

# Analysis and Control of Time-Varying and Perturbed Systems

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Advanced Nonlinear Control

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# Main Objective

## Based on:

- *Nonlinear Control* (Ch. 4): Time-varying and perturbed systems
- *Nonlinear Systems* (Ch. 9, 11.5): Stability under perturbations

## Objective:

- Analyze stability under **vanishing perturbations** using comparison functions
- Study ultimate boundedness for systems with **non-vanishing perturbations**
- Apply Lyapunov-based methods for robust analysis of nonlinear, time-varying systems
- Formulate practical and broadly applicable stability conditions

# Lyapunov Theory for Time-Varying Systems

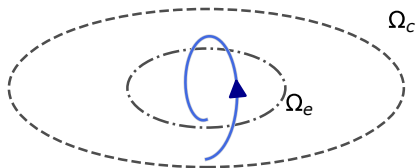
## Assumptions:

- Origin  $x = 0$  is an equilibrium point
- Lyapunov function  $V(t, x)$  is continuously differentiable, positive definite and radially unbounded
- Derivative of Lyapunov function is negative definite

## Globally uniformly exponentially stable:

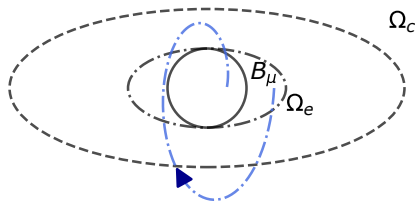
$$\begin{aligned}\exists c_i, \alpha > 0 : c_1 \|x\|^\alpha \leq V(t, x) \leq c_2 \|x\|^\alpha \\ \dot{V}(t, x) \leq -c_3 \|x\|^\alpha\end{aligned}$$

# Boundedness and Ultimate Boundedness



**Boundedness:**

$$\|x(t_0)\| \leq \alpha \Rightarrow \|x(t)\| \leq \beta,$$
$$c > 0, \alpha \in (0, c), \beta > 0, \forall t \geq t_0$$



**Ultimate Boundedness:**

$$\|x(t)\| \leq b$$
$$\forall t \geq t_0 + T$$

# Perturbation Model for Vanishing Perturbation Models

- Exact Modelling rarely feasible due to modelling errors/external disturbances or parameter drift
- $\dot{x} = f(x) + g(x, t)$ 
  - $f$  is locally Lipschitz
  - $g$  is piecewise continuous in  $t$  and locally Lipschitz
  - generally unknown but bounded
- $g(0, t)$  and  $g(x, t) = 0$  for  $t \rightarrow \infty$

# Lyapunov Stability Theorems

## Assumptions:

- Origin  $x = 0$  is an exponentially stable equilibrium point
- Perturbation vanishes
- Lyapunov function  $V(t, x)$  is continuously differentialbe, positive definite and radially unbounded

## Globally Uniformly Exponentially Stable Equilibrium:

$$\frac{\partial V}{\partial t} + \frac{\partial V}{\partial x} f(x) \leq c_3 \|x\|^2 \text{ and } \left\| \frac{\partial V}{\partial x} \right\| \leq c_4 \|x\|$$

$$\|g(x, t)\| \leq \gamma \|x\| \text{ with } 0 \leq \gamma(t) < \frac{c_3}{c_4}$$

# Comparison Functions

## Assumptions:

- Origin  $x = 0$  is an exponentially stable equilibrium point
- Perturbation vanishes

## Exponential Stability:

$$\|g(x, t)\| \leq \gamma(t)\|x\| \text{ with } \int_{t_0}^t \gamma(\tau) d\tau < \epsilon(t - t_0) + \eta$$

$$\text{with } \epsilon < \frac{c_1 c_3}{c_2 c_4}$$

# Example: Linear Time-Varying System

## Perturbed Linear Time-Varying System:

$$\dot{x} = [A(t) + B(t)]x$$

### Assumptions:

- $A(t)$  is uniformly bounded
- Origin of nominal system is uniformly exponentially stable
- $B(t) \rightarrow 0$  as  $t \rightarrow \infty$
- Lyapunov function  $V(t, x)$  is positive definite and derivative negative definite

$$g(t, x) = B(t)x \Rightarrow \|g(t, x)\| \leq \|B(t)\| \cdot \|x\| = \gamma(t)\|x\|$$

$\Rightarrow$  Exponential stability of nominal system is preserved under vanishing perturbations



# Perturbation Model for Non-Vanishing Perturbations

- $g(t, x)$  does not vanish, persistent input or modeling error
- Origin is no longer an equilibrium
- Use ultimate boundedness: trajectories enter and remain in a compact set
- Stability becomes input-to-state-like (robustness to bounded disturbances)

# Lyapunov Based Conditions for Boundedness

## Assumptions:

- Origin  $x = 0$  is an exponentially stable equilibrium point
- Non vanishing perturbation
- Lyapunov function  $V(t, x)$  is continuously differentiable, positive definite and radially unbounded
- Derivative of Lyapunov function is negative definite
- Perturbation is bounded

$$\|g(x, t)\| \leq \delta < \frac{c_3}{c_4} \sqrt{\frac{c_1}{c_2}} \theta r, \text{ with } \theta \in (0, 1), r > 0$$

# Lyapunov Based Conditions for Boundedness

## Exponential Stability:

For all initial conditions  $\|x(t_0)\| \leq \sqrt{c_1/c_2}r$

$$\|x(t)\| \leq ke^{-\gamma(t-t_0)}\|x(t_0)\|, t_0 \leq t \leq t_0 + T$$

$$\|x(t)\| \leq b \forall t \geq t_0 + T$$

$$\text{where } k = \sqrt{\frac{c_2}{c_1}}, \gamma = \frac{(1-\theta)c_3}{2c_2}, b = \frac{c_4}{c_3}k\frac{\delta}{\theta}$$

# Example: Non-Vanishing Perturbation in a Nonlinear System

- As a special case of non-vanishing perturbations
- Stability Theorems
- Case distinctions

# Conceptual Links Between Sections

- **Lyapunov Functions:** Central in both cases, quantify energy and decay
- **Main Difference:** Convergence to origin vs. convergence to a bounded set
- **Comparison Lemma:** Powerful for estimating solution bounds in both cases
- **Ultimate Boundedness:** Reflects robustness, common in practical control
- **Design Implication:** Small perturbations can be tolerated if decay is fast; persistent ones require constraint-aware design

**Takeaway:** Both frameworks enhance system analysis in the presence of uncertainty. Knowing the type of perturbation guides appropriate control strategies.

# Benefits and Drawbacks of Lyapunov-Based Analysis

## Benefits

- General applicability:
- Rigorous and systematic:
- Insightful design tool:
- Comparison lemma:
- Handles uncertainty:

## Drawbacks

- Constructing Lyapunov functions is hard:
- Results are conservative:
- Limited to local regions:
- Quantitative bounds may be loose:

# References I



Chaoqun He, Renjie Luo, Yuzhuo Bai, Shengding Hu, Zhen Leng Thai, Junhao Shen, Jinyi Hu, Xu Han, Yujie Huang, Yuxiang Zhang, Jie Liu, Lei Qi, Zhiyuan Liu and Maosong Sun.

**OlympiadBench: A Challenging Benchmark for Promoting AGI with Olympiad-Level Bilingual Multimodal Scientific Problems.** 2024. eprint: 2402.14008.



Andreas Hochlehnert, Hardik Bhatnagar, Vishaal Udandarao, Samuel Albanie, Ameya Prabhu and Matthias Bethge.

**A Sober Look at Progress in Language Model Reasoning: Pitfalls and Paths to Reproducibility.** Apr. 2025. DOI: 10.48550/arXiv.2504.07086.



Jie Huang and Kevin Chen-Chuan Chang. **Towards Reasoning in Large Language Models: A Survey.** May 2023.

DOI: 10.48550/arXiv.2212.10403.



Naman Jain, King Han, Alex Gu, Wen-Ding Li, Fanjia Yan, Tianjun Zhang, Sida Wang, Armando Solar-Lezama, Koushik Sen and Ion Stoica.

**LiveCodeBench: Holistic and Contamination Free Evaluation of Large Language Models for Code.** June 2024. DOI: 10.48550/arXiv.2403.07974.



Niklas Muennighoff, Zitong Yang, Weijia Shi, Xiang Lisa Li, Li Fei-Fei, Hannaneh Hajishirzi, Luke Zettlemoyer, Percy Liang, Emmanuel Candès and Tatsunori Hashimoto. **s1: Simple test-time scaling.** Mar. 2025. DOI: 10.48550/arXiv.2501.19393.



Simon Ott, Konstantin Hebenstreit, Valentin Liévin, Christoffer Egeberg Hother, Milad Moradi, Maximilian Mayrhauser, Robert Praas, Ole Winther and Matthias Samwald. **ThoughtSource: A central hub for large language model reasoning data.**

In: *Scientific Data* 10.1 (Aug. 2023). ISSN: 2052-4463. DOI: 10.1038/s41597-023-02433-3.



# References II



Fengli Xu, Qian Yue Hao, Zefang Zong, Jingwei Wang, Yunke Zhang, Jingyi Wang, Xiaochong Lan, Jiahui Gong, Tianjian Ouyang, Fanjin Meng, Chenyang Shao, Yuwei Yan, Qinglong Yang, Yiwen Song, Sijian Ren, Xinyuan Hu, Yu Li, Jie Feng, Chen Gao and Yong Li. **Towards Large Reasoning Models: A Survey of Reinforced Reasoning with Large Language Models.** Jan. 2025.

DOI: 10.48550/arXiv.2501.09686.



Kunhao Zheng, Jesse Michael Han and Stanislas Polu. **MiniF2F: a cross-system benchmark for formal Olympiad-level mathematics.**

Feb. 2022. DOI: 10.48550/arXiv.2109.00110.