Active Fault-tolerant Control of a Quadrotor UAV Against Actuator Faults Based on Backstepping Technique and Adaptive Observer

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**ABSTRACT**

This paper addresses the stabilization problem of an underactuated quadrotor UAV system in presence of actuator faults. First, a dynamic model of the quadcopter was established using a Lagrange approach. High-order non-holonomic constraints as well as different physical phenomena that can influence the dynamics of the structure have been taken into account. Then, for actuator faults, a new active fault tolerance strategy based on a backtracking approach and an adaptive observer is developed. The simulation results obtained illustrate the ability of the proposed control strategy to maintain performance and preserve stability in the event of actuator failure.

**KEYWORDS**

Active fault tolerant control, quadrotor unmanned aerial vehicles, nonlinear dynamical model, actuator faults, backstepping approach, adaptive observer.

# Introduction

In light of increasing needs regarding automated systems availability, safety, and performance, it is necessary to develop a diagnostic module to detect faults that may damage these systems operations and identify their origin or source. Despite the tangible progress made, researchers must still deal with severe difficulties in controlling such systems, particularly in the presence of faults. Especially in the case underactuated systems like unmanned aerial vehicles (UAV).

Quadcopters have been the subject of several studies in particular in the field of diagnosis and fault tolerance. (Freddi, Longhi and Monteriù 2010; Ouadine, et al. 2020; Ren 2020; Xulin and Yuying 2018).

The work in (Avram, Zhang and Muse 2018) presents a nonlinear robust adaptive fault-tolerant altitude and attitude tracking scheme to accommodate actuator faults in a quadrotor. In (Xulin and Yuying 2018), the authors present a fuzzy active disturbance rejection control method for controlling a quadrotor UAV with actuator faults. An active fault-tolerant tracking control system approach for actuator faults on a quadrotor was discussed in (Zhong, et al. 2019). A fault-tolerant controller was designed on basis of adaptive estimation for actuator faults in (Hasanshahi, Ahmadi and Amjadifard 2019). In (Hong-Jun , et al. 2019), the authors present the diagnosis and compensation of sensors and actuators faults in a quadrotor UAV based on a nonlinear high-gain observer. Other strategies are proposed in (Lien, Chao-Chung and Yi-Hsuan 2020; Ren 2020; Yujiang , et al. 2018).

In the field of active fault-tolerant control (AFTC), observer-based reconstruction and fault estimation (FRE) has gained increased interest in the last two decades. Its advantage is that it can estimate the faults without going through the residual generation phase. Various observer-based FRE design techniques have been presented in the literature, mainly based on sliding-mode observers, observers for singular systems, and adaptive observers (Jiang and Yu 2012). When faults are modelled in terms of parameter changes, adaptive observers can be used to estimate these faults.

This article presents a new active FTC technique on a quadrotor in the presence of actuator faults. It is based on a joint use of an adaptive observer for fault reconstruction and estimation and a backstepping approach for system control. Compared to previous work on the active FTC of a quadcopter UAV, in our work we have not neglected the non-linearity of the dynamic model of the quadcopter and the high-order non-holonomic constraints. It was used an adaptive observer proposed in (Oucief, Tadjine and Labiod 2016).

In the first section, the dynamic modelling of the quadcopter is carried out. To detect defects, an adaptive observer was developed to estimate the size of faults in the second section. Then, in the third section, a robust control strategy with actuator faults is established based on the backstepping technique and taking in to account the dynamic of faults. Finally, in the last section, simulations on MATLAB were carried out to validate the synthesized control laws. The results were conclusive in the presence of faults in the actuators.

# Quadrotor Modelling

The aerial robot under study consists of a rigid cross frame coupled with four propellers, as illustrated in figure 1. The forward/ backward left/ right and the yaw movements are generated by a differential control strategy of the thrust delivered by each rotor. The up-down motion increases or reduces the overall thrust while keeping an equal individual thrust. To minimize the yaw drift induced by the responsive torques, the quadrotor aircraft is designed so that the set of rotors (right-left) spins clockwise and the set of rotors (front-rear) spins counter-clockwise.

Let E (O, X, Y, Z) designate an inertial frame, and B (o, x, y, z) designate a frame permanently coupled to the quadrotor, as illustrated in figure 1.

The absolute location is denoted by the three coordinates () and its attitude by the three Euler’s angles () respectively called Roll angle ( rotation around x-axis), Pitch angle ( rotation around y-axis) and Yaw angle ( rotation around z-axis).

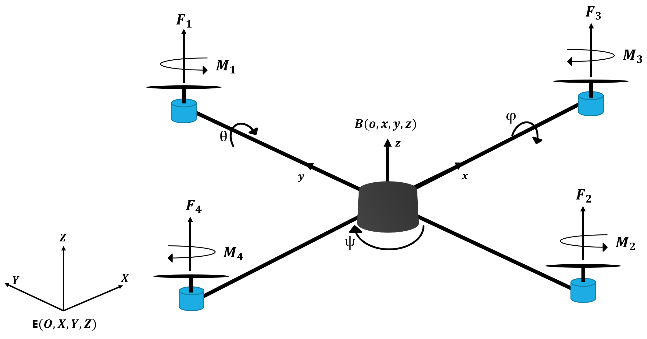


Figure 1. Quadrotor configuration

Literally, by using formalism of Newton-Euler, the quadrotor complete model (position and orientation dynamic) is provided as in (Bouadi, Bouchoucha and Tadjine 2007) by:

|  |  |  |
| --- | --- | --- |
|  |  | (1a) |
|  |  | (1b) |
|  |  | (1c) |
|  |  | (1d) |
|  |  | (1e) |
|  |  | (1f) |

Where:

* Where C and S indicate the trigonometrical functions *cos* and *sin* respectively.
* is the total mass of the structure.
* and are lift and drag coefficients respectively.
* , and are constants inertia.
* , and are the translation drag coefficients.
* , and are the aerodynamic friction coefficients around (x, y, z).
* is the distance between the quadrotor centre of mass and the rotation axis of propeller.
* is the rotor inertia.

The quadrotor object of our study is (Draganfly IV: Manufactured by Draganfly Innovations). Parameter identification is studied in (Derafa, Madani and Benallegue 2006) (Table)

|  |  |
| --- | --- |
|  |  |
|  |  |
|  |  |
|  |  |
|  |  |
|  |  |

, , and are the control inputs of the system which are written according to the angular velocities of the four rotors as follows:

()

()

From the equations of the translation dynamics in (1) we can deduce the expressions of the high-order nonholonomic constraints:

|  |  |
| --- | --- |
|  | (4a) |
|  | (4b) |

# Nonlinear adaptive observer design

## State-space model

The complete model resulting by adding the actuator faults in the model (1) can be written in the state-space form:

|  |  |
| --- | --- |
|  | (5) |

With:

is the state vector of the system, such as:

|  |  |  |
| --- | --- | --- |
|  |  | (6) |

, , and are known constant matrices.

, is the resultant vector of actuator faults related to quadrotor motions

represent the actuators faults vector, with.

is a known function matrix which may depend nonlinearly on .

is the input control vector,

is the output vector giving by .

are known nonlinear functions vectors.

Throughout this article, system model (5) satisfies the following conditions (section) :

**C0:** The pair (, ) is observable;

**C1:** The vector function is continuous in its variables;

**C2:** and satisfy the Lipschitz property with respect to *x:* there exist positive constants and such that:

|  |  |  |
| --- | --- | --- |
|  |  | (7a) |
|  |  | (7b) |

**C3:** The fault vector is piecewise constant and bounded in the following sense:

|  |  |  |
| --- | --- | --- |
|  |  | (8) |

Where is a known constant vector and is a known positive constant.

## The adaptive state observer

The typical form of the adaptive state observer dealing with the class of nonlinear systems (5) is given by the following equations (Cho and Raramani 1995; That and Ding 2014):

|  |  |
| --- | --- |
|  | (9) |

Where:

and are the state and fault estimates.

*L*is the observer gain

is a positive constant and

*F* is a matrix to be designed.

Under conditions C0, C1, C2 and C3 the state estimate converges to the actual state and converges to , if there exist a symmetric positive definite matrix *P*and a matrix *F* such that

|  |  |  |
| --- | --- | --- |
|  |  | (10) |

Equality (10) is known as the observer matching condition (Floquet, Edwards and Spurgeon 2007)(Corless and Tu 1998; Raoufi , Jose Marquez and Solo 2010), it hold if and only if:

|  |  |  |
| --- | --- | --- |
|  |  | (11) |

Unfortunately, the observer matching condition (10) is not satisfied for our system (5) and therefore we cannot use the adaptive observer of the form (9) to estimate .

In , authors have proposed a new methodology for an adaptive state observer design for a certain class of non-linear systems. This observer employs the nonlinear system model described by equation (5).

For developing the considered adaptive observer, in addition to conditions C0, C1, C2 and C3 the system model (5) has to satisfy the following conditions:

**C4:** The matrices , , and satisfy:

|  |  |  |
| --- | --- | --- |
|  |  | (12a) |
|  |  | (12b) |
|  |  | (12c) |

**C5:** The first derivative in time of is continuous and bounded provided that is bounded.

To satisfies the conditions C4, the state space (1) is rearranged. Considering the actuator faults, we obtain:

|  |  |  |
| --- | --- | --- |
|  |  | (13) |
|  |  |
|  |  |
|  |  |
|  |  |
|  |  |
|  |  |
|  |  |
|  |  |
|  |  |
|  |  |
|  |  |

With

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
|  |  |  |  |  |
|  |  |  |  |  |
|  |  |  |  |  |

When these conditions are satisfied, a stable observer for the system (5) has the form

|  |  |  |
| --- | --- | --- |
|  |  | (14a) |
|  |  | (14b) |
|  |  | (14c) |

and is the unknown parameter vector.

and are constant matrices to be designed.

is the learning rate matrix.

A sufficient condition for the asymptotic stability of the adaptive state observer is described in the following theorem **.**

**Theorem 1.** Under conditions C0, C1, C2, C3, C4 and C5, the estimate of the state converges to the real state *x* asymptotically while converges to if there are positive real constants and and matrices , and such that:

|  |  |
| --- | --- |
|  | (15a) |
|  | (15b) |

For our system, conditions C0 to C5 are satisfied, and as a consequence, the adaptation law (14) is feasible.

## Observer design

The objective is to synthesize an adaptive state observer corresponding to the model quadrotor given by (13) which can be described in the form of system given by (5).

Where:

(16a)

(16b)

(16c)

We evaluate the Lipschitz constants of the individual rows of :

|  |  |  |
| --- | --- | --- |
|  |  | (17) |

We found that we can take .

is locally Lipschitz, which means that its Lipschitz constant depends on the region where the system operates. Suppose that the control input is chosen such that it keeps the state bounded in the set

|  |  |
| --- | --- |
|  | (18) |

Then can be extended into the bounded function which coincides with in , and thus the system becomes globally Lipschitz (Farza, et al. 2009).

The computation of is given by:

|  |  |  |
| --- | --- | --- |
|  |  | (19) |

For the calculation of , we assume that the maximum value that can be reached by the faults represents 50% of the setpoint indicated in (18). The computation of is given by:

|  |  |  |
| --- | --- | --- |
|  |  | (20) |

To find the observer gains, we can turn (15a) and (15b) into an LMI optimization problem.

Let and . Using the Schur complement (Boyd, El Ghaoui and Feron 1994) for inequality (15a) we get:

|  |  |  |
| --- | --- | --- |
|  |  | (21) |

With:

.

Also, by using the same idea used in (Corless and Tu 1998) we can turn the problem of the resolution of equality (15b) into the following LMI optimization problem:

Minimize subject to:

|  |  |  |
| --- | --- | --- |
|  |  | (22) |

Where , and*δ* is a positive scalar.

Therefore, computing ,, and involves solving LMIs (21) and (22) for , . LMIs were solved using MATLAB by a convex optimization tool ( (Grant and Boyd 2014)). We get the following results:

|  |  |  |
| --- | --- | --- |
|  | |  |
|  | (23) | |
|  |
|  |
|  |
|  |

The unknown parameter vector estimates and conforming to (14) is then

|  |  |  |
| --- | --- | --- |
|  |  | (24a) |
|  |  | (24b) |

Where:

is the tracking-errors of .

and .

# Control Strategy of Quadrotor with Actuator Faults

## Control Strategy

Generally, a fault-tolerant system is composed of two cascaded modules. The first one is a monitoring module which is used to detect faults, and diagnose their location and significance in a system. The second is a recovery module taking necessary actions so that the faulty system can achieve the control objectives almost at any time (Jain, J. Yamé and Sauter 2018).

In our case, an adaptive observer is used like a monitoring module (Figure 2) and the recovery module is based on the Backstepping approach.

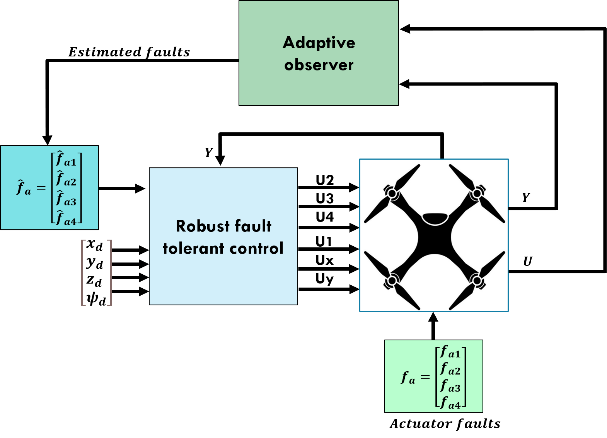


Figure 2. Fault-tolerant control system architecture

The following assumptions are needed for the analysis:

**Assumption 1.** The resultant of actuator faults related to quadrotor motions are slowly varying in time and bounded, as follows:

|  |  |  |
| --- | --- | --- |
|  |  | (25) |

Where {,,,} are positive constants.

**Assumption 2.** The unknown’s parts including the resultants of actuator faults related to the quadrotor motions are also bounded:

|  |  |  |
| --- | --- | --- |
|  |  | (26) |

Where {,,,} are positive constants.

The proposed control approach is based on two loops (internal and external loops). The internal loop has four control laws (roll, pitch, yaw, altitude). The external loop has two control laws of coordinates and .

The external control loop produces the desired roll () and pitch () via the corrective block: Equation (4). The corrective block has as a goal to correct the rotation of the roll and pitch based on the desired yaw (). The synoptic scheme (Figure 3) below illustrates this control strategy:

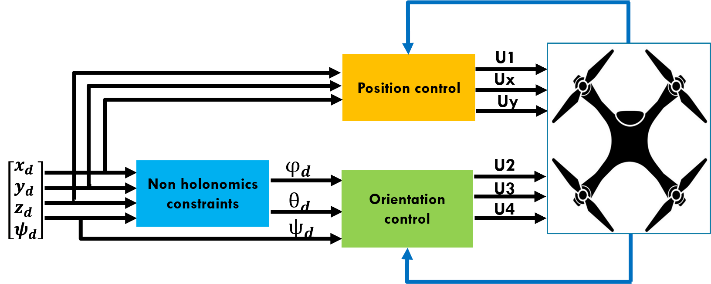


Figure 3. Synoptic scheme of the control strategy

## Control laws

Based on the backstepping technique, an iterative algorithm is used to synthesize the control laws forcing the system to follow the desired path in presence of actuator failures, we summarize all stages of calculation concerning the tracking errors and Lyapunov functions in the following way:

(27)

The related Lyapunov functions are provided by:

|  |  |  |
| --- | --- | --- |
|  |  | (28) |

The synthesized stabilizing control laws are as described in the following:

|  |  |
| --- | --- |
|  | (29a) |
|  | (29b) |
|  | (29c) |
|  | (29d) |
|  | (29e) |
|  | (29f) |

Where .

***Proof.*** Considering the first subsystem:

|  |  |  |
| --- | --- | --- |
|  |  | (30) |

The corresponding reduced order observer is:

|  |  |  |
| --- | --- | --- |
|  |  | (31) |

The calculation of the command is done in two steps.

***Step 1:*** For the first step we consider the first tracking-error given by

|  |  |  |
| --- | --- | --- |
|  |  | (32) |

Let the first Lyapunov function candidate:

|  |  |  |
| --- | --- | --- |
|  |  | (33) |

The time derivative of (33) is given by:

|  |  |  |
| --- | --- | --- |
|  |  | (34) |

According to Lyapunov’s theorem, the stabilization of can be obtained by introducing a new virtual control which represent the desired value of :

|  |  |  |
| --- | --- | --- |
|  |  | (35) |

The equation (34) is then:

|  |  |  |
| --- | --- | --- |
|  |  | (36) |

***Step 2:*** As is not a real command, we define the following tracking-error variable between the state variable and its desired value

|  |  |  |
| --- | --- | --- |
|  |  | (37) |

Let and

Then, from (5), (14) and (25) we get:

|  |  |  |
| --- | --- | --- |
|  |  | (38a) |
|  |  | (38b) |

Where and .

The augmented Lyapunov function is given by:

|  |  |  |
| --- | --- | --- |
|  |  | (39) |
|  |  |

The time derivative of is given by:

|  |  |  |
| --- | --- | --- |
|  |  | (40) |

The time derivative of is given by:

|  |  |  |
| --- | --- | --- |
|  |  | (41) |

Substituting (38a) and (38b) into (41), we obtain

|  |  |
| --- | --- |
|  | (42) |

Using (15b), we get:

(43)

With , the Lipschitz conditions (7a) to (7b) and inequality (8), we obtain the following inequalities

|  |  |  |
| --- | --- | --- |
|  |  | (44) |
|  |  |

And

|  |  |  |
| --- | --- | --- |
|  |  | (45) |
|  |  |

Where and are positive constants. Substituting (44) and (45) into (43), we obtain:

|  |  |  |
| --- | --- | --- |
|  |  | (46) |

With:

Substituting (46) into (40) and replacing by his estimate , (40) yields

|  |  |  |
| --- | --- | --- |
|  |  | (47) |
|  |  |
|  |  |

Where  ( is the minimum eigenvalue of ).

The stabilization of (, ) can be obtained by introducing the input control :

(48)

Finally, the equation (47) becomes

|  |  |  |
| --- | --- | --- |
|  |  | (49) |

The same steps are followed to extract , , , and .

# Simulation Results

To evaluate the performance of the controller proposed in this work, we executed simulations in MATLAB. We carried out the simulations in two stages: A simulation with a healthy system and another simulation where we introduced faults on the actuators.

Results of simulation without faults are shown in Figure 4, Figure 5.



Figure 4. Trajectories along roll (𝜑), pitch (θ), yaw angle (ψ), and attitude Z (Test 1)

From this simulation (Figure 4), it can be seen that the true and estimated state by using this adaptive observer are matched perfectly.



Figure 5. Tracking errors: roll (𝜑), pitch (θ), yaw angle (ψ), and attitude Z (Test 1)

The estimation errors (Figure 5) quickly converge to 0 and remain below , which clearly illustrates the good performance and robustness of the control system with the backstepping approach

In the second simulation, four actuator faults related to roll, pitch, yaw and altitude (φ, θ, ψ, z) commands are introduced. The simulation results are presented in Figure 6, 7, 8, 9,10 and 12.



Figure 6. Fault estimation

According to Figure 6, there is very excellent estimation of the actuator faults. The estimates of and converges rapidly to the real values. Meanwhile, the estimate of converge rapidly after a response time of around .

The mean of the estimation error of and respectively are , , and . Therefore, the proposed observer can give a fast and accurate fault estimation.

The state estimates in faulty case are shown in Figure 7.



Figure 7. Trajectories along roll (𝜑), pitch (θ), yaw angle (ψ), and attitude Z axis (with actuators faults)

As shown in Fgure. 7, trajectory tracking is very excellent even after the appearance of actuator faults. Small transient variations in the movements of roll, pitch, yaw and altitude at the instants of appearance of the faults at the instants:10s, 20s, 30s and 40s.

Figure 8. Tracking errors of roll (𝜑), pitch (θ), yaw angle (ψ), and Attitude Z (with actuators faults)

As shown in Figure 8, the roll (φ), pitch (θ) and yaw (ψ) estimation errors remain close to zero (). For the attitude estimation error z remains below 0.2 m which represents 2% of the desired attitude. Despite the faults, we can conclude that the trajectory tracking of our system is assured.

Figure 9 and Figure 10 illustrate the inputs control , *,*  of our system. It is easy to notice the transient peaks in all controllers. Despite it, the stability of the closed-loop dynamics of the quadrotor is assured. Furthermore, we can observe input control signals provided by this control strategy are acceptable and physically realizable.



*Figure 9. Control inputs of actuators in normal case (Without faults)*



*Figure 10. Control inputs of actuators in faulty case (Actuator faults)*

To numerically evaluate the results obtained during the simulations, we will calculate two numerical criteria: the RMS error (Root Mean Square) and the average error of the desired coordinates. (Table 1).

Table 1. Numerical evaluation of the control strategy in Test 1 and Test 2

|  |  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- | --- |
|  | **RMS** | | | | **Error mean** | | | |
| **𝜑** | **θ** | **ψ** | **z** | **𝜑** | **θ** | **ψ** | **z** |
| **Test 1** |  |  |  |  |  |  |  |  |
| **Test 2** |  |  |  |  |  |  |  |  |

The quantitative analysis confirms that the proposed strategy not only ensures a satisfactory tracking performance of the state estimation but also preserves a low energy consumption with small control inputs.

Figure 11 and Figure 12 illustrate the 3D trajectory of the quadrotor aircraft throughout the flight. The simulation resultsindicate high performances and resilience towards stability and tracking even after the occurrence of actuator faults (Figure 12), which shows the efficacy of the control method suggested in this work.



Figure 11. Global trajectory of the quadrotor in 3D along the (X, Y, Z) axis (*Without faults*)



Figure 12. Global trajectory of the quadrotor in 3D along the (X, Y, Z) axis (*Actuator faults*)

# Conclusion

This paper presents a novel active fault-tolerant control strategy for diagnosing the actuator faults for a quadrotor vehicle. This approach is based on the observer-based fault reconstruction and estimation (FRE) technique using an adaptive observer. Firstly, we introduced a complete nonlinear dynamical model of the quadrotor, taking into consideration several physics phenomena that might impact our system's navigation in space. Secondly, an adaptive observer has been developed to estimate simultaneously the system state used in feedback control and actuator faults used in the FDI task. Thirdly we presented a stabilizing control law, in the presence of actuator faults, based on backstepping technique.

Many test simulations in MATLAB have been executed to evaluate the performance of the proposed strategy. In the first test, the motion of the quadrotor is considered normal without faults. In the second test, we created multiple actuator failures relating to roll, pitch, yaw, and altitude motions. Simulation results clearly illustrates good performances and robustness towards stability and tracking of this control strategy with respect to the backstepping approach in the absence of faults. There is very excellent estimation of the actuator faults and the intended trajectories even after the appearance of actuator faults. We can see that the estimation errors of roll (), pitch () and yaw () still keep close to zero (), meanwhile, the estimation error of attitude still remain below which represent 2% of the desired attitude. Moreover, we can observe that the input control signals provided by this control strategy are physically realizable.

The contribution of this work, firstly, is the use of a complete model of the quadrotor considering the non-linearities and the high-order nonholonomic constraints of the system which simulate a real behaviour of the quadrotor, especially in faulty cases. Secondly, It’s the first use of the adaptive observer proposed in in the field of active FTC for quadrotor UAV. This observer can estimate the system state and actuator faults simultaneously, which can be used respectively in feedback control and the FDI task of the actuator faults. Another advantage of the use of this observer structure lies in the fact that neither the conventional adaptive state observer nor any other alternative to the adaptive observer can be used in FTC in the case of our complete nonlinear dynamical model of the quadrotor UAV because of the non-satisfaction of the persistent excitation condition by model used. While using an adaptive observer as proposed in (Oucief, Tadjine and Labiod 2016) does not require the system structure to satisfy the standard observer matching condition required in the conventional adaptive state observer. Finally, the use of these observers allows the estimation of any number of faults, regardless of the number of measured outputs, and it can estimate additive and multiplicative faults. The observer gains can be solved together with the Lyapunov inequality using LMI-based computations and do not require to change the system model into a special form or the resolution of a system of partial differential equations like in (Stamnes, Aamo and Kaasa 2011).

This strategy can be easily applied to other nonlinear systems faults tolerant where several faults occur simultaneously. or the process itself in the case where sensor faults occur.

# References

Avram, Remus, Xiaodong Zhang, and Jonathan Muse. 2018. “Nonlinear Adaptive Fault-Tolerant Quadrotor Altitude and Attitude Tracking With Multiple Actuator Faults.” *IEEE Transactions on Control Systems Technology* 26 (2): 701–707. doi:10.1109/TCST.2017.2670522.

Bouadi, H, M Bouchoucha, and M Tadjine. 2007. “Modelling and stabilizing control laws design based on backstepping for an UAV type-quadrotor.” *IFAC Proceedings Volumes* 40 (15): 245-250. doi:10.3182/20070903-3-FR-2921.00043.

Boyd, S, L El Ghaoui, and E Feron. 1994. *Linear Matrix Inequalities in System and Control Theory.* Philadelphia : Society for Industrial and Applied Mathematics.

Cho, Young Man, and Rajesh Raramani. 1995. “A Systematic Approach to Adaptive Observer Synthesis for Nonlinear Systems.” *Proceedings of Tenth International Symposium on Intelligent Control* 487-482. doi:10.1109/isic.1995.525102.

Corless, Martin, and Jay Tu. 1998. “State and Input Estimation for a Class of Uncertain Systems.” *Automatica* 34 (6): 757-764. doi:10.1016/S0005-1098(98)00013-2.

Derafa, L, t Madani, and A Benallegue. 2006. “Dynamic modelling and experimental identification of four rotor helicopter parameters.” *2006 IEEE International Conference on Industrial Technology* 1834-1839. doi:10.1109/ICIT.2006.372515.

Farza, M, M M'Saad, T Maatoug, and M Kamoun. 2009. “Adaptive observers for nonlinearly parameterized class of nonlinear systems.” *Automatica* 45 (10): 2292-2299. doi:10.1016/j.automatica.2009.06.008.

Floquet, T, C Edwards, and S.K Spurgeon. 2007. “On Sliding Mode Observers for Systems with Unknown Inputs.” *International Journal of Adaptive Control and Signal Processing* 21 (8-9): 638-656. doi:10.1002/acs.958.

Freddi, Alessandro, Sauro Longhi, and Andrea Monteriù. 2010. “Actuator fault detection system for a mini-quadrotor.” *IEEE International Symposium on Industrial Electronics* 2055-2060. doi:10.1109/ISIE.2010.5637750.

Grant, Michael, and Stephen Boyd. 2014. *CVX: Matlab Software for Disciplined Convex Programming.* March. Accessed 06 16, 2023. http://cvxr.com/cvx/.

Hasanshahi, Mahsa, Aliakbar Ahmadi, and Roya Amjadifard. 2019. “Robust Fault Tolerant Position Tracking Control for a Quadrotor UAV in Presence of Actuator Faults.” *Proceedings of the 2019 6th International Conference on Control, Instrumentation and Automation (ICCIA)* 1-6. doi:10.1109/ICCIA49288.2019.9030838.

Hong-Jun , Ma, Liu Yanli, Li Tianbo , and Yang Guang-Hong. 2019. “Nonlinear High-Gain Observer-Based Diagnosis and Compensation for Actuator and Sensor Faults in a Quadrotor Unmanned Aerial Vehicle.” *IEEE Transactions on Industrial Informatics* 15 (1): 550-562. doi:10.1109/TII.2018.2865522.

Jain, Tushar , Joseph J. Yamé, and Dominique Sauter. 2018. *Active Fault-Tolerant Control Systems-A Behavioral System Theoretic Perspective.* 1. Springer Cham. doi:10.1007/978-3-319-68829-9.

Jiang, Jin, and Xiang Yu. 2012. “Fault-tolerant control systems: A comparative study between active and passive approaches.” *Annual Reviews in Control* 36 (1): 60-72. doi:10.1016/j.arcontrol.2012.03.005.

Lien, Yu-Hsuan, Peng Chao-Chung , and Chen Yi-Hsuan . 2020. “Adaptive Observer-Based Fault Detection and Fault-Tolerant Control of Quadrotors under Rotor Failure Conditions.” *Applied Sciences* 10 (10): 3503. doi:10.3390/app10103503.

Ouadine , Ahmed Youssef , Mostafa Mjahed , Hassan Ayad , and Abdeljalil El Kari. 2020. “UAV Quadrotor Fault Detection and Isolation Using Artificial Neural Network and Hammerstein-Wiener Model.” *Studies in Informatics and Control* 29 (3): 317-328. doi:10.24846/v29i3y202005.

Oucief, Nabil , Mohamed Tadjine, and Salim Labiod . 2016. “A new methodology for an adaptive state observer design for a class of nonlinear systems with unknown parameters in unmeasured state dynamics.” *Transactions of the Institute of Measurement and Control* 40 (4): 1297-1308. doi:10.1177/0142331216680288.

Raoufi , Reza, Horacio Jose Marquez , and Alan Solo. 2010. “ℋ︁∞ sliding mode observers for uncertain nonlinear Lipschitz systems with fault estimation synthesis.” *International Journal of Robust and Nonlinear Control* 20 (16): 1785-1801. doi:10.1002/rnc.1545.

Ren, Xiao-Lu. 2020. “Observer Design for Actuator Failure of a Quadrotor.” *IEEE Access* 8: 152742-152750. doi:10.1109/ACCESS.2020.3017522.

Stamnes, Oyvind Nistad, Ole Morten Aamo, and Glenn-Ole Kaasa. 2011. “Redesign of adaptive observers for improved parameter identification in nonlinear systems.” *Automatica* 47 (2): 403-410. doi:10.1016/j.automatica.2010.11.005.

That , Long Ton, and Zhengtao Ding. 2014. “ADAPTIVE LIPSCHITZ OBSERVER DESIGN FOR A MAMMALIAN MODEL.” *Asian Journal of Control* 16 (2): 335–344. doi:10.1002/asjc.731.

Xulin, Liu, and Guo Yuying. 2018. “Fault tolerant control of a quadrotor UAV using control allocation.” *Proceedings of the 2018 Chinese Control And Decision Conference (CCDC)* 1818-1824. doi:10.1109/CCDC.2018.8407422.

Yujiang , Zhong, Zhang Youmin, Zhang Wei, Zuo Junyi, and Zhan Hao. 2018. “Robust Actuator Fault Detection and Diagnosis for a Quadrotor UAV With External Disturbances.” *IEEE Access* 6: 48169-48180. doi:10.1109/ACCESS.2018.2867574.

Zhong, Yu-jiang, Zhixiang Liu, Youmi Zhang, Wei Zhang, and Junyi Zuo. 2019. “Active fault-tolerant tracking control of a quadrotor with model uncertainties and actuator faults.” *Frontiers of Information Technology & Electronic Engineering* 20: 95-106. doi:10.1631/FITEE.1800570.