

# Robust Backstepping Sliding Mode Control and Observer-Based Fault Estimation for a Quadrotor UAV

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**Abstract**—This study gives the mathematic model of a quadrotor unmanned aerial vehicle (UAV) and then proposes a robust nonlinear controller which combines the sliding mode control technique and the backstepping control technique. To achieve Cartesian position trajectory tracking capability, the construction of the controller can be divided into two stages: a regular SMC controller for attitude subsystem (inner loop) is first developed to guarantee fast convergence rapidity of Euler angles and the backstepping technique is applied to the position loop until desired attitudes are obtained and then the ultimate control laws. The stability of the closed-loop system is guaranteed by stabilizing each of the subsystems step by step and the robustness of the controller against model uncertainty and external disturbances is investigated. In addition, an adaptive observer-based fault estimation scheme is also considered for taking off mode. Simulations are conducted to demonstrate the effectiveness of the designed robust nonlinear controller and the fault estimation scheme.

**Index Terms**—Backstepping control, Sliding mode control, Quadrotor UAV, Robustness, Fault estimation

## I. INTRODUCTION

AS a new-born member of the small unmanned aerial vehicle (UAV) family, quadrotor is becoming a popular topic among researchers and scholars because of its extensive utility in several important applications such as military surveillance, disaster monitoring, agricultural mapping, and rescue mission and so on [1]. The main advantages of quadrotor UAVs lie in flying in any direction, taking off and landing vertically and hovering at a desired altitude. However, its characteristics, such as nonlinearity,

strong coupling and under-actuating, make designing an effective controller for position trajectory tracking control difficult. What's more, taking robustness of the trajectory tracking controller into consideration poses a bigger challenge.

In the past few years, researchers have paid attention to control and have proposed different kinds of control methods, such as linear and adaptive PID controls [2-3], dynamic inversion Control [4], feedback linearization control [5-6], model predictive control [7], backstepping control [8-9], sliding mode control (SMC) [10-11], LMI-based linear control [12], observer-based control [13-18], adaptive control [19-21] and nonlinear control [22-23], just to numerate a few. Recently, with the development of intelligence control theory, some advanced or modified controllers, which combine the aforementioned basic methods with intelligent control methods, such as fuzzy control and switching topology, have been presented for a few complicated tasks [24-26]. Fault tolerant control and robust control methods are also becoming much more popular for fault tolerance and robustness against disturbances, respectively [20, 27-31]. The performances of these controllers are satisfying in terms of attitude or position trajectory tracking capability when some particular circumstances are considered. For instance, [1] proposed a new feedback control scheme for attitude stabilization of a quadrotor; [2] designed a flight control system capable of not only stabilizing attitude but also tracking a trajectory accurately for an under-actuated quadrotor aircraft; [18] presented a robust attitude tracking controller for quadrotors based on the nonlinear disturbance observer; [29] proposed a robust controller consist of a nominal feedforward controller, a nominal linear LQR controller and a robust compensator to deal with robust attitude control problem. The proposed controllers aforementioned are demonstrated to be effective in some specific cases. However, all the works above are limited to attitude trajectory tracking control problem and lack focus on position control which has highly important practical utilities. On the other hand, many existing works that aimed at position trajectory tracking control have either paid little attention to robustness and fault tolerance of the proposed controllers or preferred a simplified form of dynamic equation, for example, [6] presented a feedback linearization controller and noticed its sensitivity to noise and modeling uncertainly, so further presented a sliding mode controller to overcome it, however, the quadrotor dynamics is a simplified one and the coupling between attitudes and positions was omitted; [9] applied backstepping

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control on the Lagrangian form of the dynamics of a quadrotor and also adopted a simplified form of dynamics; [23] presented a nonlinear adaptive state feedback controller that stabilizes the closed-loop system in the presence of force disturbances, however, the adopted model was a nominal one and the disturbance was assumed to be constant which is often not the case. Motivated by these, this study investigates the possibility of a robust backstepping sliding mode controller for the nonlinear strong coupled model of a quadrotor UAV.

As an extensively used nonlinear control method, SMC features excellent performance properties and robustness for a specific class of nonlinear tracking problems [32-33] and is also used in some areas to address practical and important issues [36-38]. SMC utilizes a high frequency switching control signal to enforce the system trajectories onto a surface, that is, the so-called sliding surface or sliding manifold which is designed to achieve desired specifications. In a finite time, the system trajectories arrive and remain within the vicinity of the sliding surface. The sliding mode control is robust with respect to internal and external disturbances, as well as parameter uncertainties which accounts for its popularity among the nonlinear control methods. Quadrotor position control is a typical nonlinear control problem because complicated nonlinearity and coupling exist in its dynamics. And further, researchers have developed some improved sliding mode controllers with robustness against perturbation and uncertainties [10, 27]. However, most of the existing literatures have focused on using SMC methods to address attitude control of a quadrotor UAV rather than position trajectory tracking control design since the transformed dynamics equation has a preferable form for attitude control. However, position control has highly important practical utilities and correspondingly increased difficulties in control design.

Backstepping technique starts the control design process in feedback form at the origin of a complicated system and backs out to new controllers that stabilize each outer subsystem step by step. Virtual controls are utilized in these steps to guarantee the stability of each subsystem and they are finally determined by performance specifications. The backing out process continues until the ultimate control law is obtained. However, in many existing research works, backstepping control is utilized alone for position control of a quadrotor UAV with very little discussion about robustness against external disturbances and model uncertainties [8]. In this study, we consider a position trajectory tracking control for a quadrotor UAV with the consideration of robustness against disturbances and model uncertainties. As the study will show, we combine the SMC and the backstepping control techniques to develop a robust backstepping sliding mode controller which is constructed by two stages. A regular sliding mode controller for attitude control is first designed to guarantee fast attitude trajectory tracking capability. Then, we use backstepping technique for Cartesian position trajectory tracking control and obtain a group of equations by mathematically solving which the desired Euler angles are given. Applying the desired angles as input for the attitude system, the final robust sliding mode control laws are then developed. In constructing the robust nonlinear controller, "virtual control" and sliding manifold are used to guarantee position trajectory tracking capability as well as robustness against disturbances and uncertainties.

Compared with traditional aircrafts, a quadrotor UAV is more likely to encounter problems such as actuator faults, which threaten the system's stability and reliability. These potential problems pose a challenge to the applications of a quadrotor UAV in some significant situations [34]. A lot of work has been done in fault diagnosis field. In the past few years, model-dependent methods have been studied and mature [35]. Recently, new data-driven methods are being researched and some results have been produced [38-41]. Inspired by the methods applied to handle conventional fixed-wing aircraft's actuator faults, fault tolerant control is a possible option for a quadrotor UAV. The key issue for fault tolerant control rests with obtaining enough information about faults where the fault estimation scheme plays an important role. Thus, apart from robust controller design, fault estimation scheme is also considered in some certain cases to supplement the controller design.

This paper is organized as follows. A non-simplified mathematic model is first developed. Afterwards, a sliding mode controller for attitude trajectory control is presented, followed by a robust backstepping sliding mode controller for position trajectory tracking control. In addition, an observer-based fault estimation scheme is considered and the conditions and constraints for its applicability are provided. Simulations are conducted in different scenarios to demonstrate the effectiveness of the proposed robust backstepping sliding mode controller and the fault estimation scheme. Conclusions are drawn in the end of this paper.

## II. DYNAMIC MODEL

Quadrotor is an under-actuated system because it has six degrees of freedom but only four actual inputs. The six degrees of freedom include translational motion in three directions and rotational motion around three axes. The schematic configuration of a quadrotor we adopted in this study is shown in Fig.1. Four rotors, each of which is driven by a motor, are mounted at two orthogonal directions and rotates in a direction as the circular arrow shows. The body-fixed frame which is represented by  $Ox_b y_b z_b$  with its origin at the center of the mass and the inertia frame which is represented by  $Ox_e y_e z_e$  are as shown in Fig.1. The front (Rotor 1) and the rear (Rotor 3) rotors rotate counter clockwise, while the lateral rotors (Rotor 2 and Rotor 4) rotate clockwise to generate the lift force and to balance the yaw torque as needed. On varying the rotor speeds altogether with the same quantity, the lift forces will change, affecting the altitude of the vehicle in this case. Pitch rotation can be obtained by changing the speeds of the front and the rear rotors conversely while coupled with translation along  $Ox_b$ . Roll rotation and translation along  $Oy_b$  can be obtained in a similar way by varying the speeds of the lateral ones conversely. The equations describing the dynamics of a quadrotor are basically those of a rotating rigid body with six degrees of freedom which can be derived with Newton-Euler formulas. The complicated motions of a quadrotor can be described by two typical groups of equations, each of which represents a subsystem with coupled terms. The first group is related to the translational positions and the second group is related to the rotational angles. In this section, we will give the quadrotor UAV's mathematic model including the navigation equations and the moment equations [1].

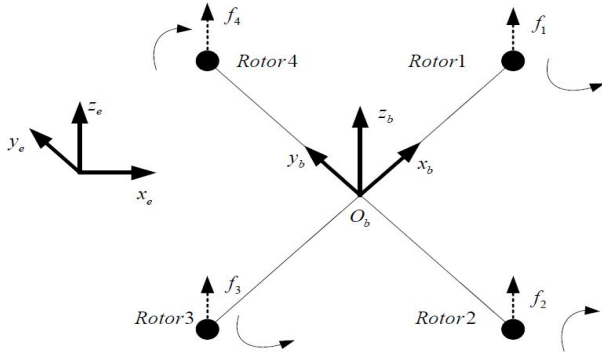


Fig.1 Quadrotor configuration frame scheme with body fixed and the inertia frames

Define  $\xi = [\phi, \theta, \psi]^T \in \mathbb{R}^3$  and  $\omega_b = [p, q, r]^T$ , where  $\phi$ ,  $\theta$  and  $\psi$  denote the angle of roll, pitch and yaw with respect to the inertia frame and  $p$ ,  $q$  and  $r$  denote the angular velocity of roll, pitch and yaw with respect to the body-fixed frame. The rotation matrix from body-fixed frame to inertia frame can be obtained as:

$$R_t = \begin{bmatrix} C_\psi C_\theta & C_\psi S_\theta S_\phi - S_\psi C_\phi & C_\psi S_\theta C_\phi + S_\psi S_\phi \\ S_\psi C_\theta & S_\psi S_\theta S_\phi + C_\psi C_\phi & S_\psi S_\theta C_\phi - C_\psi S_\phi \\ -S_\theta & C_\theta S_\phi & C_\theta C_\phi \end{bmatrix} \quad (1a)$$

where  $S_{(\cdot)}$  and  $C_{(\cdot)}$  denote  $\sin(\cdot)$  and  $\cos(\cdot)$  respectively. In particular,  $R_t$  is an orthogonal matrix which means that  $R_t^T = R_t^{-1}$  when  $R_t$  is nonsingular. In particular, the transformation matrix  $R_t$  mentioned in (1a) is special because the following equation holds

$$\dot{R}_t = R_t S(\omega_b) \quad (1b)$$

where  $S(\omega_b)$  is a skew symmetric matrix which means  $S^T(\omega_b) = -S(\omega_b)$  and  $S(\omega_b)$  is given as:

$$S(\omega_b) = \begin{bmatrix} 0 & -r & q \\ r & 0 & -p \\ -q & p & 0 \end{bmatrix} \quad (1c)$$

For any given  $V = [v_1 \ v_2 \ v_3]^T \in \mathbb{R}^3$ , it is easy to verify

$$S(\omega_b) \cdot V = \omega_b \times V \quad (2a)$$

According to the rotation matrix  $R_t$  the relationship between  $\dot{\xi}$  and  $\omega_b$  can be described as:

$$\omega_b = R_t \dot{\xi} = \begin{bmatrix} 1 & 0 & -S_\theta \\ 0 & C_\phi & C_\theta S_\phi \\ 0 & -S_\phi & C_\theta C_\phi \end{bmatrix} \begin{bmatrix} \dot{\phi} \\ \dot{\theta} \\ \dot{\psi} \end{bmatrix} \quad (2b)$$

$M_b$  is defined as the torque provided by the rotors with respect to the body-fixed frame, and is described as follows:

$$M_b = \begin{bmatrix} M_{bx} \\ M_{by} \\ M_{bz} \end{bmatrix} = \begin{bmatrix} l\kappa(\Omega_4^2 - \Omega_2^2) \\ l\kappa(\Omega_3^2 - \Omega_1^2) \\ \tau(\Omega_2^2 + \Omega_4^2 - \Omega_1^2 - \Omega_3^2) \end{bmatrix} \quad (3a)$$

where  $\Omega_i$ ,  $i=1,2,3,4$  denote rotary speed of the front, right, rear, and left rotors, respectively;  $l$  is the distance between a rotor and the center of mass of the quadrotor;  $\kappa$  is the drag force coefficient; and  $\tau$  is reverse moment coefficient.

Using the Newton-Euler equation, the rotational dynamic equation of the quadrotor is obtained as follows:

$$M_b = J_b \dot{\omega}_b + \omega_b \times J_b \omega_b + M_g + M_d \quad (3b)$$

where  $J_b = \text{diag}(J_x, J_y, J_z)$  is a symmetric positive definite constant matrix with  $J_x$ ,  $J_y$  and  $J_z$  being the rotary inertia with respect to the  $O_b x_b$ ,  $O_b y_b$ , and  $O_b z_b$  axes, respectively; the notation  $\times$  denotes cross multiplication;  $M_g$  and  $M_d$  are the resultant torques due to the gyroscopic effects and the resultant of aerodynamic frictions torque. They are given as:

$$M_g = \sum_{i=1}^4 \omega_b \times J_r [0, 0, (-1)^{i+1} \Omega_i]^T \quad (3c)$$

$$M_d = \text{diag}(d_\phi, d_\theta, d_\psi) \dot{\xi}$$

where  $J_r$  is the moment of inertia of each rotor;  $d_\phi$ ,  $d_\theta$  and  $d_\psi$  are the corresponding aerodynamic drag coefficients.

Considering that  $J_b$  is nonsingular and transforming (3b), the following equation can be obtained:

$$\dot{\omega}_b = J_b^{-1} [M_b - M_g - M_d - \omega_b \times (J_b \omega_b)] \quad (4)$$

Furthermore, with the aid of (1c), (2a)-(2b) and approximation of Euler angles at equilibrium point, the following dynamic equations can be obtained:

$$\begin{aligned} \ddot{\phi} &= (\dot{\theta}\dot{\psi}(J_y - J_z) - J_r \dot{\theta}\dot{\psi} - d_\phi \dot{\phi} + M_{bx}) / J_x \\ \ddot{\theta} &= (\dot{\phi}\dot{\psi}(J_z - J_x) - J_r \dot{\phi}\dot{\psi} - d_\theta \dot{\theta} + M_{by}) / J_y \\ \ddot{\psi} &= (\dot{\phi}\dot{\theta}(J_x - J_y) - d_\psi \dot{\psi} + M_{bz}) / J_z \end{aligned} \quad (5)$$

where  $\varpi = \Omega_4 + \Omega_3 - \Omega_2 - \Omega_1$  can be obtained easily online.

**Remark 2.1:** It is worth nothing to point out that the roll and pitch angles are limited to  $(-\pi/2, \pi/2)$  which is physically meaningful. In particular, the yaw angle in this study is also limited to  $(-\pi/2, \pi/2)$ .

Let  $P = [x, y, z]^T \in \mathbb{R}^3$  denote the position with respect to the inertia frame. The translational dynamic equations of the quadrotor is given by

$$m\ddot{P} = R_t \cdot F + \begin{bmatrix} 0 \\ 0 \\ -mg \end{bmatrix} - \begin{bmatrix} d_x \dot{x} \\ d_y \dot{y} \\ d_z \dot{z} \end{bmatrix} \quad (6)$$

where  $R_t$  is given in (1a), and  $d_x$ ,  $d_y$  and  $d_z$  are the air drag coefficients which are added in (6) to model the drag force caused by translational motions;  $F$  is the lift force generated by rotors with respect to the body-fixed frame

$$F = \begin{bmatrix} 0 \\ 0 \\ \kappa(\Omega_1^2 + \Omega_2^2 + \Omega_3^2 + \Omega_4^2) \end{bmatrix} \quad (7)$$

By combining (5) and (6), a compact affine nonlinear equation of the quadrotor UAV is given

$$\dot{X} = f(X) + g(X)U \quad (8)$$

where  $X = [\phi, \dot{\phi}, \theta, \dot{\theta}, \psi, \dot{\psi}, x, \dot{x}, y, \dot{y}, z, \dot{z}]^T \in \mathbb{R}^{12}$  is the state variable,  $f(X)$  and  $g(X)$  are smooth functions on  $X$ . The Equation (8) is expended as follows:

$$\begin{cases} \dot{x}_1 = x_2 \\ \dot{x}_2 = a_1 x_4 x_6 + a_2 \varpi x_4 - a_3 x_2 + U_1 \\ \dot{x}_3 = x_4 \\ \dot{x}_4 = a_4 x_2 x_6 + a_5 \varpi x_2 - a_6 x_4 + U_2 \\ \dot{x}_5 = x_6 \\ \dot{x}_6 = a_7 x_2 x_4 - a_8 x_6 + U_3 \\ \dot{x}_7 = x_8 \\ \dot{x}_8 = (C_{x_1} S_{x_3} C_{x_5} + S_{x_1} S_{x_5}) U_4 - a_9 x_8 \\ \dot{x}_9 = x_{10} \\ \dot{x}_{10} = (C_{x_1} S_{x_3} S_{x_5} - S_{x_1} S_{x_5}) U_4 - a_{10} x_{10} \\ \dot{x}_{11} = x_{12} \\ \dot{x}_{12} = -g + (C_{x_1} C_{x_3}) U_4 - a_{11} x_{12} \end{cases} \quad (9)$$

where  $\varpi$ ,  $d_\phi$ ,  $d_\theta$ ,  $d_\psi$ ,  $l$ ,  $\kappa$ ,  $C_{(\cdot)}$ ,  $S_{(\cdot)}$  and  $m$  are same as those mentioned earlier and  $g$  is the gravity acceleration.  $U_i$ ,  $i=1,2,3,4$  are the control inputs defined as follows:

$$\begin{aligned} U_1 &= l\kappa(\Omega_4^2 - \Omega_2^2)/J_x \\ U_2 &= l\kappa(\Omega_3^2 - \Omega_1^2)/J_y \\ U_3 &= l\kappa(\Omega_4^2 + \Omega_2^2 - \Omega_3^2 - \Omega_1^2)/J_z \\ U_4 &= l\kappa(\Omega_1^2 + \Omega_2^2 + \Omega_3^2 + \Omega_4^2)/m \end{aligned} \quad (10)$$

and  $a_i$ ,  $i=1,2,\dots,11$ , are the normalized parameters defined as follows:

$$\begin{aligned} a_1 &= \frac{J_y - J_z}{J_x}, \quad a_2 = \frac{J_r}{J_x}, \quad a_3 = \frac{d_\phi}{J_x}, \quad a_4 = \frac{J_z - J_x}{J_y}, \quad a_5 = \frac{J_r}{J_y}, \\ a_6 &= \frac{d_\theta}{J_y}, \quad a_7 = \frac{J_x - J_y}{J_z}, \quad a_8 = \frac{d_\psi}{J_z}, \quad a_9 = \frac{d_x}{m}, \quad a_{10} = \frac{d_y}{m}, \quad a_{11} = \frac{d_z}{m}. \end{aligned}$$

### III. BACKSTEPPING SLIDING MODE CONTROLLER FOR POSITION TRAJECTORY TRACKING CONTROL

In this section, a regular sliding mode controller for attitude trajectory tracking control is first presented. The SMC technique is then combined with the backstepping control technique to develop a robust backstepping sliding mode controller for position trajectory tracking control. The regular sliding mode controller is designed to ensure the trajectory tracking capability of  $\{\phi(t), \theta(t)\}$  along the desired roll and pitch angle trajectory  $\{\phi_d(t), \theta_d(t)\}$ . The robust backstepping sliding mode controller constructed based on the regular controller aims to ensure the position and yaw angle tracking performance along the reference input  $\{x_r(t), y_r(t), z_r(t), \psi_r(t)\}$  robustly. The block diagram of control structure is shown in Fig.2.

#### A. Regular Sliding Mode Controller for Attitude Trajectory Tracking Control

The sliding mode control features acceptable robustness in the presence of model uncertainties and external disturbances for a specific class of nonlinear tracking problems and becomes a good choice for attitude control of a quadrotor UAV. SMC focuses on designing a sliding manifold and discontinuous control laws that drive the system toward the desired dynamic behavior and ensure the tracking performance. This kind of direct nonlinear control strategy for a complicated coupled nonlinear system avoids

the inaccuracy caused by linearization around the operation points, and is extensively applied in nonlinear control problems. In this part, a sliding mode controller for attitude trajectory tracking control is developed. The laws presented in this part will be used directly in the position trajectory tracking controller design part.

Part of the state variable  $X$  is selected to form a new sub-state variable  $\chi$ , and the considered attitude dynamic system can be extracted as follows:

$$\begin{cases} \dot{\chi}_1 = \chi_2 \\ \dot{\chi}_2 = a_1 \chi_4 \chi_6 + a_2 \varpi \chi_4 - a_3 \chi_2 + U_{s1} \\ \dot{\chi}_3 = \chi_4 \\ \dot{\chi}_4 = a_4 \chi_2 \chi_6 + a_5 \varpi \chi_2 - a_6 \chi_4 + U_{s2} \\ \dot{\chi}_5 = \chi_6 \\ \dot{\chi}_6 = a_7 \chi_2 \chi_4 - a_8 \chi_6 + U_{s3} \end{cases} \quad (11)$$

where  $\chi = [\phi, \dot{\phi}, \theta, \dot{\theta}, \psi, \dot{\psi}]^T \in \mathbb{R}^6$  and the parameters are same as the above mentioned ones in (9). Our control objective for the attitude system (11) is to ensure the trajectory tracking of  $\{\phi(t), \theta(t), \psi(t)\}$  along the desired attitude trajectory  $\{\phi_d(t), \theta_d(t), \psi_d(t)\}$  by designing SMC

laws  $U_s = [U_{s1}, U_{s2}, U_{s3}]^T$ .

The regular sliding mode controller for attitude trajectory tracking control starts with the roll angle. A proper sliding manifold for roll motion control is chosen as follows:

$$s_1 = \dot{\epsilon}_1 + c_1 \epsilon_1 \quad (12)$$

where  $\epsilon_1 = \chi_{1d} - \chi_1$ ,  $c_1$  is a positive constant and  $\chi_{1d}$  is the desired roll angle command. The derivative of  $s_1$  with respect to time is

$$\begin{aligned} \dot{s}_1 &= \dot{\epsilon}_1 + c_1 \dot{\epsilon}_1 \\ &= \ddot{\chi}_{1d} - \ddot{\chi}_1 + c_1 \dot{\epsilon}_1 \\ &= \ddot{\chi}_{1d} + c_1 \dot{\epsilon}_1 - (a_1 \chi_4 \chi_6 + a_2 \varpi \chi_4 - a_3 \chi_2) - U_{s1} \\ &= \ddot{\chi}_{1d} + c_1 \dot{\chi}_{1d} - (a_1 \chi_4 \chi_6 + a_2 \varpi \chi_4 - a_3 \chi_2 - c_1 \chi_2) - U_{s1} \end{aligned} \quad (13)$$

where  $\dot{\chi}_{1d}$  and  $\ddot{\chi}_{1d}$  are the derivative and second derivative of the desired roll angle command which can be obtained approximately by stable first and second order differential filters. Assume that the state variables in (11) are measurable and the sliding mode control law for roll motion is designed as follows:

$$U_{s1} = \ddot{\chi}_{1d} + c_1 \dot{\epsilon}_1 - (a_1 \chi_4 \chi_6 + a_2 \varpi \chi_4 - a_3 \chi_2) + k_1 \text{sign}(s_1) \quad (14)$$

where  $\text{sign}(\cdot)$  is the sign function;  $k_1$  is a positive constant to be determined and is also often in accordance to the requirement of performance and can be acquired through trial and error.

A Lyapunov candidate is chosen as follow:

$$V_{s1} = \frac{1}{2} s_1^2 \quad (15)$$

The derivative of (15) with respect to time can be written as (16) when  $U_{s1}$  is substituted.

$$\begin{aligned} \dot{V}_{s1} &= s_1 \dot{s}_1 \\ &= s_1 (\ddot{\chi}_{1d} + c_1 \dot{\epsilon}_1 - (a_1 \chi_4 \chi_6 + a_2 \varpi \chi_4 - a_3 \chi_2) - U_{s1}) \\ &= -s_1 * k_1 \text{sign}(s_1) \\ &= -k_1 |s_1| \leq 0 \end{aligned} \quad (16)$$

The combination of (15) and (16) implies that the sliding manifold designed in (12) is reachable.

Similar design procedures can be conducted to design SMC laws for trajectory tracking control of pitch angle ( $\chi_3$ ) and yaw angle ( $\chi_5$ ). The corresponding control laws are designed as follows:

$$\begin{cases} U_{s2} = \ddot{\chi}_{3d} + c_2\dot{\varepsilon}_3 - (a_4\chi_2\chi_6 + a_5\chi_2 - a_6\chi_4) + k_2\text{sign}(s_2) \\ U_{s3} = \ddot{\chi}_{5d} + c_3\dot{\varepsilon}_5 - (a_7\chi_2\chi_4 - a_8\chi_6) + k_3\text{sign}(s_3) \end{cases} \quad (17)$$

where  $s_2 = \dot{\varepsilon}_3 + c_2\varepsilon_3$  and  $s_3 = \dot{\varepsilon}_5 + c_3\varepsilon_5$  are the properly designed sliding manifolds for pitch angle and yaw angle trajectory control.  $\varepsilon_3 = \chi_{3d} - \chi_3$  and  $\varepsilon_5 = \chi_{5d} - \chi_5$  are state errors.  $c_2, c_3, c_4, k_2, k_3$  and  $k_4$  are positive constants to be determined and need to be tuned by trial and error to reach the requirement of performance of the system.

The Lyapunov candidate for attitude system (11) is chosen as

$$V_s = \frac{1}{2}(s_1^2 + s_2^2 + s_3^2) \quad (18)$$

The derivative of (18) with respect to time can be written as (19) with  $U_{s1}, U_{s2}, U_{s3}$  being substituted by (14) and (17).

$$\begin{aligned} \dot{V}_s &= \sum_{i=1}^3 s_i \dot{s}_i \\ &= \sum_{i=1}^3 -s_i * k_i \text{sign}(s_i) \\ &= \sum_{i=1}^3 -k_i |s_i| \leq 0 \end{aligned} \quad (19)$$

Thus, the stability of the attitude system (11) is guaranteed by (18)-(19), ensuring attitude trajectory tracking capability.

In this part, regular SMC laws for attitude trajectory tracking control of a quadrotor UAV are presented. Sliding surfaces are chosen similarly for each Euler angle and control laws are designed under the assumption that state variables in (11) are measurable. Lyapunov-based analysis shows the stability of such an attitude subsystem. On the basis of the preceding regular SMC laws, a robust backstepping sliding mode controller for position trajectory tracking control is presented in the following part.

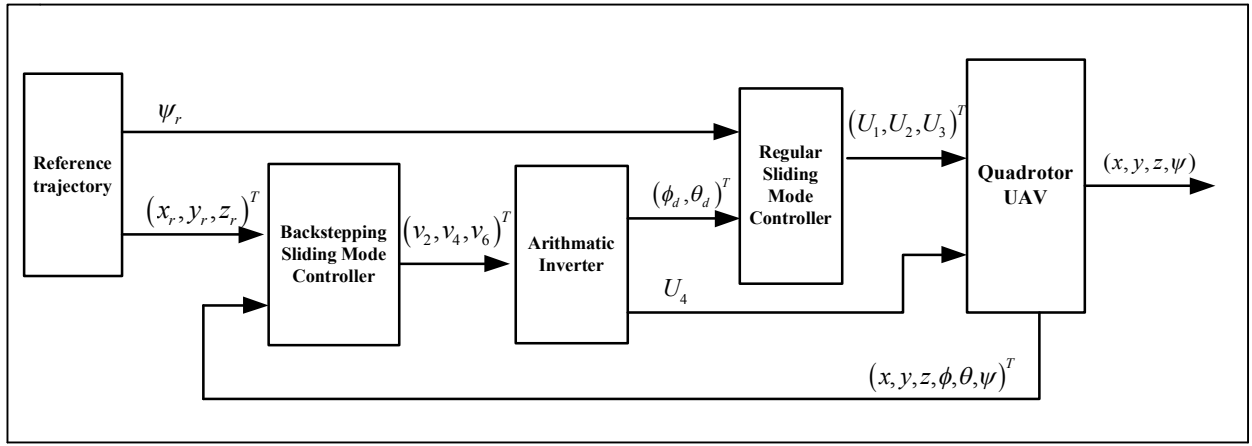


Fig.2 Block diagram of the control structure

### B. Robust Backstepping Sliding Mode Control Design for Position Trajectory Tracking Control

The regular sliding mode controller for attitude control of a quadrotor UAV presented above is effective in terms of attitude trajectory tracking. It guarantees stability of the attitude subsystem and provides tracking capability. However, the controller has some deficiencies in realizing position control since that the dynamic model has an adverse form. In (9), control input  $U_4$  is explicitly related to positions  $\{x(t), y(t), z(t)\}$  and is the single external excitation for each of these three controlled variables. This means designing a separate sliding control law for each position variable is impossible. Thus, SMC alone is not suitable for position trajectory tracking control. This part presents a robust backstepping sliding mode controller based on the aforementioned regular sliding mode control law. The forthcoming controller is a combination of sliding mode control technique and backstepping control technique. On the one hand, backstepping control starts from Cartesian position trajectory tracking, utilizes virtual controls, stabilizes each subsystem and stops when the desired inner Euler angles are obtained. On the other hand, we apply the desired Euler angle commands to the attitude subsystem and

obtain the final sliding mode control law. Thus, the proposed controller has advantages in both dealing with nonlinear system (a merit of backstepping control) and possessing robustness against model uncertainties and disturbances (the feature and advantage of sliding mode control). Our control objective with the proposed controller is to ensure that the position trajectory  $\{x(t), y(t), z(t), \psi(t)\}$  converges to the reference trajectory  $\{x_r(t), y_r(t), z_r(t), \psi_r(t)\}$ . The control strategy adopted in this section is that the inner loop ( $\phi(t), \theta(t), \psi(t)$ ) is controlled by sliding mode laws, and a fast convergence is ensured to the desired trajectory  $\{\phi_d(t), \theta_d(t), \psi_d(t)\}$  while the outer loop ( $x(t), y(t), z(t)$ ) is stabilized progressively until the ultimate backstepping control laws are obtained.

The backstepping sliding mode control design of position trajectory tracking control for system (9) starts with position  $x$  in the inertia frame. A sliding mode manifold  $s_{p1} = v_1 - x_8$  is first introduced, where  $v_1$  is a virtual control to be determined. The Lyapunov candidate chosen for this step is given by

$$V_1 = \frac{1}{2}z_7^2 \quad (20)$$

where  $z_7 = x_{7r} - x_7$  is the tracking error between reference position  $x_r$  and actual position  $x$ . The derivative of (20) with respect to time is given by

$$\dot{V}_1 = z_7 \dot{z}_7 = z_7(\dot{x}_{7r} - \dot{x}_8) = z_7(\dot{x}_{7r} - v_1 + s_{p1}) \quad (21)$$

To stabilize  $V_1$ , the virtual control  $v_1$  is designed as  $v_1 = \dot{x}_{7r} + \rho_1 z_7$ , where  $\rho_1$  is a positive constant. By substituting it into (21), the following equation can be obtained

$$\dot{V}_1 = z_7 \dot{z}_7 = z_7(-\rho_1 z_7 + s_{p1}) = -\rho_1 z_7^2 + z_7 s_{p1} \quad (22)$$

After  $v_1$  is designed, the state error dynamics of  $z_7$  can be obtained as follows:

$$\dot{z}_7 = -\rho_1 z_7 + s_{p1} \quad (23)$$

A second Lyapunov function involving state error  $z_7$  is considered as follows:

$$V_2 = \frac{1}{2}(z_1^2 + s_{p1}^2) \quad (24)$$

and a second virtual control  $v_2 = (C_{x_1} S_{x_3} C_{x_5} + S_{x_1} S_{x_5})U_4$  is introduced. Up to now, as a control input,  $U_4$  is still the one to be determined. The derivative of (24) with respect to time is then given by

$$\begin{aligned} \dot{V}_2 &= z_7 \dot{z}_7 + s_{p1} \dot{s}_{p1} \\ &= z_7(-\rho_1 z_7 + s_{p1}) + s_{p1}(\dot{v}_1 - \dot{x}_8) \\ &= -\rho_1 z_7^2 + s_{p1}(z_7 + \dot{v}_1 - (C_{x_1} S_{x_3} C_{x_5} + S_{x_1} S_{x_5})U_4 + a_9 x_8) \\ &= -\rho_1 z_7^2 + s_{p1}(\dot{v}_1 + z_7 + a_9 x_8 - v_2) \end{aligned} \quad (25)$$

The stabilization of  $V_2$  can be obtained by designing the virtual control as  $v_2 = \dot{v}_1 + z_7 + a_9 x_8 + \rho_2 s_{p1}$ . By substituting it into (25), the following inequality can be obtained.

$$\begin{aligned} \dot{V}_2 &= -\rho_1 z_7^2 + s_{p1}(\dot{v}_1 + z_7 + a_9 x_8 - v_2) \\ &= -\rho_1 z_7^2 - \rho_2 s_{p1}^2 \\ &\leq 0 \end{aligned} \quad (26)$$

where  $\rho_2$  is a positive constant.

For the trajectory tracking control of  $x_9$  along with  $x_{9r}$ , a sliding manifold  $s_{p2} = v_3 - x_{10}$  is introduced, where,  $v_3$  is a virtual control to be determined to guarantee the tracking capability of  $x_9$  along with  $x_{9r}$ . A Lyapunov candidate in this step is considered as

$$V_3 = \frac{1}{2}z_9^2 \quad (27)$$

where  $z_9 = x_{9r} - x_9$  is the tracking error between reference position  $y_r$  and actual position  $y$ . The derivative of  $V_3$  with respect to time is given by

$$\dot{V}_3 = z_9 \dot{z}_9 = z_9(\dot{x}_{9r} - \dot{x}_{10}) = z_9(\dot{x}_{9r} - v_3 + s_{p2}) \quad (28)$$

To stabilize  $V_3$ , the virtual control  $v_3$  is designed as  $v_3 = \dot{x}_{9r} + \rho_3 z_9$ , where  $\rho_3$  is a positive constant. By substituting  $v_3$  into (28), the following equation can be obtained:

$$\dot{V}_3 = z_9(\dot{x}_{9r} - v_3 + s_{p2}) = -\rho_3 z_9^2 + z_9 s_{p2} \quad (29)$$

Similar to (23), the state error dynamics is obtained as follows:

$$\dot{z}_9 = -\rho_3 z_9 + s_{p2} \quad (30)$$

A Lyapunov function involving the state error  $z_9$  is chosen as

$$V_4 = \frac{1}{2}(z_7^2 + s_{p1}^2 + z_9^2 + s_{p2}^2) \quad (31)$$

and a virtual control  $v_4 = (C_{x_1} S_{x_3} S_{x_5} - S_{x_1} S_{x_5})U_4$  is introduced.

Until now,  $U_4$  is undetermined and need to be evaluated. The derivative of (31) with respect to time is

$$\begin{aligned} \dot{V}_4 &= z_7 \dot{z}_7 + s_{p1} \dot{s}_{p1} + z_9 \dot{z}_9 + s_{p2} \dot{s}_{p2} \\ &= \dot{V}_2 + z_9(-\rho_3 z_9 + s_{p2}) + s_{p2}(\dot{v}_3 - \dot{x}_{10}) \\ &= \dot{V}_2 - \rho_3 z_9^2 + s_{p2}(z_9 + \dot{v}_3 - (C_{x_1} S_{x_3} S_{x_5} - S_{x_1} S_{x_5})U_4 + a_{10} x_{10}) \\ &= \dot{V}_2 - \rho_3 z_9^2 + s_{p2}(z_9 + \dot{v}_3 - v_4 + a_{10} x_{10}) \end{aligned} \quad (32)$$

The stabilization of  $V_4$  can be obtained by setting the virtual control as  $v_4 = \dot{v}_3 + z_9 + a_{10} x_{10} + \rho_4 s_{p2}$ , where  $\rho_4$  is a positive constant. Then, (32) becomes

$$\begin{aligned} \dot{V}_4 &= \dot{V}_2 - \rho_3 z_9^2 + s_{p2}(z_9 + \dot{v}_3 - v_4 + a_{10} x_{10}) \\ &= \dot{V}_2 - \rho_3 z_9^2 - \rho_4 s_{p2}^2 \leq 0 \end{aligned} \quad (33)$$

The backstepping procedure continues and a new sliding mode manifold is introduced as  $s_{p3} = v_5 - x_{12}$ , where  $v_5$  is a virtual to be determined. At this step, the error measure is chosen as

$$V_5 = \frac{1}{2}z_{11}^2 \quad (34)$$

where  $z_{11} = x_{11r} - x_{11}$  is tracking error between reference position  $z_r$  and actual position  $z$ . The derivative of (34) with respect to time is given by

$$\dot{V}_5 = z_{11} \dot{z}_{11} = z_{11}(\dot{x}_{11r} - \dot{x}_{12}) = z_{11}(\dot{x}_{11r} - v_5 + s_{p3}) \quad (35)$$

To stabilize  $V_5$ , the virtual control  $v_5$  is designed as  $v_5 = \dot{x}_{11r} + \rho_5 z_{11}$ , where  $\rho_5$  is a positive constant. By substituting  $v_5$  into (35), the following equation can be obtained:

$$\dot{V}_5 = z_{11}(\dot{x}_{11r} - v_5 + s_{p3}) = -\rho_5 z_{11}^2 + z_{11} s_{p3} \quad (36)$$

Similar to (23), the tracking error dynamics are then obtained as follows:

$$\dot{z}_{11} = -\rho_5 z_{11} + s_{p3} \quad (37)$$

A Lyapunov function involving state error  $z_{11}$  is considered as

$$V_6 = \frac{1}{2}(z_7^2 + s_{p1}^2 + z_9^2 + s_{p2}^2 + z_{11}^2 + s_{p3}^2) \quad (38)$$

and a virtual control  $v_6 = (C_{x_1} C_{x_3})U_4$  is introduced where  $U_4$  is unknown till now. The derivative of (38) with respect to time is given by

$$\begin{aligned} \dot{V}_6 &= z_7 \dot{z}_7 + s_{p1} \dot{s}_{p1} + z_9 \dot{z}_9 + s_{p2} \dot{s}_{p2} + z_{11} \dot{z}_{11} + s_{p3} \dot{s}_{p3} \\ &= \dot{V}_4 + z_{11}(-\rho_5 z_{11} + s_{p3}) + s_{p3}(\dot{v}_5 - \dot{x}_{12}) \\ &= \dot{V}_4 - \rho_5 z_{11}^2 + s_{p3}(z_{11} + \dot{v}_5 + g + a_{12} x_{12} - v_6) \end{aligned} \quad (39)$$

The stabilization of  $V_6$  can be obtained by designing the virtual control as  $v_6 = \dot{v}_5 + z_{11} + g + a_{12} x_{12} + \rho_6 s_{p3}$ , where  $\rho_6$  is a positive constant. Then, (39) becomes

$$\begin{aligned} \dot{V}_6 &= \dot{V}_4 + \dot{V}_5 + s_{p3}(z_{11} + \dot{v}_5 + g + a_{12} x_{12} - v_6) \\ &= \dot{V}_4 - \rho_5 z_{11}^2 - \rho_6 s_{p3}^2 \leq 0 \end{aligned} \quad (40)$$

The combination of (38) and (40) indicates that the stability of the overall system is guaranteed and position trajectory tracking capability is ensured. Ultimate control laws are finally designed from the virtual controls introduced in these backstepping steps.

By associating  $v_2$ ,  $v_4$  and  $v_6$ , a group of virtual controls

is obtained as:

$$\begin{aligned} v_2 &= (C_{x_1} S_{x_3} C_{x_5} + S_{x_1} S_{x_3}) U_4 \\ v_4 &= (C_{x_1} S_{x_3} S_{x_5} - S_{x_1} S_{x_3}) U_4 \\ v_6 &= (C_{x_1} C_{x_3}) U_4 \end{aligned} \quad (41)$$

**Remark 3.1:**  $v_2$ ,  $v_4$ , and  $v_6$  are combinations of available terms either by measuring or given directly as aforementioned. Thus, they can be regarded as known in the controlled system (11) and the control input  $U_4$  can be solved.

Notably, equation (41) has 4 degrees of freedom, namely,  $x_1$ ,  $x_3$ ,  $x_5$  and  $U_4$ . We also point out that a reference trajectory for yaw angle, i.e.  $x_{5r}$ , is usually given in advance as an extra reference signal and the corresponding SMC law  $U_3$  is directly designed to ensure the rapid convergence of  $x_5$  to  $x_{5r}$ . Thus,  $x_5$  is regarded as known and can be replaced by  $x_{5r}$  in this case, and the degrees of freedom in (41) is reduced such that  $x_1$ ,  $x_3$ , and  $U_4$  can be solved. The solutions are as follows:

$$\begin{cases} x_{1d} = \arctan\left(\frac{av_6 - dv_4}{av_6\sqrt{c^2 + d^2}}\right) \\ x_{3d} = \arctan\left(\frac{c}{d}\right) \\ U_4 = \frac{a^2(c^2 + d^2)v_6^2 + (av_6 - dv_4)^2}{ad} \end{cases} \quad (42)$$

where  $x_{1d}$  and  $x_{3d}$  are desired the roll angle trajectory and pitch angle trajectory for the inner loop, and  $U_4$  is part of the ultimate control laws, and  $a = \cos(x_{5r})$ ,  $b = \sin(x_{5r})$ ,  $c = (v_2 + v_4)/v_6$  and  $d = a + b$ .

**Remark 3.2:** According to Remark 2.1, it is trivial to point out that solutions in (42) are meaningful. In particular, the denominator of  $U_4$  will not cause singularity since the yaw angle is limited to  $(-\pi/2, \pi/2)$ .

By taking  $\{x_{1d}, x_{3d}, x_{5r}\}$  as desired attitude trajectory, the ultimate control laws can be obtained by applying the preceding SMC law design procedure and are presented as follows:

$$\begin{cases} U_1 = \ddot{x}_{1d} + c_1\dot{z}_1 - (a_1x_4x_6 + a_2\varpi x_4 - a_3x_2) + k_1\text{sign}(s_1(z_1)) \\ U_2 = \ddot{x}_{3d} + c_2\dot{z}_3 - (a_4x_2x_6 + a_5\varpi x_2 - a_6x_4) + k_2\text{sign}(s_2(z_3)) \\ U_3 = \ddot{x}_{5r} + c_3\dot{z}_5 - (a_7x_2x_4 - a_8x_6) + k_3\text{sign}(s_3(z_5)) \end{cases} \quad (43)$$

where  $s_i(z_{2i-1}) = \dot{z}_{2i-1} + c_{2i}z_{2i-1}$ ,  $i = 1, 2, 3$  are the accordingly chosen sliding mode manifolds.

Ultimate control laws are finally obtained from (42) and (43) via the SMC and the backstepping control techniques. In addition, progressive stability analysis of the closed loop system guarantees the stability of the overall system (9) and ensures the position trajectory tracking capability of the proposed backstepping sliding mode controller.

#### IV. FAULT ESTIMATION FOR A QUADROTOR UAV

The presented backstepping sliding mode controller is a typical passive fault tolerant controller that handles disturbances and uncertainties well without knowing their information. This section investigates a method that estimates the fault information to complement our passive robust position trajectory tracking controller design. We

know that the vertical taking off and the hovering mode are the common and important flight modes for a quadrotor UAV. And any fault, especially actuator fault, during the taking off phase may cause catastrophic consequence and even crash, resulting in great loss. Thus, we give a practical linear fault estimation module for the simplified model in taking off and hovering mode instead of a nonlinear one for the complicated complete dynamics. The fault estimation scheme will work as an alarming module and give a rough estimate of the occurring faults, indicating the changing trend and amplitude. Based on these estimates, a fault tolerant control algorithm may be developed. An adaptive observer based fault estimation is adopted for its fast implementation.

A multiple-input and multiple-output (MIMO) linear dynamic system with faults can be written as follows:

$$\begin{cases} \dot{x}(t) = Ax(t) + Bu(t) + Ef(t) \\ y = Cx(t) \end{cases} \quad (44)$$

where  $x(t) \in \mathbb{R}^n$  is the state variable vector,  $u(t) \in \mathbb{R}^m$  is the input vector,  $y(t) \in \mathbb{R}^p$  is the output vector, and  $f(t) \in \mathbb{R}^r$  is the fault.  $A, B, C, D$  are known matrices of dimension  $n \times n, n \times m, p \times n, p \times m$ .  $E \in \mathbb{R}^{n \times r}$  is fault distribution matrix.

In this part, a fault estimation scheme for system (44) is considered with the following assumptions.

**Assumption 1:** The fault is bounded and the derivative of the fault is bounded, that is

$$\|f(t)\| \leq f_0, \quad \|\dot{f}(t)\| \leq f_1 \quad (45)$$

where  $f_0$  and  $f_1$  are two positive scalars.

**Remark 4.1:** Assumption 1 is a little strict in light of that no priori knowledge about actual faults signal bound is known in real application. Yet, we know for sure that the upper bound of fault signal exists.

The fault estimation observer constructed for the faulty system (44) is given by

$$\begin{cases} \hat{\dot{x}}(t) = A\hat{x}(t) + Bu(t) + E\hat{f}(t) - L(\hat{y}(t) - y(t)) \\ \hat{y}(t) = C\hat{x}(t) \end{cases} \quad (46)$$

where  $L \in \mathbb{R}^{n \times p}$  is the observer gain matrix.

Equation (47) is defined as follows:

$$\begin{cases} e_x(t) = \hat{x}(t) - x(t) \\ e_y(t) = \hat{y}(t) - y(t) \\ e_f(t) = \hat{f}(t) - f(t) \end{cases} \quad (47)$$

Considering the derivative of (47) and the error dynamics is obtained as follows:

$$\begin{cases} \dot{e}_x(t) = (A - LC)e_x(t) + Ee_f(t) \\ \dot{e}_y(t) = Ce_x(t) \end{cases} \quad (48)$$

The adaptive fault estimation algorithm is chosen as

$$\dot{\hat{f}}(t) = -\Gamma F e_y(t) \quad (49)$$

where  $\Gamma > 0$  is the learning rate matrix and  $F$  is the matrix to be designed. Notably, such algorithm is only suitable for constant faults, that is  $\dot{f}(t) = 0$ .

The derivative of  $e_f(t)$  with respect to time is given by

$$\dot{e}_f(t) = \dot{\hat{f}}(t) \quad (50)$$

**THEOREM 4.1:** If there exists a symmetric positive definite matrix  $P \in \mathbb{R}^{n \times n}$  and matrices  $Y \in \mathbb{R}^{n \times p}$ ,  $F \in \mathbb{R}^{r \times p}$  such that the following condition holds



$$\begin{cases} PA + A^T P - YC - C^T Y^T < 0 \\ E^T P = FC \end{cases} \quad (51)$$

The fault estimation algorithm (49) then guarantees that the state error  $e_x(t)$  and fault estimation error  $e_f(t)$  converge to zero asymptotically. In this paper, the observer gain matrix is chosen as

$$L = P^{-1}Y \quad (52)$$

The proof of Theorem 1 is as follow.

**Proof:** Considering the Lyapunov candidate

$$V(t) = e_x^T(t)Pe_x(t) + e_f^T(t)\Gamma^{-1}e_f(t) \quad (53)$$

The derivative of  $V(t)$  with respect to time is

$$\begin{aligned} \dot{V}(t) = & \dot{e}_x^T(t)Pe_x(t) + e_x^T(t)P\dot{e}_x(t) \\ & + \dot{e}_f^T(t)\Gamma^{-1}e_f(t) + e_f^T(t)\Gamma^{-1}\dot{e}_f(t) \end{aligned} \quad (54)$$

Then, with the aid of (47), we have

$$\begin{aligned} \dot{V}(t) = & e_x^T(t)(P(A-LC) + (A-LC)^T P)e_x(t) \\ & + 2e_x^T(t)PEe_f(t) - 2e_f^T(t)Fe_y(t) \\ = & e_x^T(t)(P(A-LC) + (A-LC)^T P)e_x(t) \\ & + 2e_x^T(t)PEe_f(t) - 2e_f^T(t)FCe_x(t) \end{aligned} \quad (55)$$

By combining (50) and (55), the following equation is obtained:

$$\dot{V}(t) = e_x^T(t)(P(A-LC) + (A-LC)^T P)e_x(t) \quad (56)$$

Thus, if condition  $PA + A^T P - YC - C^T Y^T < 0$  holds, the state estimation error  $e_x(t)$  and fault estimation error  $e_f(t)$  converge to zero asymptotically. The proof for Theorem 4.1 is completed. This fault estimation scheme is used in vertical taking off and landing phase to avoid crash landing by estimating the faults rapidly.

## V. SIMULATION

In this section, simulations are conducted to verify the effectiveness of the presented robust backstepping sliding mode controller and adaptive observer based fault estimation. Different scenarios are considered, including normal case, cases under model uncertainty and external disturbances, and even the case under Gaussian white noise, to demonstrate the robustness of the proposed controller. The nominal parameters for the quadrotor UAV adopted in this study are same as those described in Table 1 [2, 12].

### A. Simulation for Backstepping Sliding Mode Controller

The effectiveness of the proposed backstepping sliding mode controller is verified by comparing the position trajectory tracking performance in normal case and in cases with uncertain parameters and even with white Gaussian noise. The desired trajectory of the position and yaw angle  $\{x_r(t), y_r(t), z_r(t), \psi_r(t)\}$  in the simulation study is chosen as  $\{\cos(t), \sin(t), 0.5t, \sin(0.5t)\}$ . One thing should be noted is that the parameters in the robust controller design procedure ( $\rho_i, i=1,2,3,4,5,6$ ) should be tuned carefully given that these sliding mode surfaces. Therefore, these parameters affect the performance of the presented controller. To demonstrate the effectiveness of the developed robust controller, the following different scenarios are considered in the simulations and brief comparisons are conducted.

TABLE 1 QUADROTOR PARAMETERS

symbol	case 1	case 2	case 3	units
$m$	2	2	2	kg
$l$	0.2	0.2	0.2	m
$\kappa$	2.98	2.98	2.98	$10^{-6}Ns^2rad^{-2}$
$\tau$	1.14	1.14	1.14	$10^{-7}Ns^2rad^{-2}$
$d_\phi$	1.2	1.2	1.2	$10^{-2}Nsrad^{-1}$
$d_\theta, d_\psi$	1.2	1.2	1.2	$10^{-2}Nsrad^{-1}$
$J_x$	1.25	0.94	1.25	$Ns^2rad^{-1}$
$J_y$	1.25	0.94	1.25	$Ns^2rad^{-1}$
$J_z$	2.50	1.88	3.63	$Ns^2rad^{-1}$

### Scenario I: Normal case

In this case, we assume that the parameters of the quadrotor vehicle are normal, and neither the model uncertainty nor the disturbance is considered. Simulations for different initial states are conducted.

### Scenario II: Uncertainty (25% subtracted) in rotary inertia

In this case, we consider model uncertainty 25% subtracted in all three body-frame axes. We also conduct simulations with the proposed controller and the existing backstepping controller in the same situations to compare their robustness against model uncertainties.

### Scenario III: Uncertainty (45% added) in rotary inertia as well the presence of external disturbance

In this case, we give a consideration to both uncertainty in rotary inertia and external disturbance. The uncertainty in yaw axis is 45% more and the disturbance is band-limited white noise of which the power is 0.005 and the sample time is 0.01 and the amplitude is limited to 0.3.

### Scenario I: Trajectory tracking control in normal case

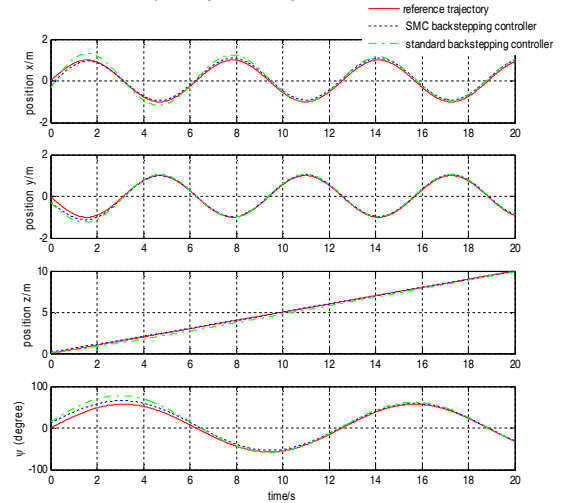


Fig.3 Normal case (red: reference trajectory, blue: SMC backstepping controller, green: standard backstepping controller)



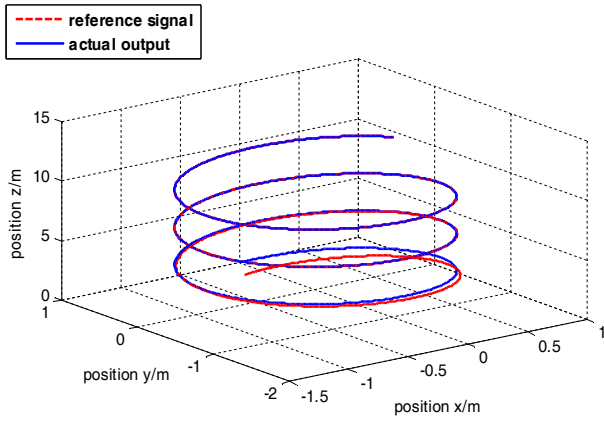


Fig. 4 Space diagram of position in normal case

Scenario II: Trajectory tracking control with uncertainty (25% subtracted) in rotary inertia

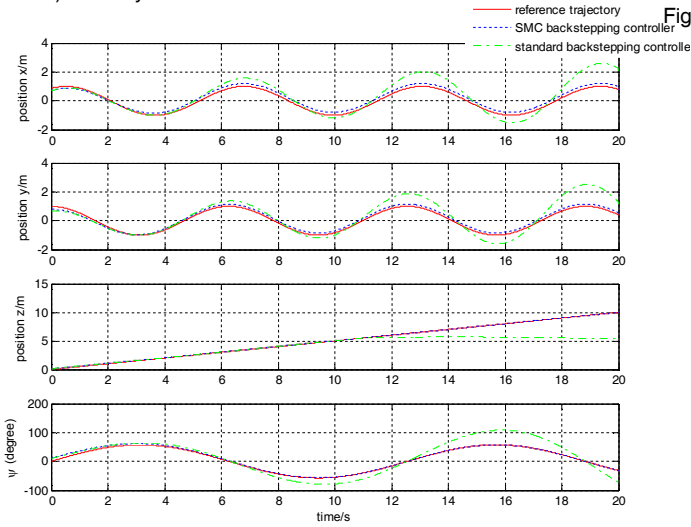


Fig. 5(a) Uncertainty 25% subtracted in all three axes

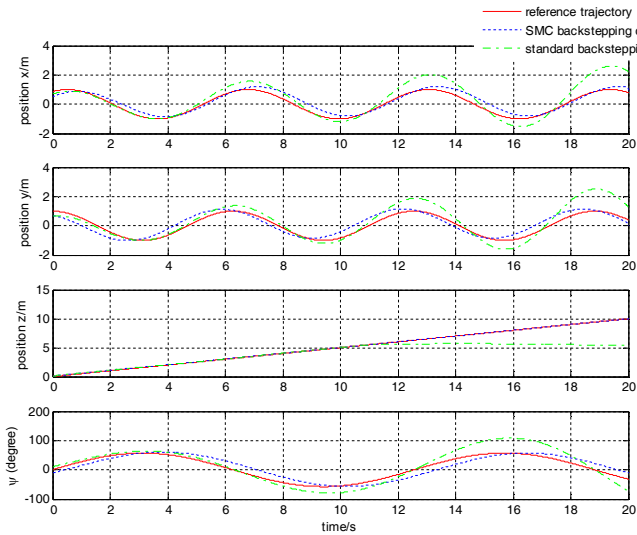


Fig. 5(b) Uncertainty 25% subtracted in all three axes with filtered control signals

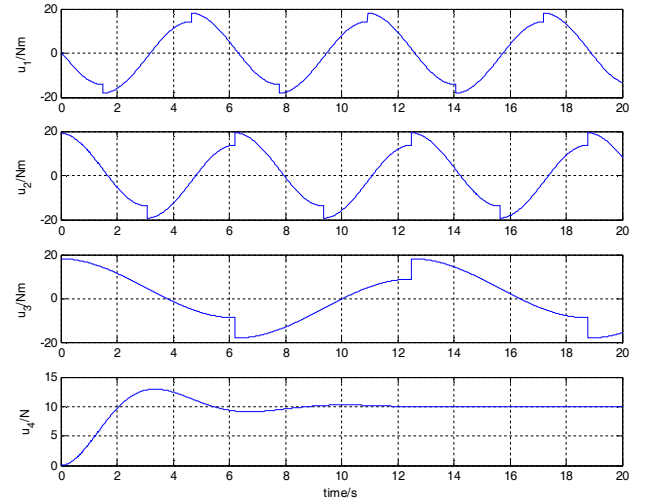


Fig. 5(c) Control signals

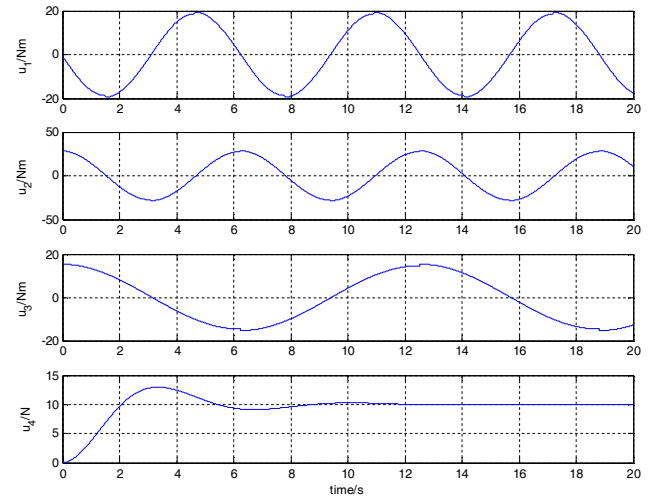


Fig. 5(d) Filtered control signals

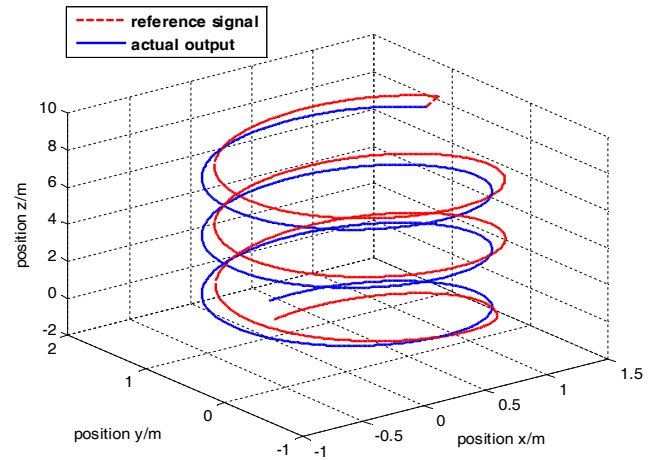


Fig. 6 Space diagram of position with uncertainty

Scenario III: Trajectory tracking control with uncertainty (45% added) and white noise

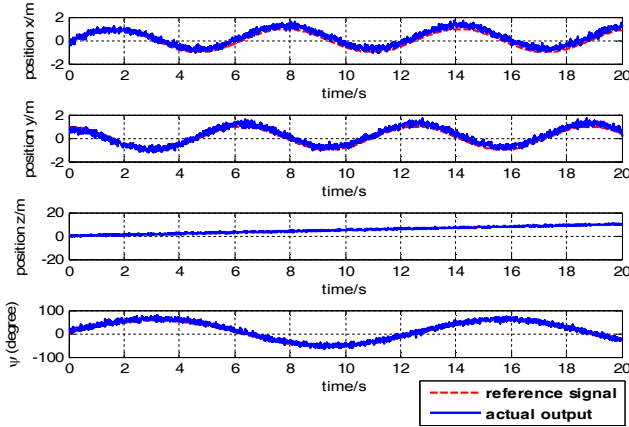


Fig. 7 Uncertainty 45% added in one rotating axis

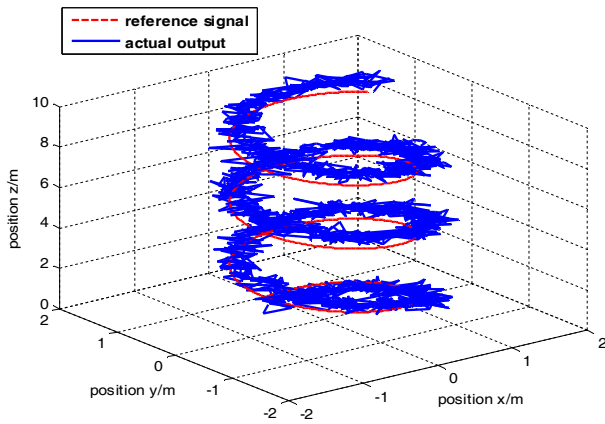


Fig. 8 Space diagram of position with uncertainty and white Gaussian noise

The position and yaw angle trajectories in Figs. 3 and 4 show satisfactory performances, that is, trajectory tracking is achieved in a few seconds. Moreover, we can see from Fig. 3 that both standard backstepping controller and the proposed SMC backstepping controller work well in a normal case in terms of convergence rapidity and tracking error. Fig. 5 (a)-(b) show the simulation results with standard and SMC backstepping controllers in an initial state different from normal case. Compared with Fig. 5 (a), Fig. 5 (b) shows a minor delay effect caused by the filtered control signal with a first order filter. Both 5(a) and 5(b) show that standard backstepping controller fails to handle such a case with uncertainty 25% subtracted in all three axes while the proposed controller succeeds in dealing with this case. Fig. 5 (c) and Fig. 5 (d) show the unfiltered control signals and the filtered ones by  $8/(s+8)$ . Figs. 7 and 8 illustrate the simulation results of the quadrotor system in scenarios which are under 45% uncertainty added in one of the three rotating axes in the presence of white Gaussian noise. From Fig. 7 we can see that the proposed controller possesses trajectory tracking capability even under circumstances in which uncertainty in rotary inertia is up to 45% along with white noise at the same time. From Fig. 3 to Fig. 8, we can conclude that the proposed controller is robust to uncertainty of different levels and even with white noise simultaneously.

The trajectory tracking performances in these simulations demonstrate the effectiveness of the robust controller against adverse condition.

### B. Simulation for Fault Estimation Scheme

The vertical take-off phase of a quadrotor UAV is considered to verify the effectiveness of the adaptive observer based fault estimation scheme. The dynamic model represented by (6) and (7) can be simplified in this case. The take-off manipulation is assumed to be obtained by increasing the rotor speeds with identical quantity which means  $\Omega_i, i=1,2,3,4$  have the identical quantity. Thus, the control inputs  $U_i, i=1,2,3$  in (11) are zeros.  $U_4$  is bound to be positive enough to overcome the mass of the quadrotor. Another assumption is that the initial states of the attitude angles and their derivatives with respect to time are all zeros, and this often makes sense. Under the preceding assumptions, the taking off dynamic model is reduced to

$$\begin{cases} \dot{x}_{11} = x_{12} \\ \dot{x}_{12} = -g + (C_{x1} C_{x3})U_4 - a_{11}x_{12} \end{cases} \quad (57)$$

which can be rewritten as a typical linear second-order system in the assumption that  $C_{x1} = C_{x3} = 1$ , which means the roll angle and the pitch angle are near zeros. The second order dynamics is

$$\ddot{z} = U_4 - g - d_z \dot{z} / m \quad (58)$$

where  $z = x_{11}$  is the state variable. Thus, the aforementioned adaptive fault estimation method can be applied to estimate the fault information in this dynamic system.

Denote  $z_1 = z$ ,  $\dot{z}_1 = \dot{z}$ ,  $U = U_4 - g$ , and consider possible actuator fault, the dynamic system becomes

$$\begin{aligned} \begin{bmatrix} \dot{z}_1 \\ \dot{z}_2 \end{bmatrix} &= \begin{bmatrix} 0 & 1 \\ -d_z/m & 0 \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u + \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ f_1(t) \end{bmatrix} \\ &= \begin{bmatrix} 0 & 1 \\ -0.005 & 0 \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} (u + f_1) \end{aligned} \quad (59)$$

$$w = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \end{bmatrix} \quad (60)$$

where  $w$  is output.

The parameters given in Theorem 1 can be obtained through LMI toolbox in MATLAB. The corresponding observer gain matrix  $L$  and the corresponding  $F$  is solved as

$$L = \begin{bmatrix} 0.5000 & -22.5519 \\ 23.5469 & 0.5000 \end{bmatrix}, F = \begin{bmatrix} 0 & 1 \end{bmatrix} \quad (61)$$

Simulations are conducted with these parameters and the results for fault estimation are shown in the following figures.

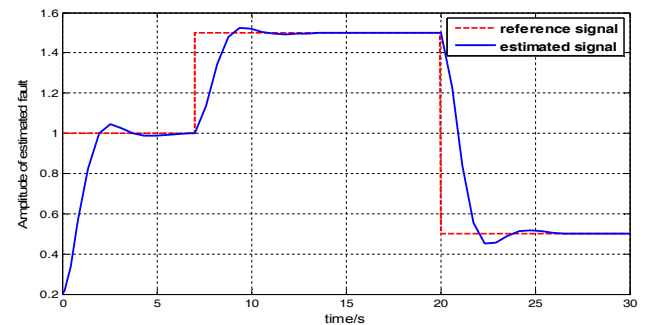


Fig. 9 Estimation of invariant fault

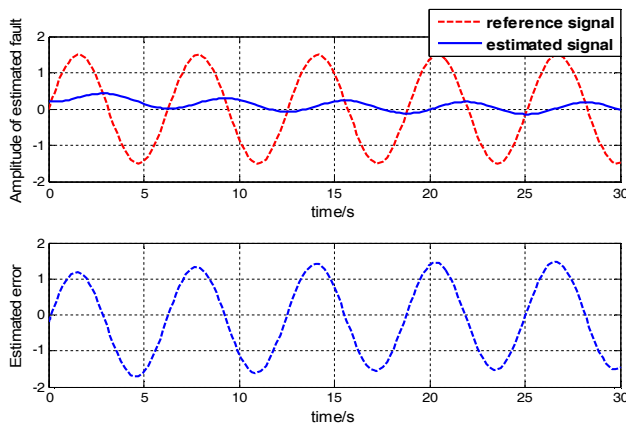


Fig. 10 Estimation of time-varying fault and estimation error

Fig. 9 shows the simulation result of employing adaptive observer to estimate the invariant fault signal

$$f_1(t) = \begin{cases} 1; & t < 7(\text{sec}) \\ 1.5; & t \geq 7(\text{sec}) \\ 0.5; & t > 20(\text{sec}) \end{cases}$$

It is clear that the adaptive observer can estimate the invariant fault signal satisfyingly. Fig. 10 shows the simulation result of the system with a fault as  $f(t) = 1.5\sin(t)$ , and the estimation error response illustrates that this fault estimation method fails to deal with time-varying faults.

## VI. CONCLUSIONS

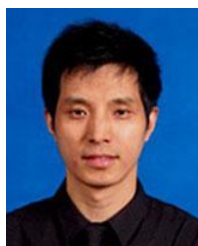
In this paper, we first present a regular sliding mode controller design for attitude control of a quadrotor UAV. On the basis of this design, we combine SMC technique and backstepping control technique to design a robust backstepping sliding mode controller for position and yaw angle control. Virtual control inputs are introduced in the robust controller design to guarantee the trajectory tracking performance. Sliding manifolds are designed to drive the actual variables to approximate the virtual ones. We present the stability analysis of the overall system through Lyapunov stability theory. Fault estimation is also considered in this study and is applied to the take-off phase of a quadrotor. The simulation results show that, robustness and improved tracking performance can be achieved with the proposed robust sliding mode controller. The simulation results also demonstrate the effectiveness of the presented fault estimation method. Our future works includes considering actuator fault and developing an active fault-tolerant controller based on the control design and fault estimation scheme proposed in this paper.

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