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Distributed Fixed-Time Formation Tracking Control for the Multi-Agent System and an Application in Wheeled Mobile Robots

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Abstract: This work addresses the issue of multi-agent system (MAS) formation control under external disturbances and a directed communication topology. Firstly, a new disturbance observer is proposed to effectively reconstruct and compensate for external disturbances within a short period of time. Then, the integral terminal sliding mode technology is introduced to devise a novel distributed formation control protocol, ultimately realizing the stability of the MAS within a fixed time. Moreover, by means of rigorous Lyapunov theory analyses, a faster formation convergence rate and more accurate consensus accuracies are achieved in the proposed fixed-time strategy with variable exponent form. Finally, the formation tracking control scheme is applied to a multi-wheeled mobile robot (WMR) system. The experimental results strongly support the fine effectiveness of the control scheme designed in this work.

Keywords: multi-agent system; formation tracking; disturbance observer; multi-wheeled mobile robot system



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1. Introduction

The multi-agent system (MAS) is one of the current research hotspots with a broad range of applications, among which are spacecraft or unmanned aerial vehicle formation flying, multi-robot transporting, and so on [1–3]. As a fundamental control problem of the MAS, the formation control aims to coordinate MASs to execute tasks in the desired pattern. Due to the disturbances derived from the real environment and the communication constraints within the formation, designing a practical and reliable formation control scheme remains an important research topic.

The key issue of the successful formation control is the effective communication among the multiple agents. Communication plays a pivotal role in sharing information, exchanging data, and synchronizing actions within the formation process. At present, the main communication approaches for formation control include the centralized method [4], decentralized method [5], and distributed method [6]. In large-scale MAS formation, the distributed method has the advantages of scalability, robustness, and adaptability, while striking a balance between resource consumption and communication efficiency. Thus, it has been widely utilized in practical applications [7–9]. Based on an undirected topological graph, ref. [7] formulated distributed observers for each follower within a category of multi-agent systems characterized by nonlinear uncertainties. In [8], a distributed control approach based on digital twin technology was introduced to counter composite attacks in multi-agent systems, including Denial-of-Service and actuation attacks, with its effectiveness demonstrated through simulation and experimentation. The distributed communication was integrated into the state estimators in [9], which resulted in the controller being capable of predefining the system's convergence performance. Simultaneously, the

controller introduced a repulsive potential function to ensure collision avoidance during the formation process.

It is worth noting that the rate of convergence is imperative in evaluating the performance of consensus-based formation tracking control of MASs. Therefore, finite-time control strategies emerged [10] and were extensively applied in formation control of MASs [11–14]. Based on the homogeneous theory, ref. [15] investigated the finite-time formation control problem for MAS. In [16], a distributed finite-time bearing-only formation control method for MASs was proposed based on orientation estimation, which attained almost global finite-time convergence of the actual formation to the specified formation shape. Through the application of the adaptive law to the gradient term of the potential energy function, the authors introduced a novel finite-time controller designed for the distributed anti-jamming formation control of multiple unmanned helicopters [17]. To address the time-varying formation control challenges of multiple maritime surface vessels, ref. [18] devised a novel finite-time control algorithm utilizing the sliding mode control approach. This proposed resolution not only mitigates uncertainties and input saturation constraints within the system but also effectively handles issues arising from actuator faults.

It is essential to highlight that in the aforementioned studies, the determination of the upper bound of the settling time is contingent upon the initial states of the control system. However, it is very difficult to obtain the initial states accurately in practical applications. To address this issue, ref. [19] proposed the fixed-time stability theory, which makes the settling time unaffected by the initial conditions. Currently, this theory has been applied to research on various control issues, such as actuator failures in quadrotor and spacecraft [20,21] and attitude stabilization of aircraft [22,23] and has consistently demonstrated superior control performance in these applications. In [24], an improved continuous fixed-time sliding-mode control law was investigated, which sped up the theoretical convergence time of the spacecraft system while ensuring that the controller was chattering-free and nonsingular. A combination of fixed-time strategy and backstepping control was employed in [25], along with the use of a filter to address the computational complexity in backstepping design. The method resulted in the realization of precise trajectory tracking for underwater vehicles.

With the maturation of fixed-time theory, it has been proficiently applied in the research of formation control in multi-agent systems [26–28]. In [29], a distributed fixed-time protocol has been proposed for an MAS chain structure by utilizing the backstepping control method. Ref. [30] investigated the cross-dimensional formation for a class of second-order multi-dimensional heterogeneous MASs. Furthermore, to deal with the external disturbance problem in MASs, ref. [31] integrated the disturbance observer with the super-twisting control method to achieve the stability of the MAS. For multiple unmanned ground vehicles with mismatched disturbances and parameter uncertainties, the authors proposed a time-varying formation control scheme [32]. In addition, ref. [33] addressed the input delay that exists in the system by designing a state observer and transforming the nonholonomic mobile robot model into two subsystems. Subsequently, the distributed controllers were conceived for the subsystems separately by integrating the estimation information of the future state attained from the observer to achieve fixed-time stable formation tracking.

Motivated by the above results, constructing a high-performance multi-agent formation control scheme holds practical significance. However, the multi-agent formation control system inevitably faces the adverse effect deriving from external disturbances in practical engineering applications. To address this dilemma, this work introduces a disturbance-resistant fixed-time formation control algorithm. The main contributions of this paper are summarized in the following points:

- (1) A terminal sliding-mode surface is constructed by using local information among leader–follower agents. Furthermore, a new form of sliding-mode observer incorporating a Gaussian error function is proposed that can effectively estimate external disturbances and compensate for their impact on the system.

(2) A new distributed fixed-time formation control protocol with newly added variable exponential and variable coefficient terms is proposed to enhance the convergence rate and accuracy, in which the parameters of the controller can be adjusted according to the state of the system.

(3) A remarkable point is that the formation tracking experiments are conducted on a multi-wheeled mobile robot (WMR) experimental platform. The introduction of novel error variables facilitated the achievement of fixed-time formation tracking control of multi-WMRs. Experimental results substantiate the practical engineering effectiveness of the designed formation control scheme.

Notation 1. Define $\mathbf{x} = [x_1, x_2, \dots, x_n]^T \in \mathbb{R}^N$, $|\mathbf{x}| = [|x_1|, |x_2|, \dots, |x_n|]^T \in \mathbb{R}^N$ and $\text{sign}(\mathbf{x}) = [\text{sign}(x_1), \text{sign}(x_2), \dots, \text{sign}(x_n)]^T \in \mathbb{R}^N$ with $\text{sign}(x_i) = 1 (x_i \geq 0)$ and $\text{sign}(x_i) = -1 (x_i < 0)$. For the sake of simplicity, define $[\mathbf{x}]^\alpha = [|x_1|^\alpha \text{sign}(x_1), \dots, |x_n|^\alpha \text{sign}(x_n)]^T \in \mathbb{R}^N$ for $\mathbf{x} \in \mathbb{R}^N$, $\alpha \in \mathbb{R}^+$. $\mathbf{1}_N = [1, 1, \dots, 1]^T \in \mathbb{R}^N$. Denote $\|\cdot\|$ to be Euclidean norm with $\cdot$ being an arbitrary and $\otimes$ being the Kronecker product.

2. Preliminaries and Problem Statements

2.1. Graph Theory Preliminaries

This paper describes a formation system comprising n agents, employing a directed digraph $G = \{V, \varepsilon, A\}$ to depict the communication topology that elucidates the information interchange among the individual agents. $V = \{v_1, v_2, \dots, v_n\}$ denotes a vertex set with n nodes; $\varepsilon \subseteq V \times V$ and $A = [a_{ij}] \in \mathbb{R}^{N \times N}$ represent the collection of linking edges and the weighted adjacency matrix, respectively. When agent i is capable of receiving information from agent j , then $a_{ij} = 1$, otherwise $a_{ij} = 0$. Denote the diagonal matrix $\mathbf{b} = \text{diag}\{b_1, b_2, \dots, b_n\}$ as the adjacent matrix linking the leader and followers, and $b_i = 1$ indicates that the i th follower can obtain information from the leader, otherwise $b_i = 0$.

Assumption 1. The digraph G contains a directed spanning tree where the node of the leader robot serves as the root.

2.2. Mathematical Preliminaries

Lemma 1 ([34]). Define $H = L + B$, then the matrix H is a positive stable matrix whose eigenvalues have positive real parts if the digraph G has a directed spanning tree.

Lemma 2 ([11]). The following inequality is given as $\sum_{i=1}^N x_i^m \geq N^{1-m} \left(\sum_{i=1}^N x_i \right)^m$ with $x_i \geq 0$ and $m > 1$.

Lemma 3 ([35]). For any $\kappa > 0$ and $\vartheta \in \mathbb{R}$

$$|x| - \frac{\vartheta}{\kappa} \leq x \tanh(\kappa x)$$

where $\vartheta = 0.2785$.

Lemma 4 ([36]). The Gaussian error function is defined as $\text{erf}(x) = \frac{2}{\sqrt{\pi}} \int_0^x e^{-t^2} dt$, where e is the natural constant. And the Gaussian error function satisfies $\frac{1}{2}x \leq \text{erf}(x) \leq 2x$ for $0 \leq x \leq 1$.

Lemma 5 ([37]). The following chain of inequalities hold: $x \tanh < x \text{erf}(\frac{x}{\varphi}) < |x|$ for $x \in \mathbb{R}$ and $\varphi > 0$.

2.3. HMAS Model Descriptions

Consider the high-order MAS (HMAS) of n followers

$$\begin{cases} \dot{q}_{i,1} = q_{i,2} \\ \dot{q}_{i,2} = q_{i,3} \\ \vdots \\ \dot{q}_{i,m} = u_i + d_i \end{cases} \quad (1)$$

where $q_i = [q_{i,1}, q_{i,2}, \dots, q_{i,m}]^T \in \mathbb{R}^M$ represents the state vector and $i = 1, 2, \dots, n$. $u_i \in \mathbb{R}$ and $d_i \in \mathbb{R}$ denote the control input and external disturbance of the i th follower, respectively.

The dynamic model of the virtual leader agent is described as

$$\begin{cases} \dot{q}_{0,1} = q_{0,2} \\ \dot{q}_{0,2} = q_{0,3} \\ \vdots \\ \dot{q}_{0,m} = u_0 \end{cases} \quad (2)$$

where $u_0 \in \mathbb{R}$ and $q_0 = [q_{0,1}, q_{0,2}, \dots, q_{0,m}]^T \in \mathbb{R}^M$ represent the control input of the leader and the state vector, respectively.

Assumption 2. The disturbance d_i in (1) is continuous and bounded, that is, $\|d\| \leq \bar{k}_0$, where $\bar{k}_0$ is a positive constant.

Assumption 3. The control input u_0 in (2) of the virtual leader is unknown and bounded, expressed as $|u_0| \leq \bar{u}_0$, where $\bar{u}_0$ is a positive constant.

3. Fixed-Time ISMC-Based Formation Control for HMASs

In this section, a new fixed-time formation control protocol is designed for HMASs in (1) and (2). Considering the communication structure between multi-agents, we first introduce the following consensus error variables

$$e_{i,k} = \sum_{j=1}^N a_{ij}(q_{i,k} - q_{j,k}) + b_i(q_{i,k} - q_{0,k}) \quad (3)$$

with $k = 1, 2, \dots, m$. Define error vector $e_k = [e_{1,k}, e_{2,k}, \dots, e_{n,k}]^T \in \mathbb{R}^N$ and input vector $u = [u_1, u_2, \dots, u_n]^T \in \mathbb{R}^N$, the error dynamic model can be rewritten as follows

$$\begin{cases} \dot{e}_1 = e_2 \\ \dot{e}_2 = e_3 \\ \vdots \\ \dot{e}_m = H(u - \mathbf{1}_N \otimes u_0 + d) \end{cases} \quad (4)$$

3.1. Fixed-Time Disturbance Observer

In view of the external disturbances existing in (1), the auxiliary variable is considered as $z_i = q_{i,m} - \sigma_i$, where σ_i satisfies the following equation

$$\dot{\sigma}_i = u_i + l_1 \operatorname{erf}\left(\frac{z_i}{\epsilon_1}\right) + l_2 \lceil z_i \rceil^{\gamma_1 + \gamma_2 \operatorname{sign}(|z_i| - 1)} + l_3 \lceil z_i \rceil^{\gamma_3} \quad (5)$$

in which l_1, l_2, l_3 , and ϵ_1 are positive constants and $l_1 > \bar{k}_0$. The parameters $\gamma_1, \gamma_2, \gamma_3$ satisfy $\gamma_1 + \gamma_2 > 1, 0 < \gamma_1 - \gamma_2 < 1, \gamma_3 > 1$.

Theorem 1. For the i th agent in the error system (4), a continuous disturbance observer is designed as

$$\hat{d}_i = l_1 \operatorname{erf}\left(\frac{z_i}{\epsilon_1}\right) + l_2 [z_i]^{\gamma_1 + \gamma_2 \operatorname{sign}(|z_i| - 1)} + l_3 [z_i]^{\gamma_3} \quad (6)$$

Then, the estimation error $\tilde{d}_i = d_i - \hat{d}_i$ will converge to a small neighborhood around the origin in a fixed time T_{d0} .

Proof. Define $\mathbf{z} = [z_1, z_2, \dots, z_n]^T \in \mathbb{R}^N$ and $\mathbf{d} = [d_1, d_2, \dots, d_n]^T \in \mathbb{R}^N$, one selects a positive definite Lyapunov function as $V_1 = \mathbf{z}^T \mathbf{z}$. Differentiating V_1 , according to Lemma 3 and Lemma 5, one has

$$\begin{aligned} \dot{V}_1 &\leq -2\left(l_1 \mathbf{z}_1^T \operatorname{erf}\left(\frac{\mathbf{z}}{\epsilon_1}\right) + l_2 \mathbf{z}^T \mathbf{Q}(\mathbf{z}) + l_3 \mathbf{z}^T [\mathbf{z}]^{\gamma_3} - \mathbf{z}^T \mathbf{d}\right) \\ &\leq -2\left(l_1 \mathbf{z}_1^T \tanh\left(\frac{\mathbf{z}}{\epsilon_1}\right) + l_2 \mathbf{z}^T \mathbf{Q}(\mathbf{z}) + l_3 \mathbf{z}^T [\mathbf{z}]^{\gamma_3} - \mathbf{z}^T \mathbf{d}\right) \\ &\leq -2\left(l_1 \|\mathbf{z}\| - N l_1 \rho_1 \epsilon_1 + l_2 \mathbf{z}^T \mathbf{Q}(\mathbf{z}) + l_3 \mathbf{z}^T [\mathbf{z}]^{\gamma_3} - \bar{k}_0 \|\mathbf{z}\|\right) \\ &\leq -2l_2 \mathbf{z}^T \mathbf{Q}(\mathbf{z}) - 2l_3 \mathbf{z}^T [\mathbf{z}]^{\gamma_3} + 2N l_1 \rho_1 \epsilon_1 \\ &\leq -2l_2 \sum_{i=1}^N |z_i|^{\gamma_1 + \gamma_2 \operatorname{sign}(|z_i| - 1) + 1} - 2l_3 N^{-\gamma_3} \|\mathbf{z}\|^{\gamma_3 + 1} + 2l_1 \rho_1 \epsilon_1 \end{aligned} \quad (7)$$

where ρ_1 and ϵ_1 are positive constants. Denote $\mathbf{Q}(\mathbf{z}) = [Q_1(z_1), Q_2(z_2), \dots, Q_n(z_n)]^T$ with $Q_i = [z_i]^{\gamma_1 + \gamma_2 \operatorname{sign}(|z_i| - 1)}$.

Case 1 When $V_1 \geq 1$, it can be obtained that $\|\mathbf{z}\| \geq 1$ and $\dot{V}_1 \leq -2l_3 N^{-\gamma_3} \|\mathbf{z}\|^{\gamma_3 + 1} + 2l_1 \rho_1 \epsilon_1$. Define $\bar{\epsilon}_1 = l_1 \rho_1 \epsilon_1$, one can achieve $\dot{V}_1 \leq -2(\bar{N}_1 - \bar{\epsilon}_1) V_1^{\frac{\gamma_3 + 1}{2}}$ with $\bar{N}_1 = l_3 N^{-\gamma_3}$. Hence, all the solutions of $\{V_1 \geq 1\}$ will reach the set $\{V_1 < 1\}$ in a settling time, which is given by $t_{d1} \leq \frac{1}{(\gamma_3 - 1)(\bar{N}_1 - \bar{\epsilon}_1)}$.

Case 2 In the converse case $V_1 < 1$, there is $\|\mathbf{z}\| < 1$. Inequality (7) can be redrafted as $\dot{V}_1 \leq -2\kappa_1 \bar{N}_2 V_1^{\frac{\gamma_1 - \gamma_2 + 1}{2}} - 2(1 - \kappa_1) \bar{N}_2 V_1^{\frac{\gamma_1 - \gamma_2 + 1}{2}} + \bar{\epsilon}_1$ with $\bar{N}_2 = l_2 N^{\gamma_2 - \gamma_1}$ and $0 < \kappa_1 < 1$. When $\bar{\epsilon}_1 - 2(1 - \kappa_1) \bar{N}_2 V_1^{\frac{\gamma_1 - \gamma_2 + 1}{2}} \leq 0$, then $\dot{V}_1$ is simplified as $\dot{V}_1 \leq -2\kappa_1 \bar{N}_2 V_1^{\frac{\gamma_1 - \gamma_2 + 1}{2}}$. Consequently, the solution of V_1 will reach in a compact set given by $\Theta = \left\{ \mathbf{z} | V_1(\mathbf{z}) \leq \left(\frac{\bar{\epsilon}_1}{2(1 - \kappa_1) \bar{N}_2} \right)^{\frac{2}{\gamma_1 - \gamma_2 + 1}} \right\}$ within a fixed time $t_{d2} \leq \frac{1}{\kappa_1 \bar{N}_2 (1 - \gamma_1 + \gamma_2)}$.

Therefore, the estimation error $\tilde{d}_i$ will converge to a small set Θ within $T_{d0} \leq t_{d1} + t_{d2}$. $\square$

3.2. Fixed-Time Formation Control Protocol

To handle the problem of the fixed-time formation control of HMAss, an integral sliding-mode surface is introduced as follows [26]:

$$s_i = e_{i,m} + \int_0^t \sum_{j=1}^M k_j ([e_{i,j}]^{p_j} + [e_{i,j}]^{q_j}) d\tau \quad (8)$$

where the parameters k_j , p_j , and q_j are chosen to satisfy the constraints in Lemma 2 in [26].

Theorem 2. Consider the MAS with Assumptions 1–2, one introduces a sliding-mode surface as (8), and designs a formation control protocol in the form of

$$u_i = \frac{1}{\sum_{i=1}^N a_{ij} + b_j} \left(\sum_{i=1}^N a_{ij} u_j + b_i u_0 - \sum_{j=1}^M k_j (\lceil e_{i,j} \rceil^{p_j} + \lceil e_{i,j} \rceil^{q_j}) - \rho_0 \text{sign}(s_i) - (a_1 + a_2 e^{-|s_i|}) \lceil s_i \rceil^{b_1 + b_2 \text{sign}(|s_i| - 1)} - a_3 \lceil s_i \rceil^{b_3} \right) - \hat{d}_i \quad (9)$$

where a_1, a_2, a_3 are positive constants and b_1, b_2, b_3 are selected by $b_1 + b_2 > 1, 0 < b_1 - b_2 < 1, b_3 > 1, \rho_0 > \|H\| \Theta$. Then, the error system (4) achieves fixed-time stability within a settling time given by $t_0 \leq \frac{2}{\bar{a}_1(b_3-1)} + \frac{2}{\bar{a}_2(1-b_1+b_2)}$.

Proof. Let $\mathbf{s} = [s_1, s_2, \dots, s_n]^T$. The formation control protocol can be rephrased in a compact form as follows

$$\mathbf{u} = H^{-1} \left(\mathbf{b} u_0 - \sum_{j=1}^M k_j (\lceil \mathbf{e}_j \rceil^{p_j} + \lceil \mathbf{e}_j \rceil^{q_j}) - \rho_0 \text{sign}(\mathbf{s}) - W(\mathbf{s}) - a_3 \lceil \mathbf{s} \rceil^{b_3} \right) - \hat{\mathbf{d}} \quad (10)$$

where $W(\mathbf{s}) = [W_1(s_1), W_2(s_1), \dots, W_n(s_n)]^T$ with $W_i(s_i) = (a_1 + a_2 e^{-|s_i|}) \lceil s_i \rceil^{b_1 + b_2 \text{sign}(|s_i| - 1)}$. Choose a Lyapunov function as $V_2 = \mathbf{s}^T \mathbf{s}$, one has

$$\begin{aligned} \dot{V}_2 &= 2\mathbf{s}^T \left(\dot{\mathbf{e}}_n + \sum_{j=1}^M k_j (\lceil \mathbf{e}_j \rceil^{p_j} + \lceil \mathbf{e}_j \rceil^{q_j}) \right) \\ &= 2\mathbf{s}^T \left(H \cdot \left(\mathbf{u} + \mathbf{d} + \sum_{j=1}^M k_j (\lceil \mathbf{e}_j \rceil^{p_j} + \lceil \mathbf{e}_j \rceil^{q_j}) \right) \right) \\ &= 2\mathbf{s}^T \left(-W(\mathbf{s}) - a_3 \lceil \mathbf{s} \rceil^{b_3} + (H \cdot \tilde{\mathbf{d}} - \rho_0 \text{sign}(\mathbf{s})) \right) \\ &\leq 2\mathbf{s}^T \left(-W(\mathbf{s}) - a_3 \lceil \mathbf{s} \rceil^{b_3} \right) \end{aligned} \quad (11)$$

Case 1 When $\|\mathbf{s}\| > 1$, one has $V_2 > 1$. Consequently, (11) can be rewritten as

$$\begin{aligned} \dot{V}_2 &\leq -2a_1 N^{-b_3} \left(\sum_{i=1}^N s_i \right)^{b_3+1} \\ &\leq -2a_1 N^{-b_3} \|\mathbf{s}\|^{b_3+1} \\ &= -\bar{a}_1 V_2^{\frac{b_3+1}{2}} \end{aligned} \quad (12)$$

where $\bar{a}_1 = 2a_1 N^{-b_3}$. Hence, solving the equation $\dot{V}_2 = -\bar{a}_1 V_2^{\frac{b_3+1}{2}}$ provides the upper limit of the settling time. Then, the state will converge into the set $\{\mathbf{s} | V_2 \leq 1\}$ within $t_1 \leq \frac{2}{\bar{a}_1(b_3-1)}$.

Case 2 When $V_2 \leq 1, \|\mathbf{s}\| \leq 1$, then (11) can be reformulated as

$$\begin{aligned} \dot{V}_2 &\leq -2a_1 N^{b_2-b_1} \left(\sum_{i=1}^N s_i \right)^{b_1-b_2+1} \\ &\leq -2a_1 N^{b_2-b_1} \|\mathbf{s}\|^{b_1-b_2+1} \\ &= -\bar{a}_2 V_2^{\frac{b_1-b_2+1}{2}} \end{aligned} \quad (13)$$

with $\bar{a}_2 = 2a_1N^{b_2-b_1}$. Similarly, the settling time can be calculated by

$$t_2 = \frac{2}{\bar{a}_2(1-b_1+b_2)} V_2^{1-\frac{b_1-b_2+1}{2}}(0) \leq \frac{2}{\bar{a}_2(1-b_1+b_2)}$$

In summary, it can be obtained that the error system (4) will converge to origin within a fixed time $t_0 \leq t_1 + t_2$. $\square$

Remark 1. The designed formation control protocol (9) not only facilitates the realization of fixed-time control, allowing the estimation of an upper bound on the settling time in scenarios where the initial system states are unknown, but also serves to augment the convergence rate of the HMAS (1) and (2). When $\|s\| \leq 1$, the variable coefficient term $a_1 + a_2e^{-|s_i|}$ mainly achieves a fast convergence rate of the system while $\|s\| > 1$, and the variable exponent term $b_1 + b_2\text{sign}(|s_i| - 1)$ mainly serves to regulate the convergence rate. In consequence, the control protocol designed in this work possesses the capability to achieve both fixed-time stability and faster convergence speed.

4. Fixed-Time Formation Control for a Multi-WMR System

Take into account an MAS composed of N WMRs. All WMRs in possession of the identical mechanical structure are depicted in Figure 1.

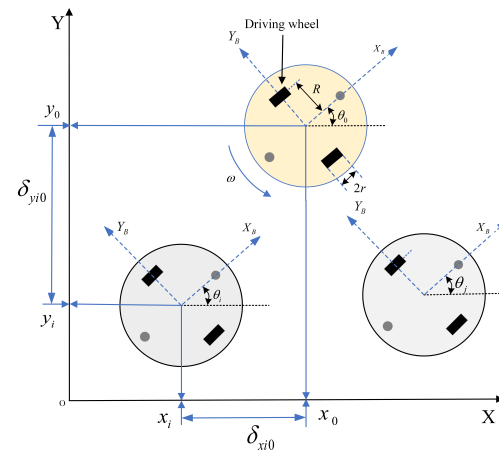


Figure 1. Leader–follower formation model of multi-WMRs.

The dynamic model of the i th WMR is described by [15]

$$\begin{cases} \dot{x}_i = v_i \cos \theta_i \\ \dot{y}_i = v_i \sin \theta_i \\ \dot{\theta}_i = \omega_i \\ \dot{\omega}_i = u_{1i} + d_{i1} \\ \dot{v}_i = u_{2i} + d_{i2} \end{cases} \quad (14)$$

where v_i, ω_i are the linear and angular velocity. Define the actual posture of the i th WMR as $P_c = [x_i, y_i, \theta_i]^T$, (x_i, y_i) is the position, θ_i is the attitude information, $d_1 = [d_{11}, d_{21}, \dots, d_{n1}]^T$ and $d_2 = [d_{12}, d_{22}, \dots, d_{n2}]^T$ are external disturbances of the WMR. Moreover, the dynamic model of the leader $P_l = [x_0, y_0, \theta_0]^T$ is given as

$$\begin{cases} \dot{x}_0 = v_0 \cos \theta_0 \\ \dot{y}_0 = v_0 \sin \theta_0 \\ \dot{\theta}_0 = \omega_0 \\ \dot{\omega}_0 = u_{10} \\ \dot{v}_0 = u_{20} \end{cases} \quad (15)$$

To achieve the formation mission, introduce the following desired formation pattern

$$\begin{cases} P_{ix} = x_i - x_0 - \delta_{x_{i0}} \\ P_{iy} = y_i - y_0 - \delta_{y_{i0}} \\ P_{i\theta} = \theta_i - \theta_0 \end{cases} \quad (16)$$

where $\delta_{x_{i0}}$ and $\delta_{y_{i0}}$ are constant values that denote desired distance of the i th followers from the leader (see in Figure 2), respectively. Then, based on the geometric relationship, the error variables of WMR formation tracking are defined as follows

$$\begin{bmatrix} x_{ie} \\ y_{ie} \\ \theta_{ie} \end{bmatrix} = \begin{bmatrix} \cos \theta_i & \sin \theta_i & 0 \\ -\sin \theta_i & \cos \theta_i & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} P_{ix} \\ P_{iy} \\ P_{i\theta} \end{bmatrix} \quad (17)$$

Furthermore, the i th WMR's error dynamics system can be altered as

$$\begin{cases} \dot{x}_{ie} = \omega y_{ie} - v_i + v_0 \cos \theta_{ie} \\ \dot{y}_{ie} = v_0 \sin \theta_{ie} - \omega_i x_{ie} \\ \dot{\theta}_{ie} = \omega_i - \omega_0 \\ \dot{\omega}_i = u_{i1} + d_{i1} \\ \dot{v}_i = u_{i2} + d_{i2} \end{cases} \quad (18)$$

Then, by introducing the following state transformation [15]: $x_{i1} = \theta_{ie}$, $x_{i2} = \bar{\omega}_{ie}$, $x_{i3} = y_{ie}$, $x_{i4} = -\omega_0 x_{ie}$, $x_{i5} = -\omega_0^2 x_{i3} + \omega_0(v_i - v_0) + \dot{\omega}_0 \frac{x_{i2}}{\omega_0}$, with $\bar{\omega}_{ie} = \omega_0 - \omega_i$, (18) can be transformed into two subsystems

$$\begin{cases} \dot{x}_{i1} = x_{i2} \\ \dot{x}_{i2} = \dot{\omega}_0 - u_{i1} - d_{i1} \end{cases} \quad (19a)$$

$$\begin{cases} \dot{x}_{i3} = x_{i4} - \frac{1}{\omega_0} x_{i4} x_{i2} + v_0 \sin x_{i1} \\ \dot{x}_{i4} = x_{i5} + x_{i3} x_{i2} \omega_0 - \omega_0 v_0 (\cos x_{i1} - 1) \\ \dot{x}_{i5} = -\omega_0 \dot{\omega}_0 x_{i3} + \left(\frac{\ddot{\omega}_0}{\omega_0} - \omega_0^2 - \frac{2\dot{\omega}_0^2}{\omega_0^2} \right) x_{i4} \\ \quad + \omega_0 u_{i2} + \omega_0 d_{i2} - \omega_0 \dot{v}_0 + x_{i4} x_{i2} \omega_0 + \frac{2\dot{\omega}_0}{\omega_0} x_{i5} \\ \quad - \omega_0^2 v_0 \sin x_{i1} + \dot{\omega}_0 x_{i3} x_{i2} - \dot{\omega}_0 v_0 (\cos x_{i1} - 1) \end{cases} \quad (19b)$$

In this part, the aim of the control is to design distributed formation protocols that ensure the stability of two subsystems and achieve multi-WMR formation tracking control.

Assumption 4. d_1 and d_2 exist upper bounds defined by $\|d_1\| \leq \bar{k}_1$, $\|d_2\| \leq \bar{k}_2$, where $\bar{k}_1$, $\bar{k}_2$ are positive constants.

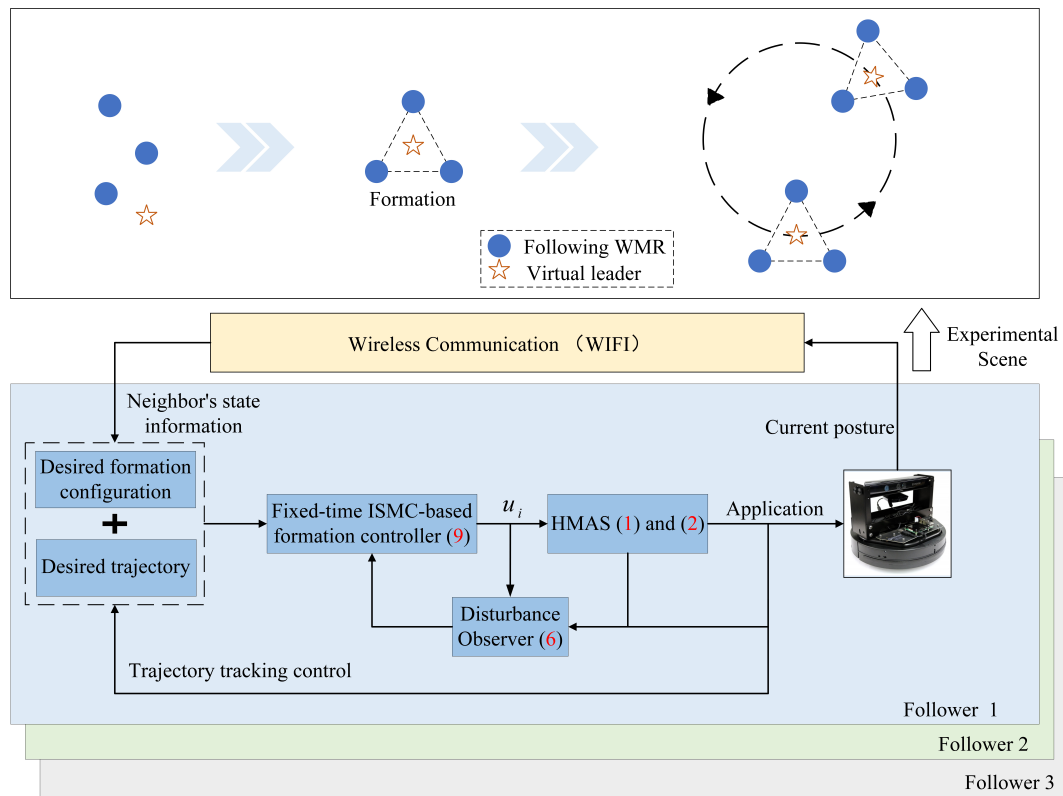


Figure 2. Frame of control scheme and QBot 2e platform.

4.1. Fixed-Time Disturbance Observers

To design the disturbance observers, we define the auxiliary variables as $\varsigma_{i1} = \omega_i - \omega_{i1}$ and $\varsigma_{i2} = v_i - \omega_{i2}$, where $\omega_{i\iota} (\iota = 1, 2)$ satisfies

$$\dot{\omega}_{i\iota} = u_{i\iota} + l_{i1} \operatorname{erf}\left(\frac{\varsigma_{i\iota}}{\epsilon_i}\right) + l_{i2} [\varsigma_{i\iota}]^{\gamma_{i1} + \gamma_{i2} \operatorname{sign}(|\varsigma_{i\iota}| - 1)} + l_{i3} [\varsigma_{i\iota}]^{\gamma_{i3}} \quad (20)$$

where l_{i1}, l_{i2}, l_{i3} are positive constants and $l_{i1} > k_i$, ϵ_i are small positive constants. $\gamma_{i1}, \gamma_{i2}, \gamma_{i3}$ meet the constraints $\gamma_{i1} + \gamma_{i2} > 1$, $0 < \gamma_{i1} - \gamma_{i2} < 1$, and $\gamma_{i3} > 1$.

For the i th agent, the continuous disturbance observers are designed as

$$\hat{d}_{i\iota} = l_{i1} \operatorname{erf}\left(\frac{\varsigma_{i\iota}}{\epsilon_i}\right) + l_{i2} [\varsigma_{i\iota}]^{\gamma_{i1} + \gamma_{i2} \operatorname{sign}(|\varsigma_{i\iota}| - 1)} + l_{i3} [\varsigma_{i\iota}]^{\gamma_{i3}} \quad (21)$$

Similar to Theorem 1, the estimation error $\tilde{d}_{i\iota} = d_{i\iota} - \hat{d}_{i\iota}$ will converge to a small neighborhood around the origin in the fixed time T_{d_i} .

4.2. Distributed Formation Control for the Second-Order Subsystem

For the attitude error systems (19a), we define two error variables as $\tilde{\theta}_i = \theta_i - \theta_0$, $\tilde{\omega}_i = \omega_i - \omega_0$. Furthermore, the consensus errors are introduced as

$$\begin{cases} e_{i1} = \sum_{j=1}^N a_{ij}(x_{i1} - x_{j1}) + b_i x_{i1} \\ e_{i2} = \sum_{j=1}^N a_{ij}(x_{i2} - x_{j2}) + b_i x_{i2} \end{cases} \quad (22)$$

Obviously, one can deduce from (22) that $\dot{e}_{i1} = e_{i2}$. In order to design a distributed controller for the second-order subsystem, we introduce a fixed-time sliding-mode surface, which is defined as $s_{i1} = e_{i2} + \int_0^t \sum_{j=1}^2 k_j ([e_{ij}]^{p_j} + [e_{ij}]^{q_j}) d\tau$, where $i = 1, 2, \dots, n$, k_j are

positive constants. The parameters p_j, q_j are positive odd integers, satisfying $0 < p_j < 1$, $q_j > 1$.

Then, the fixed-time control protocol for the second-order system is formulated as follows:

$$u_{i1} = -\hat{d}_{i1} + \frac{1}{\sum_{i=1}^N a_{ij} + b_j} \left\{ \sum_{j=1}^N a_{ij} u_{j1} + b_i \dot{\omega}_0 - \sum_{j=1}^2 k_j (\lceil e_{ij} \rceil^{p_j} + \lceil e_{ij} \rceil^{q_j}) - (m_2 + m_1 e^{-|s_{i1}|}) \lceil s_{i1} \rceil^{\alpha_1 + \alpha_2 \text{sign}(|s_{i1}| - 1)} - m_3 \lceil s_{i1} \rceil^{\alpha_3} - \rho_1 \text{sign}(s_{i1}) \right\} \quad (23)$$

where m_1, m_2, m_3 are positive constants. $\alpha_1, \alpha_2, \alpha_3$ satisfy $\alpha_1 + \alpha_2 > 1$, $0 < \alpha_1 - \alpha_2 < 1$, and $\alpha_3 > 1$. Select ρ_1 to meet the constraint $\rho_1 \geq \|H\| \|\tilde{d}_1\|$, which is relevant to the external disturbance d_{i1} . The proposed distributed formation protocol (23) guarantees the consensus errors e_{i1} can converge to be stable along s_{i1} within a fixed time, which is denoted as T_1 .

Then, we can obtain that attitude error variables θ_{ie} and ω_{ie} can converge to zero in a fixed time. When $t > T_1$, the third-order subsystem (19b) can be simplified as

$$\begin{cases} \dot{x}_{i3} = x_{i4} \\ \dot{x}_{i4} = x_{i5} \\ \dot{x}_{i5} = -\omega_0 \dot{\omega}_0 x_{i3} + \left(\frac{\ddot{\omega}_0}{\omega_0} - \omega_0^2 - \frac{2\dot{\omega}_0^2}{\omega_0^3} \right) x_{i4} \\ \quad + \omega_0 u_{i2} + \omega_0 d_{i2} - \omega_0 \dot{v}_0 + \frac{2\dot{\omega}_0}{\omega_0} x_{i5} \end{cases} \quad (24)$$

4.3. Distributed Formation Control for the Third-Order Subsystem

Similar to the procedure followed in designing the control protocol for the second-order subsystem, we can define the subsequent error variables:

$$\begin{cases} e_{i3} = \sum_{i=1}^N a_{ij} (x_{i3} - x_{j3}) + b_i x_{i3} \\ e_{i4} = \sum_{i=1}^N a_{ij} (x_{i4} - x_{j4}) + b_i x_{i4} \\ e_{i5} = \sum_{i=1}^N a_{ij} (x_{i5} - x_{j5}) + b_i x_{i5} \end{cases} \quad (25)$$

Likewise, it can be obtained that (25) is a continuous third-order system. By designing a fixed-time sliding-mode surface for (25) as $s_{i2} = e_{i5} + \int_0^t \sum_{j=3}^5 k_j (\lceil e_{ij} \rceil^{p_j} + \lceil e_{ij} \rceil^{q_j}) d\tau$, we construct a fixed-time control protocol for the system (25) in the form of

$$u_{i2} = \frac{1}{\sum_{i=1}^N a_{ij} + b_j} \left\{ \sum_{j=1}^N a_{ij} u_{j2} + \frac{1}{\omega_0} \left(b_i \dot{v}_0 - \sum_{j=3}^5 k_j (\lceil e_{ij} \rceil^{p_j} + \lceil e_{ij} \rceil^{q_j}) \right) + \left(\frac{\ddot{\omega}_0}{\omega_0} - 2 \frac{\dot{\omega}_0^2}{\omega_0^2} - \omega_0^2 \right) x_{i4} + 2 \frac{\dot{\omega}_r x_{i5}}{\omega_0} - \dot{\omega}_0 \omega_0 x_{i3} - \rho_2 \text{sign}(s_{i2}) - (m_5 + m_4 e^{-|s_{i2}|}) \lceil s_{i2} \rceil^{\beta_1 + \beta_2 \text{sign}(|s_{i2}| - 1)} - m_6 \lceil s_{i2} \rceil^{\beta_3} \right\} - \hat{d}_{i2} \quad (26)$$

where m_4, m_5, m_6 are positive constants. $\beta_1, \beta_2, \beta_3$ satisfy $\beta_1 + \beta_2 > 1$, $0 < \beta_1 - \beta_2 < 1$, and $\beta_3 > 1$. ρ_2 is selected as a positive constant surpassing $\tilde{d}_{i2}$, that is, $\rho_2 \geq \|H\| \|\tilde{d}_2\|$.

According to Theorem 2, for the third-order error system in (25), the sliding-mode surface s_2 will converge to a small set within a fixed time. Furthermore, within a fixed time marked as T_2 , the position error variables can converge to zero along the sliding-mode surface.

5. Experimental Results

To confirm the efficiency of the proposed control scheme, the formation trajectory tracking experiment is implemented on the QBot 2e mobile robot platform. The configuration block diagram illustrating the control scheme and the application scenario are depicted in Figure 2. The experimental platform primarily comprises three QBot 2e robots, one central host computer, a set of infrared motion capture systems for real-time pose information acquisition, and a wireless communication router to enable wireless connectivity.

The experiment aims to realize that three QBot 2e robots start from random positions and maneuver to achieve a stable formation in the shape of an equilateral triangle. The robots autonomously converge to form a stable equilateral triangle formation and track a circular trajectory with a predefined radius of 0.4 m.

The directed topology graph of multiple WMRs is depicted in Figure 3, with the leader identified as 0 and three followers labeled as 1, 2, and 3. The connected topology indicates that only part of the followers have access to information transmitted by the leader. In the experiment, the relevant physical parameters of QBot 2e are given as $m = 4$ kg, $J = 2.5$ kg · m². The parameters of the designed fixed-time disturbance observers for two subsystems are selected as $\gamma_1 = 1$, $\gamma_2 = 0.01$, $\gamma_3 = 1$, $\gamma_4 = 0.01$, $\epsilon_1 = 0.1$, $\epsilon_2 = 0.1$. And the main parameters of the proposed distributed formation controllers for multi WMRs are given by $k_1 = 4.5$, $k_2 = 4$, $k_3 = 2$, $k_4 = 74$, $k_5 = 4$, $\alpha_1 = 0.01$, $\beta_1 = 1$, $\alpha_2 = 0.01$, $\beta_2 = 1$, $p_1 = 0.53$, $q_1 = 1.3$, $p_2 = 0.53$, $q_2 = 1.3$.

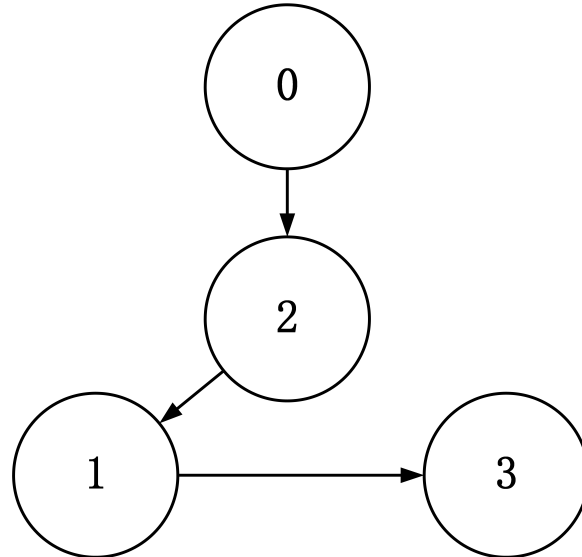


Figure 3. Communication topology among WMRs.

Figure 4 plots the results of the formation tracking experiment of three WMRs, where the green circle represents the desired trajectory of the virtual leader. The red, blue, and cyan circles portray the trajectory of three followers labeled by 1, 2, and 3. The black dashed lines depict the configurations of three QBot 2e robots at their initial positions, which form an irregular triangle.

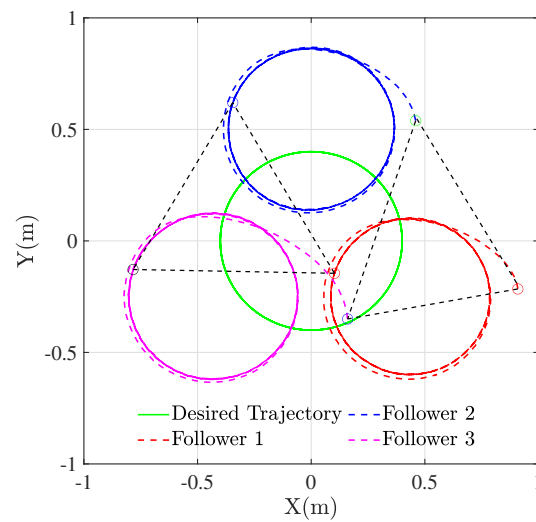


Figure 4. Trajectory tracking for multiple WMRs.

Through the designed control framework, it can be observed that, once the system stabilizes, the entire formation system of WMRs forms the desired equilateral triangle. Moreover, they all track the expected circular trajectory with a radius of 0.4 m. Figures 5–7 illustrate the experimental tracking errors X_e , Y_e , and Θ_e between the virtual leader and followers. As can be observed from the figure, the tracking error of the formation system can be stabilized and converged to a smaller value around 0 in about 15 s, which indicates a fast convergence rate, high accuracy, and small variability. This is achieved through the variable coefficient and variable exponent terms proposed by the controller (26). Furthermore, the efficacy of the proposed formation control scheme is validated through the results of tracking errors.

As depicted in Figure 8, the control inputs in the experiments can oscillate within an appropriate range by utilizing the designed formation protocols. At the same time, the stable control inputs ensure that each WMR can establish a formation configuration while tracking the desired trajectory of the leader.

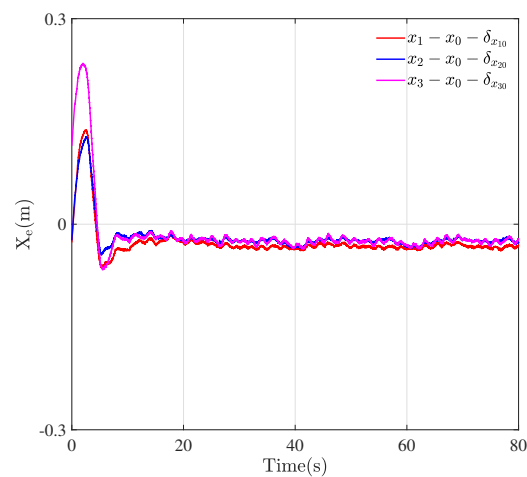


Figure 5. Tracking errors of X_e .

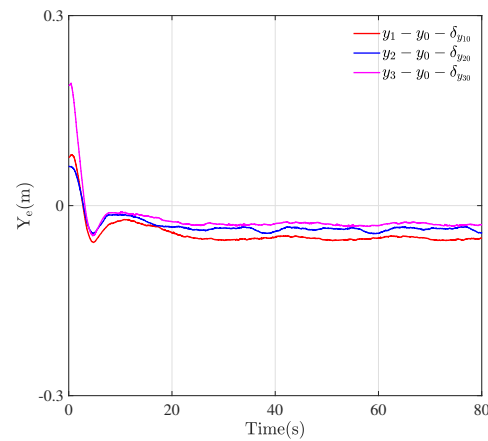


Figure 6. Tracking errors of Y_e .

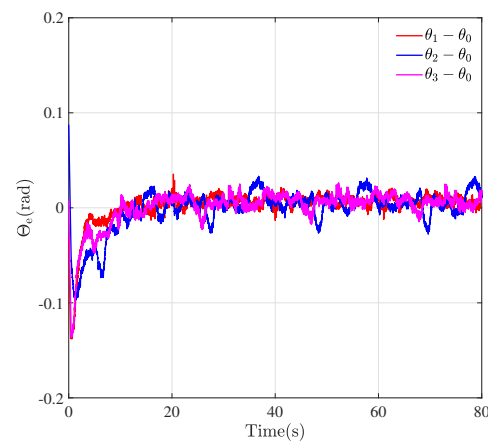


Figure 7. Tracking errors of Θ_e .

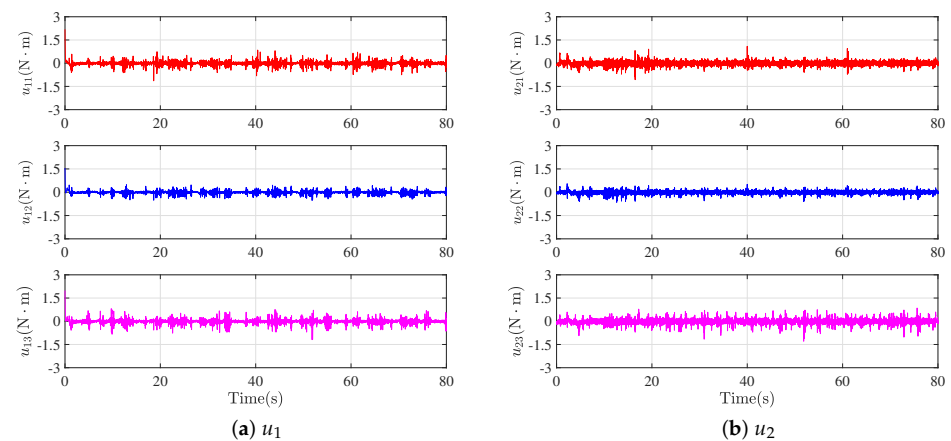


Figure 8. Response curves of control inputs in experiment.

The observed values $\hat{d}_1$ and $\hat{d}_2$ in the experimental environment are provided in Figure 9. In this experiment, the external disturbances may arise from factors such as uneven laboratory ground, wind effects, sensor measurement errors, and so on. The data displayed in Figure 9 clearly show that the error states are bounded after the system is stable. The observers demonstrate a high degree of accuracy in their estimation of the disturbances in the external environment.

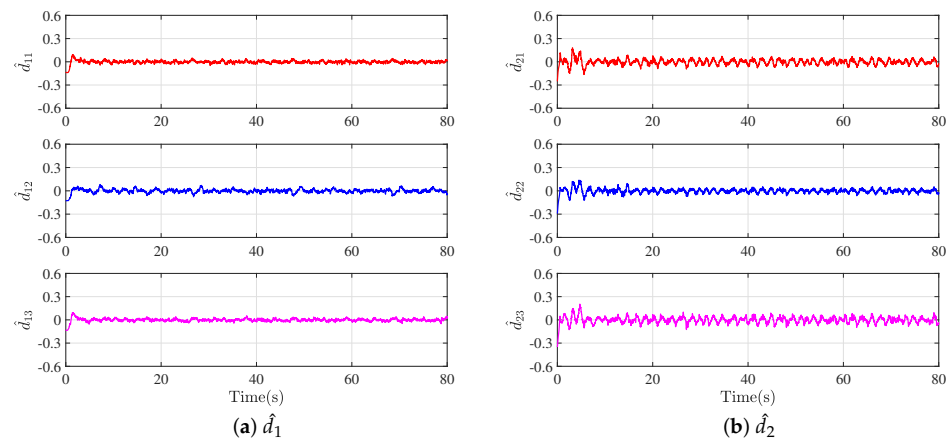


Figure 9. External disturbance estimation in experiment.

6. Conclusions

This article investigates the problem of distributed fixed-time tracking control for MASs and applies it to a multi-WMR system. Firstly, a fixed-time disturbance observer is designed for a third-order continuous integration system with external disturbances, which can effectively attenuate the chattering phenomenon. Then, a distributed controller is devised to achieve the formation tracking of a third-order MAS by combining the integral sliding-mode methodology. Finally, the designed control scheme is applied to a mobile robot platform for experimental validation. Nevertheless, the design and experimental implementation of the entire control method rely on the assumption of ideal communication conditions. Further study will focus on communication delays in formation control for MASs.

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Abbreviations

The following abbreviations are used in this manuscript:

MAS	Multi-agent system
WMR	Wheeled mobile robot
HMAS	High-order multi-agent system

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