

## # Exercise 1: Asymptotic & Lyapunov Stability

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(a)  $x(k+1) = \begin{bmatrix} 1 & 0 \\ -0.5 & 0.5 \end{bmatrix} x(k) + \begin{bmatrix} 1 \\ -1 \end{bmatrix} u(k)$

**A**                           **B**

Finding eigenvalues of A:

$$\begin{bmatrix} 1 & 0 \\ -0.5 & 0.5 \end{bmatrix} \quad \left. \begin{array}{l} \lambda_1 = 1 \\ \lambda_2 = 0.5 \end{array} \right\} \text{upper triangular matrix}$$

For stability of a discrete LTI:

Criteria I:  $|\lambda_i| \leq 1 \Rightarrow \lambda_1 = 1 \leq 1 \quad \checkmark$

$$\lambda_2 = 0.5 \leq 1 \quad \checkmark$$

Criteria II:  $|\lambda_i| = 1 \rightarrow \text{non-defective}$

Here,  $\lambda_1 = 1$  is non-defective as its algebraic multiplicity is equal to its geometric multiplicity.

$\therefore$  The given system is **Lyapunov stable** as it satisfies both the criteria.

For the system to be Asymptotically stable,  $\lambda_i < 1$

$\lambda_1 = 1$  is not less than 1.

$\therefore$  The system is **not Asymptotically stable**.

(b)

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

**A**                           **B**

Finding eigenvalues of A:  $|\lambda I - A| = 0$

$$\begin{vmatrix} \lambda+7 & 2 & -6 \\ -2 & \lambda+2 & 2 \\ 2 & 2 & \lambda-1 \end{vmatrix} = 0$$

$$(\lambda+7)[(\lambda+2)(\lambda-1)-4] - 2[-2(\lambda-1)-4] - 6[-4-2(\lambda+2)] = 0$$

$$\lambda^3 + 9\lambda^2 + 23\lambda + 15 = 0$$

$$\lambda_1 = -1$$

$$\lambda_2 = -3$$

$$\lambda_3 = -5$$

$$\lambda_i = \text{Re} + \text{Im}j$$

For Asymptotic stability,  $\text{Re } \lambda_i < 0$

For the above system,  $\lambda_1, \lambda_2 \notin \lambda_3$  satisfies the above condition.

$\therefore$  This system is **asymptotically stable**.

$\therefore$  It is also **Lyapunov stable**

## #Exercise 2: Stabilizability

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

$$y = [1 \ 1 \ 1] x$$

controllable form:

$$P = [B \ AB \ A^2B]$$

$$\text{rank}(P) = 2 < 3$$

$\therefore$  The system is not controllable

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

Using the Controllable Kalman decomposition:

$$M = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 2 & 1 \end{bmatrix} \quad r(M) = 3$$

$\downarrow$  1<sup>st</sup> two linearly independent columns of P.

arbitrary column to make M full rank.

$$\hat{A} = M^{-1}AM = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 2 & 1 \end{bmatrix}^{-1} \begin{bmatrix} 0 & -1 & 1 \\ 1 & -2 & 1 \\ 0 & 1 & -1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 2 & 1 \end{bmatrix}$$

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

$$\hat{B} = M^{-1}B = \begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 1 & -2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$$

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

$$= [3 \ 3 \ 1]$$

$\therefore$  The controllable form is:

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

$$y = [3 \ 3] \tilde{x}_c$$

Observability:

$$Q = \begin{bmatrix} C \\ CA \end{bmatrix} = \begin{bmatrix} 3 & 3 \\ 0 & 0 \end{bmatrix}$$

$$\text{rank}(Q) = 1 < 2$$

$\therefore$  The decomposed system is  
not observable.

Stabilizability:

The system is stabilizable if its uncontrollable modes are Lyapunov stable.

The decomposed equation is fully controllable, making it automatically stabilizable.

Detectability:

Eigenvalues of  $A \begin{bmatrix} 0 & 1 \\ 0 & -1 \end{bmatrix}$  are:  $\lambda_1 = 0$   
 $\lambda_2 = -1$

$$\lambda_i = \text{Re} + \text{Im}j$$

Finding eigenvectors:

$$AV_1 = \lambda_1 V_1$$

$$(A - \lambda_1 I) V_1 = 0$$

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

$$V_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$AV_2 = \lambda_2 V_2$$

$$(A - \lambda_2 I) V_2 = 0$$

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

$$V_2 = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

Finding the unobservable mode. ( $Cv=0$  means that mode is unobservable)

$$\lambda_1 = 0, \quad Cv_1 = \begin{bmatrix} 3 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = 3 \neq 0 \quad \therefore \text{Observable}$$

$$\lambda_2 = -1, \quad Cv_2 = \begin{bmatrix} 3 & 3 \end{bmatrix} \begin{bmatrix} -1 \\ 1 \end{bmatrix} = 0 = 0 \quad \therefore \text{Not observable}$$

The system is detectable if its unobservable modes are Lyapunov stable.

$$\lambda_2 = -1 \quad \operatorname{Re}_2 = -1 \quad \left. \right\} \text{unobservable modes}$$

$\therefore \# \lambda_i, \operatorname{Re} < 0$  is satisfied  $\Rightarrow$  The system is asymptotically stable

Hence, the uncontrollable modes are also lyapunov stable.

$\therefore$  The decomposed system is detectable.

### # Exercise 3: Stability

$$\begin{aligned} m\ddot{x} &= -u_1 \sin \theta + \epsilon u_2 \cos \theta \\ m\ddot{y} &= u_1 \cos \theta + \epsilon u_2 \sin \theta - mg \\ J\ddot{\theta} &= u_2 \end{aligned}$$

$$\dot{\theta} = \dot{\theta}$$

$$\ddot{x} = -u_1 \frac{\sin \theta}{m} + u_2 \frac{\epsilon \cos \theta}{m} \quad -\textcircled{1}$$

$$\ddot{y} = u_1 \frac{\cos \theta}{m} + u_2 \frac{\epsilon \sin \theta}{m} - g \quad -\textcircled{2}$$

$$\ddot{\theta} = \frac{u_2}{J} \quad -\textcircled{3}$$

Equilibrium points :  $\bar{x}(t), \bar{y}(t), \bar{\theta}(t) = 0$   
 $\bar{u}_1(t) = mg \quad \bar{u}_2(t) = 0$

State vector:

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} \theta \\ \dot{x} \\ \dot{y} \\ \dot{\theta} \end{bmatrix}$$

$x_1 = \theta$
$x_2 = \dot{x}$
$x_3 = \dot{y}$
$x_4 = \dot{\theta}$

$\therefore$  The equations become  $\Rightarrow$

$$\dot{x}_1 = x_4$$

$$m\dot{x}_2 = -u_1 \sin(x_1) + \epsilon u_2 \cos(x_1)$$

$$m\dot{x}_3 = u_1 \cos(x_1) + \epsilon u_2 \sin(x_1) - mg$$

$$\dot{x}_4 = u_2 / J$$

State,  $z = \begin{bmatrix} \theta \\ \dot{x} \\ \dot{y} \\ \dot{\theta} \end{bmatrix}$

Input,  $u = \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}$

Finding the Jacobian matrix for the above system:

$$J_A = \frac{\partial f}{\partial z} = \begin{bmatrix} 0 & 0 & 0 & 1 \\ -u_1 \cos(x_1)/m & -\varepsilon u_2 \sin(x_1)/m & 0 & 0 \\ -u_1 \sin(x_1)/m + \varepsilon u_2 \cos(x_1)/m & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

Jacobian @ equilibrium point:

$$\left. \begin{array}{l} \ddot{x}(t), \ddot{y}(t), \ddot{\theta}(t) = 0 \\ \ddot{u}_1(t) = mg, \quad \ddot{u}_2(t) = 0 \end{array} \right\} \left. \begin{array}{l} \dot{x}(t), \dot{y}(t), \dot{\theta}(t) = 0 \\ \ddot{u}_1(t) = mg, \quad \ddot{u}_2(t) = 0 \end{array} \right\} \left. \begin{array}{l} x_1 = 0 \\ x_2 = 0 \\ x_3 = 0 \\ x_4 = 0 \end{array} \right\}$$

$$J_A \Big|_{z=z_0} = \begin{bmatrix} 0 & 0 & 0 & 1 \\ -g & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

Doing the same for the input matrix B:

$$J_B = \frac{\partial f}{\partial u} = \begin{bmatrix} 0 & 0 \\ -\sin(x_1)/m & \varepsilon \cos(x_1)/m \\ \cos(x_1)/m & \varepsilon \sin(x_1)/m \\ 0 & 1/J \end{bmatrix}$$

$$J_B \Big|_{z=z_0} = \begin{bmatrix} 0 & 0 \\ 0 & \varepsilon/m \\ 1/m & 0 \\ 0 & 1/J \end{bmatrix}$$

$\therefore$  The linearized system about the equilibrium point is:

$$\begin{bmatrix} \dot{\theta} \\ \ddot{x} \\ \ddot{y} \\ \ddot{\theta} \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & 1 \\ -g & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} \dot{\theta} \\ \ddot{x} \\ \ddot{y} \\ \ddot{\theta} \end{bmatrix} + \begin{bmatrix} 0 & 0 \\ 0 & \epsilon/m \\ 1/m & 0 \\ 0 & I/J \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}$$

**A**                                   **B**

Stability analysis:

Finding eigenvalues:  $|\lambda I - A| = 0$

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

$$\lambda \begin{vmatrix} \lambda & 0 & 0 \\ 0 & \lambda & 0 \\ 0 & 0 & \lambda \end{vmatrix} - (-1) \begin{vmatrix} g & \lambda & 0 \\ 0 & 0 & \lambda \\ 0 & 0 & 0 \end{vmatrix} = 0$$

$$\lambda [\lambda(\lambda^2)] + 1 [g(0) - \lambda(0) 0] = 0$$

$$\lambda^4 = 0$$

$\therefore$  The eigenvalues of the system are  $\lambda_1 = \lambda_2 = \lambda_3 = \lambda_4 = 0$

We see that  $\exists \lambda_i, \operatorname{Re} \lambda_i = 0, m > 0$

indicating that the system is **unstable**.

## # Exercise 4: Lyapunov's direct method

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

$$\text{Given, } V = x_1^2 + x_2^2$$

Lyapunov's Direct Method :

### Theorem

The origin of  $\dot{x} = f(x)$  is stable if

- $V(x) = 0$  if and only if  $x = 0$
- $V(x) > 0$  if and only if  $x \neq 0$
- $\dot{V}(x) = \frac{d}{dt}V(x) = \sum_{i=1}^n \frac{\partial V}{\partial x_i} f_i(x) = \nabla V \cdot f(x) \leq 0$  for all values of  $x \neq 0$ .

Note: for asymptotic stability,  $\dot{V}(x) < 0$  for  $x \neq 0$  is required.

$$1. \quad V(x) = 0 \quad \text{if and only if} \quad x_1 = 0 \quad \& \quad x_2 = 0$$

$\because x_1^2 \text{ and } x_2^2$  are always positive

$$2. \quad V(x) > 0 \quad \& \quad x_1, x_2 \neq 0$$

Asymptotic stability :

$$\dot{x}_1 = ax_1 \quad \dot{V}(x) = 2x_1 \cdot \dot{x}_1 + 2x_2 \cdot \dot{x}_2$$

$$\dot{x}_2 = x_1 - x_2 \quad = 2 [x_1(ax_1) + x_2(x_1 - x_2)]$$

$$= 2 [ax_1^2 + x_1x_2 - x_2^2]$$

$$\dot{V}(x) < 0 \quad \text{for } x_1 \neq 0 \quad \& \quad x_2 \neq 0$$

$$\dot{V}(x) < 0$$

$$2 \left[ \alpha x_1^2 + x_1 x_2 - x_2^2 \right] < 0$$

$$\alpha x_1^2 + x_1 x_2 - x_2^2 < 0$$

$$\alpha x_1^2 < x_2^2 - x_1 x_2$$

$$\alpha x_1^2 < x_2^2 + \frac{1}{4} x_1^2 - \frac{1}{4} x_1^2 - 2 \times \frac{1}{2} x_1 x_2$$

$$\alpha x_1^2 < \left[ x_2 - \frac{1}{2} x_1 \right]^2 - \frac{1}{4} x_1^2$$

$$\alpha x_1^2 < \left[ x_2 - \frac{1}{2} x_1 \right]^2 - \left[ \frac{x_1}{2} \right]^2$$

if  $x_1, x_2 \neq 0$ , the minimum value of RHS occurs when  $x_2 = \frac{x_1}{2}$

The minimum value is  $-\frac{x_1^2}{4}$

If  $\alpha x_1^2$  is less than minimum of RHS, then it must satisfy the inequality for all other values of RHS.

$$\therefore \alpha x_1^2 < -\frac{x_1^2}{4}$$

$$x_1^2 \left( \alpha + \frac{1}{4} \right) < 0 \Rightarrow \alpha + \frac{1}{4} < 0 \Rightarrow \alpha < -\frac{1}{4}$$

$\therefore \alpha < -\frac{1}{4}$ . the above system is asymptotically stable

## # Exercise 5: Stability of non-linear systems

$$\dot{x}_1 = x_2 - x_1 x_2^2$$

$$\dot{x}_2 = -x_1^3$$

(a)  $\dot{x} = \begin{bmatrix} x_2 - x_1 x_2^2 \\ -x_1^3 \end{bmatrix}$

Finding Equilibria:  $\begin{cases} x_2 - x_1 x_2^2 = 0 \\ -x_1^3 = 0 \end{cases} \Rightarrow \begin{array}{l} x_2(1 - x_1 x_2) = 0 \\ x_1 = 0 \\ x_2 = 0. \end{array}$

∴ Equilibrium point:  $x_0 = (0, 0)$

$$\frac{\partial f}{\partial x} = \begin{bmatrix} -x_2^2 & 1 - 2x_1 x_2 \\ -3x_1^2 & 0 \end{bmatrix}$$

$$\left. \frac{\partial f}{\partial x} \right|_{x_0} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$$

The linearized system is:

$$\dot{x} = f(x_0) + \left. \frac{\partial f}{\partial x} \right|_{x_0} \delta_x = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \delta_x$$

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \delta x_1 \\ \delta x_2 \end{bmatrix}$$

$A$

Eigenvalues of  $A$ :  $\lambda_1 = 0, \lambda_2 = 0$

$\operatorname{Re}(\lambda_i) = 0 \Rightarrow \therefore$  The linearized system is ristey

Cannot say if it is stable or unstable using Lyapunov's indirect method. (inconclusive)

$$(b) \quad V(x_1, x_2) = x_1^4 + 2x_2^2$$

$$\begin{aligned}\dot{V}(x_1, x_2) &= 4x_1^3 \cdot \dot{x}_1 + 4x_2 \cdot \dot{x}_2 \\ &= 4x_1^3 [x_2 - x_1 x_2^2] + 4x_2 (-x_1^3) \\ &= \cancel{4x_1^3} x_2 - 4x_1^4 x_2^2 - \cancel{4x_1^3} x_2 = -4x_1^4 x_2^2\end{aligned}$$

condition 1:  $V(x) = 0$  only when  $x_1 = 0, x_2 = 0$   
as  $x_1^4 + 2x_2^2$  is always positive.

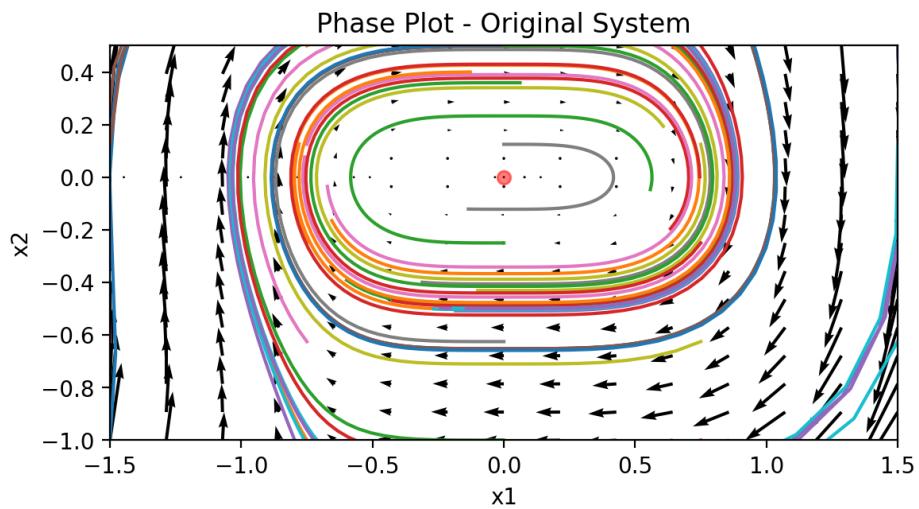
condition 2:  $\dot{V}(x) > 0$  only when  $x_1 \neq 0 \text{ or } x_2 \neq 0$   
as  $x_1^4 = 0$  only when  $x_1 = 0$   
 $2x_2^2 = 0$  only when  $x_2 = 0$

condition 3:  $\dot{V}(x) = -4x_1^4 x_2^2$

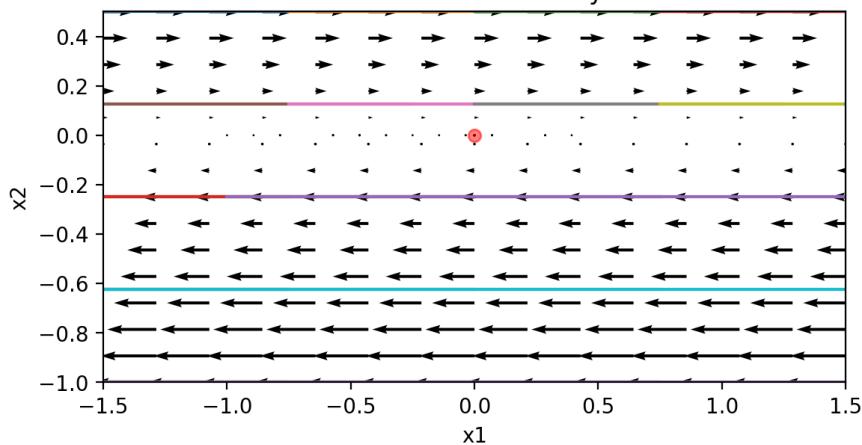
$$\dot{V}(x) < 0 \quad \forall x_1, x_2 \neq 0$$

$\therefore$  The system satisfies the 3 conditions of Lyapunov's direct method.  
the system is **Asymptotically stable.**

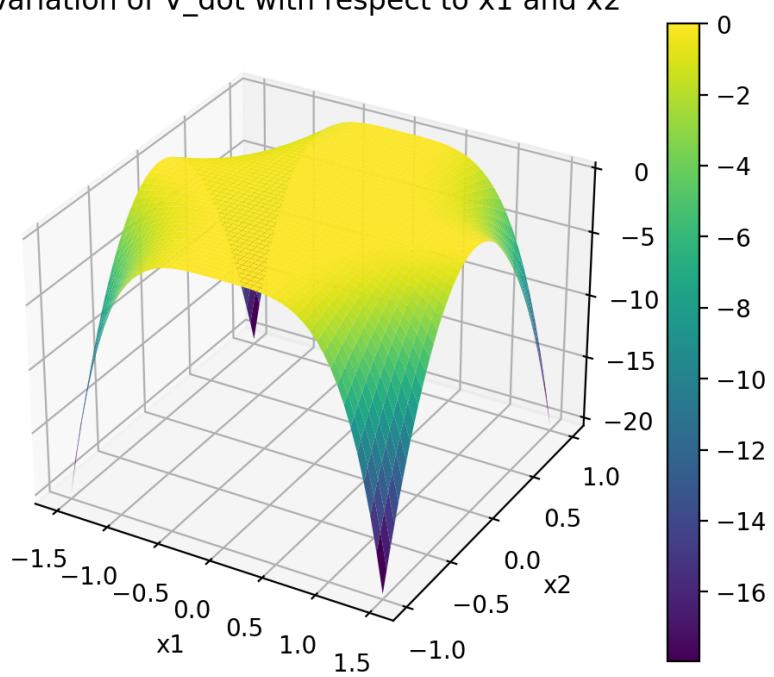
(c)



Phase Plot - Linearized System



(d)

Variation of  $V_{\text{dot}}$  with respect to  $x_1$  and  $x_2$ 

## # Exercise 6: BIBO Stability

(a)

$$x(k+1) = \begin{bmatrix} 1 & 0 \\ -0.5 & 0.5 \end{bmatrix} x(k) + \begin{bmatrix} 1 \\ -1 \end{bmatrix} u(k)$$

$$y(k) = [5 \ 5] x(k)$$

For the discrete system above:

$$G_D(z) = C(zI - A)^{-1}B + D$$

$$= [5 \ 5] \left[ \begin{bmatrix} z & 0 \\ 0 & z \end{bmatrix} - \begin{bmatrix} 1 & 0 \\ -0.5 & 0.5 \end{bmatrix} \right]^{-1} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$= [5 \ 5] \begin{bmatrix} z-1 & 0 \\ 0.5 & z-0.5 \end{bmatrix}^{-1} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$= [5 \ 5] \times \frac{1}{(z-1)(z-0.5)} \times \begin{bmatrix} (z-0.5) & 0 \\ -0.5 & (z-1) \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$= 0$$

Since  $G(s) = 0$ , it means that the input has no change to the output.

Output is always zero (bounded) for any input signal as it does not affect the output.

∴ The system is BIBO stable.

(b)

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

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

Converting to transfer function:

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

$$= \begin{bmatrix} -1 & -1 & 2 \\ 1 & 1 & -1 \end{bmatrix} \begin{bmatrix} s+7 & 2 & -6 \\ -2 & s+3 & 2 \\ 2 & 2 & s-1 \end{bmatrix}^{-1} \begin{bmatrix} 1 & 1 \\ 1 & -1 \\ 1 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} -1 & -1 & 2 \\ 1 & 1 & -1 \end{bmatrix} \left[ \begin{array}{ccc} \frac{s^2+2s-7}{(s+5)(s+1)(s+3)} & \frac{-2}{(s+1)(s+3)} & \frac{2(3s+11)}{(s+5)(s^2+4s+3)} \\ \frac{2}{(s+5)(s+3)} & \frac{1}{(s+3)} & \frac{-2}{(s+5)(s+3)} \\ \frac{-2}{s^2+4s+3} & \frac{-2}{s^2+4s+3} & \frac{s+5}{s^2+4s+3} \end{array} \right]$$

On solving:

$$G(s) = \begin{bmatrix} 0 & 0 \\ \frac{1}{s+3} & 0 \end{bmatrix}$$

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

Poles of  $G(s)$ :  $s = -3$  ( $G_{11}, G_{12}, G_{21}$  are constants and do not contribute any poles)

$\therefore$  Every pole of  $G(s)$  has negative real part.

$\therefore$  The above system is BIBO stable.

## # Exercise 7: BIBO Stability

$$V_C \frac{dT_C}{dt} = f_C(T_{Ci} - T_C) + \beta(T_H - T_C)$$

$$V_H \frac{dT_H}{dt} = f_H(T_{Hi} - T_H) + \beta(T_C - T_H)$$

$$f_C = f_H = 0.1$$

$$y_1 = T_C$$

$$u_1 = T_{Ci}$$

$$\beta = 0.2$$

$$y_2 = T_H$$

$$u_2 = T_{Hi}$$

$$V_H, V_C = 1$$

1.

$$\begin{cases} \frac{dT_C}{dt} = 0.1 [T_{Ci} - T_C] + 0.2 [T_H - T_C] \\ \frac{dT_H}{dt} = 0.1 [T_{Hi} - T_H] + 0.2 [T_C - T_H] \end{cases}$$

$$\begin{cases} \dot{y}_1 = 0.1 [u_1 - y_1] + 0.2 [y_2 - y_1] \\ \dot{y}_2 = 0.1 [u_2 - y_2] + 0.2 [y_1 - y_2] \end{cases}$$

$$\begin{cases} \dot{y}_1 = 0.1 u_1 + 0.2 y_2 - 0.3 y_1 \\ \dot{y}_2 = 0.1 u_2 - 0.3 y_2 + 0.2 y_1 \end{cases}$$

Considering state  $x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$  where  $x_1 = T_C$   $\Rightarrow x_1 = y_1$   
 $x_2 = T_H$   $\Rightarrow x_2 = y_2$

$$\left\{ \begin{array}{l} \dot{x}_1 = -0.3x_1 + 0.2x_2 + 0.1u_1 \\ \dot{x}_2 = 0.2x_1 - 0.3x_2 + 0.1u_2 \end{array} \right.$$

$\therefore$  The state space equation form is as follows:

$$\dot{\mathbf{x}}(t) = \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} -0.3 & 0.2 \\ 0.2 & -0.3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0.1 & 0 \\ 0 & 0.1 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}$$

$$y(t) = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

2. If input is absent,  $\begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$

$$\begin{pmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \end{pmatrix} = \underbrace{\begin{bmatrix} -0.3 & 0.2 \\ 0.2 & -0.3 \end{bmatrix}}_A \begin{pmatrix} x_1(t) \\ x_2(t) \end{pmatrix}$$

Finding eigenvalues of the system:

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

$$\begin{vmatrix} \lambda + 0.3 & -0.2 \\ -0.2 & \lambda + 0.3 \end{vmatrix} = 0 \Rightarrow (\lambda + 0.3)^2 - 0.04 = 0$$

$$\lambda^2 + 0.09 + 0.6\lambda - 0.04 = 0$$

$$\lambda^2 + 0.6\lambda + 0.05 = 0$$

$$(\lambda + 0.5)(\lambda + 0.1) = 0$$

$$\therefore \lambda_1 = -0.5 \quad \& \quad \lambda_2 = -0.1$$

Finding eigenvectors:

$$AV_1 = \lambda_1 V_1$$

$$(A - \lambda_1 I) V_1 = 0 \Rightarrow \begin{bmatrix} 0.2 & 0.2 \\ 0.2 & 0.2 \end{bmatrix} V_1 = 0$$

$$V_1 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$(A - \lambda_2 I) V_2 = 0 \Rightarrow \begin{bmatrix} -0.2 & 0.2 \\ 0.2 & -0.2 \end{bmatrix} V_2 = 0$$

$$V_2 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

General solution:  $\begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} = c_1 \begin{bmatrix} 1 \\ -1 \end{bmatrix} e^{-0.5t} + c_2 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{-0.1t}$

$$\therefore y_1(t) = x_1(t)$$

$$y_2(t) = x_2(t)$$

$\therefore$  The values of  $y_1(t)$  &  $y_2(t)$  are:

$$y_1(t) = c_1 e^{-0.5t} + c_2 e^{-0.1t}$$

$$y_2(t) = -c_1 e^{-0.5t} + c_2 e^{-0.1t}$$

$c_1$  &  $c_2$  are constants that can be determined from the initial condition of the system.

3. To check if the system is BIBO stable, we need to convert it to a transfer function from state space.

$$\dot{x}(t) = \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} -0.3 & 0.2 \\ 0.2 & -0.3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0.1 & 0 \\ 0 & 0.1 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}$$

$A$        $B$

$$y(t) = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$C$

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

$$= \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} s+0.3 & -0.2 \\ -0.2 & s+0.3 \end{bmatrix}^{-1} \begin{bmatrix} 0.1 & 0 \\ 0 & 0.1 \end{bmatrix}$$

$$= \frac{1}{s^2 + 0.6s + 0.05} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} s+0.3 & 0.2 \\ 0.2 & s+0.3 \end{bmatrix} \begin{bmatrix} 0.1 & 0 \\ 0 & 0.1 \end{bmatrix}$$

$$G(s) = \frac{1}{s^2 + 0.6s + 0.05} \begin{bmatrix} 0.1(s+0.3) & 0.02 \\ 0.02 & 0.1(s+0.3) \end{bmatrix}$$

Poles of the above transfer function:  $s^2 + 0.6s + 0.05 = 0$   
 $(s+0.1)(s+0.5) = 0$

$\therefore$  The poles are  $-0.1$  &  $-0.5$

$\therefore$  Every pole of the transfer function has a negative real part,  
the system is **BIBO stable**.

# hw5\_abijunai

October 12, 2024

```
[ ]: import os
import numpy as np
import matplotlib.pyplot as plt
from control.phaseplot import phase_plot
from numpy import pi
from mpl_toolkits.mplot3d import Axes3D
plt.close('all')

[ ]: def org_system(x, t):
    # x1, x2 = X
    dx1dt = x[1] - x[0] * x[1]**2
    dx2dt = -x[0]**3
    return [dx1dt, dx2dt]

def lin_system(x,t):
    dx1dt = x[1]
    dx2dt = x[0]*0
    return np.array([dx1dt, dx2dt])

[ ]: plt.figure(dpi = 200)
plt.clf()
plt.gca().set_aspect('equal', adjustable='box')
plt.axis([-1.5, 1.5, -1, 0.5])

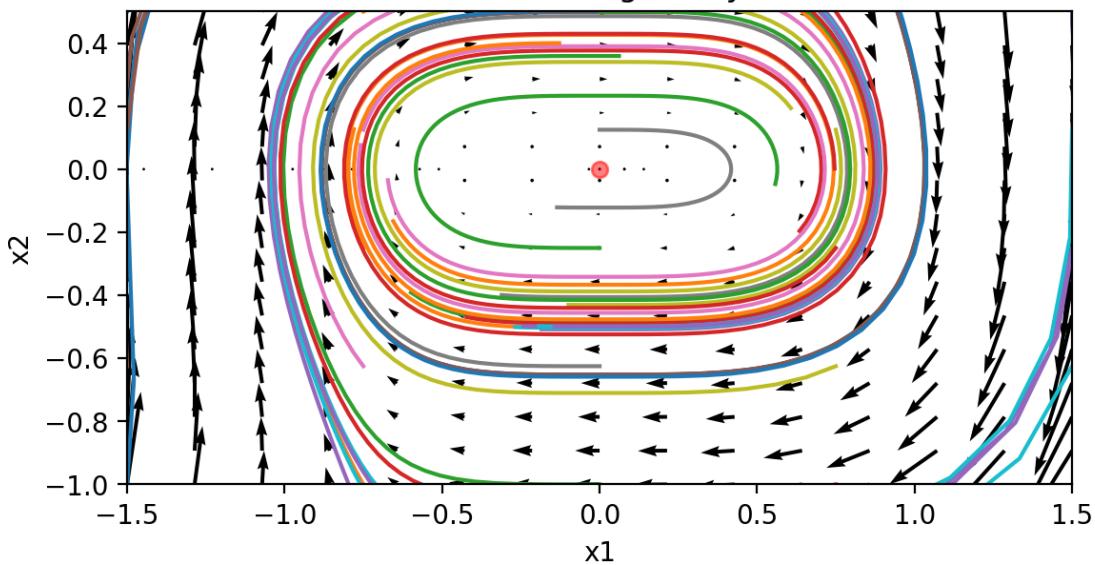
plt.title('Phase Plot - Original System')
x1 = np.linspace(-1.5, 1.5, 5)
x2 = np.linspace(-1, 0.5, 5)
X1, X2 = np.meshgrid(x1, x2)
X0 = np.hstack((X1.reshape(-1,1), X2.reshape(-1,1)))

# Outer trajectories
phase_plot(org_system, X0 = X0, X = (-1.5, 1.5, 15), Y = (-1, 0.5, 15), T = 10)

plt.scatter(x = [0], y = [0], c='r', alpha=0.5)

plt.show()
```

Phase Plot - Original System



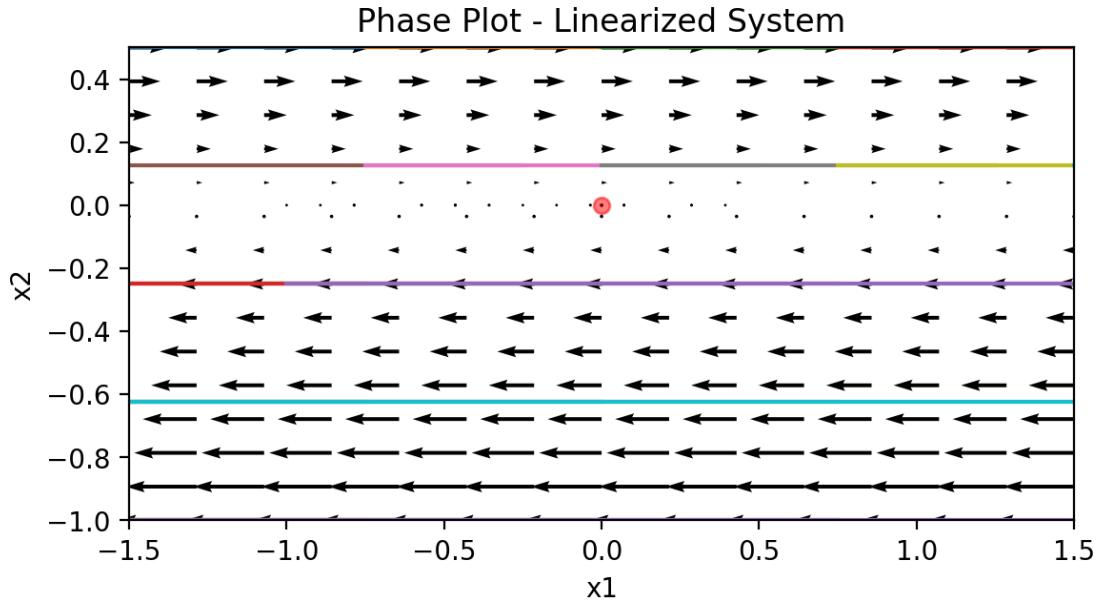
```
[ ]: plt.figure(dpi = 200)
plt.clf()
plt.gca().set_aspect('equal', adjustable='box')
plt.axis([-1.5, 1.5, -1, 0.5])

plt.title('Phase Plot - Linearized System')
x1 = np.linspace(-1.5, 1.5, 5)
x2 = np.linspace(-1, 0.5, 5)
X1, X2 = np.meshgrid(x1, x2)
X0 = np.hstack((X1.reshape(-1,1), X2.reshape(-1,1)))

# Outer trajectories
phase_plot(lin_system, X0 = X0, X = (-1.5, 1.5, 15), Y = (-1, 0.5, 15), T = 10)

plt.scatter(x = [0], y = [0], c='r', alpha=0.5)

plt.show()
```



```
[ ]: def V(x1, x2):
    return x1**4 + 2*x2**2

def V_dot(x1, x2):
    dV_dx1 = 4 * x1**3
    dV_dx2 = 4 * x2
    x1_dot = x2 - x1 * (x2**2)
    x2_dot = -x1**3
    return dV_dx1 * x1_dot + dV_dx2 * x2_dot

x1 = np.linspace(-1.5, 1.5, 50)
x2 = np.linspace(-1, 1, 50)
X1, X2 = np.meshgrid(x1, x2)

vdot = V_dot(X1, X2)

# Create the 3D plot
fig = plt.figure(dpi=200)
ax = fig.add_subplot(111, projection='3d')

# Plot the surface
surf = ax.plot_surface(X1, X2, vdot, cmap='viridis')

# Set labels and title
ax.set_xlabel('x1')
```

```

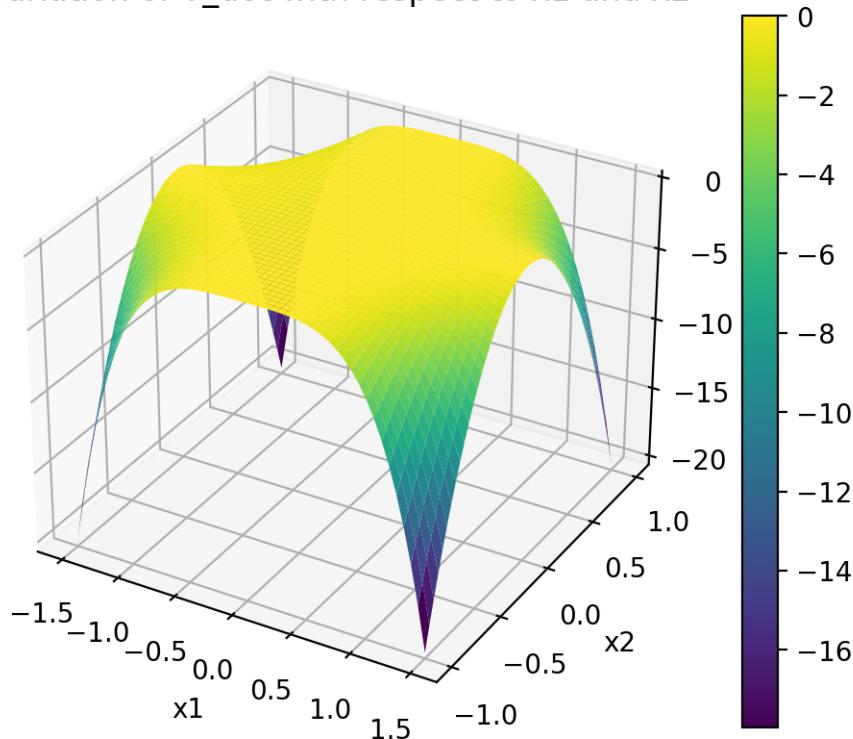
ax.set_ylabel('x2')
ax.set_zlabel('V_dot')
ax.set_title('Variation of V_dot with respect to x1 and x2')

# Add a color bar
fig.colorbar(surf)

# Show the plot
plt.show()

```

Variation of V\_dot with respect to x1 and x2



```

[ ]: #####
NOTEBOOK_NAME = "hw5_abijunai.ipynb"
!apt-get install -y pandoc > /dev/null 2>&1
!sudo apt-get install texlive-xetex texlive-fonts-recommended
  ↵texlive-plain-generic > /dev/null 2>&1
!pip install nbconvert > /dev/null 2>&1

from google.colab import drive, files
drive.mount('/content/drive')
notebook_path = rf'/content/drive/MyDrive/Colab Notebooks/{NOTEBOOK_NAME}'

```

```
import os
if os.path.exists(notebook_path):
    !jupyter nbconvert --to pdf "{notebook_path}" > /dev/null 2>&1
    pdf_filename = notebook_path.replace('.ipynb', '.pdf')
    files.download(pdf_filename)
else:
    print("Notebook not found. Please check the path.")
#####
```