

Ch7: Stability for Non-Autonomous Systems

Concepts of Stability

Autonomous vs. Non-Autonomous Systems

The fundamental difference between autonomous and non-autonomous systems lies in the fact that the **state trajectory of an autonomous system is independent of the initial time t_0** , while that of a non-autonomous system generally is **not**.

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

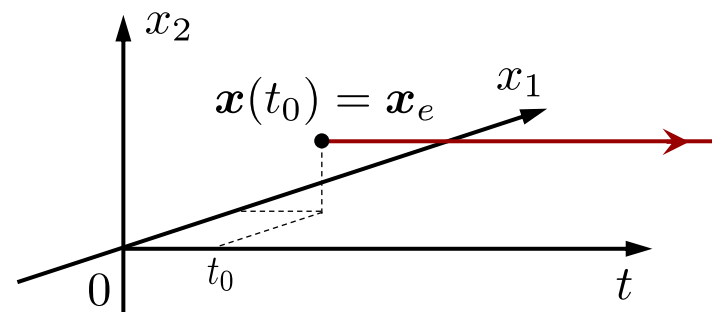
This difference requires us to **consider the initial time t_0 explicitly** in defining stability concepts for non-autonomous systems and makes the analysis more difficult than that of autonomous systems.

Non-autonomous systems appear in robot control when the desired task is to follow a time-varying trajectory, i.e. in **motion control**, or when there is uncertainty in the physical parameters and therefore, an **adaptive control** approach may be used.

Equilibrium Point

A state \mathbf{x}_e is an **Equilibrium Point** (or **Equilibrium State**) if the system starts there (initial state $\mathbf{x}(t_0) = \mathbf{x}_e$) it will remain there for all future time.

$$\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}_e, t) = \mathbf{0} \quad \forall t \geq t_0$$



For example, the system $\dot{x} = -\frac{a(t)x}{1+x^2}$ has an equilibrium point at $x = 0$.

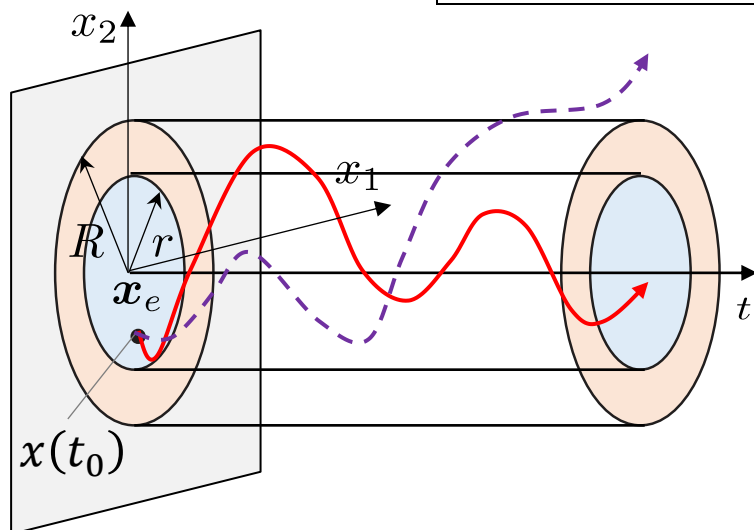
However, the system $\dot{x} = -\frac{a(t)x}{1+x^2} + b(x)$, $b(x) \neq 0$ does not have an equilibrium point.

Extensions of Stability Concepts

The concepts of stability for non-autonomous systems are quite similar to those of autonomous systems. However, the definitions include the **initial time** t_0 explicitly.

The equilibrium point x_e is said to be **Stable** at t_0 if for any $R > 0$, there exists $r = r(R, t_0) > 0$, such that if $\|x(t_0) - x_e\| < r$, then $\|x(t) - x_e\| < R$ for all $t \geq t_0$. Otherwise, the equilibrium point is **Unstable**.

$$\forall R > 0, \exists r > 0 : \|x(t_0) - x_e\| < r \Rightarrow \|x(t) - x_e\| < R, \forall t \geq t_0$$



(we can keep the state in a ball of arbitrarily small radius R by starting the state trajectory in a ball of sufficiently small radius r)

The equilibrium point x_e is said to be **Uniformly Stable**, if r can be chosen independently of the initial time t_0 .

Extensions of Stability Concepts (cont.)

The equilibrium point x_e is said to be **Asymptotically Stable** at t_0 if (1) it is **Lyapunov Stable**, and (2) there exists $r=r(t_0) > 0$ such that if $\|x(t_0) - x_e\| < r$, then $\|x(t) - x_e\| \rightarrow 0$ as $t \rightarrow \infty$.

$$\boxed{\exists r > 0 : \|x(t_0) - x_e\| < r \Rightarrow \|x(t) - x_e\| \rightarrow 0, \text{ as } t \rightarrow \infty}$$

The equilibrium point x_e is said to be **Uniformly Asymptotically Stable**, if it is **Uniformly Stable** (i.e., r can be chosen independently of the initial time t_0) where

$$\boxed{\exists r > 0 : \|x(t_0) - x_e\| < r \Rightarrow \|x(t) - x_e\| \rightarrow 0, \text{ as } t \rightarrow \infty}$$

Example: $\dot{x} = -\frac{x}{(1+t)} \longrightarrow x(t) = \frac{1+t_0}{1+t}x(t_0)$

The origin is asymptotically stable but not uniformly asymptotically stable, because a larger t_0 requires a longer time to get close to the origin.

★ Non-autonomous systems with uniform properties have some desirable ability to withstand disturbances.

Extensions of Stability Concepts (cont.)

The equilibrium point x_e is said to be **Exponentially Stable** if there exist $\alpha, \lambda, r > 0$ such that if $\|x(t_0) - x_e\| < r$, then $\|x(t) - x_e\| < \alpha \|x(t_0) - x_e\| e^{-\lambda(t-t_0)} \quad \forall t \geq t_0$.

$$\boxed{\exists \alpha, \lambda, r > 0 : \|x(t_0) - x_e\| < r \Rightarrow \|x(t) - x_e\| \leq \alpha \|x(t_0) - x_e\| e^{-\lambda(t-t_0)}}$$

If asymptotic (or exponential) stability holds for **any initial states** $x(t_0) \in \mathbb{R}^n$, the equilibrium point is said to be **Globally Asymptotically (or Exponentially) Stable**.

★ It can be shown that **exponential stability** always implies **uniform asymptotic stability**.

Example: A First-Order Linear Time-varying System

Consider the first-order system $\dot{x}(t) = -a(t)x(t)$

Its solution is $x(t) = x(t_0)e^{-\int_{t_0}^t a(r)dr}$

The system is stable if $a(t) \geq 0, \forall t \geq t_0$. It is asymptotically stable if $\int_0^\infty a(r)dr = +\infty$.

For Example:

$\dot{x} = -\frac{x}{(1+t)^2}$: The origin is stable (but not asymptotically stable), because $\int_0^\infty \frac{1}{(1+r)^2} dr = 1$.

$\dot{x} = -\frac{x}{1+t}$: The origin is asymptotically stable, because $\int_0^\infty \frac{1}{1+r} dr = +\infty$.

$\dot{x} = -tx$: The origin is exponentially stable, because $x = c_1 e^{-t^2/2}$.

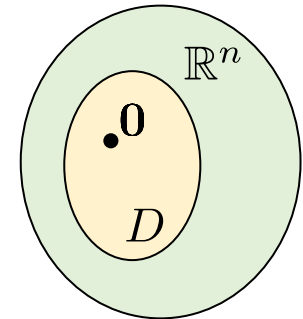
Lyapunov Analysis

Time-Varying Positive Definite Functions

A scalar, time-varying function $V(\mathbf{x}, t)$ ($V: D \times \mathbb{R}_+ \rightarrow \mathbb{R}$, $D \subset \mathbb{R}^n$, $\mathbf{0} \in D$) is said to be **Locally Positive Definite** if

- 1) $V(\mathbf{0}, t) = 0 \quad \forall t \geq t_0$
- 2) $V(\mathbf{x}, t) \geq V_0(\mathbf{x}) \quad \forall t \geq t_0, \forall \mathbf{x} \in D$

where $V_0(\mathbf{x})$ ($V_0: D \rightarrow \mathbb{R}$) is a **time-invariant positive definite** function.



$V(\mathbf{x}, t)$ is said to be **Globally Positive Definite** if $D = \mathbb{R}^n$.

\Rightarrow A scalar time-variant function $V(\mathbf{x}, t)$ is positive definite if it dominates a time-invariant positive definite function.

- A function $V(\mathbf{x}, t)$ is **positive semi-definite** if $V_0(\mathbf{x})$ is positive semi-definite.
- A function $V(\mathbf{x}, t)$ is **negative (semi-)definite** if $-V(\mathbf{x}, t)$ is positive (semi-)definite.

Decrescent Function

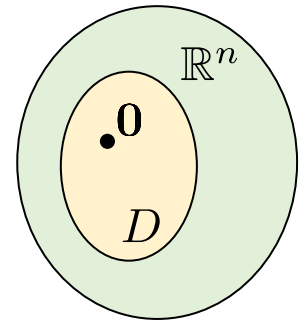
A scalar function $V(\mathbf{x}, t)$ ($V: D \times \mathbb{R}_+ \rightarrow \mathbb{R}$, $D \subset \mathbb{R}^n$, $\mathbf{0} \in D$) is said to be **Locally Decrescent** if

- 1) $V(\mathbf{0}, t) = 0 \quad \forall t \geq t_0$
- 2) $V(\mathbf{x}, t) \leq V_1(\mathbf{x}) \quad \forall t \geq t_0, \forall \mathbf{x} \in D$

where $V_1(\mathbf{x})$ ($V_1: D \rightarrow \mathbb{R}$) is a **time-invariant positive definite** function.

$V(\mathbf{x}, t)$ is said to be **(Globally) Decrescent** if $D = \mathbb{R}^n$.

\Rightarrow A scalar time-variant function $V(\mathbf{x}, t)$ is decrescent if it is dominated by a time-invariant positive definite function.



Example: $V(\mathbf{x}, t) = (1 + \sin^2 t) (x_1^2 + x_2^2)$

$$V_0(\mathbf{x}) = x_1^2 + x_2^2 \quad V_1(\mathbf{x}) = 2 (x_1^2 + x_2^2)$$

The function is positive definite and decrescent.

Lyapunov's Direct Method for Non-Autonomous Systems

Consider a non-autonomous system, $\dot{x} = f(x, t)$, with an equilibrium point at origin, $x = 0$. If there exists a scalar function $V(x, t)$ ($V: D \times \mathbb{R}_+ \rightarrow \mathbb{R}$, $D \subset \mathbb{R}^n$, $0 \in D$) with continuous partial derivatives such that

1) $V(x, t)$ is **positive definite** (locally in D),

2) $\dot{V}(x, t)$ is **negative semi-definite** (locally in D),

the equilibrium point 0 is **Stable** (and V is called a Lyapunov function).

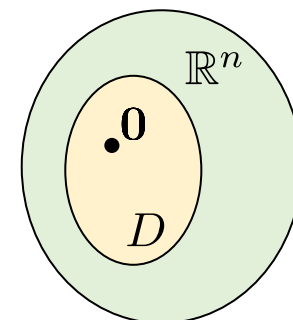
3) $V(x, t)$ is **decreascent** (locally in D),

the equilibrium point 0 is **Uniformly Stable**. If $\dot{V}(x, t)$ is **negative definite** (locally in D), the equilibrium point 0 is **Uniformly Asymptotically Stable**.

4) $D = \mathbb{R}^n$,

5) $V(x, t)$ is **radially unbounded**, i.e., $V(x, t) \rightarrow \infty$ as $\|x\| \rightarrow \infty$.

the equilibrium point 0 is **Globally Uniformly (Asymptotically) Stable**



Note:

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

Example

Example: Determine the stability of the equilibrium point at **0**.

$$\begin{aligned}\dot{x}_1 &= -x_1 - e^{-2t}x_2 \\ \dot{x}_2 &= x_1 - x_2\end{aligned}$$

Let's choose this scalar function:

$$V(\mathbf{x}, t) = x_1^2 + (1 + e^{-2t}) x_2^2$$

$$x_1^2 + x_2^2 \leq V(\mathbf{x}, t) \leq x_1^2 + 2x_2^2 \quad \therefore \text{The function is positive definite and decrescent.}$$

$$\dot{V}(\mathbf{x}, t) = -2 [x_1^2 - x_1x_2 + x_2^2 (1 + 2e^{-2t})]$$

$$\dot{V} \leq -2 (x_1^2 - x_1x_2 + x_2^2) = -(x_1 - x_2)^2 - x_1^2 - x_2^2 \quad \therefore \dot{V} \text{ is negative definite.}$$

$V(\mathbf{x}, t)$ is radially unbounded, i.e., $V(\mathbf{x}, t) \rightarrow \infty$ as $\|\mathbf{x}\| \rightarrow \infty$.

\therefore The point **0 is globally uniformly asymptotically stable.**

Stability of Linear Time-Varying Systems

Consider linear time-varying (LTV) systems of the form $\dot{\mathbf{x}} = \mathbf{A}(t)\mathbf{x}$.

LTI systems are asymptotically stable if their eigenvalues all have negative real parts. However, none of the standard approaches for analyzing LTI systems applies to LTV systems.

Example:
$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} -1 & e^{2t} \\ 0 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \Rightarrow \lambda_{1,2} = -1, \quad \forall t \geq 0$$

However, the system is unstable

$$\begin{aligned} x_2 &= x_2(0)e^{-t} \\ \dot{x}_1 + x_1 &= x_2(0)e^t \end{aligned}$$

Stability of Linear Time-Varying Systems

The LTV system $\dot{x} = \mathbf{A}(t)x$ is **asymptotically stable** if the eigenvalues of the symmetric matrix $\mathbf{A}(t) + \mathbf{A}^T(t)$ (all of which are real) remain strictly in the left-half complex plane:

$$\exists \lambda > 0, \quad \forall i, \quad \forall t \geq 0, \quad \lambda_i(\mathbf{A}(t) + \mathbf{A}^T(t)) \leq -\lambda$$

- Note that the result provides a **sufficient** condition for asymptotic stability.

Lyapunov-Like Analysis

Barbalat's Lemma

For autonomous systems, the invariant set theorems are powerful tools to study stability, because they allow asymptotic stability conclusions to be drawn even when \dot{V} is only negative semi-definite. However, the invariant set theorems are not applicable to non-autonomous systems. Instead, Barbalat's lemma can be used for non-autonomous systems.

Barbalat's Lemma:

If the differentiable function $f(t)$ has a finite limit as $t \rightarrow \infty$, and if \dot{f} is uniformly continuous, then $\dot{f}(t) \rightarrow 0$ as $t \rightarrow \infty$.



A sufficient condition for a differentiable function to be uniformly continuous is that its derivative be bounded.



\Rightarrow If the differentiable function $f(t)$ has a finite limit as $t \rightarrow \infty$, and is such that \ddot{f} exists and is bounded, then $\dot{f} \rightarrow 0$ as $t \rightarrow \infty$.

Lyapunov-Like Stability Analysis Using Barbalat's Lemma

If a scalar function $V(\mathbf{x}, t)$ satisfies the following conditions

- $V(\mathbf{x}, t)$ is lower bounded,
 - $\dot{V}(\mathbf{x}, t)$ is negative semi-definite,
 - $\dot{V}(\mathbf{x}, t)$ is uniformly continuous in time (i.e., $\ddot{V}(\mathbf{x}, t)$ is bounded),
- then $\dot{V}(\mathbf{x}, t) \rightarrow 0$ as $t \rightarrow \infty$.

Therefore, V approaches a finite limiting value V_∞ , such that $V_\infty \leq V(\mathbf{x}(t_0), 0)$.

Example

The closed-loop error dynamics of an adaptive control system for a first-order plant with one unknown parameter is

$$\begin{aligned}\dot{e} &= -e + \theta w(t) \\ \dot{\theta} &= -ew(t)\end{aligned}$$

where e and θ are the two states of the closed-loop dynamics, representing tracking error and parameter error, and $w(t)$ is a bounded continuous function.

Consider Lyapunov function $V = e^2 + \theta^2$. The time derivative is

$$\dot{V} = 2e(-e + \theta w) + 2\theta(-ew) = -2e^2 \leq 0$$

Based on Lyapunov theory, the system is stable, and therefore, e and θ are bounded.

Example (cont.)

To use Barbalat's lemma, we must check the uniform continuity of \dot{V} .

$$\ddot{V} = -4e(-e + \theta w)$$

The derivative of \dot{V} (i.e., \ddot{V}) is bounded, since w is bounded by hypothesis, and e and θ were shown to be bounded. Hence, \dot{V} is uniformly continuous, and application of Barbalat's lemma indicates that $e \rightarrow 0$ as $t \rightarrow \infty$ (since $\dot{V}(x, t) \rightarrow 0$ as $t \rightarrow \infty$).

Note: Although e converges to zero, the system is not asymptotically stable, because θ is only guaranteed to be bounded.

Simulation with

$$w(t) = 1/(1 + t),$$

$$e(0) = \theta(0) = 0.1$$

