

## Class Handout: Chapter 5 Input-Output Stability

2006 Fall

### I. $\mathcal{L}$ STABILITY

$$y = Hu$$

where  $y : [0, \infty) \rightarrow \mathbb{R}^q$  and  $u : [0, \infty) \rightarrow \mathbb{R}^m$  are functions, and  $H$  is a mapping (or operator) between two signals.

To measure the size of signals, let us introduce the norm function  $\|\cdot\|$ , with which it holds that

- $\|u\| = 0$  if and only if  $u(t) \equiv 0$ ,
- $\|au\| = |a|\|u\|$ ,
- $\|u_1 + u_2\| \leq \|u_1\| + \|u_2\|$ .

For example,

$$\|u\|_{\mathcal{L}_\infty} := \sup_{t \geq 0} \|u(t)\|$$

$$\|u\|_{\mathcal{L}_p} := \left( \int_0^\infty \|u(t)\|^p dt \right)^{1/p}$$

(Here, the norm inside the integral is not necessarily equal to the  $p$ -norm, but usually it is.)

The set of all piecewise continuous functions whose  $\mathcal{L}_p$  ( $p \in [1, \infty]$ ) norm is finite, is denoted by  $\mathcal{L}_p^m$ . Sub-, or super-script may be omitted if clear from the context.

In fact, ‘piecewise continuous’ can be replaced by ‘measurable’.

“Extended space”:

$$\mathcal{L}_e^m = \{u : u_\tau \in \mathcal{L}^m, \forall \tau \in [0, \infty)\}$$

where  $u_\tau$  is a truncation of  $u$  defined by

$$u_\tau(t) = \begin{cases} u(t), & 0 \leq t \leq \tau, \\ 0, & t > \tau \end{cases}$$

The extended space is a linear space that contains the unextended space as a subset. It allows us to deal with unbounded growing signals (e.g.,  $u(t) = t$  belongs to  $\mathcal{L}_{\infty, \epsilon}$ ).

“Causal” mapping  $H$ :

$$(Hu)_\tau = (Hu_\tau)_\tau, \quad \forall \tau \geq 0.$$

Causality is an intrinsic property of dynamical systems represented by state models.

#### Definition 5.1

A mapping  $H : \mathcal{L}_e^m \rightarrow \mathcal{L}_e^q$  is  $\mathcal{L}$  stable if  $\exists$  a class- $\mathcal{K}$  function  $\alpha$ , and  $\beta > 0$  s.t.

$$\|(Hu)_\tau\|_{\mathcal{L}} \leq \alpha(\|u_\tau\|_{\mathcal{L}}) + \beta$$

for all  $u \in \mathcal{L}_e^m$  and  $\tau \in [0, \infty)$ .

It is finite-gain  $\mathcal{L}$  stable if  $\exists$  nonnegative constants  $\gamma$  and  $\beta$  s.t.

$$\|(Hu)_\tau\|_{\mathcal{L}} \leq \gamma\|u_\tau\|_{\mathcal{L}} + \beta \quad (1)$$

for all  $u \in \mathcal{L}_e^m$  and  $\tau \in [0, \infty)$ .

- $\beta$  is a bias term, which possibly considers the initial condition.
- If (1) holds, then the system has an  $\mathcal{L}$  gain less than or equal to  $\gamma$ . The smallest  $\gamma$  s.t. (1) holds is called the gain of the system.
- For causal,  $\mathcal{L}$  stable systems, we have

$$u \in \mathcal{L}^m \Rightarrow Hu \in \mathcal{L}^q$$

and

$$\|Hu\|_{\mathcal{L}} \leq \alpha(\|u\|_{\mathcal{L}}) + \beta, \quad \forall u \in \mathcal{L}^m.$$

- $\mathcal{L}_\infty$  stability is the familiar notion of BIBO stability.

**Example 5.1** Let

$$h(u) = a + b \tanh cu = a + b \frac{e^{cu} - e^{-cu}}{e^{cu} + e^{-cu}}$$

Then

$$|h(u)| \leq a + bc|u|.$$

Hence,  $h$  is finite-gain  $\mathcal{L}_\infty$  stable with  $\gamma = bc$  and  $\beta = a$ .

If  $a = 0$ ,  $h$  is  $\mathcal{L}_p$  stable with zero bias, gain  $\gamma = bc$ , for each  $p \in [1, \infty]$  since

$$\int_0^\infty |h(u(t))|^p dt \leq (bc)^p \int_0^\infty |u(t)|^p dt.$$

If  $h(u) = u^2$ , it is  $\mathcal{L}_\infty$  stable with zero bias and  $\alpha(r) = r^2$ , but is not finite-gain  $\mathcal{L}_\infty$  stable.

**Example 5.2** Consider a causal convolution

$$y(t) = \int_0^t h(t - \sigma)u(\sigma)d\sigma$$

where  $h(t) = 0$  for  $t < 0$ . Suppose that  $h \in \mathcal{L}_{1,e}$ ; that is, for every  $\tau \in [0, \infty)$ ,

$$\|h_\tau\|_{\mathcal{L}_1} = \int_0^\infty |h_\tau(\sigma)|d\sigma = \int_0^\tau |h(\sigma)|d\sigma < \infty.$$

If  $u \in \mathcal{L}_{\infty, e}$  and  $\tau \geq t$ , then

$$\begin{aligned} |y(t)| &\leq \int_0^t |h(t-\sigma)| |u(\sigma)| d\sigma \\ &\leq \int_0^t |h(t-\sigma)| d\sigma \sup_{0 \leq \sigma \leq \tau} |u(\sigma)| = \int_0^t |h(s)| ds \sup_{0 \leq \sigma \leq \tau} |u(\sigma)| \end{aligned}$$

Consequently,

$$\|y_\tau\|_{\mathcal{L}_\infty} \leq \|h_\tau\|_{\mathcal{L}_1} \|u_\tau\|_{\mathcal{L}_\infty}, \quad \forall \tau \in [0, \infty).$$

This does not mean  $\mathcal{L}$  stability of the system, because  $\|h_\tau\|_{\mathcal{L}_1}$  is not *uniformly* bounded with respect to  $\tau$ . (For example,  $h(t)$  may diverge.)

Assume that  $h \in \mathcal{L}_1$ ; that is,

$$\|h\|_{\mathcal{L}_1} = \int_0^\infty |h(\sigma)| d\sigma < \infty.$$

Then, it holds that

$$\|y_\tau\|_{\mathcal{L}_\infty} \leq \|h\|_{\mathcal{L}_1} \|u_\tau\|_{\mathcal{L}_\infty}, \quad \forall \tau \in [0, \infty),$$

which means that the system is finite-gain  $\mathcal{L}_\infty$  stable.

In fact, this assumption implies finite-gain  $\mathcal{L}_p$  stability for any  $p \in [1, \infty]$ . See the textbook pp. 199 and 200.

**Definition 5.2** (Small-signal  $\mathcal{L}$  stability)

A mapping  $H : \mathcal{L}_e^m \rightarrow \mathcal{L}_e^q$  is small-signal (finite-gain)  $\mathcal{L}$  stable if  $\exists r > 0$  s.t. it is (finite-gain)  $\mathcal{L}$  stable for all  $u \in \mathcal{L}_e^m$  with  $\sup_{0 \leq t \leq \tau} \|u(t)\| \leq r$ .

**Example 5.3** See the textbook.

## II. $\mathcal{L}$ STABILITY OF STATE MODELS

Consider

$$\begin{aligned} \dot{x} &= f(t, x, u), & x(0) &= x_0, \\ y &= h(t, x, u) \end{aligned}$$

where  $x \in \mathbb{R}^n$ ,  $u \in \mathbb{R}^m$ ,  $y \in \mathbb{R}^p$ ,  $f : [0, \infty) \times D \times D_u \rightarrow \mathbb{R}^n$  is piecewise continuous in  $t$  and locally Lipschitz in  $(x, u)$ ,  $h : [0, \infty) \times D \times D_u \rightarrow \mathbb{R}^q$  is piecewise continuous in  $t$  and continuous in  $(x, u)$ ,  $D$  is a domain containing  $x = 0$ ,  $D_u$  is a domain containing  $u = 0$ .

\* For each fixed  $x_0 \in D$ , the above system defines an operator  $H$  that assigns to each input signal  $u(t)$  the corresponding output signal  $y(t)$ .

Assume that  $x = 0$  of

$$\dot{x} = f(t, x, 0)$$

is an equilibrium.

**Theorem 5.1**

Let  $r > 0$  and  $r_u > 0$  s.t.  $\{\|x\| \leq r\} \subset D$  and  $\{\|u\| \leq r_u\} \subset D_u$ .

- IF  $x = 0$  of  $\dot{x} = f(t, x, 0)$  is LES, and  $\exists V(t, x)$  s.t.

$$c_1\|x\|^2 \leq V(t, x) \leq c_2\|x\|^2 \quad (2)$$

$$\frac{\partial V}{\partial t} + \frac{\partial V}{\partial x} f(t, x, 0) \leq -c_3\|x\|^2 \quad (3)$$

$$\left\| \frac{\partial V}{\partial x}(t, x) \right\| \leq c_4\|x\| \quad (4)$$

for all  $(t, x) \in [0, \infty) \times D$ , and,

IF

$$\|f(t, x, u) - f(t, x, 0)\| \leq L\|u\|$$

$$\|h(t, x, u)\| \leq \eta_1\|x\| + \eta_2\|u\|$$

for all  $(t, x, u) \in [0, \infty) \times D \times D_u$ ,

THEN, for each  $\|x_0\| \leq r\sqrt{c_1/c_2}$ , the system is small-signal finite-gain  $\mathcal{L}_p$  stable ( $p \in [1, \infty)$ ).

In particular, for each  $u \in \mathcal{L}_{p,e}$  with  $\sup_{0 \leq t \leq \tau} \|u(t)\| \leq \min\{r_u, c_1 c_3 r / (c_2 c_4 L)\}$ , the output  $y(t)$  satisfies

$$\|y_\tau\|_{\mathcal{L}_p} \leq \gamma \|u_\tau\|_{\mathcal{L}_p} + \beta \quad (5)$$

for all  $\tau \in [0, \infty)$ , with

$$\gamma = \eta_2 + \frac{\eta_1 c_2 c_4 L}{c_1 c_3}, \quad \beta = \eta_1 \|x_0\| \sqrt{\frac{c_2}{c_1}} \rho$$

where

$$\rho = \begin{cases} 1, & \text{if } p = \infty, \\ \left(\frac{2c_2}{c_3 p}\right)^{1/p}, & \text{if } p \in [1, \infty) \end{cases}.$$

- IF the origin is GES and  $D = R^n$ ,  $D_u = R^m$ , THEN the above holds for any  $x_0$  and  $u(t)$ .

*Proof.* Derivative of  $V$  along  $\dot{x} = f(t, x, u)$  is

$$\begin{aligned} \dot{V} &= \frac{\partial V}{\partial t} + \frac{\partial V}{\partial x} f(t, x, 0) + \frac{\partial V}{\partial x} [f(t, x, u) - f(t, x, 0)] \\ &\leq -c_3\|x\|^2 + c_4 L \|x\| \|u\| \end{aligned}$$

(At time point, the proof for global case is actually done already. Can you see?)

(For numerical analysis, we continue...) Take  $W(t) = \sqrt{V(t, x(t))}$ .

If  $V(t, x(t)) \neq 0$ ,

$$\begin{aligned}\dot{W} &= \frac{\dot{V}}{2\sqrt{V}} \leq \frac{1}{2\sqrt{V}} (-c_3\|x\|^2 + c_4L\|x\|\|u\|) \\ &\leq \frac{1}{2\sqrt{V}} \left( -\frac{c_3}{c_2}V(t, x) + c_4L\frac{\sqrt{V}}{\sqrt{c_1}}\|u\| \right) \\ &= -\frac{c_3}{2c_2}W + \frac{c_4L}{2\sqrt{c_1}}\|u\|\end{aligned}$$

If  $V(t, x(t)) = 0$  (i.e.,  $x(t) = 0$ ), it can be shown that

$$D^+W(t) \leq \frac{c_4L}{2\sqrt{c_1}}\|u(t)\|.$$

Hence, for all  $V(t, x(t))$ , we have

$$D^+W(t) \leq -\frac{1}{2}\frac{c_3}{c_2}W + \frac{c_4L}{2\sqrt{c_1}}\|u(t)\|.$$

By the comparison lemma,

$$W(t) \leq e^{-t\frac{c_3}{2c_2}}W(0) + \frac{c_4L}{2\sqrt{c_1}}\int_0^t e^{-(t-\tau)\frac{c_3}{2c_2}}\|u(\tau)\|d\tau.$$

We then obtain (since  $\sqrt{c_1}\|x(t)\| \leq W(t) \leq \sqrt{c_2}\|x(t)\|$ )

$$\begin{aligned}\|x(t)\| &\leq \sqrt{\frac{c_2}{c_1}}\|x_0\|e^{-t\frac{c_3}{2c_2}} + \frac{c_4L}{2c_1}\int_0^t e^{-(t-\tau)\frac{c_3}{2c_2}}\|u(\tau)\|d\tau \\ &= \sqrt{\frac{c_2}{c_1}}\|x_0\|e^{-t\frac{c_3}{2c_2}} + \frac{c_4L}{2c_1}\frac{2c_2}{c_3}\left(1 - e^{-t\frac{c_3}{2c_2}}\right)\left(\sup_{0 \leq \sigma \leq t}\|u(\sigma)\|\right)\end{aligned}$$

(We now check if  $x(t) \in B_r$  for all  $t \geq 0$  so that the whole analysis is valid.) Then, since

$$\|x_0\| \leq r\sqrt{\frac{c_1}{c_2}}, \quad \sup_{0 \leq \sigma \leq t}\|u(\sigma)\| \leq \frac{c_1c_3r}{c_2c_4L},$$

we have

$$\|x(t)\| \leq re^{-t\frac{c_3}{2c_2}} + \left(1 - e^{-t\frac{c_3}{2c_2}}\right)r = r.$$

(We now obtain (5).) From the assumption, we have

$$\|y(t)\| \leq k_1e^{-at} + k_2\int_0^t e^{-a(t-\tau)}\|u(\tau)\|d\tau + k_3\|u(t)\|$$

where

$$k_1 = \sqrt{\frac{c_2}{c_1}}\|x_0\|\eta_1, \quad k_2 = \frac{c_4L\eta_1}{2c_1}, \quad k_3 = \eta_2, \quad a = \frac{c_3}{2c_2}.$$

Set

$$y_1(t) = k_1e^{-at}, \quad y_2(t) = k_2\int_0^t e^{-a(t-\tau)}\|u(\tau)\|d\tau, \quad y_3(t) = k_3\|u(t)\|.$$

Then, for any  $p \in [1, \infty]$ , we have

$$\|y_{2,\tau}\|_{\mathcal{L}_p} \leq \frac{k_2}{a}\|u_\tau\|_{\mathcal{L}_p}$$

Exercise 5.6 and Exercise 3.24 are employed here.

since  $\|h\|_{\mathcal{L}_1} = 1/a$  and from Example 5.2, and

$$\|y_{3,\tau}\|_{\mathcal{L}_p} \leq k_3 \|u_\tau\|_{\mathcal{L}_p}.$$

For the term  $y_1$ ,

$$\|y_{1,\tau}\|_{\mathcal{L}_p} \leq k_1 \rho$$

where

$$\rho = \begin{cases} 1, & \text{if } p = \infty, \\ \left(\frac{1}{ap}\right)^{1/p}, & \text{if } p \in [1, \infty) \end{cases}.$$

Thus, by the triangular inequality,

$$\gamma = k_3 + \frac{k_2}{a}, \quad \beta = k_1 \rho.$$

On the other hand, the global case follows easily.

### Corollary 5.1

IF the origin of  $\dot{x} = f(t, x, 0)$  is GES(LES) and

$$\begin{aligned} \|f(t, x, u) - f(t, x, 0)\| &\leq L \|u\| \\ \|h(t, x, u)\| &\leq \eta_1 \|x\| + \eta_2 \|u\| \end{aligned}$$

THEN the original system is (small-signal) finite-gain  $\mathcal{L}_p$  stable for each  $p \in [1, \infty]$ .

### Corollary 5.2

A LTI system is finite-gain  $\mathcal{L}_p$  stable if  $A$  is Hurwitz.

(In particular,

$$\begin{aligned} \gamma &= \|D\|_2 + \frac{2\lambda_{\max}^2(P)\|B\|_2\|C\|_2}{\lambda_{\min}(P)} \\ \beta &= \|C\|_2 \|x_0\| \sqrt{\frac{\lambda_{\max}(P)}{\lambda_{\min}(P)}} \rho \\ \rho &= \begin{cases} 1, & p = \infty, \\ \left(\frac{2\lambda_{\max}(P)}{p}\right)^{\frac{1}{p}}, & p \in [1, \infty). \end{cases} \end{aligned}$$

where  $A^T P + PA = -I$ .)

### Example 5.4

$$\begin{aligned} \dot{x} &= -x - x^3 + u, & x(0) &= x_0, \\ y &= \tanh x + u \end{aligned}$$

With  $V(x) = x^2/2$ , we can show that the system is finite-gain  $\mathcal{L}_p$  stable.

Let us consider LUAS case restricting to  $\mathcal{L}_\infty$  stability.

**Theorem 5.2** (local version)

Let  $r > 0$  and  $r_u > 0$  s.t.  $\{\|x\| \leq r\} \subset D$  and  $\{\|u\| \leq r_u\} \subset D_u$ .

IF  $x = 0$  of  $\dot{x} = f(t, x, 0)$  is LUAS, and  $\exists V(t, x)$  s.t.

$$\begin{aligned} \alpha_1(\|x\|) &\leq V(t, x) \leq \alpha_2(\|x\|) \\ \frac{\partial V}{\partial t} + \frac{\partial V}{\partial x} f(t, x, 0) &\leq -\alpha_3(\|x\|) \\ \left\| \frac{\partial V}{\partial x}(t, x) \right\| &\leq \alpha_4(\|x\|) \end{aligned}$$

for all  $(t, x) \in [0, \infty) \times D$ , and,

IF

$$\begin{aligned} \|f(t, x, u) - f(t, x, 0)\| &\leq \alpha_5(\|u\|) \\ \|h(t, x, u)\| &\leq \alpha_6(\|x\|) + \alpha_7(\|u\|) + \eta \end{aligned}$$

for all  $(t, x, u) \in [0, \infty) \times D \times D_u$ ,

THEN, for each  $\|x_0\| \leq \alpha_2^{-1}(\alpha_1(r))$ , the system is small-signal  $\mathcal{L}_\infty$  stable.

*Proof.*

Derivative of  $V$  along  $\dot{x} = f(t, x, u)$  is

$$\begin{aligned} \dot{V} &= \frac{\partial V}{\partial t} + \frac{\partial V}{\partial x} f(t, x, 0) + \frac{\partial V}{\partial x} [f(t, x, u) - f(t, x, 0)] \\ &\leq -\alpha_3(\|x\|) + \alpha_4(\|x\|)\alpha_5(\|u\|) \\ &\leq -(1 - \theta)\alpha_3(\|x\|) - \theta\alpha_3(\|x\|) + \alpha_4(r)\alpha_5 \left( \sup_{0 \leq t \leq \tau} \|u(t)\| \right) \end{aligned}$$

where  $0 < \theta < 1$ . Set

$$\mu = \alpha_3^{-1} \left( \frac{\alpha_4(r)\alpha_5 \left( \sup_{0 \leq t \leq \tau} \|u(t)\| \right)}{\theta} \right).$$

We consider only such  $u(t)$  that  $\sup_{0 \leq t \leq \tau} \|u(t)\|$  is small enough for  $\mu < \alpha_2^{-1}(\alpha_1(r))$ . Then,

$$\dot{V} \leq -(1 - \theta)\alpha_3(\|x\|), \quad \forall \|x\| \geq \mu.$$

From Theorem 4.18,

$$\|x(t)\| \leq \beta(\|x_0\|, t) + \gamma \left( \sup_{0 \leq t \leq \tau} \|u(t)\| \right)$$

for all  $0 \leq t \leq \tau$ . Hence,

$$\begin{aligned} \|y(t)\| &\leq \alpha_6 \left( \beta(\|x_0\|, t) + \gamma \left( \sup_{0 \leq t \leq \tau} \|u(t)\| \right) \right) + \alpha_7(\|u(t)\|) + \eta \\ &\leq \alpha_6(2\beta(\|x_0\|, t)) + \alpha_6 \left( 2\gamma \left( \sup_{0 \leq t \leq \tau} \|u(t)\| \right) \right) + \alpha_7(\|u(t)\|) + \eta \end{aligned}$$

Therefore, we have

$$\begin{aligned} \|y_\tau\|_{\mathcal{L}_\infty} &\leq \gamma_0(\|u_\tau\|_{\mathcal{L}_\infty}) + \beta_0 \\ \gamma_0 &= \alpha_6 \circ 2\gamma + \alpha_7, \quad \beta_0 = \alpha_6(2\beta(\|x_0\|, 0)) + \eta \end{aligned}$$

**Theorem 5.3** (global version)

IF  $D = \mathbb{R}^n$ ,  $D_u = \mathbb{R}^m$ , the system  $\dot{x} = f(t, x, u)$  is ISS, and

$$\|h(t, x, u)\| \leq \alpha_1(\|x\|) + \alpha_2(\|u\|) + \eta,$$

THEN, the system is  $\mathcal{L}_\infty$  stable.

*Proof.* Trivial. Isn't it?

\* Think about why the global asymptotic stable case needs so strong property (ISS)?

Consider

$$\dot{x} = -\frac{x}{1+x^2} + u,$$

which is GAS (but not GES), and is not  $\mathcal{L}_\infty$  stable (also not ISS).



### III. $\mathcal{L}_2$ GAIN

“ $\mathcal{L}_2$  stability plays a special role in systems analysis. It is natural to work with square-integrable signals, which can be viewed as finite-energy signals. In many control problems, the system is represented as an input-output map, from a disturbance input  $u$  to a controlled output  $y$ , which is required to be small.”

Here we study how to calculate the  $\mathcal{L}_2$  gain for TI systems.

**Theorem 5.4** Consider

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

$$y = Cx + Du$$

where  $A$  is Hurwitz. Let  $G(s) = C(sI - A)^{-1}B + D$ . Then, the  $\mathcal{L}_2$  gain is

$$\sup_{\omega \in \mathbb{R}} \|G(j\omega)\|_2 = \sup_{\omega \in \mathbb{R}} \sqrt{\lambda_{\max}[G^T(-j\omega)G(j\omega)]}.$$

*Proof.* Due to linearity, we set  $x(0) = 0$ . From Fourier transform theory, for a causal signal  $y \in \mathcal{L}_2$ ,

$$Y(j\omega) = \int_0^{\infty} y(t)e^{-j\omega t} dt, \quad Y(j\omega) = G(j\omega)U(j\omega).$$

By Parseval's theorem,

$$\begin{aligned} \|y\|_{\mathcal{L}_2}^2 &= \int_0^{\infty} y^T(t)y(t)dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} Y^*(j\omega)Y(j\omega)d\omega \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} U^*(j\omega)G^T(-j\omega)G(j\omega)U(j\omega)d\omega \\ &\leq \left( \sup_{\omega \in \mathbb{R}} \|G(j\omega)\|_2 \right)^2 \frac{1}{2\pi} \int_{-\infty}^{\infty} U^*(j\omega)U(j\omega)d\omega \\ &= \left( \sup_{\omega \in \mathbb{R}} \|G(j\omega)\|_2 \right)^2 \|u\|_{\mathcal{L}_2}^2 \end{aligned}$$

which shows that the  $\mathcal{L}_2$  gain is less than or equal to  $\sup_{\omega \in \mathbb{R}} \|G(j\omega)\|_2$ . See Appendix C.10 to show that the  $\mathcal{L}_2$  gain is, in fact, equal to  $\sup_{\omega \in \mathbb{R}} \|G(j\omega)\|_2$ .

Consider

$$\dot{x} = f(x) + G(x)u, \quad x(0) = x_0,$$

$$y = h(x)$$

where  $f$  is locally Lipschitz,  $G$  and  $h$  are continuous, and  $f(0) = 0$  and  $h(0) = 0$ .

**Theorem 5.5**

IF  $\exists C^1$  positive semidefinite function  $V(x)$  and  $\gamma > 0$  s.t.

$$\mathcal{H}(V, f, G, h, \gamma) := \frac{\partial V}{\partial x} f(x) + \frac{1}{2\gamma^2} \frac{\partial V}{\partial x} G(x) G^T(x) \left( \frac{\partial V}{\partial x} \right)^T + \frac{1}{2} h^T(x) h(x) \leq 0,$$

THEN, for each  $x_0 \in \mathbb{R}^n$ , the system is finite-gain  $\mathcal{L}_2$  stable and the gain  $\leq \gamma$ .

\* Hamilton-Jacobi inequality

\* Compare with Theorem 5.1—no exponential stability, but needs a solution to HJ eq.

*Proof.*

$$\begin{aligned} \dot{V} &= L_f V(x) + L_G V(x) u = -\frac{1}{2} \gamma^2 \|u\|^2 - \frac{1}{\gamma^2} (L_G V)^T(x) \|^2_2 + L_f V(x) \\ &\quad + \frac{1}{2\gamma^2} L_G V(x) (L_G V)^T(x) + \frac{1}{2} \gamma^2 \|u\|^2_2 \\ &\leq \frac{1}{2} \gamma^2 \|u\|^2_2 - \frac{1}{2} \|y\|^2_2 - \frac{1}{2} \gamma^2 \|u\|^2 - \frac{1}{\gamma^2} (L_G V)^T(x) \|^2_2 \\ &\leq \frac{1}{2} \gamma^2 \|u\|^2_2 - \frac{1}{2} \|y\|^2_2 \end{aligned}$$

Thus,

$$V(x(\tau)) - V(x(0)) \leq \frac{1}{2} \gamma^2 \int_0^\tau \|u(t)\|^2_2 dt - \frac{1}{2} \int_0^\tau \|y(t)\|^2_2 dt.$$

Since  $V(x) \geq 0$ , we have

$$\int_0^\tau \|y(t)\|^2_2 \leq \gamma^2 \int_0^\tau \|u(t)\|^2_2 dt + 2V(x_0).$$

Then,

$$\|y_\tau\|_{\mathcal{L}_2} \leq \gamma \|u_\tau\|_{\mathcal{L}_2} + \sqrt{2V(x_0)}.$$

### Example 5.8

$$\begin{aligned} \dot{x}_1 &= x_2 \\ \dot{x}_2 &= -ax_1^3 - kx_2 + u, \quad a > 0, k > 0, \\ y &= x_2 \end{aligned}$$

Let  $V(x) = \alpha(ax_1^4/4 + x_2^2/2)$ . Then,  $L_f V = -\alpha k x_2^2$ ,  $L_G V = \alpha x_2$ ,  $h(x) = x_2$ , so that

$$\mathcal{H} = \left( -\alpha k + \frac{\alpha^2}{2\gamma^2} + \frac{1}{2} \right) x_2^2.$$

That is, if

$$\gamma^2 \geq \frac{\alpha^2}{2\alpha k - 1}$$

then the system is  $\mathcal{L}_2$  stable. The right-hand side has a minimum  $1/k^2$  for  $\alpha = 1/k$ , so with  $\gamma = 1/k$ , we conclude that the system is finite-gain  $\mathcal{L}_2$  stable with the gain less than or equal to  $1/k$ .

**Example 5.9** Consider the system with the property

$$\begin{aligned} L_f W(x) &\leq -kh^T(x)h(x), & k > 0 \\ L_G W(x) &= h^T(x) \end{aligned}$$

where  $W(x)$  is a  $C^1$  positive semidefinite function.

Let  $V(x) = \alpha W(x)$ . Then,

$$\mathcal{H} = \left( -\alpha k + \frac{\alpha^2}{2\gamma^2} + \frac{1}{2} \right) h^T(x)h(x).$$

By Example 5.8, we conclude that the system is finite-gain  $\mathcal{L}_2$  stable with the gain  $\leq 1/k$ .

**Example 5.10** Consider the system with the property

$$\begin{aligned} L_f W(x) &\leq 0 \\ L_G W(x) &= h^T(x) \end{aligned}$$

where  $W(x)$  is a  $C^1$  positive semidefinite function.

Let an output feedback control

$$u = -ky + v, \quad k > 0.$$

Then, the closed-loop system becomes

$$\begin{aligned} \dot{x} &= f(x) - kG(x)G^T(x) \left( \frac{\partial W}{\partial x} \right)^T + G(x)v \\ y &= h(x) \end{aligned}$$

This system is, in fact, the case of Example 5.9. (Verify!)

So, this system is finite-gain  $\mathcal{L}_2$  stable with the gain  $\leq 1/k$  from  $v$  to  $y$ . Note that the gain can be arbitrarily assigned with the feedback.

**Example 5.11** Consider

$$\begin{aligned} \dot{x} &= Ax + Bu \\ y &= Cx \end{aligned}$$

Suppose  $\exists P \geq 0$  s.t.

$$PA + A^T P + \frac{1}{\gamma^2} PBB^T P + C^T C = 0$$

(the Riccati equation), with some  $\gamma > 0$ .

Then,  $V(x) = \frac{1}{2}x^T P x$  satisfies the HJ equation for this system, i.e.,  $\mathcal{H} = 0$ . Thus, the system is finite-gain  $\mathcal{L}_2$  stable with the gain  $\leq \gamma$ . (In fact, the Riccati equation has a solution  $P \geq 0$  if and only if the system's  $\mathcal{L}_2$  gain is less than or equal to  $\gamma$ .)

- From the proof of Theorem 5.5, we note that, if the assumptions hold only on a finite domain  $D$ , then we obtain the same conclusion as long as the solution  $x(t)$  stays in  $D$ . (**Corollary 5.4.**) This possibly restricts the class of input signal, which also depends on the initial condition  $x_0$ .
- If  $\dot{x} = f(x)$  is AS and if  $\|x_0\|$  and  $\sup_{0 \leq t \leq \tau} \|u(t)\|$  are sufficiently small (relative to  $D$ ), then the solution  $x(t)$  remains in the neighborhood of the origin. This leads to the following Lemma.

Consider

$$\begin{aligned}\dot{x} &= f(x) + G(x)u, & x(0) &= x_0, \\ y &= h(x)\end{aligned}$$

where  $f$  is  $C^1$ ,  $G$  and  $h$  are continuous on a set  $D \subset \mathbb{R}^n$ , and  $f(0) = 0$  and  $h(0) = 0$ .

**Lemma 5.1**

IF the origin of  $\dot{x} = f(x)$  is AS, and  $\exists C^1$  positive semidefinite function  $V(x)$  and  $\gamma > 0$  s.t.

$$\mathcal{H}(V, f, G, h, \gamma) := \frac{\partial V}{\partial x} f(x) + \frac{1}{2\gamma^2} \frac{\partial V}{\partial x} G(x) G^T(x) \left( \frac{\partial V}{\partial x} \right)^T + \frac{1}{2} h^T(x) h(x) \leq 0$$

on a domain  $D$ ,

THEN  $\exists k_1 > 0$  s.t. for each  $\|x_0\| \leq k_1$ , the system is small-signal finite-gain  $\mathcal{L}_2$  stable with the gain  $\leq \gamma$ .

*Proof.* We can apply Corollary 5.4, if we show that  $x(t)$  stays in a neighborhood of the origin.

By the converse theorem,  $\exists$  a  $C^1$  function  $W(x)$  and  $r_0 > 0$  s.t.

$$\begin{aligned}\alpha_1(\|x\|) &\leq W(x) \leq \alpha_2(\|x\|) \\ \frac{\partial W}{\partial x} f(x) &\leq -\alpha_3(\|x\|)\end{aligned}$$

for all  $x \in B_{r_0}$ , and without loss of generality, we assume that  $B_{r_0} \subset D$ .

Let  $k$  and  $L$  be an upper bound of  $\|\partial W/\partial x\|$  and  $\|G(x)\|$ , respectively. Then,

$$\begin{aligned}\dot{W}(x, u) &\leq \frac{\partial W}{\partial x} f(x) + \frac{\partial W}{\partial x} G(x)u \leq -\alpha_3(\|x\|) + kL\|u\| \\ &\leq -(1-\theta)\alpha_3(\|x\|) - \theta\alpha_3(\|x\|) + kL \sup_{0 \leq t \leq \tau} \|u(t)\| \\ &\leq -(1-\theta)\alpha_3(\|x\|), \quad \forall \|x\| \geq \alpha_3^{-1} \left( kL \sup_{0 \leq t \leq \tau} \|u(t)\|/\theta \right)\end{aligned}$$

where  $0 < \theta < 1$ .

This implies that (you may recall Theorem 4.18),  $\exists k_1, k_2 > 0$  s.t., for every  $\|x_0\| < k_1$  and for inputs such that  $\sup_{0 \leq t \leq \tau} \|u(t)\| \leq k_2$ , the solution  $x(t) \in B_{r_0}$ . Hence, the conclusion follows from Corollary 5.4.

**Lemma 5.2**

IF  $\exists C^1$  positive *semidefinite* function  $V(x)$  and  $\gamma > 0$  s.t.

$$\mathcal{H}(V, f, G, h, \gamma) := \frac{\partial V}{\partial x} f(x) + \frac{1}{2\gamma^2} \frac{\partial V}{\partial x} G(x) G^T(x) \left( \frac{\partial V}{\partial x} \right)^T + \frac{1}{2} h^T(x) h(x) \leq 0$$

on  $D$ , and ‘no solution of  $\dot{x} = f(x)$  can stay identically in  $S = \{x \in D : h(x) = 0\}$  other than the trivial solution  $x(t) \equiv 0$ ’,

THEN the origin of  $\dot{x} = f(x)$  is AS, and  $\exists k_1 > 0$  s.t. for each  $\|x_0\| \leq k_1$ , the system is small-signal finite-gain  $\mathcal{L}_2$  stable with the gain  $\leq \gamma$ .

- In this lemma the Lyapunov function  $V(x)$  came from the HJ inequality.
- How to interpret the phrase: ‘no solution of  $\dot{x} = f(x)$  can stay identically in  $S = \{x \in D : h(x) = 0\}$  other than the trivial solution  $x(t) \equiv 0$ ’? (Not clear in the textbook. Discuss it in the proof.) I would say ‘there is no *locally invariant set* in  $S$  except the origin’. By a locally invariant set  $L$  in  $S$ , I mean that, if  $x(0) = x_0 \in L$ , there exist nonnegative constants  $\delta_1$  and  $\delta_2$ , not both zero, s.t.  $x(t) \in L$  for  $t \in [-\delta_1, \delta_2]$ .

Locally invariant set  $\subset$   
Invariant set

*Proof.* The proof is done by Lemma 5.1 if we show that the origin of  $\dot{x} = f(x)$  is AS.

From the HJ inequality,

$$\frac{\partial V}{\partial x} f(x) \leq -\frac{1}{2\gamma^2} \frac{\partial V}{\partial x} G(x) G^T(x) \left( \frac{\partial V}{\partial x} \right)^T - \frac{1}{2} h^T(x) h(x) \leq -\frac{1}{2} h^T(x) h(x),$$

which seemingly means that the origin is stable, and by employing the LaSalle’s theorem, we have the conclusion. (In particular, the set  $\{x \in D : \frac{\partial V}{\partial x} f(x) = 0\} \subset S$ . In addition, since there is no locally invariant set in  $S$  except the origin, the only invariant set in  $S$  is the origin. So, the LaSalle’s theorem can be applied.) This is true if  $V(x)$  is positive definite.

However, since  $V(x)$  is assumed just positive semidefinite, we now prove that, from the assumption, the function  $V(x)$  is in fact positive definite.

Let  $r > 0$  be s.t.  $B_r \subset D$  (strict inclusion). Then, for each  $x \in B_r$ ,  $\exists \delta > 0$  s.t. the solution from  $x$ ,  $\phi(t; x)$ , stays in  $D$  for  $t \in [0, \delta]$ . Integrating the above inequality over  $[0, \delta]$ , we obtain

$$V(\phi(\delta; x)) - V(x) \leq -\frac{1}{2} \int_0^\delta \|h(\phi(t; x))\|_2^2 dt.$$

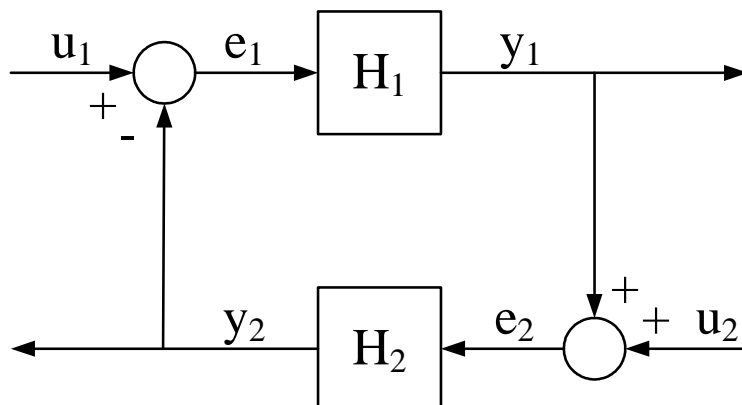
Using  $V(\phi(\delta; x)) \geq 0$ , we have

$$V(x) \geq \frac{1}{2} \int_0^\delta \|h(\phi(t; x))\|_2^2 dt.$$

Now suppose that  $\exists x \neq 0$  s.t.  $V(x) = 0$  (not positive definite). Then, the above implies that  $h(\phi(t; x)) = 0$  for  $t \in [0, \delta]$ . This means that  $x \neq 0$  is an element of locally invariant set. However, since the only element of the locally invariant set is the origin,  $x = 0$  (contradiction).

**Example 5.12** DIY.

#### IV. FEEDBACK SYSTEMS: THE SMALL-GAIN THEOREM



From the textbook,

- The formalism of input-output stability is particularly useful in studying stability of interconnected systems, since the gain of a system allows us to track how the norm of a signal increases or decreases as it passes through the system.
- See Figure 5.1.  $H_1 : \mathcal{L}_e^m \rightarrow \mathcal{L}_e^q$  and  $H_2 : \mathcal{L}_e^q \rightarrow \mathcal{L}_e^m$ . Suppose both system are finite-gain  $\mathcal{L}$  stable; that is,

$$\begin{aligned} \|y_{1,\tau}\|_{\mathcal{L}} &\leq \gamma_1 \|e_{1,\tau}\|_{\mathcal{L}} + \beta_1, & \forall e_1 \in \mathcal{L}_e^m, \forall \tau \in [0, \infty) \\ \|y_{2,\tau}\|_{\mathcal{L}} &\leq \gamma_2 \|e_{2,\tau}\|_{\mathcal{L}} + \beta_2, & \forall e_2 \in \mathcal{L}_e^q, \forall \tau \in [0, \infty) \end{aligned}$$

(finite-gain for simplicity).

- Suppose that the feedback system is *well defined* in the sense that for every pair of inputs  $u_1 \in \mathcal{L}_e^m$  and  $u_2 \in \mathcal{L}_e^q$ ,  $\exists$  unique outputs  $e_1, y_2 \in \mathcal{L}_e^m$  and  $e_2, y_1 \in \mathcal{L}_e^q$ .
- Define  $u = [u_1; u_2]$ ,  $y = [y_1; y_2]$ ,  $e = [e_1; e_2]$ . We ask if the feedback system, viewed as a mapping from  $u$  to  $y$  (or to  $e$  since they are equivalent in the sense of Exercise 5.21, your homework), is finite-gain  $\mathcal{L}$  stable.
- If the external input and output of interest is just  $u_1$  and  $y_1$ , why should we consider  $\mathcal{L}$  stability from both  $(u_1, u_2)$  to both  $(y_1, y_2)$ ? The reason is that, by considering all the pairs of inputs and outputs, any possibly hidden internal mode comes out. (See Exercise 5.20.)

**Theorem 5.6 (Small-gain Theorem)**

IF  $\gamma_1\gamma_2 < 1$ ,

THEN the feedback system is finite-gain  $\mathcal{L}$  stable.

*Proof.*

$$e_{1\tau} = u_{1\tau} - (H_2 e_2)_\tau, \quad e_{2\tau} = u_{2\tau} + (H_1 e_1)_\tau$$

Then,

$$\begin{aligned} \|e_{1\tau}\|_{\mathcal{L}} &\leq \|u_{1\tau}\|_{\mathcal{L}} + \|(H_2 e_2)_\tau\|_{\mathcal{L}} \leq \|u_{1\tau}\|_{\mathcal{L}} + \gamma_2 \|e_{2\tau}\|_{\mathcal{L}} + \beta_2 \\ &\leq \|u_{1\tau}\|_{\mathcal{L}} + \gamma_2 (\|u_{2\tau}\|_{\mathcal{L}} + \gamma_1 \|e_{1\tau}\|_{\mathcal{L}} + \beta_1) + \beta_2 \\ &= \gamma_1 \gamma_2 \|e_{1\tau}\|_{\mathcal{L}} + (\|u_{1\tau}\|_{\mathcal{L}} + \gamma_2 \|u_{2\tau}\|_{\mathcal{L}} + \beta_2 + \gamma_2 \beta_1) \end{aligned}$$

Since  $\gamma_1\gamma_2 < 1$ ,

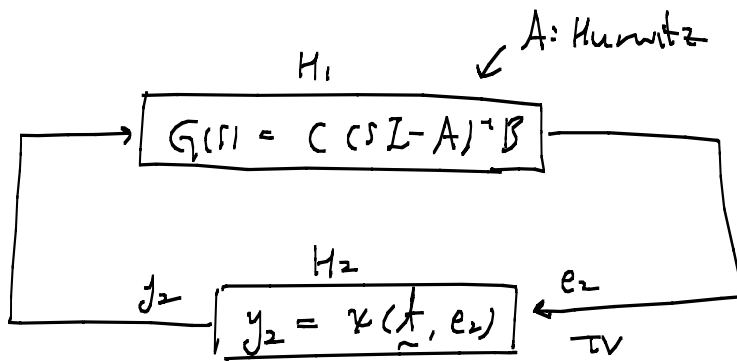
$$\|e_{1\tau}\|_{\mathcal{L}} \leq \frac{1}{1 - \gamma_1\gamma_2} (\|u_{1\tau}\|_{\mathcal{L}} + \gamma_2 \|u_{2\tau}\|_{\mathcal{L}} + \beta_2 + \gamma_2 \beta_1)$$

for all  $\tau \in [0, \infty)$ . Similarly,

$$\|e_{2\tau}\|_{\mathcal{L}} \leq \frac{1}{1 - \gamma_1\gamma_2} (\|u_{2\tau}\|_{\mathcal{L}} + \gamma_1 \|u_{1\tau}\|_{\mathcal{L}} + \beta_1 + \gamma_1 \beta_2)$$

for all  $\tau \in [0, \infty)$ . The proof completes since  $\|e\|_{\mathcal{L}} \leq \|e_1\|_{\mathcal{L}} + \|e_2\|_{\mathcal{L}}$ .

Example 5.17



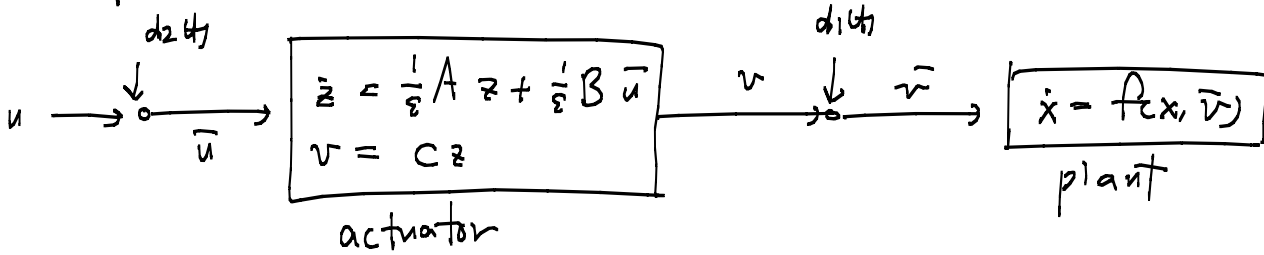
$$\|\gamma(t, e_2)\|_2 \leq \gamma_2 \|e_2\|_2$$

$H_1$ : finite-gain  $L_2$  gain:  $\gamma_1 = \sup \|G(j\omega)\|_2$

$H_2$ : " " :  $\gamma_2$

$\therefore \gamma_1 \gamma_2 < 1 \Rightarrow$  feedback sys. is finite-gain  $L_2$  stable

Example 5.14



$A$ : Hurwitz,  $\epsilon$ : small positive  
 $-CA^{-1}B = I$

$d_1, d_2 \in \mathcal{L}^p$   
 $\uparrow$   
 $\frac{p}{\epsilon}, \mathcal{L}^p (p \in [0, \infty])$

Goal: attenuate the effect of  $d_1(t), d_2(t)$  on  $x(t)$

i.e.,  $\begin{bmatrix} d_1(t) \\ d_2(t) \end{bmatrix} \xrightarrow[\text{gain} < \delta]{L \text{ stable}} x(t)$



\* Why can we ignore the fast actuator?

$$\begin{aligned} \text{Let } \varepsilon = 0. \quad z &= -A^{-1}B(u + d_2) \\ \Rightarrow v &= -CA^{-1}B(u + d_2) = u + d_2 \\ \Rightarrow \dot{x} &= f(x, \underbrace{u + d_1 + d_2}_{\equiv d(t)}) \end{aligned}$$

Controller  $u = \gamma(x)$  only considers  $x$  (not  $z$ ).

Ass:  $\gamma(x)$  is designed s.t.

$$\dot{x} = f(x, \gamma(x) + d)$$

$$\|x\|_{\mathcal{L}} \leq \lambda \|d\|_{\mathcal{L}} + \beta, \quad \lambda < \sigma$$

What about the actuator dynamics?

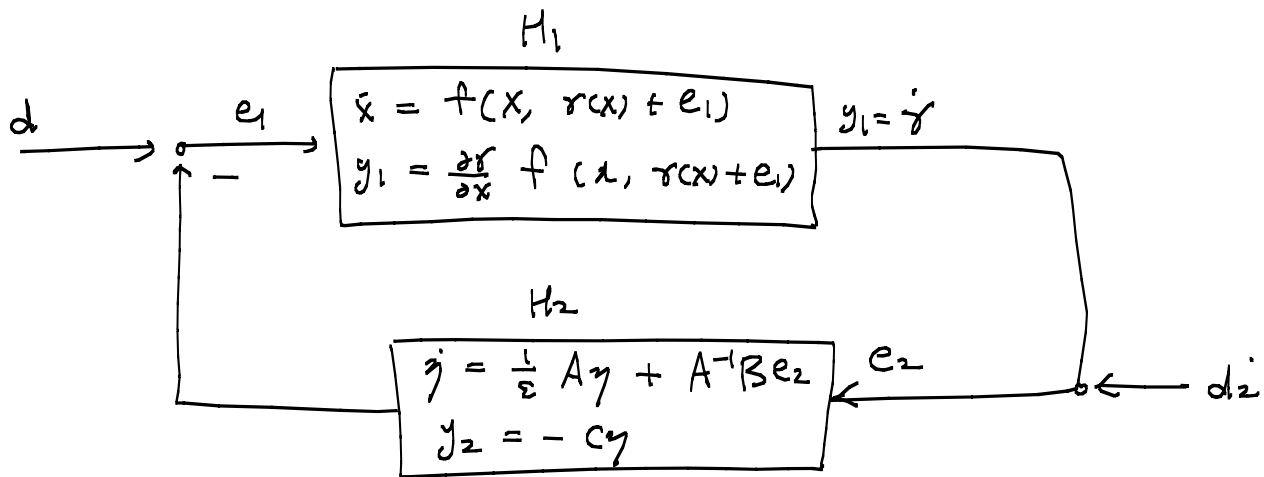
\* Robustness to unmodeled actuator dynamics

$$\begin{aligned} \dot{x} &= f(x, Cz + d_1) \\ \varepsilon \dot{z} &= Az + B(\gamma(x) + d_2) \end{aligned}$$

Ass:  $d_2(t)$  is differentiable &  $d_2 \in \mathcal{L}$

Trick begins by  $\eta \equiv z + A^{-1}B[\gamma(x) + d_2]$

$$\begin{aligned} \dot{x} &= f(x, C(\eta - A^{-1}B(\gamma(x) + d_2)) + d_1) \\ &= f(x, C\eta + \gamma(x) + d_1 + d_2) \\ \varepsilon \dot{\eta} &= A(\eta - A^{-1}B(\gamma + d_2)) + B(\gamma + d_2) + \varepsilon A^{-1}B(\dot{\gamma} + \dot{d}_2) \\ &= A\eta + \varepsilon A^{-1}B(\dot{\gamma} + \dot{d}_2) \\ (\dot{\gamma} &= \frac{\partial f}{\partial x} f(x, \gamma + d + C\eta)) \end{aligned}$$



Ans:  $\left\| \frac{\partial \gamma}{\partial x} \alpha f(x, \gamma(x) + e_1) \right\| \leq \alpha \|x\| + c_2 \|e_1\|$

H1:  $\|y_1\|_{\mathcal{L}} \leq \underbrace{(c_1 \lambda + c_2)}_{= \lambda_1} \|e_1\|_{\mathcal{L}} + \underbrace{c_1 \rho}_{= \rho_1}$

H2: Hurwitz LTI (Corollary 5.2)

$$\|y_2\|_{\mathcal{L}} \leq \frac{2\lambda_{\max}^2(p) \|B\|_2 \|C\|_2}{\lambda_{\min}(p)} \|e_2\|_{\mathcal{L}} + \rho_2$$

$$= \epsilon \lambda_f \|e_2\|_{\mathcal{L}} + \rho_2$$

where  $p(A/\epsilon) + (A/\epsilon)^T p = -I$

$$\Rightarrow p = \epsilon \bar{p}, \quad \bar{p}A + A^T \bar{p} = -I$$

$\therefore$  if  $\lambda_1 \epsilon \lambda_f < 1$  then ok

$\uparrow$  for sufficiently small  $\epsilon$ .