^2 + 8A + 19I_3 + 12A^{-1} = 0$$

$$\therefore A^{-1} = -\frac{A^2 + 8A + 19I_3}{12}$$

Tuesday 19/11/2019

* Eigen values properties and characteristics: (See notes)

① $\lambda[M^T] = \lambda[M] = \{\lambda_1, \lambda_2, \dots, \lambda_n\}$; IF $M = M^T$ eigen values are real numbers

② $\lambda[\alpha M] = \alpha \lambda[M]$

③ $\lambda[MM] = \lambda[NM]$

④ $\lambda[P^{-1}MP] = \lambda[M]$; Similarity transformation

⑤ $\lambda[M^{-1}] = \frac{1}{\lambda[M]}$

⑥ $\lambda[M + \alpha I] = \lambda[M] + \alpha$

* Structural properties of System's Stability

⇒ a stable system doesn't grow without bound due to a bounded excitation

⇒ Stability of linear systems is determined by the eigen values of A

⇒ Demonstration

Unit Step	Different A Certain B, C, D	Scope
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check stability visually and using $\text{eig}(A)$

⇒ -ve real part of $\lambda \in A$ then the system is stable Sunday 24/11

$$\text{Ex: } \dot{x} = \begin{bmatrix} -5 & 3 & 3 \\ -6 & 3 & 4 \\ 0 & 1 & 0 \end{bmatrix} x + \begin{bmatrix} 5 \\ 2 \\ 1 \end{bmatrix} u$$

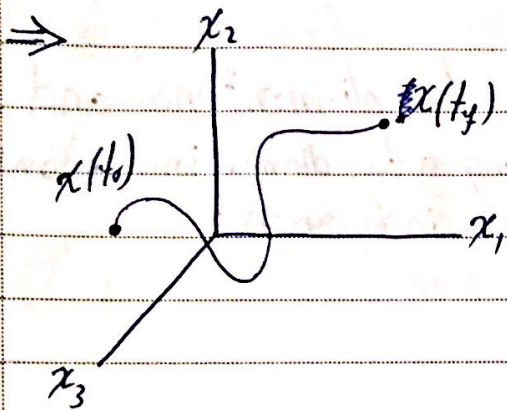
$\text{trace}(A) = -5 + 3 + 0 = -2$; not conclusive to determine stability

Suppose we change -5 to -2

$\text{trace}(A) = -2 + 3 + 0 = 1$; unstable because there at least a positive real part
conclusive decision

* Controllability :-

⇒ A system is called controllable if it is possible to move the states of the system $x(t_0)$ at $t = t_0$ to any arbitrary final state $x(t_f)$ in finite time using an unconstrained control law



$x(t_0)$ & $x(t_f)$ are not contained within a plane
⇒ controllable in \mathbb{R}^3

Sunday 11/12/2019

⇒ Test for controllability (CC)

Def^o: n vectors v_1, v_2, \dots, v_n are independent

$$\text{if } c_1 v_1 + c_2 v_2 + \dots + c_n v_n = 0$$

$$\text{implies } c_1 = c_2 = \dots = c_n = 0$$

Rank^o: The maximum number of independent columns of an matrix $n \times m$

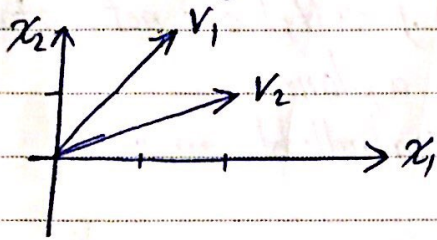
$$\text{Ex^o: } v_1 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}, v_2 = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

$$\text{Solve } \begin{bmatrix} 1 \\ 2 \end{bmatrix} c_1 + \begin{bmatrix} 2 \\ 1 \end{bmatrix} c_2 = 0 \Rightarrow \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = 0$$

$$\Rightarrow \frac{1}{3} \begin{bmatrix} 1 & -2 \\ -2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \frac{1}{3} \begin{bmatrix} 1 & -2 \\ -2 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad \text{inv. b. l. j. p. *}$$

$$I_2 \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} c_2 \\ c_1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad ; \text{ then they are independent}$$

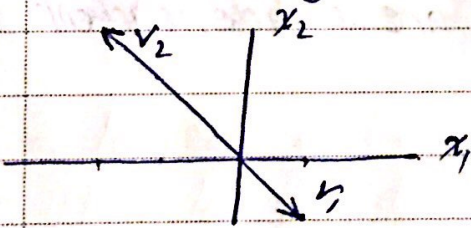
\Rightarrow geometrically :-



being two dimensional and
Spanning a two dimensional plane
hence independent.

$$\text{Ex}^o \quad v_1 = \begin{bmatrix} 1 \\ -2 \end{bmatrix} \quad , \quad v_2 = \begin{bmatrix} -2 \\ 4 \end{bmatrix}$$

\Rightarrow geometrically :-



dependent

\Rightarrow mathematically :-

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

satisfied For all $c_1 = 2c_2 \neq 0 \Rightarrow$ dependent

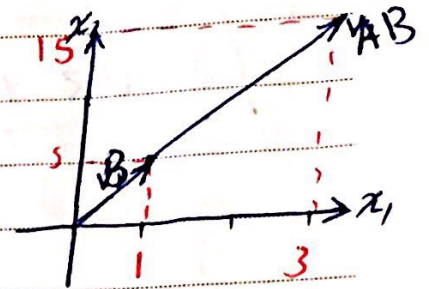
NO. Controllability

[1] The Rank test: A system is completely (CC) iff
 $\text{rank}(M) = \text{rank}([B \ AB \ A^2B \ \dots \ A^{n-1}B]) = n$

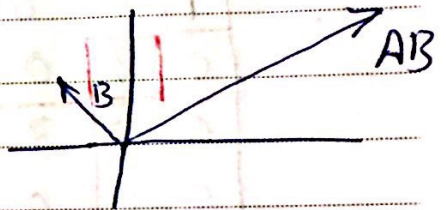
Exo- $\dot{x} = \begin{bmatrix} -1 & 2 \\ 3 & 4 \end{bmatrix} x + \begin{bmatrix} 1 \\ 3 \end{bmatrix} u$

Solution: $M = [B \ AB] = \begin{bmatrix} 1 & 5 \\ 3 & 15 \end{bmatrix}$

$\text{rank}(M) = 1 \Rightarrow$ uncontrollable



If $B = \begin{bmatrix} -1 \\ 3 \end{bmatrix} \Rightarrow \text{rank}(M) = \text{rank} \begin{bmatrix} -1 & 7 \\ 3 & 9 \end{bmatrix} = 2$ controllable



Ex. $\text{rank} \begin{bmatrix} 1 & 2 \\ 0 & 2 \end{bmatrix} = 2$

② $\text{rank} \begin{bmatrix} -7 & 21 \\ 8 & -24 \end{bmatrix} = 1$

③ $\text{rank} \begin{bmatrix} 4 & -8 & 4 \\ 2 & -4 & 2 \\ -3 & 6 & -3 \end{bmatrix} = 1$

N.B.: For cases where $M_{\text{rank}} = [B \ AB \ \dots \ A^{n-1}B]$ like single-input system then the system is CC iff $|M| \neq 0$

Tuesday 3/12/2019

[2] Diagonalization test: CC can be judged visually (by inspection) after diagonalizing the system through a similarity transformation

$\Rightarrow x = Sz$, $|S| \neq 0$

$z = S^{-1}x$

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

$$S\dot{z} = ASz + Bu$$

$$S^{-1}S\dot{z} = S^{-1}ASz + S^{-1}Bu$$

$$\dot{z} = S^{-1}ASz + S^{-1}Bu$$

$$\Rightarrow \dot{z} = Jz + B_z u$$

$$\text{where } J = S^{-1}AS$$

$$B_z = S^{-1}B$$

S contains the eigen vectors of A as columns

$\Rightarrow J$ is a block diagonal (i.e. of the form)

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

\Rightarrow To judge CC using J & B_z :

1) Know to Jordan blocks should be associated the same eigen value

2) the last row in B_z should be non-zero

3) For distinct eigen value the row of z associated with it should be non-zero

$$\text{Exo: } J = \begin{bmatrix} -2 & 0 & 0 & 0 \\ 0 & -3 & 0 & 0 \\ 0 & 0 & 4 & 1 \\ 0 & 0 & 0 & 4 \end{bmatrix}, B_z = \begin{bmatrix} -1 \\ 0 \\ -7 \\ 0 \end{bmatrix}$$

solⁿ -2 is CC

-3 is CC

4 is CC

∴ the system is uncontrollable

Exo: 9.12 (p 650)

* Judging controllability through the (T.F):

⇒ Given $\dot{x} = Ax + Bu \quad x \in \mathbb{R}^n$
 $y = Cx + Du \quad u \in \mathbb{R}^m$

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

assuming the system is observable; then the system is completely controllable iff there is no poles-zeros cancellation in the T.F

Ex: $G(s) = \frac{s+3}{s^2-3s-10} = \frac{s+3}{(s+2)(s-5)}$; no cancellation ⇒ CC

$G(s) = \frac{s+4}{s^2+7s+12} = \frac{s+4}{(s+4)(s+3)} = \frac{1}{s+3}$; cancellation ⇒ $\bar{C}C$

Ex: $\dot{x} = \begin{bmatrix} -8 & 5 & 3 \\ -6 & 0 & 4 \\ 0 & 1 & -3 \end{bmatrix} x + \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} u$

⇒ $A = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix}$; $B = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix}$; $C = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix}$;
 $D = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix}$;

⇒ $G = C * \text{inv}(sI - A) * B + D$

⇒ pretty (collect (G))

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

$\frac{s+6}{s+3}$

∴ it must be third order to be CC

Sunday 8/12

* Observability: A system is observable if it's possible to obtain the states at (t_s, t_f) from observation of the output over a finite period of time

* Test For Observability: (OO)

□ The Rank test: A system is completely OO iff $\text{rank}(N) = \text{rank} \begin{pmatrix} CA \\ CA \\ \vdots \\ C^{n-1}A \end{pmatrix} = n$

$$\text{Ex} \dot{x} = \begin{bmatrix} -2 & 0 & 0 \\ 0 & 0 & -2 \\ 0 & 1 & -2 \end{bmatrix} x + \begin{bmatrix} 5 \\ 2 \\ 0 \end{bmatrix} u$$

$$y = [0 \quad 1 \quad -2] x$$

$$\text{rank}(N) = \begin{pmatrix} C \\ CA \\ CA^2 \end{pmatrix} = \text{rank} \begin{pmatrix} 0 & 1 & -2 \\ 0 & -2 & 2 \\ 0 & 2 & 0 \end{pmatrix} = 2 < n \Rightarrow \text{OO}$$

□ Diagonalization Test:

$$\text{let } x = Sz \Rightarrow z = S^{-1}x$$

and

$$\dot{z} = S^{-1}ASz + S^{-1}Bu$$

$$y = CSz + Du$$

where columns of S are the eigen vectors of A

⇒ The same criteria used for controllability applies except now columns of CT are used.

Ex^o Consider the previous example. $S = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1+j & 1-j \\ 0 & 1 & 1 \end{bmatrix} = [\text{eigen vector}]$

then

$$\Rightarrow \dot{z} = \begin{bmatrix} -20 & 0 & 0 \\ 0 & -1+j & 0 \\ 0 & 0 & -1-j \end{bmatrix} z + \begin{bmatrix} 5 \\ -j \\ j \end{bmatrix} u$$

$$y = [0 \quad -1+j \quad -1-j] z + D u \Rightarrow -20 \text{ is } \bar{00}$$

[3] Cancellation test :- examine the transfer function after cancelling poles with zeros

⇒ If the order of $G(s)$ is less than n then the system is $\bar{00}$

Ex^o Consider the previous example.

$$G(s) = C[sI_n - A]^{-1} B + D = \frac{2s}{s^2 + 2s + 2} \Rightarrow \text{order } 2 < n \text{ hence } \bar{00}$$

N.B. Stability, Controllability and Observability don't implate each other; a system can be stable, controllable and observable etc where the total combination is 8 up to unstable, uncontrollable and unobservable.

NO. _____

N.13.0:- Cancellation may be due to uncontrollable or unobservable or both, if you have cancellation check one of them in order to make right conclusion.

Exercise:- Consider the following system and determine cc & oo using the three methods.

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

$$y = [0 \quad 1 \quad -1] x$$

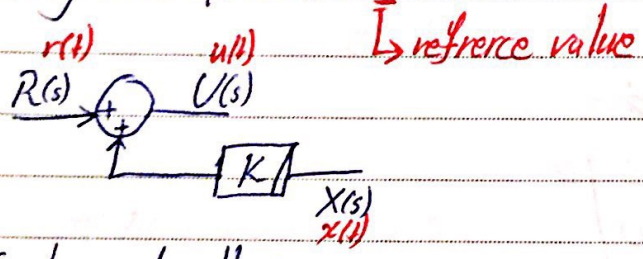
Thursday 10/12

#ch 10° Design of control System (state Feed Back)

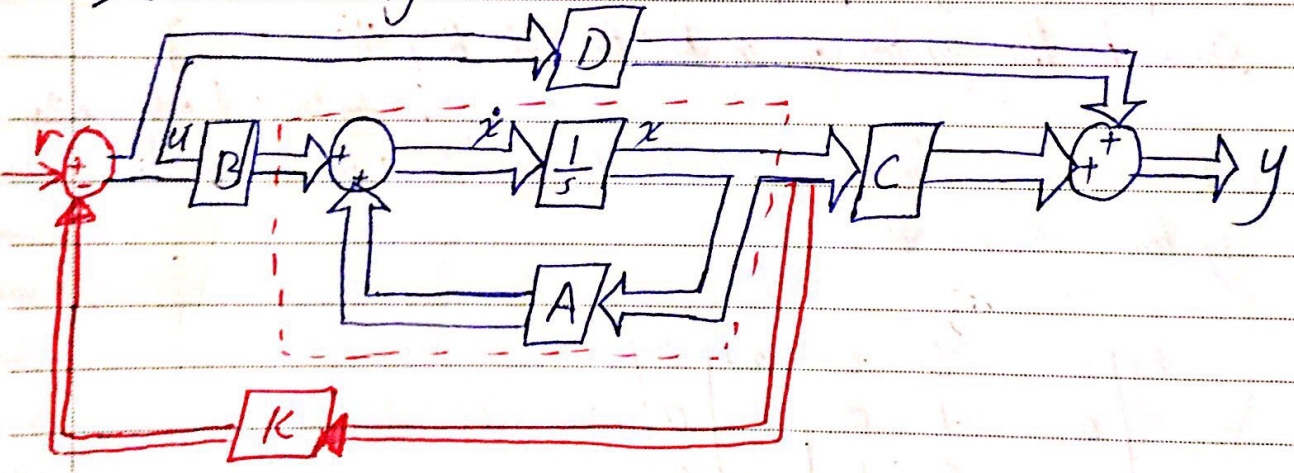
-use Feed Back the states not the output

⇒ Consider a system given by: $\dot{x} = Ax + Bu$; $x \in R^n$; $u \in R^n$

In state feed back method the states are feed-back through the input signal u (i.e $u = r - Kx$)



⇒ Schematically:-



before introducing of K the system is without Feed back

12/12 Thursday

⇒ stability of linear system is not dictated by the exciting input

$$\Rightarrow u = r - Kx$$

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

هو الذي يعتمد ال stability

State feed back seeks a K which results in acceptable system dynamics

motivational example :- ~~is~~

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

Simulate this system when i) $K=0$ ii) $K = [-4/3 \quad 3 \quad 5]$

Thursday 12/12/2019 الالصابي

Choosing K Frobenius (controllable form) :- in this form the system is described as:

$$\dot{x} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ -a_4 & -a_3 & -a_2 & -a_1 \end{bmatrix} x + \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix} u$$

① → ≠ 0

Pro. of this matrix

$$\textcircled{1} C.E = \lambda^4 + a_1 \lambda^3 + a_2 \lambda^2 + a_3 \lambda + a_4 = 0$$

$$\textcircled{2} \text{eigen vector} = \begin{bmatrix} 1 \\ \lambda_i \\ \lambda_i^2 \\ \lambda_i^3 \end{bmatrix}$$

⇒ Jor distinct eigen value

⇒ in which case

$$A_K = A - BK = A - \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix} [k_4 \ k_3 \ k_2 \ k_1]$$

$$A_K = A - \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ k_4 & k_3 & k_2 & k_1 \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ -a_4 - k_4 & -a_3 - k_3 & -a_2 - k_2 & -a_1 - k_1 \end{bmatrix}$$

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

$$\dot{x} = A_K x + Br$$

A_K is required to have desirable eigen value say $(\lambda_1, \lambda_2, \lambda_3$ and $\lambda_4)$ giving a closed loop C.E

$$\phi(\lambda) = (\lambda - \lambda_1)(\lambda - \lambda_2)(\lambda - \lambda_3)(\lambda - \lambda_4)$$

$$\phi(\lambda) = \lambda^4 + \delta_1 \lambda^3 + \delta_2 \lambda^2 + \delta_3 \lambda + \delta_4$$

$$A_K = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ -\delta_4 & -\delta_3 & -\delta_2 & -\delta_1 \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ -a_4 - k_4 & -a_3 - k_3 & -a_2 - k_2 & -a_1 - k_1 \end{bmatrix}$$

there for $\boxed{-a_i - k_i = -\delta_i}$ $i = 1, 2, 3, 4$; solve for $\underline{k_i = \delta_i - a_i}$

Ex: Consider the following

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

$\underline{-a_3} \quad \underline{-a_2} \quad \underline{-a_1}$

assign $-1, -2, -3$

$$\text{trace}(A) = 2 \quad \text{unstable}$$

$$\lambda^3 - 2\lambda^2 - 3\lambda - 4 = 0 \quad \text{unstable}$$

$$\lambda^3 - 2\lambda^2 + 3\lambda + 4 = 0 \quad \text{unstable}$$

$$\lambda^3 + 2\lambda^2 + 3\lambda + 4 = 0 \quad \text{idk}$$

$$d(\lambda) = (\lambda+1)(\lambda+2)(\lambda+3) = \lambda^3 + \underbrace{6}_{\tilde{\gamma}_1} \lambda^2 + \underbrace{11}_{\tilde{\gamma}_2} \lambda + \underbrace{6}_{\tilde{\gamma}_3}$$

$$K_1 = 6 + 2 = 8, \quad K_2 = 11 + 9 = 20, \quad K_3 = 6 + 4 = 10$$

$$\therefore K = [10 \quad 20 \quad 8]$$

$$\Rightarrow e^A = \text{cig}(A), \quad e_K = \text{cig}(A - BK)$$

Tuesday 17/12

Eigen values assignment for general A and B of controllable

⇒ Given $\dot{x} = Ax + Bu$ of no special form, $x \in \mathbb{R}^n$
 $u \in \mathbb{R}^1$ (single input)

then to calculate K :

i) Evaluate the C.E, put it in the form of
 $\lambda^n + a_1 \lambda^{n-1} + a_2 \lambda^{n-2} + \dots + a_n$; a_i given by
 $\Rightarrow \text{poly}(A)$

ii) let $M_{n \times n} = [B \quad AB \quad A^2B \quad \dots \quad A^{n-1}B]$; $\Rightarrow \text{ctrb}(A, B)$

iii) let $T = MW = M \begin{bmatrix} a_3 & a_2 & a_1 & 1 \\ a_2 & a_1 & 1 & 0 \\ a_1 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix}$ decom
conv

iv) Calculate \bar{K} as for Frobenius form

$$v) K = \bar{K}T^{-1}$$

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

Assign $-2, -3, \text{ and } -4$, for more exercies change $B = \begin{bmatrix} 1 \\ 2 \\ 5 \end{bmatrix}$

* Ackermann's method:

Given $\dot{x} = Ax + Bu$; $x \in \mathbb{R}^n$, $u \in \mathbb{R}^1$, system is CC

\Rightarrow to assign $\lambda_1, \lambda_2, \lambda_3, \dots, \lambda_n$ as eigen values. Generate $\phi(\lambda)$

$$\phi(\lambda) = (\lambda - \lambda_1)(\lambda - \lambda_2)(\lambda - \lambda_3) \dots (\lambda - \lambda_n) \quad ; \text{Closed loop C.E}$$

$$= \lambda^n + \alpha_1 \lambda^{n-1} + \alpha_2 \lambda^{n-2} + \dots + \alpha_n \quad ; \text{using conv in matlab}$$

$$\Rightarrow K = [0 \ 0 \ \dots \ 0 \ 1] [B \ AB \ \dots \ A^{n-1}B]^{-1} \phi(A)$$

$$= [0 \ 0 \ \dots \ 0 \ 1] [B \ AB \ \dots \ A^{n-1}B]^{-1} (A^n + \alpha_1 A^{n-1} + \alpha_2 A^{n-2} + \dots + \alpha_n I_n)$$

Example: Given $\dot{x} = \begin{bmatrix} -7 & 3 & 3 \\ -6 & 1 & 4 \\ 0 & 1 & -2 \end{bmatrix} x + \begin{bmatrix} 1 \\ 2 \\ 5 \end{bmatrix} u$

Thursday 20/12

assign $-2 \pm j3$ & -5

$$\text{sol: } K = [0 \ 0 \ 1] [BAB \ A^2B]^{-1} \phi(A)$$

$$\phi(\lambda) = (\lambda + 2 - j3)(\lambda + 2 + j3)(\lambda + 5) = \lambda^3 + 9\lambda^2 + 3\lambda + 65$$

\Rightarrow syms lam

$$\Rightarrow (\text{lam} + 2 - j3) * (\text{lam} + 2 + 3j) * (\text{lam} + 5)$$

ans

$$\Rightarrow \text{lam}^3 + 9 * \text{lam}^2 + 33 * \text{lam} + 65$$

$$\Rightarrow K = [0 \ 0 \ 1] * \text{inv}(\text{ctrb}(A, B)) * (A^3 + 9 * A^2 + 33 * A + 65 \text{eig}(\alpha_3))$$

$$K = \begin{bmatrix} 97 & -61 & 14 \\ 45 & 45 & 45 \end{bmatrix}$$

To check the correctness of $K \Rightarrow \text{eig}(A - B * K)$

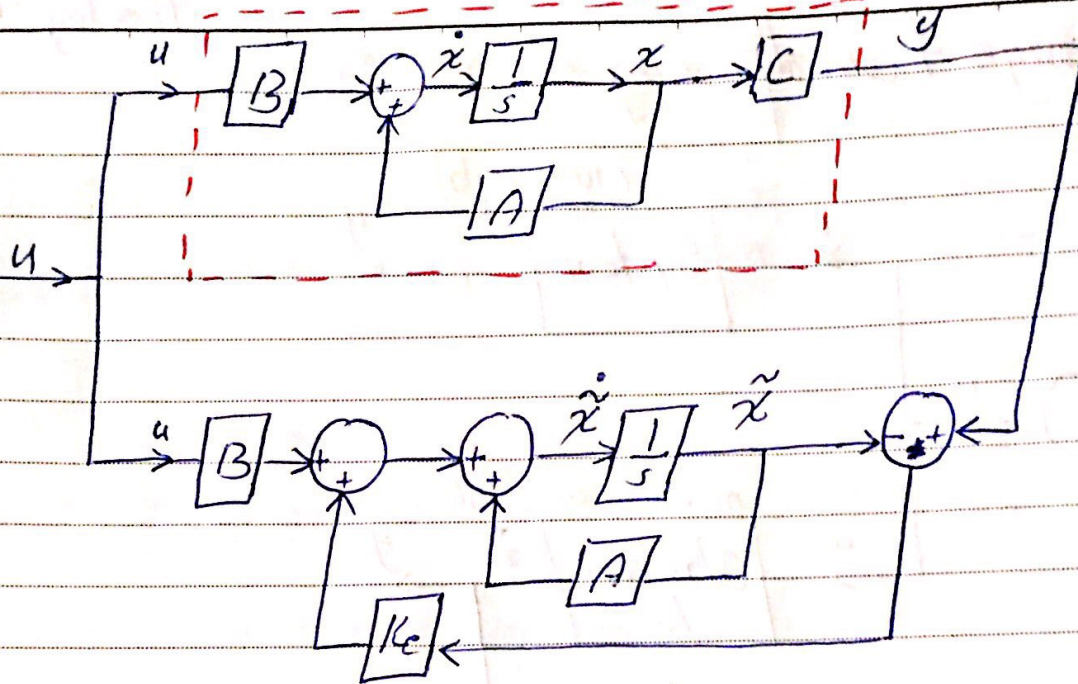
*** State Observers:** Sunday 22/12 Used to estimate the states from the system outputs in cases, where the state are available due to inaccessibility or differences in measurement.

Consider the system $\dot{x} = Ax + B u$
 $y = Cx$

and ^{an} observer the following form:

$$\begin{aligned} \dot{\tilde{x}} &= A \tilde{x} + B u + K_e (y - \tilde{y}) \\ \tilde{y} &= C \tilde{x} \end{aligned}$$

NO. Inaccessible



⇒ The design in observers center on a proper selection of K_e such that the estimated state \tilde{x} approach those inaccessible system state.

⇒ For the selection of K_e the ackermann's method can be adopted to have desirable eigen value

⇒ Ackermann's method:
$$K_e = \phi(A) \begin{bmatrix} C \\ CA \\ \vdots \\ CA^{n-1} \end{bmatrix}^{-1} \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$$

Ex: Consider the system

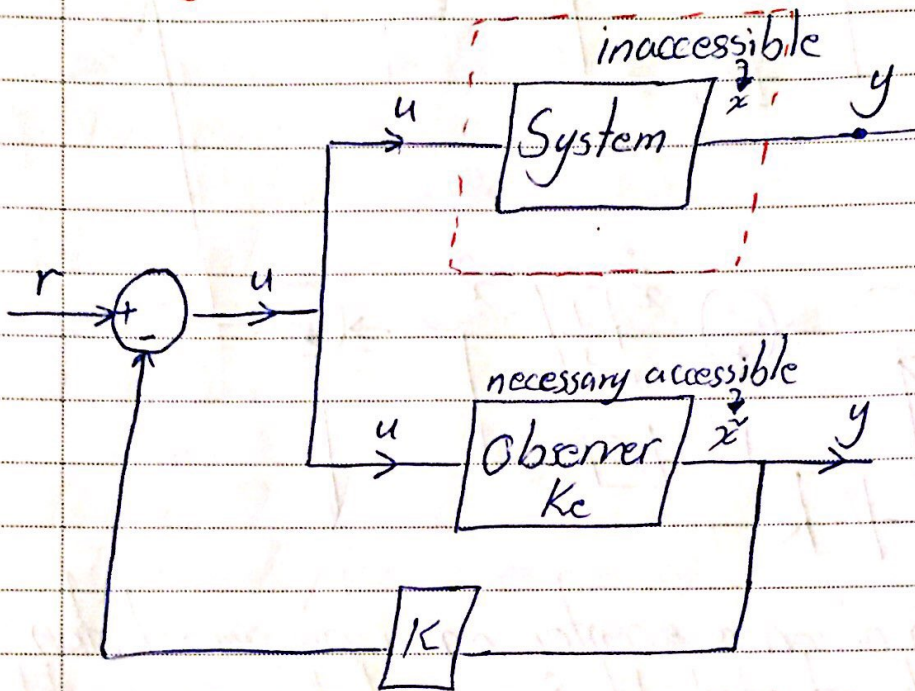
$$\dot{x} = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} x + \begin{bmatrix} 0 \\ 2 \end{bmatrix} u$$

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

⇒ $K_e = \text{acker}(A', C', [-4, -5])'$

Choose K_e for $(A - K_e C)$ to have eigen value $-4, -5$

*1) Design of controller using observers :



⇒ to calculate K use ackermann's method as usual ending with desirable eigen value for $A-BK$

⇒ to calculate K_e use ackermann's method as usual either after adaptation or using matlab ending with desirable eigen values for $A-K_eC$

* Quadratic optimal control:

⇒ Given $\dot{x} = Ax + Bu$, optimal control seeks a controller $u = -Kx$ to minimize a performance index.

$$J = \int_0^{\infty} (\underbrace{x^T Q x}_{\text{state}} + \underbrace{u^T R u}_{\text{control}}) dt$$

⇒ It can be shown that: $K = R^{-1} B^T P$

where P is positive definite matrix satisfying

$$A^T P + PA - PBR^{-1}B^T P + Q = 0$$

⇒ It can be shown that $J = x(0)^T \underbrace{P}_{(n \times n)} \underbrace{x(0)}_{(n \times 1)} = \text{scalar}_{(1 \times 1)}$

Example: See book

using matlab $[K, P, E] = \text{lqr}(A, B, Q, R)$

Exo: $\dot{x} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} x + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u$, $Q = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$, $R = 1$

⇒ $[K, P, E] = \text{lqr}(A, B, Q, R)$

$$K = [1 \quad 1.7321]$$

$$P = \begin{bmatrix} 1.7321 & 1 \\ 1 & 1.7321 \end{bmatrix} \Rightarrow \begin{bmatrix} \sqrt{3} & 1 \\ 1 & \sqrt{3} \end{bmatrix} = P^T$$

$$c = \begin{bmatrix} -0.866 + j0.5 \\ -0.866 - j0.5 \end{bmatrix}$$

⇒ For positive definiteness^{o-}

① Symtric

② $P_{11} > 0$

③ $\begin{vmatrix} P_{11} & P_{12} \\ P_{21} & P_{22} \end{vmatrix} \checkmark \text{ det.}$