Improved Altitude Control Algorithm for Quadcopter Unmanned Aerial Vehicles

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Abstract: Quadcopter unmanned aerial vehicles continue to play important roles in several applications and the improvement of their control performance has been explored in a great number of studies. In this paper, we present an altitude control algorithm for quadcopters that consists of a combination of nonlinear and linear controllers. The smooth transition between the nonlinear and linear modes are guaranteed through controller gains that are obtained based on mathematical analysis. The proposed controller takes advantage and addresses some known shortcomings of the conventional proportional–integral–derivative control method. The algorithm is simple to implement, and we prove its stability through the Lyapunov theory. By prescribing certain flight conditions, we use numerical simulations to compare the control performance of our control method to that of a conventional proportional–derivative–integral approach. Furthermore, we use a DJI-F450 drone equipped with a laser ranging sensor as the experimental quadcopter platform to evaluate the performance of our new controller in real flight conditions. Numerical simulation and experimental results demonstrate the effectiveness of the proposed algorithm.

Keywords: quadcopter; unmanned aerial vehicle; altitude control; position control; PID control

1. Introduction

In recent decades, unmanned aerial vehicles (UAVs) have gained popularity in the scientific community due to their numerous advantages such as high maneuverability, diverse applicability, reliability, and economy. UAVs are used in a broad range of applications such as scientific research, civil engineering, military applications, aerial mapping, search and rescue operations, and risk zone inspection [1–4]. Quadcopters are an important class of UAVs that are superior in many ways: they have a simple structure, are quick to manufacture, and are inexpensive [4, 5]. Controlling their position, in general, and altitude, in particular, has been the subject of numerous studies [6–23] with ever more advanced algorithms having been proposed to address the quadcopter control problem.

1.1. Related Works

Many applications of quadcopters in the real world have been discussed in recent years. The most widely mentioned topics include autonomous landing on moving platforms [6–9], precision landing for autonomous docking and wireless self-charging system [10, 11], and autonomous delivery tasks [12–14]. These applications are all, somehow, based on a reliable altitude controller. Therefore, designing an altitude controller with superior performance is one of the most common and important efforts that researchers all around the world are nowadays undertaking.

To that end, several papers have addressed the altitude control problem in the literature. Santos et al. [15] presented an algorithm to accomplish quadcopter trajectory tracking tasks by controlling the altitude through an adaptive dynamic controller that was capable of dealing with
uncertainties in model parameters. Jayakrishnan [16] and Xiong et al. [17] used the sliding mode control (SMC) technique to control the horizontal position and attitude while also providing a significant improvement of altitude control. In other studies [18–20], the second order SMC method was also used to improve quadcopter altitude control performances. Yet another method was presented by Muliadi et al. [21], where the authors proposed a neural network approach to control UAV altitude dynamics. The results obtained with this method were verified through comparisons with a conventional proportional–integral–derivative (PID) control system. However, these approaches [15–21] have a common disadvantage in that the SMC technique generates a high chattering control signal method which reduces the lifetime of the entire system.

Although several control methods have been proposed in the literature, PID control has become the most widely used technique in a variety of applications all over the world because it is simple and easy to design and typically delivers a satisfactory performance. The PID approach was used in many studies [22–28] to achieve not only quadcopter altitude control, but also attitude stabilization and horizontal position control. In recent studies [29–32], the authors proposed the use of a multi-loop control architecture (i.e., inner-loop and outer-loop) to control quadcopters in specific applications. The outer-loop controllers were designed in different ways while the inner-loop controllers were all implemented using the PID control law.

Nevertheless, the conventional PID controller has several limitations. First, their fixed gains limit system performance over a wider operational range. When the required range of operation is large, the conventional PID controller is prone instability, because the nonlinearities in the system cannot be properly dealt with. Second, as conventional PIDs are based on a linear model, their performance may suffer in a nonlinear system like a quadcopter. Several studies have attempted to overcome these shortcomings. Phi et al. [33] presented a gain scheduling PID controller which determines the PID gains by linearly adjusting the gain as a function of tracking errors. In another approach [34], the authors used a pickup table to schedule the PID gains in a quadrotor fault tolerant control task. Both methods were able to improve the control performance under different operating conditions. However, it is still a linear control law which means that it may not perform well in non-linear systems. Further, scheduling the gains results in discontinuous transitions which may result in sudden jerks or oscillations. A quadcopter self-tuning fuzzy PID controller where the gains are tuned nonlinearly through fuzzy logic have been proposed to address this problem [35]. Goodarzi et al. [36] introduced a nonlinear PID control to achieve asymptotic system stability under the presence of uncertainties. Another way to cope with the difficulties in control of a nonlinear system is to represent the dynamical system by the Takagi–Sugeno (T–S) fuzzy model [37]. Based on this approach, Liu et al. [38] addressed the problem of finite-time stability and stabilization for a class of nonlinear systems with time-varying delay. However, both the T–S fuzzy and linear parameter varying (LPV) approaches [39,40] require considerably greater computational resources compared with a conventional PID controller.

The references indicate that most of the existing methods either are complex to design and implement or require great computational resources. Meanwhile, PID control law appears to play an important role for finding a simple and efficient control method for a variety of systems. Designing a controller that makes use of the inherent advantages of the PID approach while tackling its shortcomings is therefore an important and worthwhile challenge.

1.2. Main Contributions

Motivated by the above observations, in this article, we propose an altitude control algorithm for quadcopter unmanned aerial vehicles. The main advantages of this new method are twofold: (1) while our method inherit the advantages of the PID approach, i.e., it is simple and straightforward, its multi-loop approach ensures that unlike other control algorithms [33–38], our controller requires less computational resources and exhibits a smooth performance at all times; and (2) compared to conventional PID controllers, our control algorithm delivers a greatly improved performance with regard to both raising time and error convergence speed.
We use the Lyapunov stability theory to validate the system’s stability. We prescribe a maximum acceleration/deceleration of the vehicle which allows the controller to have the vehicle move at an appropriate acceleration before fast and hard approaching a target. This contributes to a safer and smoother flight performance of both single vehicles performing specific missions \([5,14,41,42]\) and multi-agent systems flying in formation or switching their inter-communication topology \([43,44]\). For safety reasons, we firstly evaluate the performance of our new controller with numerical simulations and then we will conduct experimental flights to demonstrate the effectiveness of the proposed algorithm in real flight conditions.

1.3. Organization

The remainder of this paper is organized as follows: Section 2 presents the dynamical quadcopter model and Section 3 the control algorithm while the numerical simulation and experimental demonstration results and their discussions are provided in Section 4, before ending with the main conclusions.

2. Preliminaries

2.1. Quadcopter Dynamics Model

Here, we briefly summarize the quadcopter dynamics model as it has already been presented previously \([45,46]\).

Let \(\phi\), \(\theta\), and \(\psi\) denote the three Euler angles roll, pitch, and yaw, respectively, where, \(|\phi| < \pi/2\), \(|\theta| < \pi/2\), and \(|\psi| \leq \pi\) (see Figure 1). The symbols \(x\), \(y\), and \(z\) respectively represent the position of the quadcopter along the \(x\), \(y\), and \(z\) axes in an earth-fixed frame \([E]\). \(J_x\), \(J_y\), and \(J_z\) denote the moments of inertia along the \(x\), \(y\), and \(z\) axes, respectively; \(m\) represents the mass, \(l\) the arm length of the vehicle, and \(g\) the gravitational acceleration. The quadcopter dynamics model can be described as follows \([45,46]\):

\[
\begin{align*}
\ddot{\phi} &= \left(\frac{J_y - J_z}{J_x}\right)\dot{\theta}\dot{\psi} + \frac{1}{J_x}U_2 \\
\ddot{\theta} &= \left(\frac{J_z - J_x}{J_y}\right)\dot{\phi}\dot{\psi} + \frac{1}{J_y}U_3 \\
\ddot{\psi} &= \left(\frac{J_x - J_y}{J_z}\right)\dot{\phi}\dot{\theta} + \frac{1}{J_z}U_4 \\
\ddot{x} &= \frac{1}{m}(\cos \phi \sin \theta \cos \psi + \sin \phi \sin \psi)U_1 \\
\ddot{y} &= \frac{1}{m}(\cos \phi \sin \theta \sin \psi - \sin \phi \cos \psi)U_1 \\
\ddot{z} &= g - \frac{1}{m}(\cos \phi \cos \theta)U_1
\end{align*}
\]  

(1)

where \(U_i (i = 1, 2, 3, 4)\) denotes the control inputs which are:

\[
\begin{align*}
U_1 &= F_1 + F_3 + F_2 + F_4 \\
U_2 &= F_4 - F_2 \\
U_3 &= F_3 - F_1 \\
U_4 &= C_d (F_1 + F_3 - F_2 - F_4)
\end{align*}
\]  

(2)

where \(F_i = C_t \Omega_i^2\) denotes the thrust force generated by motor \(i\); \(\Omega_i\) is the speed of motor \(i\); \(C_t\) and \(C_d\) are the thrust and drag coefficients, respectively.
A multi-mode control law is chosen for the outer-loop of the altitude controller as:

\[ z_d - v_z = -\dot{z} \]  (4) 

A new algorithm is proposed for altitude control. It is worth noting that our new algorithm can also be used for horizontal position control.

### 3. Controller Design

The multi-loop PID-PID control scheme for quadcopter altitude control was introduced and verified in some existing studies [28,33,47]. In this section, we present our new altitude control algorithm which also consists of a multi-loop controller, i.e., outer-loop and inner-loop. We derive the outer-loop as a multi-mode (i.e., nonlinear and linear modes) control law while the inner-loop is a PID controller.

Let \( z_d \) denote the desired attitude, and \( v_z \) represent the vertical velocity of the vehicle. The altitude tracking error is then given by:

\[ e_z = z_d - z \]  (3) 

Then the derivative of \( e_z \) is:

\[ \dot{e}_z = z_d - v_z \]  (4)
A multi-mode control law is chosen for the outer-loop of the altitude controller as:

\[
v_z = \begin{cases} 
-\sqrt{2a_{\text{max}}(-e_z - e_0)} + \dot{z}_d, & e_z < -d_l \\
ke_z + \dot{z}_d, & -d_l \leq e_z \leq d_l \\
\sqrt{2a_{\text{max}}(e_z - e_0)} + \dot{z}_d, & e_z > d_l 
\end{cases}
\]  

where, \(a_{\text{max}}\) is a positive constant which represents the desired maximum acceleration of the vehicle; \(d_l\) a positive constant at which the line \((\Delta)\): \(v_z = ke_z + \dot{z}_d\) to be tangential to the curve \((S)\): \(v_z = \begin{cases} 
-\sqrt{2a_{\text{max}}(-e_z - e_0)} + \dot{z}_d, & e_z < -d_l \\
\sqrt{2a_{\text{max}}(e_z - e_0)} + \dot{z}_d, & e_z > d_l 
\end{cases}\); \(k, e_0\) positive constants, and \(e_0 < d_l\); \(v_z\) the control action which will be fed into the inner-loop (see Figure 3) as a velocity reference [31].

![Figure 3](image-url)  
**Figure 3.** The multi-loop altitude control scheme of the quadcopter. The proposed algorithm is implemented in the outer-loop while a PID controller is used in the inner-loop.

Figure 4 shows \(v_z\) as a function of \(e_z\) from Expression (5) corresponding to \(\dot{z}_d = 0\). For \(\dot{z}_d \neq 0\), the graphs are shifted along the \(v_z\) axis an amount of \(\dot{z}_d\) without any change of the relations between \((S)\) and \((\Delta)\). In order for the line \((\Delta)\) to be tangential to the curve \((S)\) at \(e_z = d_l\), the followings must be satisfied:

\[
\begin{align*}
ke_z &= \sqrt{2a_{\text{max}}(d_l - e_0)} \\
k &= \frac{\sqrt{2a_{\text{max}}}}{2 \sqrt{d_l - e_0}}
\end{align*}
\]  

Manipulating (6) yields:

\[
\begin{align*}
d_l &= \frac{a_{\text{max}}}{k^2} \\
e_0 &= \frac{d_l}{2}
\end{align*}
\]  

![Figure 4](image-url)  
**Figure 4.** Graphical illustration of the curve \((S)\) (blue), the line \((\Delta)\) (green), and their relations.
**Theorem 1.** If the control law is designed as Expression (3) with the parameters satisfying (6) then the system (1) is asymptotically stable and the tracking error is forced to zero.

**Proof.** Choose a Lyapunov candidate function as:

\[
V = \begin{cases} 
2 \sqrt{-e_z - e_0}, & e_z < -d_l \\
\frac{1}{2} e_z^2, & -d_l \leq e_z \leq d_l \\
2 \sqrt{e_z - e_0}, & e_z > d_l 
\end{cases} 
\]  

(8)

It is clear that \( V > 0, \forall e_z \neq 0 \). We have:

\[
\dot{V} = \begin{cases} 
-\dot{e}_z / \sqrt{-e_z - e_0}, & e_z < -d_l \\
\dot{e}_z e_z / \sqrt{-e_z - e_0}, & -d_l \leq e_z \leq d_l \\
\dot{e}_z / \sqrt{e_z - e_0}, & e_z > d_l 
\end{cases} 
\]  

(9)

Substituting (4) into (9) and choosing control laws as (5), we have:

\[
\dot{V} = \begin{cases} 
-\sqrt{2a_{\text{max}}}, & e_z < -d_l \\
-k \dot{e}_z^2, & -d_l \leq e_z \leq d_l \\
-\sqrt{2a_{\text{max}}}, & e_z > d_l 
\end{cases} 
\]  

(10)

Clearly, we have \( \dot{V} < 0, \forall e_z \neq 0 \) which means that the tracking error, \( e_z \), converges to zero as time \( t \to \infty \). This completes the proof. \( \square \)

**Remark 1.** When \( e_z < -d_l \) or \( e_z > d_l \), the controller generates control actions that follow the curve \( (S) \). While this is the case, the slope of \( (S) \) is smaller than that of \( (\Delta) \) which means that the acceleration is limited, which in turn results in fewer overshoots and vehicle jerks. Whereas, when \(-d_l \leq e_z \leq d_l\), the control action follows the line \( (\Delta) \) which has a higher slope. This guarantees that the tracking error converges to zero in a prompt and hard manner.

**Remark 2.** When \(-d_l \leq e_z \leq d_l\), the proposed controller becomes a proportional-PID (P-PID) control scheme, which implies that the parameter \( k \) can be obtained through a conventional PID tuning process. Moreover, at \( e_z = -d_l \) and \( e_z = d_l \), the line \( (\Delta) \) becomes a tangent to the curve \( (S) \), resulting in smooth transitions when the controller switches control modes (between nonlinear and linear modes), with regard to both acceleration and velocity.

4. Numerical Simulation, Experimental Results and Discussions

In this section, we demonstrate the effectiveness of the proposed altitude control algorithm. The quadcopter dynamics from Expression (1) are used to verify the stability and advanced performance of our controller. By comparing our new method to a conventional PID controller, we highlight and discuss the effectiveness of the new method.
4.1. Experimental Platform Description

4.1.1. Hardware and Avionics Devices

We use a DJI-F450 quadcopter as the experimental platform (Figure 5) which is operated by an onboard flight computer unit (FCU) Pixhawk [48,49]. The quadcopter attitude and acceleration are provided by an inertial navigation system (INS). We use a commercially available laser ranging sensor LidarLite V3 to measure the altitude and a commercial global positioning system (GPS) receiver to determine the position of the vehicle. Additionally, a power supplying system (including battery, power adapter module), a set of remote control transmitter/receiver for the manual pilot, and a set of radio telemetry transmitter/receiver for ground station monitoring are used. The quadcopter dynamical parameters are listed in Table 1.

![Quadcopter platform](image)

**Figure 5.** Quadcopter platform used for the experimental demonstration.

**Table 1.** System parameters used for the simulations.

| Parameter | Value       | Unit |
|-----------|-------------|------|
| m         | 2.0         | kg   |
| J_x       | 0.0121      | kg.m^2 |
| J_y       | 0.0119      | kg.m^2 |
| J_z       | 0.0223      | kg.m^2 |
| l         | 0.23        | m    |
| g         | 9.81        | m/s^2 |
| C_t       | 7.732 (10^{-6}) | Ns^2 |
| C_b       | 1.275 (10^{-7}) | Nms^2 |

4.1.2. Software

We implemented all vehicle altitude, position, and attitude controllers on the Pixhawk FCU running at frequencies of 100 Hz, 100 Hz, and 400 Hz, respectively. The altitude controller gains and reference values for altitude are listed in Table 2. It is worth noting that the horizontal position controller and attitude controller are beyond the scope of this paper and will not be presented in this section. The system signal flows are briefly described by the block diagram in Figure 6.
The controllers are implemented based on Matlab/Simulink with the sampling time is chosen identically as the operating frequency of the controllers in the experimental system. The initial vehicle altitude and vertical velocity are all zero, i.e., $z(0) = 0$ and $\dot{z}(0) = 0$. The vehicle is armed on a horizontal level surface with the attitude stabilization controller enabled.

We present two test scenarios to evaluate the effectiveness of our proposed algorithm, namely the tracking of (1) a step altitude reference; and (2) a moving altitude reference. For comparison, we also simulated the performance of a conventional PID controller for both scenarios.

### 4.2.1. Scenario 1: Tracking Step Altitude References

This scenario consists of three stages: (i) initialization; (ii) take-off; and (iii) descent. During the first five seconds the vehicle is initialized, i.e., it is armed and prepares for take-off. The altitude, vertical velocity, and acceleration are all zero (Figures 7–10) and the control signal, $U_t$, increases from the initial value of zero to the hovering value of about 19.62 N (Figure 11).

During the second stage (from $t = 5$ to $t = 30$ s), the quadcopter takes off and ascends to the desired altitude of 15 m. The altitude responses are nearly identical for both controllers during the early stage of the flight ($5 \leq t < 9$ s in Figure 7). Initially, the conventional PID controller responds slightly faster than our new controller (see Inset 1 in Figure 7). This is reversed during the later stages of the flight when the tracking error becomes smaller than $d_l$ ($e_z \leq d_l \approx 5.1$ m) and the controller smoothly switches to the linear mode. For $9 \leq t < 30$ s our new controller exhibits a significantly superior performance compared to the conventional PID. With our algorithm, the tracking error converges to a small range closed to zero ($e_z$ is less than 0.5%) at $t \approx 21$ s while the conventional PID only reaches this range around $t \approx 25$ s (Inset 1 in Figure 8). The velocity, acceleration, and control signal of the conventional PID controller exhibit large amplitudes and slow converging speeds (Figures 9–11).

In order to evaluate the performance of our controller when the step amplitude is smaller than $d_l$, during the third stage ($30 \leq t \leq 50$ s), the quadcopter is commanded to descend from the current altitude of 15 m to the desired altitude of 12 m (corresponding to a step of 3 m). If we look at the insets 2 in Figures 7 and 8, it is clearly seen that our controller aggressively responds and quickly reaches the altitude reference at $t \approx 45$ s while the conventional PID controller only reaches this reference at $t \approx 49$ s.

### 4.2. Numerical Simulation and Discussions

The experimental quadcopter parameters and controller gains are used for the simulation. We implemented all vehicle altitude, position, and attitude controllers on the Pixhawk FCU during the early stage of the flight ($5 \leq t < 30$ s in Figure 7). Initially, the conventional PID controller responds slightly faster than our new controller (see Inset 1 in Figure 7). This is reversed during the later stages of the flight when the tracking error becomes smaller than $d_l$ ($e_z \leq d_l \approx 5.1$ m) and the controller smoothly switches to the linear mode. For $9 \leq t < 30$ s our new controller exhibits a significantly superior performance compared to the conventional PID. With our algorithm, the tracking error converges to a small range closed to zero ($e_z$ is less than 0.5%) at $t \approx 21$ s while the conventional PID only reaches this range around $t \approx 25$ s (Inset 1 in Figure 8). The velocity, acceleration, and control signal of the conventional PID controller exhibit large amplitudes and slow converging speeds (Figures 9–11).

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### Table 2. Controller parameters and the altitude references.

| Parameters | Value and Unit |
|------------|----------------|
| $a_{\text{max}}$ | 2.5 m/s² |
| $k$ | 0.7 |
| $k_p$ | 1.1 |
| $k_i$ | 0.002 |
| $k_d$ | 0.2 |
| $z_d$ | 0, 15, 12, and 10 sin (0.1t) m |

### Figure 6. The system signal flow block diagram.

**Table 2. Controller parameters and the altitude references.**

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| $k$ | 0.7 |
| $k_p$ | 1.1 |
| $k_i$ | 0.002 |
| $k_d$ | 0.2 |
| $z_d$ | 0, 15, 12, and 10 sin (0.1t) m |
and the controller while the conventional PID controller only reaches this range around $t \approx 20$. If we look at the error $e_z$ (corresponding to a step of $15 \text{ m}$ to the desired altitude of $3 \text{ m}$), the quadcopter is commanded to descend from the current altitude $z_{led} \geq 30 \text{ m}$ while the conventional controller continues to descend to reach the altitude reference at $z_{led} = 3 \text{ m}$. Figure 11 shows a comparison of error $e_z$ between our new controller and a conventional PID controller for Scenario 1. It is clearly seen that our controller aggressively responds and quickly converges to a small range closed to zero ($\approx 0.5 \text{ m}$). The proposed controller is less than $0.5 \text{ m}$ during the third stage ($t = 9 \text{ s}$ to $t = 15 \text{ s}$). Between our new controller and a conventional PID, the tracking error is less than $0.5 \text{ m}$. In Figure 8, it is clearly seen that our controller exhibits a significantly superior performance compared to the conventional PID. With our algorithm, the tracking error of the flight when the tracking error becomes smaller than $0.5 \text{ m}$, the tracking error of the conventional PID controller exhibit large amplitudes and slow converging, while the proposed controller smoothly switches to the linear mode. For the second stage ($t = 15 \text{ s}$ to $t = 30 \text{ s}$), the tracking error is slightly faster than our new controller (see Inset 1 in Figure 7). This is reversed during the later stages ($t = 30 \text{ s}$ to $t = 45 \text{ s}$). Figure 9 and 10 show a comparison of the vertical velocity performance and vertical acceleration performance respectively between our new controller and a conventional PID controller for Scenario 1.

**Figure 7.** Comparison of altitude performance between our new controller and a conventional PID controller for Scenario 1 (tracking a step altitude reference).

**Figure 8.** Comparison of altitude tracking error convergence between our new controller and a conventional PID controller for Scenario 1.

**Figure 9.** Comparison of the vertical velocity performance between our new controller and a conventional PID controller for Scenario 1.
4.2.2. Scenario 2: Tracking a Moving Altitude Reference

Here, the vehicle tracks a moving altitude reference using the same controller gains as in Scenario 1. The moving altitude reference is given as a sine function of time \( t \) as:

\[
 z_d = 10 \sin(0.1t) \tag{11}
\]

Clearly, the altitude performance of our new algorithm is superior to a conventional PID controller (Figure 12). While the moving reference altitude reaches 10 m altitude at time \( t \approx 15.7 \) s, our controller reaches the peak about 3.8 s later (at \( t \approx 19.5 \) s), while the conventional PID controller lags by 9.7 m, reaching the maximum altitude at \( t \approx 21.1 \) s. Due to this significant delay, the conventional PID does not even reach the full altitude. In addition, the tracking error of the PID controller exhibits large amplitude oscillation and a slow response compared to our algorithm (Figure 13).
4.3. Experimental Demonstration and Discussions

In this subsection, we present the experimental results of a step-altitude-reference-tracking flight to evaluate the performance and demonstrate the effectiveness of our new controller in real flight conditions. After arming, the quadcopter stays on the ground and prepares for take-off during the first ten seconds (Figure 14). The altitude reference of 15 m is commanded at time $t \approx 10$ s leading the vehicle to ascend. It is clear that, initially, the conventional PID controller performance is considerably smaller. If we look at the inset in Figure 14, it is seen that the response of our proposed controller exhibits a superior performance compared with the PID controller when the tracking error becomes faster than our proposed controller. However, similarly to the simulation results, our new controller reaches the peak about 9.7 m, reaching the maximum altitude at $t \approx 19.5$ s leading the vehicle to ascend. It is clear that, initially, the conventional PID controller performance is considerably faster than our proposed controller. However, similarly to the simulation results, our new controller exhibits a superior performance compared with the PID controller when the tracking error becomes smaller. If we look at the inset in Figure 14, it is seen that the response of our proposed controller is about 5 s faster than that of the conventional PID controller. The experimental vertical velocity and acceleration performance (Figures 15 and 16) also prove the stability and the effectiveness of the proposed algorithm.
Figure 14. Comparison of experimental altitude performance between our new controller and a conventional PID controller.

Figure 15. Comparison of the experimental vertical velocity performance between our new controller and a conventional PID controller.

Figure 16. Comparison of the experimental vertical acceleration performance between our new controller and a conventional PID controller.
The way our new algorithm exhibits the performance promises a smooth system response when the vehicle starts a movement and a fast target approaching speed when the tracking error becomes small. These properties are useful in performing tasks that require high accuracy and smooth performance such as precision landing, obstacle avoidance, and autonomous delivery to high buildings, etc.

5. Conclusions

This paper presented a new altitude control algorithm for quadcopters. The experimental validation and the numerical simulation results clearly demonstrated that our algorithm performs significantly better compared to a conventional PID controller. While taking advantage of the strength of conventional PID approaches, our method successfully tackles some of their shortcomings. Our algorithm is simple to design and implement, and can be applied to a large range of position control applications for unmanned vehicles. Future work is dedicated to precision landing control design with our proposed controller for an autonomous quadcopter docking and self-charging system.

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