

1 Important Definitions

1.1 Norm

The norm $|x|$ of a vector x is a real valued function with the following properties:

- (i) $|x| \geq 0$ with $|x| = 0$ if and only if $x = 0$
- (ii) $|\alpha x| = |\alpha| |x|$ for any scalar α
- (iii) $|x + y| \leq |x| + |y|$ (triangle inequality)

The norm $|x|$ of a vector x can be thought of as the size or length of the vector x . Similarly, $|x - y|$ can be thought of as the distance between the vectors x and y .

1.2 Induced Norm

Let $|\cdot|$ be a given vector norm. Then for each matrix $A \in \mathcal{R}^{m \times n}$, the quantity $\|A\|$ defined by $\|A\| \triangleq \sup_{\substack{x \neq 0 \\ x \in \mathcal{R}^n}} \frac{|Ax|}{|x|} = \sup_{|x| \leq 1} |Ax| = \sup_{|x|=1} |Ax|$ is called the induced (matrix) norm of A corresponding to the vector norm $|\cdot|$.

Some of the properties of the induced norm that we will often use in this book are summarized as follows:

- (i) $|Ax| \leq \|A\| |x|, \quad \forall x \in \mathcal{R}^n$
- (ii) $\|A + B\| \leq \|A\| + \|B\|$
- (iii) $\|AB\| \leq \|A\| \|B\|$

1.3 \mathcal{L}_p spaces

for functions of time, we define the \mathcal{L}_p norm

$$\|x\|_p \triangleq \left(\int_0^\infty |x(\tau)|^p d\tau \right)^{1/p}$$

for $p \in [1, \infty)$ and say that $x \in \mathcal{L}_p$ when $\|x\|_p$ exists (i.e., when $\|x\|_p$ is finite). The \mathcal{L}_∞ norm is defined as

$$\|x\|_\infty \triangleq \sup_{t \geq 0} |x(t)|$$

and we say that $x \in \mathcal{L}_\infty$ when $\|x\|_\infty$ exists. In the above $\mathcal{L}_p, \mathcal{L}_\infty$ norm definitions, $x(t)$ can be a scalar or a vector function. If x is a scalar function, then $|\cdot|$ denotes the absolute value. If x is a vector function in \mathcal{R}^n then $|\cdot|$ denotes any norm in \mathcal{R}^n .

Example 1 (\mathcal{L}_p spaces) Consider the function $f(t) = \frac{1}{1+t}$. Then,

$$\|f\|_\infty = \sup_{t \geq 0} \left| \frac{1}{1+t} \right| = 1, \quad \|f\|_2 = 1$$

Hence, $f \in \mathcal{L}_2 \cap \mathcal{L}_\infty$ but $f \notin \mathcal{L}_1$; f , however, belongs to \mathcal{L}_{1e} , i.e., for any finite $t \geq 0$, we have

$$\int_0^t \frac{1}{1+\tau} d\tau = \ln(1+t) < \infty$$

- A function f belongs to \mathcal{L}_1 if the integral of the absolute value of f over its entire domain is finite. Mathematically, this can be expressed as:

$$\int |f(x)| dx < \infty$$

In \mathcal{L}_1 , functions are required to have a finite " \mathcal{L}_1 norm."

- A function f belongs to \mathcal{L}_2 if the square of the absolute value of f is Lebesgue integrable, meaning:

$$\int |f(x)|^2 dx < \infty$$

In \mathcal{L}_2 , functions are required to have a finite "L norm."

- A function f belongs to \mathcal{L}_∞ if it is bounded, meaning there exists a constant M such that $|f(x)| \leq M$ for all x . In \mathcal{L}_∞ , functions are required to have a finite "norm," which is essentially the supremum (maximum) of the absolute value of the function.

2 Fundamental Properties

2.1 Existence and Uniqueness

2.1.1 Local Existence and Uniqueness

Let $f(t, x)$ be piecewise continuous in t and satisfy the Lipschitz condition

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

$\forall x, y \in B = \{x \in \mathbb{R}^n \mid ||x - x_0|| \leq r\}, \forall t \in [t_0, t_1]$. Then, there exists some $\delta > 0$ such that the state equation $\dot{x} = f(t, x)$ with $x(t_0) = x_0$ has a unique solution over $[t_0, t_0 + \delta]$.

2.1.2 Lipschitz Condition

Let $f : [a, b] \times D \rightarrow \mathbb{R}^n$ be continuous on some domain $D \subset \mathbb{R}^n$. Suppose that $[\partial f / \partial x]$ exists and is continuous on $[a, b] \times D$. If, for a convex subset $W \subset D$, there is a constant $L \geq 0$ such that

$$\left\| \frac{\partial f}{\partial x}(t, x) \right\| \leq L$$

on $[a, b] \times W$, then

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

for all $t \in [a, b]$, $x \in W$, and $y \in W$.

2.1.3 Locally Lipschitz

If $f(t, x)$ and $[\partial f / \partial x](t, x)$ are continuous on $[a, b] \times D$, for some domain $D \subset \mathbb{R}^n$, then f locally Lipschitz in x on $[a, b] \times D$.

2.1.4 Globally Lipschitz

If $f(t, x)$ and $[\partial f / \partial x](t, x)$ are continuous on $[a, b] \times \mathbb{R}^n$, then f is globally Lipschitz in x on $[a, b] \times \mathbb{R}^n$ if and only if $[\partial f / \partial x]$ is uniformly bounded on $[a, b] \times \mathbb{R}^n$.

2.1.5 Global Existence and Uniqueness

Suppose that $f(t, x)$ is piecewise continuous in t and satisfies

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

$\forall x, y \in \mathbb{R}^n, \forall t \in [t_0, t_1]$. Then, the state equation $\dot{x} = f(t, x)$, with $x(t_0) = x_0$, has a unique solution over $[t_0, t_1]$.

2.1.6 Existence and Uniqueness Theorem

3 Second-Order Systems

3.1 Limit Cycles

We consider a two-dimensional dynamical system of the form

$$\dot{x}(t) = V(x(t))$$

where $V : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ is a smooth function. A trajectory of this system is some smooth function $x(t)$ with values in \mathbb{R}^2 which satisfies this differential equation. Such a trajectory is called closed (or periodic) if it is not constant but returns to its starting point, i.e., if there exists some $t_0 > 0$ such that $x(t + t_0) = x(t)$ for all $t \in \mathbb{R}$. An orbit is the image of a trajectory, a subset of \mathbb{R}^2 . A closed orbit, or cycle, is the image of a closed trajectory. A limit cycle is a cycle which is the limit set of some other trajectory.

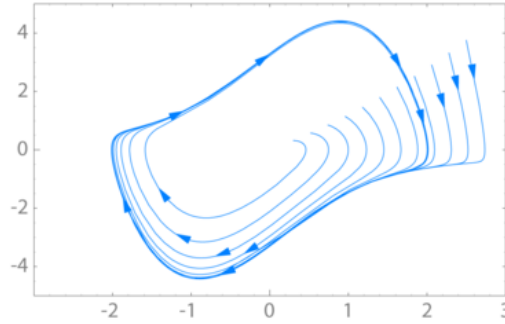


Figure 1: Stable Limit Cycle

3.1.1 Poincaré-Bendixson Criterion

Consider the system $\dot{x} = f(x)$ and let M be a closed bounded subset of the plane such that

- M contains no equilibrium points or contains only one equilibrium point such that the Jacobian matrix $[\partial f / \partial x]$ at this point has eigenvalues with positive real parts. (Hence, the equilibrium point is unstable focus or unstable node.)
- Every trajectory starting in M stays in M for all future time.

Then, M contains a periodic orbit of $\dot{x} = f(x)$.

3.1.2 Benixson Criterion

If, on a simply connected region (\mathcal{C}^1) D of the plane, the expression $\partial f_1 / \partial x_1 + \partial f_2 / \partial x_2$ is not identically zero and does not change sign, then $\dot{x} = f(x)$ has no periodic orbits lying entirely in D .

4 Lyapunov Stability

4.1 Autonomous Systems

4.1.1 Stability Principles

The equilibrium point $x = 0$ of $\dot{x} = f(X)$ is

- stable if, for each $\epsilon > 0$, there is a $\delta = \delta(\epsilon) > 0$ such that

$$\|x(0)\| < \delta \Rightarrow \|x(t)\| < \epsilon, \quad \forall t \geq 0$$

- unstable if it is not stable
- asymptotically stable if it is stable and δ can be chosen such that

$$\|x(0)\| < \delta \Rightarrow \lim_{t \rightarrow \infty} x(t) = 0$$

4.1.2 Quadratic Lyapunov Function

$$V(x) = \frac{1}{2}x^T P x, \quad P > 0 \quad (1)$$

and

$$\begin{aligned} \frac{d}{dt}V(x) &= \frac{1}{2}x^T P A x + (x^T P A x)^T \\ &= \frac{1}{2}2x^T P \dot{x} \\ &= x^T P \dot{x} \end{aligned} \quad (2)$$

$$\dot{V}(x) = x^T P \dot{x} \quad (3)$$

In other words

$$\dot{V}(x) = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}^T \begin{bmatrix} p_{11} & p_{21} \\ p_{12} & p_{22} \end{bmatrix} \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} \quad (4)$$

4.1.3 Asymptotically Stable

Let $x = 0$ be an equilibrium point and $D \subset \mathbb{R}^n$ be a domain containing $x = 0$. Let $V : D \rightarrow \mathbb{R}$ be a continuously differentiable function such that

$$V(0) = 0 \quad \text{and} \quad V(x) > 0 \text{ in } D - \{0\} \quad (5)$$

$$\dot{V}(x) \leq 0 \text{ in } D \quad (6)$$

Then $x = 0$ is stable. Moreover, if

$$\dot{V}(x) < 0 \text{ in } D - \{0\} \quad (7)$$

then $x = 0$ is asymptotically stable.

4.1.4 Globally Asymptotic stable

Let $x = 0$ be an equilibrium point or $\dot{x} = f(x)$. Let $V : \mathbb{R}^n \rightarrow \mathbb{R}$ be a continuously differentiable function such that

$$V(0) = 0 \text{ and } V(x) > 0, \quad \forall x \neq 0$$

$$\|x\| \rightarrow \infty \Rightarrow V(x) \rightarrow \infty$$

4.1.5 Radially Unbounded Function

A radially unbounded function is a function $f : \mathbb{R}^n \rightarrow \mathbb{R}$ for which

$$\|x\| \rightarrow \infty \Rightarrow f(x) \rightarrow \infty$$

Or equivalently

$$\forall c > 0 : \exists r > 0 : \forall x \in \mathbb{R}^n : [\|x\| > r \Rightarrow f(x) > c]$$

4.1.6 Region of attraction

The region of attraction is also called the region of asymptotic stability. The region of attraction may be hard to calculate exactly, but Lyapunov functions can be used to estimate this region.

First of all, the region of attraction must satisfy the following: from the definition of asymptotic stability we see that if there is a Lyapunov function that satisfy the condition of asymptotic stability over a domain \mathbb{D} and if $\Omega_c = \{x \in \mathbb{R}^n | V(x) \leq c\}$ is bounded and contained in \mathbb{D} , then every trajectory starting in Ω_c remains in Ω_c and approaches the origin as $t \rightarrow \infty$

Let $x = 0$ be an asymptotically stable equilibrium point of the system $\dot{x} = f(x)$, where $f : \mathbb{D} \rightarrow \mathbb{R}^n$ is locally lipschitz and $\mathbb{D} \subset \mathbb{R}^n$ is a domain that contains the origin. Let $t \rightarrow \phi(t; x_0)$ be the solution of $\dot{x} = f(x)$ that starts at initial state x_0 at time $t = 0$. The region of attraction of the origin, denoted by R_A , is defined by

$$R_A = \{x_0 \in \mathbb{D} : \phi(t; x_0) \text{ is defined } \forall t \geq 0 \text{ and } \phi(t; x_0) \rightarrow 0 \text{ as } t \rightarrow \infty\}$$

I.e. the region of attraction R_A is the set of all points x_0 in \mathbb{D} such that the solution of

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

is defined for all $t \geq 0$ and converges to the origin as $t \rightarrow \infty$.

Example 2 (Example of region of attraction)

Our domain \mathbb{D} is defined as

$$\mathbb{D} = \{x \in \mathbb{R}^2 : x_1 + x_2 < 1\}$$

where $\dot{V}(x) < 0$. Our region of attraction is defined as

$$\Omega_c = \{x \in \mathbb{R}^2 : V(x) \leq c\}$$

where $c < \min(V(x))$ and $x \in \partial c$ (x is on the border of c).

$$\begin{aligned} x_1 + x_2 &= 1 \\ x_1 &= 1 - x_2 \end{aligned}$$

Then we minimize $V(x)$ with respect of x_2

$$\begin{aligned} \min(V(x))_{x_2} \Big|_{x_1=1-x_2} &= \min_{x_2} \left(\frac{1}{2}((1-x_2)^2 + x_2^2) \right) = 0 \\ \min_{x_2} \left(\frac{1}{2} - x_2 + x_2^2 \right) &= \frac{\partial}{\partial x_2} \left(\frac{1}{2} - x_2 + x_2^2 \right) = 0 \\ -1 + 2x_2 &= 0 \\ x_2 &= \frac{1}{2} \rightarrow x_1 = \frac{1}{2} \end{aligned}$$

Further, we must satisfy $c < V(\frac{1}{2}, \frac{1}{2}) = \frac{1}{4}$. We have therefore the following region of attraction

$$\Omega_{\frac{1}{4}} = \{x \in \mathbb{R}^2 : V(x) \leq \frac{1}{4}\}$$

4.2 The Invariance Principle

4.2.1 Invariant Set

A set M is said to be an invariant set with respect to $\dot{x} = f(x)$ iff

$$x(0) \in M \Rightarrow x(t) \in M, \quad \forall t \in \mathbb{R}$$

4.2.2 Positively Invariant Set

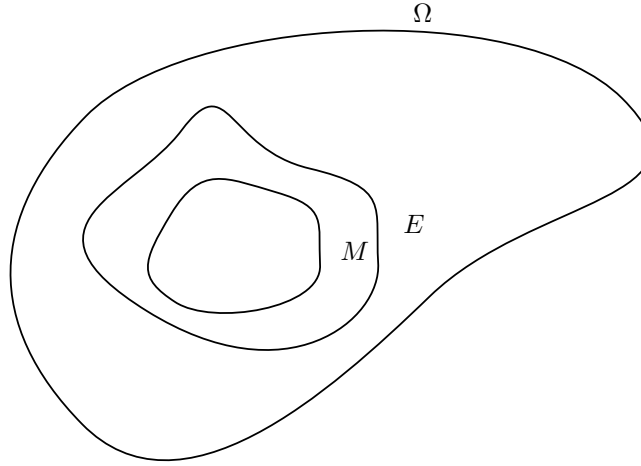
A set M is a positively invariant set with respect to $\dot{x} = f(x)$ iff

$$x(0) \in M \Rightarrow x(t) \in M, \quad \forall t \geq 0$$

4.2.3 Lasalle's Theorem

Lasalle's invariance principle can be used to show the asymptotic stability of an equilibrium point when \dot{V} is negative semi-definite, i.e. $\dot{V} \leq 0$. \dot{V} is semidefinite if \dot{V} is not strictly 0 in origin.

Let $\Omega \subset D$ be a compact set that is positively invariant with respect to $\dot{x} = f(x)$. Let $V : D \rightarrow \mathbb{R}$ be a continuously differentiable function such that $\dot{V}(x) \leq 0$ in Ω . Let E be the set of all points in Ω where $\dot{V}(x) = 0$. Let M be the largest invariant set in E . Then every solution starting in Ω approaches M as $t \rightarrow \infty$.



4.2.4 Barbashin Theorem

Let $x = 0$ be an equilibrium point for

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

Let $V : D \rightarrow \mathbb{R}$ be a continuously differentiable positive definite function on a domain D containing the origin $x = 0$, such that $\dot{V}(x) \leq 0$ in D . Let $S = \{x \in D \mid \dot{V}(x) = 0\}$ and suppose that no solution can stay identically in S , other than the trivial solution $x(t) \equiv 0$. Then, the origin is asymptotically stable. If a A matrix is Hurwitz, we can conclude asymptotic stability.

4.2.5 Krasovskii Theorem

Let $x = 0$ be an equilibrium point for the system

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

. Let $V : \mathbb{R}^n \rightarrow \mathbb{R}$ be a continuously differentiable, radially unbounded, positive definite function such that $V(x) \leq 0$ for all $x \in \mathbb{R}^n$. Let $S = \{x \in \mathbb{R}^n \mid V(x) = 0\}$ and suppose that no solution can stay identically in S , other than the trivial solution $x(t) = 0$. Then, the origin is globally asymptotically stable.

4.2.6 Dominating cross-terms

Completion of squares

$$x_1 x_2 \leq \frac{1}{2}(x_1^2 + x_2^2) = \frac{1}{2} \|x\|_2^2$$

Another alternative is to write \dot{V} as $-x^T Q x$, where $Q = Q^T$ is positive definite

Young's inequality

$$xy \leq \epsilon x^2 + \frac{1}{4\epsilon} y^2, \quad \forall \epsilon > 0 \quad \forall x, y \in \mathbb{R}$$

4.2.7 Handling terms with indeterminate sign

Cauchy-Schwarz Inequality

$$|a_1 x_1 + a_2 x_2 + \dots + a_n x_n| \leq \sqrt{(a_1^2 + a_2^2 + \dots + a_n^2)} \|x\|_2$$

4.3 Linear Systems and Linearization

4.3.1 Hurwitz matrix / Stability matrix

A matrix A is Hurwitz; that is, $\text{Re}\{\lambda_i\} < 0$ for all eigenvalues of A , if and only if for any given positive definite symmetric matrix Q there exists a positive definite symmetric matrix P that satisfies the *Lyapunov equation* (8). Moreover, if A is Hurwitz, then P is the unique solution of (8).

$$PA + A^T P = -Q \quad (8)$$

Another way one can prove asymptotic stability in the origin is by using this symmetric matrix Q . We can write $\dot{V}(x)$ as the following

$$\dot{V}(x) = x^T P \dot{x} + \dot{x}^T P x = -x^T Q x$$

If Q is positive definite, we can conclude asymptotic stability at the origin.

Example 3 (Example using Hurwitz matrix)

We have the following $\dot{V}(x)$

$$\dot{V}(x) \leq -x_2^2 \left(2 - \frac{1}{4\epsilon}\right) - x_1 x_2 - x_1^2(1 - \epsilon)$$

We can then formulate the following Q matrix

$$Q = \begin{bmatrix} 1 - \epsilon & \frac{1}{2} \\ \frac{1}{2} & 2 - \frac{1}{4\epsilon} \end{bmatrix}$$

To ensure that Q is positive definite we can evaluate the principle of minors

$$(1 - \epsilon) \left(w - \frac{1}{4\epsilon}\right) - \frac{1}{4} > 0$$

while $(1 - \epsilon) > 0$. We can solve the last inequality, and ϵ must satisfy

$$\frac{2 - \sqrt{2}}{4} < \epsilon < \frac{2 + \sqrt{2}}{4}$$

to ensure that the origin is asymptotically stable.

4.3.2 Indirect Method

Let $x = 0$ be an equilibrium point for the nonlinear system

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

where $f : D \rightarrow \mathbb{R}^n$ is continuously differentiable and D is a neighborhood of the origin. Let

$$A = \left. \frac{\partial f}{\partial x_i} \right|_{x=0}$$

Then,

1. The origin is asymptotically stable if $\text{Re}(\lambda_i) < 0$ for all eigenvalues λ_i of A .
2. The origin is unstable if $\text{Re}(\lambda_i) > 0$ for one or more of the eigenvalues λ_i of A .

4.4 Comparison Functions

4.4.1 Class \mathcal{K} and Class \mathcal{K}_∞ Definitions

A continuous function $\alpha : [0, a) \rightarrow [0, \infty)$ is said to belong to class \mathcal{K} if it is strictly increasing and $\alpha(0) = 0$. It is said to belong to class \mathcal{K}_∞ if $a = \infty$ and $\alpha(r) \rightarrow \infty$ as $r \rightarrow \infty$.

4.4.2 Class \mathcal{KL} Definition

A continuous function $\beta : [0, a) \times [0, \infty) \rightarrow [0, \infty)$ is said to belong to class \mathcal{KL} if, for each fixed s , the mapping $\beta(r, s)$ belongs to \mathcal{K} with respect to r and, for each fixed r , the mapping $\beta(r, s)$ is decreasing with respect to s and $\beta(r, s) \rightarrow \infty$ as $s \rightarrow \infty$.

4.4.3 Class \mathcal{K} , \mathcal{K}_∞ and \mathcal{KL} Rules

Let α_1 and α_2 be class \mathcal{K} functions on $[0, a)$, α_3 and α_4 be class \mathcal{K}_∞ functions on $[0, \infty)$.

Example 4 (Class \mathcal{K})

$\alpha(r) = \tan^{-1} r$ is strictly increasing since $\dot{\alpha}(r) = \frac{1}{(1+r^2)} > 0$. It belongs to class \mathcal{K} , but not to class \mathcal{K}_∞ since $\lim_{r \rightarrow \infty} \alpha(r) = \frac{\pi}{2} < \infty$.

Example 5 (Class \mathcal{KL})

$\beta(r, s) = \frac{r}{(ksr+1)}$, for any positive real number k , is strictly increasing in r since

$$\frac{\partial \beta}{\partial r} = \frac{ks}{(ksr+1)^2} > 0$$

and strictly decreasing in s since

$$\frac{\partial \beta}{\partial s} = \frac{-kr^2}{(ksr+1)^2} < 0$$

Moreover $\beta(r, s) \rightarrow 0$ as $s \rightarrow \infty$

4.4.4 Introducing class \mathcal{K} functions in Lyapunov's stability theorem

Let α_1 and α_2 be class \mathcal{K} functions on $[0, a)$, α_3 and α_4 be class \mathcal{K}_∞ functions, and β be a class \mathcal{KL} function. Denote the inverse of α_i by α_i^{-1} . Then,

- α_1^{-1} is defined on $[0, \alpha_1(a))$ and belongs to class \mathcal{K} .
- α_3^{-1} is defined on $[0, \infty)$ and belongs to class \mathcal{K}_∞ .
- $\alpha_1 \circ \alpha_2$ belongs to class \mathcal{K} .
- $\alpha_3 \circ \alpha_4$ belongs to class \mathcal{K}_∞ .
- $\sigma(r, s) = \alpha_1(\beta(\alpha_2(r), s))$ belongs to class \mathcal{KL} .

Class \mathcal{K} and class \mathcal{KL} functions enter into Lyapunov analysis through the next subsections

4.4.5 Lyapunov's analysis with Class \mathcal{K} functions

Let $V : D \rightarrow \mathbb{R}$ be a continuous positive definite function defined on a domain $D \subset \mathbb{R}^n$ that contains the origin. Let $B_r \subset D$ for some $r > 0$. Then, there exist class \mathcal{K} functions α_1 and α_2 , defined on $[0, r]$, such that

$$\alpha_1(\|x\|) \leq V(x) \leq \alpha_2(\|x\|)$$

for all $x \in B_r$. If $D = \mathbb{R}^n$, the functions α_1 and α_2 will be defined on $[0, \infty)$ and the foregoing inequality will hold for all $x \in \mathbb{R}^n$. Moreover, if $V(x)$ is radially unbounded, then α_1 and α_2 can be chosen to belong to class \mathcal{K}_∞ .

4.4.6 Lyapunov's analysis with Class \mathcal{KL} functions

Consider the scalar autonomous differential equation

$$\dot{y} = -\alpha(y), \quad y(t_0) = y_0$$

where α is a locally Lipschitz class \mathcal{K} function defined on $[0, a)$. For all $0 \leq y_0 < a$, this equation has a unique solution $y(t)$ defined for all $t \geq t_0$. Moreover,

$$y(t) = \sigma(y_0, t - t_0)$$

where σ is a class \mathcal{KL} function defined on $[0, a) \times [0, \infty)$.

4.5 Nonautonomous Systems

Throughout this section we will consider the following Nonautonomous system

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

4.5.1 Uniformity

Proving uniformity means that the initial condition x_0 has no effect on the stability of the system.

Uniformity \mapsto convergence rate not dependent on t_0 .

4.5.2 Stability definitions

The equilibrium point $x = 0$ of 4.5 is

- stable if, for each $\varepsilon > 0$, there is $\delta = \delta(\varepsilon, t_0) > 0$ such that

$$\|x(t_0)\| < \delta \Rightarrow \|x(t)\| < \varepsilon, \quad \forall t \geq t_0 \geq 0$$

- uniformly stable if, for each $\varepsilon > 0$, there is $\delta = \delta(\varepsilon) > 0$, independent of t_0 , such that (4.16) is satisfied.
- unstable if it is not stable.
- asymptotically stable if it is stable and there is a positive constant $c = c(t_0)$ such that $x(t) \rightarrow 0$ as $t \rightarrow \infty$, for all $\|x(t_0)\| < c$.
- Asymptotically stable (AS) if

it is stable

there exists $c > 0$ s.t. for all $t_0 \geq 0$ and $\eta > 0$, there exists $T(t_0, \eta) > 0$ satisfying

$$\|x(t_0)\| \leq c \Rightarrow \|x(t)\| \leq \eta \quad \forall t \geq t_0 + T(t_0, \eta)$$

- Uniformly asymptotically stable (UAS) if

it is uniformly stable (US)

there exists $c > 0$ s.t. for all $t_0 \geq 0$ and $\eta > 0$, there exists $T(\eta) > 0$ satisfying

$$\|x(t_0)\| \leq c \Rightarrow \|x(t)\| \leq \eta \quad \forall t \geq t_0 + T(\eta)$$

- Globally uniformly asymptotically stable (GUAS) if

it is globally uniformly stable (GUS)

for all $\eta, c > 0$ and $t_0 \geq 0$ there exists $T(\eta, c) > 0$ s.t.

$$\|x(t_0)\| \leq c \Rightarrow \|x(t)\| \leq \eta \quad \forall t \geq t_0 + T(\eta, c)$$

4.5.3 Equilibrium Points

$x^* \in \mathbb{R}^n$ is an equilibrium point of $\dot{x} = f(t, x)$ if $f(t, x^*) = 0$ for all $t \geq 0$.

Example 6 (Example of equilibrium point)

$\dot{x} = -\frac{a(t)x}{1+x^2}$ has the equilibrium point $x^* = 0$. We can see this by evaluating $f(t, x^*)$

$\dot{x} = -\frac{a(t)}{1+x^2} + b(t)$ where $b(t) \neq 0 \forall t$ has no equilibrium points

4.5.4 Stability definitions with \mathcal{K} and \mathcal{KL} classes

For $\dot{x} = f(t, x)$, the equilibrium point $x^* = 0$ is

- Uniformly stable (US) if $\exists \alpha \in \mathcal{K}, \exists c > 0$ such theoremseparator

$$\|x(t)\| \leq \alpha(\|x(t_0)\|) \quad \forall t \geq t_0, \quad \|x(t_0)\| \leq c, \quad \forall t_0 \geq 0$$

- globally uniformly stable (GUS) if $\exists \alpha \in \mathcal{K}_\infty$ such that

$$\|x(t)\| \leq \alpha(\|x(t_0)\|) \quad \forall t \geq t_0, \quad \forall t_0 \geq 0$$

- exponentially stable if $\exists c, k, \lambda > 0$

$$\|x(t)\| \leq k\|x(t_0)\|e^{-\lambda(t-t_0)} \quad t \geq t_0 \geq 0, \forall \|x(t_0)\| < c$$

- globally exponentially stable if $\exists k, \lambda > 0$

$$\|x(t)\| \leq k\|x(t_0)\|e^{-\lambda(t-t_0)} \quad t \geq t_0 \geq 0, \forall x(t_0)$$

4.5.5 Time-varying Lyapunov function candidates

$V(t, x)$ is positive definite if

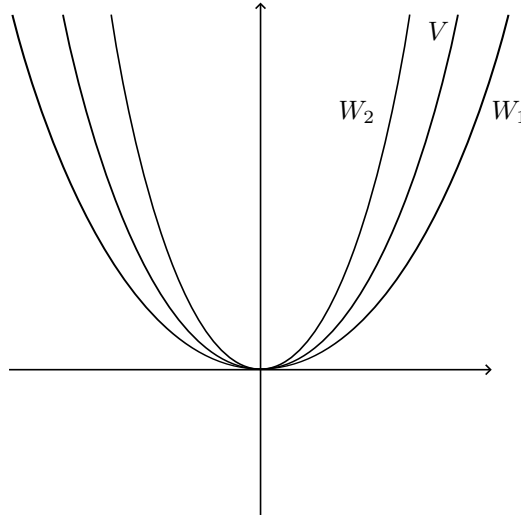
$$\left. \begin{array}{l} V(t, 0) = 0 \\ V(t, x) \geq W_1(x) \end{array} \right\} \forall t \geq 0, \text{ for some positive definite function } W_1(x)$$

and

- $(t, x) \rightarrow V(t, x)$ is positive semidefinite if $x \rightarrow W_1(x)$ positive semidefinite
- $(t, x) \rightarrow V(t, x)$ is radially unbounded if $x \rightarrow W_1(x)$ radially unbounded

$V(t, x)$ is decrescent if

$$\left. \begin{array}{l} V(t, 0) = 0 \\ V(t, x) \leq W_2(x) \end{array} \right\} \forall t \geq 0, \text{ for some positive definite function } W_2(x)$$



4.5.6 Lyapunov theorem: Uniform Stability

If there exist a function $V(t, x) : [0, \infty) \times \mathbb{D} \rightarrow \mathbb{R}$ and $W_i : \mathbb{D} \rightarrow \mathbb{R}$ continuous positive definite such that

- V is C^1
- $W_1(x) \leq V(t, x) \leq W_2(x) \quad \forall (t, x) \in [0, \infty) \times \mathbb{D}$
- $\dot{V}(t, x) \leq 0 \quad \forall (t, x) \in [0, \infty) \times \mathbb{D}$

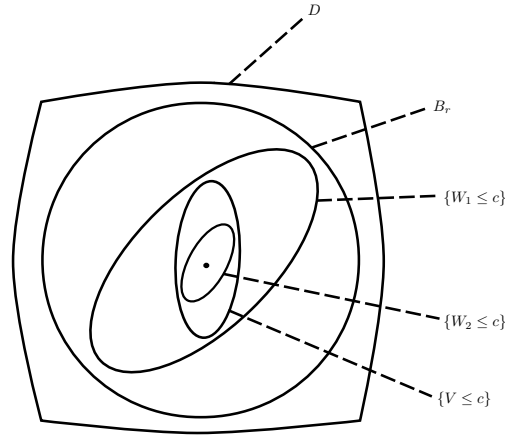
then $x = 0$ is uniformly stable.

Proof 1 The derivative of V along the trajectories of (4.15) is given by

$$\dot{V}(t, x) = \frac{\partial V}{\partial t} + \frac{\partial V}{\partial x} f(t, x) \leq 0$$

Choose $r > 0$ and $c > 0$ such that $B_r \subset D$ and $c < \min_{\|x\|=r} W_1(x)$. Then, $\{x \in B_r \mid W_1(x) \leq c\}$ is in the interior of B_r . Define a time-dependent set $\Omega_{t,c}$ by

$$\Omega_{t,c} = \{x \in B_r \mid V(t, x) \leq c\}$$



The set $\Omega_{t,c}$ contains $\{x \in B_r \mid W_2(x) \leq c\}$ since

$$W_2(x) \leq c \Rightarrow V(t, x) \leq c$$

On the other hand, $\Omega_{t,c}$ is a subset of $\{x \in B_r \mid W_1(x) \leq c\}$ since

$$V(t, x) \leq c \Rightarrow W_1(x) \leq c$$

Thus,

$$\{x \in B_r \mid W_2(x) \leq c\} \subset \Omega_{t,c} \subset \{x \in B_r \mid W_1(x) \leq c\} \subset B_r \subset D$$

for all $t \geq 0$. These five nested sets are sketched in Figure 4.7. The setup of Figure 4.7 is similar to that of Figure 4.1, except that the surface $V(t, x) = c$ is now dependent on t , and that is why it is surrounded by the time-independent surfaces $W_1(x) = c$ and $W_2(x) = c$.

4.5.7 Lyapunov theorem: Uniform Asymptotical Stability

If there exist a function $V(t, x) : [0, \infty) \times \mathbb{D} \rightarrow \mathbb{R}$ and $W_i : \mathbb{D} \rightarrow \mathbb{R}$ continuous positive definite such that

- i) V is C^1
- ii) $W_1(x) \leq V(t, x) \leq W_2(x) \quad \forall (t, x) \in [0, \infty) \times \mathbb{D}$
- iii) $\dot{V}(t, x) \leq -W_3(x) \quad \forall (t, x) \in [0, \infty) \times \mathbb{D}$

then $x = 0$ is uniformly asymptotically stable.

4.5.8 Lyapunov theorem: Global Uniformly Asymptotical Stability

If there exist a function $V(t, x) : [0, \infty) \times \mathbb{R}^2 \rightarrow \mathbb{R}$ and $W_i : \mathbb{R}^2 \rightarrow \mathbb{R}$ continuous positive definite such that

- i) V is C^1
- ii) $W_1(x) \leq V(t, x) \leq W_2(x) \quad \forall (t, x) \in [0, \infty) \times \mathbb{D}$
- iii) $\dot{V}(t, x) \leq -W_3(x) \quad \forall (t, x) \in [0, \infty) \times \mathbb{D}$
- iv) W_1 is radially unbounded

then $x = 0$ is globally uniformly asymptotically stable.

4.5.9 Lyapunov theorem: Exponential Stability

If there exist a function $V(t, x) : [0, \infty) \times \mathbb{D} \rightarrow \mathbb{R}$ and constants $a, k_1, k_2, k_3 > 0$ such that

- i) V is C^1
- ii) $k_1 \|x\|^a \leq V(t, x) \leq k_2 \|x\|^a \quad \forall (t, x) \in [0, \infty) \times \mathbb{D}$
- iii) $\dot{V}(t, x) \leq -k_3 \|x\|^a \quad \forall (t, x) \in [0, \infty) \times \mathbb{D}$

then $x = 0$ is exponentially stable.

4.5.10 Barbalat's Lemma

If there exist functions $V : [0, \infty) \times \mathbb{R}^n \rightarrow \mathbb{R}$, $W_i : \mathbb{R}^n \rightarrow \mathbb{R}$ continuous positive definite and radially unbounded and $W : \mathbb{R}^n \rightarrow \mathbb{R}^n C^1$ and positive semidefinite satisfying

- i) V is C^1
- ii) $W_1(x) \leq V(t, x) \leq W_2(x) \quad \forall (t, x) \in [0, \infty) \times \mathbb{R}^n$
- iii) $\dot{V}(t, x) \leq -W(x) \quad \forall (t, x) \in [0, \infty) \times \mathbb{R}^n$
- iv) for every $k > 0$ there exists $c > 0$ s.t. $\|x\| \leq k \implies |\dot{W}(t, x)| \leq c$ for all $t \geq 0$

then the origin is globally uniformly stable and all solutions approach $E = \{x \in \mathbb{R}^n : W(x) = 0\}$.

4.5.11 LaSalle-Yokishawa Theorem

$\dot{x} = f(t, x)$ where $f : [0, \infty) \times \mathbb{D} \rightarrow \mathbb{R}^n$ is piecewise continuous locally Lipschitz in x and $x = 0$ is an eq.point.

If there exist functions $V : [0, \infty) \times \mathbb{D} \rightarrow \mathbb{R}$, $W_i : \mathbb{D} \rightarrow \mathbb{R}$ continuous positive definite, $W : \mathbb{D} \rightarrow \mathbb{R}$ continuous positive semidefinite, such that

- i) V is C^1
- ii) $W_1(x) \leq V(t, x) \leq W_2(x) \quad \forall (t, x) \in [0, \infty) \times \mathbb{D}$
- iii) $\dot{V}(t, x) \leq -W(x) \quad \forall (t, x) \in [0, \infty) \times \mathbb{D}$
- iv) for every $k > 0$ there exists $r > 0$ s.t. $\|x\| \leq k \implies \|f(t, x)\| \leq r$ for all $t \geq 0$

then the origin is uniformly stable, and $\exists c > 0$ such that all solutions with $\|x(t_0)\| < c$ approach $E = \{x \in \mathbb{D} : W(x) = 0\}$

4.5.12 Global LaSalle-Yokishawa Theorem

$\dot{x} = f(t, x)$ where $f : [0, \infty) \times \mathbb{D} \rightarrow \mathbb{R}^n$ is piecewise continuous locally Lipschitz in x and $x = 0$ is an eq.point.

If there exist functions $V : [0, \infty) \times \mathbb{R}^n \rightarrow \mathbb{R}$, $W_i : \mathbb{R}^n \rightarrow \mathbb{R}$ continuous positive definite and radially unbounded, $W : \mathbb{R}^n \rightarrow \mathbb{R}$ continuous positive semidefinite

- i) V is C^1
- ii) $W_1(x) \leq V(t, x) \leq W_2(x) \quad \forall (t, x) \in [0, \infty) \times \mathbb{R}^n$
- iii) $\dot{V}(t, x) \leq -W(x) \quad \forall (t, x) \in [0, \infty) \times \mathbb{R}^n$
- iv) for every $k > 0$ there exists $r > 0$ s.t. $\|x\| \leq k \implies \|f(t, x)\| \leq r$ for all $t \geq 0$

then the origin is globally uniformly stable and all solutions approach $E = \{x \in \mathbb{R}^n : W(x) = 0\}$.

4.6 Linear Time-Varying Systems and Linearization

4.6.1 Global uniform asymptotic stability (T 4.11)

The equilibrium point $x = 0$ of $\dot{x}(t) = A(t)x$ is (globally) asymptotically stable if and only if the state transition matrix satisfies the inequality

$$\|\Phi(t, t_0)\| \leq ke^{-\lambda(t-t_0)}, \quad \forall t \geq t_0 \geq 0$$

for some positive constant k and λ .

4.7 Converse Theorems

4.7.1 Exponential Stability (T 4.15)

Let $x = 0$ be an equilibrium point for the nonlinear system

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

where $f : [0, \infty) \times D \rightarrow \mathbb{R}^n$ is continuously differentiable, $D = \{x \in \mathbb{R}^n \mid \|x\|_2 < r\}$, and the Jacobian matrix $[\partial f / \partial x]$ is bounded and Lipschitz on D , uniformly in t .

Let

$$A(t) = \frac{\partial f}{\partial x}(t, x)|_{x=0}$$

Then, $x = 0$ is an exponentially stable equilibrium point for the nonlinear system if and only if it is an exponentially stable equilibrium point for the linear system

$$\dot{x} = A(t)x$$

Note that, for linear time-varying systems, uniform asymptotic stability cannot be characterized by the location of the eigenvalue of the matrix $A(t)$. I.e. if $A(t)$ varies with time, we must use a Lyapunov analysis to conclude exponential stability.

5 Input-To-State Stability

5.1 Definition

Consider:

$$\Sigma : \dot{x} = f(t, x, u) \tag{9}$$

The system Σ is ISS if $\exists \beta \in \mathcal{KL}$ and $\exists \gamma \in \mathcal{K}$ such that for any $t_0 \geq 0$, any $x_0 = x(t_0) \in \mathbb{R}^n$ and any bounded input $t \mapsto u(t)$, the solution $t \mapsto x(t)$ exists $\forall t \geq t_0$ and satisfies

$$\|x(t)\| \leq \beta(\|x_0\|, t - t_0) + \gamma(\sup \|u(\tau)\|) \tag{10}$$

5.2 Application

If a system is ISS stable, then:

- i) for $u=0$, the origin is GUAS (0-GUAS)
- ii) for a bounded input $t \mapsto u(t)$, every solution $t \mapsto x(t)$ is bounded.

If one of these is not satisfied, the system is **not** ISS.

5.3 ISS (T 4.19)

Let $V : [0, \infty) \times \mathbb{R}^n \rightarrow \mathbb{R}$ be a continuously differentiable function such that

$$\begin{aligned} \alpha_1(\|x\|) &\leq V(t, x) \leq \alpha_2(\|x\|) \\ \frac{\partial V}{\partial t} + \frac{\partial V}{\partial x} f(t, x, u) &\leq -W_3(x), \quad \forall \|x\| \geq \rho(\|u\|) > 0 \end{aligned}$$

$\forall (t, x, u) \in [0, \infty) \times \mathbb{R}^n \times \mathbb{R}^m$, where α_1, α_2 are class \mathcal{K}_∞ functions, ρ is a class \mathcal{K} function, and $W_3(x)$ is a continuous positive definite function on \mathbb{R}^n . Then, the system (4.44) is input-to-state stable with $\gamma = \alpha_1^{-1} \circ \alpha_2 \circ \rho$.

5.4 ISS (L 4.6)

Suppose $f(t, x, u)$ is continuously differentiable and globally Lipschitz in (x, u) , uniformly in t . If the unforced system (4.45) has a globally exponentially stable equilibrium point at the origin $x = 0$, then the system (4.44) is input-to-state stable.

5.5 ISS with cascade and x_2 as input

Under the stated assumptions, if the system

$$\dot{x}_1 = f_1(t, x_1, x_2) \quad (11)$$

$$\dot{x}_2 = f_2(t, x_2) \quad (12)$$

with x_2 as input, is input-to-state stable and the origin of (12) is globally uniformly asymptotically stable, then the origin of the cascade system (11) and (12) is globally uniformly asymptotically stable.

Example 7 (Example of ISS) *The system is given by*

$$\begin{aligned}\dot{x}_1 &= -x_1 + x_1^2 x_2 \\ \dot{x}_2 &= -x_1^3 - x_2 + u\end{aligned}$$

Let $V(x)$ be given by

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

which is a \mathcal{K}_∞ function. First we will prove that the origin is 0-GUAS, i.e. if the system is GUAS when $u = 0$. Then we will prove that the system is ISS.

$$\begin{aligned}\dot{V}(x) &= x_1 \dot{x}_1 + x_2 \dot{x}_2 \\ &= x_1(-x_1 + x_1^2 x_2) + x_2(-x_1^3 - x_2 + u) \\ &= -x_1^2 - x_2^2\end{aligned}$$

We can further apply the rules of GUAS, and prove that the system is 0-GUAS. Further, we will prove ISS, i.e. when $x \neq 0$. ”

$$\begin{aligned}\dot{V}(x) &= x_1 \dot{x}_1 + x_2 \dot{x}_2 \\ &= x_1(-x_1 + x_1^2 x_2) + x_2(-x_1^3 - x_2 + u) \\ &= -x_1^2 + x_1^3 x_2 - x_1^3 x_2 - x_2^2 + u x_2 \\ &= -x_1^2 - x_2^2 + u x_2 \\ &= -\|x\|_2^2 + u x_2\end{aligned}$$

and upper bounded as

$$\begin{aligned}\dot{V}(x) &\leq -\|x\|_2^2 + |u x_2| \\ &= -\|x\|_2^2 + |u| |x_2| \\ &\leq -\|x\|_2^2 + |u| \|x\|_2 \\ &= -\|x\|_2^2 + |u| \|x\|_2 + \theta \|x\|_2^2 - \theta \|x\|_2^2 \\ &= -(1 - \theta) \|x\|_2^2 + |u| \|x\|_2 - \theta \|x\|_2^2 \\ &= -(1 - \theta) \|x\|_2^2 - (\theta \|x\|_2 - |u|) \|x\|_2 \\ &\leq -(1 - \theta) \|x\|_2^2 \forall \theta \|x\|_2 - |u| \geq 0 \\ &= -(1 - \theta) \|x\|_2^2 \forall \|x\|_2 \geq \frac{|u|}{\theta}\end{aligned}$$

where $\theta \in (0, 1)$. Hence, by Theorem 4.19, the system is input-to-state stable (ISS) with $\rho(|u|) = \frac{|u|}{\theta}$.

5.5.1 Some important things when evaluating ISS

- The θ trick is almost always used
- $\|x\|_2 = \sqrt{x_1^2 + x_2^2}$
- $x_1 \leq \|x\|_2$
- If possible the system can be viewed as cascade.

Example 8 (Example of ISS with cascade) *The system is given by*

$$\dot{x}_1 = -x_1 + x_2^2 \quad (13)$$

$$\dot{x}_2 = -x_2 \quad (14)$$

The system (12) is a linear system and its eigenvalue is negative. The origin of (12) is therefore GES. We now want to show that the system (11) is ISS when x_2 is viewed as input. Let $V = \frac{1}{2}x_1^2$. Then

$$\begin{aligned} \dot{V} &= x_1(-x_1 + x_2^2) = -x_1^2 + x_1x_2^2 \\ &= -(1-\theta)x_1^2 - \theta x_1^2 + x_1x_2^2 \\ &\leq -(1-\theta)x_1^2 - \theta x_1^2 + |x_1|x_2^2 \\ &\leq -(1-\theta)x_1^2 \quad \forall |x_1| \geq \frac{x_2^2}{\theta} \end{aligned}$$

where $0 < \theta < 1$. Since $\gamma(r) = \frac{r^2}{\theta}$ is a class \mathcal{K} function, by Theorem 4.19 the system (5) is ISS with x_2 as input. Hence by Lemma 4.7, the origin of the cascade system (5)-(6) is globally asymptotically stable. (Note: While it is tempting to use Lemma 4.6 to conclude that the system (5) is ISS, the function $f(t, x_1, x_2) = -x_1 + x_2^2$ is not globally Lipschitz in x_2 , and the lemma cannot be applied.)

6 Input-Output Stability

In input-output stability we can only measure the input and output of the system. We cannot measure the state of the system. This is useful when we want to control a system, but we cannot measure the state of the system. We can then use input-output stability to prove that the system is stable. I.e. the system is viewed as a black box that can be accessed only through its input and output terminals.

6.1 \mathcal{L} stability

We consider the input-output relation

$$y = Hu \quad (15)$$

where H is some mapping. u belongs to a space of signals that map the time interval $[0, \infty)$ into the Euclidean space \mathbb{R}^m . The norm $\|u\|$ have to satisfy the same properties as introduced earlier in this text.

- For the space of piecewise continuous signals, bounded functions:

$$\|u\|_{\mathcal{L}_\infty} = \sup_{t \geq 0} \|u(t)\| < \infty \quad \Rightarrow \quad \mathcal{L}_\infty^m \quad (16)$$

- For the space of piecewise continuous signals, square integrable functions:

$$\|u\|_{\mathcal{L}_2} = \sqrt{\int_0^\infty u^T(t)u(t)dt} < \infty \quad \Rightarrow \quad \mathcal{L}_2^m \quad (17)$$

More generally, for the space of piecewise continuous signals, p -integrable functions:

$$\|u\|_{\mathcal{L}_p} = \left(\int_0^\infty \|u(t)\|^p dt \right)^{\frac{1}{p}} < \infty \quad \Rightarrow \quad \mathcal{L}_p^m \quad (18)$$

where \mathbf{m} is the dimension of u , and \mathbf{q} is the dimension of y .

If we think of $u \in \mathcal{L}^m$ as a "well behaved" system, the question to ask is whether the put-put $y \in \mathcal{L}^q$ is also "well behaved".

We cannot define H as a mapping from \mathcal{L}^m to \mathcal{L}^q , because we have to deal with systems which are unstable. H is therefore usually defined as a mapping from an **extended space** \mathcal{L}_e^m to an **extended space** \mathcal{L}_e^q where \mathcal{L}_e^m

$$\mathcal{L}_e^m = \{u | u_\tau \in \mathcal{L}^m, \forall \tau \in [0, \infty)\} \quad (19)$$

and u_τ is defined by

$$u_\tau(t) = \begin{cases} u(t) & \text{if } t \geq \tau \\ 0 & \text{if } t < \tau \end{cases} \quad (20)$$

Example 9 (Example of extended space)

$$\begin{aligned} u(t) = \sin \omega t &\Rightarrow u \in L_{2e} & \text{but } u \notin L_2 \\ y(t) = t &\Rightarrow y \in L_{\infty e} & \text{but } y \notin L_\infty \end{aligned}$$

I.e. we can deal with unbounded signals, such that a unbounded signal can be bounded in an extended space, and a non-integrable function can be integrable in an extended space.

For causal, \mathcal{L} stable systems, it can be shown by a simple argument that

$$u \in \mathcal{L}^m \Rightarrow Hu \in \mathcal{L}^q \quad (21)$$

We also have to consider that an input-output system H is a mapping from m -dimensional \mathcal{L}_e space to q -dimensional \mathcal{L}_e space, i.e.,

$$H : \mathcal{L}_e^m \rightarrow \mathcal{L}_e^q \quad (22)$$

$H : \mathcal{L}_e^m \rightarrow \mathcal{L}_e^q$ is causal if $\forall \tau < \infty, (H(u)_\tau)_\tau = (H(u))_\tau, \forall u \in \mathcal{L}_e^m$. This means that the non-truncated outputs do not have future values, but only past and present. I.e. future values of u do not effect the mapping, so the current value of the output does not depend upon future values of the input.

Further we can introduce that $H : \mathcal{L}_e^m \rightarrow \mathcal{L}_e^q$ is **finite gain \mathcal{L} -stable** if:

- (i) $H : \mathcal{L}_e^m \rightarrow \mathcal{L}_e^q$
- (ii) $\|(Hu)_\tau\| \leq \gamma \|u_\tau\| + \beta$ for some $|\gamma| < \infty$ and $|\beta| < \infty$

where β is called the bias term (zero biased : $\beta = 0$).

To summary we have that $H : \mathcal{L}_{pe}^m \rightarrow \mathcal{L}_{pe}^q$ is \mathcal{L}_p stable iff

- i) $\exists \alpha$ class \mathcal{K} $\alpha : [0, \infty) \rightarrow [0, \infty)$
- ii) \exists constant $\beta \geq 0$

such that

$$\|(Hu)_\tau\|_{\mathcal{L}_p} \leq \alpha(\|u_\tau\|_{\mathcal{L}_p}) + \beta, \quad \forall u \in \mathcal{L}_{pe}^m, \forall \tau \in [0, \infty) \quad (23)$$

Where it is BIBO stability $\equiv \mathcal{L}_\infty$ stability.

Example 10 (Input-Output stability) Consider the following system

$$y = u^{\frac{1}{3}}$$

which is a \mathcal{K}_∞ class function. Further we can assume that $u \in \mathcal{L}_{\infty e}^m$, which implies

$$\sup |u_\tau(t)| < \infty, \quad \forall t \in [0, \infty)$$

We then have

$$\|(Hu)_\tau\|_{\mathcal{L}_\infty} = \sup |u_\tau(t)^{\frac{1}{3}}| = \sup |u_\tau|^{\frac{1}{3}} = (\sup |u_\tau|)^{\frac{1}{3}} = (\|u_\tau\|_{\mathcal{L}_\infty})^{\frac{1}{3}}$$

we can now choose a $\alpha = r^{\frac{1}{3}}$ which is a class \mathcal{K} function. We then have

$$\|(Hu)_\tau\|_{\mathcal{L}_\infty} = \alpha(\|u_\tau\|_{\mathcal{L}_\infty})$$

which satisfies the definition of \mathcal{L}_∞ stability. We also have zero bias here where $\beta = 0$. We can further analyse the finite gain stability. If we take a closer look at the theorem we see that

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

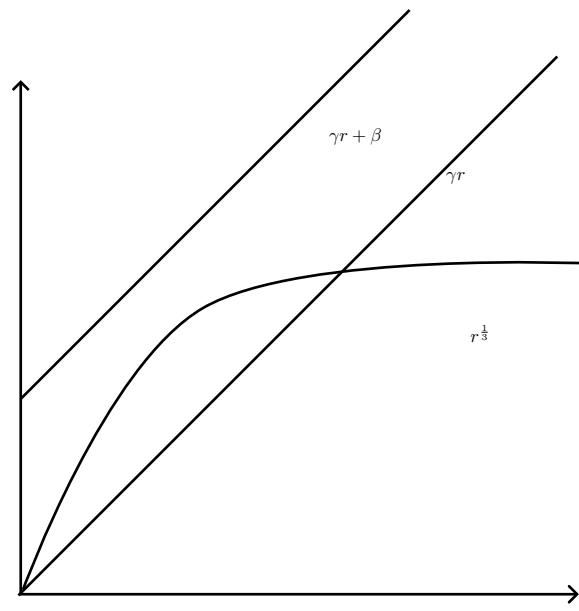


Figure 2: Example 10

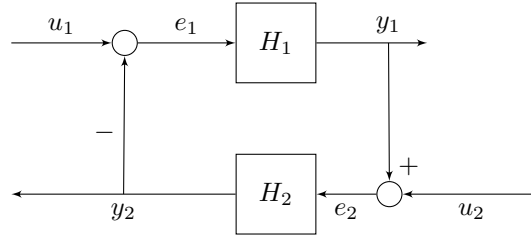
I.e. we must find a β such that this statement holds (see figure 2).

$$r^{\frac{1}{3}} = \gamma r \Rightarrow r = \gamma^{-\frac{2}{3}} \Rightarrow \gamma r = \gamma^{-\frac{1}{2}}$$

which leads to

$$\|(Hu)_\tau\|_{\mathcal{L}_\infty} \leq \gamma \|u_\tau\|_{\mathcal{L}_\infty} + \frac{1}{\sqrt{\gamma}}$$

6.2 The small-Gain Theorem



The interconnection where H_1 and H_2 are finite-gain \mathcal{L} -stable, i.e

$$\begin{aligned} \|y_{1\tau}\|_{\mathcal{L}} &\leq \gamma_1 \|e_{1\tau}\|_{\mathcal{L}} + \beta_1 \\ \|y_{2\tau}\|_{\mathcal{L}} &\leq \gamma_2 \|e_{2\tau}\|_{\mathcal{L}} + \beta_2 \end{aligned}$$

is finite-gain \mathcal{L} -stable if

$$\gamma_1 \gamma_2 < 1 \tag{24}$$

7 Passivity

Passivity is a tool for analysis of nonlinear systems, with an input-output description. It also has an interesting energy interpretation. It also relates nicely to Lyapunov stability and \mathcal{L}_2 stability.

7.1 Memoryless functions

In a simple sense passivity for a memoryless function $y = h(t, u)$ where $h : [0, \infty) \times \mathbb{R}^p \rightarrow \mathbb{R}^p$ is passive if $u^T y \geq 0$ for the sector $h \in [0, \infty]$. The sector $h \in [0, \infty]$ is i.e. the space between two lines with a slope of 0 and ∞ .

I.e. is this sector the first and third quadrant of the $u - y$ plane. For vector cases we can decouple $h(t, u)$, such that $h_i(t, u)$ only depends on u_i

$$h(t, u) = \begin{bmatrix} h_1(t, u_1) \\ h_2(t, u_2) \\ \vdots \\ h_p(t, u_p) \end{bmatrix}$$



Figure 3: Picture of the sector $h \in [0, \infty]$

A extreme case of passivity happens when $u^T y = 0$. In this case, we say that the system is **lossless**.

Futher consider a function h satisfying $u^T y \geq u^T \phi(u)$ for some function $\phi(u)$. When $u^T \phi(u) > 0$ for all $u \neq 0$, h called **input strictly passive**. The term $u^T \phi(u)$ represents the "excess" of passivity, and if $u^T \phi(u)$ is negative for some value u , then the function h is not necessarily passive. The term $u^T \phi(u)$ represents the "shortage" of passivity. If we have a scalar example where $\phi(u) = \epsilon u$, then h belongs to the sector $[\epsilon, \infty]$. If $\epsilon > 0$ we have excess passivity, and when $\epsilon < 0$ we have shortage of passivity. Excess or shortage of passivity can be removed by the input-output feedforward operation. With the new output defined as

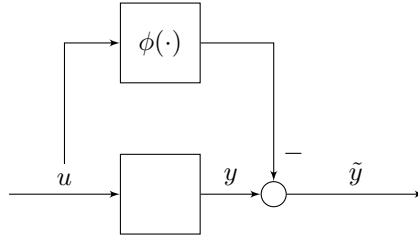


Figure 4: Removal of excess or shortage of passivity by input-feedforward operation

new output defined as $\tilde{y} = y - \phi(u)$

$$u^T \tilde{y} = u^T y - u^T \phi(u) = 0 \quad (25)$$

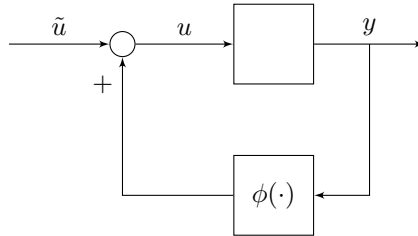


Figure 5: Removal of excess or shortage of passivity by output-feedback operation

We can also remove excess or shortage of passivity by output-feedback operation. With the new input defined as $\tilde{u} = u + \phi(y)$

$$\tilde{u}^T y = [u^T y + \rho(y)]^T y \geq y^T \rho(y) - y^T \rho(y) = 0 \quad (26)$$

To summarize we have the following

- passive if $u^T y \geq 0$
- lossless if $u^T y = 0$

- input-feedforward passive if $u^T y \geq u^T \phi(u)$ for some function $\phi(u)$
- input strictly passive if $u^T y \geq u^T \phi(u)$ for some function $\phi(u)$ and $u^T \phi(u) > 0$ for all $u \neq 0$
- output-feedback passive if $u^T y \geq u^T \rho(y)$ for some function $\rho(y)$
- output strictly passive if $u^T y \geq u^T \rho(y)$ for some function $\rho(y)$ and $u^T \rho(y) > 0$ for all $y \neq 0$

7.2 State Models

Consider the system

$$\begin{aligned}\dot{x} &= f(t, x, u) \\ y &= h(t, x, u)\end{aligned}\tag{27}$$

where $f : \mathbb{R}^n \times \mathbb{R}^p \rightarrow \mathbb{R}^n$ is Locally Lipschitz, $h : \mathbb{R}^n \times \mathbb{R}^p \rightarrow \mathbb{R}^p$ is continuous $f(0, 0) = 0, h(0, 0) = 0$

7.2.1 Passivity for dynamical systems (D 6.3)

The system (27) is said to be passive if there exists a continuously differentiable positive function $V(x)$ (called the storage function) such that

$$u^T y \geq \dot{V} = \frac{\partial V}{\partial x} f(x, u), \quad \forall (x, u) \in \mathbb{R}^n \times \mathbb{R}^p\tag{28}$$

- lossless if $u^T y = \dot{V}$.
- input-feedforward passive if $u^T y \geq \dot{V} + u^T \varphi(u)$ for some function φ .
- input strictly passive if $u^T y \geq \dot{V} + u^T \varphi(u)$ and $u^T \varphi(u) > 0, \forall u \neq 0$
- output-feedback passive if $u^T y \geq \dot{V} + y^T \rho(y)$ for some function ρ .
- output strictly passive if $u^T y \geq \dot{V} + y^T \rho(y)$ and $y^T \rho(y) > 0, \forall y \neq 0$
- strictly passive if $u^T y \geq \dot{V} + \psi(x)$ for some positive definite function ψ

In all cases, the inequality should hold for all (x, u) .

7.3 \mathcal{L}_2 and Lyapunov stable

7.3.1 Finite Gain \mathcal{L}_2 stability (L 6.5)

If the system (27) is strictly passive with $\rho(y) = \delta y, \delta > 0$ then the system is finite gain \mathcal{L}_2 stable, with \mathcal{L}_2 gain $\delta \leq \frac{1}{\delta}$

Proof 2 (Proof of finite-gain \mathcal{L}_2 stability) Assume output strict passivity with $\rho(y) = \delta y$

$$\begin{aligned}u^T y &\leq \dot{V} + \delta y^T y \\ \dot{V} &\leq u^T y - \delta y^T y \\ &\leq \frac{\delta}{2} y^T y + \frac{1}{2\delta} u^T u - \delta y^T y \quad \text{Young's inequality} \\ &= \frac{1}{2\delta} u^T u - \frac{\delta}{2} y^T y \\ y^T y &\leq \frac{1}{\delta^2} u^T u - \frac{2}{\delta} \dot{V}\end{aligned}\tag{29}$$

We then integrate over $[0, \tau]$ with respect to time

$$\int_0^\tau y(\tau)^T y(\tau) d\tau \leq \frac{1}{\delta^2} \int_0^\tau u(\tau)^T u(\tau) d\tau - \frac{2}{\delta} V(x(\tau)) + \frac{2}{\delta} V(x(0))\tag{30}$$

this states that $\|y_\tau\|_{\mathcal{L}_2}^2 \leq \frac{1}{\delta^2} \|u_\tau\|_{\mathcal{L}_2}^2 + \frac{2}{\delta} V(x(0))$ which can be formed on the form of finite gain stability

$$\|y_\tau\|_{\mathcal{L}_2} \leq \frac{1}{\delta} \|u\|_{\mathcal{L}_2} + \sqrt{\frac{2}{\delta} V(x(0))}$$

7.3.2 Lyapunov stable (0-stable) (L 6.6)

If the system (27) is passive with a positive storage function $V(x)$, then the origin is Lyapunov stable (0-stable), $f(x, 0)$.

Proof 3 (Positive definite storage function) Consider $\exists V$ s.t.

i) V is C^1

ii) V is positive definite

iii) $u^T y \geq \dot{V}$

if $u = 0 \rightarrow 0 \geq \dot{V}$ (Standard Lyapunov theory).

7.4 Asymptotically stable (0-AS) (L 6.7)

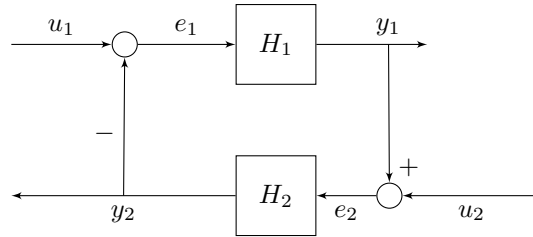
Lemma 6.7 Consider the system (27). The origin of $\dot{x} = f(x, 0)$ is asymptotically stable if the system is

- strictly passive or
- output strictly passive and zero-state observable.

Furthermore, if the storage function is radially unbounded, the origin will be globally asymptotically stable.

(27) is **zero-state observable** iff no solution $\dot{x} = f(x, 0)$ can stay identically $S0\{x \in \mathbb{R}^n | h(x, 0) = 0\}$ other than the trivial solution $x(t) = 0$.

7.5 Feedback Systems: Passivity Theorems (T 6.1)



where

$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}, \quad u = \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}, \quad y = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} \quad (31)$$

and

$$\begin{aligned} e_1 &= u_1 - h_2(x_2, e_2) \\ e_2 &= u_2 + h_1(x_1, e_1) \end{aligned} \quad (32)$$

(The feedback connection of two passive systems is passive.)

Proof 4 (Proof of the passivity of feedback connection) Let $V_1(x_1)$ and $V_2(x_2)$ be the storage functions for H_1 and H_2 , respectively. If either component is a memoryless function, take $V_i = 0$. Then,

$$e_i^T y_i \geq \dot{V}_i$$

From the feedback connection of Figure 6.11, we see that

$$e_1^T y_1 + e_2^T y_2 = (u_1 - y_2)^T y_1 + (u_2 + y_1)^T y_2 = u_1^T y_1 + u_2^T y_2$$

Hence,

$$u^T y = u_1^T y_1 + u_2^T y_2 \geq \dot{V}_1 + \dot{V}_2$$

Taking $V(x) = V_1(x_1) + V_2(x_2)$ as the storage function for the feedback connection, we obtain

$$u^T y \geq \dot{V}$$

Using Theorem 6.1 and the results of the previous section on stability properties of passive systems, we can arrive at some straightforward conclusions on stability of the feedback connection. We start with \mathcal{L}_2 stability.

7.5.1 Feedback Systems: Finitegain \mathcal{L}_2 stable (T 6.2)

Theorem 6.2 Consider the feedback connection of Figure 6.11 and suppose each feedback component satisfies the inequality

$$e_i^T y_i \geq \dot{V}_i + \varepsilon_i e_i^T e_i + \varepsilon_i y_i^T y_i, \text{ for } i = 1, 2$$

for some storage function $V_i(x_i)$. Then, the closed-loop map from u to y is finite gain \mathcal{L}_2 stable if

$$\varepsilon_1 + \delta_2 > 0 \text{ and } \varepsilon_2 + \delta_1 > 0$$

7.5.2 Asymptotically stability of feedback systems (T 6.3)

Theorem 6.3 Consider the feedback connection of two time-invariant dynamical systems. The origin of the closed-loop system (when $u = 0$) is asymptotically stable if

- both feedback components are strictly passive,
- both feedback components are output strictly passive and zero-state observable, or
- one component is strictly passive and the other one is output strictly passive and zero-state observable.

Furthermore, if the storage function for each component is radially unbounded, the origin is globally asymptotically stable.