

24677

PS-5

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1.)

a) let: $\dot{x}_k = Ax_k + Bu_k$ | $A = \begin{bmatrix} 1 & 0 \\ -\frac{1}{2} & \frac{1}{2} \end{bmatrix}$, $B = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$

$\lambda_1 = 1, \lambda_2 = \frac{1}{2}$, $\therefore |\lambda_i| \leq 1$, and $M = I = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ \therefore algebraic multiplicity = geometric multiplicity \rightarrow equal $\Rightarrow \lambda_1$ is not defective

\therefore SYSTEM IS LYAPUNOV STABLE BUT NOT ASYMPTOTIC STABLE

$\therefore |\lambda_j| \neq 1 \forall \lambda_j$

b.) let: $\dot{x} = Ax + Bu$ | $A = \begin{bmatrix} -7 & -2 & 6 \\ 2 & -3 & -2 \\ -2 & -2 & 1 \end{bmatrix}$, $B = \begin{bmatrix} 1 & 1 \\ 1 & -1 \\ 0 & 0 \end{bmatrix}$

let: $|A - \lambda I| = 0 \Rightarrow \begin{vmatrix} -\lambda & -2 & 6 \\ 2 & -\lambda & -2 \\ -2 & -2 & 1-\lambda \end{vmatrix} = (-\lambda) \begin{vmatrix} -3-\lambda & -2 \\ -2 & 1-\lambda \end{vmatrix} + 2 \begin{vmatrix} 2 & -2 \\ -2 & 1-\lambda \end{vmatrix} + 6 \begin{vmatrix} 2 & -3-\lambda \\ -2 & -2 \end{vmatrix} = 0$

 $\Rightarrow -\lambda^3 - 9\lambda^2 - 23\lambda - 15 = 0$
 $\Rightarrow \lambda_1 = -1, \lambda_2 = -5, \lambda_3 = -3$
 $\therefore \operatorname{Re}(\lambda_i) < 0, \text{ SYSTEM IS ASYMPTOTIC STABLE} \rightarrow \text{SYSTEM IS THEREFORE ALSO LYAPUNOV STABLE}$

2.) Let: $\begin{cases} \dot{x} = Ax + Bu \\ y = Cx \end{cases}$ | $A = \begin{bmatrix} 0 & -1 & 1 \\ 1 & -2 & 1 \\ 0 & 1 & -1 \end{bmatrix}$, $B = \begin{bmatrix} 1 & 0 \\ 1 & 2 \\ 1 & 1 \end{bmatrix}$, $C = \begin{bmatrix} 1 & 1 \end{bmatrix}$

Let: $\begin{cases} \hat{A} = M^{-1}AM, \hat{B} = M^{-1}B \\ \hat{C} = CM, \hat{D} = D \end{cases}$

Let: $M = M_c$ is • first $n_c = \text{rank}(P_c)$ columns are lin. indep columns of P_c
• last $(n - n_c)$ columns are arbitrary | $\text{rank}(M) = n$

$$P_c = [A^0 B_0 \mid A^1 B_1 \cdots A^{n-1} B_n]$$

Let: $P_c = [B \mid AB/A^2 B]$

$$\therefore A^2 = AA = \begin{bmatrix} -1 & 3 & -2 \\ -2 & 4 & -2 \\ 1 & -3 & 2 \end{bmatrix}, P_c = \begin{bmatrix} 1 & 0 & 0 & 1 & 0 & -1 \\ 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 2 & 0 & -1 & 0 & 1 \end{bmatrix}, \therefore \text{rank}(P_c) = 2, M = \begin{bmatrix} 1 & 0 & 1 \\ 1 & 1 & 3 \\ 1 & 2 & 2 \end{bmatrix}$$

$$\therefore M^{-1} = \frac{1}{\text{adj}(M)} = \frac{1}{\begin{vmatrix} 1 & 3 \\ 2 & 2 \end{vmatrix} + \begin{vmatrix} 0 & 1 \\ 1 & 3 \end{vmatrix}} \begin{bmatrix} 1 & 3 & -1 & 0 & 1 \\ -1 & 3 & 0 & 1 & 3 \\ 1 & 2 & -1 & 0 & 1 \\ 1 & 1 & -1 & 1 & 3 \\ 1 & 0 & 1 & 0 & 1 \end{bmatrix}$$

$$\therefore \hat{A} = \frac{-1}{3} \begin{bmatrix} -1 & 2 & -1 \\ 1 & -2 & 1 \\ 1 & -2 & 1 \end{bmatrix} \begin{bmatrix} 0 & -1 & 1 \\ 1 & -2 & 1 \\ 1 & 1 & 3 \end{bmatrix} \begin{bmatrix} 1 & 0 & 1 \\ 0 & -1 & 2 \\ 0 & 0 & -2 \end{bmatrix} = \begin{bmatrix} 0 & 1 & 1 \\ 0 & -1 & 2 \\ 0 & 0 & -2 \end{bmatrix}, \hat{B} = M^{-1}B = \frac{-1}{3} \begin{bmatrix} -4 & 2 & -1 \\ 1 & 1 & -2 \\ 1 & -2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}, \hat{C} = CM = \begin{bmatrix} 1 & 1 \\ 1 & 1 \\ 1 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 1 \\ 1 & 1 & 3 \\ 1 & 2 & 2 \end{bmatrix} = [3 \ 3 \ 6]$$

Let: $\hat{A} = \begin{bmatrix} A_c & A_{c\bar{c}} \\ \emptyset & A_{\bar{c}\bar{c}} \end{bmatrix}$, $\hat{B} = \begin{bmatrix} B_c \\ \emptyset_{n-n_c} \end{bmatrix}$, $\hat{C} = \begin{bmatrix} C_c & C_{\bar{c}} \end{bmatrix}$ | $A_c \in \mathbb{R}^{n_c \times n_c}$, $A_{\bar{c}} \in \mathbb{R}^{(n-n_c) \times (n-n_c)}$, etc.

$\therefore A_c = \begin{bmatrix} 0 & 1 \\ 0 & -1 \end{bmatrix}$, $B_c = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$, $C_c = [3 \ 3]$

CONTROLLABLE FORM: $\begin{cases} \dot{\hat{x}}_c = A_c \hat{x}_c + B_c u, \\ \dot{\hat{y}} = C_c \hat{x}_c + D u \end{cases}$ | $A_c = \begin{bmatrix} 0 & 1 \\ 0 & -1 \end{bmatrix}$, $B_c = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$, $C_c = [3 \ 3]$

check observability: $Q_c = \begin{bmatrix} C_c \\ C_c A_c \end{bmatrix} = \begin{bmatrix} 3 & 3 \\ 0 & 0 \end{bmatrix}$

$\therefore n_c \leq \text{rank}(Q_c) = 1 < n_c$, controllable form is not observable

check stabilizability: The controllable reduced form by definition has no uncontrollable modes, therefore all (zero) of its uncontrollable modes (which there are none) are Lyapunov stable, therefore the controllable canonical form is stabilizable

check detectability: First, find the undetectable modes (given by $A_{c\bar{c}}$)

Let: M_{co}^{-1} be first n_{co} rows of Q_c + arbitrary rows so that $\text{rank}(M_{co}^{-1}) = n_c$,

$$\Rightarrow M_{co}^{-1} = \begin{bmatrix} 3 & 3 \\ 1 & 2 \end{bmatrix}, \therefore M_{co} = (M_{co}^{-1})^{-1} = \frac{1}{\text{adj}(M_{co}^{-1})} \text{adj}(M_{co}^{-1}) = \frac{1}{3} \begin{bmatrix} 2 & -3 \\ -1 & 3 \end{bmatrix}$$

$$\text{Let: } \hat{A}_c = M_{co}^{-1} A_c M_{co} = \frac{1}{3} \begin{bmatrix} 3 & 3 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 0 & -1 \end{bmatrix} \begin{bmatrix} 2 & -3 \\ -1 & 3 \end{bmatrix} = \frac{1}{3} \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

Let: $\hat{A}_c = \begin{bmatrix} A_{c\bar{c}} & \emptyset \\ \emptyset & A_{\bar{c}\bar{c}} \end{bmatrix} \therefore A_{c\bar{c}} = [-1] \leftarrow \text{represents uncontrollable modes of } A_c$

$\therefore \lambda_{c\bar{c}} = -1$, $\text{Re}(\lambda_{c\bar{c}}) = -1 < 0 \rightarrow \lambda_{c\bar{c}}$ is asymptotically stable (\therefore also Lyapunov stable),

\therefore All undetectable modes of A_c are Lyapunov stable, the controllable canonical form is detectable

Q3.)

$$\text{let: } \dot{\mathbf{z}} = \begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \end{bmatrix}, \mathbf{u} = \begin{bmatrix} v_1 \\ v_2 \end{bmatrix}, \begin{cases} m\ddot{x} = -v_1 \sin(\theta) + v_2 \cos(\theta), \\ m\ddot{y} = v_1 \cos(\theta) + v_2 \sin(\theta) - mg, \\ J\ddot{\theta} = v_2, \end{cases}$$

$$\dot{\mathbf{z}} = \begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \end{bmatrix} = \begin{bmatrix} \dot{x} \\ v_1 \sin(\theta) + \frac{v_2}{m} \cos(\theta) \\ v_1 \cos(\theta) + \frac{v_2}{m} \sin(\theta) - g \end{bmatrix}$$

LINARIZE ABOUT: $\tilde{\mathbf{z}}(t) = \mathbf{0}(t) = \dot{\mathbf{z}}(0) = \mathbf{0}, \tilde{\mathbf{u}} = mg, \tilde{v}_2 = 0, \therefore \dot{\theta}(t) = 0$

$$\text{let: } \tilde{\mathbf{z}} = \begin{bmatrix} \tilde{x} \\ \tilde{y} \\ \tilde{\theta} \end{bmatrix}, \tilde{\mathbf{u}} = \begin{bmatrix} \tilde{v}_1 \\ \tilde{v}_2 \end{bmatrix}, \delta_z = \begin{bmatrix} \dot{\delta}_\theta \\ \dot{\delta}_x \\ \dot{\delta}_y \\ \dot{\delta}_\theta \end{bmatrix} = \begin{bmatrix} \dot{\delta}_\theta \\ \dot{x} - \tilde{x} \\ \dot{y} - \tilde{y} \\ \dot{\theta} - \tilde{\theta} \end{bmatrix} = \begin{bmatrix} \dot{\delta}_\theta \\ \mathbf{0} \\ \mathbf{0} \\ \mathbf{0} \end{bmatrix}, \delta_u = \begin{bmatrix} \delta_{v_1} \\ \delta_{v_2} \end{bmatrix} = \begin{bmatrix} v_1 - \tilde{v}_1 \\ v_2 - \tilde{v}_2 \end{bmatrix} = \begin{bmatrix} v_1 - mg \\ 0 \end{bmatrix}$$

$$\text{let: } A = \left[\frac{\partial f}{\partial z} \right]_{(\tilde{z}, \tilde{u})} = \left[\begin{array}{cccc} \frac{\partial f}{\partial \tilde{x}} & \frac{\partial f}{\partial \tilde{y}} & \frac{\partial f}{\partial \tilde{\theta}} & \frac{\partial f}{\partial \delta_x} \\ \frac{\partial f}{\partial \tilde{x}} & \frac{\partial f}{\partial \tilde{y}} & \frac{\partial f}{\partial \tilde{\theta}} & \frac{\partial f}{\partial \delta_y} \\ \frac{\partial f}{\partial \tilde{x}} & \frac{\partial f}{\partial \tilde{y}} & \frac{\partial f}{\partial \tilde{\theta}} & \frac{\partial f}{\partial \delta_\theta} \\ \frac{\partial f}{\partial \tilde{x}} & \frac{\partial f}{\partial \tilde{y}} & \frac{\partial f}{\partial \tilde{\theta}} & \frac{\partial f}{\partial \delta_\theta} \end{array} \right]_{(\tilde{z}, \tilde{u})} = \left[\begin{array}{cc|cc|cc|c} 0 & \frac{v_2}{m} \sin(\theta) - \frac{v_1}{m} \cos(\theta) & 0 & 0 & 0 & 1 \\ -v_1/m \cos(\theta) - v_2/m \sin(\theta) & 0 & 0 & 0 & 0 & 0 \\ 0 & -v_1/m \sin(\theta) + v_2/m \cos(\theta) & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{array} \right]_{(\tilde{z}, \tilde{u})} = \left[\begin{array}{cccc} 0 & 0 & 0 & 1 \\ -9 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right],$$

$$\text{let: } B = \left[\frac{\partial f}{\partial u} \right]_{(\tilde{z}, \tilde{u})} = \left[\begin{array}{cc} \frac{\partial f}{\partial v_1} & \frac{\partial f}{\partial v_2} \\ \frac{\partial f}{\partial v_1} & \frac{\partial f}{\partial v_2} \\ \frac{\partial f}{\partial v_1} & \frac{\partial f}{\partial v_2} \\ \frac{\partial f}{\partial v_1} & \frac{\partial f}{\partial v_2} \end{array} \right]_{(\tilde{z}, \tilde{u})} = \left[\begin{array}{cc} 0 & 0 \\ -\frac{1}{m} \sin(\theta) & \frac{v_1}{m} \cos(\theta) \\ \frac{v_1}{m} \cos(\theta) & \frac{v_1}{m} \sin(\theta) \\ 0 & \frac{1}{J} \end{array} \right]_{(\tilde{z}, \tilde{u})} = \left[\begin{array}{cc} 0 & 0 \\ 0 & \frac{1}{m} \\ \frac{1}{m} & 0 \\ 0 & \frac{1}{J} \end{array} \right]$$

$$\text{LINEARIZED (LTI) MODEL: } \dot{\delta}_z = A\delta_z + B\delta_u \Rightarrow \begin{bmatrix} \dot{\delta}_\theta \\ \dot{\delta}_x \\ \dot{\delta}_y \\ \dot{\delta}_\theta \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & 1 \\ -9 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} \theta \\ x \\ y \\ \theta \end{bmatrix} + \begin{bmatrix} 0 & 0 \\ 0 & \frac{1}{m} \\ 0 & 0 \\ 0 & \frac{1}{J} \end{bmatrix} \begin{bmatrix} v_1 - mg \\ v_2 \end{bmatrix}$$

DETERMINE LTI MODEL STABILITY ABOUT $\{\tilde{z}, \tilde{u}\}$:

$$\text{let: } |A - \lambda I| = 0 \Rightarrow -\lambda \begin{vmatrix} -\lambda & 0 & 0 \\ 0 & -\lambda & 0 \\ 0 & 0 & -\lambda \end{vmatrix} - 0 + 0 - 1 \begin{vmatrix} 0 & 0 & -\lambda \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{vmatrix} = -\lambda(-\lambda^3) - (0 - 0 + 0) = \lambda^4$$

$\therefore \lambda_1 = 0, M_1 = 4 \leftarrow \text{algebraic multiplicity}$

FIND REAL EIGENVECTORS:

$$\text{let: } \phi = (\lambda - \lambda_1 I)v_i = \begin{bmatrix} -0 & 0 & 0 & 1 \\ -9 & -0 & 0 & 0 \\ 0 & 0 & -0 & 0 \\ 0 & 0 & 0 & -0 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} \Rightarrow \begin{cases} d = 0, \\ ga = 0, \\ ab = 0, \\ bc = 0 \end{cases} \Rightarrow \text{any } v_i = \begin{bmatrix} 0 \\ b \\ 0 \\ 0 \end{bmatrix} \text{ for any } \{b, c\} \text{ is a valid eigenvector}$$

... linearly independent eigenvectors of λ_1 will be, for example: $v_1 = \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}, v_2 = \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix}$

$$q_1 = 2$$

$\therefore \text{Re}(\lambda_1) \geq 0$, the model is not asymptotically stable

$q_1 < M_1$, λ_1 is degenerate; so, the model is not Lyapunov stable

the LTI model is not stable around the given equilibrium solution

(NOTE: if we were trying to use this LTI model to assess the stability of the original nonlinear model about the given equilibrium, we wouldn't have even considered the degeneracy of λ_1 and, instead, immediately considered it unstable b/c $\text{Re}(\lambda_1) = 0$ as per Lyapunov's Indirect Method.)

Q4.) Let: $\dot{x} = \begin{bmatrix} a & 0 \\ 1 & -1 \end{bmatrix}x$, $x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$, $V(x) = x_1^2 + x_2^2$

• (Clearly, $V(x) > 0 \forall x \neq 0 \therefore V(x)$ is positive definite

SOLVE CONDITION FOR $\dot{V}(x)$ NEGATIVE DEFINITENESS (CRITERION FOR ASYMPTOTIC STABILITY):

Let: $0 > \dot{V}(x) = 2x_1\dot{x}_1 + 2x_2\dot{x}_2$

$$\Rightarrow 0 > 2x_1(x_1) + 2x_2(x_1 - x_2) = 2ax_1^2 + 2x_1x_2 - 2x_2^2$$

$$\Rightarrow 0 > 2ax^2 + 2x_2 - 2 \quad | x = \frac{x_1}{x_2}$$

$$\Rightarrow 1 - \gamma > ax^2$$

$$\Rightarrow a < \frac{1}{x^2} - \frac{1}{\gamma} = \left(\frac{x_2}{x_1}\right)^2 - \frac{x_2}{x_1} = \frac{x_2^2 - x_1x_2}{x_1^2}$$

$\therefore \dot{V}(x)$ is negative definite (and therefore the model is asymptotically stable about the origin) when $a < \frac{x_2^2 - x_1x_2}{x_1^2} \quad \forall \{x_1, x_2\} \neq 0$.

Q5.) Let: $\begin{cases} \dot{x}_1 = x_2 - x_1 x_2^2, \\ \dot{x}_2 = -x_1^3 \end{cases} \Rightarrow \dot{x} = f(x) = \begin{bmatrix} x_2 - x_1 x_2^2 \\ -x_1^3 \end{bmatrix} \mid x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$

a.) FIND EQUILIBRIUM:

$$\text{let: } \bar{x} = f(\bar{x}) = 0 \Rightarrow \begin{cases} \bar{x}_2 - \bar{x}_1 \bar{x}_2^2 = 0, \\ -\bar{x}_1^3 = 0 \end{cases} \Rightarrow \begin{cases} \bar{x}_1 = 0, \\ \bar{x}_2 = 0 \end{cases} \Rightarrow \bar{x} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\text{let: } A = \left[\frac{\partial f_i}{\partial x_j} \right]_{\bar{x}} = \left[\begin{array}{cc} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} \end{array} \right]_{\bar{x}} = \left[\begin{array}{cc} -x_2^2 & 1-2x_1 \\ -3x_1^2 & 0 \end{array} \right]_{\bar{x}} = \left[\begin{array}{cc} 0 & 1 \\ 0 & 0 \end{array} \right]$$

$$\therefore \text{LINEARIZED SYSTEM: } \dot{\delta}_x = A \delta_x = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \delta_x \mid \delta_x = x - \bar{x} = x$$

$$\therefore \text{eig}(\lambda) \rightarrow \lambda_1 = 0, m_1 = 2$$

$\therefore \text{Re}(\lambda_1) = 0$, The system should not be considered stable according to Lyapunov's indirect method.

b.) Let: $V(x) = x_1^4 + 2x_2^2$

• Clearly, $V(x) > 0 \quad \forall x \neq 0 \quad \therefore V(x)$ is positive definite.

CHECK $\dot{V}(x)$ NEGATIVE DEFINITENESS:

$$\text{let: } 0 \leq \dot{V}(x) = 4x_1^3 \dot{x}_1 + 4x_2 \dot{x}_2 = 4x_1^3(x_2 - x_1 x_2^2) + 4x_2(-x_1^3) \\ = 4x_1^3 x_2 - 4x_1^4 x_2^2 - 4x_1^3 x_2 \\ 0 \leq -4x_1^4 x_2^2$$

\therefore it is clear that $\dot{V}(x) = -4x_1^4 x_2^2 < 0 \quad \forall x \neq 0$

$\therefore V(x)$ is positive definite and $\dot{V}(x)$ is negative definite.

\therefore the model is asymptotically stable about the origin according to Lyapunov's direct method.

c.-d.) [SEE NEXT PAGE]

q5

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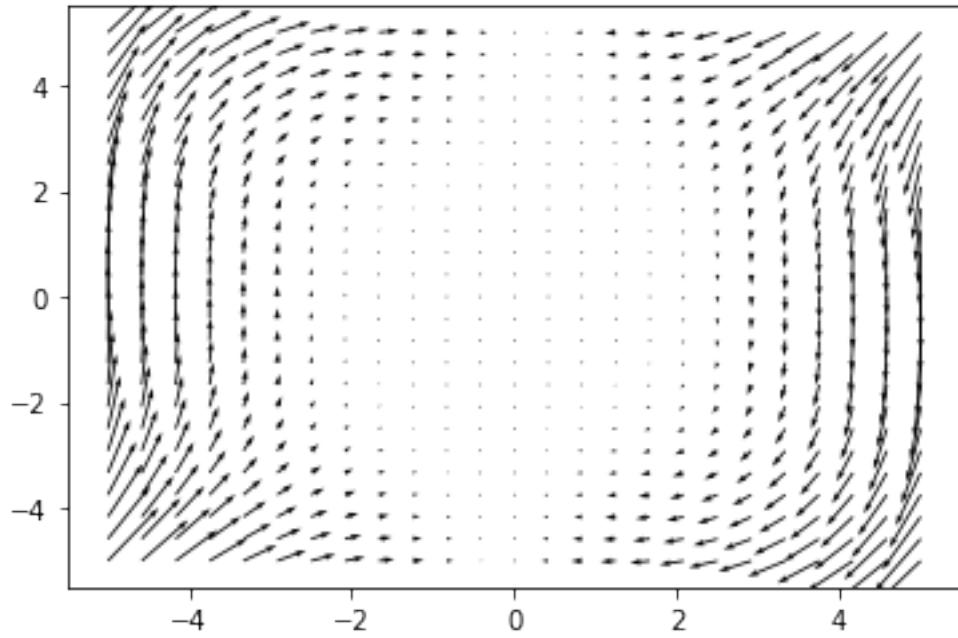
```
[63]: import numpy as np
import matplotlib.pyplot as plt
from mpl_toolkits.mplot3d import Axes3D
from matplotlib import cm
```

```
[50]: def sys(X):
    x1, x2 = X
    return [x2 - x1*np.power(x2,2), -np.power(x1,3)]
def linsys(X):
    x1, x2 = X
    return [-x2, 0]
```

```
[51]: def phase_portrait2D(f, xmin=-5, xmax=5, ymin=-5, ymax=5, num=25):
    x1 = np.linspace(xmin,xmax, num=num)
    x2 = np.linspace(ymin,ymax, num=num)
    x1, x2 = np.meshgrid(x1,x2)
    u, v = np.zeros(x1.shape), np.zeros(x2.shape)
    Nr, Nc = x1.shape
    for r in range(Nr):
        for c in range(Nc):
            x = x1[r,c]
            y = x2[r,c]
            xd = f([x,y])
            u[r,c] = xd[0]
            v[r,c] = xd[1]
    Q = plt.quiver(x1,x2, u,v)
```

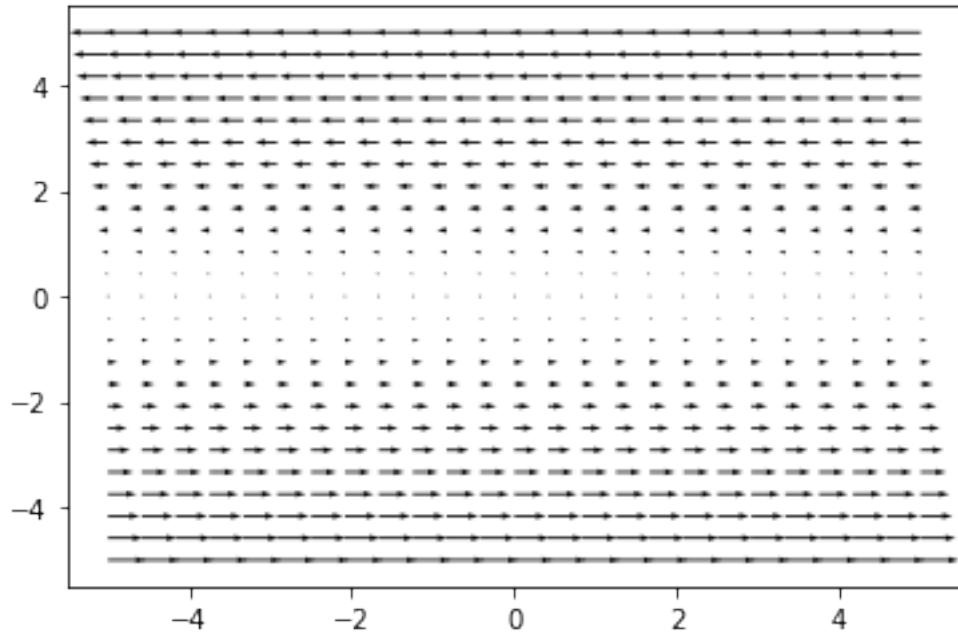
0.1 c.i. Original System Phase Portrait Plot

```
[52]: phase_portrait2D(sys)
```



0.2 c.ii. Linearized System Phase Portrait Plot

```
[111]: phase_portrait2D(linsys)
```



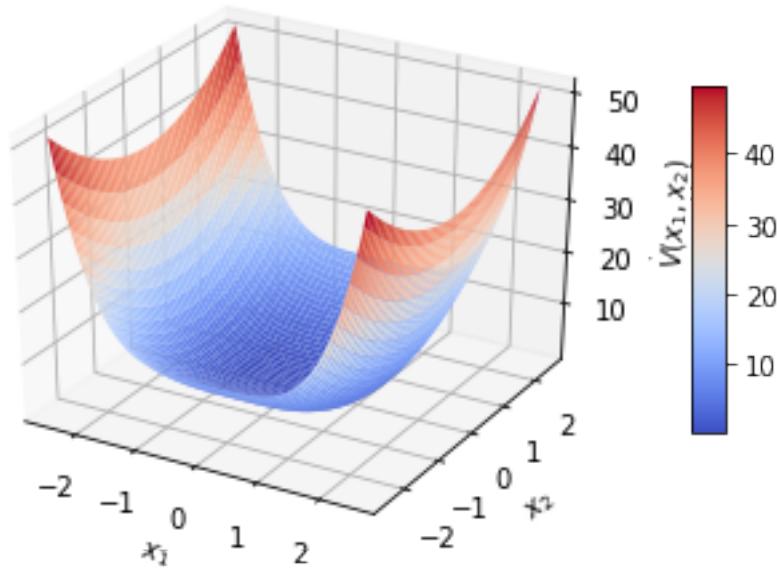
0.3 d. 3D Plot of Lyapunov Function Temporal Derivative

```
[123]: fig = plt.figure()
ax = fig.gca(projection='3d')

x1 = np.linspace(-2.5,2.5, num=100)
x2 = np.linspace(-2.5,2.5, num=100)
x1, x2 = np.meshgrid(x1,x2)
Vd = x1**4 + 2*x2**2

surf = ax.plot_surface(x1,x2,Vd, cmap=cm.coolwarm, linewidth=2, antialiased=True)
ax.set_xlabel(r'$x_1$')
ax.set_ylabel(r'$x_2$')
ax.set_zlabel(r'$\dot{V}(x_1,x_2)$')

# Add a color bar which maps values to colors.
fig.colorbar(surf, shrink=0.6, aspect=10)
plt.show()
```



```
[ ]:
```

Q6.)

a.) let: $\begin{cases} \dot{X}_k = AX_k + BU_k, \\ Y_k = CX_k + DU_k \end{cases} \quad | \quad A = \begin{bmatrix} 1 & 0 \\ -\frac{1}{2} & \frac{1}{2} \end{bmatrix}, B = \begin{bmatrix} 1 \\ -1 \end{bmatrix}, C = \begin{bmatrix} 5 & 5 \end{bmatrix}, D = \emptyset$

$$\begin{aligned} \text{let: } G(z) &= C(zI - A)^{-1}B + D \\ \Rightarrow G(z) &= \begin{bmatrix} z-1 & 0 \\ \frac{1}{2} & z-\frac{1}{2} \end{bmatrix}^{-1} \begin{bmatrix} 1 \\ -1 \end{bmatrix} \\ &= \begin{bmatrix} 5 & 5 \end{bmatrix} \left(\frac{2}{(2z-1)(z-1)} \right) \begin{bmatrix} z-\frac{1}{2} & 0 \\ -\frac{1}{2} & z-1 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix} \\ &= \begin{bmatrix} 5 & 5 \end{bmatrix} \left(\frac{2}{(2z-1)(z-1)} \right) \begin{bmatrix} z-\frac{1}{2} \\ \frac{1}{2}-z \end{bmatrix} \\ &= \left(\frac{2}{(2z-1)(z-1)} \right) [\emptyset] \end{aligned}$$

$\Rightarrow G(z) = \emptyset$
 $\because G(z) = \frac{Y(z)}{U(z)}$, the output $(Y(z))$ will be bounded for any input $(U(z))$, including finite bounded inputs,
because the output will always be \emptyset .

b.) let: $\begin{cases} \dot{X} = AX + BU, \\ Y = CX + DU \end{cases} \quad | \quad A = \begin{bmatrix} -7 & -2 & 6 \\ 2 & -3 & -2 \\ -2 & -2 & 1 \end{bmatrix}, B = \begin{bmatrix} 1 & 1 \\ 1 & -1 \\ 1 & 0 \end{bmatrix}, C = \begin{bmatrix} -1 & -1 & 2 \\ 1 & 1 & -1 \end{bmatrix}, D = \emptyset$

$$\begin{aligned} \text{let: } G(s) &= C(sI - A)^{-1}B + D \\ &= \begin{bmatrix} -1 & -1 & 2 \\ 1 & 1 & -1 \end{bmatrix} \begin{bmatrix} s+7 & 2 & -6 \\ -2 & s+3 & 2 \\ 2 & 2 & s+1 \end{bmatrix}^{-1} \begin{bmatrix} 1 & 1 \\ 1 & -1 \\ 1 & 0 \end{bmatrix} \end{aligned}$$

$\therefore (sI - A)^{-1} = \frac{1}{|sI - A|} \text{adj}(sI - A)$, $|sI - A|$ will contain set of all possible poles of $G(s)$ (though, some may be canceled in $G(s)$)
if the real part of all roots of $|sI - A|$ are < 0 , the real part of all poles will necessarily be < 0 .

$$\begin{aligned} |sI - A| &= 2 \left| \begin{smallmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{1}{2} & -\frac{1}{2} \end{smallmatrix} \right| - 2 \left| \begin{smallmatrix} s+7 & -6 \\ -2 & s+3 \end{smallmatrix} \right| + (s-1) \left| \begin{smallmatrix} s+7 & 2 \\ 2 & s+1 \end{smallmatrix} \right| \\ &= 2(4+6s+18) - 2(-2s+14-12) + (s-1)((s+7)(s+3)+4) \\ &= 12s^2 + 44s + 4 - 4s^2 - 12s + 25 \\ &= s^3 + 9s^2 + 23s + 15 \end{aligned}$$

$$= (s+1)(s+3)(s+5) \Rightarrow s_0 = -1, s_1 = -3, s_2 = -5$$

$\therefore \text{Re of all roots of } |sI - A| \text{ are } < 0, \text{ Re of all poles of } G(s) \text{ are } < 0,$
 $\therefore \text{System is BiBO stable}$

Q7.)

a.)

$$X = \begin{bmatrix} T_C \\ T_H \end{bmatrix} = AX + BU, \quad A = \begin{bmatrix} -\frac{f_C-B}{V_C} & \frac{B}{V_C} \\ \frac{-B}{V_H} & \frac{B-f_H}{V_H} \end{bmatrix}, \quad B = \begin{bmatrix} f_C & 0 \\ 0 & f_H \end{bmatrix}, \quad X = \begin{bmatrix} T_C \\ T_H \end{bmatrix}, \quad U = \begin{bmatrix} V_1 \\ V_2 \end{bmatrix} = \begin{bmatrix} T_{C_1} \\ T_{H_1} \end{bmatrix},$$

$$Y = \begin{bmatrix} Y_1 \\ Y_2 \end{bmatrix} = \begin{bmatrix} T_C \\ T_H \end{bmatrix} = CX + DU \quad C = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad D = 0, \quad f_C = f_H = 0.1 \text{ m}^3/\text{min}, \quad B = 0.2 \text{ m}^3/\text{min}, \quad V_H = V_C = 1 \text{ m}^3$$

b.) Let $U = \emptyset$

$$\therefore Y = X = e^{A(t-t_0)} X_0 + \int_{t_0}^t e^{A(t-\tau)} B U d\tau = e^{A(t-t_0)} X_0$$

NOTE: $A = \begin{bmatrix} -\frac{f_C-B}{V_C} & \frac{B}{V_C} \\ \frac{-B}{V_H} & \frac{B-f_H}{V_H} \end{bmatrix} = \begin{bmatrix} -\frac{3}{10} & \frac{2}{10} \\ -\frac{2}{10} & \frac{1}{10} \end{bmatrix}, \quad B = \begin{bmatrix} f_C & 0 \\ 0 & f_H \end{bmatrix} = \begin{bmatrix} 1/10 & 0 \\ 0 & 1/10 \end{bmatrix}$

Let: $A = M^{-1}JM, \quad \because \text{eig}(A) \rightarrow \lambda_1 = -1/10, \quad \lambda_2 = 2$

REAL EIGENVECTORS:

Let: $D = (A - \lambda_1 I)V_1 = \begin{bmatrix} \frac{2}{10} & \frac{2}{10} \\ \frac{2}{10} & \frac{1}{10} \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} \Rightarrow \begin{cases} a+b=0 \\ a=0 \end{cases} \Rightarrow V_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}; \quad q_1 = 1 \Rightarrow J \text{ is Type II}_2 \text{ Jordan Matrix}, \quad J = \begin{bmatrix} \lambda_1 & 1 \\ 0 & \lambda_2 \end{bmatrix}$

$(M, -q_1) = 1$ GENERATED E-VECTORS: $\begin{bmatrix} \frac{2}{10} & \frac{2}{10} \\ \frac{2}{10} & \frac{1}{10} \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} \Rightarrow \begin{cases} a+b=5 \\ a=5 \end{cases} \Rightarrow V_2 = \begin{bmatrix} -5 \\ 0 \end{bmatrix}$

$\therefore M = [V_1 \mid V_2] = \begin{bmatrix} 1 & -5 \\ 1 & 0 \end{bmatrix}, \quad M^{-1} = \frac{1}{10} \begin{bmatrix} 0 & 5 \\ -1 & 1 \end{bmatrix}, \quad J = \frac{1}{10} \begin{bmatrix} -1 & 10 \\ 0 & -1 \end{bmatrix}$

$\therefore e^{At} = Me^{Jt}M^{-1} = \frac{1}{5} \begin{bmatrix} 1 & -5 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} e^{-t/10} & te^{-t/10} \\ 0 & e^{-t/10} \end{bmatrix} \begin{bmatrix} 0 & 5 \\ -1 & 1 \end{bmatrix}$

$\Rightarrow e^{At} = \frac{1}{5} \begin{bmatrix} (5-t)e^{-t/10} & te^{-t/10} \\ -te^{-t/10} & (t+5)e^{-t/10} \end{bmatrix}$

$\therefore Y = \begin{bmatrix} Y_1(t) \\ Y_2(t) \end{bmatrix} = \frac{1}{5} \begin{bmatrix} (5-t)e^{(t-t_0)/10} & (t-t_0)e^{(t-t_0)/10} \\ (-t-t_0)e^{(t-t_0)/10} & (t-t_0+5)e^{(t-t_0)/10} \end{bmatrix} X_0$

c.)

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

$$= \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} s+\frac{3}{10} & \frac{2}{10} \\ -\frac{2}{10} & s-\frac{1}{10} \end{bmatrix}^{-1} \begin{bmatrix} 1/10 & 0 \\ 0 & 1/10 \end{bmatrix}$$

$\therefore (sI - A)^{-1} = \frac{1}{|sI - A|} \text{adj}(sI - A), \quad |sI - A| \text{ will contain set of all possible poles of } G(s) \quad (\text{though, some may be canceled in } G(s))$
 $\therefore \text{if the real part of all roots of } |sI - A| \text{ are } < 0, \text{ the real part of all poles will necessarily be } < 0.$

$\therefore |sI - A| = (s + \frac{3}{10})(s - \frac{1}{10}) - \left(\frac{-2}{10}\right)\left(\frac{2}{10}\right) = s^2 + \frac{1}{3}s + \frac{1}{100} = \frac{1}{100}(s + \frac{1}{10})^2, \quad s_1 = s_2 = -\frac{1}{10} \leftarrow \text{double root of } |sI - A|$

$\therefore \text{Re of all roots of } |sI - A| \text{ are } < 0, \quad \text{Re of all poles of } G(s) \text{ are } < 0,$

$\therefore \text{System is BiBO stable}$