



# Finite-Time Stabilization-Based Adaptive Fuzzy Control Design

Prepared for Adaptive Control Project  
Professor Bagheri

Presentation  
Murtaza Asaadi

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# Outline

## Outline

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# What is the Problem?

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- Controlling nonlinear systems is challenging due to uncertainties and complex dynamics.
- Traditional controllers guarantee stability over an infinite time horizon (asymptotic stability).



# Why Finite-Time Control?

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Many real-world applications (robotics, aerospace) require systems to reach their desired state within a specific, finite time.

## Benefits of Finite-Time Stability:

- Faster convergence rates.
- Higher precision in tracking.
- Improved robustness against disturbances.



# Contribution

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- Develops an adaptive fuzzy control strategy for a class of nonlinear systems.
- Guarantees that the system's tracking error converges to a small region around zero in finite time.
- Ensures all signals in the closed-loop system remain bounded and stable.



# System Representation

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$$\begin{cases} \dot{z}_i = \phi_i(\bar{z}_i) + \varphi_i(\bar{z}_i)z_{i+1}, & i = 1, \dots, n-1 \\ \dot{z}_n = \phi_n(z) + \varphi_n(z)q(v) \\ y = z_1 \end{cases}$$

- $z_i$ : System states.
- $\phi_i, \varphi_i$ : Unknown nonlinear functions.
- $v$ : The control input we design.
- $q(v)$ : A quantizer, which models the digital nature of controllers (discrete signal levels).
- $y$ : The system output.



# Control Objective

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Design a **control law**  $v$  such that the **system output**  $y$  follows a desired **reference signal**  $y_r$  in finite time, despite the unknown functions and the quantizer.



# Key Components of Control Strategy

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The controller is designed using a combination of techniques:

- Backstepping
- Fuzzy Logic Systems (FLS)
- Adaptive Control
- Hysteretic Quantizer





# Backstepping

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A recursive design methodology. It breaks down the complex  $n$ -dimensional system into a series of 1-D problems, designing a "virtual controller" at each step.



# Fuzzy Logic Systems

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Used as universal approximators. Since the functions  $\phi$  and  $\phi_i$  are unknown, an FLS is used to estimate their behavior online.



# Adaptive Control

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The parameters of the Fuzzy Logic System are not fixed; they are "adapted" or tuned in real-time by adaptive laws to improve approximation accuracy.



# Hysteretic Quantizer

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The control input  $v$  is passed through a quantizer  $q(v)$ . This models the constraints of digital hardware and communication channels.



# Controller and Adaptive Law Design

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The design follows the backstepping procedure, step-by-step.

- Step 1: Virtual Controller  $\alpha_1$
- Step 2: Actual Controller  $v$
- Adaptive Laws



# Controller and Adaptive Law Design

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The design follows the backstepping procedure, step-by-step.

- Step 1: Virtual Controller  $\alpha_1$ 
  - Define the first error surface:  $\eta_1 = z_1 - y_r$ .
  - Design a virtual controller  $\alpha_1$  to stabilize this error.
  - $\alpha_1$  includes terms to drive the error to zero and a fuzzy logic term to cancel nonlinearities.
- Step 2: Actual Controller  $v$
- Adaptive Laws



# Controller and Adaptive Law Design

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The design follows the backstepping procedure, step-by-step.

- Step 1: Virtual Controller  $\alpha_1$
- Step 2: Actual Controller  $v$ 
  - Define the final error surface:  $\eta_2 = z_2 - \alpha_1$ .
  - Design the actual control input  $v$  to stabilize  $\eta_2$ .
- Adaptive Laws



# Controller and Adaptive Law Design

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The design follows the backstepping procedure, step-by-step.

- Step 1: Virtual Controller  $\alpha_1$
- Step 2: Actual Controller  $v$
- Adaptive Laws
  - For each fuzzy system, an adaptive law updates its weight parameter  $\theta_i$ .
  - The goal is to minimize the function approximation error.





# Stability Analysis

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A total Lyapunov function  $V$  is constructed for the entire closed-loop system. The paper proves that the derivative of the Lyapunov function satisfies:

$$\dot{V} \leq -\lambda_0 V - \lambda_1 V^h + b_0$$

- $\lambda_0 V$ : Guarantees exponential stability.
- $\lambda_1 V^h$ : Guarantees finite-time stability.
- $b_0$ : A small positive constant due to approximation errors.



# Theorem 1 (Main Result)

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The designed controller ensures that:

- The system is practically stable in finite time. The tracking error converges to a small, bounded region around zero.
- The size of this region and the convergence time can be calculated.
- All signals within the system (states, adaptive parameters) remain bounded.



# Python Implementation Overview

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The simulation was reproduced using Python with SciPy and Matplotlib.

## Key Implementation Steps:

- Define Parameters
- Hysteretic Quantizer
- Fuzzy Basis Functions
- System Model Function
- ODE Solver
- Plotting



# System Output vs. Reference Signal

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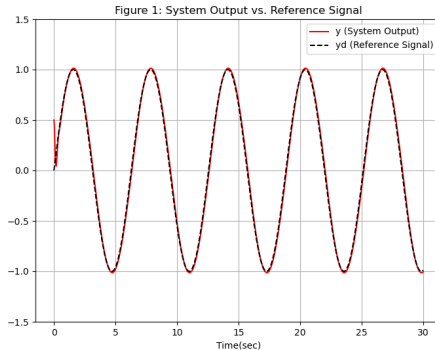
## Adaptive

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- The plot shows the system output  $y$  quickly converging to and tracking the sinusoidal reference signal  $y_d$ .
- This demonstrates the effectiveness of the tracking control.



# State Variable $z_2$

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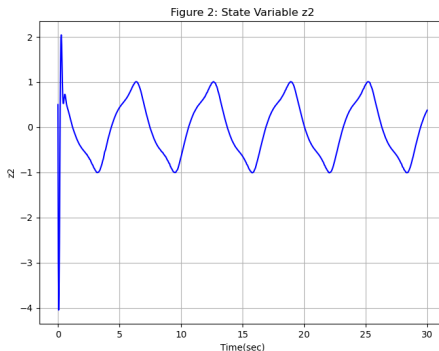
## Adaptive

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- The state  $z_2$  remains bounded throughout the simulation, confirming the stability of all system signals.



# Adaptive Parameters

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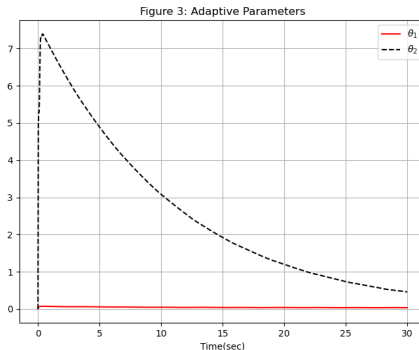
## Adaptive

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- The adaptive parameters  $\theta_1$  and  $\theta_2$  converge to stable values.
- This indicates that the fuzzy logic systems have successfully learned to approximate the unknown nonlinearities.



# Quantizer Output

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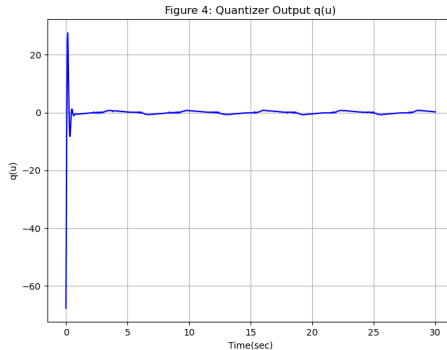
## Adaptive

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- This plot shows the discrete output of the hysteretic quantizer.
- It highlights that the controller operates effectively even with a non-continuous, quantized input signal.



# Conclusion

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- Successfully simulated an adaptive fuzzy controller that achieves finite-time stability for a class of nonlinear systems.
- The use of backstepping and fuzzy logic systems effectively handles system uncertainties.
- The design explicitly accounts for input quantization, making it more practical for digital implementation.
- Stability analysis proves that all signals are bounded and the tracking error converges to a small region in finite time.
- The Python simulation verifies the theoretical results, showing excellent tracking performance and stability.





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Thank you.  
Any questions?