Nonlinear Control

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Intro

Goals of Course

- overview over moder nonlinear analyses and control concepts
- modern methodologies to analyze nonlinear systems
- apply a wide range of nonlinear control method learn the mathmatic basis

Differential equations $\dot{x} = f(x)$

Nonlinear differential equtiion $\dot{x} = f(t, x)$

System with input $\dot{x} = f(x, u)$

System with input and output $\dot{x} = f(x, u), \quad y = g(x, u)$

Input-output methods

Scope

[1] Khalil Nonlinear System, Prentice Hall, 2002

[2] Sepulchre Constructive Nonlinear Control, Springer-Verlag 1997

1 Differential equations

Consider differential equality

$$\frac{d}{dt}x(t) = \dot{x}(t) = f(x(t)),\ x(0) = x_0 \tag{1}$$

Where $f : D \to R^n,\ D \subset R^n$ is open, [here we should explain, what means open set].

Solution to 1 $x : I_{x_0} \to D,\ t \to x(t)$ is differentiable

Interval existence solution

Questions:

existence of solution

"how large" is I_{x_0}

uniqueness of solution

Usually we will add some restrictions on f functions, like continuous.

1.1 Existence of solutions

Definition. Function $f : D \to R^n$ is continuous at $x' \in D$ if for $\forall \epsilon > 0$ $\exists \delta > 0$ such that for $\forall x \in D$, $\|x - x'\| < \delta => \|f(x) - f(x')\| < \epsilon$

Function $f : D \to R^n$ is continuous on D if it's continuous at $\forall x' \in D$

Theorem 1.1 (Piano)**.** If $f : D \to R^n$ continuous, then for each $x_0 \in D$ $\exists x : (-\epsilon, \epsilon) \to D,\ \epsilon > 0$ satisfying (1).

Further, given a compact set $U \subset D$, then $\exists \alpha > 0$ s.t. $\forall x_0 \in U$ $\exists x : (-\epsilon, \epsilon) \to D$ satisfying (1).

Example. Consider equation $\dot{x}(t) = x(t)^2,\ \ x(0) = x_0 = 0$. Solution $x(t) = -\frac{1}{t-c},\ \ c = \frac{1}{x_0}$. In this example solution exist in interval $(-c, c)$.

But, what about the number of solutions? Which conditions we should add to garantie uniqueness of solution?

1.2 Uniquence of solutions

Definition. Function $f : D \to R^n$ is locally Lipshitz (continuous???) on D if $\forall x \in D$ there is a neighborhood $N(x) \subset D$ and $\exists L > 0$ s.t.

$$||f(x_1) - f(x_2)|| \leq L||x_1 - x_2|| \tag{2}$$

For all $x_1, x_2 \in N$.

- Lipschitz on $W \in D$ if (2) holds $\forall x_1, x_2 \in W$ (with same L)
- globally Lipschitz if (2) holds $\forall x_1, x_2 \in R^n$ (with same L)

We have

locally Lipschitz functions are continuous

differentiable functions are locally Lipschitz

locally Lipschitz functions are Lipschitz on each compact subset of D (Khalil Ex 3.19)

Lemma 1 (Cromwall)**.** Suppose that $0 \leq \phi(t) \leq c + L\int_0^t \phi(\tau)d\tau$, $c, L > 0$, ϕ – continuous. Then $\phi(t) \leq ce^{Lt}$.

Proof. $c + L\int_0^t \phi(\tau)d\tau := \psi(t)$, $\dot{\psi}(t) = L\phi(t) \leq L\psi(t)$.

Consider $\frac{d}{dt}\left(\psi(t)e^{-LT}\right) = e^{-Lt}\left(\dot{\psi}(t) - L\psi(t)\right) \leq 0$, thus $\psi(t)e^{-LT}$ is decreased, and as a result we have $\phi(t)e^{-Lt} \leq \psi(t)e^{-Lt} \leq \psi(0) = c$

$\square$

Theorem 1.2 (Picard Lindelof)**.** If function $f : D \to R^n$ is localy Lipschitz then for $\forall x_0 \in D$ $\exists! x : (-\epsilon, \epsilon) \to D,\ \epsilon > 0$ satisfying (1).

Proof. * existence from Piano theorem

Proof of uniqueness

Consider two solutions $x_1(.)$ and $x_2(.)$ to (1). $\dot{x}_1 - \dot{x}_2 = f(x_1) - f(x_2)$, $x_1(0) = x_2(0)$. Then we can integrate equality: $x_1(t) - x_2(t) = \int_0^t f(x_1(\tau)) - f(x_2(\tau))d\tau$. $|x_1(t) - x_2(t)| \leq \int_0^t |f(x_1(\tau)) - f(x_2(\tau))|d\tau \leq L\int_0^t |x_1(\tau) - x_2(\tau)|d\tau$. Now we can apply Cromwall's lemma with $c = 0$ and $\phi(t) = |x_1(t) - x_2(t)|$, then $\phi(t) \leq 0$, then $x_1(t) = x_2(t),\ \forall t \in (-\epsilon, \epsilon)$ $\square$

Example.

$$\dot{x} = \begin{cases} \sqrt{x}, & \text{if } x \geq 0 \\ 0, & \text{else } x < 0 \end{cases}$$

$$\textit{Solutions } x(t) = \begin{cases} \frac{1}{4}(t-c)^2, & \text{if } t \geq c \geq 0 \\ 0, & \text{else} \end{cases}$$

Global existance & uniqueness

- sufficient condition: f globally Lipschitz
- another sufficient condition: solution entirely lies in a coplex set
- forward completeness has equivalent Lyapunov-like characterization: system is forward-complete iff $\exists$ solution $V : R^n \to R \geq 0$ s.t. $\frac{\partial V}{\partial x} f(x) \leq V(x)$, $\forall x \in R^n$

1.3 Lyapunov stability

If functions $\dot{V}(x) < 0$, $\forall x \in D \setminus \{0\}$, then x^* is asymptotically stable.

Definition. Equilibrium point $x = 0$ is stable if $\forall \epsilon > 0$ $\exists \delta > 0$ s.t. from $\|x_0\| < \delta$ follows $\|x(t)\| \leq \epsilon, \ \forall t \geq 0$.

Definition. Equilibrium point $x = 0$ is asymptotically stable if it is stable and exist $\delta > 0$ s.t. from $\|x_0\| < \delta$ follows $\lim_{t\to\infty} x(t) \to 0$.

Theorem 1.3 (Lyapunov's direct method)**.** Let $x^* = 0 \in D$ be an equilibrium point of (1), i.e., $f(0) = 0$. Let $f : D \to R^n$ is continuous. If there exists a differentiable $V : D \to R$ s.t.

1. $V(x^*) = 0$, $V(x) > 0$, $\forall x \in D \setminus \{0\}$
2. $\dot{V}(x) = \frac{\partial V}{\partial x} f(x) \leq 0$, $\forall x \in D$

then $x^* = 0$ is stable.

Proof. Fix compact $U = \{x : V(x) \leq c\}$ s.t. $U \in D$. By Piano: exist $\alpha > 0$ s.t. any solution x with $x_0 \in U$ exists at least on the interval $[0, \alpha)$.

TODO proof is not full □

Lyapunovs direct method gives us:

- stability
- convergence (if $V < 0$)
- subset of the region of attraction (all compact $U = \{x : V(x) \leq c\} \in D$)
- existance of solution for all times

2 Nonlinear systems

In this section we consider function $f : R \times D \to R^n$, where $D \subseteq R^n$, and D is open.

$$\dot{x}(t) = f(t, x(t)), \quad t \geq t_0 \geq 0, \quad x(t_0) = x_0 \tag{3}$$

The origin $x^* \in D$ is an equilibrium point for (3), if $f(t, 0) = 0, \quad \forall t \geq 0$.

Remark: EP (equilibrium point) $x^* = 0$ can be translation of a nonzero solution.

Suppose $\overline{y}$ is a solution of $\dot{y} = g(t, y)$.

Change of coordinates: $x(t) = y(t) - \overline{y}(t)$, then $\dot{x}(t) = \dot{y}(t) - \dot{\overline{y}}(t) = g(t, x(t) + \overline{y}(t)) - \dot{\overline{y}}(t) := f(t, x(t))$. Since $\dot{\overline{y}}(t) = g(t, \overline{y}(t))$, then $f(t, 0) = 0, \ \forall t \geq 0$.

Existence and uniqueness of solution to (3):

- if f continuous, then exist local solution
- if f continuous and locally Lipschitz in x^*, then exist local unique solution

Now we need new stability definitions.

Definition. Point $x^* = 0$ is stable if $\forall \epsilon > 0$ and $\forall t_0 \geq 0, \ \exists \delta > 0$ s.t. from $||x_0|| < \delta$ follows $||x(t)|| < \epsilon$, $\forall t \geq t_0$.

Definition. Point $x^* = 0$ is uniformly stable if $\forall \epsilon > 0 \ \ \exists \delta > 0$, s.t $\forall t_0 \geq 0$, from $||x_0|| < \delta$ follows $||x(t)|| < \epsilon$, $\forall t \geq t_0$.

Definition. Point $x^* = 0$ asymptotically stable if it is stable and $\forall t_0 \geq 0 \ \ \exists c > 0$, s.t from $||x_0|| < c$ follows $\lim_{t \to \infty} ||x(t)|| \to 0$.

Definition. Point $x^* = 0$ uniformly asymptotically stable if it is uniformly stable and $\exists c > 0$, s.t $\forall t_0 \geq 0$ from $||x_0|| < c$ follows $\lim_{t \to \infty} ||x(t)|| \to 0$.

Definition. Convergence: $\forall \eta > 0 \ \ \forall t_0 \geq 0$, $\exists T > 0$ such that $\forall t \geq t_0 + T$ follows $||x(t)|| < \eta$.

Definition. Uniform convergence: $\forall \eta > 0 \ \ \exists T > 0$ such that $\forall t_0 \geq 0$ and $\forall t \geq t_0 + T$ follows $||x(t)|| < \eta$.

Example. Consider next equation:

$$\dot{x}(t) = -\frac{x(t)}{1+t}, \quad t_0 \geq 0$$

Solution $x(t) = x(t_0)\frac{1+t_0}{1+t}$. It is uniformly stable, because we can choose $\delta = \epsilon$. But does $x(t)$ convergence uniformly? Answer is no.

Definition. Point $x^* = 0$ is globally uniformly asymptotically stable if it is uniformly stable with $\delta \to \infty$ for $\epsilon \to \infty$ and $\forall c, \eta \quad \exists T > 0$ such that $\forall t_0 \geq 0$ from $||x_0|| < c$ follows $\|x(t)\| < \eta$, $\forall t \geq t_0 + T$.

2.1 Lyapunov's direct method

Consider some function $V : [0, \infty) \times D \to R$, $(t, x) \to V(t, x)$ such that $\dot{V}(t, x) = \frac{\partial}{\partial t}V(t, x) + \frac{\partial}{\partial x}V(t, x)f(t, x)$.

Theorem 2.1 (Lyapunov's direct method)**.** Let $f : [0, \infty) \times D \to R^n$ is continuous and let $x^* = 0$ be equilibrium point. If there is a differentiable function $V : [0, \infty) \times D \to R$ with:

- $W_1(x) \leq V(t, x) \leq W_2(x)$, $\forall t \geq 0, \ x \in D$

- $\dot{V}(t,x) \le 0,\ \forall t \ge 0,\ \ x \in D$

where $W_1, W_2 : D \to R$ continuous and positive definite, then $x^* = 0$ is uniformly stable.

If further $\dot{V}(t,x) \le -W_3(x),\ \forall t \ge 0,\ \ x \in D$ with $W_3 : D \to R$ continuous and positive definite, the $x^* = 0$ is uniformly asymptotically stable.

If $D = R^n$ and W_1 is radialy unbounded then $X^* = 0$ is globally uniformly asymptotically stable.

Example. Consider the equation $\dot{x}(t) = -(1+t)x(t)$.

Check function $V(t,x) = \frac{1}{2}x^2$ as candidate for Lyapunov's function. Then $W_1(x) = W_2(x) = \frac{1}{2}x^2$ and $\dot{V}(t,x) = -(1+t)x^2 \le -x^2(t) =: W_3(x)$. Then from theorem we have, that $X^* = 0$ is globally uniformly asymptotically stable.

2.2 Exponential stability

Definition. Point $X^* = 0$ is an exponentially stable EP of (3) if $\exists \lambda, c, k > 0$ s.t. $t \ge t_0 \ge 0$ and all $\|x_0\| < c$ follows $\|x(t)\| \le K\|x(t_0)\|e^{\lambda(t-t_0)}$.

Remark: from exponential stability follows uniformly asymptotically stability.

Lemma 2 (Auxiliary result)**.** Let $\dot{x}(t) = f(t, x(t))$, f scalar and $\dot{\xi}(t) \le f(t, \xi(t))$ with $\xi(t_0) \le x(t_0)$. Then $\xi(t) \le x(t) \quad \forall t \ge t_0$.

Theorem 2.2. Let $f : [0,\infty) \times D \to R^n$ be continuous and $x^* = 0 \in D$ be an EP.

If there is a differentiable function $V : [0,\infty) \times D \to R$ and constants $k_1, k_2, k_3, a > 0$ s.t.

1. $k_1||x||^a \le V(t,x) \le k_2\|x\|^a,\ \forall t \ge 0,\ \ x \in D$
2. $\dot{V}(t,x) \le -k_3\|x\|^a$

then $x^* = 0$ is exponentially stable.

If $D = R^n$, then X^* is globally exponential stable.

Proof. For $c > 0$ small enough, trajectories initialized in $\{x : k_2||x||^a < c\}$ remain bounded and in D. From 1) and 2) we can conclude $\dot{V} \le -\frac{k_3}{k_2}V$. Then from previous Lemma $V(t, x(t)) \le V(t_0, x(t_0))e^{-\frac{k_3}{k_2}(t-t_0)}$. Then $||x(t)|| \le [from(1)] \le \left(\frac{V(t,x(t))}{k_1}\right)^{\frac{1}{a}} \le \left(\frac{V(t_0,x(t_0))e^{-\frac{k_3}{k_2}(t-t_0)}}{k_1}\right)^{\frac{1}{a}} \le \left(\frac{k_2||x(t_0)||^a e^{-\frac{k_3}{k_2}(t-t_0)}}{k_1}\right)^{\frac{1}{a}} = \left(\frac{k_2}{k_1}\right)^{\frac{1}{a}} ||x(t_0)||e^{-\frac{k_3}{k_2 a}(t-t_0)}$ □

Example. Consider the equation $\dot{x}(t) = -(1+t)x(t)$.

Here $V(t,x) = \frac{1}{2}x^2$ then X^* is exponentially stable.

2.3 Comparison function

Definition. A function $\alpha : [0, \delta) \to [0, \infty)$ is (of) "class K" if it is continuous, strictly increasing, and $\alpha(0) = 0$.

Definition. A function $\alpha : [0, \delta) \to [0, \infty)$ is "class K_∞" if $\alpha \in K$ and $\lim_{r\to\infty} \to \infty$.

Example. Function $\alpha(r) = \tan^{-1}(r)$ – class K

Function $\alpha(r) = r^k$ – class K_∞

Definition. A function $\beta : [0, \delta) \times [0, \delta) \to [0, \infty)$ is "class KL if it is continuous , $\beta(\cdot, s) \in K$ for all fixed s, and for each fixed r, $\beta(r, \cdot)$ is strictly decreasing: $\lim_{s\to\infty} \beta(r, s) = 0$

Example. Function $\beta(x, s) = max(r, r^2)e^{-s}$ belongs to class KL.

Properties of comparison functions:

- If $\alpha \in K$ on $[0, \delta)$, then α^{-1} is defined on $[0, \alpha(\delta))$ and $\alpha^{-1} \in K$.
- If $\alpha \in K_\infty$, then $\alpha^{-1} \in K_\infty$
- If $\alpha_1, \alpha_2 \in K$, then $\alpha_1 \circ \alpha_2 \in K$ (same for K_∞)
- If $\alpha_1, \alpha_2 \in K$, $\beta \in KL$ then $\alpha_1(\beta(\alpha_2(r), s)) \in KL$

Now we consider comparison functions and stability definitions.

Lemma 3. The equilibrium $x^* = 0$ of $\dot{x}(t) = f(t, x(t))$ is uniformly stable iff $\exists \alpha \in K$ and $c > 0$ s.t. $\forall t \geq t_0$, $\forall ||x(t_0)|| < c$ and $||x(t)|| \leq \alpha(||x(t_0)||)$.

(only sufficiency). Given $\epsilon > 0$ choose $\delta < \min(c, \alpha^{-1}(\epsilon))$. Then from $||x(t_0)|| < \delta$ follows $||x(t)|| \leq \alpha(||x(t_0)||) < \alpha(\alpha^{-1}(\epsilon)) = \epsilon$. □

Lemma 4. The equilibrium $x^* = 0$ of $\dot{x}(t) = f(t, x(t))$ is uniformly asymptotically stable iff $\exists \beta \in KL$ and $c > 0$ s.t. $\forall t \geq t_0$, $\forall ||x(t_0)|| < c$ and $||x(t)|| \leq \beta(||x(t_0)||, t - t_0)$.

(only sufficiency). Let $||x(t_0)|| < c$. Then $||x(t)|| \leq \beta(||x(t_0)||, t - t_0) < \beta(c, t - t_0)$. This mean uniform convergence. $||x(t)|| \leq \beta(||x(t_0)||, t - t_0) < \beta(||x_{t_0}||, 0)$. This gives us uniform stability. □

Lemma 5. The equilibrium $x^* = 0$ of $\dot{x}(t) = f(t, x(t))$ is globally uniformly asymptotically stable iff previous lemma holds for all $x_0 \in R^n$.

Now consider comparison functions and Lyapunov functions

If $W : R^n \to R$ is continuous and positive definite, then $\forall r > 0 \ \ \exists \alpha_1, \alpha_2 \in K$ s.t. $\alpha_1(||x||) \leq W(x) \leq \alpha_2(|x||)$ for all $x \in B_r(0) = \{x|||x|| \leq r\}$.

If W is radially unbounded, then $\exists \alpha_1, \alpha_2 \in K_\infty$ s.t. $\alpha_1(||x||) \leq W(x) \leq \alpha_2(|x||)$ for all $x \in R^n$.

Lemma 6 (Auxiliary)**.** Consider $\dot{y} = \alpha(y), \quad y(t_0) = y_0 > 0, \ \alpha \in K$. Then $\exists \beta \in KL$ s.t. $y(t) = \beta(y_0, t - t_0)$.

Sketch of proof of Lyapunov's direct method:

$$\begin{cases} W_1(x) \leq V(t,x) \leq W_2(x) \\ \dot{V} \leq -W_3(x) \end{cases}$$

Where W_1, W_2, W_3 – continuous and positive defined.

Then $\exists \alpha_1, \alpha_2, \alpha_3 \in K$ such that $\alpha_1(||x||) \leq V(t,x) \leq \alpha_2(||x||)$ and $\dot{V}(t,x) \leq -\alpha_3(||x||)$.

Proof uniform stability:

$||x(t)|| \leq \alpha_1^{-1}(V(t, x(t)) \leq [\alpha_1 \in K] \leq \alpha_1^{-1}(V(t_0, x(t_0)) \leq \alpha_1^{-1}(\alpha_2(||x(t_0)||))$.

Proof uniform convergence

$\dot{V} \leq -\alpha_3(||x||) \leq -\alpha_3(\alpha_2^{-1}(V))$. We know, that $\alpha_3 \circ \alpha_2^{-1} \in K$. By comparison lemma, $V(t, x(t)) \leq W(t)$, where W solves $\dot{W} = -\alpha_3(\alpha_2^{-1}(W))$ with $W(t_0) = V(t_0, x(t_0))$. By auxiliary lemma $\exists \beta \in KL$ s.t. $V(t, x(t)) \leq \beta(V(t_0, x(t_0)), t - t_0)$, then $||x(t)|| \leq \alpha_1^{-1}(V(t, x(t))) \leq \alpha_1^{-1}(\beta(V(t_0, x(t_0)), t - t_0)) \leq \alpha_1^{-1}(\beta(\alpha_2(||x(t_0)||), t - t_0)) =: \bar{\beta}(||x(t_0)||, t - t_0)$. From this follows uniform asymptotic stability since $\bar{\beta} \in KL$.

2.4 Converse theorems

Theorem 2.3. Let $X^* = 0$ be an EP of $\dot{x}(t) = f(t, x(t))$ with $f : [0, \infty) \times R^n \to R^n$ continuously differentiable and $\frac{\partial f}{\partial x}$ bounded in R^n, uniformly in t ($||\frac{\partial f}{\partial x}(t,x)|| \leq L$ for all $x \in R^n, \ t \geq 0, \ L > 0$.

If $x^* = 0$ is globally exponentially stable, then exists differentiable $V : [0, \infty) \times R^n \to R$ and $c_1, c_2, c_3, c_4 > 0$ s.t. $c_1||x||^2 \leq V(t,x) \leq c_2||x||^2$, $\dot{V}(t,x) \leq -c_3||x||^2$ and $\left\|\frac{\partial V}{\partial x}\right\| \leq c_4||x||$.

Proof. Let $\Phi(\tau; t, x)$ – solution to $\dot{x}(t) = f(t, x(t))$ which is static at (t, x).

$V(t,x) = \int_t^{t+\delta} \Phi^T(\tau; t, x)\Phi(\tau; t, x) d\tau, \quad \delta > 0$. Upper bound: $V(t,x) = \int_t^{t+\delta} ||\Phi(\tau; t, x)||_2^2 d\tau \leq$ [*exponential stability*] $\leq \int_t^{t+\delta} k^2 e^{-2\lambda(\tau - t)} d\tau ||x||_2^2 = \frac{k^2}{2\lambda}(1 - e^{-2\lambda\delta})||x||_2^2$.

Lower bound: since $\left\|\frac{\partial V}{\partial x}\right\| \leq L$, then $||f(t,x)||_2 \leq L||x||_2$. Thus by comparation lemma $||\Phi(\tau; t, x)||_2^2 \geq ||x||_2^2 e^{-2L(\tau - t)}$. Set it in $V(t,x) \geq \int_t^{t+\delta} e^{-2L(\tau - t)} d\tau ||x||_2^2 = \frac{1}{2L}(1 - e^{-2L\delta})||x||_2^2$.

Decrease conditions: $\dot{V}(t,x) = \cdots \leq -(1 - k^2 e^{-2\lambda\delta})||x||_2^2$. $\square$

3 System with inputs

Consider equation:

$$\dot{x}(t) = f(x(t), u(t)), \quad x(0) = x_0 \tag{4}$$

where $f : R^n \to R^n$.

Assumption: f in locally Lipschitz.

Exogenous signal $u : R \to R^n$.

Input can be "bad" (disturbance) or "good" (control).

3.1 Input-to-state stability

Motivation: LTI system $\dot{x} = Ax + Bu, \quad x(0) = x_0$.

Solution: $x(t) = e^{At}x_0 + \int_0^t e^{A(t-\tau)}Bu(\tau)d\tau$. If A is Hurwitz, then $||e^{At}|| \leq ce^{-\lambda t}$ for some $c, \lambda > 0$.

How large can x grow for some bounded u? $||x(t)|| \leq ||e^{At}||||x_0|| + \int_0^t ||e^{A(t-\tau)}||||B||||u(\tau)||d\tau \leq e^{-\lambda t}c||x_0|| + \int_0^t e^{-\lambda(t-\tau)}c||B||||u(\tau)||d\tau = ce^{-\lambda t}||x_0|| + (1 - e^{-\lambda t})\ \frac{c}{\lambda}||B|| \sup_{\tau\in[0,t]} ||u(\tau)||$.

- $ce^{-\lambda t}||x_0||$ class KL in $(||x_0||, t)$
- $(1 - e^{-\lambda t})$ less than 1
- $\frac{c}{\lambda}||B|| \sup ||u(\tau)||$ class K

If $\sup_{\tau\in[0,t]} ||u(\tau)||$ is bounded than $\dot{x}$ remains bounded. Even more: the smaller $\sup_{\tau\in[0,t]} ||u(\tau)||$, the smaller $||x(t)||$.

Definition. System (4) is input-to-state stable (ISS) if $\exists\beta \in KL$, $\gamma \in K$ s.t. $\forall x_0 \in R^n$, $\forall t \geq 0$ follows $||x(t)|| \leq \beta(||x_0||, t) + \gamma(\sup_{\tau\in[0,t]} ||u(\tau)||)$.

Remarks:

- From ISS follows O-GAS (global assymptotical stability of $x = 0$ for $\dot{x} = f(x, 0)$)
- γ can be interpreted as "gain" w.r.t. u
- if $\lim_{t\to\infty} u(t) = 0$ then $\lim_{t\to\infty} x(t) = 0$

Example. Consider equation $\dot{x} = -x + xu$. System is O-GASS, not ISS (for example $u \equiv \alpha \Rightarrow \dot{x} = x(\alpha - 1)$ all solution diverge).

Example. Consider equation $\dot{x} = -3x + (1+2x^2)u$. System is O-GASS, not ISS (for example $u \equiv 1$, $x_0 = 2$, $x(t) = \frac{3-e^t}{3-2e^t}$ has a finite escape time.

Theorem 3.1. Suppose that there exists a continuously differentiable function $V : R^n \to R$ and $\alpha_1, \alpha_2 \in K_\infty$ and $\alpha_3, \rho \in K$ such that $\alpha_1(||x||) \leq V(x) \leq \alpha_2(||x||)$, $\forall x \in R^n$ and $\frac{\partial V}{\partial x} f(x,u) \leq -\alpha_3(||x||)$, $\forall x : ||x|| \geq \rho(||u||)$. Then (4) is ISS with $\gamma = \alpha_1^{-1} \circ \alpha_2 \circ \rho$

Proof. Idea: same as Lyapunovs direct method when x is "outside" of ball $\{x | ||x|| \leq \rho(||u||)\}$

TODO Picture □

Example. Consider equality $\dot{x} = -x^3 + u$. Let $V(x) = \frac{1}{2}x^2$, then $\dot{V} = -x^4 + xu = [0 < \Theta < 1] = -(1-\Theta)x^4 - \Theta x^4 + xu \leq -(1-\Theta)x^4$ for all $x : ||x|| \geq \left(\frac{||u||}{\Theta}\right)^{\frac{1}{3}}$. Thus, system is ISS with $\gamma(v) = \rho(v) = \left(\frac{v}{\Theta}\right)^{\frac{1}{3}}$.

Remarks:

- Existence of V is both neccessary and sufficient for ISS;
- (**??**) is equivalent to $\frac{\partial V}{\partial x} f(x,u) \leq -\alpha_4(||x||) + \alpha_5(||u||), \forall x, u$ for some $\alpha_4, \alpha_5 \in K$;
- If $x_1 = 0$ is a globally asymptotically stable EP of Σ_1 and Σ_2 is ISS w.r.t. "input" x_1, then $(x_1, x_2) = (0,0)$ is a globally asymptotically stable EP for the cascaded system.

Theorem 3.2. Assume that:

- f is globally Lipschitz;
- $x = 0$ is a globally exponentially stable EP for $\dot{x} = f(x,0)$

Then the system (4) is ISS.

Proof. Sketch: $\exists$ continuous differentiable V:

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

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

$$||\frac{\partial V}{\partial x}|| \leq c_4||x||$$

Then:
$\frac{\partial V}{\partial x} f(x,u) = \frac{\partial V}{\partial x} f(x,0) + \frac{\partial V}{\partial x}(f(x,u) - f(x,0) \leq -c_3||x||^2 + c_4||x||L||u|| = -c_3(1-\theta)||x||^2 - \theta||x||^2 + c_4L||x||||u|| \leq -c_3(1-\theta)||x||^2$
if $||x|| \geq \frac{c_4 L}{\theta c_3}||u||$. □

3.2 Control Lyapunov functions

Motivation: Lyapunov theory for control systems.

(input affine systems)
$\dot{x} = f(x) + \sum_{i=1}^{m} g_i(x)u_i = f(x) + G(x)u$,
$f : R^n \to R^n, g : R^n \to R^n, G : R^n \to R^{n\times m}$
$u : t \to u(t), R \to R^m$ is a control signal (decision variable).

Definition. A function $V : R^n \to R$ is a control Lyapunov function (CLF) if it's differentiable positive definite, radially unbounded and

$$\forall x \neq 0 \quad \inf_u (\nabla V(x) \cdot (f(x) + G(x)u)) < 0 \tag{5}$$

Remark:
Concept can be generalized to systems $\dot{x} = f(x, u)$. Then 5 becomes

$$\forall x \neq 0 \quad \inf_u (\nabla V(x) \cdot f(x, u)) < 0$$

Theorem 3.3 (Artstein)**.** There exists $k : R^n \to R^m$ (state feedback) which is continuous on $R^n \backslash \{0\}$ s.t. $x^* = 0$ is globally asymptotically stable EP for $\dot{x} = f(x) + G(x)k(x)$ iff there exists a CLF.

How to find CLFs?

Proposition:
Condition (5) is equivalent to

$$\forall x \neq 0, \quad \frac{\partial V}{\partial x} G(x) = 0 \implies L_f V(x) < 0 \tag{6}$$

Remark:

$$\frac{\partial V}{\partial x} G(x) = (\nabla V(x) g_1(x), \ldots \nabla V(x) g_m(x)) =: L_G V(x)$$
$$(6) \iff \forall x \neq 0, \quad L_f V(x) \geq 0 \implies L_G V(x) \neq 0$$

Proof. $\Longleftarrow$:
Assume (6) holds. Then:

$$\inf_u (\nabla V(x) \cdot (f(x) + G(x)u)) = \inf_u L_f V(x) + L_G V(x) u < 0$$

Why?

- If $L_G V(x) = 0$, then by (6) $L_f V(x) < 0$;
- If $L_G V(x) \neq 0$, then (at least) for one i we have $\nabla V(x) \cdot g_i(x) \neq 0 \implies$ set $u_i = -c\nabla V(x) \cdot g_i(x)$.

$\Longrightarrow$:
If (5) holds for some x with $L_G V(x) = 0$, then we must have $L_f V(x) < 0$. $\square$

Example (discontinuous control):

$$\dot{x} = \begin{cases} 1 - u, & u \geq 1 \\ -1 - u, & u \leq -1 \\ 0, & else \end{cases}$$

If you want to move the system you need to apply control $|u| \geq 1$.
Using

$$u(x) = \begin{cases} x + 1, & x > 0 \\ x - 1, & x \leq 0 \end{cases}$$

results in closed loop $\dot{x} = -x$ - asymptotically stable.
$V(x) = x^2$ is a CLF.

Theorem 3.4. There exists a continuous $k : R^n \to R^m$, smooth on $R^n \backslash \{0\}$ s.t. $x^* = 0$ is globally asymptotically stable EP for $\dot{x} = f(x) + G(x)k(x)$ iff:

- there exists a (smooth)CLF V;
- $\forall \varepsilon > 0 \;\; \exists \delta > 0 : \;\; \forall x : 0 < ||x|| < \delta$
  $\exists u \in R^m : ||u|| < \varepsilon$ s.t. $L_f V(x) + L_G V(x) u < 0$

How to construct a globally stabilizing state feedback k from knowledge of a CLF?

"Sontag's formula"
Fix $c \geq 0, a(x) := L_f V(x), b(x) := (L_G V(x))^T$

$$k(x) = \begin{cases} -cb(x) - \frac{a(x) + \sqrt{a(x)^2 + (b(x)^T b(x))^2}}{b(x)^T b(x)}, & b(x) \neq 0 \\ 0, & b(x) = 0 \end{cases}$$

Proposition: Let $V : R^n \to R$ be a CLF and k as above. Then $x^* = 0$ is globally asymptotically stable EP for $\dot{x} = f(x) + G(x)k(x)$

Proof. $\dot{V} = L_f V(x) + L_G V(x) k(x) = a(x) - cb(x)^T b(x) - \frac{a(x) + \sqrt{a(x)^2 + (b(x)^T b(x))^2}}{b(x)^T b(x)} b(x)^T b(x) = -cb(x)^T b(x) - \sqrt{a(x)^2 + (b(x)^T b(x))^2} < 0 \;\; \forall x \neq 0$ s.t. $L_G V(x) \neq 0$

$\dot{V} = L_f V(x) + L_G V(x) \cdot 0 < 0 \;\; \forall x \neq 0$ s.t. $L_G V(x) = 0$ (since V is CLF)

$\Longrightarrow$ V - Lyapunov function $\Longrightarrow$ ... $\square$

Remarks:

- Sontag's formula is smooth on $R^n \backslash \{0\}$;
- Sontag's formula is continuous at $x = 0$ iff small control property holds.

$$\forall x \neq 0 : \inf_u \frac{\partial V}{\partial x} f(x,u) < 0 \ \dot{x} = f(x) + G(x)u$$

So this leads to

$$\forall x \neq 0 L_G V(x) = 0 \Rightarrow L_f V(x) \neq 0$$

Remark: The last formula is "optimal" if minimize:

$$\int_0^\infty \frac{1}{2} p(x) b(x)^T b(x) + \frac{1}{2p(x)} u^T u dt$$

$b(x) := (L_G V(x))^T$

where $c > 0$

$$p(x) = \begin{cases} c + \frac{a(x) + \sqrt{a(x)^2 + (b(x)^T b(x))^2}}{b(x)^T b(x)} & b(x) \neq 0 \\ c & b(x) = 0 \end{cases}$$

It still works if $u = \lambda h(x)$ with $\lambda \in [\frac{1}{2}; \infty)$ is applied (large "gain margin")

4 Backstepping

Integrator backstepping

$$\dot{x}_1 = f_1(x_1) + g_1(x_1) x_2 \qquad (7)$$
$$\dot{x}_2 = u$$

where $x_1 \in \mathbb{R}^m,\ x_2,\ u \in \mathbb{R}$ (single input)

image to be inserted

Assumption: we know (smooth) "feedback" $\alpha_1 : \mathbb{R}^n \to \mathbb{R}$, and positive definite, differentiable $v_1 : \mathbb{R}^m \to \mathbb{R}$

s.t. $L_{f_1 + g_1 \alpha_1} V_1(x)$ is negative definite $\Rightarrow$ origin of $\dot{x_1} = f_1(x_1) + g_1(x_1)\alpha_1(x_1)$ is asymptotically stable

Goal: Compute feedback $u = k(x)$ which stabilises (7). Backstepping constructs $u = \alpha_2(x_1, x_2)$ s.t. $(e_1, e_2) = (x_1 - 0, x_2 - \alpha_1(x_1)) = 0$ error coordinates

Rewrite (7) :

$$\dot{x}_1 = f_1(x_1) + g_1(x_1)\alpha_1(x_1) + g_1(x_1)(x_2 - \alpha_1(x_1))$$

$$\dot{x}_2 = u$$

image to be inserted

In error coordinates

$$\dot{e}_1 = f_1(e_1) + g_1(e_1)\alpha_1(e_1) + g_1(e_1)e_2 \tag{8}$$
$$\dot{e}_2 = u - \dot{\alpha_1} = u - \frac{\partial \alpha_1}{\partial e_1}\dot{e_1}$$

"backstepping" α_1 through the integrator

Define $V_2(e_2) := \frac{1}{2}e_2^2$, and

$$V(e_1, e_2) = V_1(e_1) + V_2(e_2)$$
$$\dot{V}(e_1, e_2) = \frac{\partial V_1}{\partial e_1}(f_1(e_1) + g_1(e_1)\alpha_1(e_1)) + \frac{\partial V_1}{\partial e_1}g_1(e_1)e_2 + \frac{\partial V_2}{\partial e_2}(u - \dot{\alpha_1})$$

as far as $L_{f_1+g_1\alpha_1}V_1$ -negative definite and $\frac{\partial V_2}{\partial e_2} \to e_2$

Choose

$$u = (-\frac{\partial V_1}{\partial e_1}g_1(e_1) + \dot{\alpha_1})(\text{"}canaling\ terms\text{"}) - k_2e_2(\text{"}stabilizing\ term\text{"}),\ k_2 > 0 \tag{9}$$

$\Rightarrow$ Then $\dot{V}(e_1, e_2) = L_{f_1+g_1\alpha_1}V_1(e_1) - k_2e_2^2 < 0,\ \forall(e_1, e_2) \neq 0$

$\Rightarrow (e_1, e_2) = (0, 0)$ is an asymptotically stable EP for (8) with u as in (9)

Remark: $(e_1, e_2) \to (0, 0)$ doesnot necessarily imply that $(x_1, x_2) \to 0$ for $u = \alpha_2(x_1, x_2) = -\frac{\partial V_1}{\partial x_1}g_1(x_1) + \dot{\alpha_1} - k_2(x_2 - \alpha_1(x_1))$

where $u \leftarrow$ (9) the original coordinates and $\dot{\alpha_1} \leftarrow \frac{\partial \alpha_1}{\partial x_1}(f_1(x_1) + g_1(x_1)x_2)$

But $(x_1, x_2) = (0, 0)$ is asymptotically stable if $\alpha_1(0) = 0$ why? $(e_1, e_2) \to 0 \Rightarrow x_1 \to 0$ $x_2 \to \alpha_1(0) = 0$

Example.

$$\dot{x_1} = x_1x_2$$
$$\dot{x_2} = u$$

Choose $\alpha_1(x_1) = -k$ $(k > 0) \to \dot{x_1} = -kx_1 \Rightarrow V_1(x_1) = \frac{1}{2}x_1^2$

Then:

$$e_1 = x_1 - 0,\ \dot{e_1} = e_1(e_2 - k)$$
$$e_2 = x_2 + k,\ e_2 = u$$

Backstepping yields: $u = -e_1^2 - k_2 e_2,\ k_2 > 0 \Rightarrow (e_1, e_2) = (0,0)$ is asymptotically stabilized

$(x_1, x_2) = (0, -k)$ is asymptotically stabilized

Can we choose different α_1 s.t. $(x_1, x_2) = (0,0)$ is stabilized?

Yes, e.g.

$$\alpha_1(x_1) = -x_1^2 \Rightarrow \dot{x_1} = -x_1^3,\ V_1(x_1) = \frac{1}{2}x_1^2$$

So we have equations

$$e_1 = x_1 - 0,\ \dot{e_1} = e_1(e_2 - e_1^2)$$
$$e_2 = x_2 + x_1^2,\ \dot{e_2} = u + 2e_1^2(e_2 - e_1^2)$$

Backstepping results in

$$u = -e_1^2 - 2e_1^2(e_2 - e_1^2) - k_2 e_2,\ k_2 > 0 \Rightarrow (e_1, e_2) \to (0,0),\ (x_1, x_2) \to (0,0)$$

Generalization-1

$$\dot{x_1} = f_1(x_1) + g_1(x_1)x_2$$
$$\dot{x_2} = f_2(x_1, x_2) + g_2(x_1, x_2)u$$

Assumption: $g_2(x_1, x_2) \neq 0,\ \forall x_1, x_2 \Rightarrow$ Input transformation: $u = \frac{1}{g_2(x_1,x_2)}(V - f_2(x_1, x_2)) \Rightarrow \dot{x_1} = f_1(x_1) + g_1(x_1)x_2,\ \dot{x_2} = V \Rightarrow$ can apply integrator backstepping to determine V results in

$$u = \alpha_2(x_1, x_2) = \frac{1}{g_2(x_1, x_2)}(-\frac{\partial V_1}{\partial x_1}g_1(x_1) + \dot{\alpha_1} - k_2(x_2 - \alpha_1(x_1)) - f_2(x_1, x_2))$$

<u>Generalization 2:</u> (Backstepping through 2 integrators)

$$\dot{x}_1 = f_1(x_1) + g_1(x_1)x_2, \quad x_1 \in R^{n_1}$$
$$\dot{x}_2 = f_2(x_1, x_2) + g_2(x_1, x_2)x_3, \quad x_2, x_3 \in R$$
$$\dot{x}_3 = f_3(x_1, x_2, x_3) + g_3(x_1, x_2, x_3)u, \quad u \in R$$

Assumption: g_2, g_3 nowhere zero.

Shown before: $\exists \alpha_2$: for $x_3 = \alpha_2(x_1, x_2)$ $\ (e_1, e_2) \to 0$
Thus $e_3 := x_3 - \alpha_2(x_1, x_2)$

Input transformation:

$$u = \frac{1}{g_3(x_1, x_2, x_3)}(V - f_3(x_1, x_2, x_3))$$

$\implies \dot{x}_1 = \ldots, \dot{x}_2 = \ldots, \dot{x}_3 = V \implies$ can apply backstepping once more.

In "error" coordinates:

$$\dot{e}_1 = f_1(e_1) + g_1(e_1)(e_2 + \alpha_1(e_1)$$

$$\dot{e}_2 = f_2(e_1, e_2 + \alpha_1(e_1)) + g_2(e_1, e_2 + \alpha_1(e_1))(e_3 + \alpha_2(e_1, e_2 + \alpha_1(e_1))) - \dot{\alpha}_1$$

$$\dot{e}_3 = V - \dot{\alpha}_2$$

Define $V_3(e_3) = \frac{1}{2}e_3^2, V(e_1, e_2, e_3) = \sum\limits_{i=1}^{3} V_i(e_i)$

$\dot{V}(e_1, e_2, e_3) = \underline{\frac{\partial V_1}{\partial e_1}(f_1(e_1) + g_1(e_1)(e_2 + \alpha_1(e_1)) + e_2(f_2(e_1, e_2 + \alpha_1(e_1)) + g_2(e_1, e_2 + \alpha_1(e_1))(e_3 +}$ $\underline{\alpha_2(e_1, e_2 + \alpha_1(e_1))) - \dot{\alpha}_1)} + e_3(V - \dot{\alpha}_2)$

All the underlined terms were designed (previously) to be $= L_{f_1+g_1\alpha_1}V_1(e_1) - k^2e_2^2 < 0$

So: $\dot{V}(e_1, e_2, e_3) = L_{f_1+g_1\alpha_1}V_1(e_1) - k^2e_2^2 + e_2g_2(e_1, e_2 + \alpha_1(e_1))e_3 + e_3(V - \dot{\alpha}_2)$

Structurally it is exactly the same as it was in backstepping through 1.

Choose:

$$V = \dot{\alpha}_2 - e_2g_2(e_1, e_2 + \alpha_1(e_1)) - k_3e_3$$

$\dot{\alpha}_2 - e_2g_2(e_1, e_2 + \alpha_1(e_1))$ - "cancelling terms".
k_3e_3 - "stabilizing term".

In original coordinates:

$$u = \frac{1}{g_3(x_1, x_2, x_3)}(\dot{\alpha}_2 - (x_2 - \alpha_1(x_1))g_2(x_1, x_2) - k_3(x_3 - \alpha_2(x_1, x_2)) - f_3(x_1, x_2, x_3))$$

We need α_1, α_2 to compute u.

<u>General backstepping recursion:</u>
Systems in "strict feedback form":
$\dot{x}_1 = f_1(x_1) + g_1(x_1)x_2, \quad x_1 \in R^{n_1}$
$\dot{x}_2 = f_2(x_1, x_2) + g_2(x_1, x_2)x_3$
$\ldots$
$\dot{x}_k = f_k(x_1, \ldots x_k) + g_k(x_1, \ldots x_k)u, \quad x_2, \ldots x_k, u \in R$
$g_2, \ldots g_k$ nowhere zero, f_i, g_i (sufficiantly) smooth, as it is needed in α_i.

Backstepping recursion:

1. "Input data": a CLF V_1 for $\dot{x}_1 = f_1(x_1) + g_1(x_1)u_1$ with a (smooth) feedback $u_1 = \alpha_1 x_1$ which as. stabilizes the origin of $\dot{x}_1 = f_1(x_1) + g_1(x_1)u_1$.

2. for $i = 2, \dots k$:

   construct a CLF $V_i(e_i) = \frac{1}{2}e_i^2, V = \sum\limits_{j=1}^{i} V_j(e_j)$ and a feedback α_1 which as. stabilizes origin of $(e_1, \dots e_i) = (x_1, x_2 - \alpha_1(x_1), \dots, x_i - \alpha_{i-1}(x_1, \dots x_{i-1}))$

$$\alpha_i(x_1, \dots x_i) = \frac{1}{g_i}(\dot{\alpha}_{i-1} - \frac{\partial V_{i-1}}{\partial e_{i-1}} g_{i-1} - k_i(x_i - \alpha_{i-1} - f_i)$$

3. apply $u = \alpha_k(x_1, \dots x_k)$

Backstepping and CLFs:
Backstepping is sensitive to uncertainties in f_i, g_i (due to cancelling terms)
$\implies$ Sontag's formula is more practical $\implies$ we can use it since V is CLF.

Error system is input affine (using input transformation)
$\dot{e} = f(e) + g(e)V$

with $f(e) = \begin{pmatrix} f_1(e_1) + g_1(e_1)(e_2 + \alpha_1(e_1)) \\ \dots \\ -\alpha_{k-1} \end{pmatrix}, g(e) = \begin{pmatrix} 0 \\ 0 \\ \dots \\ 0 \\ 1 \end{pmatrix}$

Claim:
$V(e) = \sum\limits_{i=1}^{k} V_i(e_i)$ is a CLF.

Proof. For input affine systems we need to show $L_g V = 0 \implies L_f V < 0, \quad \forall e \neq 0$.
$\dot{V}(e) = L_{f_1 + g_1\alpha_1} V_1(e_1) - \sum\limits_{i=2}^{k-1} k_i e_i^2 + e_{k-1} g_{k-1}(\dots) e_k - e_k \dot{\alpha}_{k-1} + e_k u$.
Here $e_k u = L_g V$ and the rest is $L_f V$.
Assume $L_g V = 0 \iff e_k = 0$
$\implies L_f V = L_{f_1 + g_1\alpha_1} V_1(e_1) - \sum\limits_{i=2}^{k-1} k_i e_i^2 < 0 \;\; \forall e \neq 0$ with $e_k = 0$. □

$\implies$ We can apply Sontag's formula to construct V.

This theory can be extended to systems with $x_2, \dots x_k, u \in R^m$ ("block backstepping").

5 Systems with inputs and outputs

Study/control systems $\dot{x} = f(x, u)$ with "output" $y(t) = W(x(t))$

5.1 Sliding mode control

Motivating example

$$\dot{x_1} = x_2$$

$$\dot{x_2} = u \Rightarrow \dot{y} = x_2 + u$$

$$y = x_1 + x_2$$

Choose:

$$u = \begin{cases} -x_2 - 1, & y > 0 \\ -x_2 + 1, & y < 0 \\ -x_2, & y = 0 \end{cases}$$

$$\Rightarrow \dot{y} = \begin{cases} -1, & y > 0 \\ +1, & y < 0 \\ 0, & y = 0 \end{cases}$$

Solutions(Laratheodory) are if $y(0) > 0$, then

$$y(t) = \begin{cases} y(0) - t, & t \leq y(0) \\ 0, & t > y(0) \end{cases}$$

If $y(0) < 0$, then

$$y(t) = \begin{cases} y(0) + t, & t \leq y(0) \\ 0, & t > -y(0) \end{cases}$$

Key property: choose u s.t. $y(t)$ goes to zero in finite time $\Rightarrow x(t)$ tends $\{(x_1, x_2) \in \mathbb{R}^2 | x_1 + x_2 = 0\}$ in finite time

Consider dynamics on S

$$\begin{cases} \dot{x_1} = x_2 (x_2 = -x_1 \text{ if } y = 1) = -x_1 \\ \dot{x_2} = u = -x_2 \end{cases}$$

globally as stable

Two "phases"

1. solutions converge to S in finite time
2. solutions converge to zero ("on S") asymptotically

$\rightsquigarrow$ "sliding mode" control

Remark: in (1) "finite time convergence is crucial"

General procedure:

$$\dot{x} = f(x) + g(x)uy = h(x) = s(x)$$

$f : \mathbb{R}^n \to \mathbb{R}^n,\ y : \mathbb{R}^n \to \mathbb{R}^n,\ s : \mathbb{R}^n \to \mathbb{R}$

u - scalar input, $s(x)$ - sliding

single input, single output

Assumptions: y has relative degree 1, well - defined globally, i.e. $L_g s(x) \neq 0 \; \forall \in \mathbb{R}^n$

Two-step approach:

1. Bring $x(t)$ to $S := \{x \in \mathbb{R}^n | S(x) = 0\}$ in finite time
2. Have $x(t)$ going to zero asymptotically (on S)
    - switching between nodes 1 and 2
    - mode 2 is "sliding mode"

How are 1 + 2 achieved?

- Design of sliding manifolds crucial!

  Need: For $y(t) = 0$ for all $t \geq 0$. All solutions convergeto the origin, i.e., "zero dynamics" have globally asymptotically stable origin.

  How? e.g. systems in "regular form" $x = [\eta \xi]'$

$$\dot{\eta} = f_1(\eta, \xi)$$
$$\dot{\xi} = f_2(\eta, \xi) + g_2(\eta, \xi)u$$

  Choose $s(x) = \eta - \phi(\eta)$, where ϕ asymptotically stabilizes zero dynamics $\dot{\eta} = f_1(\eta, \phi(\eta))$ (and $\phi(0) = 0$) Ex. 1.9 in Khalil

- Converging to sliding manifold in finite time: $\rightsquigarrow \dot{y} = L_f s(x) + L_g s(x) u$, where $L_g s(x) \neq 0$. Obvious choice to render S invariant is $u = -\frac{L_f S(x)}{L_g s(x)}$ (mode 2, behaviour on S)

  Asin motivating example, add
$$\begin{cases} -\hat{u}/L_g s(x) & y > 0 \\ \hat{u}/L_g s(x) & y < 0 \end{cases}$$

  where $\hat{u} > 0$

$$u = -\frac{1}{L_g s(x)} (L_f s(x) + \hat{u} sgn(s(x)))$$

  $\rightsquigarrow \dot{y} = -\hat{u} sgn(y)$

  $\rightsquigarrow$ (caratheodory) solutions converge to zero in finite time

  $\rightsquigarrow x(t)$ converges to S in finite time

Control Lyapunov perspective

$V(X) = \frac{1}{2} s(x)^2$

$$\dot{V}(x) = s(x)\dot{s}(x) = s(x)(L_f s(x) + L_g s(x) u) = -s(x) sgn(s(x))\hat{u} = |s(x)|\hat{u} < 0 for s(x) \neq 0$$

Consider $W = \sqrt{2v} \rightsquigarrow^{s\neq 0} \dot{w} = \sqrt{2}, \ \frac{1}{2\sqrt{v}}\dot{v} = -\hat{u}$

$\rightsquigarrow$ w converges to 0 in finite time $\Rightarrow$ V converges to 0 in finite time $\Rightarrow$ $S(x(t))$ converges to 0 in finite time.

Example.

$$\begin{aligned} \dot{x_1} &= x_2 + x_1 sin(x_2) \\ \dot{x_2} &= x_2^2 + x_1 + u \end{aligned} \tag{10}$$

Choose $s(x) = x_2 + 2x_1$, where $+2x_! := \phi(x_1)$ on S: $\dot{x_1} = -2x_1 + x_1 sin(-2x_1) \rightsquigarrow$ asymptotically stable

$\dot{s} = x_2^2 + x_1 + u - 2x_2 - 2x_1 sin(x_2) \rightsquigarrow u = -(x_2^2 + x_1 - 2x_2 - 2x_1 sin(x_2) + \hat{u} sgn(x_2 - 2x_1)), \ \hat{u} > 0$

$\rightsquigarrow$ yields finite-time convergence to S.

Alternative sliding mode controllers

$$u = -\frac{1}{L_g s(x)}(L_f s(x) + \hat{u} sgn(s(x))), \ \hat{u} > 0$$

In particular

$$u = -\frac{1}{L_g s(x)}(L_f s(x) + \hat{u}|L_g s(x)| sgn(s(x)))$$

$\rightsquigarrow$ ensure robustness w.r.t. "matched uncertainties"

$$\dot{x} = f(x) + g(x)\sigma(x) + g(x)u$$

$\sigma : \mathbb{R}^n \to \mathbb{R}$, bounded (i.e., $|\sigma(x)| \leq c \ \forall x \in \mathbb{R}^n$)

Why? $V(x) = \frac{1}{2}s(x)^2$

$\dot{V} = s(x)(L_f s(x) + L_g s(x) u + L_g s(x)\sigma(x)) = -s(x) sgn(s(x))\hat{u}|L_g s(x)| + s(x) L_g s(x)\sigma(x) \leq -|s(x)||L_g s(x)|(\hat{u} - c)$

$$u = -\frac{1}{L_g s(x)}(L_f s(x) + (\hat{u} + \beta(x)|L_g s(x)|) sgn(s(x)))$$

ensures robustness w.r.t. matched uncertainties s.t. $\sigma(x) \leq \beta(x) \ \forall x \in \mathbb{R}^n$

Example 2

Example.

$$\begin{aligned} \dot{x_1} &= x_2 + x_1 sin(x_2) \\ \dot{x_2} &= \theta x_2^2 + x_1 + u \end{aligned}$$

$|\theta| \leq 2 \rightsquigarrow |\theta x_2^2| \leq 2x_2^2 = \beta(x)$

$$\dot{s} = \theta x_2^2 + x_1 + u + 2x_1 + 2x_1 sin x_2$$

$$u = -(x_1 + 2x_1 + 2x_1 sin x_2 + \hat{u} sign(s(x)) + 2x_2^2 sgn(s(x)))$$

$L_f s(x) = x_1 + 2x_1 + 2x_1 sin x_2$

$\rightsquigarrow \dot{s} = -\hat{u} sgn(s(x)) + x_2^2(\theta - 2sgn(s(x))) \Rightarrow$ finite -time convergence to S.

Remedy: replace sign-function by saturated slope (continuous approximation)

can be extended to multi-input systems $u \in \mathbb{R}^m \to s : \mathbb{R}^n \to \mathbb{R}^m$

5.2 Dissipativity

Dissipativity: Generalization of Lyapunov theory to systems w inputs and outputs

$$\begin{array}{lll} \dot{x} = f(x,u) & x(0) = x_0 & f : \mathbb{R}^n \times \mathbb{R}^m \to \mathbb{R}^n \\ y = h(x) & & h : \mathbb{R}^n \to \mathbb{R}^p \end{array} \tag{11}$$

Definition:

- storage function $s : \mathbb{R}^n \to \mathbb{R}$, $x \to S(x)$ nonnegative (i.e., $s(x) \geq 0 \ \forall x \in \mathbb{R}^n$)
- supply rate $s : \mathbb{R}^m \times \mathbb{R}^p \to \mathbb{R}$, $(u, y) \to s(u, y)$

Definition: System (11) is dissipative w.r.t. the supply rate s if there exists a storage function S s.t. $\forall x_0 \in \mathbb{R}$, $\forall t \geq 0$, $\forall u : [0, t] \to \mathbb{R}^m$

$$S(x(t)) \leq S(x_0) + \int_0^t s(u(\tau), y(\tau)) d\tau$$

First item - storage at time t, second item - initial storage,the last item - supply delivered over [0, t]

"dissipation inequality" (DIE)

Interpretation:

- "Dissipative systems dissipate storage/stored energy"
- "No storage/energy can be created internally"
- positive s "supplied" energy/ storage
  negative s "extracted" energy / storage

Remark:

- If S is differentiable, DIE is equivalent to $\dot{S}(x) \leq s(u, y) \ \forall u, x$
- Dissipation (rate) is defined as $d(x, u) = s(u, h(x)) - \dot{S}(x) \geq 0$

Examples of dissipative systems:

	supply rate	input	output	storage function
electrical	$u \cdot i$	voltage	current	energy storage in all capacitors and inductors
mechanical	$F \cdot V$	force	velocity	Hamiltonian $=$ kinetic $+$ potential energy
thermo-dynamics	$Q + W$	rate of hate	rate of work	internal energy
	$-\frac{a}{T}$		temperature	entropy

How do we computer storage functions?

- in general difficult (similar to computing Lyapunov functions)
- characterization via optimization problem

Introduce "available storage"

$$S_a(x) := sup_{u:[0,T]\to\mathbb{R}^m, T\geq 0, x(0)=0}(-\int_0^T s(u(\tau), y(\tau)))$$

the maximum of energy we can extract

Theorem 5.1. System (11) is dissipative w.r.t. the supply rate s iff $S_a(x) < \infty$ for all $x \in \mathbb{R}^n$

Moreover, if $S_a(x) < \infty$ for all $x \in \mathbb{R}^n$, then S_a is a storage function and $S(x) \geq S_a(x)$ $\forall x \in \mathbb{R}^n$ for all storage functions S.

Proof. Sketch of proof. "$S_a(x) < \infty \Rightarrow$ dissipativity". $S_a(x) \geq 0$ $\forall x \in \mathbb{R}^n$ by definition (can take $T = 0$)

$$S_a(x) = sup_{u[0,T]\to\mathbb{R}^m, T\geq 0, x(0)=0} - \int_0^T s(u(\tau), y(\tau))d\tau \geq^* - \int_0^t s(u(\tau), y(\tau))d\tau + sup_{u[t,t+T]\to\mathbb{R}^m, T\geq 0, x(t)=x(t)} - \int_t^{t+T} s$$

the last item is $S_a(x(t))$,

$$\Rightarrow = S_a(x(t)) - \int_0^t s(u(\tau), y(\tau))d\tau$$

and this is DIE $\Rightarrow S_a$ is a storage function

Note for (*): "suboptimal" to first transfer system to $x(t)$ and then extract maximum energy starting of $x(t)$

"Dissip. $\Rightarrow S_a(x) < \infty$"

From DIE: $S(x_0) \geq S(x(T)) - \int_0^T s(u(\tau), y(\tau))d\tau \geq -\int_0^T s(u(\tau), y(\tau))d\tau$

for all x_0, for all $T \geq 0$, all $u(\cdot) \Rightarrow S(x_0) \geq sup_{u:[0,T]\to\mathbb{R}^m,\ T\geq 0,\ x(0)=x_0} - \int_0^T s(u(\tau), y(\tau))d\tau = S_a(x)$

$\Rightarrow S_a(x) < \infty$ $\forall x \in \mathbb{R}^n$ and $S \geq S_a$ for all storage function S. $\square$

Another special supply rate: "required supply"

$$S_r(x) := inf_{u:[-T,0]\to\mathbb{R}^m,\ T\geq 0,\ x(-T)=x^*,\ x(0)=x} \int_T^0 s(u(\tau), y(\tau))d\tau$$

Theorem 5.2. Assume that end state $x \in \mathbb{R}^n$ is readable from x^*. If system (11) is dissipative w.r.t. the supply rate s, then for all storage functions S

$$S(x) \leq S_r(x) + S(x^*)\ \forall \in \mathbb{R}^n$$

Furthermore, $S_r(x) + S(x^*)$ is a storage function.

Proof. Sketch of proof.

Consider $u : [-T, 0] \to \mathbb{R}^n$ which transfers the system from x^* to x

$$S(x) - S(x^*) \leq [byDIE]inf_{u[-T,0]\to\mathbb{R}^n,\ T\geq 0,\ x(-T)=x^*,\ x(0)=x} \int_{-T}^0 s(u(\tau), y(\tau))d\tau = S_r(x)$$

□

Remark: Set of all storage functions is convex, i.e. , $\alpha S_1 + (1-\alpha)S_2$, $\alpha \in [0, 1]$ is a storage function (for $S1,\ S2$ storage functions)

Dissipativity widely used in control theory

If system is dissipative with positive definite storage S and if there exists a (continuous) $k : \mathbb{R}^n \to \mathbb{R}^n$ s.t.

$$s(k(x), h(x)) < 0,\ \forall x \neq 0$$

then $x = 0$ is asymptotically stable under $u = k(x)$

Why? Take S as Lyapunov function

$$\dot{S} \leq s(u, y) =^{u=k(x)} s(k(x), h(x)) < 0,\ \forall x \neq 0$$

L_2 - gain via supply rate

$$s(u, y) = \frac{1}{2}\gamma^2\|u\|^2 - \frac{1}{2}\|y\|^2$$

$\rightsquigarrow$ from dissipation inequality

$$\frac{1}{2}\int_0^t \gamma^2\|u(\tau)\|^2 + \|y(\tau)\|^2 d\tau \geq S(x(t)) - S(x(0)) \geq -S(x(0))$$

$$\Rightarrow \int_0^t \|y(\tau)\|^2 d\tau \leq \gamma^2 \int_0^t \|u(\tau)\|^2 d\tau + 2S(x(0))$$

$\Rightarrow$ system has L_2 - gain γ

Classify optional $l(x, u^{\leftarrow x})$ operating conditions $s(u, y) = l(x, u) - l(x^*, u^*)$

Example.

$$\dot{x} = u,\ y = x$$

$S(x) = \frac{1}{2}x^2,\ \dot{S} = xu = uy \rightsquigarrow$ system is dissipative w.r.t. supply rate $s(u, y) = uy$.

Example. "part-Hamiltonian systems"

$$\dot{x} = [F(x) - R(x)]\nabla H(x) + g(x)u$$

$y = y(x)^T \nabla H(x)$, H - Hamiltonian total stored energy in system

$F(x) = -F^T(x)$ internal interconnection structure (power conserving) $R(x) \geq 0$ dissipation structure

Take $S(x) = H(x)$

$$\begin{aligned} \dot{S}(x) &= \nabla H(x) \cdot [F(x) - R(x)]\nabla H(x) + \nabla H(x) \cdot g(x)u \\ &= -\nabla H(x) \cdot R(x)\nabla H(x) + yu \leq yu \end{aligned}$$

as far as $-\nabla H(x) \cdot R(x)\nabla H(x) \leq 0 \Rightarrow$ dissipative w.r.t. $s(u, y) = u^T y$

5.3 Passivity

$$\begin{aligned} \dot{x} = f(x, u),\ x \in \mathbb{R}^n,\ u \in \mathbb{R}^m \\ y = h(x),\ y \in \mathbb{R}^m \end{aligned} \tag{12}$$

(same number of inputs and outputs)

Definition. System (12) is passive if it is dissipative w.r.t. supply rate $s(u, y) = u^T y$

Why "passive"? From circuit theory passive compared to "active" ones as diods or transistors

Examples: electrical, mechanical systems

Stabilization of passive systems

Definition. System (12) is zero-state observable (ZSO) if (for $u(t) = 0$) $y(t) = 0$ for all $t \geq 0 \Rightarrow$ $x(t) = 0$ for all $t \geq 0$

"trivial solution $x(t) \equiv 0$ is observable from the output"

Remark: can be related to zero-state detectability

Theorem 5.3. Let system (12) be

i) passive in differentiable storage set

ii)ZSO

Then the feedback $u = -Py,\ P > 0$ renders the origin asymptotically stable

Proof. Sketch of proof From passivity

$$\dot{S} \leq u^T y = -y^T P y \leq 0 \tag{13}$$

$$S(x(t)) - S(x(0)) \leq -\int_0^t y(\tau)^T P y(\tau) d\tau, \ \forall t \geq 0$$

$S(x(t)) \geq 0$

$$S(x_0) \geq \int_0^t y(\tau)^T P y(\tau) d\tau, \ \forall t \geq 0 \tag{14}$$

$y(\tau)^T P y(\tau) \geq 0$. Want to show $S(x_0) > 0$ for all $x_0 \neq 0$. By contradiction. Suppose $\exists \bar{x} \neq 0$ with $S(\bar{x}) = 0$.

From (14) $\Rightarrow y(\tau) = 0 \ \forall \tau \geq 0$

By ZSO $\Rightarrow x(\tau) = 0 \ \forall \tau \geq 0 \Rightarrow \bar{x} = 0$

$\Rightarrow S$ is positive definite. $\Rightarrow$ Lyapunov stability together with (13)

For convergence, use (13) together with La Salle's invariance principle and ZSO □

Advantage. We have (static) output feedback (no observer needed)

Passivity of interconnections

1. Parallel interconnections of two passive systems are passive
   Take $S(x_1, x_2) = S_1(x_1) + S_2(x_2)$.
   $$\dot{S} \leq u_1^T y_1 + u_2^T y_2 = u^T(y_1 + y_2) = u^T y$$
2. Feedback interconnection of passive systems are passive
   Take $S(x_1 + x_2) = S_1(x_1) + S_2(x_2)$
   $$S \leq u_1^T y_1 + u_2^T y_2 =^{x_1 = u_2 = y} y^T(u_1 + y_2) = y^T u$$

Remark:

- does not work for serious interconnections
- can construct possibly large networks of passive systems

Stability if feedback interconnections:

Main idea: "shortage" of passivity of H_1 can be compensated by nexcess of passivity of H_2

Theorem 5.4. Consider feedback interconnection (2) with $u \equiv 0$. Assume that H_1 and H_2 are (i) ZSO and dissipative with differentiable $S_1,\ S_2$ w.r.t. the supply rates

$$S_i(u_i, y_i) = u_i^T y_i - \rho_i y_i^T y_i - \nu_i u_i^T u_i,\ i = 1, 2,\ \rho, \nu \in \mathbb{R} \tag{15}$$

Then the origin $(x_1, x_2) = (0, 0)$ for interconnection is asymptotically stable if $\nu_1 + \rho_2 > 0$ and $\nu_2 + \rho_1 > 0$.

Proof. Take $S(x) = S_1(x_1) + S_2(x_2)$.

$$\begin{aligned} \dot{S}(x) \leq^{(15)} & u_1^T y_1 - \rho_1 y_1^T y_1 - \nu_1 u_1^T u_1 \\ & + u_2^T y_2 - \rho_2 y_2^T y_2 - \nu_2 u_2^T u_2 \\ = & -(\rho_1 + \nu_2) y_1^T y_1 - (\rho_2 + \nu_1) y_2^T y_2 \end{aligned}$$

$u_1^T y_1$ and $u_2^T y_2$ can be excluded as $u_1 = y_2,\ u_2 = y_1$.

$\Rightarrow$ can show as in previous theorem that S is positive definite *Rightarrow* Lyapunov stability

For using La Salle:

$$\begin{aligned} y_1 \equiv 0 \Rightarrow u_2 \equiv 0 \Rightarrow^{ZSO} x_2 \equiv 0 \\ y_2 \equiv 0 \Rightarrow u_1 \equiv 0 \Rightarrow^{ZSO} x_1 \equiv 0 \end{aligned}$$

□

Remark

- If (3) is sabisfied with $v_i = 0$: "output - feedback passive" $\Rightarrow p_i > 0$ - "excess" of passivity, $p_i < 0$ - "shortage" of passivity ($|p_i|$).
- If (3) satisfied with $p_i = 0$: "input - feadforward passive" $\Rightarrow v_i > 0$ - "excess" of passivity, $v_i < 0$ - "shortage" of passivity ($|v_i|$).
- Comment on terminology "output feedback passive" $\dot{s} \leq u^T y - \rho y^T y =$ [output feedback $u = \bar{u} - ky] = \bar{u}^T y - (\rho + k) y^T y \leq \bar{u}^T y$ In other words, system can be made passive by output feedback.
- Similar for feedforward passivity, system can be made passive by feedback forward the input: $\bar{y} = y + ku$.

Remark:

Feedforward interconnection b) can be extended to allow h_1 or h_2 to explicitly depend on $u \Rightarrow$ includes static systems (controllers) e.g. state output feedback $y = h(u)$ e.g. $y_2 = ku_2 \Rightarrow$ input-strictly passive controller.

Extension of output-feedback/input-feedforward passivity:

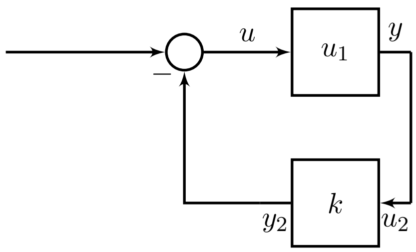
$$s(u, y) = u^T - \rho(y)^T y$$

with $\rho(y) = [\rho_1(y_1), \ldots, \rho_n(y_n)]^T$ with ρ_i sectionnonliniarities, $\rho_i : \mathbb{R} \to \mathbb{R}$

Example.

$$H_1 : \qquad \begin{aligned} \dot{x}_1 &= x2 \\ \dot{x}_2 &= -x_1^3 + x_2 + u \\ y &= x_2 \end{aligned}$$

Take $S_1(x) = \frac{1}{4}x_1^4 + \frac{1}{2}x_2^2 \Rightarrow \dot{S}_1 = x_1^3 x_2 - x_1^3 + x_2^2 + x_2 u$, define $y^2 := x_2^2$ then $yu = x_2 u \Rightarrow$ output - feedback passive with shortage of passivity $1 \Rightarrow \rho_1 = -1$ and $v_1 = 0$.

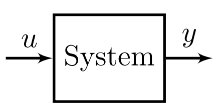


$y_2 u_2 = k u_2 = \gamma k u_2^2 + \frac{1-\gamma}{k} y_2^2$, $0 < \gamma < 1 \Rightarrow \rho 2 = \frac{1-\gamma}{k}$, $v_2 = \gamma k$

$\Rightarrow v_2 + \rho 1 > 0$ for $k > 1$ and γ close enough to 1 $v_1 + \rho_2 = \rho_2 > 0$. $\Rightarrow$ with ZSO, the origin is a stable.

6 Input/Output Methods

References: Desoev, Vidyasagar "Feedback Systes Input-output properties"



$$\begin{aligned} &u : u \to y \\ &u, y : [0, \infty] \to \mathbb{R}^m \\ &t \to u(t), y(t) \end{aligned}$$

6.1 Sygnals and Systems

- How to define "stability" in input/output setting?
- Which signals are "good"?

Definition. Lp-spaces, $p \in [1, \infty]$. $Lp[0, \infty) = \{\Phi : [0, \infty) \to \mathbb{R}^m, masurable | \int_0^\infty ||\Phi(t)||^p dt < \infty\}$

Interpretation: "finite energy sygnal" (p=2).

Remark: "measurable" = pointwise limit of a sequence of piecewise constant functions (except on a set of measure o)

Example. :

- continuous function
- functions with "few enough" discontinuities

Lp is a real vector space ("signals $\Phi(\cdot)$ are vectors") i.e., for $\Phi, \Phi_1, \Phi_2 \in Lp$, $\alpha \in \mathbb{R}$ vector addition: $\Phi_1 + \Phi_2 : t \to \Phi_1(t) + \Phi_2(t) \in Lp$. Scalar multiplication: $\alpha\Phi : t \to \alpha\Phi(t) \in Lp$

Zero element is signal $\Phi \equiv 0$.

Lp is a normed vector space with norm $||\Phi||_{Lp} = \sqrt[p]{\int_0^\infty ||\Phi(t)||^p dt}$ for $\Phi \in Lp \Rightarrow$

- $||\Phi||_{Lp} = 0 \iff \Phi = 0$, else $||\Phi||_{Lp} > 0$
- for $\alpha \in mathbbR$, $||\alpha\Phi||_{Lp} = |\alpha| ||\Phi||_{Lp}$
- for $\Phi_1, \Phi_2 \in Lp$ $\;||\Phi_1 + \Phi_2||_{Lp} \leq ||\Phi_1||_{Lp} + ||\Phi_2||_{Lp}$

7 Exercises

7.1 Exercise 1

Problem 1:

Proof. For any $t \geq 0$, we have

$$\frac{d}{dt}V(x(t)) = \frac{d}{dt}(V \circ x)(t) = \langle \nabla V(x(t)), \frac{d}{dt}x(t) \rangle = \langle \nabla V(x(t)), f(x(t)) \rangle = L_f V(x(t))$$

□

Problem 2:

Proof.

Lemma 7. Given the assumptions in Problem 2, if there exists a solution $x : [0, +\infty] \to R^n, t \to x(t)$, of $\dot{x} = f(x)$ s.t. $x(t) \in K$ for any $t \geq 0$, where $k \subset R^n$ is a compact with $O \in K$ (O - origin), then $x(t) \xrightarrow{t \to +\infty} 0$.

Clearly, for any $c > 0, lev_{\leq c}V$ is positive invariant w.r.t $\dot{x} = f(x)$. Given $c > 0$, let $x_0 \in lev_{\leq c}V$, i.e., $V(x_0) \leq c$. Then, for any $t \geq 0$

$$V(x(t)) = V(x_0) + \int_0^t \frac{d}{dt}V(x(\tau))d\tau < V(x_0) \leq c,$$

i.e. $x(t) \in lev_{\leq c}V$ for any $t \geq 0$.
Then, for any $x_0 \in lev_{\leq c}V$ there exists a solution $x : [0, +\infty] \to R^n$ of $\dot{x} = f(x)$ s.t. $x(t) \in lev_{\leq c}V$ for all $t \geq 0$. Clearly, $O \in lev_{\leq c}V$. We conclude by using the above Lemma ($K = lev_{\leq c}V$). □

Problem 3:

Proof. Let $r > 0$. By assumption, there exists $c > 0$ s.t. $\overline{B(0, r)} \subset lev_{\leq c}V$.
Since any bounded set $lev_{\leq c}V$ is a subset of the region of attraction, and since the sublevel sets are arbitrary large, R^n is also the region of attraction.
A condition that ensures that for any $c > 0, lev_{\leq c}V$ is bounded is $V(x) \xrightarrow{||x|| \to +\infty} +\infty$. □

Problem 4:

Proof. Let $P : R^2 \to R^2$ be continuously differentiable. Consider

$$m\dot{v} = -g\nabla P(q).$$

Consider $x = (q, v), \dot{q} = v, \dot{v} = -\frac{g}{m}\nabla P(q)$. Let $H : R^2 \to R$ be defined by

$$H(q, v) = \frac{1}{2}||v||^2 + \frac{g}{m}P(q).$$

We have

$$\begin{pmatrix} \dot{q} \\ \dot{v} \end{pmatrix} = \begin{pmatrix} & I \\ -I & \end{pmatrix} \nabla H(q, v)$$

Since P is positive definite, then H is positive definite.
Then

$$L_{\begin{pmatrix} & I \\ -I & \end{pmatrix}\nabla H} H(q, v) = \langle \nabla H(q, v), \begin{pmatrix} & I \\ -I & \end{pmatrix} \nabla H(q, v) \rangle = 0 \quad \forall (q, v) \in R^2 \times R^2$$

$\implies$ the origin is stable. □

Problem 5:

Proof. For any $t \geq 0$, we have
$\frac{d}{dt}V(t,x(t)) = \frac{d}{dt}(V \circ (id_R, x))(t) = [id_R : R \to R, t- > t] = \langle \begin{pmatrix} \frac{\partial}{\partial t}V(t,x(t)) \\ \frac{\partial}{\partial x}V(t,x(t)) \end{pmatrix}, \frac{d}{dt}(id_R(t), x(t)) \rangle = \langle \begin{pmatrix} \frac{\partial}{\partial t}V(t,x(t)) \\ \frac{\partial}{\partial x}V(t,x(t)) \end{pmatrix}, \begin{pmatrix} 1 \\ f(t,x(t)) \end{pmatrix} \rangle = \frac{\partial}{\partial t}V(t,x(t)) + \langle \frac{\partial}{\partial x}V(t,x(t)), f(t,x(t)) \rangle = L_{\begin{pmatrix} 1 \\ f \end{pmatrix}} V(x(t))$.

$g(t,x(t)) := \begin{pmatrix} 1 \\ f(t,x(t)) \end{pmatrix}$ □

Problem 6:

Proof. Consider $\dot{x} = a\sin(\omega t), \quad x(0) = x_0 \in R \quad a, \omega > 0$.
This is solved by $x(t) = -\frac{a}{\omega}\cos(\omega t) + \frac{a}{\omega} + x_0$.
Clearly, x is bounded on $[0, +\infty]$ since $x(t) \geq x_0$, and $x(t) \leq x_0 + 2\frac{a}{\omega}$ for any $t \geq 0$.
Choose $\varepsilon = \frac{a}{\omega}$ and $t_0 = 0$. Then $\forall \delta > 0 \;\; \exists x_0 \in B(0,\delta)$, namely x_0, s.t. $\exists t \geq t_0$, namely $t = \frac{\pi}{\omega}$, with $x(t) \notin B(0,\varepsilon) \quad (x(\frac{\pi}{\omega}) = 2\frac{a}{\omega} > \varepsilon)$. □

Short notes:

Problem 7:
Take $V(t,x) = \frac{1}{2}x^2$.

Problem 8:
Take $V(t,x) = x_1^2 + (1 + e^{-2t})x_2^2$.

7.2 Exercise 2

Problem 1:

Proof. a) Since α_1 is continuous and strictly increasing:

$$\forall x, y \in [0,\delta), x < y \quad \alpha_1(x) < \alpha_1(y)$$

$\implies \alpha_1$ is injective, i.e.

$$\forall x, y \in [0,\delta), x \neq y \implies \alpha_1(x) \neq \alpha_1(y).$$

Clearly, $\alpha_1 : [0,\delta) \to \alpha_1([0,\delta))$ is surjective, i.e.

$$\forall y \in \alpha_1([0,\delta)) \;\; \exists x \in [0,\delta) : \quad \alpha_1(x) = y$$

Thus α_1 is bijective.
Define $\alpha_1^{-1} : [0, \alpha_1(\delta)) \to [0,\delta)$ by $\alpha_1^{-1}(\alpha_1(x)) = x$.

b) From a) we have $\alpha_3^{-1} \in K$. Since $\alpha_3 \in K_\infty, \alpha_3{-}1$ is defined om $[0, +\infty)$ and $\alpha_3^{-1}(r) \xrightarrow{r\to\infty} \infty$

c) Let $\alpha = \alpha_1 \circ \alpha_2$. Then we have $\alpha(0) = \alpha_1(\alpha_2(0)) = 0$ and $\alpha(r) > 0$ whenever $r > 0$. Moreover, for any x, y:

$$x < y \implies \alpha_2(x) < \alpha_2(y) \implies \alpha(x) = \alpha_1(\alpha_2(x)) < \alpha_1(\alpha_2(y)) = \alpha(y)$$

It is continuous (as composition of continuous functions).

d) From c) we have $\alpha := \alpha_3 \circ \alpha_4 \in K, \alpha$ is defined on $[0, +\infty)$ since $\alpha_3, \alpha_4 \in K_\infty$ and

$$r \to +\infty \implies \alpha_4(r) \to +\infty \implies \alpha(r) \to +\infty$$

e) For each $s, r \mapsto \beta(\alpha_2(r), s)$ is of class K.
Thus $r \mapsto \alpha_1(\beta(\alpha_2(r), s)) \in K$.
For each $r, s \mapsto \beta(\alpha_2(r), s)$ decreases.
Hence, $s \mapsto \alpha_1(\beta(\alpha_2(r), s))$ decreases.
Moreover,

$$\alpha_1(\beta(\alpha_2(r), s)) \xrightarrow{s\to+\infty} 0$$

□

Problem 3:

Proof. For $u = 0$ the origin is UGAS. Consider $V : [0, +\infty) \times R \to R, \quad (t, x) \mapsto \frac{1}{2}x^2$.
We have

$$\frac{\partial}{\partial t}V(t,x)+\frac{\partial}{\partial x}V(t,x)f(t,x,u) = (\sin(t){-}2)x^2{+}xu \leq -x^2{+}|x||u| = -(1{-}\theta)x^2{-}\theta x^2{+}|x||u|, \;\; \theta \in (0,1)$$

Hence, whenever $|x| \geq \frac{|u|}{\theta}$, the system is ISS with $\gamma = \frac{r}{\theta}$. □

Problem 4:

Proof.

$$\dot{x} = -x + (x^2 + 1)d \tag{16}$$
$$\dot{x} = -2x - x^3 + (x^2 + 1)d \tag{17}$$

System (**??**): Clearly, the system is 0-GAS. However, for $d = 1$ and $x > 1$ we have $x^2 + 1 > x$.

$$f(x, 1) = -x + (x^2 + 1) > 0$$

and thus $\dot{x} > 0$. Hence, if $x(0) = x_0 > 1$, the solution diverges (in finite time).
$\implies$ System (**??**) isn't ISS.

System (**??**): It is 0-GAS. Moreover, for any finite d there exists a "large" x s.t.

$$2x + x^3 > (x^2 + 1)d$$

$$\implies f(x,d) = -2x - x^3 + (x^2+1)d < 0$$

and $\dot{x} < 0 \implies$ System **??** is ISS.
Consider $V : R \to R, x \mapsto \frac{1}{2}x^2$ s.t

$$V'(x)f(x,d) = -2x^2 - x^4 + x(x^2+1)d \leq -x^2 - x^2(x^2+1) + (x^2+1)|x||d|$$

Hence, whenever $|x| \geq |d|$,

$$V'(x)f(x,d) \leq -x^2$$

s.t. system (**??**) is ISS with $\gamma(r) = r$. □

Problem 5:

Proof.

$$\langle \nabla V(x), -\nabla V(x) + \delta u \rangle \leq -||\nabla V(x)||^2 + |\langle \nabla V(x), \delta u \rangle| \leq [YI] \leq -||\nabla V(x)||^2 + \frac{1}{2}||\nabla V(x)||^2 + \frac{\delta^2}{2}||u||^2$$

Young's inequality:

$$\forall x, y: \quad |\langle x, y \rangle| \leq \varepsilon \frac{||x||^p}{p} + \frac{||y||^q}{\varepsilon q}, \quad p, q > 1, \frac{1}{p} + \frac{1}{q} = 1, \varepsilon > 0$$

Hence, whenever $||x|| > \frac{\delta}{\sqrt{c}}||u||, t \mapsto ||x(t)||$ is decreasing.
Moreover whenever $||x|| \geq \frac{\delta}{\sqrt{c\theta}}||u||, \theta \in (0,1)$, we have $\langle \nabla V(x), -\nabla V(x) + \delta u \rangle \leq -\frac{c}{2}(1-\theta)||x||^2 \implies$ ISS. □

7.3 Exercise 3

Motivation: Lyapunov Theory

$$\dot{x} = f(x,u)$$

$f : \mathbb{R}^n \times \mathbb{R}^m \to \mathbb{R}^n$

Definition. (CLF) A function $V : \mathbb{R}^n \to \mathbb{R}$ is a CLF if it is continuous differentiable, positive definite, radially unbounded and $\forall x \neq 0 \ \inf_u < \nabla V(x), f(x,u) >< 0$

In order to find CLFs, we restrict our analysis to input -affine systems

$$\dot{x} = f(x) + G(x)u$$

where $f : \mathbb{R}^n \to \mathbb{R}^n, \ G : \mathbb{R}^n \to \mathbb{R}^{n \times m}$

Proposition: A continuous, differentiable, positive definite and radially unbounded. $V : \mathbb{R}^n \to \mathbb{R}$ is a CLF iff

$$\forall x \neq 0 \ L_G V(x) = 0 \Rightarrow L_f V(x) < 0$$

Image to be inserted

Problem 1

Consider $\dot{x} = cos(x) + (1 + e^x)u$ where $f(x) = cos(x)$- drift and $g(x) = 1 + e^x$

Let $V : \mathbb{R} \to \mathbb{R},\ x \mapsto \frac{1}{2}x^2$. Clearly, continuous differentiable, positive definite and radially unbounded. Moreover, for any nonzero x, we have $L_G V(x) \neq 0$.

Thus, for any $x \neq 0$, there exists a control that readers $< \nabla V(x), f(x) + g(x)u >$ negative. Givn this CLF, there exists a state feedback $u = u(x)$, e.g.

$$u(x) = -\frac{kx + cos(x)}{1 + e^x},\ k > 0$$

Problem

Consider

$$\dot{x_1} = -x_1^3 + x_2 e^{x_1} cos(x_2)$$
$$\dot{x_2} = x_1^5 sin(x_2) + u$$

Take $V : \mathbb{R}^2 \to \mathbb{R},\ (x_1, x_2) \mapsto \frac{1}{2}(x_1^2 + x_2^2)$

For any $x \neq 0$, we have

$$\inf_{u \in \mathbb{R}} (L_f V(x) + L_G V(x)u) = \begin{cases} L_f V(x), & if\ L_G V(x) = 0 \\ -\infty & else \end{cases}$$

In particular,

$$L_f V(x) = \cdots = x_1(-x_1^3 + x_2 e_1^x cos(x_2)) + x_2 x_1^5 sin(x_2)$$
$$L_G V(x) = \cdots = x_2$$

However,

$$L_f V(x)|_{x_2=0} = -x_1^4 < 0\ \forall x_1 \neq 0$$

Image to be inserted

Concluding that V is a CLF.

Problem 2:

$\dot{x} = Ax + Bu$, input defined system where (A, B) is stabilizable, there exists $K \in \mathbb{R}^{m \times n}$ s.t. $A + BK$ is Hurwitz (cf. KRT).The latter is equivalent to the existance $P = P^T > 0$ s.t. $P(A + BK) + (A + BK)^T P < 0$ (cf. Khalil theorem 4,6)

Let $V : \mathbb{R}^n \to \mathbb{R}, x \mapsto < x, Px >$. Moreover, $\forall x \neq 0 \exists u = Kx$ s.t. $< \nabla V(x), Ax + Bu >< 0$, since

$$< \nabla V(x), Ax + Bu >=^{u=Kx} < x, (P(A + BK) + (A + BK)^T P)x >< 0$$

In addition,

$$\forall \epsilon > 0 \exists \delta = \frac{\epsilon}{\|K\|} > 0 \ \forall x \neq 0, \ \|x\| < \delta \ \exists u = Kx \ \|u\| < \epsilon$$

s.t. $L_f V(x) + L_G V(x) u < 0$ since $\|u\| = \|Kx\| \leq \|K\| \|x\| < \|K\| \delta = \epsilon$

Problem 3

Let $P : \mathbb{R}^2 \to \mathbb{R}$ be continuous, differentiable consider

$$m\dot{v} = -g \triangledown P(q) + F, \ m, g > 0$$

a) Hamiltonian form. Let $x := (q, v)$. Then $\dot{x} = (-\frac{g}{m} \triangledown P(q) + \frac{1}{m} F) = \begin{bmatrix} & I \\ -I & \end{bmatrix} \begin{bmatrix} \frac{g}{m} \triangledown P(q) \\ v \end{bmatrix} + \begin{bmatrix} \\ \frac{1}{m} I \end{bmatrix} F = \begin{bmatrix} & I \\ -I & \end{bmatrix} \triangledown H(x) + G(x)$ given $H(x) = \frac{1}{2} \|\nu\|^2 + \frac{g}{m} P(q)$

b) "CLF". Take H as a CLF candidate. Then, for any x

$$< \triangledown H(x), \begin{bmatrix} & I \\ -I & \end{bmatrix} \triangledown H(x) + G(x) F >= < \triangledown H(x), \begin{bmatrix} & I \\ -I & \end{bmatrix} \triangledown H(x) > + < \triangledown H(x), G(x) F >=$$

$$[< \triangledown H(x), \begin{bmatrix} & I \\ -I & \end{bmatrix} \triangledown H(x) >= L_f H(x) = 0] = \frac{1}{m} < v, F >$$

Strictly speaking, H is no CLF, but it reveals how to choose F s.t. the origin is GAS.

For any point x for which there exists no control F s.t. $< \triangledown H(x), \begin{bmatrix} & I \\ -I & \end{bmatrix} \triangledown H(x) + G(x) F >< 0$

Choose $F = 0$. Why? Using the Krasovsky-Lasallle inv. principle, we conclude that the origin is GAS, since any solution in $\{x | \dot{H}(x) = 0\}$ verifies $v(t) \equiv 0$, implying $\dot{v}(t) \equiv 0$ s.t.

$$0 = -\frac{g}{m} \triangledown P(q(t)) + \frac{1}{m} P(t)$$

The last part equals 0. Since $F = 0$ (by choice) and $\triangledown P(q) = 0$ iff $q = 0$ we conclude that $\dot{H}(x) = 0$ can only be "maintained" at the origin.

Problem 4

Consider

$$\dot{x_1} = x_2$$
$$\dot{x_2} = -u x_2 + u^3$$

show that $V(x) = \frac{1}{2} x_1^2 + \frac{1}{2}(x_1 + x_2)^2$ is CLF and let $V : \mathbb{R}^n \to \mathbb{R}$ be defined by

$$\ddot{x} + u\dot{x} - u^3 = 0$$

For any x and u, we have $< \nabla V(x), f(x,u) > = \cdots = x_1(2x_2 - ux_2 + u^3) + x_2(x_2 - ux_2 + u^3) = x_1 h_1 + x_2 h_2$

Image to be inserted

Hence if $u < 0$ and $-u$ "large", then we can render $< \nabla V(x), f(x,u) > < 0$.

7.4 Exercise 4

Consider

$$\begin{cases} \dot{x}_1 = f_1(x_1) + g_1(x_1)x_2 \\ \dot{x}_2 = f_2(x_1) + g_2(x_1,x_2)u \end{cases} \tag{18}$$

Using the "preliminary control"

$$\begin{cases} \dot{x}_1 = f_1(x_1) + g_1(x_1)x_2 \\ \dot{x}_2 = \check{u} \end{cases} \tag{19}$$

$$u = \frac{1}{g_2(x_1,x_2)}(\check{u} - f_2(x_1,x_2))$$

Idea: Look at the upper(-most) system only and consider x_2 as a "virtual control".

Assumptions: Suppose

- $\exists$ CLF V_1;
- $\exists$ (smooth) feedback α_1 s.t. $L_{f_1+g_1\alpha_1}V_1 < 0$.

Now, add and subtract $g_1\alpha_1$ in **??** s.t.

$$\begin{cases} \dot{x}_1 = f_1(x_1) + g_1(x_1)\alpha_1(x_1) + g_1(x_1)(x_2 - \alpha_1(x_1)) \\ \dot{x}_2 = \check{u} \end{cases} \tag{20}$$

Next, introduce $(e_1, e_2) := (x_1 - 0, x_2 - \alpha_1(x_1))$ s.t.

$$\begin{cases} \dot{e}_1 = f_1(e_1) + g_1(e_1)\alpha_1(e_1) + g_1(e_1)e_2 \\ \dot{e}_2 = \check{u} - \dot{\alpha}_1(e_1) \end{cases} \tag{21}$$

Problem 1:

$$\begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \begin{pmatrix} 0 \\ 1 \end{pmatrix} u$$

Proof. 1. Choose "virtual control":

$$x_2 = -(k+1)x_1 =: \alpha_1(x_1), \quad k > 0$$

The origin of $\dot{x}_1 = -kx_1$ is GAS.
(Take $V_1 : R \to R, \quad x_1 \mapsto \frac{1}{2}x_1^2$ s.t. $\dot{V}_1(x_1) = -kx_1^2 < 0$ for all $x_1 \neq 0$)

2. Error coordinates:
   Let $(e_1, e_2) := (x_1 - 0, x_2 - \alpha_1(x_1))$ s.t.

$$\dot{e}_1 = -ke_1 + e_2$$

$$\dot{e}_2 = u + (k+1)(-ke_1 + e_2)$$

3. "Composite CLF":
   Define $V : R \times R \to R, \quad (e_1, e_2) \mapsto V_1(e_1) + \frac{1}{2}e_2^2$ s.t.

$$\dot{V}(e_1, e_2) = -ke_1^2 + e_2(u + (k+1)(-ke_1 + e_2) + e_1)$$

4. Choose control:
   Let $u = -e_1 - (k+1)(e_2 - ke_1) - ke_2$
   s.t. $\dot{V}(e_1, e_2) = -ke_1^2 - ke_2^2 < 0$ for all $(e_1, e_2) \neq (0, 0)$

Remark: The closed-loop system reads:

$$\begin{pmatrix} \dot{e}_1 \\ \dot{e}_2 \end{pmatrix} = \begin{pmatrix} -k & 1 \\ -1 & -k \end{pmatrix} \begin{pmatrix} e_1 \\ e_2 \end{pmatrix}$$

□

Problem 2:

$$\dot{x}_1 = x_1(x_2 - k), \quad k > 0$$

$$\dot{x}_2 = u$$

Proof. 1. $x_2 = 0 =: \alpha_1(x_1)$
   The origin of $\dot{x}_1 = -kx_1$ is GAS ($V_1(x_1) = \frac{1}{2}x_1^2$)

2. $(e_1, e_2) := (x_1, x_2)$ s.t.

$$\dot{e}_1 = e_1(e_2 - k)$$

$$\dot{e}_2 = u$$

3. $V(e_1, e_2) = V_1(e_1) + \frac{1}{2}e_2^2$ s.t.

$$\dot{V}(e_1, e_2) = -ke_1^2 + e_2(e_1^2 + u)$$

4. $u = -e_1^2 - ke_2$

□

Problem 3:

$$\dot{x}_1 = x_1(x_2 - k)$$

$$\dot{x}_2 = x_2(x_3 - k) - x_1^2$$

$$\dot{x}_3 = u$$

Proof. 1. From problem 2:
$\dot{x}_2 = x_2(x_3 - k) - x_1^2 = -x_1^2 - kx_2 = u$ in Problem 2.
The origin of

$$\dot{x}_1 = x_1(x_2 - k)$$
$$\dot{x}_2 = x_2(x_3 - k) - x_1^2$$

is GAS.
And this is true for $x_3 = 0 =: \alpha_2(x_1, x_2)$.

2. $(e_1, e_2, e_3) := (x_1 - 0, x_2 - \alpha_1(x_1), x_3 - \alpha_2(x_1, x_2))$ s.t.

$$\dot{e}_1 = e_1(e_2 - k)$$
$$\dot{e}_2 = e_2(e_3 - k) - e_1^2$$
$$\dot{e}_3 = u$$

3. $V(e_1, e_2, e_3) = V_1(e_1) + \frac{1}{2}e_2^2 + \frac{1}{2}e_3^2$ s.t.

4. $u = -e_2^2 - ke_3$

□

Problem 4:

$$\dot{x}_1 = x_1(x_2 - k)$$
$$\dot{x}_2 = x_2(x_3 - k) - x_1^2$$
$$\dot{x}_3 = x_3(x_4 - k) - x_2^2$$
$$\dot{x}_4 = u$$

Proof. 1. Is GAS for

$$x_3(x_4 - k) - x_2^2 = -x_2^2 - kx_3$$

which is attained for $x_4 = 0 =: \alpha_3(x_1, x_2, x_3)$.

2.

$$\dot{e}_1 = e_1(e_2 - k)$$
$$\dot{e}_2 = e_2(e_3 - k) - e_1^2$$
$$\dot{e}_3 = e_3(e_4 - k) - e_2^2$$
$$\dot{e}_4 = u$$

...

3. $u = -e_3^2 - ke_4$

□

Problem 5:

$$\dot{x}_1 = x_1(x_2 - k)$$
$$\dot{x}_2 = x_2(x_3 - k) - x_1^2$$
$$\dots$$
$$\dot{x}_i = x_i(x_{i+1} - k) - x_{i-1}^2$$
$$\dots$$
$$\dot{x}_n = u$$

Proof. We will always have $u = e_{n-1}^2 - ke_n$.
Let $V : R \times \cdots \times R \to R, \quad (e_1, \dots e_n) \mapsto \sum_{i=1}^n V_i(e_i)$, where $V_i(e_i) = \frac{1}{2}e_i^2, \quad i = 2, \dots n.$
We have $\dot{V}(e_1, \dots e_n) = L_{f_1+g_1\alpha_1}V_1(e_1) - k\sum_{i=2}^{n-1} e_i^2 + e_n u + e_{n-1}g_{n-1}(x_1, \dots x_{n-1})e_n - e_n\dot{\alpha}_{n-1}(x_1, \dots x_{n-1})$.
We observe that for α_i being zero, the inequality

$$e_{n-1}g_{n-1}(x_1, \dots x_{n-1})e_n - e_n\dot{\alpha}_{n-1}(x_1, \dots x_{n-1}) + e_n u < 0$$

hence $e_{n-1}^2 e_n + e_n u < 0$ for non-zero e.
It is solved by $u = e_{n-1}^2 - ke_n, \quad k > 0$. □

7.5 Exercise 5

Consider the SISO system

$$\dot{x} = f(x) + g(x)(u + \sigma(x))$$
$$y = s(x)$$

$f, g : R^n \to R^n, \quad \sigma : R^n \to R$ and bounded, $\quad s : R^n \to R$

Design steps for SMC:

1. If no output is provided, design a sliding surface $S := \{x \in R^n | s(x) = 0\}$ s.t.
   (a) the system has rel. degree one;
   (b) for $y(t) \equiv 0$, all solutions converge to the origin ("zero dynamics" have GAS origin)
2. Choose a control s.t. the sliding surface is reached (in finite time), e.g.

$$v(x) = -\frac{1}{L_g s(x)}(L_f s(x) + \hat{u} \cdot sgn(s(x))), \quad \hat{u} > 0$$

Problem 1:

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

Sliding surface $S, \quad s : R^2 \to R, \quad (x_1, x_2) \mapsto x_2$

Proof. (a) For the given S, we have $L_g s(x) = 1$ for any $x \in R^2$.
Moreover, from

$$\dot{s}(x) = L_f s(x) + L_g s(x)u$$

(we want $= 0$) we have that for

$$u = -\frac{L_f s(x)}{L_g s(x)} = -x_2$$

the "dynamics on S" (i.e. $x_2 = 0$) reduced to

$$\dot{\eta} = -\eta^3$$

whose origin is GAS.

(b) Consider

$$u = -\frac{1}{L_g s(x)}(L_f s(x) + \hat{u} \cdot sgn(s(x))) = -x_2 - \hat{u} \cdot sgn(x_2), \quad \hat{u} > 0$$

such that $x(t)$ "tends to S" in finite time (phase 1). Moreover, "on S", $x(t)$ converges to the origin $t \to +\infty$ (phase 2).

□

Remark: Given a system in regular form

$$x = (\eta, \xi)^T$$

$$\dot{\eta} = f_1(\eta, \xi)$$

$$\dot{\xi} = f_2(\eta, \xi) + g_2(\eta, \xi)u$$

choose $s(x) = \xi - \Phi(\eta)$, s.t. Φ as. stabilizes $\dot{\eta} = f_1(\eta, \Phi(\eta))$.

Problem 2:

$$\dot{x}_1 = -x_1 \cos x_2 + x_1 x_2$$

$$\dot{x}_2 = x_1 \cos x_1 + \sigma(x) + u$$

Proof. (a) (For the design of sliding surface pretend that uncertainty $\sigma(x) = 0$)
Let $S := \{x \in R^2 | s(x) = 0\}$ be def. by $s : R^2 \to R, \quad (x_1, x_2) \mapsto x_2(-\Phi(x_1) = 0)$. We have $L_g s(x) = 1$ for all $x \in R^2$.
From

$$\dot{s}(x) = L_f s(x) + L_g s(x)u$$

(we want $= 0$) s.t. for $u = -\frac{L_f s(x)}{L_g s(x)}(= -x_1 \cos x_1)$ the "dynamics on S" (i.e. $x_2 = 0$) reads

$$\dot{\eta} = -\eta$$

whose origin is GAS.

(b) Take

$$u = -\frac{1}{L_g s(x)}(L_f s(x)+(\hat{u}+\beta(x)|L_g s(x)|)\cdot sgn(s(x)))(= -x_1 \cos x_1-(\hat{u}+(x_1^2+x_2^2))\cdot sgn(x_2)), \ \ \hat{u} > 0$$

Consider the Lyapunov(-like) function $V(x) = \frac{1}{2}s(x)^2$ s.t.

$$\dot{V}(x) = s(x)(L_f s(x) + L_g s(x)(u + \sigma(x)))$$

Choosing u as above
$\dot{V}(x) = s(x)(-(\hat{u} + \beta(x)|L_g s(x)|) \cdot sgn(s(x)) + \sigma(x)L_g s(x)) \leq -(\hat{u} + \beta(x)|L_g s(x)|)|s(x)| + |\sigma(x)||L_g s(x)||s(x)| \leq -\hat{u}|s(x)| < 0$ for $s(x) \neq 0$

□

Problem 3:

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

$$\dot{x}_2 = -x_1^3 + \sigma(x) + u$$

$s(x) = x_2 + x_1, \quad u = -x_2 + x_1^3 - 2 \cdot sgn(s(x))$

Proof. (a) Given S, we have $L_g s(x) = 1$ for all $x \in R^2$. The "dynamics on S" (i.e. $x_1 + x_2 = 0$) reads

$$\dot{\eta}_1 = -\eta_1$$

$$\dot{\eta}_2 = -\eta_2$$

whose origin is GAS.

(b) Take $V(x) = \frac{1}{2}s(x)^2$ s.t.
$\dot{V}(x) = s(x)(L_f s(x)+L_g s(x)(u+\sigma(x))) \leq -\hat{u}|L_g s(x)||s(x)|+|\sigma(x)||L_g s(x)||s(x)| \leq [|\sigma(x)| \leq c] \leq -(\hat{u} - c)|L_g s(x)||s(x)|$.
Hence, for $c < \hat{u} = 2$ there exists $\varepsilon > 0$ s.t. $\dot{V}(x) \leq -\varepsilon|s(x)| < 0$ for $s(x) \neq 0$

□

7.6 Exercise 6

Problem 1:

$$\dot{x} = xu(x^2 + u)$$

$$\dot{y} = h(x)$$

$s : R \times R \to R, \quad (u, y) \mapsto uy^2 + u^2 y$
$S : R \to R, \quad x \mapsto \frac{x^2}{2}$

Proof. Clearly, S is non-negative. Moreover:
$\dot{S}(x) = x^2 u(x^2 + u) = x^4 u + x^2 u^2 = [h(x) = x^2] = s(u, x^2)$
for all $x, u \in R$ with $h : R \to R, x \mapsto x^2$. □

Problem 2:

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

$$y = x$$

$s : R^n \times R^n \to R, \quad (u, y) \mapsto < u, y >$

Proof. For any $x_0 \in R^n$, we have

$$S_a(x_0) = \sup_{u:[0,t]\to R^n,\ t\geq 0,\ x(0)=x_0} (-\int_0^t < u(\tau), y(\tau) > d\tau) =$$

$$= \sup_{-//-} (-\frac{1}{2}\int_0^t \frac{d}{d\tau}||x(\tau)||^2 d\tau) = \sup_{-//-} (-\frac{1}{2}||x(t)||^2 + \frac{1}{2}||x(0)||^2) \leq \frac{1}{2}||x_0||^2$$

$\Longrightarrow$ av. storage is finite $\Longrightarrow$ system is dissipative. Moreover, we have for any $x_0 \in R^n$,

$$S_r(x_0) = \inf_{u:[-t,0]\to R^n,\ t\geq 0,\ x(-t)=0,\ x(0)=x_0} \int_{-t}^0 < u(\tau), y(\tau) > d\tau = \inf_{-//-} (\frac{1}{2}||x_0||^2 - \frac{1}{2}||x(-t)||^2) = \frac{1}{2}||x_0||^2$$

($S_a = S_r \Longrightarrow$ this is a unique stor. func.)
Hence the (lossless) system is reachable (from 0 to any x_0). □

Problem 3:

Proof. Consider the Lyapunov func. cand. $V(x) = S_1(x_1) + S_2(x_2)$ s.t.
$\dot{V}(x) \leq S_1(u_1, y_1) + S_2(u_2, y_2) = S_1(u_1, y_1) + S_2(y_1, -u_1) = 0 \Longrightarrow$ origin is stable.

□

Remark: the above problem captures many stability results (in the frequency domain). Particular choices of supply rates are:

- $s_i(u_i, y_i) = ||u_i||^2 - ||y_i||^2, i = 1, 2$ (small-gain theorem);
- $s_i(u_i, y_i) = < u_i, y_i >, i = 1, 2$ (positive operator theorem);
- $s_1(u_1, y_1) = < u_1 + ay_1, u_1 + by_1 >$
  $s_2(u_2, y_2) = -ab < u_2 - \frac{1}{a}y_2, u_2 - \frac{1}{b}y_2 >$ (conic operator theorem).

Problem 4:

$$\dot{x} = f(x) + G(x)u$$

$$y = h(x)$$

$s : R^m \times R^m \to R, \quad (u, y) \mapsto ||u||^2 - ||y||^2$

Proof. Take $V = S$ s.t.

$$\dot{V}(x) \leq ||u||^2 - ||h(x)||^2, \ \forall x \in R^n, \ \forall u \in R^m$$

Then the (continuous) state feedback $u = \gamma h(x)$ for some $|\gamma|^2 < 1$, s.t.

$$\dot{V}(x) \leq (|\gamma|^2 - 1)||h(x)||^2 < 0, \ \forall x \neq 0$$

□

Problem 5:

Proof. Take $S(x) =< x, P_x >$ s.t.

$$\dot{S}(x) =< x, (PA + A^T P)x > +2 < x, PBu >$$

Add and subtract $\gamma^2||u||^2$ and $\frac{1}{\gamma^2}||B^T Px||^2$.

$$\dot{S}(x) =< x, (PA + A^T P + \frac{1}{\gamma^2}PBB^T P)x > +\gamma^2||u||^2 - \gamma^2||u\frac{1}{\gamma^2}B^T Px||^2$$

Add and subtract $||y||^2$.

$$\dot{S}(x) =< x, (PA + A^T P + \frac{1}{\gamma^2}PBB^T P + C^T C)x > +\gamma^2||u||^2 - ||y||^2 - \gamma^2||u - \frac{1}{\gamma^2}B^T Px||^2$$

$$\dot{S}(x) \leq \gamma^2||u||^2 - ||y||^2$$

□

7.7 Exercise 7

Definition. A mapping $\Phi : R \to R, \ u \mapsto \Phi(u)$, belongs to the sector

- $[0, +\infty]$ if $u\Phi(u) \geq 0, \ \forall u \in R$;
- $[\alpha, +\infty]$ if $u(\Phi(u) - \alpha u) \geq 0, \ \forall u \in R$ and some $\alpha \in R$;
- $[0, \beta]$ if $\Phi(u)(\Phi(u) - \beta u) \leq 0, \ \forall u \in R$ and some $\beta \in R$;
- $[\alpha, \beta]$ if $(\Phi(u) - \alpha u)(\Phi(u) - \beta u) \leq 0, \ \forall u \in R$ and some $\alpha, \beta \in R$;

Notation: we write, e.g., $\Phi \in [0, +\infty]$.

Problem 1:

$$\dot{x} = x^3 - kx + u, \ k > 0$$

$$y = x$$

Proof. Take, e.g., $S : R \to R, x \mapsto \frac{x^2}{2}$ $(S \geq 0)$ s.t.

$$\dot{S}(x) = x^2(x^2 - k) + yu \leq yu$$

whenever $x \in [-\sqrt{k}, \sqrt{k}]$.

Let $\bar{x} \in R$ and take $u = -\bar{x}^3 + k\bar{x}$ with init. condition $x(0) = \bar{x}$, s.t. we have $x(t) = \bar{x}$ for all $t \geq 0$. If the system is passive, then along this (constant) solution we must have

$$S(x(t)) - S(\bar{x}) \leq \int_0^t u(\tau)y(\tau)d\tau, \ t \geq 0$$

This inequality, however, is violated for $\bar{x} \notin [-\sqrt{k}, \sqrt{k}]$ and hence $[-\sqrt{k}, \sqrt{k}]$ must be the largest interval. □

Problem 2:

$$\dot{x} = -x + \frac{1}{\beta}h(x) + u, \ \beta > 0$$
$$y = h(x)$$

$S(x) = \int_0^x h(\sigma)d\sigma, \ h \in [0, \beta]$

Proof. Clearly, we have $S \geq 0$ since $h \in [0, \beta]$.
Moreover,

$$\dot{S}(x) = S'(x)\dot{x} = \dot{x}\frac{d}{dx}\int_0^x h(\sigma)d\sigma = h(x)\dot{x} = \frac{1}{\beta}h(x)(h(x) - \beta x) + yu \leq yu$$

since $h \in [0, \beta]$. □

Problem 3:

$$H_1 : \begin{cases} \dot{x}_1 = x_2 \\ \dot{x}_2 = -x_1 + kx_2 + u, \ k > 0 \\ y = x_2 \end{cases}$$

Proof. Take $S : R^2 \to R, \ (x_1, x_2) \mapsto \frac{x_1^2}{2} + \frac{x_2^2}{2}$ s.t. $\dot{S}(x) = uy + ky^2$.
Let $u = -\Phi(y), \ \Phi : R \to R$ satisfying $\Phi \in [l, +\infty]$ for some $l > k$ $(\nu_2 + \rho_1 > 0)$ s.t.

$$\dot{S}(x) = -y\Phi(y) + ky^2 \leq -(l - k)y^2$$

Since the system H_1 is ZSO the origin is GAS. □

Problem 4:

Proof. Take $S(x) = S_1(x_1) + S_2(x_2)$ s.t.

$$\dot{S}(x) \leq < u_1, y_1 > -\rho_1||y_1||^2 - \nu_1||u_1||^2 + < u_2, y_2 > -\rho_2||y_2||^2 - \nu_2||u_2||^2$$

Using that

$$< u_1, y_1 > + < u_2, y_2 >=< u - y_2, y_1 > + < v + y_1, y_2 >=< u, y_1 > + < v, y_2 >$$

and

$$||u_1||^2 = ||u||^2 - 2 < u, y_2 > +||y_2||^2$$
$$||u_2||^2 = ||v||^2 + 2 < v, y_1 > +||y_1||^2$$

we obtain

$\dot{S}(x) = - < \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}, \begin{pmatrix} (\nu_2+\rho_1)I_m & \\ & (\nu_1+\rho_2)I_m \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} > - < \begin{pmatrix} u \\ v \end{pmatrix}, \begin{pmatrix} \nu_1 I_m & \\ & \nu_2 I_m \end{pmatrix} \begin{pmatrix} u \\ v \end{pmatrix} > + < \begin{pmatrix} u \\ v \end{pmatrix}, \begin{pmatrix} I_m & 2\nu_1 I_m \\ -2\nu_2 I_m & I_m \end{pmatrix} \begin{pmatrix} u \\ v \end{pmatrix} > \leq [Coshi - Schwarz] \leq -a||(y_1, y_2)||^2 + b||(u,v)||||(y_1,y_2)|| + c||(u,v)||^2$

with $a = \min\{\nu_2+\rho_1, \nu_1+\rho_2\} > 0,\ b = ||N|| \geq 0$ and $c = ||M|| \geq 0$.

Hence,

$\dot{S}(x) \leq -\frac{1}{2a}(b||(u,v)|| - a||(y_1,y_2)||)^2 + \frac{b^2}{2a}||(u,v)||^2 - \frac{a}{2}||(y_1,y_2)||^2 + c||(u,v)||^2 \leq \frac{b^2+2ac}{2a}||(u,v)||^2 - \frac{a}{2}||(y_1,y_2)||^2$ □

Problem 5:

Proof. Take $V(x) =< x, Px >$ s.t.

$$\dot{V}(x) =< x, (PA + A^T P)x > -2\Phi(y) < x, PB >$$

Add and subtract $2\Phi(y)^2$ and $2\Phi(y)BCx$ yields

$\dot{V}(x) = -\varepsilon < x, Px > - < x, L^T Lx > -2\Phi(y) < x, PB - BC^T > -2\Phi(y)^2 + 2\Phi(y)(\Phi(y) - By) = -\varepsilon < x, Px > -|Lx - \sqrt{2}\Phi(y)|^2 + 2\Phi(y)(\Phi(y) - By) \leq -\varepsilon < x, Px >.$ □