

5.11

(1)

$$\text{Let } V = \frac{1}{2}(x_1^2 + x_2^2)$$

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

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

$$\leq -\|x\|_2^2 + \|x\|_2 |u|$$

$$= -(1-\theta)\|x\|_2^2 - \theta\|x\|_2^2 + \|x\|_2 |u|, \quad 0 < \theta < 1$$

$$\leq -(1-\theta)\|x\|_2^2, \quad \forall \|x\|_2 \geq \frac{|u|}{\theta}$$

Thus, by Theorem 4.19, the system is input to state stable.

$\|h(t, x, u)\| \leq \alpha_1(\|x\|)$ is satisfied where α_1 is class K .

\therefore by Theorem 5.3, L_∞ is stable

5.11

(3)

$$\text{Let } V = \frac{1}{2}(x_1^2 + x_2^2)$$

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

$$= -x_1^2 - x_2^2 + (x_1^2 + x_2^2) \|x\|_2^2 - x_1 u (1 - \|x\|_2^2)$$

$$\leq -\|x\|_2^2 + \|x\|_2^4 + \|x\|_2 |u| (1 - \|x\|_2^2), \quad \forall \|x\|_2 < 1$$

$$\leq -(1 - \theta) \|x\|_2^2 (1 - \|x\|_2^2) - \theta \|x\|_2^2 (1 - \|x\|_2^2) + \|x\|_2 |u| (1 - \|x\|_2^2), \quad 0 < \theta < 1$$

$$\leq -(1 - \theta) \|x\|_2^2, \quad \forall 1 > \|x\|_2 \geq \frac{|u|}{\theta}$$

Thus, by Theorem 4.19, the system is locally input to state stable.

$\|h(t, x, u)\| \leq \alpha_1(\|x\|)$ is satisfied where α_1 is class K .

\therefore by Theorem 5.2, small- signal L_∞ stable.

5.11

(5)

$$\text{Let } V = \frac{1}{2}(x_1^2 + x_2^2)$$

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

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

$$\leq -\|x\|_2^2 + x_1 x_2 (x_1^2 + 1) + \|x\|_2 |u|$$

$$\leq -\|x\|_2^2 + \frac{1}{2}\|x\|_2^2 (\|x\|_2^2 + 1) + \|x\|_2 |u|$$

$$\leq -\frac{1}{2}(1-\theta)\|x\|_2^2 (1-\|x\|_2^2) - \frac{1}{2}\theta\|x\|_2^2 (1-\|x\|_2^2) + \|x\|_2 |u|, \quad 0 < \theta < 1$$

$$\leq -(1-\theta)\|x\|_2^2, \quad \forall 1 > \|x\|_2 \geq \frac{2|u|}{\theta}$$

Thus, by Theorem 4.19, the system is locally input to state stable.

$\|h(t, x, u)\| \leq \alpha_1(\|x\|) + \alpha_2(\|u\|)$ is satisfied where α_1 and α_2 are class K .

\therefore by Theorem 5.2, small- signal L_∞ stable.

5.16

(a) let $V(x) = \frac{1}{2}(x_1^2 + x_2^2) + \int_0^{x_1} \sigma(s) ds$

• Using $\sigma(s) \leq L|s|$, we have

$$\frac{1}{2} \|x\|_2^2 \leq V(x) \leq \frac{1+L}{2} \|x\|_2^2$$

• $\frac{\partial V}{\partial x} f(x) \leq -x_1^2 - x_2^2 + |x_2| |u|$

$$\leq -\|x\|_2^2 + \|x\|_2 |u|$$

$$= -(1-\theta) \|x\|_2^2 - \|x\|_2 (\theta \|x\|_2 - |u|) \quad 0 < \theta < 1$$

$$\leq -(1-\theta) \|x\|_2^2 \quad \forall \|x\|_2 \geq \frac{|u|}{\theta}$$

Thus, by theorem 4.1P, the system is input-to-state stable

$\|h(t, x, u)\| \leq \alpha_1(\|x\|)$ is satisfied where α_1 is class K.

\therefore by theorem 5.3, L_∞ stable

(b) Use theorem 5.5

$$H(V, f, g, h, r) \stackrel{\text{def}}{=} \frac{\partial V}{\partial x} f(x) + \frac{1}{2r^2} \frac{\partial V}{\partial x} g(x) g^T(x) \left(\frac{\partial V}{\partial x} \right)^T + \frac{1}{2} h^T(x) h(x) \leq 0$$

$$\left\{ \begin{array}{ll} \frac{\partial V}{\partial x} f(x) = -x_1^2 - x_2^2 - x_1 \sigma(x_1) & g(x) g^T(x) = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \\ \frac{\partial V}{\partial x} = [x_1 + \sigma(x_1) & x_2] & h^T(x) h(x) = x_2^2 \end{array} \right\}$$

By substitution, we have

$$-x_1^2 - x_2^2 - x_1 \sigma(x_1) + \frac{1}{2r^2} x_2^2 + \frac{1}{2} x_2^2 \leq 0$$

$$-x_1^2 - x_1 \sigma(x_1) - x_2^2 \left(1 - \frac{1}{2r^2} - \frac{1}{2} \right) \leq 0$$

$$1 - \frac{1}{2r^2} - \frac{1}{2} \geq 0 \Rightarrow r \geq 1$$

\therefore system is finite-gain L_2 stable, L_2 gain ≤ 1

8.6

(2)

$$\text{eq.pt. : } \begin{bmatrix} 0 \\ 0 \end{bmatrix} \rightarrow A = \left. \frac{\partial f}{\partial x} \right|_{0,0} = \begin{bmatrix} 2ax_1 & -2x_2 \\ 2x_1 + x_2 & -1 + x_1 \end{bmatrix} \bigg|_{0,0} = \begin{bmatrix} 0 & 0 \\ 0 & -1 \end{bmatrix}$$

$$\Rightarrow T = T^{-1} = I_{2 \times 2}$$

$$\begin{bmatrix} y \\ z \end{bmatrix} = Tx = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$\dot{y} = ay^2 - z^2 \Rightarrow A_1 = 0, g_1(y, z) = ay^2 - z^2$$

$$\dot{z} = -z + y^2 + yz \Rightarrow A_2 = -1, g_2(y, z) = y^2 + yz$$

$$\bullet z = h(y), h(0) = 0, \frac{\partial h}{\partial y}(0) = 0$$

• center manifold equation

$$0 = -h(y) + g_2(y, z) - \frac{\partial h}{\partial y}(y)[g_1(y, z)]$$

$$= -h(y) + (y^2 + yh(y)) - \frac{\partial h}{\partial y}(y)[ay^2 - h(y)^2]$$

$$\therefore h(y) = y^2 + O(|y|^3) \quad \dot{y} = ay^2 + O(|y|^3)$$

$$\dot{y} = -(y^2 + O(|y|^3))^2 = -y^4 + O(|y|^5)$$

\therefore unstable.

(6),(8) : the same method (2) (ref, lecture note 6-1~6-8)

8.6 (8)

$$eq.pt. : \begin{bmatrix} 0 \\ 0 \end{bmatrix} \rightarrow A = \frac{\partial f}{\partial x} \Big|_{0,0} = \begin{bmatrix} -2 & -3 & 1 \\ 1 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

Since we know the eigenvalues of A have imaginary parts

$(0 \text{ and } -0.5 \pm j\frac{\sqrt{3}}{2})$, we don't need to have the A as Jordan

form. Alternatively, we divide A matrix to two sub-parts: stable part and uncertain part. Thus, we take T as follows:

$$T = T^{-1} = I_{3 \times 3}$$

$$\begin{bmatrix} y \\ z_1 \\ z_2 \end{bmatrix} = Tx = \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 1 \\ 0 & 1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$

Hence, we show the following system.

$$\begin{bmatrix} \dot{y} \\ \dot{z}_1 \\ \dot{z}_2 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & -2 & -3 \\ 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} y \\ z_1 \\ z_2 \end{bmatrix} + \begin{bmatrix} (z_1 - y)^2 \\ (z_1 - y)^2 \\ 0 \end{bmatrix}$$

$$\bullet z = h(y), \quad h(0) = 0, \quad \frac{\partial h}{\partial y}(0) = 0$$

$$\begin{bmatrix} z_1 \\ z_2 \end{bmatrix} = \begin{bmatrix} h_1(y) \\ h_2(y) \end{bmatrix} = \begin{bmatrix} h_{13}y^2 + O(|y|^3) \\ h_{23}y^2 + O(|y|^3) \end{bmatrix}$$

•center manifold equation

$$0 = \begin{bmatrix} -2 & -3 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} h_1(y) \\ h_2(y) \end{bmatrix} + \begin{bmatrix} (h_1(y) - y)^2 \\ 0 \end{bmatrix} - \begin{bmatrix} \frac{\partial h_1}{\partial y}(y) \\ \frac{\partial h_2}{\partial y}(y) \end{bmatrix} \begin{bmatrix} (h_1(y) - y)^2 \\ 0 \end{bmatrix}$$

From the above equation, we show following two equations:

$$i) (-2h_{13} - 3h_{23})y^2 + y^2 + O(|y|^3) = 0$$

$$ii) (h_{13} + h_{23})y^2 + O(|y|^3) = 0$$

$$\text{since } (h_1(y) - y)^2 = y^2 + O(|y|^3), \quad \frac{\partial h_1}{\partial y}(y)(h_1(y) - y)^2 = O(|y|^3).$$

Hence, since $h_{13} = -1$ and $h_{23} = 1$, we have

$$\begin{bmatrix} z_1 \\ z_2 \end{bmatrix} = \begin{bmatrix} h_1(y) \\ h_2(y) \end{bmatrix} = \begin{bmatrix} -y^2 + O(|y|^3) \\ y^2 + O(|y|^3) \end{bmatrix}$$

Finally,

$$\dot{y} = y^2 + O(|y|^3)$$

\Rightarrow we conclude that this system is unstable

8.15

See Example 8-10.

$$(a) \quad \dot{V}(x) = -(2x_2 + x_1)^2 + 2x_2^2(2x_2 + x_1)^2 - 3x_1^2 - 6x_2^2 \\ \leq -(2x_2 + x_1)^2(1 - 2x_2^2)$$

$$1 - 2x_2^2 > 0 \rightarrow -\frac{1}{\sqrt{2}} < x_2 < \frac{1}{\sqrt{2}}$$

 \therefore asymptotically stable

$$(b) \quad i) x_2 = 1 \rightarrow \dot{V}(x) = -2(x_1 - 1)^2$$

$$ii) x_2 = -1 \rightarrow \dot{V}(x) = -2(x_1 + 1)^2$$

Let $\sigma = x_2$

$$i) \sigma = 1 \rightarrow x_1 \geq -1$$

$$ii) \sigma = -1 \rightarrow x_1 \leq 1$$

$$\therefore (-1, 1), (1, -1)$$

$$V(-1, 1) = 5, \quad V(1, -1) = 5$$