

### Abstract

In this poster, an adaptive control strategy is presented for quadrotors under parametric uncertainties. The adaptive control scheme learns the quadrotor's inverse model with a Lyapunov-based adaptation law. For that, a robustness feedback loop is used to stabilize the quadrotor at start-up. Therefore, the controller achieves accurate motion tracking with parametric uncertainties. Unlike many controllers, the proposed adaptive control scheme's stability is guaranteed by Lyapunov direct method. The proposed controller's performance in coping with parameter variations is highlighted in different operating conditions.

## Introduction

Unmanned Aerial Vehicles (UAVs) have attracted the attention of the scientific community from diverse disciplines due to their versatile applications. Their popularity has increased exponentially, particularly quadrotors, which is mainly due to their ability to hover and maneuver in tight and dangerous places. Consequently, they have become widely used in many applications such as surveillance, exploration, rescue missions and payload transportation.

Studies have shown that the design of robust controllers for mathematically ill-defined systems that may be subjected to structured and unstructured uncertainties is made possible with computational intelligence tools, such as artificial neural networks and fuzzy logic systems. The approximation capabilities have been the main driving force behind the increasing popularity of such methods as they are theoretically capable of uniformly approximating any continuous real function to any degree of accuracy. This has led to the recent advances in the area of intelligent control. Satisfactory performance is achieved with various neural network models for complex systems control.

The contribution of this research is to achieve motion tracking for quadrotors in the presence of parametric uncertainties. The proposed control scheme makes use of adaptive control theory to cope with parameter variations while a robustness feedback loop copes with the time-varying modeling and disturbance uncertainties. Unlike many controllers, the closed-loop control scheme's stability is guaranteed by Lyapunov direct method. This paper deals with high motion tracking performance under both structured and unstructured uncertainties



# **Dynamics**

 $\ddot{x} = -(\cos\phi\sin\theta\cos\psi + \sin\phi\sin\psi)u_1$  $\ddot{y} = -(\cos\phi\sin\theta\sin\psi - \sin\phi\cos\psi)u_1$  m = mass of the quadrotor

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Where

$$u_{1} = b(\Omega_{1}^{2} + \Omega_{2}^{2} + \Omega_{3}^{2} + \Omega_{4}^{2})$$
$$u_{2} = b(\Omega_{2}^{2} - \Omega_{4}^{2})$$
$$u_{3} = b(\Omega_{3}^{2} - \Omega_{1}^{2})$$
$$u_{4} = b(\Omega_{1}^{2} - \Omega_{2}^{2} + \Omega_{3}^{2} - \Omega_{4}^{2})$$
$$\Omega_{r} = \Omega_{1} - \Omega_{2} + \Omega_{3} - \Omega_{4}$$

- l = length from the rotors to the center of mass
- g = gravitational constant
- $\phi$  = roll angle of the quadrotor
- $\theta$  = pitch angle of the quadrotor  $\psi$  = yaw angle of the quadrotor
- $u_i = \text{control inputs}$
- $I_r$  = moments of inertia of the propeller blades  $\Omega_r$  = angular velocity of the propeller blades  $I_x$ ,  $I_y$ ,  $I_z$  = moments of interia of the quadrotor
- $u_1$  provides thrust on the body in the z-axis.  $u_2$  and  $u_3$  are the roll and pitch inputs.  $u_4$  is used for the yaw control.  $\Omega_1, \Omega_2, \Omega_3$ , and  $\Omega_4$  are speed of the motors



# **Control Design**

Parameter	Quadrotor 1	Quadrotor 2	Quadrotor 3	Unit
Mass	0.68	0.80	1.50	(kg)
Arm length	0.18	0.60	1	(m)
Thrust coefficient	$6.9 \cdot 10^{-6}$	$192 \cdot 10^{-7}$	$1.48\cdot 10^{-7}$	$(N \cdot s^2)$
Drag coefficient	$9.3 \cdot 10^{-8}$	$4 \cdot 10^{-8}$	$2.9 \cdot 10^{-9}$	(N·m·s <sup>2</sup> )
Rotor inertia	$1.4 \cdot 10^{-5}$	$6 \cdot 10^{-5}$	$2 \cdot 10^{-4}$	(kg·m <sup>2</sup> )
Inertia on x axis	$3.5 \cdot 10^{-3}$	$15.67 \cdot 10^{-3}$	$8.5 \cdot 10^{-3}$	(kg·m <sup>2</sup> )
Inertia on y axis	$4.2 \cdot 10^{-3}$	$15.67 \cdot 10^{-3}$	$8.5 \cdot 10^{-3}$	(kg·m <sup>2</sup> )
Inertia on z axis	$7.5 \cdot 10^{-3}$	$28.34 \cdot 10^{-3}$	$16.33 \cdot 10^{-3}$	(kg·m <sup>2</sup> )

# ominal Case Quadrotor 1



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Let 
$$e_2 = x = x^*, e_2 = \phi - \phi^*, e_3 = \theta - \phi^*, e_9 = \psi - \psi^*$$
 denotes the system's tracking errors.  
Let's define the following reference model:  

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x_5 = e$$

 $\widehat{P}_z = -\Gamma_z \Upsilon_z s_z$  $\widehat{P_{\phi}} = -\Gamma_{\phi} \Upsilon_{\phi} s_{\phi}$  $\widehat{P_{\theta}} = -\Gamma_{\theta} \Upsilon_{\theta} s_{\theta}$  $\widehat{P_{\psi}} = -\Gamma_{\psi} \Upsilon_{\psi} s_{\psi}$ 

Where  $\Gamma = [\gamma_1, \gamma_2, \dots, \gamma_m]^T$  and  $\gamma_i$  is a positive constant gain,  $i = 1, \dots, m$ , with *m* being the dimension of  $\gamma_0$  or  $P_0$ .

# **Adaptive Motion Control of Quadrotors under Parametric Uncertainties** with Stability

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# **Results and Discussion**

## **Table 1.** Physical parameters of several quadrotors



# ominal Case Quadrotor 2



# **Control under disturbances**



# **Control under noise**



A Lyapunov-based adaptive controller is proposed for high-performance control of quadrotors. Results shows that the proposed adaptive controller is more efficient in dealing with parametric uncertainties

## **Nominal Case Quadrotor 3**