

Stability Analysis

- Introduction
- Stability analysis for LTI systems
- Floquet theory
- Lyapunov theory

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Stability

The linear differential equation

$$\dot{x}(t) = A(t)x(t) \quad \text{with} \quad x(0) = x_0$$

is called stable, if there exists for every $\epsilon > 0$ a $\delta > 0$ such that for every $x_0 \in \mathbb{R}^{n_x}$ with $\|x_0\| \leq \delta$ the function $x(t)$ satisfies $\|x(t)\| \leq \epsilon$ for all $t \geq 0$.

Asymptotic Stability

The linear differential equation

$$\dot{x}(t) = A(t)x(t) \quad \text{with} \quad x(0) = x_0$$

is called asymptotically stable, if it is stable and additionally satisfies

$$\lim_{t \rightarrow \infty} x(t) = 0 .$$

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Stability of LTI systems

LTI system $\dot{x}(t) = Ax(t)$ is stable if and only if

- all eigenvalues of the matrix A have non-positive real part and
- all purely imaginary eigenvalues have algebraic multiplicity 1.

Stability of LTI systems

Proof:

- **Step 1:** Write A in Jordan normal form:

$$A = T \text{diag}(J_1, \dots, J_{n_j}) T^{-1}$$

such that

$$x(t) = e^{At} x_0 = T \text{diag} \left(e^{J_1 t}, \dots, e^{J_{n_j} t} \right) T^{-1} x_0 .$$

Stability of LTI systems

Proof:

- **Step 2:** The block matrices

$$e^{J_i t} = e^{\lambda_i t} \begin{pmatrix} 1 & t & \dots & \frac{t^{m_i-1}}{(m_i-1)!} \\ 0 & 1 & \ddots & \vdots \\ \vdots & \ddots & 1 & t \\ 0 & \dots & 0 & 1 \end{pmatrix}$$

are uniformly bounded for all $t \geq 0$ if either λ_i has a strictly negative real part or if λ_i is purely imaginary and $m_i = 1$.

Stability of LTI systems

Proof:

- **Step 3:** Stability follows then from the estimate

$$\|x(t)\| \leq \|e^{At}\| \|x_0\| \leq \epsilon$$

for all x_0 with $\|x_0\| \leq \delta$ and $\delta = \frac{\epsilon}{\max_{t \geq 0} \|e^{At}\|}$, since the maximum exists.

Stability of LTI systems

Proof:

- **Step 4 (other direction):** if there exists an imaginary eigenvalue with algebraic multiplicity larger than 1 or an eigenvalue with strictly positive real part, we can find $0 \neq c \in \mathbb{R}^{n_x}$ with

$$\lim_{t \rightarrow \infty} \|e^{At}c\| \rightarrow \infty . \quad (\text{why?})$$

This implies that the system is unstable.

Asymptotic Stability of LTI systems

A linear time invariant system is asymptotically stable if and only if all eigenvalues of A have strictly negative real part.

Proof: Exercise.

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Stability analysis for periodic systems

Let A be periodic, $A(t + T) = A(t)$. The linear time varying system

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

is stable if and only if

- the eigenvalues of the associated monodromy matrix $G(T, 0)$ are contained in the closed unit disk and
- all eigenvalues that are on the unit circle have algebraic multiplicity 1.
- The eigenvalues of $G(T, 0)$ are also called “Floquet multipliers”.

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Stability analysis for periodic systems

Proof (sketch): Introduce the Jordan normal form

$$G(T, 0) = T \operatorname{diag} (J_1, \dots, J_{n_j}) T^{-1}$$

Since we have

$$x(kT) = G(T, 0)^k x_0$$

for all integer $k \in \mathbb{N}$, we are interested in analyzing the k -th power of the monodromy matrix, which can be written as

$$G(T, 0)^k = T \operatorname{diag} \left(J_1^k, \dots, J_{n_j}^k \right) T^{-1} .$$

Stability analysis for periodic systems

Proof (sketch):

The k -th power of the i -th Jordan block can be worked out explicitly

$$J_i^k = \begin{pmatrix} \lambda_i^k & k\lambda_i^{k-1} & \cdots & \frac{k!\lambda_i^{k-m_i+1}}{(k-m_i+1)!} \\ 0 & \lambda_i^k & \ddots & \vdots \\ \vdots & \ddots & \ddots & k\lambda_i^{k-1} \\ 0 & \cdots & 0 & \lambda_i^k \end{pmatrix}.$$

(from here the proof is straightforward)

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- **Lyapunov theory**

Lyapunov function

Idea: Instead of analyzing A directly, introduce a Lyapunov function

$V : \mathbb{R}^{n_x} \rightarrow \mathbb{R}$ that decreasing along the trajectories of the system.

- V is assumed to be differentiable; Notation:

$$\dot{V}(x(t)) = \nabla_x V(x(t))^\top \dot{x}(t)$$

- $x(t)$ denotes the solution of a differential equation
- $\nabla_x V$ denotes the gradient of V

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Lyapunov theory

Main idea of Lyapunov theory: formulate conditions on the functions V and $\dot{V}(x(t))$ which imply desired stability or boundedness properties of the state trajectory x .

Advantages:

- the concepts can be applied to nonlinear systems
- Lyapunov functions are closely related to invariant sets

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Advantages:

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- Lyapunov functions are closely related to invariant sets

Examples for conditions on V

- *Positive definiteness.* V is called positive definite, if $V(x) \geq 0$ for all $x \in \mathbb{R}$ but $V(x) = 0$ if and only if $x = 0$.
- *Monotonicity.* V is called monotonically decreasing, if

$$\dot{V}(x(t)) \leq 0$$

for all initial values x_0 .

- *Strict monotonicity.* V is called strictly monotonically decreasing, if

$$\dot{V}(x(t)) < 0$$

for $x(t) \neq 0$ and all initial values x_0 .

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Examples for conditions on V

- *Unboundedness.* V is called unbounded, if $\lim_{\|x\| \rightarrow \infty} V(x) = \infty$.

Equivalent: all sublevel sets of V are bounded.

- *Positive quadratic.* V is called positive quadratic, if there exists

$P \in \mathbb{S}_{++}^{n_x}$ such that $V(x) = x^T P x$.

- *Exponential contractivity.* V is called exponentially contractive, if there exists a $\alpha > 0$ such that $\dot{V}(x(t)) \leq -\alpha V(x(t))$.

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Examples for Lyapunov statements

- If V is monotonically decreasing, then all sublevel sets of the form

$$S_\alpha = \{s \in \mathbb{R}^{n_x} \mid V(s) \leq \alpha\}$$

are for all $\alpha \in \mathbb{R}$ invariant sets of the differential equation for x .

Proof: Use the inequality

$$V(x(t)) = V(x(0)) + \underbrace{\int_0^t \dot{V}(x(\tau)) \, d\tau}_{\leq 0} \leq V(x(0)) .$$

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Examples for Lyapunov statements

- If the function V is positive definite, unbounded, and monotonically decreasing, then $x(t)$ is bounded.

Proof: Use that the invariant sublevel set

$$\{s \in \mathbb{R}^{n_x} \mid V(s) \leq V(x(0))\}$$

is bounded; i.e., $x(t)$ is uniformly bounded for all $t \in \mathbb{R}$.

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Examples for Lyapunov statements

- If V is positive definite, unbounded, and strictly monotonically decreasing, then $x(t)$ is bounded and converges to zero for $t \rightarrow \infty$, $\lim_{t \rightarrow \infty} x(t) = 0$.

Proof: Boundedness is already established. Since $V(x(t))$ is strictly monotonically decreasing and bounded, $V_\infty = \lim_{t \rightarrow \infty} V(x(t))$ must exist. If we would have $V_\infty > 0$, then

$$\lim_{t \rightarrow \infty} V(x(t)) = V(x(0)) + \lim_{t \rightarrow \infty} \int_0^t \dot{V}(x(\tau)) d\tau = -\infty,$$

which is a contradiction to the positive definiteness of V . Thus, $\lim_{t \rightarrow \infty} V(x(t)) = 0$, which implies $\lim_{t \rightarrow \infty} x(t) = 0$.

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$\lim_{t \rightarrow \infty} V(x(t)) = 0$, which implies $\lim_{t \rightarrow \infty} x(t) = 0$.

Examples for Lyapunov statements

- If V is positive quadratic, $V(x) = x^\top P x$ with $\lambda_{\min}(P) > 0$, and exponentially contractive with dissipation rate $\alpha > 0$, $\dot{V} \leq -\alpha V$, then x is exponentially stable. I.e., there exists a constant

$$\|x(t)\|_2 \leq \sqrt{\frac{V(x(0))}{\lambda_{\min}(P)}} e^{-\alpha t/2} \leq \sqrt{\frac{\lambda_{\max}(P)}{\lambda_{\min}(P)}} \|x(0)\| e^{-\alpha t/2}$$

for all $t \geq 0$.

Proof: Use the estimate

$$\lambda_{\min}(P) \|x(t)\|_2^2 \leq V(x(t)) \leq V(x(0)) e^{-\alpha t}.$$

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Lyapunov theory for LTI systems

The LTI system $\dot{x}(t) = Ax(t)$ is stable if and only if there exists a positive definite quadratic Lyapunov function $V(x) = x^T Px$, which proves it, i.e.,

$$\exists P \succ 0 : \quad P = P^T \quad \text{and} \quad A^T P + PA \preceq 0 .$$

Proof: If we can find a positive definite P with $A^T P + PA \preceq 0$, we have

$$\dot{V} = \dot{x}^T Px + x^T P \dot{x} = x^T (A^T P + PA)x \leq 0 ,$$

i.e., V is a Lyapunov function proving stability.

The other direction is slightly more difficult to prove...

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Lyapunov theory for LTI systems

We can always write A in Jordan normal form

$$A = T \text{diag}(J_1, \dots, J_{n_j}) T^{-1}.$$

$$A^T P + P A \preceq 0 \quad \Leftrightarrow \quad J_i^T Q_i + Q_i J_i \preceq 0$$

for all $i \in \{1, \dots, n_j\}$ with $P = (T^T)^{-1} \text{diag}(Q_1, \dots, Q_{n_j}) T^{-1}$.

- **Case 1:** If A stable and $m_i = 1$, we have $J_i = \lambda_i \leq 0$, i.e.,

$$J_i^T Q_i + Q_i J_i \preceq 0 \text{ is satisfied for any } Q_i > 0.$$

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Lyapunov theory for LTI systems

- **Case 2:** If A stable but $m_i > 1$, we must have $\lambda_i < 0$. In this case, the integral

$$Q_i = \int_0^{\infty} (e^{J_i t})^{\top} e^{J_i t} dt$$

exist and satisfies $J_i^{\top} Q_i + Q_i J_i = -I$.

Thus, if A is stable, we can construct a positive definite solution of $A^{\top} P + P A \preceq 0$. This completes the proof.

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