

- differentiation $d\left(\frac{ds(t)}{dt}\right) = s_d(s(t)) - s(0)$
- final value $\lim_{t \rightarrow \infty} s(t) = \lim_{s \rightarrow 0} sF(s)$ these are useful in studying

LTI systems

$$d\left(\frac{d^n s}{dt^n}\right) = s_d\left(\frac{ds}{dt}\right) - \frac{ds}{dt}(0) = S\left[s_d(s) - s(0)\right] - \frac{ds}{dt}(0) =$$

generally taking identical route as in

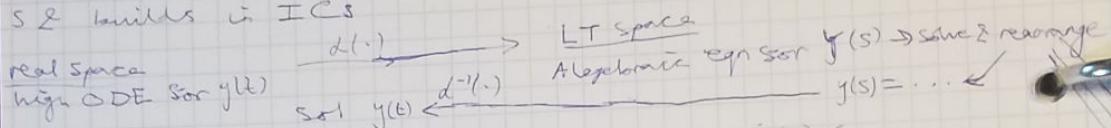
$$S^2 d(s) - s s(0) - \dot{s}(0)$$

$$\frac{d^n s}{dt^n} : d\left(\frac{d^n s}{dt^n}\right) = S^n d(s) - S^{n-1} s(0) - S^{n-2} \dot{s}(0) - \dots - \overset{\uparrow}{S^{n-1} \dot{s}}(0) - \dots - \ddot{s}(0)$$

with assumption $s(0) = \dot{s}(0) = \ddot{s}(0) = \dots = 0$

$d\left(\frac{d^n s}{dt^n}\right) = S^n d(s)$ L.T. converts derivatives into multiplication by

S^n builds in ICs



$$\frac{d^2 x(t)}{dt^2} + 2\zeta \omega_n \frac{dx}{dt} + \omega_n^2 x(t) = \dot{x}_n^2(t) \quad \text{ICs } x(0) = \dot{x}(0) = 0$$

$$\text{Taking Laplace transform } d\left[\frac{d^2 x}{dt^2} + 2\zeta \omega_n \frac{dx}{dt} + \omega_n^2 x\right] = d[\dot{x}_n^2(t)]$$

$$S^2 X(s) + 2\zeta \omega_n S X(s) + \omega_n^2 X(s) = \dot{x}_n^2 F(s) \quad , X(s) = d(x(t)) \& F(s) = d(\dot{x}(t))$$

$$(S^2 + 2\zeta \omega_n S + \omega_n^2) X(s) = \omega_n^2 F(s)$$

$$G(s) = \frac{X(s)}{F(s)} = \frac{\omega_n^2}{S^2 + 2\zeta \omega_n S + \omega_n^2}$$

$G(s)$ is ratio of polys in 's' is called Transfer

func of Z system

$$\text{numerical ex} / \frac{d^2 y}{dt^2} + 5 \frac{dy}{dt} + 6y = e^{4t} \quad y(0) = 10 \quad \frac{dy}{dt}(0) = 8$$

for considering generic case

$$\text{Taking L.T. } d\left\{\frac{d^2 y}{dt^2}\right\} + S d\left[\frac{dy}{dt}\right] + 6d[y] = d[e^{4t}] \quad \therefore$$

$$(S^2 Y(s) - S Y(0) - \frac{dy}{dt}(0)) + (5S Y(s) - 5Y(0)) + 6Y(s) = \frac{1}{S-4}$$

$$(S^2 + 5S + 6)Y(s) = 8Y(0) + 5Y(0) + \frac{dy}{dt}(0) + \frac{1}{S-4} = (S+5)10 + 8 + \frac{1}{S-4}$$

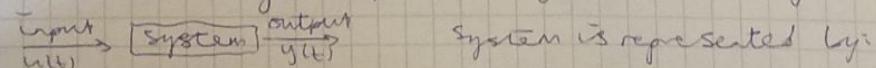
$$Y(s) = \frac{1}{(S+2)(S+3)(S-4)} + \frac{10S + 58}{(S+2)(S+3)} = \frac{10S^2 + 18S - 231}{(S+2)(S+3)(S-4)} =$$

$$\frac{227}{6(S+2)} - \frac{195}{7(S+3)} + \frac{1}{42(S-4)} \quad (\text{doing partial fractions}) \quad \therefore \text{taking } d^{-1}(\cdot)$$

$$\text{Set C. our ODE is } y(t) = \frac{227}{6} e^{-2t} - \frac{195}{7} e^{-3t} + \frac{1}{42} e^{4t}$$

From ICS part: C.F. From Z S z transform part: P.I.

Consider a single input - single output dynamical system



$$\begin{aligned}\dot{x}(t) &= Ax(t) + bu(t) \\ y(t) &= Cx(t)\end{aligned}$$

$x(t) \in \mathbb{R}^n$ (states)
 $u(t) \in \mathbb{R}$ (control input) $y(t) \in \mathbb{R}$ (output)

$A \in \mathbb{R}^{n \times n}$ $b \in \mathbb{R}^{n \times 1}$ $C \in \mathbb{R}^{1 \times n}$ assume IC $x(0)=0$ (unless)

Specified, may consider $\neq 0$)

Taking Laplace of $\textcircled{1}$ yields $sX(s) = Ax(s) + bu(s)$, $Y(s) = CX(s)$
 $\therefore sI_n - AX(s) = bu(s)$ (identity mat $\neq 0$ \Rightarrow n dimensions)

$$\therefore X(s) = (sI_n - A)^{-1}bu(s)$$

output $Y(s) = CX(s) = \underbrace{C}_{\text{transf. func. representation}} \underbrace{(sI_n - A)^{-1}bu(s)}_{\text{input}}$

$$\frac{Y(s)}{U(s)} = C(sI_n - A)^{-1}b$$

This is transfer func between $Y(s) \in U(s)$

\Rightarrow is a ratio of poly in 's'.

$$\text{revisit } \frac{d^2x}{dt^2} + 2\zeta\omega_n \frac{dx}{dt} + \omega_n^2 x = \omega_n^2 s(t) \quad \xrightarrow{\text{3(b) input}} \text{system} \xrightarrow{\text{output}} y(t) = x(t)$$

State space form $\dot{x} = Ax + Bu$ $y = cx$ Triplet (A, B, C)

$$A = \begin{bmatrix} 0 & 1 \\ -\omega_n^2 & -2\zeta\omega_n \end{bmatrix}, B = \begin{bmatrix} 0 \\ \omega_n^2 \end{bmatrix}, C = \begin{bmatrix} 1 & 0 \end{bmatrix} \quad x = \begin{bmatrix} x \\ \dot{x} \end{bmatrix}$$

\Rightarrow Converting to TF (transfer func) representation

$$\frac{Y(s)}{U(s)} = C(sI_n - A)^{-1}B \quad \text{use } A, B, C \text{ from above}$$

$$F(s) = (sI_n - A)^{-1} = \begin{bmatrix} s & 1 \\ \omega_n^2 & s + 2\zeta\omega_n \end{bmatrix}^{-1} = \frac{1}{s^2 + 2\zeta\omega_n s + \omega_n^2} \begin{bmatrix} s + 2\zeta\omega_n & 1 \\ -\omega_n^2 & s \end{bmatrix}$$

$$B = \begin{bmatrix} 0 \\ \omega_n^2 \end{bmatrix} \quad \& \quad C = \begin{bmatrix} 1 & 0 \end{bmatrix} \quad \therefore \frac{Y(s)}{F(s)} = \begin{bmatrix} 1 & 0 \end{bmatrix} \left(\frac{1}{s^2 + 2\zeta\omega_n s + \omega_n^2} \begin{bmatrix} s + 2\zeta\omega_n & 1 \\ -\omega_n^2 & s \end{bmatrix} \right) \begin{bmatrix} 0 \\ \omega_n^2 \end{bmatrix} =$$

$$G(s) = \frac{\omega_n^2}{s^2 + 2\zeta\omega_n s + \omega_n^2}$$

$$\frac{d^2x}{dt^2} + 2\zeta\omega_n \frac{dx}{dt} + \omega_n^2 x = \omega_n^2 s(t) \quad , \quad x = 0 \Rightarrow x(0) = 0$$

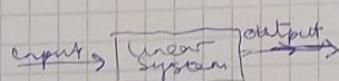
transfer func

$$G(s) = \frac{\omega_n^2}{s^2 + 2\zeta\omega_n s + \omega_n^2}$$

equivalent

state space form $\dot{x} = Ax + Bu$ $y = cx$ $X = \begin{bmatrix} x \\ \dot{x} \end{bmatrix}^T$

$$A = \begin{bmatrix} 0 & 1 \\ -\omega_n^2 & -2\zeta\omega_n \end{bmatrix}, B = \begin{bmatrix} 0 \\ \omega_n^2 \end{bmatrix}, C = \begin{bmatrix} 1 & 0 \end{bmatrix}$$



ordinary differential eqn (ODE) $\xrightarrow{\text{Laplace transform}}$ $\text{input} \xrightarrow{\text{linear system}} \text{output}$

compact form

state space model

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

represented also by (A, B, C) triplet

$$y = Cx + Du$$

transfer func form

$$G(s) = C(sI_n - A)^{-1}B$$

ratio of polys 's'

$$G(s) = \frac{N(s)}{D(s)}$$

open loop half plane

stability: eigenvalues of A

by charac eqn $\det(sI_n - A) = 0$ λ is eigenvals $\operatorname{Re}(\lambda) < 0$

For transser func form: charact eqn $D(s) = 0$ roots (poly)
 & poles of transser func $\text{Re}(\text{poles}) < 0$

Ex / state space $A = \begin{bmatrix} -7 & -12 \\ 1 & 0 \end{bmatrix}$ $B = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ $C = \begin{bmatrix} 1 & 2 \end{bmatrix} \Rightarrow G(s) = C(sI - A)^{-1}B$

transser func $\frac{x(s)}{u(s)} = \frac{s+2}{s^2 + 7s + 12}$

eigvals $\det(sI - A) = 0$ (charact eqn) $\lambda_{1,2} = \{-3, -4\}$
 eig vals stable node at equilib

$G(s) = \frac{s+2}{s^2 + 7s + 12}$ (charact eqn $\lambda^2 + 7\lambda + 12 = 0$)

$D(s) = s^2 + 7s + 12$. charact eqn
 $D(s) = 0$ & solving for s yields poles
 $s = -3, -4$

For a triple (A, B, C) $\dot{x} = Ax + Bu$ $x \in \mathbb{R}^n \rightarrow [LT]$

$y = Cx$ $y \in \mathbb{R}$, $u \in \mathbb{R}$
 associated transser func representation $G(s) = C(sI - A)^{-1}B$

this representation is unique

but it is possible to have many different triple representation

(A, B, C) give rise to same transser func

uniqueness of state space representation: suppose T is a square & invertible mat let T has dimens $\leq A$

(A, B, C) state space triple \rightarrow transser func representation \rightarrow

$T \rightarrow (TAT^{-1}, TB, CT^{-1})$ transformed state space triple using square invertible $T \rightarrow$ lead to same transser func representation $TAT^{-1} = I_n$

$$G(s) = CT^{-1}(sI_n - TAT^{-1})^{-1}TB = CT^{-1}[STT^{-1} - TAT^{-1}]^{-1}TB$$

$$= CT^{-1}[T(sI_n - A)T^{-1}]^{-1}TB$$

$$G(s) = C(sI - A)^{-1}B$$

$$= CT^{-1}[T(sI_n - A)T^{-1}]^{-1}TB =$$

$$(AB)^{-1} = B^{-1}A^{-1}$$

$$CT^{-1}[(sI_n - A)T^{-1}]^{-1}T^{-1}TB$$

not unique

$$= CT^{-1}T(sI_n - A)^{-1}T^{-1}TB = C(sI_n - A)^{-1}B \Leftarrow (A, B, C) \text{ triple}$$

Ex / non unique state space representation / output mat \downarrow
 output mat \downarrow $A = \begin{bmatrix} 0 & 1 \\ -1 & -2 \end{bmatrix}$
 $B = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$

$C = \begin{bmatrix} 1 & 0 \end{bmatrix}$ triple (A, B, C) $\tilde{A} = \begin{bmatrix} -1 & 0 \\ -1 & -1 \end{bmatrix}$ $\tilde{B} = \begin{bmatrix} -1 \\ -1 \end{bmatrix}$ $\tilde{C} = \begin{bmatrix} -1 & -1 \end{bmatrix}$ triple $(\tilde{A}, \tilde{B}, \tilde{C})$

Both (A, B, C) & $(\tilde{A}, \tilde{B}, \tilde{C})$ yield same transser func use $D = 0$

$$G(s) = \frac{1}{s^2 + 2s + 1} \leftarrow \text{this is unique}$$

\ controllability LTI system: $\dot{x}(t) = Ax(t) + Bu(t)$, $x(t) \in \mathbb{R}^n$

$u(t) \in \mathbb{R}^m$ output eqn is not important now

so state space LTI system is $x(t) = e^{At}x_0$

$$x(t) = e^{A(t-t_0)}x(t_0) + \int_{t_0}^t e^{A(t-\tau)}Bu(\tau)d\tau$$

depending on $x(t_0)$ and $u(\tau)$

Let us desire a set which represents all those pts that can be reached in time T from origin reachable set

$$R(T) = \left\{ x \in \mathbb{R}^n \mid x = \int_{t_0}^T e^{A(t-\tau)}Bu(\tau)d\tau \right\} \quad (\because e^{A(t-t_0)}x_0 = 0)$$

for some control $u(t) \in U$ during time interval $[t_0, T]$ system is

controllable if $R(T) = \mathbb{R}^n$ ie any pt in \mathbb{R}^n is reachable in time T if it is possible to find a control signal $u(t)$ which drive x

State from origin at time $t=t_0$ to anywhere in \mathbb{R}^n during $[t_0, T]$ is this possible? given (A, B, C) under what conditions?

system is said to be controllable?

For $\dot{x}(t) = Ax(t) + Bu(t)$ know set is $x(t) = e^{A(t-t_0)}x(t_0) + \int_{t_0}^t e^{A(t-\tau)}Bu(\tau)d\tau$

at time T , $x(T) = x^* = e^{A(T-t_0)}x(t_0) + \int_{t_0}^T e^{A(T-\tau)}Bu(\tau)d\tau = x^* =$

$$= e^{AT}x - e^{-At}x(t_0) + \int_{t_0}^T e^{A(T-\tau)}Bu(\tau)d\tau$$

$$e^{-AT}x^* = e^{-At}x(t_0) + e^{-AT} \int_{t_0}^T e^{A(T-\tau)}Bu(\tau)d\tau$$

cancel e^{-AT} crucial part

$$e^{-AT}x^* - e^{-At}x(t_0) = \int_{t_0}^T e^{-A\tau}Bu(\tau)d\tau$$

recall Cayley-Hamilton Thm: $e^{-AT} = \sum_{i=0}^{n-1} \alpha_i(T)A^i$ where

$\alpha_i(T)$, $i=0, 1, \dots, n-1$ scalar time series, depending on co-effs of A

charac poly of A { $\lambda_1, \lambda_2, \dots, \lambda_n$ }

$$e^{-AT}x^* - e^{-At}x(t_0) = \int_{t_0}^T e^{-A\tau}Bu(\tau)d\tau = \int_{t_0}^T \sum_{i=0}^{n-1} \alpha_i(\tau)A^iBu(\tau)d\tau =$$

$$\sum_{i=0}^{n-1} A^i B \int_{t_0}^T \alpha_i(\tau)u(\tau)d\tau = [A^0 B; A^1 B; \dots; A^{n-1} B] \begin{bmatrix} \int_{t_0}^T \alpha_0(\tau)u(\tau)d\tau \\ \int_{t_0}^T \alpha_1(\tau)u(\tau)d\tau \\ \vdots \\ \int_{t_0}^T \alpha_{n-1}(\tau)u(\tau)d\tau \end{bmatrix}$$

known: const vec

$$= e^{-AT}x^* - e^{-At}x(t_0)$$

const vec $n \times 1 = M^{n \times m}$

$$\begin{bmatrix} g_1(u(\tau)) \\ \vdots \\ g_{n-1}(u(\tau)) \end{bmatrix} \quad n \times m$$

set exists iff $\text{rank}(M) = n$

comps
es
es
req
ctrl
inputs
(t/f)
 $t \in (t_0, T)$

\Def: A matrix (A, B) that forms an n -dim LTI System, is said to be controllable iff controllability mat $M = [B : AB : A^2B : \dots : A^{n-1}B]$ (here n denotes state) is full rank i.e. $\text{rank}(M) = n$

also (A, B) is controllable iff $2 \{s_1, T\}$ eigenvals of $(A - BK)$ can be assigned arbitrarily by choice of K . For single I/P single O/P, test for controllability is that $\det(M) \neq 0$.

\Ex/ consider $\dot{x}_1 = x_2$ $\dot{x}_2 = u$ i.e. $A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$ $B = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$
controllability mat is $M = [B : AB] \quad \{\because 2 \text{ dim system}\}$
 $= \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}$ clearly $\text{rank}(M) = 2$ check for controllability $\det(M) = -1 \neq 0 \therefore$ controllable

\Ex/ consider $\tilde{A} = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$ $\tilde{B} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$ controllability mat $M = [B \ AB]$
 $\det(M) = \det \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} = 0 \therefore$ not controllable \therefore $(A - BK)$ can't be assigned arbitrarily

\Week 5/ Oliver control laws State feedback design \rightarrow pole placement design \circledast Lipschitz Condition

Oliver control laws state feedback - pole placement design/
consider LTI system $\textcircled{1} \quad \{\dot{x} = Ax + Bu \ x \in \mathbb{R}^n \Rightarrow r \in \mathbb{R}^m \text{ reference}\}$ $\textcircled{2} \quad \{y = cx \quad (A, B, C)\}$
possible which sys is expected to follow this tracking problem
 $r(t) = 0$ is a special case, called regulation problem

consider single I/P single O/P ($SISO$) case: Desire $\textcircled{2}$: $u = -Kx$ is state feedback control with $\textcircled{1}$ in $\textcircled{1}$ yields:

$\dot{x} = Ax - BKx \quad \dot{x} = (A - BK)x \quad \text{Des. this is closed loop syst}$

Suppose $r = 0$: looking at regulation problem given $\dot{x} = Ax + Bu$

$u = -Kx$ to be determined \Rightarrow objective: with 'u': a closed loop syst

All is stable \rightarrow syst (closed loop) possess specific time \rightarrow response behaviour, which is indicated by desired signals or pole locations

$$\dot{x} = \frac{A_{cl}}{(A-BK)}x + A_{cl} + \text{Hurwitz is stable } \operatorname{Re}(\lambda(A_{cl})) < 0$$

charac eqn (poly in 'S') $\det(SI - (A-BK)) = 0$
Identity mat & correct dims

how to achieve this?

(*) pole placement prob / $\dot{x} = Ax + Bu \quad x \in \mathbb{R}^n \quad y = Cx$

find $u = -Kx$ st closed loop poles (eigenvalues) are at $\mu_1, \mu_2, \dots, \mu_n$
desired locations defines 2 specific behaviour for syst

assumptions: • (A, B) controllable $\Rightarrow M = [B : AB : \dots : A^{n-1}B]$

rank(M) full rank

+ assume $\mu_i, i=1, \dots, n$ are distinct

entire state available for feedback control design

• Control I/P is unconstrained. (ie no amplitude or rate limits on signal u)

pole placement Design Steps / $\dot{x} = Ax + Bu$

step 1: check controllability of (A, B) $M = [B : AB : A^2B : \dots : A^{n-1}B]$

rank(M) = n in single I/P single O/P else $\det(M) \neq 0$ also ok

step 2: design $K = [k_1, k_2, \dots, k_n] \because x \in \mathbb{R}^n$ vec of vars k_i

step 3: closed loop syst $\dot{x} = (A-BK)x$ charac poly is $\det(SI - (A-BK))$

$= \prod_{i=1}^n (S - \mu_i)$ known desired poles given which is 2 desired charac poly $\because \mu_1, \mu_2, \dots, \mu_n$ are

distinct desired poles

step 4: solve for vars k_i 's by matching coeffs of polys

$u = -Kx$ state feedback control law

$$\text{Ex 1: } \dot{x} = Ax + Bu \quad y = Cx \quad A = \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix} \quad B = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \quad C = \begin{bmatrix} 1 & 0 \end{bmatrix}$$

find $u = -Kx$ st closed loop eigenvalues are $-4, 2, -5$:

$$\text{step 1: } [B : AB] : M = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \cdot \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 & 2 \\ 1 & 1 \end{bmatrix} \quad \text{rank}(M) = \text{full rank (2)}$$

also $\det(M) \neq 0$ (-2) [single I/P sing O/P] \therefore controllable

$$\text{step 2: } K = [k_1, k_2]$$

$$\text{step 3: desired charac poly } (S+4)(S+5) = S^2 + 9S + 20 \quad \textcircled{1}$$

$$\text{charac poly of closed loop syst } A_{cl}^+ = \begin{bmatrix} 1 & 2 \\ 2-k_1 & 1-k_2 \end{bmatrix} \text{ is}$$

$$S^2 + (k_2 - 2)S - 3 + 2k_1 - k_2 \quad \textcircled{2}$$

$$A_{cl} = (A - BK) = \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix} - \begin{bmatrix} 0 \\ 1 \end{bmatrix} \begin{bmatrix} k_1 & k_2 \end{bmatrix}$$

$$\text{Comparing coeffs of } \textcircled{1} \text{ & } \textcircled{2} \quad -2+k_2=9 \quad \therefore k_2=11 \quad -3+2k_1-k_2=20$$

$$\therefore k_1=17 \quad \therefore \text{Control law } u = [-17 \quad -11]x$$

$$\text{Towards a general case: } \dot{x} = Ax + Bu \quad n \in \mathbb{R}^n \quad u = -Kx$$

$u \in \mathbb{R}^m, m=1 \quad y = cx \quad$ suppose A, B possess a specific structure as follows: $A = \begin{bmatrix} 0 & 1 & 0 & 0 & \cdots & 0 \\ 0 & 0 & 1 & 0 & \cdots & 0 \\ 0 & 0 & 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & 0 & \cdots & 1 \\ a_n & a_{n-1} & a_{n-2} & \cdots & a_1 \end{bmatrix} \quad B = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 1 \end{bmatrix}$ Controllable Canonical form

$$\det(SI-A) = \rightarrow S^n + a_1S^{n-1} + a_2S^{n-2} + \cdots + a_n$$

or So what happens? write: $\dot{x} = (A+BK)x = A_{cl}x \quad K = [k_1 \ k_2 \ \cdots \ k_n]$

$$\text{where } A_{cl} = \begin{bmatrix} 0_{n-1} & I_{n-1} \\ - & - & - & - & - & - \\ (-a_n-k_1) & (-a_{n-1}-k_2) & (-a_{n-2}-k_3) & \cdots & (-a_1-k_n) \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0_{n-1} & 0_{n-1} & 0_{n-1} & \cdots & 0_{n-1} \end{bmatrix} - \begin{bmatrix} 0_{n-1} & 0_{n-1} \\ \vdots & \vdots \\ k_1 & k_2 & \cdots & k_n \end{bmatrix} \quad \begin{array}{l} (\text{say I/P sing o/p}) \\ \text{only one row} \end{array}$$

$$A_{cl} = \begin{bmatrix} 0_{n-1} & I_{n-1} \\ - & - & - & - & - & - \\ (-a_n-k_1) & (-a_{n-1}-k_2) & (-a_{n-2}-k_3) & \cdots & (-a_1-k_n) \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0_{n-1} & 0_{n-1} & 0_{n-1} & \cdots & 0_{n-1} \end{bmatrix}$$

known variables to be determined

\therefore know k_1, \dots, k_n are distinct desired poles

$$\sum_{i=1}^n (s-\mu_i) = s^n + d_1 s^{n-1} + d_2 s^{n-2} + \cdots + d_n \quad \text{like earlier}$$

\therefore know k_1, \dots, k_n are distinct desired poles

$$\sum_{i=1}^n (s-\mu_i) = s^n + d_1 s^{n-1} + d_2 s^{n-2} + \cdots + d_n \quad \text{like earlier}$$

associated syst mat will be $\begin{bmatrix} 0_{n-1} & I_{n-1} & - & - & - & - \\ -d_n & -d_{n-1} & -d_{n-2} & \cdots & -d_1 \end{bmatrix} \equiv A_{cl}$ Comparing two

mat yields

$$a_n + k_1 = d_n \quad \Rightarrow \quad k_1 = d_n - a_n$$

$$a_{n-1} + k_2 = d_{n-1} \quad \Rightarrow \quad k_2 = d_{n-1} - a_{n-1}$$

$$a_1 + k_n = d_1 \quad \text{known} \quad k_n = d_1 - a_1$$

associate with $\det(SI-A)$ straight forward calc

how to obtain this form is controllable canonical form \therefore

desire square invertible mat T let $x = Tx \quad \Rightarrow \quad x \in \mathbb{R}^n \quad \dot{x}$

represents states in transformed coords take their w.r.t time $\dot{x} = T\dot{x} \quad \therefore T$ is invertible $\dot{x} = T^{-1}\dot{x} = T^{-1}(Ax + Bu) \quad \leftarrow \dot{x} = Ax + Bu$

$$\therefore T^{-1}Ax + T^{-1}Bu = T^{-1}AT\dot{x} + T^{-1}Bu \quad (\because x = Tx)$$

$\dot{x} = (T^{-1}AT)\dot{x} + (T^{-1}B)u$ is in transformed coords

how determine T ?

$[T]$ is square, invertible mat $\dot{x} = Ax + Bu$ desire $T = MW$:

M is controllability mat $M = [B : AB : A^2B : A^{n-1}B]$

V is neg definite for $\beta \leq 0$ when $\beta > 0$

$$V \leq -\gamma(\alpha - \beta) \|x\|^2, \forall \|x\|_2 \leq r$$

V is neg definite when $r^2 < \frac{\alpha}{\beta}$

$\therefore V(x) = \|x\|^2$ can write domain as \mathbb{R}^n

$$\Omega := \{x \in \mathbb{R}^n \mid \|x\|^2 \leq \frac{\alpha}{\beta}\}.$$

Backstepping control / consider a nonlinear sys

consider a nonlinear syst $\dot{y} = \delta(y) + g(y)$ ①

$\xi = u$ ② for $y \in \mathbb{R}^n$, $\xi \in \mathbb{R}$ \therefore State space $[y \ \xi]^T \in \mathbb{R}^{n+1}$

$u \in \mathbb{R}$ is control input $u \leftarrow$ single control input

$$\dot{y} = \frac{d\xi}{dt}, \quad \xi = \int \frac{d\xi}{dt}$$

① is subsyst ①, ② is Subsyst ②

$\xi(\cdot) : D \rightarrow \mathbb{R}^n$, $g(\cdot) : \mathbb{R}^n \rightarrow \mathbb{R}^n$ are smooth func, & D contains in domain $D \subseteq \mathbb{R}^n$

origin

$$y=0, \xi=0 \text{ design objective}$$

design state feedback control $\xi(u)$ to stabilize \dot{y}

nonlinear syst ①-② about origin

assumptions: $\delta(\cdot)$ is known $g(\cdot)$ is known all states

y, ξ are measured

$$\text{Step 1: in ① if } \dot{y} = \delta(y) + g(y)\xi \quad \xi \in \mathbb{R}$$

treat ξ as a "Virtual Input"

Let ① be stabilized by a smooth state feedback control $\xi = \delta(y)$

$\delta(0)=0$ implies $\dot{y} = \delta(y) + g(y)\delta(y)$ is asymptotically stable $\delta : \mathbb{R}^n \rightarrow \mathbb{R}$

Suppose have a smooth pos def Lyapunov Func $V(y)$

that satisfies $\frac{dV}{dy} [\delta(y) + g(y)\delta(y)] \leq -W(y), \forall y \in D$, where $W(y)$ is a pos def func choice to be determined

here $W(\cdot)$ introduces certain level of performance \therefore Virtual input

at this step get or determine $\delta(y)$ that stabilize Subsyst ①

a symp stable

step 2 Let us rewrite $\dot{z} = \delta(z) + g(z)\alpha$, $\alpha = u$ as
 $\dot{y} = \delta(y) + g(y)\alpha$ in original Syst

$$\therefore \dot{y} = \delta(z) + g(z)\alpha + g(z)\alpha - g(z)\alpha = g(z)\alpha$$

$\dot{y} = u$ desire $\dot{z} = \dot{y} - \alpha$ $\dot{y} - g(z)\alpha$ is

$$g(z)[\dot{y} - \alpha] = g(z)\dot{z} \quad \therefore z \text{ dynamics is rewritten as}$$

$$\dot{z} = [\delta(z) + g(z)\alpha] + g(z)\dot{z} \quad \therefore \dot{y} = \dot{z} - \alpha = u - \alpha$$

where $\alpha = \frac{\partial \delta}{\partial y} [\delta(z) + g(z)\alpha]$ this is called back stepping

Suppose $v = u - \alpha \therefore \alpha$ remains same $\dot{z} = v$

In $\dot{y} = \delta(z) + g(z)\alpha + g(z)v$, $\dot{z} = v$ has a stable origin

Step 3: desire a composite Lyapunov func which is positive definite

$$V_c(y^+, z) = V(y) + \frac{1}{2}z^2$$

taking deriv w.r.t $V_c(\cdot)$ along $\dot{y} - \dot{z}$ yields

$$\dot{V}_c(\cdot) = \frac{\partial V}{\partial y} [\delta(y) + g(y)\alpha] + \frac{\partial V}{\partial z} g(y)z + z^2 \quad z \text{ dynamics}$$

now $\leq -W(y)$ negative definite

so we can pick st $\dot{V}_c(\cdot)$ is neg def

$$\dot{V}_c(\cdot) \leq -W(y) + \frac{\partial V}{\partial z} (g(y))z + z^2 \quad \text{desire}$$

$$\text{choose } V = -\frac{\partial V}{\partial z} g(y)z - k z^2 \quad \text{known as}$$

$$\text{with this } V \quad \dot{V}_c(\cdot) \leq W(y) - k z^2 \quad \text{neg def}$$

desire impie z origin ($y^+ = 0$, $z = 0$) which is neg def

$$\text{Subs } V, z, \dot{z}(\cdot) \quad u = \frac{\partial \delta}{\partial y} [\delta(y) + g(y)\alpha] - \frac{\partial V}{\partial z} g(y) - k \left[\dot{y} - \alpha \right] \text{ is}$$

z back stepping control law to be used in $\textcircled{1} - \textcircled{2}$

$w(z) + k z^2$ is pos def is a origin $y = 0$, $z = 0$

Ex Q 1: consider $x_1 = x_1^2 - x_1^3 + x_2$ & virtual input $\dot{x}_2 = u$

$$x_1 \text{ since } y = x_2 \text{ is like } y \text{ in our theory}$$

$$\delta(y) = y^2 - y^3 \quad ; \quad g(y) = 1$$

Consider $\dot{x}_1 = x_1^2 + x_1^3 + x_2$ take x_2 as 'virtual control'

need to design $x_2 = \phi(x_1)$ Let $x_2 = \frac{-x_1^2 - x_1}{\delta(x_1)}$ is choice given or $V(\cdot) \leq W(\cdot)$ given

with this virtual control $\delta(x)$ system is \mathcal{O} stable

$$\dot{x}_1 = x_1^2 - x_1^3 - x_1^2 - x_1 = -x_1^3 - x_1$$

$$\dot{V}(x_1) = x_1 \dot{x}_1 = -x_1^2 - x_1^4 \text{ for } V(x) \leq W(x_1)$$

$\dot{V}(x) = -x_1^2 - x_1^4 \leq -2V$ so origin is stable exponentially globally

to do backstepping desire $\bar{z}_2 = x_2 - \delta(x_1)$ (is like $\delta = \phi(x)$ term in eqn)

$$\bar{z}_2 = x_2 + x_1^2 - x_1$$

$$\dot{\bar{z}}_2 = \dot{x}_2 + \dot{x}_1 + 2x_1 \dot{x}_1$$

$$\dot{x}_1 = -x_1 - x_1^3 + \bar{z}_2 \quad \dot{\bar{z}}_2 = \dot{x}_2 + \dot{x}_1 + 2x_1 \dot{x}_1 = u + (1+2x_1)(-x_1 - x_1^3 + \bar{z}_2)$$

desiring composite Lyapunov func $V_c(x_1, \bar{z}_2) = V(x_1) + \frac{1}{2}\bar{z}_2^2$

(will begin) $\dot{V}_c = \dot{x}_1 \dot{x}_1 + \bar{z}_2 \dot{\bar{z}}_2 =$

$$x_1(-x_1 - x_1^3 + \bar{z}_2) + \bar{z}_2(u + (1+2x_1)(-x_1 - x_1^3 + \bar{z}_2)) =$$

$$x_1^2 - x_1^6 + \bar{z}_2^2 + x_1 + (1+2x_1)(-x_1 - x_1^3 + \bar{z}_2) + u \quad \text{where } u \text{ to be chosen}$$

$$u \leq W(x_1) \quad \text{but } -\bar{z}_2^2 \quad \text{more}$$

$$\dot{V}_c \text{ as neg def} \quad u = -x_1 - (1+2x_1)(-x_1 - x_1^3 + \bar{z}_2) - \bar{z}_2$$

gave us $\dot{V}_c = -x_1^2 - x_1^4 - \bar{z}_2^2$ which is neg def \mathbb{R}^3 is Globally

asympt stable

\Ex/\Q1/ consider 2nd order dynamical syst $\ddot{x} = x_1 \ddot{x}_2 \quad \delta(x) = -x_1^2$

$$\dot{x}_2 = x_1 + u \quad \mathcal{O}$$

a/ show $x_2 = \delta(x_1) = -x_1^2$ is a virtual control making sub syst \mathcal{O} stable asympt

b/ define $\bar{z}_2 = x_2 - \delta(x_1)$ must use a composite Lyapunov func $V_c(x_1, \bar{z}_2) = \frac{1}{2}(x_1^2 + \bar{z}_2^2)$ & find $u = \psi(x_1, x_2)$ st

$$\dot{V}_c(x_1, \bar{z}_2) = \ddot{x} = -x_1^4 - \bar{z}_2^2$$

\a/ with $x_2 = -x_1^2$ given

$\dot{x}_1 = x_1 \dot{x}_2 \Rightarrow \dot{x}_1 = -x_1^3$ choose candidate Lyapunov func $\frac{1}{2}x_1^2$

which is p.d.s. [composite Lyapunov func $V_c(x_1, \bar{z}_2)$ given as

$\frac{1}{2}(x_1^2 + \bar{z}_2^2)$] we can show $\dot{V}(x)$ is n.d.s. globally:

$\dot{V}(x) = x_1 \dot{x}_1 = -x_1^4 \& \therefore \dot{x}_1 = x_1 \dot{x}_2$ with 'virtual control' $x_2 = -x_1^2$ is

globally asympt stab

b/ desire $\bar{z}_3 = x_2 - \delta(x_1) \quad \dot{\bar{z}}_3 = \dot{x}_2 - \delta'(x_1) \quad \left\{ \begin{array}{l} \dot{x}_1 = x_1 \dot{x}_2 \\ \dot{x}_2 = x_1 + u \end{array} \right.$

$$= \dot{x}_2 - 2x_1 \dot{x}_1 = x_1 + 2x_1^3 + x_2 + u$$

$$\begin{aligned}
 \dot{x}_1 &= x_1 \overset{\curvearrowleft}{x}_2 + \theta(x_1) \\
 \dot{x}_2 &= x_1 + 2x_1^2 + x_2 + u \\
 V_c(-) &= \frac{1}{2}(x_1^2 + x_2^2) \quad \text{posi d.s.} \\
 V_c &\rightarrow \infty \text{ as } \|x\| \rightarrow \infty \quad x = (x_1, x_2)^T \\
 \text{as } V_c &= x_1 \dot{x}_2 + x_2 \dot{x}_1 = -x_1^4 - x_2^2 \quad \text{derived (or given)} \\
 &= x_1(x_1 x_2) + x_2(x_1 + 2x_1^2 x_2 + u) \\
 &= x_1^2 x_2 + x_2 x_1 + x_2 x_1^2 x_2 + x_2 u \\
 &= x_1(x_1 x_2) + x_2(x_1 + 2x_1^2 x_2 + u) = -x_1^4 - x_2^2 \quad \text{add \& subtract} \\
 &= -x_1^4 + x_1^2 x_2 + x_1^4 + x_2(x_1 + 2x_1^2 x_2 + u) \quad \text{function to be designed} \\
 &= -x_1^4 + x_1^2(x_2 + x_1^2) + x_2(x_1 + 2x_1^2 x_2 + u) = \\
 &= -x_1^4 + [(x_1^2 + x_1 + 2x_1^2 x_2 + u) x_2] \quad \text{to have} \\
 &= -x_1^4 - x_2^2 \quad \therefore \\
 x_1^2 + x_1 + 2x_1^2 x_2 + u &= -x_2 \\
 u &= -x_2 - (x_1^2 + x_1 + 2x_1^2 x_2) \\
 &= -(x_1^2 + x_1 + 2x_1^2 x_2) \quad u = \psi(x_1, x_2) = -(x_1 + x_2 + 2x_1^2 x_2)
 \end{aligned}$$

May 2020 pp/

\ Week 6 / \ Q4 from week 6 /

$$\ddot{\theta}(t) + \dot{\theta}(t) + \frac{2}{l} \sin \theta = 0 \rightarrow \text{dynamics}$$

• Candidate Lyapunov Func

$$V(\theta, \dot{\theta}) = \frac{1}{2} \dot{\theta}^2 + \frac{2}{l} (1 - \cos \theta)$$

$V(\theta, \dot{\theta}) = 0$ only at $\theta = \dot{\theta} = 0$ in a domain $D = \{(\theta, \dot{\theta}) \mid |\theta| < \pi\}$

$V(\theta, \dot{\theta})$ is positive in $D \setminus \{(0, 0)\}$

$\therefore \dot{\theta}^2$ is +ve and $(1 - \cos \theta)$ term +ve ($\cos \theta$ upper bounded by 1)

i. Sum of two +ve terms.

$\therefore V(\theta, \dot{\theta})$ is p.d.s in D

Taking time derivative $\dot{V}(\theta, \dot{\theta}) = \dot{\theta} \ddot{\theta} + \frac{2}{l} \dot{\theta} \sin \theta$

Substituting dynamics (given)

$$= \dot{\theta} (-\dot{\theta} - \frac{2}{l} \sin \theta) + \frac{2}{l} \dot{\theta} \sin \theta = -\dot{\theta}^2 \leq 0$$

$\dot{V}(\theta, \dot{\theta})$ can't be claimed as negative definite (though $-\dot{\theta}^2$ is negative term) for the system, since

$V(\theta, \dot{\theta}) = 0$ for $\dot{\theta} = 0$ and no matter what θ values are \therefore

$\therefore \dot{V}(\theta, \dot{\theta})$ is negative semi definite & conclusion is that \mathbb{Z} origin is Lyapunov Stable

using $V(\theta, \dot{\theta}) = \frac{1}{2} \dot{\theta}^2 + \frac{2}{l} (1 - \cos \theta)$

remark: Lyapunov conditions are only sufficient conditions

$$(ii) V(\theta, \dot{\theta}) = \frac{1}{2} \dot{\theta}^2 + \frac{2}{l} (1 - \cos \theta) + \frac{1}{2} (\theta + \dot{\theta})^2$$

was considered in part (i)

$V(\theta, \dot{\theta})$ is p.d.s in D (argue like in part (i))

Taking derivative $\dot{V}(\theta, \dot{\theta}) = \dot{\theta} \ddot{\theta} + (\theta + \dot{\theta})(\dot{\theta} + \ddot{\theta}) + \frac{2}{l} \dot{\theta} \sin \theta$

Like earlier substitute for $\ddot{\theta}$ from dynamics

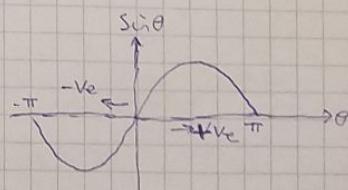
and bit of algebra yields $= -\dot{\theta}^2 - \frac{3}{l} \theta \sin \theta$
like in part (i)

$\dot{\theta}^2$ is +ve $\forall \theta \in \mathbb{R}$

$$\begin{array}{ccc} \theta & +ve & -ve \\ \sin \theta & +ve & -ve \end{array}$$

$$\begin{array}{ccc} \theta \sin \theta & +ve & +ve \end{array}$$

$$\dot{V}(\theta, \dot{\theta}) < 0 \quad \forall (\theta, \dot{\theta}) \in D \setminus \{(0, 0)\}$$



Since $V(\theta, \dot{\theta})$ is p.d.s and $\tilde{V}(\theta, \dot{\theta})$ is n.d.s. in D

the origin of the System in D is Asymptotically Stable in
 $D = \{(\theta, \dot{\theta})^T \mid |\theta| < \infty\}$

Q6 From week 6 / $\begin{cases} \dot{x}_1 = -4x_1 + x_1 x_2^2 \\ \dot{x}_2 = -x_2 + x_1 \end{cases}$

Candidate 'L' since $V(x_1, x_2) = x_1^2 + x_2^2$ $V = x_1^2 + x_2^2$ sum of squares
 $x_1 = x_2 = 0 \quad V = 0$

$V(x_1, x_2) > 0 \quad \forall (x_1, x_2) \neq (0, 0)$ $V(0, 0) = 0 \quad \therefore V(\cdot) \text{ is p.d.s}$

Letting $\{V(\cdot) > 0 \quad \forall x \in \mathbb{R}^2 \setminus \{0\}\} \quad V(0) = 0, x=0 \quad V(\cdot) \text{ is p.d.s}\}$

taking derivative of $V(\cdot)$ along $\Sigma \quad V(\cdot) = 2x_1 \dot{x}_1 + 2x_2 \dot{x}_2$ sub \dot{x}_1, \dot{x}_2

$$\text{from } \Sigma: \quad V(\cdot) = 2x_1(-4x_1 + x_1 x_2^2) + 2x_2(-x_2 + x_1) =$$

$$-8x_1^2 + 2x_1^2 x_2^2 + 2x_2 x_1 - 2x_2^2 =$$

$$-7x_1^2 - x_1^2 + 2x_1^2 x_2^2 + 2x_1 x_2 - x_2^2 - x_2^2 =$$

$$-7x_1^2 - x_1^2 + 2x_1^2 x_2^2 - (x_1^2 + x_2^2 - 2x_1 x_2) =$$

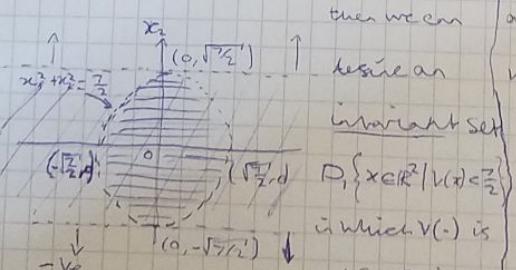
$$\begin{aligned} & -7x_1^2 - x_1^2 + 2x_1^2 x_2^2 - \underbrace{(x_1^2 + x_2^2 - 2x_1 x_2)}_{(x_1 - x_2)^2} \\ & \leq -7x_1^2 - x_1^2 + 2x_1^2 x_2^2 \end{aligned}$$

$$\tilde{V}(\cdot) \leq -7x_1^2 - x_1^2 + 2x_1^2 x_2^2$$

$$\tilde{V}(\cdot) = -x_1^2 (7 - 2x_2^2) - x_2^2 \quad \leftarrow \text{OR} \quad \tilde{V}(\cdot) = -7x_1^2 - x_1^2 (1 - 2x_2^2)$$

$\tilde{V}(\cdot) < 0$ if $(7 - 2x_2^2) < 0$ is guaranteed to be positive which implies

$$\tilde{V}(\cdot) < 0 \text{ if } x_2^2 + x_2^2 < \frac{7}{2}$$



is n.d.s. Condition holds

then we can

define an

invariant set

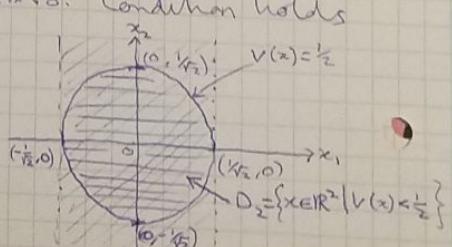
in which $V(\cdot)$ is

p.d.s $\Rightarrow \tilde{V}(\cdot)$

an invariant $D_2 = \{x \in \mathbb{R}^2 \mid V(x) \leq \frac{7}{2}\}$ in

which $V(\cdot)$ is p.d.s $\& \tilde{V}(\cdot)$ is

n.d.s. Condition holds



Week 1 Sheet

$$\checkmark \quad y^{(n)} = \frac{d^n y}{dt^n}$$

$$s = \overbrace{\dots}^n \quad y^{(1)} = g(y, u)$$

$$y^{(2)} = g(y, \dot{y}, u)$$

$$y^{(3)} = g(y, \dot{y}, \ddot{y}, u) \quad y^{(4)} = g(y, \dot{y}, \ddot{y}, \ddot{\dot{y}}, u)$$

$$\left[\begin{array}{l} y^{(1)} \\ y^{(2)} \\ y^{(3)} \\ y^{(4)} \\ \vdots \\ y^{(n)} \end{array} \right] = \left[\begin{array}{l} g(y, u) \\ g(y, \dot{y}, u) \\ g(y, \dot{y}, \ddot{y}, u) \\ g(y, \dot{y}, \ddot{y}, \ddot{\dot{y}}, u) \\ \vdots \\ g(y, \dot{y}, \dots, \dot{y}^{(n-1)}, u) \end{array} \right]$$

1.8}

$$\checkmark \quad \text{Let } x_1 = y \quad x_2 = \frac{dy}{dt} \quad \dots \quad x_n = \frac{d^{n-1} y}{dt^{n-1}}$$

taking derivatives of \mathbf{x} new variables

$$\dot{x}_1 = \dot{y} \quad \dots \quad \dot{x}_{n-1} = \frac{d^{n-1} y}{dt^{n-1}} \quad \dot{x}_n = \frac{d}{dt} \left(\frac{d^{n-1} y}{dt^{n-1}} \right) = y^{(n)}$$

$y^{(n)} = g(y, \dot{y}, \dots, \dot{y}^{(n-1)}, u)$ given. Sub variables:

$$\left. \begin{array}{l} \dot{x}_1 = x_2 \\ \dot{x}_2 = x_3 \\ \vdots \\ \dot{x}_{n-1} = x_n \\ \dot{x}_n = g(x_1, x_2, \dots, x_n, u) \end{array} \right\} \text{is represented as } \dot{\mathbf{x}} = F(\mathbf{x}, u)$$

$$\mathbf{x} = [x_1, x_2, \dots, x_n]^T$$

output is said to be y , i.e. x_1

So $y = x_1$ \checkmark $y = C\mathbf{x}$ where $C = [1 \ 0 \ \dots \ 0]_{1 \times n}$

$$\checkmark \quad \text{Let } x_1 = y \quad x_2 = \frac{dy}{dt}, \quad x_3 = \frac{d}{dt} \left(\frac{dy}{dt} \right) = \frac{d^2 y}{dt^2} \quad \dots \quad x_n = \frac{d^{n-1} y}{dt^{n-1}} \quad \therefore$$

$$\text{in } \dot{x}_1 = \dot{y} = \frac{dy}{dt} = \frac{dy}{dt}, \quad \dot{y} = x_2 \quad \therefore \quad \ddot{y} = \dot{x}_2 \quad \therefore \quad y^{(3)} = \ddot{x}_3 \quad \dots \quad y^{(n)} = \ddot{x}_n \quad \therefore$$

$$\checkmark \quad \dot{x}_{n-1} = \frac{d^{n-1} y}{dt^{n-1}} = y^{(n-1)}, \quad \therefore \quad \dot{x}_n = \frac{d}{dt} (y^{(n-1)}) = y^{(n)} = g(y, \dot{y}, \dots, \dot{y}^{(n-1)}, u) \quad \therefore$$

$$\dot{\mathbf{x}} = \left[\begin{array}{l} \dot{x}_1 \\ \dot{x}_2 \\ \vdots \\ \dot{x}_{n-1} \\ \dot{x}_n \end{array} \right] = \left[\begin{array}{l} x_2 \\ x_3 \\ \vdots \\ x_n \\ g(y, \dot{y}, \dots, \dot{y}^{(n-1)}, u) \end{array} \right] = \left[\begin{array}{l} x_2 \\ x_3 \\ \vdots \\ x_n \\ g(x_1, x_2, \dots, x_{n-1}, u) \end{array} \right]$$

$$\therefore \quad \mathbf{x} = \left[\begin{array}{l} x_1 \\ x_2 \\ \vdots \\ x_n \end{array} \right] \quad \therefore \quad y = C\mathbf{x} \quad C = [1 \ 0 \ \dots \ 0]_{1 \times n}$$

$\left. \begin{array}{l} \frac{1}{2} \\ \frac{1}{2} \end{array} \right\}$

$$2 / \ddot{\theta}_1 = -MgL \sin \theta_1 - k(\theta_1 - \theta_2)$$

$$\ddot{\theta}_1 = -\frac{1}{I} MgL \sin \theta_1 - \frac{1}{I} k(\theta_1 - \theta_2) \quad J \ddot{\theta}_2 = Jk(\theta_1 - \theta_2)$$

$$\ddot{\theta}_2 = J^{-1} u + J^{-1} k(\theta_1 - \theta_2)$$

$$\ddot{\theta}_1 - \ddot{\theta}_2 = \frac{1}{I_E} \theta_1, \quad \ddot{\theta}_1 = \frac{J^2}{I_E} \theta_1$$

$$\begin{bmatrix} \ddot{\theta}_1 \\ \ddot{\theta}_2 \end{bmatrix} = \begin{bmatrix} -\frac{1}{I} MgL \sin \theta_1 - I^{-1} k(\theta_1 - \theta_2) \\ J^{-1} u + J^{-1} k(\theta_1 - \theta_2) \end{bmatrix}$$

2 sol / Let $x_1 = \theta_1, x_2 = \dot{\theta}_1, x_3 = \theta_2, x_4 = \dot{\theta}_2$ taking derivs.

$$\dot{x}_1 = x_2, \quad \dot{x}_2 = -\frac{1}{I} MgL \sin \theta_1 - \frac{k}{I} (x_1 - x_3), \quad \dot{x}_3 = x_4$$

$$\dot{x}_4 = \frac{k}{J} (x_1 - x_3) + \frac{1}{J} u$$

state variables $x = [x_1, \dot{x}_1, x_2, \dot{x}_2] \in \mathbb{R}^4$

2nd sol (2 order ss 2 states could be different accordingly 2 RHS or $\ddot{x} = g(x)$ will change)

$$2 / \text{let } x_1 = \theta_1, x_3 = \theta_2 \quad \ddot{\theta}_1 = \frac{1}{I_E} \theta_1 = \frac{1}{I_E} x_1 = \dot{x}_1 = x_2$$

$$\ddot{\theta}_2 = \frac{1}{I_E} \theta_2 = \frac{1}{I_E} x_3 = x_4 \quad \ddot{x}_2 = \frac{1}{I_E} x_3 = x_4$$

$$\ddot{\theta}_1 = \frac{1}{I_E} x_1 = \frac{1}{I_E} \dot{x}_1 = \ddot{x}_1 = x_2$$

$$\ddot{x}_3 = \ddot{\theta}_2 = x_4$$

$$\ddot{\theta}_1 = -I^{-1} MgL \sin \theta_1 - I^{-1} k(\theta_1 - \theta_2)$$

$$\ddot{\theta}_2 = J^{-1} u + J^{-1} k(\theta_1 - \theta_2)$$

$$\ddot{x}_2 = \frac{1}{I_E} x_2 = \frac{1}{I_E} \dot{x}_1 = \ddot{x}_1 = -I^{-1} MgL \sin \theta_1 - I^{-1} k(\theta_1 - \theta_2)$$

$$\ddot{x}_4 = \frac{1}{I_E} x_4 = \frac{1}{I_E} \ddot{\theta}_2 = \ddot{x}_2 = J^{-1} u + J^{-1} k(\theta_1 - \theta_2)$$

$$x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} \theta_1 \\ \dot{\theta}_1 \\ \theta_2 \\ \dot{\theta}_2 \end{bmatrix} \quad \therefore x = [\theta_1, \dot{\theta}_1, \theta_2, \dot{\theta}_2] \in \mathbb{R}^4$$

\ week 1 Sheet /

$$3 \check{y} = y - M^{-1}ky - M^{-1}c_1\check{y} - M^{-1}c_2\check{y}|y| \quad \dots$$

$$\text{1 } x_1 = y \quad \therefore \quad \check{y} = \frac{d}{dt}y = \frac{d}{dt}x_1 = \dot{x}_1 = x_2$$

$$\check{y} = \frac{d}{dt}\check{y} = \frac{d}{dt}x_2 = \dot{x}_2 = x_3$$

$$x_1 = y \quad \therefore \quad \dot{x}_1 = \dot{y} = x_2$$

$$\check{x}_2 = \check{y} = y - M^{-1}ky - M^{-1}c_1\check{y} - M^{-1}c_2\check{y}|y| \quad \dots$$

$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} y \\ \check{y} \end{bmatrix} \quad \therefore x = [y, \check{y}] \in \mathbb{R}^2$$

$$3 \check{y} = Mg - ky - c_1y - c_2\check{y}|y| \quad \dots$$

Let $x_1 = y \quad x_2 = \check{y}$ then $\dot{x}_1 = x_2$

$$\dot{x}_2 = -\frac{k}{M}x_1 - \frac{c_1}{M}x_2 - \frac{c_2}{M}x_2|x_2| + g$$

$$\text{states } x = [y \quad \check{y}] \in \mathbb{R}^2$$

$$\downarrow \quad \downarrow$$

$$x = [x_1 \quad x_2]$$

$$3 \dot{x}_1 = y \quad \therefore \quad \dot{x}_2 = \check{y} = \frac{d}{dt}y = \frac{d}{dt}x_1 = \dot{x}_1 = x_2 \quad \dots$$

$$\check{y} = \frac{d}{dt}\check{y} = \frac{d}{dt}x_2 = \dot{x}_2 = -\frac{k}{M}x_1 - \frac{c_1}{M}x_2 - \frac{c_2}{M}x_2|x_2| + g$$

$$\therefore \text{states } [y \quad \check{y}] \in \mathbb{R} \quad \therefore x = [x_1 \quad x_2]$$

$$4 \dot{r} = \frac{d}{dt}\dot{r} = \frac{d}{dt}x_2 = \dot{x}_2 = r \quad \therefore \quad \dot{r} = \frac{d}{dt}x_2 = \dot{x}_2 = x_2$$

$$\ddot{r} = \frac{d}{dt}\dot{r} = \frac{d}{dt}x_2 = \dot{x}_2 = r \quad \ddot{r} = r(t)\ddot{\theta}^2(t) - \gamma r^{-2}(t)U_1(t)$$

$$\ddot{\theta}(t) = -2\frac{1}{r(t)}\ddot{r}(t)\dot{\theta}(t) + r(t)^{-1}U_2(t) \quad \dots$$

$$\theta = x_3 \quad \therefore \quad \dot{\theta} = \dot{x}_3 = x_4 \quad \therefore \quad \ddot{\theta} = \dot{x}_4 = -2r^{-1}(t)\ddot{r}(t)\dot{\theta}(t) + r^{-1}(t)U_2(t)$$

$$\text{compact state: } x = [\dot{r} \quad r \quad \dot{\theta} \quad \ddot{\theta}] \quad \therefore x = [x_1 \quad x_2 \quad x_3 \quad x_4]$$

$$\ddot{r}(t) = F(t)\ddot{\theta}^2(t) - \gamma r^{-2}(t) + U_1(t) \quad \text{(1)}$$

$$\cancel{\ddot{\theta}(t) = F(t)\ddot{\theta}^2(t) - \gamma r^{-2}(t) + U_2(t)} \quad \text{From (1) it can be seen that}$$

$$4 \text{ this is a nonlinear system of fourth order } \ddot{r}(t) = \frac{d^2r(t)}{dt^2}$$

$$\ddot{\theta}(t) = \frac{d^2\theta(t)}{dt^2} \quad \text{introduce state variants } x_1 = r(t) \quad x_2 = \dot{r}(t)$$

$$x_3 = \theta(t) \quad x_4 = \dot{\theta}(t) \quad \text{taking time derivatives of these state variants yields}$$

$$\dot{x}_1 = \ddot{r}(t) = r_2(t) \quad \text{from defn of state vector}$$

$$\dot{x}_2 = \ddot{\theta}(t) = r(t) \dot{\theta}^2(t) - \gamma r^{-2}(t) + u_1(t)$$

$$\dot{x}_3 = \dot{\theta}(t) = x_4(t) \quad \dot{x}_4 = \ddot{\theta}(t) = \frac{1}{r(t)} [-\dot{r}(t) \dot{\theta}(t) + u_2(t)] \quad \text{Subbing state vars}$$

Say $r(t), \dot{\theta}(t), \ddot{r}(t), \ddot{\theta}(t)$ etc

$$\dot{x}_1 = x_2 \quad \dot{x}_2 = x_3 x_4^2 - \gamma x_1^{-2} + u_1, \quad \dot{x}_3 = x_4$$

$$\dot{x}_4 = -2x_2 x_4 x_1^{-1} + u_2 x_1^{-1} \quad \text{Z above is in Z compact state space form}$$

$$\boxed{46 \text{ sols}} / x_1 = r(t), x_2 = \dot{r}(t), x_3 = \theta(t), x_4 = \ddot{\theta}(t) \quad \therefore$$

$$\dot{x}_1 = \ddot{r}(t) = x_2$$

$$\dot{x}_2 = \ddot{\theta}(t) = r(t) \dot{\theta}^2(t) - \gamma r^{-2}(t) + u_1(t)$$

$$\dot{x}_3 = \dot{\theta}(t) = x_4(t) \quad \therefore$$

$$\dot{x}_4 = \ddot{\theta}(t) = \frac{1}{r(t)} [-\dot{r}(t) \dot{\theta}(t) + u_2(t)] \quad \therefore \quad x = [x_1, x_2, x_3, x_4]^T = [r(t), \dot{r}(t), \theta(t), \ddot{\theta}(t)]^T$$

Say $r(t), \dot{\theta}(t), \ddot{r}(t), \ddot{\theta}(t)$, etc

$$\dot{x}_1 = x_2 \quad \dot{x}_2 = x_3 x_4^2 - \gamma x_1^{-2} + u_1, \quad \dot{x}_3 = x_4$$

$$\dot{x}_4 = -2x_2 x_4 x_1^{-1} + u_2 x_1^{-1} \quad \text{is compact state space form}$$

$$\boxed{46 \text{ sols}} / \tilde{x}(t) = [0, 0, \sqrt{\gamma}, 0]^T$$

$$\boxed{46 \text{ sols}} / \dot{x}(t) = F(x, u) \text{ where } x = [x_1, x_2, x_3, x_4]^T \text{ is}$$

$$[r(t), \dot{r}(t), \theta(t), \ddot{\theta}(t)]^T \quad F = [s_1(\cdot), s_2(\cdot), s_3(\cdot), s_4(\cdot)]^T$$

$$u = [u_1, u_2]^T \quad F = [s_1(\cdot), s_2(\cdot), s_3(\cdot), s_4(\cdot)]^T$$

$$\text{is } \tilde{x}(t) = P \begin{bmatrix} 1 & 0 & \dot{r}'(t) & \ddot{r}'(t) \end{bmatrix}^T \quad \tilde{u}(t) = (0, 0)^T \quad \text{Linear sol}$$

eqn about this trajectory is $\dot{x} = Ax + Bu$ where

$$A = \begin{bmatrix} \frac{\partial s_1}{\partial x_1} & \frac{\partial s_1}{\partial x_2} & \frac{\partial s_1}{\partial x_3} & \frac{\partial s_1}{\partial x_4} \\ \frac{\partial s_2}{\partial x_1} & \frac{\partial s_2}{\partial x_2} & \frac{\partial s_2}{\partial x_3} (s_2) & \frac{\partial s_2}{\partial x_4} \\ \frac{\partial s_3}{\partial x_1} & \frac{\partial s_3}{\partial x_2} & (s_3) x_3 & \frac{\partial s_3}{\partial x_4} \\ \frac{\partial s_4}{\partial x_1} & \frac{\partial s_4}{\partial x_2} & \frac{\partial s_4}{\partial x_3} & \frac{\partial s_4}{\partial x_4} \end{bmatrix} \quad \boxed{(x(0), u(0))}$$

$$\begin{bmatrix} 0 & 1 & 0 & 0 \\ x_1^2 + 2\gamma x_1^{-3} & 0 & 0 & -2x_1 x_4 \\ 0 & 0 & 0 & 1 \\ -2x_2 x_4 x_1^{-2} - u_2 x_1^{-2} & -2x_4 x_1^{-1} & 0 & -2x_2 x_1^{-1} \end{bmatrix} \quad \boxed{(\tilde{x}(0), \tilde{u}(0))} \quad \text{and}$$

$$B = \begin{bmatrix} \frac{\partial s_1}{\partial u_1} & \frac{\partial s_1}{\partial u_2} \\ \frac{\partial s_2}{\partial u_1} & \frac{\partial s_2}{\partial u_2} \\ \frac{\partial s_3}{\partial u_1} & \frac{\partial s_3}{\partial u_2} \\ \frac{\partial s_4}{\partial u_1} & \frac{\partial s_4}{\partial u_2} \end{bmatrix} \quad \boxed{(\tilde{x}(0), \tilde{u}(0))} \quad = \begin{bmatrix} 0 & 0 \\ 1 & 0 \\ 0 & 0 \\ 0 & x_1^{-1} \end{bmatrix} \quad \boxed{(\tilde{x}(0), \tilde{u}(0))} \quad = \begin{bmatrix} 0 & 0 \\ 1 & 0 \\ 0 & 0 \\ 0 & 1 \end{bmatrix}$$

Au :

Week 1 sheet / $A = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 3r & 0 & 0 & 2\sqrt{r} \\ 0 & 0 & 0 & 1 \\ 0 & -2\sqrt{r} & 0 & 0 \end{bmatrix}$ states $x = [r, \dot{r}, \theta, \dot{\theta}]$

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

$\check{x}(t) = [1 \ 0 \ \sqrt{r} \ t \ \sqrt{r}]^T \quad \tilde{u}(t) = (0, 0)^T$ linearised

long eqn about this trajectory is $\dot{x} = Ax + Bu$ where

$$A = \begin{bmatrix} \frac{\partial \dot{x}_1}{\partial x_1} & \frac{\partial \dot{x}_1}{\partial x_2} & \frac{\partial \dot{x}_1}{\partial x_3} & \frac{\partial \dot{x}_1}{\partial x_4} \\ \frac{\partial \dot{x}_2}{\partial x_1} & \frac{\partial \dot{x}_2}{\partial x_2} & \frac{\partial \dot{x}_2}{\partial x_3} & \frac{\partial \dot{x}_2}{\partial x_4} \\ \frac{\partial \dot{x}_3}{\partial x_1} & \frac{\partial \dot{x}_3}{\partial x_2} & \frac{\partial \dot{x}_3}{\partial x_3} & \frac{\partial \dot{x}_3}{\partial x_4} \\ \frac{\partial \dot{x}_4}{\partial x_1} & \frac{\partial \dot{x}_4}{\partial x_2} & \frac{\partial \dot{x}_4}{\partial x_3} & \frac{\partial \dot{x}_4}{\partial x_4} \end{bmatrix} \Big| (\tilde{x}(t), \tilde{u}(t))$$

$$\begin{bmatrix} 0 & 1 & 0 & 0 \\ r\dot{r}^2 + 2\sqrt{r}\dot{x}_1^{-3} & 0 & 0 & 2x_1\dot{x}_4 \\ 0 & 0 & 0 & 1 \\ -2x_2x_4x_1^{-2} - u_2x_1^{-1} & -2x_4x_1^{-1} & 0 & -2x_2x_1^{-1} \end{bmatrix} \Big| (\tilde{x}(t), \tilde{u}(t))$$

$$\begin{bmatrix} 0 & 1 & 0 & 0 \\ \frac{r}{2} & 0 & 0 & 2\sqrt{r} \\ 0 & 0 & 0 & 1 \\ 0 & -2\sqrt{r} & 0 & 0 \end{bmatrix} \quad B = \begin{bmatrix} (S_1)_{u_1} & (S_1)_{u_2} \\ (S_2)_{u_1} & (S_2)_{u_2} \\ (S_3)_{u_1} & (S_3)_{u_2} \\ (S_4)_{u_1} & (S_4)_{u_2} \end{bmatrix} \Big| (\tilde{x}(t), \tilde{u}(t))$$

$$\begin{bmatrix} (x_1)_{u_1} & (\dot{x}_1)_{u_2} \\ (x_2)_{u_1} & (\dot{x}_2)_{u_2} \\ (x_3)_{u_1} & (\dot{x}_3)_{u_2} \\ (x_4)_{u_1} & (\dot{x}_4)_{u_2} \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 1 & 0 \\ 0 & 0 \\ 0 & x_1^{-1} \end{bmatrix} \Big| (\tilde{x}(t), \tilde{u}(t)) \quad \begin{bmatrix} 0 & 0 \\ 1 & 0 \\ 0 & 0 \\ 0 & 1 \end{bmatrix} \quad \therefore \dot{x} = Ax + Bu$$

Distance $r(t)$ & angle $\theta(t)$ are measured so

$$y(t) = [r(t) \ \theta(t)]^T \quad \text{we need to write } y = CX \text{ where}$$

$y \in \mathbb{R}^2$ & $x \in \mathbb{R}^4$ so determine a C matrix $C \in \mathbb{R}^{2 \times 4}$ so

$$\begin{bmatrix} r(t) \\ \theta(t) \end{bmatrix} = C \begin{bmatrix} r(t) \\ \dot{r}(t) \\ \theta(t) \\ \dot{\theta}(t) \end{bmatrix} \quad \therefore \text{where } C = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}$$

$\therefore y(t) = [r(t) \ \theta(t)]^T \quad \therefore y = CX$

$$y \in \mathbb{R}^2 \quad \& \quad x \in \mathbb{R}^4 \quad \therefore \text{C matrix } C = \mathbb{R}^{2 \times 4} \quad \therefore$$

$$x = [r(t) \ \dot{r}(t) \ \theta(t) \ \dot{\theta}(t)]^T \quad \therefore$$

$$\begin{bmatrix} r(t) \\ \theta(t) \end{bmatrix} = C \begin{bmatrix} r(t) \\ \dot{r}(t) \\ \theta(t) \\ \dot{\theta}(t) \end{bmatrix} \quad \therefore C = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}$$

\checkmark Solve $m\ddot{x} = -k_1 x_1 - k_2 x_1^3$ let $x_1 = x$, $x_2 = \dot{x}$ then

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

$$\ddot{x}_2 = \ddot{x} = -\frac{k_1}{m}x_1 - \frac{k_2}{m}x_1^3 = -\frac{k_1}{m}x_1 - \frac{k_2}{m}x_1^3 = -\frac{k_1}{m}x_1 - \frac{k_2}{m}x_1^3$$

is in compact form with states $X = [x_1, x_2]$

\checkmark Solve let $x_1 = x$ i.e. $x_2 = \dot{x}$ then

$$\ddot{x} = \frac{d}{dt}x = \frac{d}{dt}x_1 = \dot{x}_1 = x_2 \quad \therefore$$

$$\ddot{x}_2 = \frac{d}{dt}x_2 = \frac{d}{dt}\dot{x} = \ddot{x} = -\frac{k_1}{m}x_1 - \frac{k_2}{m}x_1^3 = -\frac{k_1}{m}x_1 - \frac{k_2}{m}x_1^3$$

is in compact form with states $X = [x_1, x_2]$

\checkmark Solve For equilibrium pts solve

$$x_2 = 0 \quad \therefore -\frac{k_1}{m}x_1 - \frac{k_2}{m}x_1^3 = 0 \Rightarrow x_1^* = \pm \sqrt{-k_1/k_2}, 0$$

which is real only if $k_1 \neq k_2$ are of opposite signs

$$k_1 = -1, k_2 = \frac{1}{2} \text{ so } (0,0), (\sqrt{2},0), (-\sqrt{2},0)$$

\checkmark Solve For equilib pts Solve $x_2 = 0$:-

$$\frac{dx_2}{dt} = \frac{d}{dt}(0) = 0 = \dot{x}_2 \quad \therefore$$

$$-\frac{k_1}{m}x_1 - \frac{k_2}{m}x_1^3 = 0 = x_1 \left(-\frac{k_1}{m} - \frac{k_2}{m}x_1^2 \right) \quad \therefore x_1 = 0, -\frac{k_1}{m} - \frac{k_2}{m}x_1^2 = 0 \quad \therefore$$

$$-\frac{k_1}{m} = \frac{k_2}{m}x_1^2 \quad \therefore -\frac{k_1}{k_2} = x_1^2 \quad \therefore x_1 = \pm \sqrt{-\frac{k_1}{k_2}}, x_2 = -\sqrt{-\frac{k_1}{k_2}}$$

$$\therefore k_1 = -1, k_2 = \frac{1}{2} \quad \therefore x_1 = \sqrt{-\frac{-1}{\frac{1}{2}}} = \sqrt{2}, x_2 = -\sqrt{-\frac{-1}{\frac{1}{2}}} = -\sqrt{2}, x_1 = 0 \quad \therefore$$

$$(0,0), (\sqrt{2},0), (-\sqrt{2},0)$$

\checkmark Solve For linearised plant, determine Jacobian $A = \begin{bmatrix} \frac{\partial \delta_1}{\partial x_1} & \frac{\partial \delta_1}{\partial x_2} \\ \frac{\partial \delta_2}{\partial x_1} & \frac{\partial \delta_2}{\partial x_2} \end{bmatrix}$

$\delta_1(\cdot) := x_2 \quad \delta_2(\cdot) := -\frac{k_1}{m}x_1 - \frac{k_2}{m}x_1^3 \quad \therefore A = \begin{bmatrix} 0 & 1 \\ -k_1 - 3k_2(x_1)^2 & 0 \end{bmatrix}$

evaluating A about equilibrium:

$$\therefore \text{at } (0,0) : A_{(0,0)} = \begin{bmatrix} 0 & 1 \\ -k_1 & 0 \end{bmatrix} \text{ given } k_1 = -1 \text{ so } A_{(0,0)} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

$$\hat{X} = A_{(0,0)} X$$

$$\text{at } (\pm\sqrt{2}, 0) \quad A_{(\pm\sqrt{2}, 0)} = \begin{bmatrix} 0 & 1 \\ -k_1 - 3k_2(\pm\sqrt{2})^2 & 0 \end{bmatrix} \text{ since } k_1 = -1, k_2 = \frac{1}{2} \quad \therefore$$

$$\hat{X} = A_{(\pm\sqrt{2}, 0)} X \quad \text{where } A_{(\pm\sqrt{2}, 0)} = \begin{bmatrix} 0 & 1 \\ -2 & 0 \end{bmatrix}$$

\checkmark Solve For linearised plant, determine Jacobian $A = \begin{bmatrix} \frac{\partial \delta_1}{\partial x_1} & \frac{\partial \delta_1}{\partial x_2} \\ \frac{\partial \delta_2}{\partial x_1} & \frac{\partial \delta_2}{\partial x_2} \end{bmatrix}$

$$\delta_1 = x_2, \delta_2 = \dot{x}_2 = -\frac{k_1}{m}x_1 - \frac{k_2}{m}x_1^3 \quad \therefore A = \begin{bmatrix} 0 & 1 \\ -k_1 - 3k_2(x_1)^2 & 0 \end{bmatrix} \quad \therefore$$

evaluating A about equilibriums:

Week 1 sheet at $(0,0)$: $A_{(0,0)} = \begin{bmatrix} 0 & 1 \\ -k_1 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$, $\dot{x} = A_{(0,0)}x$

at $(\pm\sqrt{2}, 0)$: $A_{(\pm\sqrt{2}, 0)} = \begin{bmatrix} 0 & 1 \\ -k_1 - 3k_2 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -2 & 0 \end{bmatrix} \therefore \dot{x} = A_{(\pm\sqrt{2}, 0)}x$

Week 2 sheet

$\dot{x}_1 = -x_1 + x_2 \quad \dot{x}_2 = -0.1x_1 - 2x_2 - x_1^2 - 0.1x_1^3 \quad \therefore$
 $0 = -x_1 + x_2 \Rightarrow x_1 = x_2, 0 = -0.1x_1 - 2x_2 - x_1^2 - 0.1x_1^3 \Rightarrow$
 $-0.1x_1^3 - x_1^2 - 2.1x_1 \quad \therefore x_1(0.1x_1^2 - x_1 - 2.1) = 0$

3 equilibrium pts are $(0,0)^T, (-2.76, -2.76)^T, (-7.23, -7.23)^T$

$$\frac{\partial \dot{x}}{\partial x} = \begin{bmatrix} -1 & 1 \\ 0.1 - 2x_1 - 0.3x_1^2 & -2 \end{bmatrix} \quad \left. \frac{\partial \dot{x}}{\partial x} \right|_{(0,0)^T} = \begin{bmatrix} -1 & 1 \\ 0.1 & -2 \end{bmatrix} \Rightarrow$$

$\lambda_{1,2} = -2.09$ and $-0.9 \therefore (0,0)$ stable node

$$\left. \frac{\partial \dot{x}}{\partial x} \right|_{(-2.76, -2.76)^T} = \begin{bmatrix} -1 & 1 \\ 0.1 + 2 - 2.76 - 2.76^2 - 0.3 & -2 \end{bmatrix} \Rightarrow$$

$\lambda_{1,2} = -3.39$ and $0.393 \therefore (-2.76, -2.76)^T$ is saddle

$$\left. \frac{\partial \dot{x}}{\partial x} \right|_{(-7.23, -7.23)^T} = \begin{bmatrix} -1 & 1 \\ -1.168 & -2 \end{bmatrix} \quad \lambda_{1,2} = -1.5 \pm 0.958i \therefore$$

$\text{Re}(\lambda) < 0 \therefore (-7.23, -7.23)$ is stable focus

$$\left. \frac{\partial \dot{x}}{\partial x} \right|_0 = x_1(1+x_2) \quad 0 = -x_2 + x_2^2 + x_1x_2 - x_1^3 \quad \text{②} \quad \therefore$$

①: $x_1 = 0$ or $x_2 = -1 \therefore$

$$x_1 = 0 \Rightarrow 0 = -x_2 + x_2^2 \Rightarrow x_2 = 0 \text{ or } x_2 = 1 \quad \therefore$$

$x_2 = -1 \Rightarrow 0 = 2 - x_1 - x_1^3 \Rightarrow x_1 = 1 \quad \therefore$ have 3 equilibrium pts

$$\left. \frac{\partial \dot{x}}{\partial x} \right|_{(0,0)^T} = \begin{bmatrix} -1 & 1 \\ 0 & -1 \end{bmatrix} \quad \left. \frac{\partial \dot{x}}{\partial x} \right|_{(0,1)^T} = \begin{bmatrix} 1+x_2 & x_1 \\ x_2 - 3x_2^2 & -1+2x_2 + x_1 \end{bmatrix}$$

$$\left. \frac{\partial \dot{x}}{\partial x} \right|_{(0,0)^T} = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix} \Rightarrow \lambda_{1,2} = 1, -1 \quad \therefore (0,0)^T \text{ is saddle}$$

(+ve & -ve real signals)

$$\left. \frac{\partial \dot{x}}{\partial x} \right|_{(0,1)^T} = \begin{bmatrix} 2 & 0 \\ 1 & 1 \end{bmatrix} \Rightarrow \lambda_{1,2} = 2, 1 \quad 2+\text{signals} \quad (0,1)^T \text{ is unstable node}$$

$$\left. \frac{\partial \dot{x}}{\partial x} \right|_{(-1,-1)^T} = \begin{bmatrix} 0 & 1 \\ -4 & -2 \end{bmatrix} \Rightarrow \lambda_{1,2} = -1 \pm i\sqrt{3} \quad \text{Re}(\lambda) < 0 \therefore (-1,-1)^T \text{ is stable focus}$$

$$\left. \frac{\partial \dot{x}}{\partial x} \right|_0 = (x_1 - x_2)(1 - x_1^2 - x_2^2) \quad 0 = (x_1 + x_2)(1 - x_1^3 - x_2^3)$$

$\left. \frac{\partial \dot{x}}{\partial x} \right|_{(0,0)^T} = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}$ is an equilibrium set $(0,0)$ is an isolated equilibrium pt

$$\left. \frac{\partial \dot{x}}{\partial x} \right|_{(0,0)^T} = \begin{bmatrix} 1 - 3x_1^2 - x_2^2 + 2x_1x_2 & -2x_1x_2 - 1 + x_1^2 + 3x_2^2 \\ -1 - 3x_1^2 - x_2^2 - 2x_1x_2 & 2x_1x_2 + 1 - x_1^2 - 3x_2^2 \end{bmatrix} \quad \left. \frac{\partial \dot{x}}{\partial x} \right|_{(0,0)^T} = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}$$

\ Week 3 Sheet / $\dot{r} = r - \frac{1}{r^2}$, $\nabla V(x) \cdot \nabla V(x) = x_1^2 \geq 0$

$V(x) = 1 - \frac{1}{r^2}$, $\nabla V(x) \leq 0$ inward direction

$\dot{r}(x) = \frac{1}{2}$, $\nabla(x) \cdot \nabla V(x) \geq 0$ outward direction

Further note that system $\dot{x} = \nabla(x)$ has unique equilibri at origin $(0,0)^T$ & this is excluded in \mathbb{R} annulus $\frac{1}{\sqrt{2}} \leq r \leq 1$ considered.
 \therefore no equili in \mathbb{R} considered set is by PB (Poincaré-Bendixson)

criterion, \exists three must be atleast one periodic orbit, closed path - between 2 circles, ie' in Set M (closed bounded)

$$M = \{x \in \mathbb{R}^2 \mid \frac{1}{2} \leq V(x) = x_1^2 + x_2^2 \leq 1\} \text{ here } x_1^2 + x_2^2 = r^2 \therefore \frac{1}{\sqrt{2}} \leq r \leq 1$$

at least one periodic sol exists

\ Week 3 Sheet / $\dot{x}_1 = x_2 \cos x_1$, $\dot{x}_2 = \sin x_1$: equili plane $(\pm n\pi, 0)$

for $n=0, 1, 2, \dots$ linearisation provides $\begin{bmatrix} a & b \\ c & d \end{bmatrix}$ where a is $+1$, or -1 (slip)

inside any periodic orbit γ , there must be atleast one equili pt. Suppose 2 equili pts inside γ are not hyperbolic i.e. is Jacobians at that pt has \approx eigenvals on \mathbb{R} imaginary axis - given is N is number of nodes, & Soli S is number of saddles, it must be that $N-S=1$. A periodic orbit must enclose 1 equili pt st $N-S=1$ \therefore there are no periodic orbit.

\ Week 7 Q1 / \therefore miss expr w.r.t control input

$M(q)$ is symmetric Mat

$\bar{M}-2C$ is a skew symmetric mat

\bar{M} is \mathbb{R} total deriv of $M(q)$ wrt time t

D is pos semi definite symmetric Mat

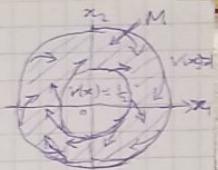
$P(q)$ is a pos def definite func

$$M(q)\ddot{q} + C(q, \dot{q})\dot{q} + D\ddot{q} + g(\dot{q}) = u$$

$$M(q) = M(q)^T \text{ symmetric P.d.S. } \forall q \in \mathbb{R}^m$$

$C(q, \dot{q})\dot{q}$ - force $[C]$ has property $\bar{M}-2C$ is skew symmetric

$$\forall q, \dot{q} \in \mathbb{R}^m$$



{skew symmetric mat $A \Rightarrow A^T = -A \quad a_{ji} = -a_{ij}$ }

$$V(\cdot) = \frac{1}{2} \dot{q}_1^T M(q) \dot{q}_1 + P(q)$$

Symmetric p.d.s
 $\forall q \in \mathbb{R}^m$

quadratic form + 1.8

$$V(\cdot) = 0 \text{ only for } q = \dot{q} = 0$$

i. $V(\cdot)$ is a p.d.s taking deriv of V along dynamics

$$\dot{V} = \frac{1}{2} \ddot{q}_1^T [M(q) \frac{d}{dt}(\dot{q}_1) + \frac{d}{dt}(M(q)) \dot{q}_1] = \frac{1}{2} \ddot{q}_1^T M(q) \dot{q}_1 + \frac{\partial f(q)}{\partial q} \dot{q}_1 =$$

$$\frac{1}{2} \ddot{q}_1^T M(q) \dot{q}_1 + \frac{1}{2} \ddot{q}_1^T \tilde{M}(q) \dot{q}_1 + \frac{1}{2} \ddot{q}_1^T M(q) \dot{q}_1 + \frac{\partial f(q)}{\partial q} \dot{q}_1 \quad \{ (AB)^T = B^T A^T \}$$

$$(\ddot{q}_1^T (M(q) \dot{q}_1))^T = (M(q) \dot{q}_1)^T \ddot{q}_1 = \ddot{q}_1^T M^T(q) \dot{q}_1 = \ddot{q}_1^T M(q) \dot{q}_1 \quad \{ M(q) = M^T(q) \}$$

$$= \ddot{q}_1^T M(q) \dot{q}_1 + \frac{1}{2} \ddot{q}_1^T \tilde{M}(q) \dot{q}_1 + g^T(\cdot) \dot{q}_1$$

we will sub from dynamics ($\because \frac{\partial f(q)}{\partial q} = g^T(\cdot)$ by def)

$$M(q) \dot{q}_1 + ((q_1 \dot{q}_1) \dot{q}_1 + D \dot{q}_1 + g(\cdot)) = 0 \quad (u = -k_1 \dot{q}_1)$$

$$= \ddot{q}_1^T [-C(q, \dot{q}) \dot{q}_1 - D \dot{q}_1 - g(\cdot)] + \frac{1}{2} \ddot{q}_1^T \tilde{M}(q) \dot{q}_1 + g^T(\cdot) \dot{q}_1 =$$

$$- \ddot{q}_1^T C(\cdot) \dot{q}_1 - \ddot{q}_1^T D \dot{q}_1 - \ddot{q}_1^T g(\cdot) + \frac{1}{2} \ddot{q}_1^T \tilde{M}(\cdot) \dot{q}_1 + g^T(\cdot) \dot{q}_1 =$$

$$\frac{1}{2} \ddot{q}_1^T (\tilde{M} - 2C) \dot{q}_1 - \ddot{q}_1^T D \dot{q}_1 - \underbrace{\ddot{q}_1^T g(\cdot)}_{\text{ok: } \tilde{M}-2C \text{ is skewsymmetric}} + g^T(\cdot) \dot{q}_1$$

$$= \ddot{q}_1^T (g^T(-) \dot{q}_1) = \ddot{q}_1^T (g^T(-))^T = \ddot{q}_1^T g(\cdot)$$

$\ddot{q}_1^T (-) \dot{q}_1$ p.s.d.s

$$\ddot{V}(\cdot) = -\ddot{q}_1^T D \dot{q}_1 \quad \ddot{V}(\cdot) \leq 0$$

$V(\cdot)$ p.d.s & $\ddot{V}(\cdot)$ neg semi \Rightarrow origin is this case u=0

is lyapunov stab if we use $u = -k_1 \dot{q}_1$ a control law gives

$$\ddot{V}(\cdot) = -\ddot{q}_1^T (k_1 \dot{q}_1 - D) \dot{q}_1 \text{ provided } k_1 \text{ p.d.s}$$

$$\ddot{V}(\cdot) \leq 0 \quad (\because \text{indep of } q)$$

$$S = \{ (q, \dot{q}) \mid \ddot{V}(\cdot) = 0 \}$$

$$\ddot{V}(\cdot) = 0 \Rightarrow \dot{q} = 0 \rightarrow \ddot{q} = 0$$

$$M(q) \dot{q}_1 + C(\cdot) \dot{q}_1 + D \dot{q}_1 + g(\cdot) = -k_1 \dot{q}_1 \Rightarrow g(\cdot) = 0 \text{ note } g(\cdot) = 0 \Rightarrow \dot{q} = 0 \text{ (by def of } g(\cdot))$$

i. by LaSalle invariance corollary we converge to

origin $(q, \dot{q}) : (0, 0)$ is asymptotic stab

Week 8 Q1 $\dot{x}_1 = x_2$ $\dot{x}_2 = x_1 + (x_1^2 - 1)x_2$ to assess non-symp stability
(method of linearisation)

- Determine region of attraction (estim)

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ x_1^2 x_2 \end{bmatrix}$$

linear part nonlinearity

$$\dot{x} = Ax + g(x), \quad x = (x_1, x_2)^T$$

2 matrix $[A]$ has eigenvalues $\frac{-1 \pm \sqrt{3}}{2}$

$\text{Re}(\lambda_i(A)) < 0, i=1,2$, can say $[A]$ is Hurwitz \therefore

2 linear part above $\dot{x} = Ax$ is asymptotic stable (global)

by converse Lyapunov thm \exists a symmetric positive definite P

satisfies $A^T P + PA = -Q$, for some $Q = Q^T > 0$

let Q be $I_{2 \times 2} \Rightarrow A^T P + PA = -I$ choose $P = \begin{bmatrix} p_{11} & p_{12} \\ p_{21} & p_{22} \end{bmatrix} \therefore P = P^T$

we know $A \geq 0 \therefore$ by solving simultaneous eqns

$$\left\{ \begin{array}{l} x_0 \in \{x \in D \mid V(x) \leq \lambda_{\min}(P) r^2\} \\ \lambda_{\min}(P) = 0.691 \end{array} \right.$$

$$P = \begin{bmatrix} 3/2 & -1/2 \\ -1/2 & 1 \end{bmatrix} \quad \lambda_{\min}(P) = 0.691 \quad (\text{min eigenval})$$

so 2 Lyapunov func $\dot{x} = Ax$ is $V = x^T P x = \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} 3/2 & -1/2 \\ -1/2 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$

$$V(x) = 1.5x_1^2 + x_2^2 - x_1 x_2$$

$$V(x) = x^T P x \quad \text{p.s.d.}$$

deriv along 2 trajectory $\dot{V}(x) = 3x_1 \dot{x}_1 + 2x_2 \dot{x}_2 - x_1 \dot{x}_2 - x_2 \dot{x}_1$

$$= (3x_1 - x_2) \dot{x}_1 + (-x_1 + 2x_2) \dot{x}_2 =$$

$$(3x_1 - x_2)(-x_2) + (-x_1 + 2x_2)(x_1 + (x_1^2 - 1)x_2) =$$

$$-3x_1 x_2 + x_2^2 - x_1^2 - x_1^3 x_2 + x_1 x_2 + 2x_1 x_2 + 2x_2^2 x_1^2 - 2x_2^2 =$$

$$-x_1^2 - x_2^2 - x_1^3 x_2 + 2x_2^2 x_1^2 =$$

$$-(x_1^2 + x_2^2) - x_1^2 x_2 (x_1 - 2x_2) \leq -||x||^2 + ||x_1|| ||x_2|| ||(x_1 - 2x_2)||$$

$$\leq -||x||^2 + ||x_1|| ||x_2|| ||x_1 - 2x_2|| \quad \begin{cases} ||x_1|| \leq ||x|| \text{ euclidean norm} \\ ||x_1, x_2|| \leq \frac{1}{2} ||x||^2 \end{cases}$$

$$\leq -||x||^2 + ||x_1|| \frac{1}{2} ||x||^2 \sqrt{5} ||x||$$

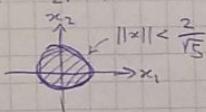
$$\leq -\left(1 - \frac{\sqrt{5}}{2} ||x||^2\right) ||x||^2 \quad \begin{cases} \text{we provide} \\ \text{we provide} \end{cases}$$

$$1 - \frac{\sqrt{5}}{2} ||x||^2 = 0 \quad 0 \leq ||x||^2 < \frac{2}{\sqrt{5}}$$

Choose $\frac{2}{\sqrt{5}} = r^2$ & following 2 arguments $\forall \alpha > 0 \exists r > 0$ s.t.

$$||x_1, x_2|| \leq \frac{1}{2} ||x||^2$$

$$||x_1 - 2x_2|| \leq \sqrt{1 - 2^2} ||x|| \leq \sqrt{5} ||x||$$



$$\|g(x)\| \leq \delta \|x\|, \forall \|x\| < r \quad g(x) = \begin{bmatrix} 0 \\ s^2 x_1 \end{bmatrix}$$

can write \mathbb{R}^2 as region of attraction

$$D_c = \left\{ x \in \mathbb{R}^2 \mid V(x) \leq c \right\} \text{ where } c = \lambda_{\min}(P) = \frac{\alpha + \beta}{\sqrt{\gamma}}$$

$$\begin{cases} 2\sin x \cos x = \sin 2x & \sin(x+y) = \sin x \cos y + \cos x \sin y \\ \sin(x-y) \sin x \cos y = \cos x \sin y & \\ \sin\left(\frac{x+y}{2}\right) = \sin\left(\frac{x+y}{2}\right) = \sin\left(\frac{\pi}{2}\right) \cos\left(\frac{\pi}{2}\right) + \cos\left(\frac{\pi}{2}\right) \sin\left(\frac{\pi}{2}\right) \end{cases}$$

$$\sin x \cos(x-y) = \sin x \sin y + \cos x \cos y \quad \therefore$$

$$\cos(x-y) = \cos\left(\frac{\pi}{2} - \frac{\pi}{2}\right) = \sin\left(\frac{\pi}{2}\right) \sin\left(\frac{\pi}{2}\right) + \cos\left(\frac{\pi}{2}\right) \cos\left(\frac{\pi}{2}\right)$$

$$2 \sin\left(\frac{x+y}{2}\right) \cos\left(\frac{x-y}{2}\right) =$$

$$2 \left(\sin\left(\frac{\pi}{2}\right) \cos\left(\frac{\pi}{2}\right) + \cos\left(\frac{\pi}{2}\right) \sin\left(\frac{\pi}{2}\right) \right) \left(\sin\left(\frac{\pi}{2}\right) \sin\left(\frac{\pi}{2}\right) + \cos\left(\frac{\pi}{2}\right) \cos\left(\frac{\pi}{2}\right) \right) =$$

$$2 \left[\sin^2\left(\frac{\pi}{2}\right) \cos\left(\frac{\pi}{2}\right) \sin\left(\frac{\pi}{2}\right) + \sin\left(\frac{\pi}{2}\right) \cos^2\left(\frac{\pi}{2}\right) \cos\left(\frac{\pi}{2}\right) + \sin\left(\frac{\pi}{2}\right) \cos\left(\frac{\pi}{2}\right) \sin\left(\frac{\pi}{2}\right) + \cos^2\left(\frac{\pi}{2}\right) \sin\left(\frac{\pi}{2}\right) \cos\left(\frac{\pi}{2}\right) \right]$$

$$\sin x \cos x = \frac{1}{2} [\sin(x+y) + \sin(x-y)]$$

$$\text{Week 5 Sheet} / \begin{cases} A/x_1 = Ax + Bu & y = Cx \\ A = \begin{bmatrix} 1 & -2 \\ 0 & 0 \end{bmatrix}, B = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \\ C = \begin{bmatrix} 0 \end{bmatrix} \therefore x_1 = \begin{bmatrix} -1 & -2 \\ 0 & 0 \end{bmatrix} x + \begin{bmatrix} 1 \\ 0 \end{bmatrix} u & y = \begin{bmatrix} 0 \end{bmatrix} x \end{cases}$$

$$\text{Controllability mat } M = [B : AB] = \begin{bmatrix} 1 & 1 & -2 \\ 0 & 1 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$$

* Check controllable or not rank(M) = 2 = full rank \therefore Controllable or $\det(M) = 1 \neq 0 \therefore$ Controllable

\b/ desired poles : $4, -5 \therefore$ associated characteristic polynomial $s^2 + 9s + 20 = 0 \therefore (s+4)(s+5)$ with $K = [k_1 \ k_2]$ (gains). \mathbb{Z} closed

loop characteristic polynomial is $\det(sI - (A - BK)) = 0 \therefore$

$$\det\left(\begin{bmatrix} s & 0 \\ 0 & s \end{bmatrix} - \begin{bmatrix} -(1+k_1) & -(2+k_2) \\ 1 & 0 \end{bmatrix}\right) = \det\left(\begin{bmatrix} s+(1+k_1) & 2+k_2 \\ 1 & s \end{bmatrix}\right) = 0 \therefore BK = \begin{bmatrix} 1 \\ 0 \end{bmatrix} [k_1 \ k_2] = \begin{bmatrix} k_1 & k_2 \\ 0 & 0 \end{bmatrix}$$

$$\therefore s^2 + (1+k_1)s + (2+k_2) = 0 \therefore k_1 = 9 \& k_2 = 20 \therefore$$

$$k_1 = 8 \& k_2 = 18 \therefore K = \begin{bmatrix} 8 & 18 \end{bmatrix} \therefore u = -[8 \ 18]x$$

\c/ transmission function associated is T.F $G(s) = C(sI - A)^{-1}B =$

$$\begin{bmatrix} 1 & 0 \end{bmatrix} \cdot \text{inv}\left(\begin{bmatrix} s+1 & 2 \\ -1 & s \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 0 \end{bmatrix}\right) = \begin{bmatrix} 1 & 0 \end{bmatrix} \cdot \frac{1}{s^2 + 9s + 20} \begin{bmatrix} s & -2 \\ 1 & s+1 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 0 \end{bmatrix} \quad \left| \det(sI - A) = \frac{1}{s^2 + 9s + 20} \right.$$

$$= \frac{1}{s^2 + 9s + 20} [s - 2] \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \frac{s}{s^2 + 9s + 20}$$

Week 5 Sheet / $n = T \tilde{x} \therefore T = MW$, M is controllability mat

$$M = \begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix} \text{ full rank} \therefore n=2 \therefore W = \begin{bmatrix} a_{11} & 1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} a_1 & 1 \\ 1 & 0 \end{bmatrix}$$

$$\det(SI - A) = S^2 + a_1 S + a_2 = \det \left(\begin{bmatrix} S & 0 \\ 0 & S \end{bmatrix} - \begin{bmatrix} -1 & -2 \\ 1 & 0 \end{bmatrix} \right) = \det \left(\begin{bmatrix} S+1 & -2 \\ -1 & S \end{bmatrix} \right) =$$

$$(S+1)S - (-1)2 = S^2 + S + 2 \therefore a_1 = 1 \therefore$$

$$W = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \therefore T = MW = \begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

$$\checkmark a/\text{controllability mat} : M = [B : AB] = \begin{bmatrix} 1 \\ 0 \end{bmatrix} : \begin{bmatrix} -1 & -2 \\ 1 & 0 \end{bmatrix} =$$

$\begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix}$ is full rank mat. $\therefore \text{rank}(M)=2$ is controllable

or $\det(M) = 1 \neq 0 \therefore \text{controllable}$

$\checkmark d/\text{desired poles: } 4 \pm 5j$. associated charac poly: $(S-4)(S+5) =$

$$S^2 + 9S + 20 = 0 \therefore \text{with } K = [k_1 \ k_2] : 2 \text{ closed loop}$$

$$\text{charac poly is } \det(SI - (4+BK)) = 0 = \det \left(\begin{bmatrix} S & 0 \\ 0 & S \end{bmatrix} - \begin{bmatrix} -1 & -2 \\ 1 & 0 \end{bmatrix} - \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} [k_1 \ k_2] \right) =$$

$$\det \left(\begin{bmatrix} S & 0 \\ 0 & S \end{bmatrix} - \begin{bmatrix} -1 & -2 \\ 1 & 0 \end{bmatrix} - \begin{bmatrix} k_1 & k_2 \\ 0 & 0 \end{bmatrix} \right) = \det \left(\begin{bmatrix} S & 0 \\ 0 & S \end{bmatrix} - \begin{bmatrix} -(1+k_1) & -2+k_2 \\ 1 & 0 \end{bmatrix} \right) =$$

$$\det \left(\begin{bmatrix} S+(1+k_1) & -2+k_2 \\ 1 & 0 \end{bmatrix} \right) = 0 = S^2 + (1+k_1)S + (2+k_2) = S^2 + 9S + 20 \therefore$$

$$1+k_1=9 \quad 2+k_2=20 \quad \therefore k_1=8 \quad k_2=18 \quad \therefore K = \begin{bmatrix} 8 & 18 \end{bmatrix} \therefore$$

$$u = -[8 \ 18]x$$

$\checkmark e/\text{transient function is T.F. } G(s) = C(SI - A)^{-1}B =$

$$[C \ 0] \left(\begin{bmatrix} S & 0 \\ 0 & S \end{bmatrix} - \begin{bmatrix} -1 & -2 \\ 1 & 0 \end{bmatrix} \right)^{-1} \begin{bmatrix} 1 \\ 0 \end{bmatrix} =$$

$$[0 \ 1] \left(\begin{bmatrix} S+1 & 2 \\ -1 & S \end{bmatrix} \right)^{-1} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = [0 \ 1] \frac{1}{S^2 + S + 2} \begin{bmatrix} S & -2 \\ 1 & S+1 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} \neq$$

$$\text{but if } C = [1 \ 0]; \quad G(s) = [1 \ 0] \frac{1}{S^2 + S + 2} \begin{bmatrix} S & -2 \\ 1 & S+1 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} =$$

$$\frac{1}{S^2 + S + 2} [S \ -2] \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \frac{1}{S^2 + S + 2} [S] = \frac{S}{S^2 + S + 2}$$

$\checkmark f/\tilde{x} = T \tilde{x} \therefore T = MW \therefore M$ is controllability mat $M = [B : AB] = \begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix}$

$$\text{is full rank} \therefore n=2 \therefore W = \begin{bmatrix} a_{11} & 1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} a_1 & 1 \\ 1 & 0 \end{bmatrix} \therefore$$

$$\det(SI - A) = S^2 + a_1 S + a_2 = \det \left(\begin{bmatrix} S & 0 \\ 0 & S \end{bmatrix} - \begin{bmatrix} -1 & -2 \\ 1 & 0 \end{bmatrix} \right) = \det \left(\begin{bmatrix} S+1 & -2 \\ -1 & S \end{bmatrix} \right) =$$

$$(S+1)S - (-1)2 = S^2 + S + 2 = S^2 + a_1 S + a_2 \therefore a_1 = 1 \therefore$$

$$W = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \therefore T = MW = \begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

✓ charac eqn is $\det \begin{bmatrix} s & -1 \\ 1 & s \end{bmatrix} = s^2 + 1 = 0$

$$\det(SI - A) = 0 = \det \left(\begin{bmatrix} s & 0 \\ 0 & s \end{bmatrix} - \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} \right) = \det \left(\begin{bmatrix} s & -1 \\ 1 & s \end{bmatrix} \right) = 0 =$$

$$s^2 - 1(-1) = s^2 + 1 = 0 \}$$

Controllability Mat: $M = [B : AB] = \left[\begin{bmatrix} 0 \\ 1 \end{bmatrix} : \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right] = \left[\begin{bmatrix} 0 \\ 1 \end{bmatrix} : \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right] =$

$$M = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \quad \{ \det(M) = 0(0) - 1(1) = -1 \neq 0 \therefore M \text{ is full rank} \}$$

$$\text{rank}(M) = 2 \therefore n=2 \therefore W = \begin{bmatrix} a_{11} & 1 \\ -1 & 0 \end{bmatrix} = \begin{bmatrix} a_1 & 1 \\ 1 & 0 \end{bmatrix} \therefore$$

$$\det(SI - A) = s^2 + a_1s + a_2 = s^2 + 1 = s^2 + 0s + 1 = 0 \therefore a_1 = 0 \therefore W = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

$$\therefore T = MW = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I = \text{identity mat} \therefore$$

A, B is in Canonical form.

✓ Controllability Mat is $M = [B : AB] = \left[\begin{bmatrix} 0 \\ 1 \end{bmatrix} : \begin{bmatrix} -1 & 0 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right] =$

$$\left[\begin{bmatrix} 0 \\ 1 \end{bmatrix} : \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right] = M = \begin{bmatrix} 0 & 0 \\ 1 & -1 \end{bmatrix} \quad \{ \det(M) = -1(0) - 1(0) = 0 - 0 = 0 \},$$

not full rank. $\therefore \text{rank}(M) \neq 2 \therefore \text{rank}(M) = 1$ so Z syst is not controllable.

$$A - BK = \begin{bmatrix} -1 & 0 \\ 1 & -1 \end{bmatrix} - \begin{bmatrix} 0 \\ 1 \end{bmatrix} \begin{bmatrix} k_1 & k_2 \end{bmatrix} = \begin{bmatrix} -1 & 0 \\ 1 & -1 \end{bmatrix} - \begin{bmatrix} k_1 & k_2 \\ -k_1 & -k_2 \end{bmatrix} \text{ where } K = \begin{bmatrix} k_1 & k_2 \end{bmatrix}$$

$$\therefore Z \text{ charac eqn is } \det(SI - (A - BK)) = 0 = \det \left(\begin{bmatrix} s & 0 \\ 0 & s \end{bmatrix} - \begin{bmatrix} -1 & 0 \\ 1 & -1 \end{bmatrix} - \begin{bmatrix} k_1 & k_2 \\ -k_1 & -k_2 \end{bmatrix} \right) =$$

$$\det \begin{bmatrix} s+1 & 0 \\ k_1 - 1 & s+1+k_2 \end{bmatrix} = 0 = (s+1)(s+1+k_2) = 0 \therefore \text{clearly } \lambda = -1 \text{ is an eigen}$$

for all k_1 and k_2 . $\therefore -1$ is a closed loop eigenval $\forall k_1$ and $k_2 \therefore \forall K$

✓ $s_1 + s_2, s_1 s_2, s_1, s_2$: $\forall x_0 \in \mathbb{R} \exists +\nu \in \mathcal{J}, L_1, L_2$; k_1 and k_2 st

$$|s_1(x) - s_1(y)| \leq L_1 |x - y| \text{ and } |s_2(x) - s_2(y)| \leq L_2 |x - y| \quad \{ \therefore s_1 \text{ and } s_2 \text{ are locally Lipschitz} \}$$

$$\therefore |s_1(x)| \leq k_1, |s_2(x)| \leq k_2 \quad \forall x, y \in \{x \in \mathbb{R} \mid |x_0 - y_0| < r\}$$

$$\text{for } S: s_1 + s_2: |S(x) - S(y)| = |s_1(x) - s_1(y) + s_2(x) - s_2(y)| \leq$$

$$|(s_1(x) + s_2(x)) - (s_1(y) + s_2(y))| = |(s_1(x) - s_1(y)) + (s_2(x) - s_2(y))| \leq$$

$$|s_1(x) - s_1(y)| + |s_2(x) - s_2(y)| \leq L_1 |x - y| + |s_2(x) - s_2(y)| \leq$$

$$L_1 |x - y| + L_2 |x - y| = (L_1 + L_2) |x - y| \therefore |S(x) - S(y)| \leq (L_1 + L_2) |x - y| \text{ for}$$

$S = s_1 + s_2 \therefore s_1 + s_2$ is locally lipschitz

$$\text{for } S = s_1 s_2: |S(x) - S(y)| = |s_1(x)s_2(x) - s_1(y)s_2(y)| =$$

$$|s_1(x)s_2(x) + (s_1(x)s_2(y) - s_1(x)s_2(x)) - s_1(y)s_2(y)| =$$

$$|(s_1(x)s_2(x) - s_1(x)s_2(y)) + (s_1(x)s_2(y) - s_1(y)s_2(y))| \leq |s_1(x)s_2(x) - s_1(x)s_2(y)| + |s_1(x)s_2(y) - s_1(y)s_2(y)|$$

Week 5 Sheet / $= |\delta_1(x)| |\delta_2(x) - \delta_2(y)| + |\delta_2(y)| |\delta_1(x) - \delta_1(y)| \leq$
 $k_1 |\delta_2(x) - \delta_2(y)| + |\delta_2(y)| |\delta_1(x) - \delta_1(y)| \leq k_1 L_2 |x-y| + |\delta_2(y)| |\delta_1(x) - \delta_1(y)| \leq$
 $\rightarrow k_1 L_2 |x-y| + k_2 |\delta_1(x) - \delta_1(y)| \leq k_1 L_2 |x-y| + k_2 L_1 |x-y| =$
 $(k_1 L_2 + k_2 L_1) |x-y| \therefore |\delta(x)| \leq k_1, |\delta_2(y)| \leq k_2 \therefore$
 $|\delta(x) - \delta(y)| \leq (k_1 L_2 + k_2 L_1) |x-y| \text{ since } \delta = \delta_1 \delta_2 \text{ is locally Lipschitz}$

Since $\delta = \delta_1 \circ \delta_2$: $|\delta(x) - \delta(y)| = |\delta_1(\delta_2(x)) - \delta_1(\delta_2(y))| \leq L_2 |\delta_1(x) - \delta_1(y)|$

$\left\{ \begin{array}{l} |\delta_1(x) - \delta_1(y)| \leq L_2 |x-y| \therefore |\delta_1(\delta_2(x)) - \delta_1(\delta_2(y))| \leq L_2 |\delta_2(x) - \delta_2(y)| \\ |\delta(x) - \delta(y)| \leq L_2 |\delta_1(x) - \delta_1(y)| \leq L_2 L_1 |x-y| \therefore \delta = \delta_1 \delta_2 \text{ is locally Lipschitz} \end{array} \right.$

$$\checkmark 5 / \begin{aligned} \delta &= \delta(x) & x &= \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} & x_1 &= x_1 \\ && && x_2 &= -x_1 + \epsilon(1-x_1^2)x_2 & \therefore \delta(x) &= \begin{bmatrix} x_2 \\ -x_1 + \epsilon(1-x_1^2)x_2 \end{bmatrix} \end{aligned}$$

$$\frac{\partial \delta}{\partial x} = \begin{bmatrix} \frac{\partial \delta_1}{\partial x_1} & \frac{\partial \delta_1}{\partial x_2} \\ \frac{\partial \delta_2}{\partial x_1} & \frac{\partial \delta_2}{\partial x_2} \end{bmatrix} \therefore \delta(x) = \begin{bmatrix} \delta_1 \\ \delta_2 \end{bmatrix} \quad x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \therefore$$

$$\frac{\partial \delta}{\partial x} = \begin{bmatrix} 0 & 1 \\ -1-2x_1x_2 & -\epsilon(1-x_1^2) \end{bmatrix} \quad \frac{\partial \delta}{\partial x} \text{ is continuous on } D_r \therefore$$

$\delta(\cdot)$ is locally Lipschitz on D_r $\forall r > 0$

k_2 $\left[\frac{\partial \delta}{\partial x} \right]$ is not globally bounded $\therefore \delta(\cdot)$ is not globally Lipschitz

$\checkmark 6 / x^2 + |x|$ $|x|$ is not continuously differentiable at $x=0$ but is globally Lipschitz $\therefore ||x|-|y|| \leq |x-y| \forall x, y \in \mathbb{R}$. term x^2 is contantly differentiable, but its partial w.r.t. x is not globally bounded.

$\therefore \delta = x^2 + |x|$ is not contantly differentiable at $x=0$. it is contantly differentiable on a domain that does not include $x=0$. it is locally Lipschitz,
 \therefore continuous but not globally Lipschitz

$\delta = \sin(x) \operatorname{sgn}(x)$ \therefore is both x, y non negative:

$$|\delta(x) - \delta(y)| = |\sin(x) \operatorname{sgn}(x) - \sin(y) \operatorname{sgn}(y)| = |\sin(x) - \sin(y)| \leq |x-y|$$

$\therefore \sin(x)$ is bounded $\therefore -1 \leq \sin(x) \leq 1$

$$\text{if } x \geq 0, y \neq 0 \text{ then } |\delta(x) - \delta(y)| = |\sin(x) - \sin(y)| = \left| 2 \sin\left(\frac{x-y}{2}\right) \cos\left(\frac{x+y}{2}\right) \right|$$

$$10) \left| 2 \sin\left(\frac{x-y}{2}\right) \cos\left(\frac{x+y}{2}\right) \right| \leq |x-y|, \text{ obviously bounded} \therefore -1 \leq \cos x \leq 1$$

\therefore have $|\delta(x) - \delta(y)| \leq |x-y| \forall x, y \therefore \delta$ is locally Lipschitz

and continuous. it is not contantly differentiable at $x=0$ \therefore two limits

exist for $x \rightarrow 0^+$ & $x \rightarrow 0^-$

$\dot{x} = -x + \alpha \sin x$: Continuously differentiable, locally Lipschitz

$\frac{d\dot{x}}{dx} = -1 + \alpha \cos x$ $\left| \frac{d\dot{x}}{dx} \right|$ bounded globally \therefore globally Lipschitz

$\left| \frac{d\dot{x}}{dx} \right|$ bounded globally \therefore globally Lipschitz $\left(\frac{d\dot{x}}{dx} \right)$ bounded globally \therefore

globally Lipschitz $\left| \frac{d\dot{x}}{dx} \right|$ bounded globally \therefore globally Lipschitz

$\left| \frac{d\dot{x}}{dx} \right|$ bounded globally \therefore globally Lipschitz $\left| \frac{d\dot{x}}{dx} \right|$ bounded globally \therefore

globally Lipschitz $\left| \frac{d\dot{x}}{dx} \right|$ bounded globally \therefore globally Lipschitz

Week 6 Sheet $A = \begin{bmatrix} 1 & 2 \\ 2 & 6 \end{bmatrix} \therefore \text{eig}(A) = 0.2984, 6.7016$ 8

$A^T = \begin{bmatrix} 1 & 2 \\ 2 & 6 \end{bmatrix}^T = \begin{bmatrix} 1 & 2 \\ 2 & 6 \end{bmatrix} = A^T \therefore \lambda_i(A) > 0 \forall i \& A = A^T \therefore A$ is positive

$P = \begin{bmatrix} 4 & 1 & 2 \\ 2 & 3 & -1 \\ -1 & 2 & 2 \end{bmatrix} \therefore \text{eig}(P) = 0.1049, 3.6027, 5.29 \& P^T = \begin{bmatrix} 2 & 1 & 2 \\ 1 & 3 & -1 \\ 2 & -1 & 2 \end{bmatrix} = P$

(symmetric $\therefore \lambda_i(P) > 0 \forall i \& P = P^T \therefore P$ is positive)

2.1/ $V(x_1, x_2) = (x_1 + x_2)^2 \therefore V(0) = V(0, 0) = 0 \& V(x_1, x_2) > 0 \forall$

$\forall x_1, x_2 \in \mathbb{R} \setminus \{0\}$ \therefore globally positive

2.2/ $V(x_1, x_2) = x_1^2 + 2x_1x_2 + 3x_2^2 = (x_1 + 2x_2)^2 + x_2^2 \geq x_1^2 + 2x_1x_2 + x_2^2 + 2x_2^2 =$

$(x_1 + x_2)^2 + 2x_2^2 > 0 \forall x_1, x_2 \in \mathbb{R} \setminus \{0\}$ and $\therefore V(0) = 0$ 8

$V(x_1, x_2) > 0 \forall x_1, x_2 \in \mathbb{R} \setminus \{0\}$ \therefore globally positive

2.3/ $V(x_1, x_2) = (x_1 - x_2)^2 \therefore V(0) = 0$ 8 is $x_1 = x_2$: ~~$V(x) = 0$~~ $\therefore V(x) = 0$

$\therefore V(x_1, x_2)$ is not positive \therefore

$V(x_1, x_2) \geq 0 \therefore V(x_1, x_2)$ is \therefore positive semi-definite

$\therefore V(x_1, x_2) \geq 0 \forall (x_1, x_2) \in \mathbb{R}^2$

3/ $V(x_1, x_2, x_3) = \frac{1}{2}(x_1^2 + x_2^2 + x_3^2) \therefore V(0) = 0 \& V(x) > 0 \text{ for } x \neq 0$

$\therefore V$ is positive $\therefore V$ is sum of square terms $\& V(0, 0, 0) = 0$ \Rightarrow

provided $x_i > 0$, at $x_1 = x_2 = x_3 = 0 \therefore$ taking deriv wrt time, or

along trajectory of dynamics: $\dot{V} = \sum_{i=1}^3 \alpha_i x_i \dot{x}_i \therefore$ Substituting \dot{x}_i from

Week 6 Sheet / $x \in \mathbb{R}^3$: $V = \sum_{i=1}^3 x_i x_i$

$$x_1 x_1 (I_{22} x_2 x_2) + x_2 x_2 (I_{33} x_3 x_3) + x_3 x_3 (I_{11} x_1 x_1) \geq V$$

$(x_1 I_{22} + x_2 I_{33} + x_3 I_{11})(x_1 x_2 x_3)$ sign definite term in \mathbb{R}^3

choice $x_1 I_{22} + x_2 I_{33} + x_3 I_{11} = 0$ $\therefore I_{11} < 0$ - it is always possible to pick a set of x_1, x_2, x_3 so V is possi defi (required $x_i > 0$)

\therefore V is nega semi defi, i.e.

$V \leq 0$ for x_1, x_2, x_3 for 2 choice of I_{ij} st $x_i I_{ii} + x_j I_{jj} + x_k I_{kk} = 0$

i.e Lyapunov stable: i.e $x_2 I_{33}$ dominates & equal to $-(x_1 I_{22} + x_3 I_{11})$

$$4\dot{\theta}^2 + \dot{\theta} + \frac{3}{4}\sin\theta = 0 \quad \text{D}, \quad V(\theta, \dot{\theta}) = \frac{1}{2}\dot{\theta}^2 + \frac{3}{4}(1-\cos\theta)$$

$V(\theta, \dot{\theta}) = 0$ only at $\theta = \dot{\theta} = 0$ in a domain $D = \{(\theta, \dot{\theta}) \mid |\theta| < \pi\}$

$V(\theta, \dot{\theta}) > 0 \quad \forall (\theta, \dot{\theta}) \in D \setminus \{(0, 0)\}$: $\dot{\theta}^2$ is the square term &

$-1 \leq \cos\theta \leq 1$, $\cos\theta$ is bounded term $\therefore V(\theta, \dot{\theta})$ is possi defi in D

taking time deriv w.r.t. θ : $\dot{V}(\theta, \dot{\theta}) = \dot{\theta}\ddot{\theta} + \frac{3}{2}\dot{\theta}\sin\theta$ summing from ① for $\dot{\theta}$:

$$\dot{\theta}^2 = -\dot{\theta} - \frac{3}{2}\sin\theta \quad \therefore \quad V(\theta, \dot{\theta}) = \dot{\theta}\ddot{\theta} + \frac{3}{2}\dot{\theta}\sin\theta =$$

$$\dot{\theta}(-\dot{\theta} - \frac{3}{2}\sin\theta) + \frac{3}{2}\dot{\theta}\sin\theta = -\dot{\theta}^2 - \frac{3}{2}\dot{\theta}\sin\theta + \frac{3}{2}\dot{\theta}\sin\theta = -\dot{\theta}^2 \leq 0,$$

$V(\theta, \dot{\theta})$ cant be claimed as nega defi for syst:

$V(\theta, \dot{\theta}) = 0$ for $\dot{\theta} = 0$ no matter what θ values are.

$\therefore V(\theta, \dot{\theta})$ is nega semi defi \therefore only conclusion is that 2

origin is Lyapunov stable for $V(\theta, \dot{\theta}) = \frac{1}{2}\dot{\theta}^2 + \frac{3}{4}(1-\cos\theta)$

\therefore Lyapunov conditions are only sufficiency conditions.

$$4.2 / \dot{\theta} + \ddot{\theta} + \frac{3}{4}\sin(\theta) = 0, \quad V(\theta, \dot{\theta}) = \frac{1}{2}\dot{\theta}^2 + \frac{1}{2}(\dot{\theta} + \ddot{\theta})^2 + \frac{3}{4}(1-\cos\theta)$$

$\therefore -1 \leq \cos\theta \leq 1 \quad \& \quad 1 - \cos\theta \leq 2 \quad \therefore \cos\theta$ is bounded by 1 & $|\cos\theta| \leq 2$

$V(\theta, \dot{\theta})$ is possi defi: each term of sum is possi & $V(\theta, \dot{\theta}) = 0$ only at $\theta = \dot{\theta} = 0$ in D . $V(\theta, \dot{\theta}) > 0 \quad \forall (\theta, \dot{\theta}) \in D \setminus \{(0, 0)\}$

V is possi defi in domain D i.e. taking deriv of V along 2 trajectory of D yields: $\dot{V}(\theta, \dot{\theta}) = \dot{\theta}\ddot{\theta} + (\dot{\theta} + \ddot{\theta})(\dot{\theta} + \ddot{\theta}) + \frac{3}{2}\dot{\theta}\sin\theta$ summing from ①:

$$= \dot{\theta}(-\dot{\theta} - \frac{3}{2}\sin\theta) + (\dot{\theta} + \ddot{\theta})(\dot{\theta} + \ddot{\theta} - \frac{3}{2}\sin\theta) + \frac{3}{2}\dot{\theta}\sin\theta =$$

$$- \dot{\theta}^2 - \frac{3}{2}\dot{\theta}\sin\theta + (\dot{\theta} + \ddot{\theta})(-\frac{3}{2}\sin\theta) + \frac{3}{2}\dot{\theta}\sin\theta =$$

$$- \dot{\theta}^2 - \frac{3}{2}\dot{\theta}\sin\theta - \frac{3}{2}\dot{\theta}\sin\theta - \frac{3}{2}\dot{\theta}\sin\theta + \frac{3}{2}\dot{\theta}\sin\theta = - \dot{\theta}^2 - \frac{3}{2}\dot{\theta}\sin\theta \quad \therefore \text{note}$$

$+\dot{\theta}^2$ is +ve (square term) $\therefore -\dot{\theta}^2$ is -ve,

$\theta \sin \theta$ is +ve term no matter what θ is \therefore product ~~of~~ & sign

8 odd terms $\therefore \sin \theta + \text{ve} - \text{ve}$

$$\dot{V}(\theta, \dot{\theta}) = -\dot{\theta}^2 - \frac{9}{2} \theta \sin \theta < 0 \quad \forall (\theta, \dot{\theta}) \in D \setminus \{(0, 0)\} \quad \therefore V \text{ is pos defi}$$

In D 2 \dot{V} is neg defi in D: 2 origin of Z Syst is an (locally) asympt stable equili in $D = \{(\theta, \dot{\theta})^T \mid |\theta| < \pi\}$

$$\begin{cases} \dot{x}_1 = -x_1 + x_2^3 \\ \dot{x}_2 = -x_1 - x_2 \end{cases} \quad V(x_1, x_2) = \frac{1}{2}x_1^2 + \frac{1}{4}x_2^4 \quad \therefore \text{sum of +ve terms}$$

$V(x)$ is pos defi $\& V(x) \rightarrow \infty$ as $\|x\| \rightarrow \infty$ (ie a radially unbd

(unbounded func) taking deriv along 2 trajec of syst:

$$\dot{V}(x) = x_1 \dot{x}_1 + \frac{3}{2}x_2^3 \dot{x}_2 = x_1(-x_1 + x_2^3) + \frac{3}{2}x_2^3(-x_1 - x_2) = 2x_1^2 x_2^3$$

$$-2x_1^2 + x_2^3 - x_1 x_2^3 - x_2^4 = -2x_1^2 - x_2^4 < 0 \quad \forall (x_1, x_2) \neq (0, 0) \text{ or}$$

$\forall (x_1, x_2) \in \mathbb{R}^2 \setminus \{(0, 0)\} \quad \therefore \dot{V}(x_1, x_2) \text{ neg defi} \quad \therefore \text{zero sol to syst}$ is globally asym stable

$$\begin{cases} \dot{x}_1 = -4x_1 + x_1 x_2^2 \\ \dot{x}_2 = -x_2 + x_1 \end{cases} \quad V(x) = x_1^2 + x_2^2 \quad \therefore \begin{cases} \dot{x}_1 = -4x_1 + x_1 x_2^2 \\ \dot{x}_2 = -x_2 + x_1 \end{cases}$$

$$V = x_1^2 + x_2^2 \quad \therefore \text{sum of squares}, V(x_1, x_2) > 0 \quad \forall (x_1, x_2) \neq (0, 0), V(0, 0) = 0$$

$\therefore V > 0 \quad \forall x \in \mathbb{R}^2 \setminus \{(0, 0)\}$, ~~as~~ $V=0$ for $x=0$ $\therefore V$ is pos defi

taking along 2 trajec of Z Syst I: $\dot{V} = 2x_1 \dot{x}_1 + 2x_2 \dot{x}_2 = 2x_1(-4x_1 + x_1 x_2^2) + 2x_2(-x_2 + x_1)$ (by sub for \dot{x}_1 & \dot{x}_2)

$$= -8x_1^2 + 2x_1 x_2^2 + 2x_1 x_2 - 2x_2^2 =$$

$$= -7x_1^2 - x_1^2 + 2x_1^2 x_2^2 - x_2^2 - x_2^2 + 2x_1 x_2.$$

$$= -7x_1^2 - x_2^2 + 2x_1^2 x_2^2 - (x_1^2 - 2x_1 x_2 + x_2^2) = -7x_1^2 - x_2^2 + 2x_1^2 x_2^2 - (x_1 - x_2)^2 \leq$$

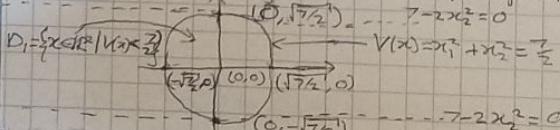
$$-7x_1^2 - x_2^2 + 2x_1^2 x_2^2 \quad (\because (x_1 - x_2)^2 \text{ is +ve term & it is Mins})$$

$$= -x_1^2 (-7 - 2x_2^2) - x_2^2 < 0 \quad \forall (-7 - 2x_2^2) \text{ is guaranteed to be +ve} \quad \therefore$$

$$-7 - 2x_2^2 > 0 \quad \therefore -7 > 2x_2^2 \quad \therefore \frac{7}{2} > x_2^2 \quad \therefore \text{is } \frac{7}{2} > x_1^2 + x_2^2 :$$

$\dot{V} < 0 \quad \forall x_1^2 + x_2^2 < \frac{7}{2} \quad \therefore \text{can define an invariant set } (x_1^2 + x_2^2) = V(x) < \frac{7}{2}$

$D_1 = \{x \in \mathbb{R}^2 \mid V(x) < \frac{7}{2}\}$ in which V is pos defi $\&$ \dot{V} is neg defi holds.)



Week 7 Sheet / $\nabla M(q)\dot{q} + C(q, \dot{q})\dot{q} + D\ddot{q} + g(q) = u$

time

$M(q)$ is a symmetric inertia mat. $M(q) = M(q)^T$, $M(q)$ is posidesi $\forall q \in \mathbb{R}^n$

(*) $C(q, \dot{q})\dot{q}$ is force $[C]$ has property $M^{-1}C$ is skew symmetric

desi mat $\forall q, \dot{q} \in \mathbb{R}^n$

is an recall skew symmetric mat $A \Rightarrow A^T = -A$ $a_{ij} = -a_{ji}$

$V(q, \dot{q}) = \frac{1}{2}\dot{q}^T M(q)\dot{q} + P(q)$ is posi desi $\therefore \dot{q}^T M(q)\dot{q}$ is quadratic \mathbb{R}^{n+1}

$\Delta M(q)$ posi desi $\therefore \dot{q}^T \dot{q}$ is posi desi & $M(q)$ is posi desi \therefore

$\dot{q}^T M(q)\dot{q}$ is posi desi $\therefore \Delta P(q)$ is posi desi since $P(q)$ is the \therefore

quad. V is posi desi \therefore taking \mathbb{R} deriv ∂V along \mathbb{R} dynamics:

$$\text{Syst: } \dot{V} = \frac{d}{dt} V = \frac{d}{dt} \left[\frac{1}{2} \dot{q}^T M(q) \dot{q} + P(q) \right] = \frac{1}{2} \frac{d}{dt} [\dot{q}^T M(q) \dot{q}] + \frac{d}{dt} [P(q)] =$$

$$\frac{1}{2} \dot{q}^T \frac{d}{dt} [M(q) \dot{q}] + \frac{1}{2} \dot{q}^T M(q) \dot{q} + \frac{\partial P(q)}{\partial q} \frac{d\dot{q}}{dt} =$$

$$\text{Syst } \frac{1}{2} \dot{q}^T \left(M(q) \frac{d}{dt} [\dot{q}] + \frac{d}{dt} [M(q)] \dot{q} \right) + \frac{1}{2} \dot{q}^T M(q) \dot{q} + \frac{\partial P(q)}{\partial q} \dot{q} =$$

$$\frac{1}{2} \dot{q}^T M(q) \dot{q} + \frac{1}{2} \dot{q}^T M(q) \dot{q} + \frac{1}{2} \dot{q}^T M(q) \dot{q} + \frac{\partial P(q)}{\partial q} \dot{q} = \dot{V}$$

note: $(AB)^T = B^T A^T$:

$$(\dot{q}^T (M(q) \dot{q}))^T = (M(q) \dot{q})^T (\dot{q}^T)^T = (M(q) \dot{q})^T \dot{q} = (\dot{q})^T (M(q))^T \dot{q} = \dot{q}^T M^T(q) \dot{q}$$

$$\therefore \dot{q}^T M^T(q) \dot{q} = \dot{q}^T M(q) \dot{q} \quad (\because M(q) = M^T(q))$$

$$\therefore g(q) = \left(\frac{\partial P(q)}{\partial q} \right)^T \therefore \frac{\partial P(q)}{\partial q} = g^T(q) \quad \therefore$$

$$\dot{V} = \frac{1}{2} \dot{q}^T M(q) \dot{q} + \frac{1}{2} \dot{q}^T M(q) \dot{q} + \frac{1}{2} \dot{q}^T M(q) \dot{q} + \frac{\partial P(q)}{\partial q} \dot{q} =$$

$$\frac{1}{2} \dot{q}^T M(q) \dot{q} + \frac{1}{2} \dot{q}^T M(q) \dot{q} + \frac{1}{2} \dot{q}^T M(q) \dot{q} + g^T(q) \dot{q} \quad \therefore$$

$$(\dot{q}^T M(q) \dot{q})^T = \dot{q}^T M(q) \dot{q} \quad \therefore \dot{q}^T M(q) \dot{q} = (\dot{q}^T M(q) \dot{q})^T \quad \therefore$$

$$\dot{V} = \frac{1}{2} \dot{q}^T M(q) \dot{q} + \frac{1}{2} \dot{q}^T M(q) \dot{q} + \frac{1}{2} \dot{q}^T M(q) \dot{q} + \frac{\partial P(q)}{\partial q} \dot{q} =$$

$$\frac{1}{2} \dot{q}^T M(q) \dot{q} + \frac{1}{2} \dot{q}^T M(q) \dot{q} + \frac{1}{2} (\dot{q}^T M(q) \dot{q})^T + \frac{\partial P(q)}{\partial q} \dot{q} =$$

$$\frac{1}{2} [\dot{q}^T M(q) \dot{q} + (\dot{q}^T M(q) \dot{q})^T] + \frac{1}{2} \dot{q}^T M(q) \dot{q} + \frac{\partial P(q)}{\partial q} \dot{q} = g^T(q) \dot{q} =$$

$\nabla(x) < \frac{1}{2}$:

$$\dot{q}^T M(q) \dot{q} + \frac{1}{2} \dot{q}^T M(q) \dot{q} + g^T(q) \dot{q} = \dot{V} \quad \therefore$$

$$M(q) \dot{q} = -C(q, \dot{q}) \dot{q} - D\ddot{q} - g(q) \quad \therefore$$

$$M(q) \dot{q} + C(q, \dot{q}) \dot{q} + D\ddot{q} + g(q) = 0 \quad \therefore u = 0 \text{ when } V(q, \dot{q}) = \frac{1}{2} \dot{q}^T M(q) \dot{q} + P(q).$$

$$\begin{aligned}\dot{V} &= \dot{q}^T [M(q)\dot{q}] + \frac{1}{2} \dot{q}^T M(q) \dot{q} + g^T(q) \dot{q} = \\ \dot{q}^T [-C(q, \dot{q})\dot{q} - D\dot{q} - g(q)] + \frac{1}{2} \dot{q}^T M(q) \dot{q} + g^T(q) \dot{q} = \\ -\dot{q}^T C(q, \dot{q})\dot{q} - \dot{q}^T D\dot{q} - \dot{q}^T g(q) + \frac{1}{2} \dot{q}^T (\dot{M}(q)) \dot{q} + g^T(q) \dot{q} = \\ \frac{1}{2} \dot{q}^T [\dot{M}(q)] \dot{q} + \frac{1}{2} \dot{q}^T [2C(q, \dot{q})] \dot{q} - \dot{q}^T D\dot{q} - \dot{q}^T g(q) + g^T(q) \dot{q} = \\ \frac{1}{2} \dot{q}^T [\dot{M}(q) - 2C] - \dot{q}^T D\dot{q} - \dot{q}^T g(q) + g^T(q) \dot{q} = \dot{V}\end{aligned}$$

i.e. at the origin: ($q=0, \dot{q}=0$):

$\dot{M} - 2C$ is skew symmetric $\therefore \dot{M}(q) - 2C = 0$ at for Skew Symmetry at everywhere including origin $\forall q, \dot{q} \in \mathbb{R}^m$ \therefore

$$\dot{V} = \frac{1}{2} \dot{q}^T [0] - \dot{q}^T D\dot{q} - \dot{q}^T g(q) + g^T(q) \dot{q} = \dot{V}$$

$$-\dot{q}^T D\dot{q} - \dot{q}^T g(q) + g^T(q) \dot{q} = \dot{V}$$

$\dot{q}^T g(q), g^T(q) \dot{q}$ are identical $\{[\dot{q}^T g(q)]^T = g^T(q) \dot{q}^T \text{ & } g(q=0)=g(0)=0\}$
 $\therefore -\dot{q}^T g(q) + g^T(q) \dot{q} = 0 \therefore$

$\dot{V} = -\dot{q}^T D\dot{q} \therefore \dot{q}^T D\dot{q}$ is in quadratic form & D is pos. semi def.

$\therefore +\dot{q}^T D\dot{q}$ is pos. semi def. $\therefore -\dot{q}^T D\dot{q}$ is neg. semi def. \therefore

$\dot{V} \leq 0 \therefore V$ is pos. def. & \dot{V} is neg. semi def. \therefore Origin is Lyapunov stable

2/ $\dot{V} = \int_{x_1}^{x_2} h_1(y) dy + \frac{1}{2} x_2^2$ \therefore given $h_1(0)=0, y h_1(y) > 0 \forall y \neq 0 \text{ & } y \in (-a, a)$
 $\therefore \int_{x_1}^{x_2} h_1(y) dy > 0 \text{ for } y < 0 \therefore h_1(y) < 0, \text{ for } y > 0 : y h_1(y) > 0 \therefore h_1(y) < 0$
 $\therefore V(x) = \int_{x_1}^{x_2} h_1(y) dy + \frac{1}{2} x_2^2 \therefore V$ is cont. diff. & pos. def. at least locally here $\because y \in (-a, a) \therefore$ for $0 < y < a : h_1(y) > 0 \therefore y > 0 \therefore h_1(y) > 0$
 $\therefore \int_{x_1}^{x_2} h_1(y) dy > 0 \text{ & for } x_2 > 0 : \frac{1}{2} x_2^2 > 0 \therefore V$ is pos. def. \therefore

$$\begin{aligned}\text{Taking time deriv: } \dot{V}(x) &= \frac{d}{dt} \left[\int_{x_1}^{x_2} h_1(y) dy \right] + \frac{d}{dt} \left(\frac{1}{2} x_2^2 \right) = \\ \frac{\partial}{\partial x_1} \left[\int_{x_1}^{x_2} h_1(y) dy \right] \frac{\partial x_1}{\partial t} + \frac{1}{2} \frac{\partial}{\partial x_2} \left[x_2^2 \right] \frac{\partial x_2}{\partial t} - \frac{\partial}{\partial t} \left[h_1(x_1) \frac{\partial x_1}{\partial x_1} - h_1(0) \frac{\partial(0)}{\partial x_1} \right] + \frac{1}{2} x_2 \dot{x}_2 = \\ [h_1(x_1) - 0] \dot{x}_1 + x_2 \dot{x}_2 = h_1(x_1) \dot{x}_1 + x_2 \dot{x}_2 = \dot{V}(x) =\end{aligned}$$

$$h_1(x_1) x_2 + x_2 (-h_1(x_1) - h_2(x_2)) = h_1(x_1) x_2 - x_2 h_1(x_1) - h_2(x_2) x_2 = h_1(x_1) x_2 - h_1(x_1) x_2 - h_2(x_2) x_2$$

$$= -x_2 h_2(x_2) = \dot{V}(x) \therefore \because y h_1(y) > 0 \therefore y h_2(y) > 0 \therefore x_2 h_2(x_2) > 0 \therefore$$

$-x_2 h_2(x_2) < 0$ $\therefore -x_2 h_2(x_2)$ is neg. semi def. \therefore

$\dot{V}(x) \leq 0$ neg. semi def. (\therefore indep. of x_1) \therefore

\Week 7 Sheet / by Lyapunov argument: V is posⁱ desⁱ & V is neg^a semi desⁱ \therefore origin is Lyapunov stable \therefore

(*) only Lyapunov stable \therefore indep of x \therefore only neg semi desⁱ $\because V(x)=0$ everywhere \rightarrow $x_1 \rightarrow x_2$

Consider a set consists of (x_1, x_2) s.t. $x_2 = 0$ & x_1 be any value in \mathbb{R} $\therefore V(x)=0$

everywhere $V(x)=0$ \therefore Suppose trajectory $X(t) \in \mathbb{R}^2$ starts in set

$$S = \{x \in D \mid V(x)=0\} \text{ then } x_2=0 \text{ from } \Sigma.$$

$\dot{x}_1=0 \therefore x_1$ is a constant but if this const is non zero, then

try out From Σ dynamics, $\dot{x}_2 = -h_1(x_1) - h_2(x_2) \therefore \dot{x}_2 \neq 0$, in turn x_2 will be a non zero val. \therefore trivial sol has to be origin \therefore by LaSalle's invariance principle argument, can conclude \exists origin Locally asymp stable.

$$\text{B/ } V(x) = \frac{1}{2}x_1^2 + \frac{1}{2}y^2 \quad b>a, y>0 \quad x_1 \neq y \quad \& \quad x_2=k(t) \quad \therefore \\ V=0=g(0)=g(y)$$

V is contly dissable & posⁱ desⁱ (sum of square terms & given facts)

$\therefore x_1^2 \geq y^2 \geq 0, (x_2-b)^2 \geq 0 \therefore V \geq 0, V(0,b)=0 \therefore V$ is posⁱ desⁱ \therefore

taking deriva of $V(x)$ along trajectories as $\dot{y}=ay+bx$ $\dot{y}=(a-k(t))y$

$$\begin{cases} \dot{y} = ay + bx \\ \dot{y} = (a - k(t))y \end{cases} \quad \begin{cases} a = -ky \\ b = ry^2 \end{cases} \quad \begin{cases} k = \alpha y \\ r = \beta y^2 \end{cases} \quad \therefore$$

$$\text{given is } \dot{V}(y, k(t)) = \frac{d}{dt} \left[\frac{1}{2}y^2 \right] + \frac{d}{dt} \left[\frac{1}{2}(k(t)-b)^2 \right] = y \dot{y} + \frac{1}{2}(k(t)-b) \dot{k}(t) =$$

$$(y)(E(k(t)-a)y) + \frac{1}{2}(k(t)-b)2y^2 = -(y^2)[k(t)-a] + (k(t)-b)y^2 =$$

$$y^2 [k(t)-b-k(t)+a]y^2 = -(b-a)y^2 \leq 0 \quad \therefore b>a \therefore b-a>0, y^2 \geq 0 \therefore$$

$-(b-a)y^2 \leq 0$ from neg^a semi desⁱ perspective \therefore

Consider $S = \{(y, k) \in D \mid V(y, k)=0\}$ when $y=0$, no matter what k .

to show \exists compact set, link this with \exists \exists \exists given $V(x) \leq 0$

V is contly dissable & posⁱ desⁱ. $S_{\text{ac}} = \{(y, k) \in D \mid V(y, k) \leq c\}$ this

S_{ac} is compact & the invariant set $\therefore V(x)$ is posⁱ desⁱ & $\dot{V}(x) \leq 0$ in

S_{ac} . $\therefore S = \{(y, k) \in S_{\text{ac}} \mid y=0\}$ is invariant set. by LaSalle's thm

every trajec in S_{ac} approaches S as $t \rightarrow \infty$ as $y(t) \rightarrow 0$ as $t \rightarrow \infty \therefore$

Locally asymp stable

$$\text{C/ } \begin{cases} \dot{x}_1 = x_1(x_1^2 - x_1^2 - x_2^2) + x_2(x_1^2 + x_2^2 + k^2) \\ \dot{x}_2 = x_1(x_1^2 + x_2^2 + k^2) + x_2(k^2 - x_1^2 - x_2^2) \end{cases} \quad k=0 \& k \neq 0: \quad V(x) = x_1^2 + x_2^2 \therefore$$

$V(x)$ is posⁱ desⁱ $\therefore x_1^2 \geq 0, x_2^2 \geq 0 \therefore V(x) \geq 0, V(0)=0 \therefore$

now taking the deriv of $V(x)$ along \mathbb{Z} trajcs \Rightarrow dynamics, i.e
in time deriv of $V(x)$: $\dot{V}(x) = 2x_1\dot{x}_1 + 2x_2\dot{x}_2 =$

$$\begin{aligned} & \dot{V}(x) = 2x_1[x_1(k^2 - x_1^2 - x_2^2) + x_2(x_1^2 + x_2^2 + k^2)] + 2x_2[-x_1(x_1^2 + x_2^2 + k^2) + x_2(k^2 - x_1^2 - x_2^2)] = \\ & 2x_1[x_1(k^2 - x_1^2 - x_2^2)] + 2x_2[x_2(k^2 - x_1^2 - x_2^2)] = \\ & 2x_1^2(k^2 - x_1^2 - x_2^2) + 2x_2^2(k^2 - x_1^2 - x_2^2) = 2(x_1^2 + x_2^2)(k^2 - x_1^2 - x_2^2) = \\ & \frac{1}{2}2(x_1^2 + x_2^2)(k^2 - (x_1^2 + x_2^2)) = 2V(k^2 - V) = \dot{V} \end{aligned}$$

two cases:

(i) if $k=0$, then $\dot{V} = 2V(0^2 - V) = -2V^2 \leq 0 \therefore V^2 \geq 0 \therefore -2V^2 \leq 0 \therefore$

$\dot{V} \leq 0 \therefore \dot{V} < 0 \forall x \in \mathbb{R}^2 \setminus \{0\}, V(0)=0$ note: $V(x) \rightarrow \infty$ as $\|x\| \rightarrow \infty$

(ii) if $k \neq 0$ then $\dot{V} < 0$ for $V > k^2$ or $\|x\| > |k|$ (i.e. $D = \{x \in \mathbb{R}^2 \mid V(x) > k^2\}$)

notice that for $k \neq 0$, have $\dot{V} < 0$ within neighbourhood of $x=0$ of radius $|k|$ and cannot find an invariant set D s.t. $x=0$ contained in D , $\therefore \dot{V} < 0$ on D : here \mathbb{Z} requirement is $\|x\| > |k|$, not $\|x\| < |k|$ which would have given us a region $V(x) \leq k^2$

if $k=0$, the origin is Globally asymptotically stable from argument in (i).
 V is positive, \dot{V} is negative.

5a/ $\begin{cases} \dot{x}_1 = -x_1 + x_2^2 \\ \dot{x}_2 = -x_2 \end{cases} \therefore V(x) = x_1^2 + x_2^2$ & equilib $x=(0,0) \therefore V(x) > 0 \forall x \in \mathbb{R}^2 \setminus \{0\}$.
 $V(0)=0 \therefore V$ is positive

$$\dot{V}(x) = 2x_1\dot{x}_1 + 2x_2\dot{x}_2 = 2x_1(-x_1 + x_2^2) + 2x_2(-x_2) = -2x_1^2 + 2x_1x_2^2 - 2x_2^2 = -2x_1^2 - 2x_2^2(1-x_1) \therefore \dot{V}(x) < 0 \text{ if } x_1 \leq 1, \therefore x=(0,0) \text{ is an asymptotically stable equili in a domain } D = \{x \in \mathbb{R}^2 \mid V(x) < 1\} \therefore x_1 \leq 1, \therefore x_1^2 \leq 1 \therefore$$

$$\{x_1^2 + x_2^2 < 1\} \subseteq \{x_1^2 \leq 1\}$$

5b/ $\begin{cases} \dot{x}_1 = (x_1 - x_2)(x_1^2 + x_2^2 - 1) \\ \dot{x}_2 = (x_1 + x_2)(x_1^2 + x_2^2 - 1) \end{cases} \therefore V(x) = x_1^2 + x_2^2 \therefore V(x) \text{ is positive} \therefore$

$$\dot{V}(x) = 2x_1\dot{x}_1 + 2x_2\dot{x}_2 =$$

$$2x_1[(x_1 - x_2)(x_1^2 + x_2^2 - 1)] + 2x_2[(x_1 + x_2)(x_1^2 + x_2^2 - 1)] =$$

$$2x_1[x_1(x_1^2 + x_2^2 - 1)] + 2x_2[x_2(x_1^2 + x_2^2 - 1)] = 2x_1^2(x_1^2 + x_2^2 - 1) + 2x_2^2(x_1^2 + x_2^2 - 1) =$$

$$2(x_1^2 + x_2^2)(x_1^2 + x_2^2 - 1) = 2V(V-1) < 0 \text{ if } V < 1 \therefore \text{origin is an asymptotically stable equili in } D = \{x \in \mathbb{R}^2 \mid V(x) < 1\}$$

$$\dot{V}(x) \leq 0 \text{ nega semi desi (i.e. indep. of } \therefore$$

, i.e. Week 7 Sheet $\begin{cases} \dot{x}_1 = -x_1 + x_1^2 x_2 \\ \dot{x}_2 = x_1 - x_2 \end{cases}$ $V(x) = x_1^2 + x_2^2 \therefore x_1^2 \geq 0, x_2^2 \geq 0 \therefore$
 $V(x) \geq 0, V(0) = 0, V(x) > 0 \forall x \in \mathbb{R}^2 \setminus \{0\} \therefore V(x)$ is positive.

• taking the derivative of V along the trajectories of the system : $\dot{V}(x) = 2x_1 \dot{x}_1 + 2x_2 \dot{x}_2 = 2x_1(-x_1 + x_2 x_1^2) + 2x_2(x_1 - x_2) =$

$$2[-(x_1^2 + x_2^2) + x_1 x_2(1+x_1^2)] \text{, if } \|x\| < 1 \text{ then: } (x_1^2 + x_2^2) > x_1 x_2(1+x_1^2)$$

$$\therefore 1+x_1^2 > 0 \text{ if } x_1 > 0, x_2 < 0; x_1 x_2 < 0 \therefore x_1 x_2(1+x_1^2) < 0 \therefore$$

$$(x_1^2 + x_2^2) > x_1 x_2(1+x_1^2) = x_1 x_2 + x_1 x_2^2 x_2 = x_1 x_2 + x_1^3 x_2 \therefore x_1 < 1, x_2 < 1 \therefore$$

$$\therefore x_1^3 < x_1^2 \text{, if } x_1^3 x_2 < x_1^2 x_2 < x_1^2 \therefore x_1^2 + x_2^2 > x_1 x_2(1+x_1^2) \therefore$$

$$\|x\| \rightarrow \infty \quad \dot{V}(x) < 0 \text{ if } \|x\| < 1 \therefore \dot{V} = 2[-(x_1^2 + x_2^2) + x_1 x_2(1+x_1^2)] \therefore$$

$\therefore -(x_1^2 + x_2^2) + x_1 x_2(1+x_1^2) < 0 \therefore \dot{V} < 0 \therefore x \in \mathbb{R}^2 \setminus \{0, 0\}$ is locally asymptotically stable.

$$D = \{x \in \mathbb{R}^2 \mid V(x) \leq 1\}$$

in D , $\begin{cases} \dot{x}_1 = -x_1 - x_2 \\ \dot{x}_2 = x_1 - x_2^3 \end{cases}$ $V(x) = x_1^2 + x_2^2 \therefore V$ is positive. \therefore taking deriv:

$$\dot{V}(x) = 2x_1 \dot{x}_1 + 2x_2 \dot{x}_2 = 2x_1[-x_1 - x_2] + 2x_2[x_1 - x_2^3] =$$

$$-2x_1^2 - 2x_1 x_2 + 2x_1 x_2 - 2x_2^4 = -2x_1^2 - 2x_2^4 = -2(x_1^2 + x_2^4) = -2(x_1^2 + (x_2^2)^2) \leq 0 \therefore$$

$\dot{V}(x) < 0 \text{ if } x \in \mathbb{R}^2 \setminus \{0\}$. $\dot{V}(0) = 0 \therefore$ origin is globally asymptotically stable.

• V is radially dissable & positive.

$\dot{V} \leq 0$ from a negative semi-definite perspective

{ Lasalle's theorem: V is radially dissable & positive &

$\dot{V} \leq 0 \therefore$ let $S = \{x \in D \mid \dot{V} = 0\}$ \therefore suppose no set S can identically stay in S other than trivial set $x(t) = 0 \therefore$ origin is ^{locally} asymptotically stable

• to make origin ~~asym~~ globally asymptotically stable: release the restriction on $y \therefore y \in \mathbb{R}$ & $V(x) \rightarrow \infty$ as $\|x\| \rightarrow \infty \therefore$ radially unbounded

• S_c is compact positive invariant set $\therefore V$ is positive &

$\dot{V} \leq 0$ in S_c : S_c is invariant set \therefore every trajec in S_c approaches S_c as $t \rightarrow \infty$ as $y(t) \rightarrow 0$ as $t \rightarrow \infty \therefore$ locally asymptotically stable.

• V positive \dot{V} negative semi-definite Lyapunov stable

V positive \dot{V} negative semi-definite \therefore Lyapunov stable

V positive \dot{V} semi-definite \therefore globally asymptotically stable

$V > 0 \quad \dot{V} \leq 0$ Lyapunov stable $V > 0 \quad \dot{V} < 0$ Lyapunov stable

$V > 0 \quad \dot{V} < 0$ globally asymptotically stable $V > 0 \quad \dot{V} \leq 0$ locally asymptotically stable

V positive \dot{V} negative \therefore globally asymptotically stable $V > 0 \quad \dot{V} \leq 0$ Lyapunov stable

Sheet week 8/11

$$\begin{aligned} \dot{x}_1 &= -x_2 \\ \dot{x}_2 &= x_1 + (x_1^2 - 1)x_2 \end{aligned} \quad \therefore \quad \dot{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -x_2 \\ x_1 + (x_1^2 - 1)x_2 \end{bmatrix}$$

$$\begin{bmatrix} -x_2 \\ x_1 + x_1^2 x_2 \end{bmatrix} = \begin{bmatrix} 0 & -1 \\ x_1 - 1 & x_2 + x_1^2 x_2 \end{bmatrix} = \begin{bmatrix} 0 & -1 \\ x_1 - 1 & x_2 + x_1^2 x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ x_1^2 x_2 \end{bmatrix} = \begin{bmatrix} 0 & -1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ x_1^2 x_2 \end{bmatrix} = A\dot{x} + g(x)$$

$$\therefore A = \begin{bmatrix} 0 & -1 \\ 1 & -1 \end{bmatrix} \quad g(x) = \begin{bmatrix} 0 \\ x_1^2 x_2 \end{bmatrix} \quad \therefore \det(A - \lambda I) = \begin{vmatrix} 0 - \lambda & -1 \\ 1 & -1 - \lambda \end{vmatrix} = \begin{vmatrix} -\lambda & -1 \\ 1 & -1 - \lambda \end{vmatrix} =$$

$$-\lambda(-1 - \lambda) - (1)(-1) = 0 = \lambda + \lambda^2 + 1 = \lambda^2 + \lambda + 1 = 0 \quad \therefore \lambda = \frac{-1 \pm \sqrt{1^2 - 4(1)(1)}}{2(1)} = \frac{-1 \pm \sqrt{-3}}{2} \quad \therefore$$

Mat A has eigenvalues $\frac{-1 \pm i\sqrt{3}}{2} = \frac{-1}{2} \pm i\sqrt{3}$ $\therefore \operatorname{Re}(\lambda_i(A)) < 0, i=1,2$ \therefore
 $[A]$ is Hurwitz \therefore Z linear part alone: $\dot{x} = A\dot{x}$ is asymptotically stable
 (origin) \therefore

by converse Lyapunov argument, $\exists P = P^T > 0 \quad \forall Q = Q^T > 0$ satisfying

Lyapunov eqn $A^T P + P A = -Q$ let Q be I $\therefore A^T P + P A = -I$ \therefore

choose $P = \begin{bmatrix} P_{11} & P_{12} \\ P_{21} & P_{22} \end{bmatrix} \quad \because P = P^T \quad \therefore$ we know $[A]$ \therefore solving $A \otimes P$ in Z

Lyapunov eqn: $P = \begin{bmatrix} 3/2 & -1/2 \\ -1/2 & 1 \end{bmatrix} \quad \left\{ \text{is } A^T = \begin{bmatrix} 0 & 1 \\ -1 & -1 \end{bmatrix} \quad \therefore A^T P + P A = -I = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix} \right\}$

$$\begin{bmatrix} 0 & 1 \\ -1 & -1 \end{bmatrix} \begin{bmatrix} P_{11} & P_{12} \\ P_{21} & P_{22} \end{bmatrix} + \begin{bmatrix} P_{11} & P_{12} \\ P_{21} & P_{22} \end{bmatrix} \begin{bmatrix} 0 & 1 \\ -1 & -1 \end{bmatrix} = \begin{bmatrix} P_{11} & P_{12} \\ P_{21} & P_{22} \end{bmatrix} + \begin{bmatrix} -P_{11} & P_{12} - P_{12} \\ -P_{21} & P_{22} - P_{12} \end{bmatrix} = \begin{bmatrix} 0 & P_{11} - P_{12} + P_{22} \\ -P_{11} - P_{12} - P_{21} & -2P_{22} \end{bmatrix} = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix} \quad \times$$

$$A = \begin{bmatrix} 0 & -1 \\ 1 & -1 \end{bmatrix} \quad A^T P + P A = -I \quad P = \begin{bmatrix} P_{11} & P_{12} \\ P_{21} & P_{22} \end{bmatrix} \quad P = P^T \quad \therefore P = \begin{bmatrix} 3/2 & -1/2 \\ -1/2 & 1 \end{bmatrix}, \quad \therefore$$

$$\lambda_{\min}(P) = 0.698 \quad (\text{min eigen}) \quad \therefore P = P^T > 0 :$$

Z Lyapunov for LT I Syst $\dot{x} = Ax$ is $V(x) = x^T P x$ where

$$x = [x_1 \ x_2]^T \quad \& \quad P = \begin{bmatrix} 1.5 & -0.5 \\ -0.5 & 1 \end{bmatrix} \quad \text{one can also write } V(x) = 1.5x_1^2 + x_2^2 - x_1 x_2$$

taking $V(x)$ along Z dynamics of system:

$$\dot{V}(x) = 3x_1 \dot{x}_1 + 2x_2 \dot{x}_2 - x_1 \dot{x}_2 - \dot{x}_1 x_2 =$$

$$(3x_1 - x_2)(-x_2) + (-x_1 + 2x_2)(x_1 + (x_1^2 - 1)x_2) =$$

$$-3x_1 x_2 + x_2^2 - x_1^2 - x_1^3 x_2 + x_1 x_2 + 2x_1 x_2 + 2x_2^2 x_1^2 - 2x_2^2 =$$

$$-x_1^2 - x_2^2 - x_1^3 x_2 + 2x_1^2 x_2^2 = -(x_1^2 + x_2^2) - x_1^2 x_2(x_1 - 2x_2) \leq$$

$$||x||^2 + ||x|| ||x_1 x_2|| |(x_1 - 2x_2)|$$

$$\text{here we know } |x_1| \leq ||x||, |x_1 x_2| \leq ||x||^2/2, |x_1 - 2x_2| \leq \sqrt{1^2 + 2^2} ||x|| \leq \sqrt{5} ||x||$$

\therefore making use of these:

$$\dot{V}(x) \leq -||x||^2 + ||x|| \frac{1}{2} ||x||^2 \sqrt{5} ||x|| \leq -||x||^2 + \frac{\sqrt{5}}{2} ||x||^4 \leq -||x||^2 \left(1 - \frac{\sqrt{5}}{2} \frac{||x||^2}{||x||^2} \right) \quad \text{+ve provided}$$

$$C \leq ||x||^2 < \frac{2}{\sqrt{5}} \quad \therefore \text{choose } \frac{2}{\sqrt{5}} = r^2 \quad \& \text{ following Z arguments as } \forall r > 0,$$

$\exists r > 0$ st $||g(x)|| < r ||x||, \forall ||x|| < r$: can write Z esti as Z region of

attraction stability as $D = \{x \in \mathbb{R}^2 | V(x) < C\}$ where $C = \lambda_{\min}(P) \cdot r^2 = \frac{(0.698)^2}{\sqrt{5}}$

Week 8 Sheet 2

$$\dot{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} x_1 \\ x_2 \\ -2x_1 - 3x_2 + x_2 |\sin x_2| \end{bmatrix} = Ax + g(x) = \begin{bmatrix} x_1 \\ x_2 \\ -2x_1 - 3x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ x_2 |\sin x_2| \end{bmatrix} =$$

$$\begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ x_2 |\sin x_2| \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \end{bmatrix} x_2 |\sin x_2| \text{ ie } \dot{x} = Ax + B\tilde{g}(x)$$

here $[A]$ is stable. by converse lyapunov argume

$$\{ A = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} \therefore \det(A - \lambda I) = \det \begin{bmatrix} -\lambda & 1 \\ -2 & -3-\lambda \end{bmatrix} = -\lambda(-3-\lambda) - 1(-2) = 3\lambda + \lambda^2 + 3 =$$

$$\lambda^2 + 3\lambda + 3 = 0 \therefore \lambda_i = \frac{-3 \pm \sqrt{9-4(3)(1)}}{2(1)} = \frac{-3}{2} \pm \frac{\sqrt{-3}}{2} = \frac{-3}{2} \pm \frac{\sqrt{3}i}{2} \therefore$$

$\operatorname{Re}(\lambda_i(A)) < 0 \quad i=1,2 \therefore A$ is stable. by converse lyapunov thm:

$\exists \alpha \operatorname{mat}[P] : P = P^T > 0$ & a choice of $Q = Q^T > 0$ satisfying

$$A^T P + PA = -Q \quad \text{let } Q = I \therefore \text{ solving } \exists \text{ Lyapunov eqn : } P = \begin{bmatrix} 5/4 & 1/4 \\ 1/4 & 1/4 \end{bmatrix}$$

$$\{ P = P^T \therefore P = \begin{bmatrix} \alpha & b \\ b & d \end{bmatrix} \therefore A = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} \therefore A^T = \begin{bmatrix} 0 & -2 \\ 1 & -3 \end{bmatrix} \therefore$$

$$\{ A^T P + PA = -I = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix} = \begin{bmatrix} 0 & -2 \\ 1 & -3 \end{bmatrix} \begin{bmatrix} \alpha & b \\ b & d \end{bmatrix} + \begin{bmatrix} \alpha & b \\ b & d \end{bmatrix} \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} = \begin{bmatrix} -2b & -2d \\ a-3b & b-3d \end{bmatrix} + \begin{bmatrix} -2b & -3b \\ a-3b & b-3d \end{bmatrix} =$$

$$\begin{bmatrix} -4b & -3b-2d \\ a-3b-2d & 2b-6d \end{bmatrix} = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix} \therefore -4b = -1 \therefore b = \frac{1}{4} \dots$$

$$-3b-2d = 0 \therefore -2d = 3(\frac{1}{4}) = \frac{3}{4} \therefore d = -\frac{3}{8}$$

$$2b-6d = -1 \therefore 6d = 2(\frac{1}{4}) + 1 = \frac{1}{2} + 1 = \frac{3}{2} \therefore d = \frac{3}{6} \cdot \frac{1}{2} = \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4}$$

$$\alpha = 3b+2d \therefore \alpha = \frac{5}{4} = 3(\frac{1}{4}) + 2(\frac{1}{4}) \therefore P = \begin{bmatrix} 5/4 & 1/4 \\ 1/4 & 1/4 \end{bmatrix} \}$$

write Z Lyapunov func $V(x) = x^T P x$ where $P = \begin{bmatrix} 5/4 & 1/4 \\ 1/4 & 1/4 \end{bmatrix}$ taking Z

deriv of V along Z trajcs of Z syst:

$$\begin{aligned} \dot{V}(x) &= x^T(PA + A^T P)x + 2x^T P g(x) = -x^T x + 2[x_1 \ x_2] \begin{bmatrix} P_{11} & P_{12} \\ P_{21} & P_{22} \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} x_2 |\sin x_2| \\ &= -x^T x + 2[x_1 \ x_2] \begin{bmatrix} P_{12} \\ P_{22} \end{bmatrix} x_2 |\sin x_2| = -x^T x + 2x_1 P_{12} x_2 |\sin x_2| + 2x_2 P_{22} x_2 |\sin x_2| = \dot{V}(x) \end{aligned}$$

$$\text{note } \frac{1}{4} = P_{22} = P_{12} > 0 \quad \& \quad | \sin x_2 | < 1 \quad \forall x_2 \quad \therefore | \sin x_2 | < 1 \quad \forall x_2 \quad \therefore$$

$$\dot{V}(x) = -x^T x + 2x_1 P_{12} x_2 |\sin x_2| + 2x_2 P_{22} x_2 |\sin x_2| \leq$$

$$-x_1^2 - x_2^2 + 2x_1 P_{12} x_2 |\sin x_2| + 2x_2 P_{22} x_2 |\sin x_2| \leq$$

$$-x_1^2 - (1-2P_{22})x_2^2 + 2x_1 P_{12} x_2 |\sin x_2|$$

con argue regd si is sign indeci

{Derive/can make use of Young's inequality: $2z^T Y \leq \frac{1}{\beta} \|z\|^2 + \beta \|Y\|^2 \quad \forall \beta > 0, \forall z, Y \in \mathbb{R}^m$ }

$$\dot{V}(x) = -x_1^2 - (1-2P_{22})x_2^2 + 2x_1 P_{12} x_2 |\sin x_2|$$

i. Making use of Young's inequality in 2nd term ie $2x_1 P_{12} x_2 |\sin x_2| \leq \frac{2}{\beta} P_{12}^2 x_1^2 + \beta x_2^2 |\sin x_2|^2$ $\leq \frac{1}{\beta} P_{12}^2 x_1^2 + \beta x_2^2$ \therefore using this:

$$\dot{V}(x) \leq -x_1^2 - (1-2P_{22})x_2^2 + \frac{1}{\beta} P_{12}^2 x_1^2 - \beta x_2^2 = -(1 - \frac{1}{\beta} P_{12}^2)x_1^2 - (1-2P_{22}-\beta)x_2^2$$

$\dot{V}(x)$ is negative definite: $1 - \frac{1}{\beta} P_{12}^2 < 0 \Leftrightarrow 1-2P_{22}-\beta > 0$

$$B > P_{12}^2 \quad \& \quad \beta < 1-2P_{22} \quad \therefore \quad \frac{1}{16} < \beta < \frac{1}{2}$$

unstable

$$\text{Q1} / \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 & -1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = A \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + g(x) \quad \text{nonlinear part}$$

2 mat $[A]$ has eigenvals $\frac{-1 \pm \sqrt{5}}{2} \quad \therefore \operatorname{Re}(\lambda_i(A)) < 0, i=1,2 \quad \therefore [A] \text{ is Hurwitz}$

\therefore 2 linear part alone $\dot{x} = Ax \quad \therefore \begin{bmatrix} \dot{x} \\ x \end{bmatrix} = \begin{bmatrix} 0 & -1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} x \\ x \end{bmatrix}$ is globally asymptotically stable (origin)

\therefore by converse Lyapunov argument, $\exists P = P^T > 0$ s.t. any $Q = Q^T > 0$ satisfies

$$\text{Lyapunov eqn } A^T P + P A = -Q$$

\therefore let $Q = I \quad \therefore A^T P + P A = -I \quad \therefore$

$$\text{choose } P = \begin{bmatrix} P_{11} & P_{12} \\ P_{21} & P_{22} \end{bmatrix} \quad \therefore P = P^T \quad \therefore$$

$$A = \begin{bmatrix} 0 & -1 \\ 1 & -1 \end{bmatrix} \quad \therefore \text{know } [A] \quad \therefore \text{Sub } A \in P \text{ in Lyapunov eqn.}$$

$$P = \begin{bmatrix} 3/2 & -1/2 \\ -1/2 & 1 \end{bmatrix} \quad \therefore$$

$$\lambda_{\min}(P) = 0.691 \quad (\text{min eigen}) \quad \therefore \approx P \approx$$

$$\therefore P = P^T > 0 \quad \therefore$$

Lyapunov func for LTI syst $\dot{x} = Ax$:

$$V(x) = x^T P x \text{ where } x = [x_1 \ x_2]^T \quad \& \quad P = \begin{bmatrix} 1.5 & -0.5 \\ -0.5 & 1 \end{bmatrix} \text{ one can write}$$

$$V(x) = 1.5x_1^2 + x_2^2 - x_1 x_2 \quad \left\{ \therefore V(x) = x^T P x = [x_1 \ x_2]^T \begin{bmatrix} 1.5 & -0.5 \\ -0.5 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \right.$$

$$\left. [1.5x_1 - 0.5x_2 \quad -0.5x_1 + x_2] \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 1.5x_1^2 + x_2^2 - x_1 x_2 \right\}$$

\therefore taking $\dot{V}(x)$ along L dynamics is original system:

$$\dot{V}(x) = 3x_1 \dot{x}_1 + 2x_2 \dot{x}_2 - x_1 \dot{x}_2 - x_2 \dot{x}_1 = (3x_1 - x_2) \dot{x}_1 + (-x_1 + 2x_2) \dot{x}_2 = \quad (\text{Sub in Syst})$$

$$(3x_1 - x_2)(-x_2) + (-x_1 + 2x_2)(x_1 + (x_1^2 - 1)x_2) =$$

$$-3x_1 x_2 + x_2^2 - x_1^2 + x_1^3 x_2 + x_1 x_2 + 2x_1 x_2 + 2x_2^2 x_1^2 - 2x_2^2 = -x_1^2 - x_2^2 - x_1^3 x_2 + 2x_1^2 x_2^2 =$$

$$- (x_1^2 + x_2^2) - (x_1^2 x_2) (x_1 - 2x_2) \quad (\because \|AB\| \leq \|A\| \|B\| \quad \therefore)$$

$$\leq -\|x\|^2 + (-x_1^2 x_2) (x_1 - 2x_2) \leq -\|x\|^2 + \| -x_1^2 x_2 \| \| x_1 - 2x_2 \| = -\|x\|^2 + \|x_1^2 x_2\| \| x_1 - 2x_2 \|$$

$$\leq -\|x\|^2 + \|x_1 (x_1 x_2)\| \|x_1 - 2x_2\| \leq -\|x\|^2 + \|x_1\| \|x_1 x_2\| \|x_1 - 2x_2\|$$

$$\text{here we know: } |x_1| \leq \|x\| \quad \& \quad |x_1 x_2| \leq \|x_1\| \|x_2\| / 2 \quad \therefore x_1^2 + x_2^2 + 2x_1 x_2 = \|x\|^2$$

$$\therefore x_1 x_2 + \frac{1}{2} (x_1^2 + x_2^2) = \|x\|^2 / 2 \quad \therefore |x_1 x_2| \leq \|x\|^2 / 2 \quad \therefore$$

$$|x_1 - 2x_2| \leq \sqrt{1^2 + (-2)^2} \|x\| = \sqrt{5} \|x\| = \sqrt{5} \|x\| \quad \therefore \text{making use of these}$$

Week 8 Sheet

$$\dot{V}(x) \leq -\|x\|^2 + \|x\| \frac{1}{2} \|x\|^2 \sqrt{\frac{5}{2}} \|x\| \leq -\|x\|^2 + \frac{\sqrt{5}}{2} \|x\|^3 \leq -\|x\|^2 \left(1 - \frac{\sqrt{5}}{2} \|x\|^2\right)$$

\therefore provided $0 \leq \|x\|^2 \leq \frac{2}{\sqrt{5}}$

$$\therefore \text{have accounted } g(x) \text{ in Z dynamics. } \therefore \dot{V}(x) \leq 0 \text{ is } 0 \leq \|x\|^2 \leq \frac{2}{\sqrt{5}}$$

\therefore asympt stable.

Choose $\frac{2}{\sqrt{5}} = r^2$ & following Z arguments as any $\gamma > 0$, $\exists r > 0$ st $\|g(x)\| < \gamma \|x\|$

$\forall \|x\| < r \therefore$

$$\text{let } \frac{2}{\sqrt{5}} = r^2 \quad \therefore 0 \leq \|x\|^2 \leq \frac{2}{\sqrt{5}} = r^2 \quad \therefore \text{for } \gamma > 0, \exists r > 0 \text{ st } \|g(x)\| < \gamma \|x\|,$$

$\forall \|x\| < r \therefore$

can write Z esti as Z region of attraction ~~of~~ stability as

$$D = \{x \in \mathbb{R}^2 | V(x) < C\} \text{ where } C = \lambda_{\min}(P) \cdot r^2 = \frac{0.691 \cdot 2}{\sqrt{5}} = 0.691 \cdot \frac{2}{\sqrt{5}}$$

$$\begin{aligned} \text{Z: } & \begin{cases} \dot{x}_1 = x_2 \\ \dot{x}_2 = -2x_1 - 3x_2 + x_2 |\sin x_2| \end{cases} \quad \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ x_2 |\sin x_2| \end{bmatrix} \\ & \text{where } A = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix}, x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}, g(x) = \begin{bmatrix} 0 \\ x_2 |\sin x_2| \end{bmatrix} \end{aligned}$$

$$A = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} \quad \therefore \det(A - \lambda I) = \det \begin{bmatrix} -\lambda & 1 \\ -2 & -3 - \lambda \end{bmatrix} = -\lambda(-3 - \lambda) - (-2) = 3\lambda + \lambda^2 + 3 = \lambda^2 + 3\lambda + 3 = 0 \quad \therefore$$

$$\lambda_i = \frac{-3 \pm \sqrt{9 - 4(3)(1)}}{2(1)} = -\frac{3}{2} \pm \frac{\sqrt{3}}{2} = -\frac{3}{2} \pm \frac{\sqrt{3}}{2} i \quad \therefore \operatorname{Re}(\lambda_i(A)) < 0, i=1,2.$$

$[A]$ is stable.

by converse Lyapunov argument, \exists a mat $[P]$ st $P = P^T > 0 \in \mathbb{R}^{2 \times 2}$

$$\text{choice of } Q = Q^T > 0 \text{ satisfying } A^T P + P A = -Q, P = \begin{bmatrix} p_{11} & p_{12} \\ p_{21} & p_{22} \end{bmatrix}, A = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} \quad \therefore$$

let $Q = I \quad \therefore$ solving Z lyapunov eqn $A^T P + P A = -I$:

$$P = \begin{bmatrix} \frac{5}{14} & \frac{1}{4} \\ \frac{1}{4} & \frac{1}{4} \end{bmatrix} \quad \therefore$$

write lyapunov func $V(x) = x^T P x$, where $P = \begin{bmatrix} \frac{5}{14} & \frac{1}{4} \\ \frac{1}{4} & \frac{1}{4} \end{bmatrix}$ \therefore taking Z deriv of

V along Z trajectories of Z dynamics:

$$\begin{aligned} \dot{V}(x) &= x^T P \dot{x} + \dot{x}^T P x = x^T P [Ax + g(x)] + [Ax + g(x)]^T P x \\ &= x^T (PA + A^T P) + 2x^T Pg(x) = -x^T Q x + 2x^T Pg(x) \quad \text{choose } \gamma < \frac{1}{2} \frac{\lambda_{\min}(Q)}{\|P\|} \end{aligned}$$

$$\therefore \dot{V}(x) = x^T (PA + A^T P) x + 2x^T Pg(x) \quad \therefore \dot{V}(x) \leq -x^T Q x + 2\|x\| \|Q\| \|g(x)\| \quad \therefore$$

$$\begin{aligned} \dot{V}(x) &= x^T (PA + A^T P) x + 2x^T Pg(x) = -x^T x + 2 \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} p_{11} & p_{12} \\ p_{21} & p_{22} \end{bmatrix} \begin{bmatrix} 0 \\ x_2 |\sin x_2| \end{bmatrix} = \\ &= -x^T x + 2 \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} p_{12} \\ p_{22} \end{bmatrix} x_2 |\sin x_2| = \end{aligned}$$

$$-x^T x + 2x_1 p_{12} x_2 |\sin x_2| + 2x_2 p_{22} x_2 |\sin x_2|$$

$$\text{note } \frac{1}{4} = p_{22} = p_{12} > 0 \quad \& \quad | \sin x_2 | < 1 \quad \forall x_2 \quad \therefore$$

$$\dot{V}(x) = -\tilde{x}^T x + 2x_1 P_{12} x_2 |\sin x_2| + 2x_2 P_{22} x_1 |\sin x_2| = -[x_1 \ x_2]^T \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + 2x_1 P_{12} x_2 |\sin x_2| + 2x_2 P_{22} x_1 |\sin x_2|$$

$$\leq -x_1^2 - x_2^2 + 2x_1 P_{12} x_2 |\sin x_2| + 2x_2 P_{22} x_1 (\because |\sin x_2| < 1 \forall x_2)$$

$$S \geq \tilde{x}^T [-x_1^2 - x_2^2 + 2x_1 P_{12} x_2 + 2x_2 P_{22} x_1] \leq -x_1^2 - (1-2P_{22})x_2^2 + 2x_1 P_{12} x_2 |\sin x_2|$$

negative desri sign indexi

∴ by young's inequality $2\tilde{x}^T Y \leq \frac{1}{\beta} \|z\|^2 + \beta \|Y\|^2 \quad \forall z, Y \in \mathbb{R}^n$

$$\therefore 2\tilde{x}^T [x_1 P_{12} x_2 |\sin x_2|] = 2\tilde{x}^T Y \leq \frac{1}{\beta} \|z\|^2 + \beta \|Y\|^2 = \frac{1}{\beta} \|P_{12} x_1\|^2 + \beta \|x_2\|^2$$

$$\frac{1}{\beta} \|P_{12} x_1\|^2 + \beta \|x_2\|^2 = \frac{1}{\beta} \|P_{12} x_1\|^2 + \beta P_{22} x_2^2 |\sin x_2|^2 = \frac{1}{\beta} P_{12}^2 x_1^2 + \beta x_2^2 |\sin x_2|^2$$

$$\leq \frac{1}{\beta} P_{12}^2 x_1^2 + \beta x_2^2, \beta > 0 \quad \therefore$$

$$\dot{V}(x) \leq -x_1^2 - (1-2P_{22})x_2^2 + 2x_1 P_{12} x_2 |\sin x_2| \leq$$

$$-x_1^2 - (1-2P_{22})x_2^2 + \frac{1}{\beta} P_{12}^2 x_1^2 + \beta x_2^2 |\sin x_2|^2 \leq$$

$$-x_1^2 - (1-2P_{22})x_2^2 + \frac{1}{\beta} P_{12}^2 x_1^2 + \beta x_2^2 = -(1 - \frac{1}{\beta} P_{12}^2)x_1^2 - (1-2P_{22}-\beta)x_2^2$$

$\dot{V}(x)$ is negar desri provided: $1 - \frac{1}{\beta} P_{12}^2$ is +ve & $1-2P_{22}-\beta$ is +ve

$$\therefore 1 - \frac{1}{\beta} P_{12}^2 > 0 \quad \therefore 1 > \frac{1}{\beta} P_{12}^2 \quad \therefore \beta > P_{12}^2 \quad \therefore \beta > \left(\frac{1}{4}\right)^2 \quad \therefore \beta > \frac{1}{16}$$

$$1-2P_{22}-\beta > 0 \quad \therefore 1-2P_{22} > \beta \quad \therefore 1-2P_{22} = 1-2\left(\frac{1}{4}\right) = 1-\frac{1}{2} = \frac{1}{2} > \beta \quad \therefore$$

$\frac{1}{16} < \beta < \frac{1}{2}$ will have asymp stability at origin.

$$\text{3/ } x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \quad \therefore x^T = [x_1 \ x_2] \quad \therefore \det(t - \lambda I) = \det \begin{bmatrix} t-\lambda & -1 \\ 1 & -t-\lambda \end{bmatrix} = \det \begin{bmatrix} -\lambda & -1 \\ 1 & -t-\lambda \end{bmatrix}$$

$$= (-\lambda)(-\lambda) - 1(-1) = \lambda + \lambda^2 + 1 = \lambda^2 + \lambda + 1 = 0 \quad \therefore \lambda = \frac{-1 \pm \sqrt{1^2 + 4(1)}}{2(1)} = \frac{-1 \pm \sqrt{5}}{2} = \frac{-1 \pm \sqrt{5}}{2} = -\frac{1}{2} \pm \frac{\sqrt{5}}{2}; \quad \therefore \operatorname{Re}(\lambda_i(A)) = -\frac{1}{2} < 0, i=1,2 \quad \therefore A \text{ is Hurwitz} \therefore$$

Linear system is globally orsymp stable.

by converse Lyapunov argument: $V(x) = x^T P x$ where $P = P^T > 0$

$$Q = P^T = 0 \text{ satisfying } PA + A^T P = -I \quad \therefore P = P^T \quad \therefore P = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

let $Q = I \quad \therefore A^T = \begin{bmatrix} 0 & 1 \\ 1 & -1 \end{bmatrix} \quad \therefore PA + A^T P = -I = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix} = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & -1 \end{bmatrix} + \begin{bmatrix} 0 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} =$

$$\begin{bmatrix} b-a-b \\ a-b-d \end{bmatrix} + \begin{bmatrix} b & d \\ -a-b & -b-d \end{bmatrix} = \begin{bmatrix} -2b & -a-b+d \\ -a-b-d & -2b-2d \end{bmatrix} = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix} \quad \therefore$$

$$-2b = -1 \quad \therefore b = -\frac{1}{2} \quad \therefore -2b - 2d = -1 = -2\left(-\frac{1}{2}\right) - 2d = 1 - 2d = -1 \quad \therefore 2 = 2d \quad \therefore d = 1 \quad \therefore$$

$$-a-b+d = 0 = -a + \frac{1}{2} + 1 = -a + \frac{3}{2} = 0 \quad \therefore a = \frac{3}{2} \quad \therefore$$

$$P = \begin{bmatrix} \frac{3}{2} & -\frac{1}{2} \\ -\frac{1}{2} & 1 \end{bmatrix} \quad \therefore V(x) = x^T P x = [x_1 \ x_2] \begin{bmatrix} \frac{3}{2} & -\frac{1}{2} \\ -\frac{1}{2} & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = [x_1 \ x_2] \begin{bmatrix} \frac{3}{2}x_1 - \frac{1}{2}x_2 \\ -\frac{1}{2}x_1 + x_2 \end{bmatrix} =$$

$$\frac{3}{2}x_1^2 - \frac{1}{2}x_1x_2 - \frac{1}{2}x_1x_2 + x_2^2 = \frac{3}{2}x_1^2 + x_2^2 - x_1x_2, \quad \det(P) = \det \begin{bmatrix} -1 & -1 \\ -1 & 1 \end{bmatrix} =$$

$$\left(\frac{3}{2} - \lambda\right)\left(1 - \lambda\right) - \left(-\frac{1}{2}\right)\left(-\frac{1}{2}\right) = \frac{3}{2} - \frac{5}{2}\lambda + \lambda^2 - \frac{1}{4} = \lambda^2 - \frac{5}{2}\lambda + \frac{5}{4} = Q \quad \therefore \lambda = \left(\frac{5}{2} \pm \sqrt{\left(\frac{5}{2}\right)^2 - 4\left(\frac{5}{4}\right)}\right)/2 \quad \therefore$$

$$\lambda_{\min}(P) = \frac{5-\sqrt{5}}{4} \approx 0.691 > 0 \quad \therefore P \text{ is positi desri.}$$

Week 8 Sheet / 4a. $\begin{cases} \dot{x}_1 = -x_1 + x_2^2 \\ \dot{x}_2 = 2x_1 - x_2 \end{cases}$ \therefore 2 dynamical sys is linear & nonlinear part
 $\ddot{x} = Ax + g(x)$ where $A = \begin{bmatrix} 0 & 1 \\ 2 & -1 \end{bmatrix}$ & $g(x) = \begin{bmatrix} x_2^2 \\ 0 \end{bmatrix}$
respectively. $\text{eig}(A) = -1, -1$ \therefore 2 origin is asympt stable if A is a Hurwitz mat $\therefore \text{Re}(\lambda_i(A)) < 0$

\therefore Solving Lyapunov eqn: $PA + A^T P = -I$ \therefore choice of Q is I.

$$\begin{cases} P = \begin{bmatrix} a & b \\ b & d \end{bmatrix} \quad \because P = P^T > 0 \quad \therefore A^T = \begin{bmatrix} -1 & 2 \\ 0 & -1 \end{bmatrix} \quad \therefore PA + A^T P = -I = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix} \\ \begin{bmatrix} a & b \\ b & d \end{bmatrix} \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix} + \begin{bmatrix} a & b \\ b & d \end{bmatrix} = \begin{bmatrix} -a+2b & -b \\ -b & -d \end{bmatrix} + \begin{bmatrix} -a+2b & -b+2d \\ -b+2d & -d \end{bmatrix} = \begin{bmatrix} -2a+4b & -2b+2d \\ -2b+2d & -2d \end{bmatrix} = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix} \end{cases}$$

$$-2d = -1 \quad \therefore d = \frac{1}{2} \quad \therefore 0 = -2b + 2d = -2b + 1 \quad \therefore 2b = 1 \quad \therefore b = \frac{1}{2} \quad \therefore$$

$$-2a + 4b = -1 = -2a + 2 \quad \therefore 2a = 3 \quad \therefore a = \frac{3}{2} \quad \therefore \left\{ \begin{array}{l} P = \begin{bmatrix} \frac{3}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{array} \\ \det(P - \lambda I) = \det \begin{bmatrix} \frac{3}{2} - \lambda & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} - \lambda \end{bmatrix} = \left(\frac{3}{2} - \lambda\right)\left(\frac{1}{2} - \lambda\right) - \frac{1}{4} = \lambda^2 - 2\lambda + \frac{1}{4} = \lambda^2 - 2\lambda + \frac{1}{2} = 0 \end{array} \right.$$

$$\therefore \lambda = \frac{2 \pm \sqrt{4 - 4(\frac{1}{2})}}{2} = 1 \pm \frac{\sqrt{2}}{2} \quad \left\{ \text{eigen vals of } P \text{ are } \frac{2 \pm \sqrt{2}}{2} \right.$$

$$P = 1 \pm \frac{\sqrt{2}}{2} \quad \therefore \|P\| = 1 + \frac{\sqrt{2}}{2} \quad \therefore 1 + \frac{\sqrt{2}}{2} > 1 - \frac{\sqrt{2}}{2}$$

2 Lyapunov func associated with $\dot{x} = Ax$ is $V = x^T P x$ where $P = \begin{bmatrix} \frac{3}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix}$

$$\therefore V(x) = x^T P x = \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} \frac{3}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} \frac{3}{2}x_1 + \frac{1}{2}x_2 & \frac{1}{2}x_1 + \frac{1}{2}x_2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} =$$

$$\frac{3}{2}x_1^2 + \frac{1}{2}x_1x_2 + \frac{1}{2}x_2^2 = \frac{3}{2}x_1^2 + x_1x_2 + \frac{1}{2}x_2^2 = \frac{1}{2}(3x_1^2 + 2x_1x_2 + x_2^2) \quad \therefore$$

\therefore to find a bound on nonlinearity/perturbation $\therefore \frac{\|g(x)\|}{\|x\|} \rightarrow 0$ as $\|x\| \rightarrow 0$, can

find $\Delta \|x\| < r$ st $\|g(x)\| < \gamma \|x\|$ \therefore here $\gamma < \frac{1}{2\|P\|}$

$$\therefore x^2 \leq \|x\|^2 \quad \therefore \frac{\|g(x)\|}{\|x\|} = \frac{\|x\|^2}{\|x\|} = \frac{\|x\|^2}{\|x\|} = \|x\| \quad \therefore$$

$$\frac{\|g(x)\|}{\|x\|} < \gamma \text{ when } \|x\| < r \quad \therefore \gamma = r = \frac{1}{2\|P\|} = \frac{1}{2 + \sqrt{2}}$$

\therefore can now write an invariant set, in general,

$$V(x) = \frac{1}{2}(3x_1^2 + 2x_1x_2 + x_2^2) \quad \therefore$$

$$D = \{x \in \mathbb{R}^2 \mid V < \lambda_{\min}(P) r^2\} \quad \text{where } \lambda_{\min}(P) = 1 - \frac{\sqrt{2}}{2}, \quad r^2 = \left(\frac{1}{2 + \sqrt{2}}\right)^2 \quad \therefore$$

$$D = \{x \in \mathbb{R}^2 \mid (3x_1^2 + 2x_1x_2 + x_2^2) < \frac{2 - \sqrt{2}}{6 + 4\sqrt{2}}\}$$

\therefore can now write an invariant set, in general,

$$V(x) = \frac{1}{2}(3x_1^2 + 2x_1x_2 + x_2^2) \quad \therefore$$

\therefore can now write an invariant set, in general,

$$V(x) = x^T P x \quad \text{where } P = \begin{bmatrix} -2 & 1 \\ 1 & -2 \end{bmatrix} \quad \therefore \lambda_i(A) = -2 \pm i \quad \therefore$$

\therefore can now write an invariant set, in general,

$$V(x) = \frac{1}{4}(x_1^2 + x_2^2) \quad \therefore$$

\therefore can now write an invariant set, in general,

$$\|g(x)\| \rightarrow 0 \text{ as } \|x\| \rightarrow 0 \quad \therefore g(x) = \begin{bmatrix} x_1^3 x_2^2 \\ x_1^2 x_2^3 \end{bmatrix} \quad \therefore \|g(x)\| = x_1^3 x_2^2 + x_1^2 x_2^3 =$$

\therefore can now write an invariant set, in general,

$$\|g(x)\| \leq \frac{\|x\|^3}{\|x\|} = \|x\|^2 \rightarrow 0 \text{ as } \|x\| \rightarrow 0$$

\therefore can now write an invariant set, in general,

$$\|g(x)\| \leq \frac{\|x\|^3}{\|x\|} = \|x\|^2 \rightarrow 0 \text{ as } \|x\| \rightarrow 0$$

Can we say $\forall \|x\| < r$ st $\|g(x)\| < \gamma \|x\|$ here $\gamma < \frac{1}{2\|x\|}$?

$$\frac{\|g(x)\|}{\|x\|} = x_1^2 x_2^2 \leq \frac{1}{2}(x_1^4 + x_2^4) \quad \left\{ g(x) = \begin{bmatrix} x_1^3 x_2^2 \\ x_1^2 x_2^3 \end{bmatrix} \right. \quad \frac{\|g(x)\|}{\|x\|} = x_1^2 x_2^2$$

$$\therefore \gamma \|g(x)\| = x_1^3 x_2^2 + x_1^2 x_2^3 = x_1^2 x_2^2 (x_1 + x_2) \quad \therefore \frac{\|g(x)\|}{\|x\| + x_2} = \frac{x_1^2 x_2^2 (x_1 + x_2)}{x_1 + x_2} = x_1^2 + x_2^2$$

$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \therefore \|x\| = x_1 + x_2$$

$$\frac{\|g(x)\|}{\|x\|} = x_1^2 x_2^2 \leq \frac{1}{2}(x_1^4 + x_2^4) \leq \frac{1}{2}(x_1^4 + 2x_1^2 x_2^2 + x_2^4) = \frac{1}{2}(x_1^2 + x_2^2)^2 = \frac{1}{2}\|x\|^4$$

$$\frac{\|g(x)\|}{\|x\|} \leq \frac{1}{2}\|x\|^4 \rightarrow 0 \text{ as } \|x\| \rightarrow 0 \quad \therefore \text{ can say } \forall \|x\| < r \text{ st } \|g(x)\| < \gamma \|x\|$$

$$\frac{\|g(x)\|}{\|x\|} < \gamma \text{ when } \|x\| < r \quad \therefore \gamma < \frac{1}{2\|P\|}$$

$$\frac{\|g(x)\|}{\|x\|} < \gamma < \frac{1}{2\|P\|} = \frac{1}{2(0.25)} = 2 \quad \text{when } \forall \|x\| < (\gamma)^{1/4} = \sqrt{2} = (2x_2)^{1/4} = 4^{1/4} = \sqrt[4]{2}$$

∴ can write an invariant set in general, $D = \{x \in \mathbb{R}^2 \mid V < \lambda_{\min}(P)r^2\}$

$$D = \{x \in \mathbb{R}^2 \mid \frac{1}{4}(x_1^2 + x_2^2) < 0.25(\sqrt{2})^2 = 0.25(2) = \frac{1}{2} \therefore x_1^2 + x_2^2 = 2\}$$

$$D = \{x \in \mathbb{R}^2 \mid x_1^2 + x_2^2 < 2\}$$

$$\cancel{5a} / \begin{aligned} x_1 &= -x_1 + x_2 \\ x_2 &= (x_1 + x_2) \sin x_1 - 3x_2 \end{aligned} \quad \therefore 0 = -x_1 + x_2 \therefore x_1 = x_2 \quad \therefore 0 = (x_1 + x_2) \sin x_1 - 3x_2 \therefore$$

$$(x_1 + x_2) \sin x_1 - 3x_2 = 0 = 2x_1 \sin x_1 - 3x_2 = x_1(2 \sin x_1 - 3) \quad \therefore x_1 = 0 \therefore x_2 = 0 \therefore$$

$$\text{origin is unique equili pt, } \frac{\partial \mathbf{x}}{\partial x} = \begin{bmatrix} \partial \mathbf{x}_1 / \partial x_1 & \partial \mathbf{x}_1 / \partial x_2 \\ \partial \mathbf{x}_2 / \partial x_1 & \partial \mathbf{x}_2 / \partial x_2 \end{bmatrix} = \begin{bmatrix} -1 & 1 \\ \sin x_1 + (x_1 + x_2) \cos x_1 & \sin x_1 - 3 \end{bmatrix}$$

$$\therefore A = \frac{\partial \mathbf{x}}{\partial x}(0) = \begin{bmatrix} -1 & 1 \\ 0 & -3 \end{bmatrix} \quad \therefore A \text{ is Hurwitz} \quad \therefore \text{origin is locally asymptotically stable.}$$

To show Global asymptotic Stability: let choose $V(x) = \frac{1}{2}(x_1^2 + x_2^2)$ which is a positive definite & radially unbounded. $\therefore V > 0 \quad \forall x \in \mathbb{R}^2 \setminus \{0\}, V(0) = 0 \quad V \rightarrow \infty$ as $\|x\| \rightarrow \infty$. taking deriv along \mathbf{z} trajec:

$$\dot{V} = \frac{1}{2} 2x_1 \dot{x}_1 + \frac{1}{2} 2x_2 \dot{x}_2 = x_1 \dot{x}_1 + x_2 \dot{x}_2 = -x_1^2 + 2x_1 x_2 (1 + \sin x_1) - (3x_2^2 - \sin x_1) x_2^2 \leq$$

$$-x_1^2 + 2|x_1||x_2| - 2x_2^2 = - \begin{bmatrix} |x_1| & 1 \\ |x_2| & -1 \end{bmatrix} \begin{bmatrix} 1 & -1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} |x_1| \\ |x_2| \end{bmatrix} \leq 0 \quad \forall x \neq 0$$

~~$$\therefore \dot{V} \leq 0$$~~ which is in quadratic form & is $\begin{bmatrix} 1 & -1 \\ -1 & 2 \end{bmatrix} = B$:

$$\text{But } B \text{ is defn.} \quad \begin{bmatrix} 1-\lambda & -1 \\ -1 & 2-\lambda \end{bmatrix} = (1-\lambda)(2-\lambda) + 1(-1) = \lambda^2 - 3\lambda + 2 = \lambda^2 - 3\lambda + 1 = 0 \therefore$$

$$\lambda = \frac{3 \pm \sqrt{9-4(1)(1)}}{2} = \frac{3 \pm \sqrt{5}}{2} > 0 \quad \therefore \lambda_{\min}(B) > 0 \quad \therefore \begin{bmatrix} 1 & -1 \\ -1 & 2 \end{bmatrix} \text{ is pos def mat.}$$

V is positive definite, radially unbounded, \dot{V} is negative definite. \therefore

The origin is globally asymptotically stable

~~$$5b / \begin{aligned} \dot{x}_1 &= -x_1^3 + x_2 \\ \dot{x}_2 &= -ax_1 - bx_2 \end{aligned} \quad \therefore \dot{x}_1 = 0 = -x_1^3 + x_2 \quad \dot{x}_2 = 0 = -ax_1 - bx_2 \quad \therefore x_2 = -\frac{b}{a}x_1 \therefore$$~~

$$-x_1^3 - \frac{b}{a}x_1 = -x_1(x_1^2 + \frac{b}{a}) = 0 \quad \therefore x_1 = 0 \quad \therefore x_2 = 0 \quad \therefore \text{origin is unique equili pt}$$

$$\therefore \frac{\partial \mathbf{x}}{\partial x} = \begin{bmatrix} \partial \mathbf{x}_1 / \partial x_1 & \partial \mathbf{x}_1 / \partial x_2 \\ \partial \mathbf{x}_2 / \partial x_1 & \partial \mathbf{x}_2 / \partial x_2 \end{bmatrix} = \begin{bmatrix} -3x_1^2 & 1 \\ -a & -b \end{bmatrix} \quad a, b > 0 \therefore$$

Week 8 Sheet / $A = \frac{\partial \dot{x}}{\partial x}(0) = \begin{bmatrix} 0 & 1 \\ -a & -b \end{bmatrix} \therefore \det(A) = \begin{vmatrix} -\lambda & 1 \\ -a & -b-\lambda \end{vmatrix} =$
 $-\lambda(-b-\lambda) - (-a) = \lambda^2 + b\lambda + a = 0 \therefore \lambda = \frac{-b \pm \sqrt{b^2 - 4a}}{2}$

$\lambda_{\min}(A) = \frac{-b - \sqrt{b^2 - 4a}}{2} < 0 \therefore \text{Re}(\lambda_i(A)) < 0, i=1,2 \therefore A \text{ is Hurwitz}$

& \therefore origin is asymptotically stable \therefore to show globally asymptotic stability:
 Let $V(x) = \frac{1}{2}(x_1^2 + \alpha x_2^2)$ $\alpha > 0 \therefore V$ is positive definite & radially unbounded

$\therefore V > 0 \quad \forall x \in \mathbb{R}^2 \setminus \{0\}, V \rightarrow \infty \text{ as } \|x\| \rightarrow \infty$

taking deriv w.r.t V :

$\dot{V} = x_1 \ddot{x}_1 + \alpha x_2 \ddot{x}_2 \quad \dot{V} = x_1 \dot{x}_1 + \alpha x_2 \dot{x}_2 = -x_1^4 + x_1 x_2(1 - \alpha x_2) - b \alpha x_2^2 \therefore$

taking $\alpha = \frac{b}{2}$: $\dot{V}(x) = -x_1^4 - \frac{b}{2} x_2^2 < 0 \quad \forall x \neq 0 \therefore$ origin is globally asymptotically stable $\therefore \frac{b}{2} > 0 \therefore V$ is positive definite & unbounded, & \dot{V} is negative definite

Week 9 Sheet / $\forall a / \begin{array}{l} \dot{x}_1 = -\alpha x_1 - \omega x_2 + (\beta x_1 - \gamma x_2)(x_1^2 + x_2^2) \\ \dot{x}_2 = \omega x_1 - \alpha x_2 + (\gamma x_1 + \beta x_2)(x_1^2 + x_2^2) \end{array} \therefore$
 linear part $\underbrace{\quad}_{\text{nonlinear part}}$

structure is indicated \therefore

note: $x_1^2 + x_2^2 = \|x\|_2^2 \therefore$ e.g. as: $\dot{x} = Ax + \|x\|_2^2 Bx$ where

$x = [x_1 \ x_2]^T, A = \begin{bmatrix} -\alpha & -\omega \\ -\omega & -\alpha \end{bmatrix}, B = \begin{bmatrix} \beta & -\gamma \\ \gamma & \beta \end{bmatrix} \therefore \|B\| = \sqrt{\beta^2 + \gamma^2}$

choose $V(x) = x^T x$ which is positive definite & Ser. nontrivial nominal

linear syst $A + A^T = -2\alpha I \quad V(x) = x^T x$ is Lyapunov funcn \therefore

(converse Lyapunov argument) $\dot{V}(x) \leq -2\alpha \|x\|^2 + 2\|B\| \|x\|^4 \leq -2\alpha \|x\|^2 + 2\sqrt{\beta^2 + \gamma^2} \|x\|^4$
 $\therefore \dot{V} = \dot{x}^T x + x^T \dot{x}$

for $\|x\| \leq r$: $\dot{V} \leq -2\alpha \|x\|^2 + 2r^2 \sqrt{\beta^2 + \gamma^2} \|x\|^2 = -2(\alpha - r^2 \sqrt{\beta^2 + \gamma^2}) \|x\|^2 \leq 0$

for $\alpha - r^2 \sqrt{\beta^2 + \gamma^2} > 0 \therefore \sqrt{\beta^2 + \gamma^2} < \frac{\alpha}{r^2} \therefore$ calc'g Z deriv & V directly

$\dot{V} = -2\alpha \|x\|^2 + 2\beta \|x\|^4$ when $\beta \leq 0 \quad \dot{V} \leq -2\alpha \|x\|^2 \therefore$ Z origin is globally exponentially stable \therefore chosen V 's also radially unbounded

$\therefore V \rightarrow \infty$ as $\|x\| \rightarrow \infty$ when $\beta > 0 \therefore$

$\dot{V} \leq -2(\alpha - \beta) \|x\|^2, \forall \|x\|_2 \leq r \therefore \dot{V}$ is negative definite when $r^2 < \frac{\alpha}{\beta} \therefore$

$V(x) = \|x\|^2$ we could conclude that Z set $\Omega = \left\{ x \in \mathbb{R}^2 \mid \|x\|^2 < \frac{\alpha}{\beta} \right\}$ is included in Z region of attraction \therefore result of (b) are less conservative than (a).

\therefore taking Z deriv wrt time:
 $\begin{cases} \dot{x}_1 = -x_1 + x_2 - x_3 \\ \dot{x}_2 = -x_1 x_3 - x_2 + u \\ \dot{x}_3 = -x_1 + u \end{cases} \quad y = x_3 \quad \therefore$ taking Z deriv wrt time:

$\dot{y} = \frac{d}{dt}(y) = \frac{d}{dt}(x_3) = \dot{x}_3 = -x_1 + u$ \therefore explicit presence of u

Σ Syst has a relative degree of 2 in \mathbb{R}^3 $\{\because \dot{y} = -x_1 + u\}$

$\sqrt{3}a$ $\begin{cases} \dot{x}_1 = -2x_1 + ax_2 + \sin x_1 \\ \dot{x}_2 = -x_2 \cos x_1 + u \cos(2x_1) \end{cases}$ Σ nonlinearity cannot be cancelled by control input u . define a new set of variables $z = [z_1 \ z_2]$

$$\begin{cases} z_1 = x_1 \\ z_2 = ax_2 + \sin x_1 \end{cases}$$

$$\dot{z}_1 = \dot{x}_1 = -2z_1 + ax_2 + \sin x_1 = -2z_1 + z_2$$

$$\dot{z}_2 = a\dot{x}_2 + \cos(x_1)\dot{x}_1 = a(-x_2 \cos x_1 + u \cos(2x_1)) + \cos(x_1)\dot{x}_1 =$$

$$-ax_2 \cos x_1 + au \cos(2x_1) + \cos(x_1)(-2x_1 + ax_2 + \sin x_1) =$$

$$-ax_2 \cos x_1 + au \cos(2x_1) + ax_2 \cos x_1 - 2x_1 \cos x_1 + \cos x_1 \sin x_1 =$$

$$-2x_1 \cos x_1 + \cos x_1 \sin x_1 + au \cos(2x_1) =$$

$$\dot{z}_2 = -2z_1 \cos z_1 + \cos z_1 \sin z_1 + u \cos(2z_1) \quad \therefore$$

$$\dot{z}_1 = -2z_1 + z_2 \quad \therefore$$

$$\dot{z}_2 = -2z_1 \cos z_1 + \cos z_1 \sin z_1 + u \cos(2z_1) \quad \therefore$$

$$\text{choose } u = \frac{1}{\cos(2z_1)} (\gamma - \cos z_1 \sin z_1 + 2z_1 \cos z_1) \quad \therefore$$

$$\dot{z}_2 = -2z_1 \cos z_1 + \cos z_1 \sin z_1 + \frac{1}{\cos(2z_1)} (\gamma - \cos z_1 \sin z_1 + 2z_1 \cos z_1) \alpha \cos(2z_1) =$$

$$-2z_1 \cos z_1 + \cos z_1 \sin z_1 + \gamma - \cos z_1 \sin z_1 + 2z_1 \cos z_1 = \dot{z}_2 = \gamma \quad \therefore$$

$$\dot{z}_1 = -2z_1 + z_2$$

\therefore it is now possible to pick feedback gains $k_1 \geq k_2$, in

$\gamma = -k_1 z_1 - k_2 z_2$ to place Σ poles properly. Then states z_1 & z_2 converge to zero.

Original states x_1, x_2 will also converge to zero.

$$\dot{x}_1 = 2x_2 - 2x_1 + x_1 - x_1^3 \quad \therefore$$

$$\dot{x}_2 = -2x_2 + 2x_1 + x_2 - x_2^3 \quad \therefore$$

$$\text{equilibrium when } (x_1, x_2) = (0, 0) \quad \therefore \text{let: } \dot{x}_1 = 2x_2 - x_1 - x_1^3 = 0 \quad \therefore \frac{1}{2}x_1^3 + \frac{1}{2}x_1 = x_2 \quad \therefore$$

$$x_2 = -x_2 + 2x_1 - x_1^3 = 0 = -(\frac{1}{2}x_1^3 + \frac{1}{2}x_1) + 2x_1 - (\frac{1}{2}x_1^3 + \frac{1}{2}x_1)^3 \quad \therefore$$

$$\text{if } x_1 = 0: -(\frac{1}{2}(0)^3 + \frac{1}{2}(0)) + 2(0) - (\frac{1}{2}(0)^3 + \frac{1}{2}(0))^3 = 0 = \dot{x}_2 \quad \therefore \frac{1}{2}(0)^3 + \frac{1}{2}(0) = 0 = x_2$$

$$\therefore \dot{x}_1 = 2(0) - (0) - (0)^3 = 0 \quad \therefore (x_1, x_2) = (0, 0) \text{ when } (x_1 = 0, x_2 = 0) = (0, 0) \text{ is an equil pt}$$

$$\text{if } x_1 = 1: -(\frac{1}{2}(1)^3 + \frac{1}{2}(1)) + 2(1) - (\frac{1}{2}(1)^3 + \frac{1}{2}(1))^3 = -1 + 2 - 1 = 0 = \dot{x}_2 \quad \therefore \frac{1}{2}(1)^3 + \frac{1}{2}(1) = x_2 = 1$$

$$\therefore \dot{x}_1 = 2(1) - 1 - 1 = 0 \quad \therefore (x_1, x_2) = (0, 0) \text{ when } (x_1 = 1, x_2 = 1) = (1, 1) \text{ is an equil pt}$$

$$\text{if } x_1 = -1: \frac{1}{2}(-1)^3 + \frac{1}{2}(-1) = x_2 = -1 \quad \therefore 2(-1) - (-1) - (-1)^3 = \dot{x}_1 = 0,$$

$$-(\frac{1}{2}(-1)^3 + \frac{1}{2}(-1)) + 2(-1) - (\frac{1}{2}(-1)^3 + \frac{1}{2}(-1))^3 = 1 - 2 + 1 = \dot{x}_2 = 0 \quad \therefore (x_1 = -1, x_2 = -1) = (-1, -1) \text{ is an equil pt.}$$

2 equilib are $(0, 0), (1, 1), (-1, -1)$

\checkmark My answers \checkmark 1b/ $\dot{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$ $\dot{S} = \begin{bmatrix} S_1 \\ S_2 \end{bmatrix} = \dot{x} = \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} 2x_2 - x_1 - x_1^3 \\ -x_2 + 2x_1 - x_2^3 \end{bmatrix}$

$$S_1 = 2x_2 - x_1 - x_1^3, S_2 = -x_2 + 2x_1 - x_2^3$$

$$\frac{\partial S_1}{\partial x_1} = -1 - 3x_1^2, \frac{\partial S_2}{\partial x_1} = 2, \frac{\partial S_1}{\partial x_2} = 2, \frac{\partial S_2}{\partial x_2} = -1 - 3x_2^2$$

$$\frac{\partial S}{\partial x} = \frac{\partial}{\partial x} \begin{bmatrix} S_1 \\ S_2 \end{bmatrix} = \begin{bmatrix} \frac{\partial S_1}{\partial x_1} & \frac{\partial S_1}{\partial x_2} \\ \frac{\partial S_2}{\partial x_1} & \frac{\partial S_2}{\partial x_2} \end{bmatrix} = \begin{bmatrix} -1 - 3x_1^2 & 2 \\ 2 & -1 - 3x_2^2 \end{bmatrix}$$

$$\text{For equilibrium } (0,0): \frac{\partial S}{\partial x} \Big|_{(0,0)} = \begin{bmatrix} -1 - 3(0)^2 & 2 \\ 2 & -1 - 3(0)^2 \end{bmatrix} = \begin{bmatrix} -1 & 2 \\ 2 & -1 \end{bmatrix}$$

$$\frac{\partial S}{\partial x} \Big|_{(0,0)} - \lambda I = \begin{bmatrix} -1 - \lambda & 2 \\ 2 & -1 - \lambda \end{bmatrix}, \det \left(\frac{\partial S}{\partial x} \Big|_{(0,0)} - \lambda I \right) = \begin{vmatrix} -1 - \lambda & 2 \\ 2 & -1 - \lambda \end{vmatrix} = (-1 - \lambda)(-1 - \lambda) - 2(2) = 1 + \lambda^2 + 2\lambda - 4 = \lambda^2 + 2\lambda - 3 = (\lambda + 3)(\lambda - 1) = 0 \therefore \lambda_1 = -3, \lambda_2 = 1 \therefore \lambda_1 \in \mathbb{R}_{<0}, \lambda_2 \in \mathbb{R}_{>0}$$

\checkmark equili $(0,0)$ is a saddle point

$$\text{For equili } (1,1): \frac{\partial S}{\partial x} \Big|_{(1,1)} = \begin{bmatrix} -1 - 3x_1^2 & 2 \\ 2 & -1 - 3x_2^2 \end{bmatrix} \Big|_{(1,1)} = \begin{bmatrix} -1 - 3(1)^2 & 2 \\ 2 & -1 - 3(1)^2 \end{bmatrix} = \begin{bmatrix} -4 & 2 \\ 2 & -4 \end{bmatrix}$$

$$\therefore \text{For its eigenvalue: } \det \left(\frac{\partial S}{\partial x} \Big|_{(1,1)} - \lambda I \right) = \begin{vmatrix} -4 - \lambda & 2 \\ 2 & -4 - \lambda \end{vmatrix} = (-4 - \lambda)(-4 - \lambda) - 2(2) =$$

$$16 + \lambda^2 - 8\lambda - 4 = \lambda^2 + 8\lambda + 12 = (\lambda + 6)(\lambda + 2) = 0 \therefore \lambda_1 = -6, \lambda_2 = -2 \therefore \lambda_1, \lambda_2 \in \mathbb{R}_{<0}$$

\checkmark equili $(1,1)$ is a stable node

$$\text{For equili } (-1,-1): \frac{\partial S}{\partial x} \Big|_{(-1,-1)} = \begin{bmatrix} -1 - 3x_1^2 & 2 \\ 2 & -1 - 3x_2^2 \end{bmatrix} \Big|_{(-1,-1)} = \begin{bmatrix} -1 - 3(-1)^2 & 2 \\ 2 & -1 - 3(-1)^2 \end{bmatrix} = \begin{bmatrix} -4 & 2 \\ 2 & -4 \end{bmatrix}$$

$$\text{Same as before: For its eigenvalue: } \det \left(\frac{\partial S}{\partial x} \Big|_{(-1,-1)} - \lambda I \right) = \begin{vmatrix} -4 - \lambda & 2 \\ 2 & -4 - \lambda \end{vmatrix} =$$

$$(\lambda + 6)(\lambda + 2) = 0 \therefore \lambda_1 = -6, \lambda_2 = -2 \therefore \lambda_1, \lambda_2 \in \mathbb{R}_{<0} \therefore \checkmark \text{ equili } (-1,-1) \text{ is a stable node}$$

\checkmark \checkmark \checkmark equili $(0,0)$ has eigenvals: $\lambda_1 = -3, \lambda_2 = 1$

\checkmark \checkmark equili $(1,1)$ has eigenvals $\lambda_1 = -6, \lambda_2 = -2$

\checkmark \checkmark equili $(-1,-1)$ has eigenval $\lambda_1 = -6, \lambda_2 = -2 \therefore$ all \checkmark equili pts

have atleast one eigenval that is nega. \therefore they all have at least one equili eigenval that is not possi. \therefore node or \checkmark equili are unstable nodes or unstable nodes. \therefore By \checkmark poincaré-Bendixson criterion; focus or unstable nodes. \therefore By \checkmark criterion for \checkmark existence of a limit cycle is not fulfilled.

\checkmark System does not have limit cycles.

$$\checkmark \text{ 1st/ } x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}, \therefore \dot{x} = \frac{d}{dt} x = \frac{d}{dt} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = Ax + Bu = \begin{bmatrix} 2(x_2 - x_1) + x_1(1-x_1^2) \\ -2(x_2 - x_1) + x_2(1-x_2^2) + u \end{bmatrix} =$$

$$\text{0} \begin{bmatrix} 2(x_2 - x_1) + x_1(1-x_1^2) \\ -2(x_2 - x_1) + x_2(1-x_2^2) + u \end{bmatrix} + \begin{bmatrix} 0 \\ u \end{bmatrix} = \begin{bmatrix} 2x_2 - x_1 - x_1^3 \\ -x_2 + 2x_1 - x_2^3 \end{bmatrix} + \begin{bmatrix} 0 \\ u \end{bmatrix} = \begin{bmatrix} (-1 - x_1^2)x_2 + 2(x_2) \\ (2)x_1 + (1 - x_2^2)x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ u \end{bmatrix} =$$

$$\begin{bmatrix} -1 - x_1^2 & 2 \\ 2 & -1 - x_2^2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ u \end{bmatrix} = A \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + Bu = Ax + Bu, \quad A = \begin{bmatrix} -1 - x_1^2 & 2 \\ 2 & -1 - x_2^2 \end{bmatrix}, \quad B = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$\therefore \text{about } Z \text{ origin is } (x_1, x_2) = (0, 0) \therefore A = \begin{bmatrix} -1-\alpha^2 & 2 \\ 2 & -1-\alpha^2 \end{bmatrix} = \begin{bmatrix} -1 & 2 \\ 2 & -1 \end{bmatrix}, B = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$\checkmark \text{e/controllability matrix is: } M = [B : AB] = \left[\begin{bmatrix} 0 \\ 1 \end{bmatrix} : \begin{bmatrix} 0 \\ 1 \end{bmatrix}; \begin{bmatrix} -1-x_1^2 & 2 \\ 2 & -1-x_1^2 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right] =$$

$$\left[\begin{bmatrix} 0 \\ 1 \end{bmatrix} : \begin{bmatrix} 2 \\ 1 \end{bmatrix} \right] = \begin{bmatrix} 0 & 2 \\ 1 & -1-x_1^2 \end{bmatrix} \therefore \text{Rank}(M) = 2 \text{ is full rank matrix. } \therefore$$

$$\det(M) = \begin{vmatrix} 0 & 2 \\ 1 & -1-x_1^2 \end{vmatrix} = 0(-1-x_1^2) - 1(2) = -2 \neq 0 \therefore \text{controllable. } \therefore Z \text{ pair } (A, B) \text{ is controllable. about } Z \text{ origin is } (x_1, x_2) = (0, 0) \therefore M = \begin{bmatrix} 0 & 2 \\ 1 & -1-\alpha^2 \end{bmatrix} = \begin{bmatrix} 0 & 2 \\ 1 & -1 \end{bmatrix} \text{ is controllable.}$$

$\text{Rank}(M) = 2 \therefore Z \text{ pair } (A, B) \text{ is controllable}$

$$\checkmark 18/ K = [k_1, k_2] \text{ eigenvalues } -5, -8 \therefore \text{desired charac poly: } (s+5)(s+8) = s^2 + 13s + 40 = s^2 + 13s + 40 \therefore A_{cl} = A - BK|_{(0,0)} = \begin{bmatrix} -1-x_1^2 & 2 \\ 2 & -1-x_1^2 \end{bmatrix} - \begin{bmatrix} 0 \\ 1 \end{bmatrix} \begin{bmatrix} k_1 & k_2 \end{bmatrix} = \begin{bmatrix} -1 & 2 \\ 2 & -1 \end{bmatrix} - \begin{bmatrix} 0 & 0 \\ 1 & k_2 \end{bmatrix} = \begin{bmatrix} -1 & 2 \\ 2-k_1 & -1-k_2 \end{bmatrix} \therefore \text{charac poly of closed loop:}$$

$$\det(sI - A_{cl}) = \det(sI - (A - BK)) = \det\left(\begin{bmatrix} s & 0 \\ 0 & s \end{bmatrix} - \begin{bmatrix} -1 & 2 \\ 2-k_1 & -1-k_2 \end{bmatrix}\right) = \det\left(\begin{bmatrix} s+1 & -2 \\ -2+k_1 & s+1+k_2 \end{bmatrix}\right)$$

$$= (s+1)(s+1+k_2) - (-2)(-2+k_1) = s^2 + s + k_2 s + s + 1 + k_2 - 4 + 2 k_1 =$$

$$s^2 + 2s + k_2 s + (-3 + 2k_1 + k_2) = s^2 + (2 + k_2)s + (-3 + 2k_1 + k_2) = s^2 + 13s + 40 \therefore$$

$$13 = 2 + k_2, \quad -3 + 2k_1 + k_2 = 40 \therefore 11 = k_2 \therefore$$

$$2k_1 + 11 = 43 \therefore 2k_1 = 32 \therefore k_1 = 16 \therefore$$

$$U = -Kx = -[k_1, k_2]x = -[16, 11]x = [-16, -11]x$$

$$\checkmark 19/ T = M \times W \quad M = \begin{bmatrix} 0 & 2 \\ 1 & -1 \end{bmatrix} \text{ at } (0,0) \therefore \text{full rank. } \therefore n=2 \therefore$$

$$W = \begin{bmatrix} \alpha_1 & 1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} \alpha_1 & 1 \\ 1 & 0 \end{bmatrix} \text{ where } \det(sI - A) = s^2 + \alpha_1 s + \alpha_2 \therefore A = \begin{bmatrix} -1 & 2 \\ 2 & -1 \end{bmatrix} \therefore$$

$$\det(sI - A) = \det\left(\begin{bmatrix} s & 0 \\ 0 & s \end{bmatrix} - \begin{bmatrix} -1 & 2 \\ 2 & -1 \end{bmatrix}\right) = \det\left(\begin{bmatrix} s+1 & -2 \\ -2 & s+1 \end{bmatrix}\right) = (s+1)(s+1) - (-2)(-2) =$$

$$s^2 + s + s + 1 - 4 = s^2 + 2s - 3 = s^2 + \alpha_1 s + \alpha_2 \therefore \alpha_1 = 2 \therefore W = \begin{bmatrix} 2 & 1 \\ 1 & 0 \end{bmatrix} \therefore$$

$$Z \text{ transformation Matrix is: } T = M \times W = \begin{bmatrix} 0 & 2 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} 2 & 1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} 2 & 0 \\ 1 & 1 \end{bmatrix} \text{ about } (0,0)$$

$$\checkmark 20/ \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \dot{x} = \begin{bmatrix} x_2 \\ -x_1 + x_2(2-3x_1^2-2x_2^2) \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} = \begin{bmatrix} x_2 \\ -x_1 + 2x_2 - 3x_1^2 x_2 - 2x_2^3 \end{bmatrix} \therefore$$

$$x_2 = 0 \therefore -x_1 + x_2(2-3x_1^2-2x_2^2) = 0 = -x_1 + 0(2-3(x_1^2) - 2(0)^2) = -x_1 + 0 = 0 = -x_1 \therefore$$

$x_1 = 0 \therefore$ equilibri pt at $(x_1, x_2) = (0, 0)$ is a unique equilibri. \therefore Jacobian:

$$\frac{\partial \mathbf{x}}{\partial x}|_{(0,0)} = \begin{bmatrix} 0 & 1 \\ -1-6x_1x_2 & 2-3x_1^2-6x_2^2 \end{bmatrix}|_{(0,0)} = \begin{bmatrix} 0 & 1 \\ -1-6(0)(0) & 2-3(0)^2-6(0)^2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -1 & 2 \end{bmatrix} \therefore \text{eigenvals by:}$$

$$\begin{vmatrix} 0-\lambda & 1 \\ -1 & 2-\lambda \end{vmatrix} = \begin{vmatrix} -\lambda & 1 \\ -1 & 2-\lambda \end{vmatrix} = -\lambda(2-\lambda) - 1(-1) = \lambda^2 - 2\lambda + 1 = (\lambda-1)(\lambda-1) = 0 \therefore \text{eigenvals: } \lambda_1 = 1, \lambda_2 = 1$$

$\therefore \lambda_1, \lambda_2 > 0 \therefore$ equilibria $(x_1, x_2) = (0, 0)$ is an unstable node. \therefore

\CW My answers / Let $V = x_1^2 + x_2^2$, $\therefore \nabla V(x) = \nabla V = \nabla V(x_1, x_2) = V(x_1^2 + x_2^2) =$

$$\left[\frac{\partial}{\partial x_1} (x_1^2 + x_2^2) \quad \frac{\partial}{\partial x_2} (x_1^2 + x_2^2) \right] = \begin{bmatrix} 2x_1 & 2x_2 \end{bmatrix} \quad \therefore \text{if } S = S(x), \therefore$$

$$S(x) \cdot \nabla V(x) = \begin{bmatrix} x_1 - x_1 + 2x_2 - 3x_1^2 x_2 - 2x_2^3 \end{bmatrix} \cdot \begin{bmatrix} 2x_1 & 2x_2 \end{bmatrix} =$$

$$2x_1 x_2 + 2x_2 (-x_1 + 2x_2 - 3x_1^2 x_2 - 2x_2^3) = 2x_1 x_2 - 2x_1 x_2 + 4x_2^2 - 6x_1^2 x_2^2 - 4x_2^4 =$$

$$4x_2^2 - 6x_1^2 x_2^2 - 4x_2^4 = 2x_2^2 (2 - 3x_1^2 - 2x_2^2) = 2x_2^2 (2 - x_1^2 - 2x_2^2) =$$

$$2x_2^2 (2 - x_1^2 - 2(x_1^2 + x_2^2)) = 2x_2^2 (2 - x_1^2 - 2V) = 8 - \nabla V \quad \therefore$$

$$2x_2^2 \geq 0, \therefore S(x) \cdot \nabla V(x) \leq 0 \quad \text{if } 2 - x_1^2 - 2V \leq 0 \quad \therefore 2 - x_1^2 \leq 2V \quad \therefore 1 - \frac{1}{2}x_1^2 \leq V$$

$$1 - \frac{1}{2}x_1^2 \leq 0 \quad \therefore 1 - \frac{1}{2}x_1^2 \leq 1 \quad \therefore 1 - \frac{1}{2}x_1^2 \leq V \quad \text{if } V \geq 1.$$

$\therefore V \geq 1: S(x) \cdot \nabla V(x) \leq 0, \therefore \text{on a closed bounded set}$

$M = \{x \in \mathbb{R}^2 | V(x) = x_1^2 + x_2^2 \leq 1\}$ which contains one equili, which is

unstable, \therefore Poincaré-Bendixson criterion holds, \therefore it is possible

to ensure all trajectories are trapped inside M . \therefore It can be

concluded \exists a periodic orbit in $M = \{x \in \mathbb{R}^2 | V(x) = x_1^2 + x_2^2 \leq 1\}$.

The system has a periodic orbit.

\3/ $S(x) = \frac{1}{x^2} = x^{-2}$, $\therefore S$ is Lipschitz cont on an interval if S is Lipschitz cont at pts in that interval. S is Lipschitz cont at pts in an interval if \exists a Lipschitz const $L > 0$ & a neighborhood that is a subset of that interval so that pt. $\exists \delta$ $|S(x) - S(y)| \leq L|x-y|$ for x, y in that neighborhood. \therefore without loss of gen

generality let $x, y \in [\frac{1}{2}, 2]$, $\therefore x, y \in \mathbb{R}$ & $x \geq y \therefore S(x) = \frac{1}{x^2}, S(y) = \frac{1}{y^2}$

$$\therefore x, y \geq \frac{1}{2} \quad \therefore x \leq 2x^2, y \leq 2y^2 \quad \therefore x^2 \leq 2x^3, y^2 \leq 2y^3 \quad \therefore$$

$$x^2 \leq 4x^2 y^2, y^2 \leq 4x^2 y^2 \quad \therefore x^2 \leq 8x^3 y^2, y^2 \leq 8x^2 y^3 \quad \therefore$$

$$\text{Let } L_S = 64: L_S(x-y) = 64(x-y) \quad \therefore S(x) - S(y) = \frac{1}{x^2} - \frac{1}{y^2} = \left(\frac{1}{x^2} - \frac{1}{y^2} \right) \frac{x^2 y^2}{x^2 y^2} =$$

$$\frac{y^2 - x^2}{x^2 y^2} = -(x^2 - y^2) \frac{1}{x^2 y^2} \quad \therefore x^2 - y^2 \leq 8x^3 y^2 - 8x^2 y^3 = 8(x^3 y^2 - x^2 y^3) \leq 64(x^3 y^2 - x^2 y^3)$$

$$\therefore x^2 > 0, y^2 > 0 \quad \therefore x, y \geq \frac{1}{2} \quad \therefore x^2 y^2 > 0 \quad \therefore$$

$$(x^2 - y^2) \frac{1}{x^2 y^2} \leq 64(x^3 y^2 - x^2 y^3) \frac{1}{x^2 y^2} \quad \therefore \left(\frac{x^2}{x^2 y^2} - \frac{y^2}{x^2 y^2} \right) = \left(\frac{1}{y^2} - \frac{1}{x^2} \right) = -\left(\frac{1}{x^2} - \frac{1}{y^2} \right) \leq$$

$$|64(x-y)| \sim |1 - (\frac{1}{x^2} - \frac{1}{y^2})| = |1 - \frac{1}{x^2} + \frac{1}{y^2}| = |S(x) - S(y)| \leq |64(x-y)| = 64|x-y|$$

$$\therefore |S(x) - S(y)| \leq 64|x-y| = L_S(x-y) \quad \therefore S(x) \text{ is Lipschitz cont on } \mathbb{R} \text{ with Lipschitz const } L_S = 64. \quad 2|S(x) - S(y)| \leq 64|x-y| \leq 128|x-y| = L_S|x-y| \quad \therefore S(x) \text{ is Lipschitz cont on } \mathbb{R} \text{ with Lipschitz const } L_S = 128.$$

cannot lie $\geq 128 \therefore 128 > 64$.

$$\checkmark V(0,0) = 0 \quad \& \quad V(x_1, x_2) = x_1^2 + x_2^2 > 0 \quad \forall (x_1, x_2) \in \mathbb{R}^2 \setminus \{(0,0)\}$$

V is possi des i \therefore taking \mathbb{Z} deriv of V along \mathbb{Z} direction $\frac{\partial}{\partial t}(x_1)$

$$\text{Syst: } \dot{V} = \dot{V}(x_1, x_2) = \frac{\partial}{\partial t} V(x_1, x_2) = \frac{\partial}{\partial t} [x_1^2 + x_2^2] = 2x_1 \frac{\partial}{\partial t}(x_1) + 2x_2 \frac{\partial}{\partial t}(x_2) = \\ 2x_1 \dot{x}_1 + 2x_2 \dot{x}_2 = 2x_1(-3x_1 + x_1 x_2) + 2x_2(x_1 - 2x_2) = -6x_1^2 + 2x_1^2 x_2 + 2x_1 x_2 - 4x_2^2 = \\ (-5x_1^2 - x_2^2) + 2x_1^2 x_2 + (-3x_2^2 - x_2^2 + 2x_1 x_2) = (-5x_1^2 + 2x_1^2 x_2 - 3x_2^2) + (-x_2^2 + 2x_1 x_2 - x_2^2) = \\ (-5x_1^2 + 2x_1^2 x_2 - 3x_2^2) - (x_2^2 - 2x_1 x_2 + x_2^2) = (-5x_1^2 + 2x_1^2 x_2 - 3x_2^2) - (x_1 - x_2)^2$$

$2(x_1 - x_2)^2 \geq 0 \therefore (x_1 - x_2)^2$ is lower bounded by zero.

$$\dot{V} = (-5x_1^2 + 2x_1^2 x_2 - 3x_2^2) - (x_1 - x_2)^2 \leq -5x_1^2 + 2x_1^2 x_2 - 3x_2^2 = -x_1^2(5 - 2x_2) - 3x_2^2 \leq$$

$$\dot{V} \leq -x_1^2(5 - 2x_2) - 3x_2^2 \therefore 2x_2^2, x_1^2 \in \mathbb{R}_{\geq 0} \therefore -3x_2^2 \leq 0, -x_1^2 \leq 0 \therefore$$

$$\text{if } 5 - 2x_2 > 0 : \dot{V} < 0 \therefore 5 > 2x_2 \therefore \cancel{\frac{5}{2} > x_2} \therefore (\frac{5}{2})^2 > (x_2)^2 \therefore \frac{25}{4} > x_2^2$$

$\therefore \frac{25}{4} > x_1^2 + x_2^2 \therefore \text{for } x_1^2 + x_2^2 < \frac{25}{4} : \dot{V} < 0 \therefore \mathbb{Z}$ origin is locally asymptotically stable with invariant set $D = \{x \in \mathbb{R}^2 \mid V(x) < \frac{25}{4}\}$ in which V is possi des i

$$\therefore V(0,0) = 0^2 + 0^2 = 0 \quad \& \quad x_1^2 \geq 0, x_2^2 \geq 0 \therefore \text{for } x_1, x_2 \neq 0 : V(x_1, x_2) = x_1^2 + x_2^2 > 0$$

$\therefore \dot{V}$ is negat des i $\therefore \dot{V}(0,0) \leq 0 \leq \dot{V} \leq 0 \leq \text{for } x_1, x_2 \neq 0 : \dot{V} \leq 0$

$$\checkmark 5a / M, g, k, c_1, c_2 \in \mathbb{R}_{>0} \quad Mij = Mg - C_1g - C_2g|j| - ky \therefore$$

$$\ddot{y} = M^{-1}Mg - M^{-1}C_1g - M^{-1}C_2g|j| - M^{-1}ky \therefore$$

$$\ddot{y} = g - M^{-1}C_1g - M^{-1}C_2g|j| - M^{-1}ky \therefore \begin{array}{l} x_1 = y - Mgk^{-1} \\ x_2 = j \end{array} \therefore$$

$$\ddot{x}_1 = \frac{d}{dt} x_1 = \frac{d}{dt} (y - Mgk^{-1}) = \frac{d}{dt} y = \ddot{y} = x_2$$

$$\ddot{x}_2 = \frac{d}{dt} (x_2) = \frac{d}{dt} (\ddot{y}) = \ddot{y} = g - M^{-1}C_1g - M^{-1}C_2g|j| - M^{-1}ky \leq (x_1 + Mgk^{-1}) = y \therefore$$

$$\ddot{x}_2 = g - M^{-1}C_1x_2 - M^{-1}C_2x_2|x_2| - M^{-1}k(x_1 + Mgk^{-1}) =$$

$$g - M^{-1}C_1x_2 - M^{-1}C_2x_2|x_2| - M^{-1}kx_1 - M^{-1}kMgk^{-1} = g - M^{-1}C_1x_2 - M^{-1}C_2x_2|x_2| - M^{-1}kx_1 - g$$

$$= -M^{-1}C_1x_2 - M^{-1}C_2x_2|x_2| - M^{-1}kx_1 = \ddot{x}_2 \therefore$$

$$\dot{x} = \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} x_1 \\ x_2 \\ -M^{-1}C_1x_2 - M^{-1}C_2x_2|x_2| - M^{-1}kx_1 \end{bmatrix} \therefore \text{compact state space: } \begin{array}{l} x = [y \quad j] \in \mathbb{R}^2 \\ x = [x_1 \quad x_2] \end{array}$$

$\checkmark 5b / \mathbb{Z}$ origin os $i = s(x)$ is globally asymptotically stable if it is Lyapunov stable $\& \forall x(0) \in \mathbb{R}^n, \lim_{t \rightarrow \infty} x(t) = 0$, \mathbb{Z} origin os $i = s(x)$ is Lyapunov stable, is

$\forall \epsilon > 0, \exists \delta = \delta(\epsilon) > 0$ st if $\|x(0)\| < \delta \therefore \|x(t)\| < \epsilon, \forall t \geq 0 \therefore$ origin is

$$(x_1, x_2) = (0,0) \therefore \dot{x}|_{(0,0)} = \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix}|_{(0,0)} = \begin{bmatrix} x_1 \\ x_2 \\ -M^{-1}C_1x_2 - M^{-1}C_2x_2|x_2| - M^{-1}kx_1 \end{bmatrix}|_{(0,0)} = \\ \begin{bmatrix} 0 \\ 0 \\ -M^{-1}C_1(0) - M^{-1}C_2(0)|0| - M^{-1}k(0) \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \therefore$$

CW my answers/ $\dot{x}_1 = 0, \dot{x}_2 = 0$ at Z origin $\therefore (x_1, x_2)$ is an equili pt

Sor $(x_1, x_2) = (0, 0), x_1^2 \geq 0 \& x_2^2 \geq 0 \therefore V(x_1, x_2) = ax_1^2 + bx_2^2 \geq 0$ for $a, b > 0$

$$V(x_1=0, x_2=0) = a(0)^2 + b(0)^2 = 0 \therefore V(x_1, x_2) > 0, \forall (x_1, x_2) \in \mathbb{R}^2 \setminus \{(0, 0)\}$$

$$\dot{V}(x_1, x_2) = \frac{d}{dt} V(x_1, x_2) = \frac{d}{dt} (ax_1^2 + bx_2^2) = a \frac{d}{dt}(x_1^2) + b \frac{d}{dt}(x_2^2) =$$

$$2ax_1 \frac{d}{dt}(x_1) + 2bx_2 \frac{d}{dt}(x_2) = 2ax_1 \dot{x}_1 + 2bx_2 \dot{x}_2 =$$

$$2ax_1 \dot{x}_1 + 2bx_2 (-M^{-1}C_1 x_2 - M^{-1}C_2 x_1 |x_2| - M^{-1}k x_1) =$$

$$2ax_1 \dot{x}_1 + 2M^{-1}C_1 b x_2^2 - 2M^{-1}C_2 b x_2^2 |x_2| - 2M^{-1}k b x_1 x_2 =$$

$$(a - M^{-1}kb) 2x_1 x_2 - 2M^{-1}(C_1 b x_2^2 + C_2 b x_2^2 |x_2|) = (a - M^{-1}kb) 2x_1 x_2 - 2M^{-1}x_2^2 (C_1 b + C_2 b |x_2|)$$

$$= (a - M^{-1}kb) 2x_1 x_2 - 2M^{-1}b x_2^2 (C_1 + C_2 |x_2|) = \dot{V}(x_1, x_2) \therefore \text{let } a = M^{-1}kb \therefore V(x_1, x_2) =$$

$$-2M^{-1}b x_2^2 (C_1 + C_2 |x_2|) \therefore x_2^2 \geq 0, |x_2| \geq 0, M > 0 \therefore M^{-1} > 0, b > 0 \therefore$$

$$-2M^{-1}b x_2^2 (C_1 + C_2 |x_2|) \leq 0 \therefore \dot{V}(x_1, x_2) \leq 0 \therefore \text{Sor } a, b > 0 \& a = M^{-1}kb.$$

Z origin $(x_1, x_2) = (0, 0)$ is an globally asymptotic stable equili pt.

~~if~~ $\dot{V}(x_1, 0) = 0 \& \dot{V} < 0$ for $x_2 \neq 0$ \therefore globally nsgn desri but only can be said to be globally asymptotic stable & not globally stable \therefore

\dot{V} is indep w/ x_1 .

Week 1 Sheet $y^{(n)} = \frac{d^n y}{dt^n} \therefore y^{(1)} = g(y, u), y^{(2)} = g(y, \dot{y}, u), y^{(3)} = g(y, \ddot{y}, \dot{y}, u)$

$$y^{(4)} = g(y, \ddot{y}, \dot{\ddot{y}}, y^{(3)}, u) \therefore \begin{bmatrix} y^{(1)} \\ y^{(2)} \\ y^{(3)} \\ y^{(4)} \end{bmatrix} = \begin{bmatrix} g(y, u) \\ g(y, \dot{y}, u) \\ g(y, \ddot{y}, \dot{y}, u) \\ g(y, \ddot{y}, \dot{\ddot{y}}, u) \end{bmatrix} \text{ let } x_1 = y, x_2 = \frac{dy}{dt} \dots x_n = \frac{d^{n-1}y}{dt^{n-1}}$$

$$\therefore \text{taking derivs of } 2 \text{ new varians} = \dot{x}_1 = \ddot{y}, \dot{x}_2 = \dot{\ddot{y}} \dots \dot{x}_{n-1} = \frac{d^{n-1}y}{dt^{n-1}}, \dot{x}_n = \frac{d}{dt} \left(\frac{d^{n-1}y}{dt^{n-1}} \right) = y^{(n)} = g(y, \ddot{y}, \dots, y^{(n-1)}, u) \therefore x = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_{n-1} \\ x_n \end{bmatrix} = \begin{bmatrix} \dot{x}_1 \\ x_2 \\ \vdots \\ x_{n-1} \\ g(y, \ddot{y}, \dots, y^{(n-1)}, u) \end{bmatrix} = \begin{bmatrix} x_2 \\ x_3 \\ \vdots \\ x_n \\ g(x_1, x_2, \dots, x_{n-1}, u) \end{bmatrix}$$

$$\therefore x = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \therefore y = cx \quad c = [1 \ 0 \ \dots \ 0]_{1 \times n} \therefore y = x,$$

$$\sqrt{2} \quad \text{let } x_1 = q_1, x_3 = q_2 \therefore \dot{q}_1 = \frac{d}{dt} q_1 = \frac{d}{dt} x_1 = \dot{x}_1 = \ddot{q}_1 = \dot{x}_2, \dot{q}_2 = \frac{d}{dt} q_2 = \frac{d}{dt} x_3 = \ddot{x}_3 = \dot{x}_4 \therefore$$

$$\dot{x}_1 = \frac{d}{dt} x_1 = \frac{d}{dt} q_1 = \dot{q}_1 = x_2 \therefore \dot{x}_3 = \dot{q}_2 = x_4 \therefore \dot{q}_2 = J^{-1}u + J^{-1}k(q_1 - q_2),$$

$$\dot{q}_1 = -I^{-1}Mg L \sin q_1 - I^{-1}k(q_1 - q_2) \therefore \dot{x}_2 = \frac{d}{dt} x_2 = \frac{d}{dt} \dot{q}_1 = \dot{\dot{q}}_1 = -I^{-1}Mg L \sin q_1 - I^{-1}k(q_1 - q_2)$$

$$\therefore \dot{x}_4 = \frac{d}{dt} x_4 = \frac{d}{dt} \dot{q}_2 = \dot{q}_2 = J^{-1}u + J^{-1}k(q_1 - q_2) \therefore x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} q_1 \\ \dot{q}_1 \\ q_2 \\ \dot{q}_2 \end{bmatrix} \therefore x = [q_1, \dot{q}_1, q_2, \dot{q}_2] \in \mathbb{R}^4$$

$$\sqrt{3} \quad \text{let } x_1 = y \therefore \dot{y} = \frac{d}{dt} y = \frac{d}{dt} x_1 = \dot{x}_1 = x_2 \therefore \dot{y} = \frac{d}{dt} \dot{y} = \frac{d}{dt} x_2 = \ddot{x}_2 = -\frac{k}{m}x_1 - \frac{c_1}{m}x_2 - \frac{c_2}{m}x_2|x_2| + g$$

$$\therefore \text{states } [y \ \dot{y}] \in \mathbb{R}^2 \therefore x = [x_1 \ x_2]$$

$$x = \begin{bmatrix} y \\ \dot{y} \end{bmatrix} \in \mathbb{R}^2$$

$$\checkmark 4a) \quad x_1 = r(t), \quad x_2 = \dot{r}(t), \quad x_3 = \ddot{r}(t), \quad x_4 = \dddot{r}(t) \quad \therefore \quad \dot{x}_1 = \ddot{r}(t) = x_2,$$

$$\dot{x}_2 = \ddot{r}(t) = r(t)\theta^2(t) - \theta r'^2(t) + u_1(t) \quad \therefore \dot{x}(t) = f(x, u)$$

$$\dot{x}_3 = \dddot{r}(t) = x_4 \quad \dot{x}_4 = \ddot{\theta}(t) = \frac{1}{r(t)} [-2\ddot{r}(t)\dot{\theta}(t) + u_2(t)] \quad \therefore x = [x_1, x_2, x_3, x_4]^T =$$

$$\begin{bmatrix} r(t) & \dot{r}(t) & \theta(t) & \dot{\theta}(t) \end{bmatrix} \text{ for } r(t), \dot{r}(t), \theta(t), \dot{\theta}(t) \text{ etc } F = [s_1, s_2, s_3, s_4]^T$$

$$\dot{x}_1 = x_2, \quad \dot{x}_2 = x_1 x_4^2 - \theta x_1^{-2} + u_1, \quad \dot{x}_3 = x_4 \quad \therefore \dot{x}_4 = -2x_1 x_4^{-1} + u_2 \text{ is compact}$$

$$(\text{State Space form } \dot{x}(t) = f(x, u) \text{ where } x = [x_1, x_2, x_3, x_4]^T \equiv [r(t), \dot{r}(t), \theta(t), \dot{\theta}(t)]^T)$$

$$(FR) F = [s_1, s_2, s_3, s_4]^T \Rightarrow u = [u_1, u_2]^T$$

\ 4b) $\tilde{x}(t) = [1 \ 0 \ \sqrt{2}t \ \sqrt{2}]^T \tilde{u}(t) F(0, 0)^T \approx \text{linearised eqn about this}$

trajectory is $\tilde{x} = Ax + Bu$ where $A = \begin{bmatrix} \frac{\partial x_1}{\partial x_1} & \frac{\partial x_1}{\partial x_2} & \frac{\partial x_1}{\partial x_3} & \frac{\partial x_1}{\partial x_4} \\ \frac{\partial x_2}{\partial x_1} & \frac{\partial x_2}{\partial x_2} & \frac{\partial x_2}{\partial x_3} & \frac{\partial x_2}{\partial x_4} \\ \frac{\partial x_3}{\partial x_1} & \frac{\partial x_3}{\partial x_2} & \frac{\partial x_3}{\partial x_3} & \frac{\partial x_3}{\partial x_4} \\ \frac{\partial x_4}{\partial x_1} & \frac{\partial x_4}{\partial x_2} & \frac{\partial x_4}{\partial x_3} & \frac{\partial x_4}{\partial x_4} \end{bmatrix} =$

$$\begin{bmatrix} 0 & 1 & 0 & 0 \\ x_4^2 + 2\theta x_1^{-3} & 0 & 0 & 2x_1 x_2 \\ 0 & 0 & 1 & 0 \\ -2x_1 x_4 x_1^{-2} u_2 x_1^{-2} & -2x_1 x_4^{-1} & 0 & -2x_2 x_1^{-1} \end{bmatrix} \begin{bmatrix} \tilde{x}(t), \tilde{u}(t) \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 3\sqrt{2} & 0 & 0 & \sqrt{2} \\ 0 & 0 & 0 & 1 \\ 0 & -2\sqrt{2} & 0 & 0 \end{bmatrix} \begin{bmatrix} s_1 u_1, s_2 u_2 \\ s_3 u_1, s_4 u_2 \\ s_3 u_1, s_4 u_2 \\ s_3 u_1, s_4 u_2 \end{bmatrix} =$$

$$\begin{bmatrix} (x_1)_{u_1} & (x_1)_{u_2} \\ (x_2)_{u_1} & (x_2)_{u_2} \\ (x_3)_{u_1} & (x_3)_{u_2} \\ (x_4)_{u_1} & (x_4)_{u_2} \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 1 & 0 \\ 0 & 0 \\ 0 & x_1^{-1} \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 1 & 0 \\ 0 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} (x_1)_{u_1}, (x_1)_{u_2} \\ (x_2)_{u_1}, (x_2)_{u_2} \\ (x_3)_{u_1}, (x_3)_{u_2} \\ (x_4)_{u_1}, (x_4)_{u_2} \end{bmatrix} \begin{bmatrix} (x_1)_{u_1}, (x_1)_{u_2} \\ (x_2)_{u_1}, (x_2)_{u_2} \\ (x_3)_{u_1}, (x_3)_{u_2} \\ (x_4)_{u_1}, (x_4)_{u_2} \end{bmatrix} \quad \therefore \tilde{x} = Ax + Bu$$

$$\checkmark 4c) \quad y(t) = [r(t) \ \theta(t)]^T \quad \therefore y = Cx \quad \therefore y \in \mathbb{R}^2 \quad x \in \mathbb{R}^4 \quad C \text{ mat } C = \mathbb{R}^{2 \times 4} \quad \therefore$$

$$x = [r(t) \ \dot{r}(t) \ \theta(t) \ \dot{\theta}(t)]^T \quad \therefore \begin{bmatrix} r(t) \\ \dot{r}(t) \\ \theta(t) \\ \dot{\theta}(t) \end{bmatrix} = C \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} \quad \therefore C = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}$$

$$\checkmark 5a) \quad \dot{x} + x_1 = x \quad \therefore x_2 = \dot{x} \quad \therefore \dot{x} = \frac{d}{dt} x = \frac{d}{dt} x_1 = \dot{x}_1 = x_2 \quad \therefore \dot{x}_2 = \frac{d}{dt} x_2 = \frac{d}{dt} \dot{x} = \ddot{x} = -\frac{k_1}{m} x_1 - \frac{k_2}{m} x_1^3$$

$$= -\frac{k_1}{m} x_1 - \frac{k_2}{m} x_1^3 \text{ is in compact form with states } x = [x_1, x_2]$$

$$\checkmark 5b) \quad \text{For equilib pts solve } x_2 = 0 \quad \therefore \frac{dx_2}{dt} = \frac{d}{dt}(0) = 0 = \dot{x}_2 \quad \therefore$$

$$-\frac{k_1}{m} x_1 - \frac{k_2}{m} x_1^3 = 0 = x_1 \left(-\frac{k_1}{m} - \frac{k_2}{m} x_1^2 \right) \quad \therefore x_1 = 0, -\frac{k_1}{m} - \frac{k_2}{m} x_1^2 = 0 \quad \therefore -\frac{k_1}{m} = \frac{k_2}{m} x_1^2 \quad \therefore -\frac{k_1}{k_2} = x_1^2 \quad \therefore$$

$$x_1 = \pm \sqrt{-\frac{k_1}{k_2}}, \quad x_2 = -\sqrt{-\frac{k_1}{k_2}} \quad \therefore k_1 = -1, \quad k_2 = \frac{1}{2} \quad \therefore k_1 = \sqrt{-\frac{1}{k_2}} = \sqrt{2} \quad \therefore x_1 = \sqrt{-\frac{1}{k_2}} = -\sqrt{2}, \quad x_1 = 0 \quad \therefore$$

$$(0, 0), (\sqrt{2}, 0), (-\sqrt{2}, 0)$$

$$\checkmark 5c) \quad \text{For linearised plant, determine Jacobian } A = \begin{bmatrix} \frac{\partial s_1}{\partial x_1} & \frac{\partial s_1}{\partial x_2} \\ \frac{\partial s_2}{\partial x_1} & \frac{\partial s_2}{\partial x_2} \end{bmatrix}$$

$$s_1 = x_1, \quad s_2 = \dot{x}_2 = -\frac{k_1}{m} x_1 - \frac{k_2}{m} x_1^3 \quad \therefore A = \begin{bmatrix} 0 & 1 \\ -k_1 - 3k_2 x_1^2 & 0 \end{bmatrix} \quad \text{evaluating } A \text{ about equilis:}$$

$$\text{at } (0, 0): A_{(0, 0)} = \begin{bmatrix} 0 & 1 \\ -k_1 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \quad \dot{x} = A_{(0, 0)} x \quad , \quad \text{at } (\pm\sqrt{2}, 0): A_{(\pm\sqrt{2}, 0)} = \begin{bmatrix} 0 & 1 \\ -k_1 - 3k_2(\pm\sqrt{2})^2 & 0 \end{bmatrix} =$$

$$\begin{bmatrix} 0 & 1 \\ -2 & 0 \end{bmatrix} \quad \therefore \dot{x} = A_{(\pm\sqrt{2}, 0)} x$$

$$\checkmark \text{Check } k_2 / \quad \begin{aligned} \dot{x}_1 &= -x_1 + x_2 \\ \dot{x}_2 &= -0.1x_1 - 2x_2 - x_1^2 - 0.1x_1^3 \quad \therefore B = -x_1 + x_2 \quad \therefore x_1 = x_2 \quad \therefore 0 = -0.1x_1 - 2x_2 - x_1^2 - 0.1x_1^3 \\ &- 0.1x_1^3 - x_1^2 - 2x_1 \quad \therefore x_1(0.1x_1^2 - x_1 - 2) = 0 \quad 3 \text{ equili pts are } (0, 0)^T, (-2.76, -2.76)^T, (-4.23, -4.23)^T \end{aligned}$$

\Week2 Sheet / \3a) $x_1 = x_2$ $\dot{x}_2 = -x_1 - 2 \tan^{-1}(x_1 + x_2)$

To determine equili pt: $\dot{x} = f(x)$ $\dot{x}_1 = 0$ i.e
 $x_2 = 0$ $-x_1 - 2 \tan^{-1}(x_1 + x_2) = 0 \Rightarrow x_1 = 2 \tan^{-1}(x_1)$; $\tan\left(\frac{x_1}{2}\right) = x_1 \Rightarrow$

$x_1 \approx \pm 2.33$ So system has 3 equili pts

they are at $(x_1^*, x_2^*)^T$ respectively $(0, 0)^T$, $(2.33, 0)^T$, $(-2.33, 0)^T$

$$\text{Jacobian matrix } \frac{\partial f}{\partial x} = \begin{bmatrix} 0 & 1 \\ -2/(1+(x_1+x_2)^2) & -2/(1+(x_1+x_2)^2) \end{bmatrix} \quad \frac{\partial f}{\partial x} \Big|_{(0,0)^T} = \begin{bmatrix} 0 & 1 \\ -1 & -2 \end{bmatrix} \Rightarrow$$

$\lambda_{1,2} = -1, -1$ multiple eigenvalues at -1 leading to stable SInn

$$\text{is } \frac{\partial f}{\partial x} \Big|_{(2.33, 0)^T} = \begin{bmatrix} 0 & 1 \\ -0.6892 & -0.3108 \end{bmatrix} \Rightarrow \lambda_{1,2} = 0.6892, -1 \text{ this means}$$

$(2.33, 0)^T$ is a saddle as one eigenvalue LHP (left hand plane) & saddle

other RHP (right hand plane)

similarly $(-2.33, 0)^T$ is also a saddle. in 2 given diagram

\3b) $x_1 = x_2$ $\dot{x}_1 = -x_1 + x_2(1 - 3x_1^2 - 2x_2^2) \Rightarrow x_2 = 0$ $\Delta x_1 = 0$ at $x_1 = 0$

there is an unique equili pt at 2 origin

$$\frac{\partial f}{\partial x} \Big|_{(0,0)} = \begin{bmatrix} 0 & 1 \\ -1 & 1 \end{bmatrix} \Rightarrow \lambda_{1,2} = \frac{1}{2} \pm i\sqrt{3}/2 \Rightarrow (0,0)^T \text{ is an unstable focus}$$

there is a periodic orbit around 2 origin from 2 diagram

(periodic orbit & existence)

\4a) $\dot{x}_1 = -x_1^2 + x_1 + \sin x_2 \leftarrow S_1$, $\dot{x}_2 = \cos x_2 - x_1^3 - 5x_2 \leftarrow S_2$

$$\text{at equili } x_0 = (1, 0)^T \quad \text{Jacobian } \frac{\partial f}{\partial x} = \begin{bmatrix} -2x_1 + 1 & \cos x_2 \\ -3x_1^2 & -\sin x_2 - 5 \end{bmatrix} \Big|_{(1, 0)} = \begin{bmatrix} -1 & 1 \\ -3 & -5 \end{bmatrix}$$

eigenvals are -2 8 -4 : both eigenval are real neg

$(1, 0)^T$ is a stable node

\5a) False : From given diagram it can be seen Q₃ is saddle

\5b) True

\5c) False : $\dot{x}_1 = x_2$ $\dot{x}_2 = -h(x_1, x_2)$ since $\dot{x}_2 = -5x_1 x_2$ has explicit appearance of 't' on RHS of dynamics. this is not autonomous System. non-autonomous time varying system

\5d) True / equili subspace comes to when one or both signals are zero. $\pm i\beta$ comes to center

Week 3 sheet /

$$\check{y} + y = \varepsilon \check{y}(1 - y_1^2 - y_2^2) \quad \text{let } x_1 = y \quad x_2 = \check{y} \quad \left. \begin{array}{l} \dot{x}_1 = x_2 \\ \dot{x}_2 = \check{y} \end{array} \right\} \therefore \dot{x}_2 = -x_1 + \varepsilon x_2(1 - x_1^2 - x_2^2)$$

$\check{y} = \check{y}(x) \quad x \in \mathbb{R}^2$ form

choose $V(x) = x_1^2 + x_2^2 = C$ dissable let us evaluate now on $V(x) = C$

$$\check{y}(x) \cdot \nabla V(x) = \frac{\partial V}{\partial x_1} \check{y}_1 + \frac{\partial V}{\partial x_2} \check{y}_2 \leftarrow \check{y}_1, \check{y}_2$$

$$= 2x_1 x_2 (1 - x_1^2 - x_2^2) = 2\varepsilon x_2^2 (1 - V) \quad \text{r.}$$

$\check{y}(x) \cdot \nabla V(x) \leq 0$ for $V(x) \geq 1$ in particular, all trajectories starting in $M = \{x \in \mathbb{R}^2 \mid V(x) \leq 1\}$ (closed bounded set) stay in M & trajectories further, M contains only one equili at origin. linearisation about origin yields matrix $\begin{bmatrix} 0 & 1 \\ 0 & \varepsilon \end{bmatrix}$ \Rightarrow origin is unstable node or source for $\varepsilon > 0$. By PB (poincare Bendixson) criterion, thus

origin in M

$\check{y}_{\text{sol}} / \dot{x}_1 = x_2 \quad \dot{x}_2 = -(1 + x_1^2 + x_2^2)x_2 - x_1 \quad \text{①}$ System given in ① is an autonomous dynamical system of dimens 2. Bendixson criterion for 2nd order autonomous is as follows:

is on a simply connected region $D \subset \mathbb{R}^2$ in x_1-x_2 plane, $\frac{\partial \check{y}_1}{\partial x_1} + \frac{\partial \check{y}_2}{\partial x_2}$ is not identically zero, & does not change sign, then \mathbb{R}^2 system has no close trajectory lying entirely in D

$$\frac{\partial \check{y}_1}{\partial x_1} = 0 \quad \frac{\partial \check{y}_2}{\partial x_2} = -(1 + x_1^2 + x_2^2) \quad \frac{\partial \check{y}_1}{\partial x_1} + \frac{\partial \check{y}_2}{\partial x_2} \leq -1 < 0 \quad \forall x_1, x_2$$

(remark: $x_1^2 + x_2^2$ pos terms. smallest possible term for $(1 + x_1^2 + x_2^2)$ is 1)
 $\frac{\partial \check{y}_1}{\partial x_1} + \frac{\partial \check{y}_2}{\partial x_2}$ does not change sign in $\mathbb{R}^2 \Rightarrow \therefore$ system has no close trajectory

$\check{y}_{\text{sol}} / \dot{x}_1 = x_1 - x_2 - x_1(x_1^2 + 2x_2^2) \quad \dot{x}_2 = x_1 + x_2 - x_2(x_1^2 + 2x_2^2) \quad \text{①}$ consider $V(x) = x_1^2 + x_2^2 = r^2$ where $\frac{1}{r^2} < r < 1$ above system is in $\check{y} = \check{y}(x)$ form. V is contly dissable vector field $\check{y}(x)$ abt r on curve $V(x)$

$$\check{y}(x) \cdot \nabla V(x) = \frac{\partial V}{\partial x_1} (\check{y}_1 - x_1(x_1^2 + 2x_2^2)) + \frac{\partial V}{\partial x_2} (\check{y}_2 - x_2(x_1^2 + 2x_2^2)) \leftarrow \check{y}_1, \check{y}_2$$

$$= 2x_1^2 - 2x_1x_2 - 2x_2^2(x_1^2 + 2x_2^2) + 2x_1^2 - 2x_2^2(x_1^2 + 2x_2^2) + 2x_1x_2$$

$$= 2(x_1^2 + x_2^2)[1 - (x_1^2 + 2x_2^2)] = 2r^2[1 - r^2 - x_2^2] \quad \left\{ \begin{array}{l} \text{common} \\ \text{cancels} \end{array} \right.$$

when $r = 1$: $\check{y}(x) \cdot \nabla V(x) = -2x_2^2 \leq 0$ likewise, subing for x_2^2 with $r^2 - x_1^2$

$$\check{y}(x) \cdot \nabla V(x) = 2r^2(1 - 2r^2 + x_1^2)$$