

Bearing-Only Formation Control With Prespecified Convergence Time

Zhenhong Li^{ID}, Hilton Tnunay^{ID}, Shiyu Zhao, Wei Meng^{ID},
Sheng Q. Xie^{ID}, *Senior Member, IEEE*, and Zhengtao Ding^{ID}, *Senior Member, IEEE*

Abstract—This article considers the bearing-only formation control problem, where the control of each agent only relies on relative bearings of their neighbors. A new control law is proposed to achieve target formations in finite time. Different from the existing results, the control law is based on a time-varying scaling gain. Hence, the convergence time can be arbitrarily chosen by users, and the derivative of the control input is continuous. Furthermore, sufficient conditions are given to guarantee almost global convergence and interagent collision avoidance. Then, a leader–follower control structure is proposed to achieve global convergence. By exploring the properties of the bearing Laplacian matrix, the collision avoidance and smooth control input are preserved. A multirobot hardware platform is designed to validate the theoretical results. Both simulation and experimental results demonstrate the effectiveness of our design.

Index Terms—Bearing-only formation control, finite-time formation control, prescribed-time consensus.

I. INTRODUCTION

FORMATION control, as an important realm of multiagent cooperative control, has been extensively studied in recent decades [1]. In the literature (see [2]–[9]), numerous control laws have been designed to achieve target formations with

the assumption that relative positions or distances between agents are measurable. However, this assumption is not always easy to satisfy, especially when agents have no access to an external localization system [10]. Recently, the bearing-only control laws have been proposed and attracted much attention (see [11]–[18]). Instead of relative positions and distances, target formations of bearing-only control laws are defined by relative bearings that can be obtained by vision sensors [19] or wireless sensor arrays [20]. Due to the accessibility of relative bearings, bearing-only control laws provide potential solutions to achieve formation control merely using onboard sensing.

In 2-D space, some early results on bearing-constrained formation control can be found in [11] and [12]. Based on the parallel rigidity theory, Bishop *et al.* [11] introduced the bearing-constrained rigidity matrix and proposed a control law with locally asymptotic stability. Although the target formation is defined by relative bearings, the measurements of relative positions are still required. This requirement is then removed by introducing a decentralized position estimator [12]. To achieve bearing-only formation control in high-dimensional space, Zhao and Zelazo [13] extended the bearing rigidity theory to arbitrary dimensions and proposed a control law for infinitesimally bearing rigid formations with almost global asymptotic stability. To further characterize the algebraic properties of bearing rigid formations, the bearing Laplacian matrix is proposed in [14]. This matrix can be used to examine the uniqueness of target formations in arbitrary dimensions. Based on this powerful tool, a new bearing-only control law is designed in the recent work [15], and global exponential convergence is guaranteed.

Due to the time requirement of many formation control tasks, convergence time is regarded as an important performance indicator. To achieve faster convergence rate, finite-time control has been widely studied in multiagent systems (see [8], [21]–[26]). However, the intrinsic nonlinearity of bearing vectors makes the finite-time convergence analysis of bearing-only control nontrivial. A few works have been done for the finite-time bearing-only formation control (see [16]–[18]). Zhao *et al.* [16] used signum functions to suppress relative bearing errors and hence achieved finite-time convergence. However, the results can only be applied to cyclic formations. Instead of signum functions, the controllers in [17] and [18] use fractional power bearing feedback and achieve almost global convergence for infinitesimally bearing rigid formations. However, the convergence time of the

Manuscript received September 26, 2019; revised January 24, 2020; accepted March 7, 2020. This work was supported in part by the Science and Technology Facilities Council of U.K. under Grant ST/N006852/1, in part by the Engineering and Physical Sciences Research Council of U.K. under Grant EP/S019219/1, and in part by the National Natural Science Foundation of China under Grant 6190020021, Grant 61903308, and Grant 51705381. This article was recommended by Associate Editor X. M. Sun. (Corresponding author: Zhengtao Ding.)

Zhenhong Li is with the School of Electronic and Electrical Engineering, University of Leeds, Leeds LS2 9JT, U.K. (e-mail: z.h.li@leeds.ac.uk).

Hilton Tnunay and Zhengtao Ding are with the Department of Electrical and Electronic Engineering, University of Manchester, Manchester M13 9PL, U.K. (e-mail: hilton.tnunay@manchester.ac.uk; zhengtao.ding@manchester.ac.uk).

Shiyu Zhao is with the School of Engineering, Westlake University, Hangzhou 310024, China, and also with the Institute of Advanced Technology, Westlake Institute for Advanced Study, Hangzhou 310024, China (e-mail: zhaoshiyu@westlake.edu.cn).

Wei Meng is with the School of Information Engineering, Wuhan University of Technology, Wuhan 430070, China, and also with the School of Electronic and Electrical Engineering, University of Leeds, Leeds LS2 9JT, U.K. (e-mail: w.meng@leeds.ac.uk).

Sheng Q. Xie is with the School of Electronic and Electrical Engineering, University of Leeds, Leeds LS2 9JT, U.K., also with the Institute of Rehabilitation Engineering, Binzhou Medical University, Yantai 264033, China, and also with the Qingdao University of Technology, Qingdao 266033, China (e-mail: s.q.xie@leeds.ac.uk).

Color versions of one or more of the figures in this article are available online at <http://ieeexplore.ieee.org>.

Digital Object Identifier 10.1109/TCYB.2020.2980963

forementioned results is all determined by initial conditions, and hence cannot be prespecified by users. Moreover, the use of signum functions and fractional power feedback will lead to nonsmooth control input. In other words, bearing-only formation control in prespecified finite time remains an open problem.

In this article, we investigate bearing-only formation control with prespecified convergence time. A new control law is proposed for leaderless formation control by introducing a time-varying gain to the regular feedback of relative bearings. Sufficient conditions are derived to achieve almost global finite-time convergence while avoiding collisions. Different from the results in [16]–[18], the convergence time can be prespecified and arbitrarily chosen by users. Furthermore, since no fractional power feedback is used, the control input is C^1 smooth everywhere. The design of time-varying gain is partly inspired by the work on finite-time regulation of nonlinear systems [27]. However, different from relative position-based formation control, for bearing-only formation control, relative bearing vectors are unit vectors, that is, a smaller position error does not imply a smaller bearing error. This phenomenon makes it difficult to establish the boundedness of control input especially when the time-varying gain is unbounded, which implies the stability analysis method in [27] cannot be directly applied to our case. Then, we design a leader–follower control structure for our proposed control law. By further exploring the properties of bearing Laplacian matrix, we prove that, with the leader–follower control structure, the global convergence can be achieved in prespecified finite time while avoiding the collisions (rather than the almost global convergence in [13]). Finally, a multirobot hardware platform is designed, and both simulation and experimental results verify the effectiveness of the proposed control laws.

The remainder of this article is organized as follows. Section II introduces some necessary preliminaries and problem setup. Sections III and IV present the main results on the control law design and stability analysis for the leaderless case and leader–follower case, respectively. The simulation results and experimental validation are given in Sections V and VI. Conclusions are drawn in Section VII.

II. PRELIMINARIES AND PROBLEM STATEMENT

A. Notations

Let $\mathbb{R}_{>0}$ denote the set of positive real numbers. $I_n \in \mathbb{R}^{n \times n}$ denotes the identity matrix, and $\mathbf{1}_n$ denotes an n -dimensional column vector with all elements are equal to one. For a series of column vectors x_1, \dots, x_n , $\text{col}(x_1, \dots, x_n)$ represents a column vector by stacking them together; $\text{span}\{x_1, \dots, x_n\}$ represents the linear span of the vectors. For a matrix A , $A > 0$ (or $A \geq 0$) means that A is positive definite (or positive semidefinite); $\lambda_i(A)$ is the i th eigenvalue of A ; and $\text{null}(A)$ and $\text{range}(A)$ are the null and range spaces of A , respectively. For a series of matrices A_1, \dots, A_n , $\text{diag}(A_i)$ denotes the block-diagonal matrix with diagonal blocks A_1, \dots, A_n . $\|\cdot\|$ represents the Euclidean norm of a vector or the spectral norm of a matrix. \otimes denotes the Kronecker product of matrices.

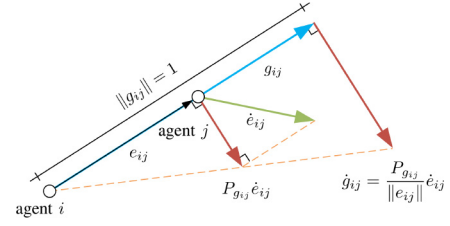


Fig. 1. Geometric relationship between g_{ij} , \dot{g}_{ij} , e_{ij} , and \dot{e}_{ij} .

B. Preliminaries

Consider a group of n mobile agents in \mathbb{R}^d ($n \geq 2$ and $d \geq 2$). Let $p_i(t) \in \mathbb{R}^d$ be the position of the agent i at time t . The *configuration* of the agents is denoted as $p = \text{col}(p_1, \dots, p_n) \in \mathbb{R}^{dn}$. The interaction among agents is described by an undirected graph $\mathcal{G} = \{\mathcal{V}, \mathcal{E}\}$, where $\mathcal{V} = \{1, \dots, n\}$ is the set of vertices and $\mathcal{E} \subseteq \mathcal{V} \times \mathcal{V}$ is the set of edges. The edge $(i, j) \in \mathcal{E}$ if agent i can measure the relative bearing of agent j . Since the graph is undirected, we have $(i, j) \in \mathcal{E} \Leftrightarrow (j, i) \in \mathcal{E}$. The *formation*, denoted as (\mathcal{G}, p) , is \mathcal{G} with each vertex $i \in \mathcal{V}$ mapped to the point p_i . The set of neighbors of agent i is denoted as $\mathcal{N}_i = \{j \in \mathcal{V} : (i, j) \in \mathcal{E}\}$. An *orientation* of an undirected graph is the assignment of a direction to each edge. An *oriented graph* is an undirected graph with an orientation. Let m be the number of undirected edges. Then, the oriented graph has m directed edges. The *incidence matrix* of the oriented graph is denoted as $H \in \mathbb{R}^{m \times n}$, where $[H]_{ki} = -1$ if vertex i is the tail of edge k ; $[H]_{ki} = 1$ if vertex i is the head of edge k ; and $[H]_{ki} = 0$ otherwise. For an undirected connected graph, it holds that $H\mathbf{1}_n = \mathbf{0}$ and $\text{rank}(H) = n - 1$ [28].

For edge (i, j) , we define the *edge vector* and the *bearing vector*, respectively, as

$$e_{ij} := p_j - p_i, \quad g_{ij} := \frac{e_{ij}}{\|e_{ij}\|}$$

where g_{ij} represents the relative bearing of p_j with respect to p_i . Obviously, we have $e_{ij} = -e_{ji}$, $g_{ij} = -g_{ji}$, and $\|g_{ij}\| = 1$. For any nonzero vector $x \in \mathbb{R}^d$, define the operator $P : \mathbb{R}^d \rightarrow \mathbb{R}^{d \times d}$ as

$$P_x := I_d - \frac{xx^T}{x^T x}$$

where P_x is an orthogonal projection matrix that can geometrically project any vector onto the orthogonal complement of x . Note that P_x is positive semidefinite, $P_x^2 = P_x$, and $\text{null}(P_x) = \text{span}(x)$. It follows that $P_x y = 0 \forall y \in \mathbb{R}^d \Leftrightarrow y$ is parallel to x . Since P_x can be used to check whether two bearings are parallel, it is widely used in bearing-based control [14], [29]. Direct evaluation gives

$$\dot{g}_{ij} = \frac{P_{g_{ij}}}{\|e_{ij}\|} \dot{e}_{ij}.$$

Together with $g_{ij}^T P_{g_{ij}} = \mathbf{0}$, we have that $g_{ij}^T \dot{g}_{ij} = \mathbf{0}$ and $e_{ij}^T \dot{g}_{ij} = \mathbf{0}$. Fig. 1 shows the geometric relationship between g_{ij} , \dot{g}_{ij} , e_{ij} , and \dot{e}_{ij} when $\|e_{ij}\| < 1$.

Suppose (i, j) corresponds to the k th directed edge in the oriented graph, where $k \in \{1, \dots, m\}$. The edge and bearing

vectors of the k th directed edge are defined as

$$e_k := e_{ij} = p_j - p_i, \quad g_k := \frac{e_k}{\|e_k\|}.$$

Similarly, we have $g_k^T \dot{g}_k = \mathbf{0}$ and $e_k^T \dot{g}_k = \mathbf{0}$. It follows from the definition of H that $e = \bar{H}p$, where $e = \text{col}(e_1, \dots, e_m)$ and $\bar{H} = H \otimes I_d$.

To characterize the properties of a formation, we introduce the *bearing Laplacian* matrix $\mathcal{B}(\mathcal{G}, p) \in \mathbb{R}^{dn \times dn}$ with the (i, j) th block of submatrix as [14]

$$[\mathcal{B}(\mathcal{G}, p)]_{ij} = \begin{cases} \mathbf{0}_{d \times d}, & i \neq j, (i, j) \notin \mathcal{E} \\ -P_{g_{ij}}, & i \neq j, (i, j) \in \mathcal{E} \\ \sum_{j \in \mathcal{N}_i} P_{g_{ij}}, & i = j, i \in \mathcal{V}. \end{cases}$$

To simplify the notation, we use \mathcal{B} instead of $\mathcal{B}(\mathcal{G}, p)$. According to the definition of bearing the Laplacian matrix, we have that $\mathcal{B} \geq 0$, $\mathcal{B}p = \mathbf{0}$, $\mathcal{B}\mathbf{1}_{dn} = \mathbf{0}$, and $\mathcal{B} = \bar{H}^T \text{diag}(P_{g_k}) \bar{H}$. Letting (\mathcal{G}, p) and (\mathcal{G}, p') be two formations with the same bearing Laplacian matrix, we give the following definition.

Definition 1 (Infinitesimal Bearing Rigidity) [13]: A formation (\mathcal{G}, p) is infinitesimally bearing rigid if $p' - p$ corresponds to translational and scaling motions $\Leftrightarrow \mathcal{B}(p' - p) = \mathbf{0}$.

The above definition implies that an infinitesimally bearing rigid formation is uniquely determined up to a translation and a scaling. Note that the definition of infinitesimal bearing rigidity in [13] is based on the bearing rigidity matrix. In this article, Definition 1 is based on the bearing Laplacian matrix \mathcal{B} .

Lemma 1 [14]: For an infinitesimally bearing rigid formation, the following properties hold:

- 1) $\text{null}(\mathcal{B}) = \text{span}\{\mathbf{1} \otimes I_d, p\}$;
- 2) $\text{rank}(\mathcal{B}) = dn - d - 1$, that is, the eigenvalues of \mathcal{B} represented as $\lambda_1(\mathcal{B}) = \dots = \lambda_{d+1}(\mathcal{B}) = 0 < \lambda_{d+2}(\mathcal{B}) \leq \dots \leq \lambda_{dn}(\mathcal{B})$;
- 3) Partition \mathcal{B} as

$$\mathcal{B} = \begin{bmatrix} \mathcal{B}_{ll} & \mathcal{B}_{lf} \\ \mathcal{B}_{lf}^T & \mathcal{B}_{ff} \end{bmatrix} \quad (1)$$

where $\mathcal{B}_{ll} \in \mathbb{R}^{dn_l}$, $\mathcal{B}_{lf} \in \mathbb{R}^{dn_l \times dn_f}$, and n_l and $n_f \in \mathbb{R}_{>0}$ satisfying $n_l + n_f = n$. Then, $\mathcal{B}_{ff} > 0$ if $n_l \geq 2$.

The above lemma bridges the gap between the rigidity of a formation and algebraic properties of the bearing Laplacian matrix, which plays an important role in the stability analysis. Lemma 1 3) implies that if more than two points of an infinitesimally bearing rigid formation are fixed then the configuration p is uniquely determined. More results on the uniqueness of infinitesimally bearing rigid formation are given in [14].

C. Problem Statement

The dynamics of mobile agents are

$$\dot{p}_i = u_i, \quad i \in \mathcal{V}$$

where $u_i \in \mathbb{R}^d$ is the velocity input of agent i . The main objective of this article is given as follows.

Problem 1: Design control input for agent $i \in \mathcal{V}$ based on the bearing vectors $\{g_{ij}(t)\}_{j \in \mathcal{N}_i}$ such that $p \rightarrow p^*$ for $t \rightarrow$

$t_0 + T$, and $p = p^*$ for $t \geq t_0 + T$, where p^* is a target configuration and $T \in \mathbb{R}_{>0}$ is a prespecified convergence time. The following assumption holds throughout this article.

Assumption 1 (Target Formation): The target formation (\mathcal{G}, p^*) is infinitesimally bearing rigid.

Remark 1: Assumption 1 is commonly used to build the connection between the target configuration p^* and the target bearing vectors $\{g_{ij}^*\}_{(i,j) \in \mathcal{E}}$ (e.g., [14] and [15]). Then, Problem 1 can be transferred into a stabilization problem of bearing vectors in prespecified finite time.

III. BEARING-ONLY LEADERLESS FORMATION CONTROL

In this section, we propose a bearing-only leaderless control law to solve Problem 1. The control law of each mobile agent is designed as

$$u_i = -\left(a + b \frac{\dot{\mu}}{\mu}\right) \sum_{j \in \mathcal{N}_i} P_{g_{ij}} g_{ij}^*, \quad i \in \mathcal{V} \quad (2)$$

where $a, b \in \mathbb{R}_{>0}$ are positive feedback gains, and $\mu : \mathbb{R}_{>0} \rightarrow \mathbb{R}_{>0}$ is a time-varying scaling function defined as

$$\mu(t) = \begin{cases} \frac{T^h}{(t_0 + T - t)^h}, & t \in [t_0, t_0 + T) \\ 1, & t \in [t_0 + T, \infty) \end{cases} \quad (3)$$

and $h \in \mathbb{R}_{>0}$ is a user-chosen parameter. Note that

$$\dot{\mu}(t) = \begin{cases} \frac{h}{T} \mu^{(1+\frac{1}{h})}, & t \in [t_0, t_0 + T) \\ 0, & t \in [t_0 + T, \infty) \end{cases}$$

where we use the right-hand derivative of $\mu(t)$ at $t = t_0 + T$ as $\dot{\mu}(t_0 + T)$. The time-varying scaling function $\mu(t)$ plays a key role in achieving prespecified finite-time control. For any $c \in \mathbb{R}_{>0}$, we have $\mu^{-c}(t_0) = 1$, $\lim_{t \rightarrow (t_0 + T)^-} \mu^{-c}(t) = 0$, and $\mu(t)^{-c}$ is monotonically decreasing on $[t_0, t_0 + T)$.

Since control law (2) is based on an implicit assumption that $g_{ij} \forall (i, j) \in \mathcal{E}$ are well defined, we make the following assumption.

Assumption 2 (Collision Avoidance): During the formation evolvment, no neighboring agents collide with each other.

Assumption 2 is widely used in the existing formation control results [30], [31], since it is nontrivial to analyze the system convergence if collision avoidance is considered. In this article, we first analyze system convergence under Assumption 2. Then, we will present sufficient conditions based on initial formation such that system convergence and collision avoidance can be simultaneously guaranteed, and hence the assumption could be dropped.

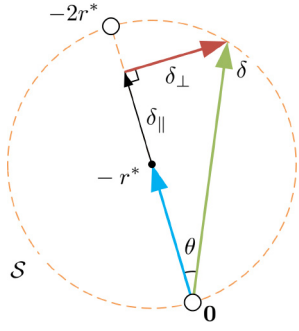
To analyze the finite-time convergence of the closed-loop system, we introduce the following lemma.

Lemma 2: Consider a continuously differentiable function $y : \mathbb{R} \rightarrow \mathbb{R}_{\geq 0}$ satisfying that

$$\dot{y}(t) \leq -\alpha y - \beta \frac{\dot{\mu}}{\mu} y, \quad t \in [t_0, \infty) \quad (4)$$

where $\alpha, \beta \in \mathbb{R}_{>0}$. Then, it follows that:

$$y(t) \begin{cases} \leq \mu^{-\beta} e^{-\alpha(t-t_0)} y(t_0), & t \in [t_0, t_0 + T) \\ \equiv 0, & t \in [t_0 + T, \infty). \end{cases}$$

Fig. 2. Geometric relationship between δ and the surface \mathcal{S} .

Proof: Multiplying μ^β on both sides of (4), we obtain

$$\mu^\beta \dot{y} \leq -\alpha \mu^\beta y - \beta \mu^{\beta-1} \dot{\mu} y.$$

Together with $[(d(\mu^\beta y))/(dt)] = \beta \mu^{\beta-1} \dot{\mu} y + \mu^\beta \dot{y}$, we have that

$$\frac{d(\mu^\beta y)}{dt} \leq -\alpha \mu^\beta y$$

which further implies that

$$\begin{aligned} \mu^\beta y(t) &\leq e^{-\alpha(t-t_0)} \mu(t_0)^\beta y(t_0) \\ &= e^{-\alpha(t-t_0)} y(t_0), \quad t \in [t_0, t_0 + T). \end{aligned} \quad (5)$$

From (5), we can obtain that $y(t) \leq e^{-\alpha(t-t_0)} \mu^{-\beta} y(t_0) \forall t \in [t_0, t_0 + T)$. By the continuity of y and $\lim_{t \rightarrow (t_0+T)^-} y(t) = 0$, we know that $y(t_0 + T) = 0$, and furthermore from $\dot{y} \leq 0 \forall t \in [t_0 + T, \infty)$, we know that $0 \leq y(t) \leq y(t_0 + T) \forall t \in [t_0 + T, \infty)$ and that $y(t) \equiv 0 \forall t \in [t_0 + T, \infty)$. ■

Before analyzing the convergence, we first show some useful properties of (2). Define *centroid* and *scale* of a formation as $\bar{p} := (1/n) \sum_{i=1}^n p_i$ and $s := \sqrt{(1/n) \sum_{i=1}^n \|p_i - \bar{p}\|^2}$.

Lemma 3: Under Assumption 2 and control law (2), \bar{p} and s are invariant. Furthermore, $\|p_i(t) - p_j(t)\| \leq 2s\sqrt{n-1} \forall i, j \in \mathcal{V} \forall t \geq t_0$.

Proof: By following the analysis in [13, Th. 9], it can be proved that $\dot{\bar{p}} \equiv 0$ and $\dot{s} \equiv 0$, which implies the invariance of \bar{p} and s .

Considering that $p_i - \bar{p} = -\sum_{j \in \mathcal{V}, j \neq i} (p_j - \bar{p})$, we can obtain that $\|p_i - \bar{p}\|^2 = \|\sum_{j \in \mathcal{V}, j \neq i} (p_j - \bar{p})\|^2 \leq (n-1) \sum_{j \in \mathcal{V}, j \neq i} \|p_j - \bar{p}\|^2$ which further implies that $\|p_i - \bar{p}\| \leq s\sqrt{n-1} \forall i \in \mathcal{V}$, and that $\|p_i - p_j\| \leq \|p_i - \bar{p}\| + \|p_j - \bar{p}\| \leq 2s\sqrt{n-1} \forall i, j \in \mathcal{V} \forall t \geq t_0$. ■

Let $\delta_i = p_i - p_i^*$, $\delta = \text{col}(\delta_1, \dots, \delta_n)$, $r = p - (\mathbf{1}_n \otimes \bar{p})$, $r^* = p^* - (\mathbf{1}_n \otimes \bar{p}^*)$, and $s^* = \sqrt{(1/n) \sum_{i=1}^n \|p_i^* - \bar{p}^*\|^2}$. With the control law (2), the dynamics of δ can be written in a compact form as

$$\dot{\delta} = \left(a + b \frac{\dot{\mu}}{\mu}\right) \bar{H}^T \text{diag}(P_{g_k}) g^*. \quad (6)$$

Lemma 3 implies that the target centroid and the target scale can be achieved by setting $\bar{p}(t_0) = \bar{p}^*$ and $s(t_0) = s^*$. To further analyze the equilibriums of the closed-loop system (6), we present the following lemma.

Lemma 4: Under Assumptions 1 and 2 and control law (2), setting $\bar{p}(t_0) = \bar{p}^*$ and $s(t_0) = s^*$, the trajectory of δ evolves

on the surface of the sphere $\mathcal{S} = \{\delta \in \mathbb{R}^n : \|\delta + r^*\| = \|r^*\|\}$, and the closed-loop system (6) has two equilibriums $\delta = 0$ and $\delta = -2r^*$. Moreover the equilibrium $\delta = -2r^*$ is unstable.

Proof: By Lemma 3, we can know that $\|r\| = \sqrt{ns}(t_0) = \sqrt{ns^*} = \|r^*\|$. Since $\bar{p} = \bar{p}^*$, we have $\delta = p - (\mathbf{1}_n \otimes \bar{p}) - (p^* - (\mathbf{1}_n \otimes \bar{p}^*))$, and consequently $\|\delta + r^*\| = \|r\| = \|r^*\|$. Hence, the trajectory of δ evolves on the surface \mathcal{S} .

Let $\hat{\delta}_i = (a + b[\dot{\mu}/\mu])f_i(\delta_i) = -(a + b[\dot{\mu}/\mu]) \sum_{j \in \mathcal{N}_i} P_{g_{ij}} g_{ij}^*$ and $f(\delta) = \text{col}(f_1(\delta_1), \dots, f_n(\delta_n)) = \bar{H}^T \text{diag}(P_{g_k}) g^*$. The equilibriums of (6) belong to \mathcal{S} and satisfy $f(\delta) = 0$. Then, it follows that:

$$\begin{aligned} (p^*)^T f(\delta) &= (p^* \bar{H})^T \text{diag}(P_{g_k}) g^* \\ &= \sum_{k=1}^m \|e_k\| (g_k^*)^T P_{g_k} g_k^* = 0. \end{aligned}$$

Due to the fact that $P_{g_k} \geq 0$, we have $g_k = \pm g_k^* \forall k = 1, \dots, m$. For the case $g_k = g_k^*$, by Assumption 1, the formation with bearing constraints $\{g_k^*\}_{k=1, \dots, m}$ is uniquely determined up to a translation and a scaling. Together with the centroid $\bar{p} = \bar{p}^*$ and the scale $s = s^*$, the formation is uniquely determined, that is, we have $p = p^*$ and $\delta = 0$. For the case $g_k = -g_k^*$, similarly, we know that the formation with bearing constraints $\{-g_k^*\}_{k=1, \dots, m}$, $\bar{p} = \bar{p}^*$ and $s = s^*$, is uniquely determined and has the same centroid, scale, and shape with (\mathcal{G}, p^*) . Furthermore, since $\|p - \mathbf{1}_n \otimes \bar{p}^*\| = \|p^* - \mathbf{1}_n \otimes \bar{p}^*\|$, we can conclude that $p = \mathbf{1}_n \otimes 2\bar{p}^* - p^*$ and $\delta = -2r^*$, which further implies that the formation with $\delta = -2r^*$ is geometrically a point reflection of (\mathcal{G}, p^*) .

For the reason that $(a + b[\dot{\mu}/\mu]) > 0$, the stability of the equilibrium $\delta = -2r^*$ is determined by the Jacobian matrix of $f(\delta)$. Let $F = [(\partial f(\delta))/(\partial \delta)]$ be the Jacobian matrix of $f(\delta)$ with the (i, j) th block of submatrix defined as

$$[F]_{ij} = \begin{cases} 0_{d \times d}, & i \neq j, (i, j) \notin \mathcal{E} \\ \frac{\partial f_i(\delta)}{\partial \delta_j}, & i \neq j, (i, j) \in \mathcal{E} \\ \frac{\partial f_i(\delta)}{\partial \delta_i}, & i = j, i \in \mathcal{V}. \end{cases}$$

Following the similar analysis in [13, Th. 9], it can be proved that $F|_{\delta=-2r^*} \geq 0$ and F has at least one positive eigenvalue. Hence, the equilibrium $\delta = -2r^*$ is unstable. ■

Remark 2: The geometric relationship between δ and the surface \mathcal{S} is shown in Fig. 2. The angle between δ and $-r^*$ is denoted as θ . Note that δ can always be decoupled as $\delta = \delta_{\parallel} + \delta_{\perp}$, where δ_{\parallel} is parallel to $-r^*$ and δ_{\perp} is perpendicular to $-r^*$.

Note that Lemmas 3 and 4 all based on the assumption that $g_{ij} \forall (i, j) \in \mathcal{E}$ is well defined. The following result will show that the interagent distances are lower bounded by γ if some initial conditions are satisfied.

Theorem 1: Under Assumption 1 and control law (2), for a constant $0 < \gamma < \min_{i, j \in \mathcal{V}, i \neq j} \|p_i^* - p_j^*\|$, the interagent distances are lower bounded by γ , that is, $\|p_i(t) - p_j(t)\| > \gamma \forall i, j \in \mathcal{V} \forall t > t_0$ if

$$\|\delta(t_0)\| < \frac{\min_{i, j \in \mathcal{V}, i \neq j} \|p_i^* - p_j^*\| - \gamma}{\sqrt{n}}. \quad (7)$$

Proof: Before analyzing the interagent distances, we first show that $\|\delta(t)\|$ is upper bounded by $\|\delta(t_0)\|$ for $t \geq t_0$.

Consider the following Lyapunov function candidate:

$$V = \frac{1}{2} \delta^T \delta.$$

By (6), the time derivative of V is obtained as

$$\begin{aligned} \dot{V} &= \left(a + b \frac{\dot{\mu}}{\mu}\right) (p - p^*)^T \bar{H}^T \text{diag}(P_{g_k}) g^* \\ &= -\left(a + b \frac{\dot{\mu}}{\mu}\right) e^{*T} \text{diag}(P_{g_k}) g^* \\ &= -\left(a + b \frac{\dot{\mu}}{\mu}\right) \sum_{k=1}^m \|e_k^*\| (g_k^*)^T P_{g_k} g_k^* \leq 0. \end{aligned} \quad (8)$$

It follows that $\|\delta(t)\| \leq \|\delta(t_0)\| \forall t \geq t_0$.

Since $p_i - p_j = (p_i - p_i^*) - (p_j - p_j^*) + (p_i^* - p_j^*)$, we obtain that

$$\begin{aligned} \|p_i - p_j\| &\geq \|p_i^* - p_j^*\| - \|p_i - p_i^*\| - \|p_j - p_j^*\| \\ &\geq \|p_i^* - p_j^*\| - \sum_{i=1}^n \|p_i - p_i^*\| \\ &\geq \|p_i^* - p_j^*\| - \sqrt{n} \|p - p^*\| \\ &\geq \|p_i^* - p_j^*\| - \sqrt{n} \delta(t) \end{aligned}$$

where we have used the fact that $n\|p - p^*\|^2 \geq \sum_{i=1}^n \|p_i - p_i^*\|^2$. Together with (7) and $\|\delta(t)\| \leq \|\delta(t_0)\|$, we can conclude that $\|p_i(t) - p_j(t)\| > \gamma \forall i, j \in \mathcal{V} \forall t > t_0$. ■

Remark 3: From (7), we can observe that the upper bound of $\|\delta(t_0)\|$ is proportional to $\min_{i,j \in \mathcal{V}} \|p_i^* - p_j^*\|$ and $(1/\sqrt{n})$. The intuitive explanation of condition (7) is that, for a large group of agents with a small target configuration, to avoid the collisions, the initial error $\delta(t_0)$ has to be small.

Now, we are in the position to give the first main result of this article.

Theorem 2: Under Assumption 1, Problem 1 is solved by control law (2) if condition (7) is satisfied, $\delta(t_0) \neq -2r^*$, $\bar{p}(t_0) = \bar{p}^*$, $s(t_0) = s^*$, and

$$\begin{aligned} &b h \lambda_{d+2}(\mathcal{B}^*) (\sin^2 \theta(t_0)) \min_{i,j \in \mathcal{V}, i \neq j} \|p_i^* - p_j^*\| \\ &> 8(n-1)(s^*)^2. \end{aligned} \quad (9)$$

Furthermore, $\|p_i(t) - p_j(t)\| > \gamma \forall i, j \in \mathcal{V}$, and the control input $u = \text{col}(u_1, \dots, u_n)$ remains C^1 smooth and uniformly bounded over the time interval $[t_0, \infty)$.

Proof: It follows from (8) that:

$$\begin{aligned} \dot{V} &= -\left(a + b \frac{\dot{\mu}}{\mu}\right) \sum_{k=1}^m \|e_k^*\| (g_k^*)^T P_{g_k} g_k^* \\ &= -\left(a + b \frac{\dot{\mu}}{\mu}\right) \sum_{k=1}^m \frac{\|e_k^*\|}{\|e_k\|^2} e_k^T P_{g_k} e_k \\ &\leq -\left(a + b \frac{\dot{\mu}}{\mu}\right) \frac{\min_{i,j \in \mathcal{V}, i \neq j} \|p_i^* - p_j^*\|}{\max_{i,j \in \mathcal{V}, i \neq j} \|p_i - p_j\|^2} p^T \bar{H}^T \text{diag}(P_{g_k^*}) \bar{H} p \end{aligned}$$

$$= -\left(a + b \frac{\dot{\mu}}{\mu}\right) \frac{\min_{i,j \in \mathcal{V}, i \neq j} \|p_i^* - p_j^*\|}{\max_{i,j \in \mathcal{V}, i \neq j} \|p_i - p_j\|^2} \delta^T \bar{H}^T \text{diag}(P_{g_k^*}) \bar{H} \delta \quad (10)$$

where we have used the facts that $(g_k)^T P_{g_k^*} g_k = (g_k^*)^T P_{g_k} g_k^*$ and $P_{g_k^*} e^* = 0$ to obtain the first and last equalities, respectively. Since $s(t) = s(t_0) = s^*$, in light of Lemma 3, we obtain that $\max_{i,j \in \mathcal{V}} \|p_i - p_j\|^2 \leq 4(n-1)(s^*)^2$, and furthermore that

$$\dot{V} \leq -\left(a + b \frac{\dot{\mu}}{\mu}\right) \frac{\min_{i,j \in \mathcal{V}, i \neq j} \|p_i^* - p_j^*\|}{4(n-1)(s^*)^2} \delta^T \mathcal{B}^* \delta$$

where $\mathcal{B}^* = \mathcal{B}(\mathcal{G}, p^*)$. By Lemma 1, we know that $\text{null}(\mathcal{B}^*) = \text{span}\{\mathbf{1} \otimes I_d, p^*\} = \text{span}\{\mathbf{1} \otimes I_d, r^*\}$. Since $\bar{p} = \bar{p}(t_0) = \bar{p}^*$ according to Lemma 3, we have $(\mathbf{1} \otimes I_d)^T \delta = 0$. Together with the facts that $\delta = \delta_{\parallel} + \delta_{\perp}$, $(\mathbf{1} \otimes I_d)^T \delta_{\parallel} = (\mathbf{1} \otimes I_d)^T r^* = 0$, and $\delta_{\perp}^T r^* = 0$, we can conclude that $\delta_{\perp} \perp \text{null}(\mathcal{B}^*)$, which further implies that $\delta^T \mathcal{B}^* \delta = \delta_{\perp}^T \mathcal{B}^* \delta_{\perp} \geq \lambda_{d+2}(\mathcal{B}^*) \delta_{\perp}^T \delta_{\perp}$. From Lemma 4, we have δ evolves on \mathcal{S} and $\delta_{\perp}^T \delta_{\perp} = \sin^2 \theta \delta^T \delta$ (see Fig. 2). It can be observed from Fig. 2 that $\theta \in [0, [\pi/2]]$. Since $\|\delta(t)\| \leq \|\delta(t_0)\| \forall t > t_0$, we know that $\theta(t) \geq \theta(t_0)$.

Based on the above analysis, we obtain that

$$\begin{aligned} \dot{V} &\leq - \underbrace{\frac{a \lambda_{d+2}(\mathcal{B}^*) (\sin^2 \theta(t_0)) \min_{i,j \in \mathcal{V}, i \neq j} \|p_i^* - p_j^*\|}{4(n-1)(s^*)^2}}_{\bar{\alpha}_1} \|\delta(t)\|^2 \\ &\quad - \underbrace{\frac{b \lambda_{d+2}(\mathcal{B}^*) (\sin^2 \theta(t_0)) \min_{i,j \in \mathcal{V}, i \neq j} \|p_i^* - p_j^*\|}{4(n-1)(s^*)^2}}_{\bar{\beta}_1} \frac{\dot{\mu}}{\mu} \|\delta(t)\|^2 \\ &= -2\bar{\alpha}_1 V - 2\bar{\beta}_1 \frac{\dot{\mu}}{\mu} V. \end{aligned}$$

Then, we can conclude from Lemma 2 that

$$\|\delta(t)\| \begin{cases} \leq \mu^{-\bar{\beta}_1} e^{-\bar{\alpha}_1(t-t_0)} \|\delta(t_0)\|, & t \in [t_0, t_0 + T) \\ \equiv 0, & t \in [t_0 + T, \infty) \end{cases} \quad (11)$$

which implies that p converges to p^* in user prespecified finite time T . In the following, we will show that u remains C^1 smooth and uniformly bounded.

By (2), we obtain that

$$\|u\| \leq \left(a + b \frac{\dot{\mu}}{\mu}\right) \|\bar{H}\| \|\text{diag}(P_{g_k}) g^*\|.$$

Since

$$\|\text{diag}(P_{g_k}) g^*\|^2 = g^T \text{diag}(P_{g_k}) g = \sum_{k=1}^m \frac{1}{\|e_k\|^2} e_k^T P_{g_k} e_k \quad (12)$$

and $\|e_k\| \geq \gamma$ (from Theorem 1), and we have

$$\|\text{diag}(P_{g_k}) g^*\| \leq \frac{1}{\gamma} \sqrt{\delta^T \bar{H}^T \text{diag}(P_{g_k^*}) \bar{H} \delta}. \quad (13)$$

Incorporating this with (11), we have

$$\|\text{diag}(P_{g_k}) g^*\| \leq \frac{1}{\gamma} \|\mathcal{B}^*\|^{\frac{1}{2}} \mu^{-\bar{\beta}_1} e^{-\bar{\alpha}_1(t-t_0)} \|\delta(t_0)\| \quad (14)$$

$\forall t \in [t_0, t_0 + T)$ and $\|\text{diag}(P_{g_k})g^*\| \equiv 0 \ \forall t \in [t_0 + T, \infty)$. Hence, we have

$$\left\| \frac{\dot{\mu}}{\mu} \text{diag}(P_{g_k})g^* \right\| \leq \frac{1}{\gamma} \|B^*\|^{\frac{1}{2}} \frac{h}{T} \mu^{-(\bar{\beta}_1 - \frac{1}{h})} e^{-\bar{\alpha}_1(t-t_0)} \|\delta(t_0)\| \quad (15)$$

$\forall t \in [t_0, t_0 + T)$, and further by $\bar{\beta}_1 - (1/h) > 0$ [from (9)], we have that $\lim_{t \rightarrow (t_0+T)^-} \|(\dot{\mu}/\mu) \text{diag}(P_{g_k})g^*\| = 0$ and that $\|(\dot{\mu}/\mu) \text{diag}(P_{g_k})g^*\| \equiv 0 \ \forall t \in [t_0 + T, \infty)$. Noting that $u = (a + b[\dot{\mu}/\mu])\bar{H}^T \text{diag}(P_{g_k})g^*$ and g_k is continuous respect to t , we can conclude that u is continuous and uniformly bounded on $[t_0, \infty)$.

Next, we will show (du/dt) is continuous on $[t_0, \infty)$. Since

$$\begin{aligned} \frac{du}{dt} = & - \left(a + b \frac{\dot{\mu}}{\mu} \right) \bar{H}^T \frac{d(\text{diag}(P_{g_k}))}{dt} g^* \\ & - \frac{bh}{T^2} \mu^{\frac{2}{h}} \bar{H}^T \text{diag}(P_{g_k})g^* \end{aligned} \quad (16)$$

it is clear that (du/dt) is continuous on $[t_0, t_0 + T)$ and $(t_0 + T, \infty)$. From (14), it can be obtained that

$$\left\| \mu^{\frac{2}{h}} \bar{H}^T \text{diag}(P_{g_k})g^* \right\| \leq \frac{1}{\gamma} \|B^*\|^{\frac{1}{2}} \mu^{-(\bar{\beta}_1 - \frac{2}{h})} e^{-\bar{\alpha}_1(t-t_0)} \|\delta(t_0)\| \quad (17)$$

$\forall t \in [t_0, t_0 + T)$, and further by $\bar{\beta}_1 - (2/h) > 0$ [from (9)], we have that $\lim_{t \rightarrow (t_0+T)^-} \|\mu^{(2/h)} \text{diag}(P_{g_k})g^*\| = 0$ and that $\|\mu^{(2/h)} \text{diag}(P_{g_k})g^*\| \equiv 0 \ \forall t \in [t_0 + T, \infty)$.

For the first term in (16), we have

$$\begin{aligned} & \left(a + b \frac{\dot{\mu}}{\mu} \right) \bar{H}^T \frac{d(\text{diag}(P_{g_k}))}{dt} g^* \\ & = \left(a + b \frac{\dot{\mu}}{\mu} \right) \frac{\partial f(\delta)}{\partial \delta} \dot{\delta} = \left(a + \frac{bh}{T} \mu^{\frac{1}{h}} \right)^2 F \bar{H}^T \text{diag}(P_{g_k})g^* \end{aligned}$$

where F is the Jacobian matrix defined in the proof of Lemma 4. From Theorem 1, we have $\|e_{ij}\| \geq \gamma$, and further due to the definition of G_{ij} and $P_{g_{ij}}$, we know that $\|[F]_{ij}|\delta\| \ \forall i, j \in \mathcal{V}$ is bounded for $\delta \in [0, \delta(t_0)]$. Thus, we can always define a positive constant κ such that $\kappa = \max_{\delta \in [0, \delta(t_0)]} \|F|\delta\|$.

It then follows that:

$$\begin{aligned} & \left\| \left(a + \frac{bh}{T} \mu^{\frac{1}{h}} \right)^2 F \bar{H}^T \text{diag}(P_{g_k})g^* \right\| \\ & \leq \kappa \|\bar{H}\| \left(a^2 + \frac{2abh}{T} \mu^{\frac{1}{h}} + \frac{b^2 h^2}{T^2} \mu^{\frac{2}{h}} \right) \|\text{diag}(P_{g_k})g^*\|. \end{aligned}$$

By using the fact $\bar{\beta}_1 - (2/h) > 0$, following similar analysis for (15) and (17), it is clear that $\lim_{t \rightarrow (t_0+T)^-} \|(du/dt)\| = 0 \ \forall t \in [t_0, t_0 + T)$ and $\|(du/dt)\| \equiv 0 \ \forall t \in [t_0 + T, \infty)$. Hence, we can conclude that (du/dt) is continuous on $[t_0, \infty)$, and furthermore the control input u is C^1 smooth and uniformly bounded over the time interval $[t_0, \infty)$. ■

Remark 4: Noting that $\delta = -2r^*$ is an unstable equilibrium of closed-loop system (6), Theorem 2 guarantees almost global formation stabilization excepting the case $\delta(t_0) = -2r^*$. Furthermore, the invariance of scale and centroid is used such that the formation control problem can be transferred into a bearing stabilization problem.

Remark 5: It is worth noting that the time-varying gain $(\dot{\mu}/\mu)$ plays an important role in achieving the formation control in finite time. From (15), we can see that the control gain $(\dot{\mu}/\mu)$ goes to infinite when $t \rightarrow t_0 + T$. However, the control input u remains bounded and C^1 smooth. Equations (12) and (13) build the connection between $\|u\|$ and $\|\delta\|$. Intuitively, condition (9) guarantees a sufficiently large b such that the decrease of $\|\delta\|$ is faster than the increase of $(\dot{\mu}/\mu)$. Different from the fractional power-based finite-time control law in [8], [21], and [32], the converge time of control law (2) does not depend on the initial condition and can be any value specified by users.

IV. BEARING-ONLY LEADER-FOLLOWER FORMATION CONTROL

To guarantee the convergence, Theorem 2 requires $\bar{p}(t_0) = \bar{p}^*$ and $s(t_0) = s^*$, which may not be easily satisfied when the system has a large number of agents. In this section, we will show that these requirements can be relaxed and the global stabilization can be achieved by using a leader-follower control structure.

Without loss of generality, suppose the first $n_l \geq 2$ agents are leaders and the remaining $n_f = n - n_l$ agents are followers. Let $\mathcal{V}_l = \{1, \dots, n_l\}$ and $\mathcal{V}_f = \{n_l + 1, \dots, n\}$ be the set of leaders and followers, respectively. The positions of agents are denoted as $p = \text{col}(p_l, p_f)$, where $p_l = \text{col}(p_1, \dots, p_{n_l})$ and $p_f = \text{col}(p_{n_l+1}, \dots, p_n)$ are the positions of leaders and followers, respectively. The leader-follower formation control problem is given as follows.

Problem 2: With leader positions $\{p_i^*\}_{i \in \mathcal{V}_l}$, design control input for agent $i \in \mathcal{V}_f$ based on the bearing vectors $\{g_{ij}(t)\}_{j \in \mathcal{N}_i}$ such that $p \rightarrow p^*$ for $t \rightarrow t_0 + T$, and $p = p^*$ for $t \geq t_0 + T$, where p^* is a target configuration and $T \in \mathbb{R}_{>0}$ is the convergence time prespecified by users.

Since the leaders are stationary, we have $\dot{p}_i = 0, i \in \mathcal{V}_l$. The control law of each following mobile agent is designed as:

$$u_i = - \left(a + b \frac{\dot{\mu}}{\mu} \right) \sum_{j \in \mathcal{N}_i} P_{g_{ij}} g_{ij}^*, \quad i \in \mathcal{V}_f. \quad (18)$$

Note that control law (18) is same as (2). In the following, we will show that control law (18) can achieve global formation stabilization.

Since $\delta_i = p_i - p_i^*$, we have $\delta = \text{col}(\delta_l, \delta_f) = \text{col}(\mathbf{0}_{n_l}, \delta_f)$, where $\delta_l = p_l - p_l^*$ and $\delta_f = p_f - p_f^*$. By (18), the dynamics of δ can be written in a compact form as

$$\dot{\delta} = \left(a + b \frac{\dot{\mu}}{\mu} \right) \begin{bmatrix} \mathbf{0}_{n_l \times n_{dl}} & \mathbf{0}_{n_l \times n_{df}} \\ \mathbf{0}_{n_f \times n_{dl}} & I_{n_f} \end{bmatrix} \bar{H}^T \text{diag}(P_{g_k})g^*.$$

Theorem 3: Under Assumption 1 and control law (18), the interagent distances are also lower bounded by γ , if condition (7) is satisfied.

Proof: Here, we just need to prove that $\|\delta\|$ is upper bounded by $\|\delta(t_0)\|$, for $t > t_0$ and the remaining proof follows similarly as in Theorem 1. Noting that $V = (1/2)\delta^T \delta$, the time derivative of V is given as

$$\dot{V} = \left(a + b \frac{\dot{\mu}}{\mu} \right) \delta^T \begin{bmatrix} \mathbf{0}_{n_l \times n_{dl}} & \mathbf{0}_{n_l \times n_{df}} \\ \mathbf{0}_{n_f \times n_{dl}} & I_{n_f} \end{bmatrix} \bar{H}^T \text{diag}(P_{g_k})g^*$$

$$\begin{aligned}
&= -\left(a + b \frac{\dot{\mu}}{\mu}\right) \delta^T \bar{H}^T \text{diag}(P_{g_k}) g^* \\
&= -\left(a + b \frac{\dot{\mu}}{\mu}\right) \sum_{k=1}^m \|e_k^*\| (g_k^*)^T P_{g_k} g_k^* \leq 0
\end{aligned} \quad (19)$$

where we have used the fact that $\delta_l = \mathbf{0}$. It follows that $\|\delta(t)\| \leq \|\delta(t_0)\| \forall t \geq t_0$. ■

Remark 6: Theorem 3 implies that although the state trajectory in the leader–follower case is different with leaderless case, the conditions for collision avoidance are same. Fixing arbitrary number of agents on the target position will not change \dot{V} , hence the condition for the collision avoidance is not related to the number of leaders.

Theorem 4: Under Assumption 1, Problem 2 is solved by control law (18) if condition (7) is satisfied and

$$\begin{aligned}
&b h \lambda_{\min}(\mathcal{B}_{ff}^*) \min_{i,j \in \mathcal{V}, i \neq j} \|p_i^* - p_j^*\| \\
&> 2 \|\bar{H}\|^2 (\|\delta(t_0)\| + \sqrt{n} s^*)^2
\end{aligned} \quad (20)$$

where $\lambda_{\min}(\mathcal{B}_{ff}^*)$ is the smallest eigenvalue of \mathcal{B}_{ff}^* . Furthermore, $\|p_i(t) - p_j(t)\| > \gamma \forall i, j \in \mathcal{V}$, and the control input $u_f = \text{col}(u_{n_l+1}, \dots, u_n)$ remains C^1 smooth and uniformly bounded over the time interval $[t_0, \infty)$.

Proof: By (18) and following the analysis for (10), we have:

$$\begin{aligned}
\dot{V} &\leq -\left(a + b \frac{\dot{\mu}}{\mu}\right) \frac{\min_{i,j \in \mathcal{V}, i \neq j} \|p_i^* - p_j^*\|}{\max_{i,j \in \mathcal{V}, i \neq j} \|p_i - p_j\|^2} \delta^T \bar{H}^T \text{diag}(P_{g_k}^*) \bar{H} \delta \\
&= -\left(a + b \frac{\dot{\mu}}{\mu}\right) \frac{\min_{i,j \in \mathcal{V}, i \neq j} \|p_i^* - p_j^*\|}{\max_{i,j \in \mathcal{V}, i \neq j} \|p_i - p_j\|^2} \delta^T \mathcal{B}^* \delta.
\end{aligned} \quad (21)$$

Due to the fact that

$$\delta^T \mathcal{B}^* \delta = \begin{bmatrix} \mathbf{0}^T & \delta_f^T \end{bmatrix} \begin{bmatrix} \mathcal{B}_{ll}^* & \mathcal{B}_{lf}^* \\ (\mathcal{B}_{lf}^*)^T & \mathcal{B}_{ff}^* \end{bmatrix} \begin{bmatrix} \mathbf{0}^T & \delta_f^T \end{bmatrix}^T$$

and in light of Lemma 1 3), we know that $\mathcal{B}_{ff}^* > 0$ and further that $\delta^T \mathcal{B}^* \delta \geq \lambda_{\min}(\mathcal{B}_{ff}^*) \delta^T \delta$.

Different from the leaderless case, for the leader–follower case, the invariance of the centroid \bar{p} and the scale s is no longer hold. Alternatively, the following inequalities are used to characterize the upper bound of $\max_{i,j \in \mathcal{V}, i \neq j} \|p_i - p_j\|^2$:

$$\begin{aligned}
\max_{i,j \in \mathcal{V}, i \neq j} \|p_i - p_j\|^2 &\leq \|e\|^2 = \|\bar{H}(p - p^* + p^*)\|^2 \\
&= \|\bar{H}(\delta + r^*)\|^2 \\
&\leq \|\bar{H}\|^2 (\|\delta(t_0)\| + \sqrt{n} s^*)^2
\end{aligned} \quad (22)$$

where we have used the facts $\|\delta(t)\| \leq \|\delta(t_0)\|$, $\bar{H}(p^* - (\mathbf{1}_n \otimes \bar{p}^*)) = \bar{H}p^*$, and $\|r^*\| = \sqrt{n} s^*$ to obtain the last inequality. It then follows from (21) and (22) that:

$$\begin{aligned}
\dot{V} &\leq - \underbrace{\frac{a \lambda_{\min}(\mathcal{B}_{ff}^*) \min_{i,j \in \mathcal{V}, i \neq j} \|p_i^* - p_j^*\|}{\|\bar{H}\|^2 (\|\delta(t_0)\| + \sqrt{n} s^*)^2}}_{\bar{\alpha}_2} \|\delta(t)\|^2 \\
&\quad - \underbrace{\frac{b \lambda_{\min}(\mathcal{B}_{ff}^*) \min_{i,j \in \mathcal{V}, i \neq j} \|p_i^* - p_j^*\|}{\|\bar{H}\|^2 (\|\delta(t_0)\| + \sqrt{n} s^*)^2}}_{\bar{\beta}_2} \frac{\dot{\mu}}{\mu} \|\delta(t)\|^2 \\
&= -2\bar{\alpha}_2 V - 2\bar{\beta}_2 \frac{\dot{\mu}}{\mu} V.
\end{aligned}$$

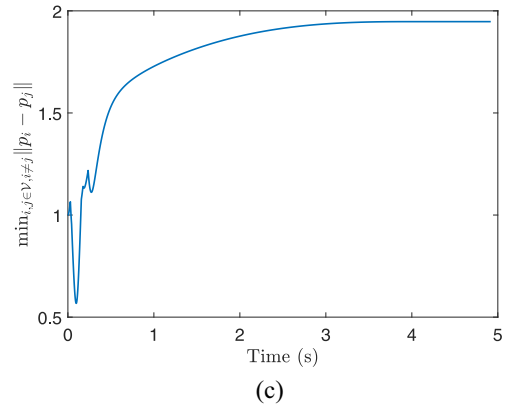
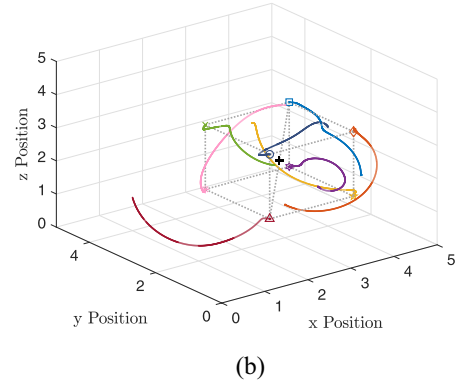
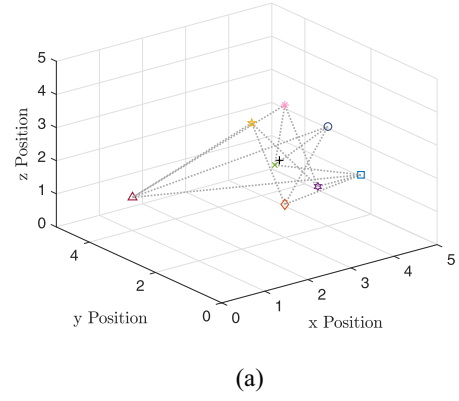


Fig. 3. Simulation results of control law (2). (a) Initial positions $p_i(0) \forall i = 1, \dots, 8$. (b) Trajectories and positions p_i at 4 s $\forall i = 1, \dots, 8$. (c) Minimum distance between agents $\min_{i,j \in \mathcal{V}, i \neq j} \|p_i - p_j\|$.

In light of Lemma 2, we have

$$\|\delta(t)\| \begin{cases} \leq \mu^{-\bar{\beta}_2} e^{-\bar{\alpha}_2(t-t_0)} \|\delta(t_0)\|, & t \in [t_0, t_0 + T) \\ \equiv 0, & t \in [t_0 + T, \infty) \end{cases}$$

which implies that p converges to p^* in a user prespecified finite time T . Note that

$$\begin{aligned}
\|u_f\| &= \left\| \begin{bmatrix} \mathbf{0}_{d_{n_l} \times d_{n_l}} & \mathbf{0}_{d_{n_l} \times d_{n_f}} \\ \mathbf{0}_{d_{n_f} \times d_{n_l}} & I_{d_{n_f}} \end{bmatrix} \bar{H}^T \text{diag}(P_{g_k}) g^* \right\| \\
&\leq \|\bar{H}\| \|\text{diag}(P_{g_k}) g^*\|.
\end{aligned}$$

Following the similar analysis in Theorem 2, it can be proved that u_f is C^1 smooth and uniformly bounded over the time interval $[t_0, \infty)$. ■

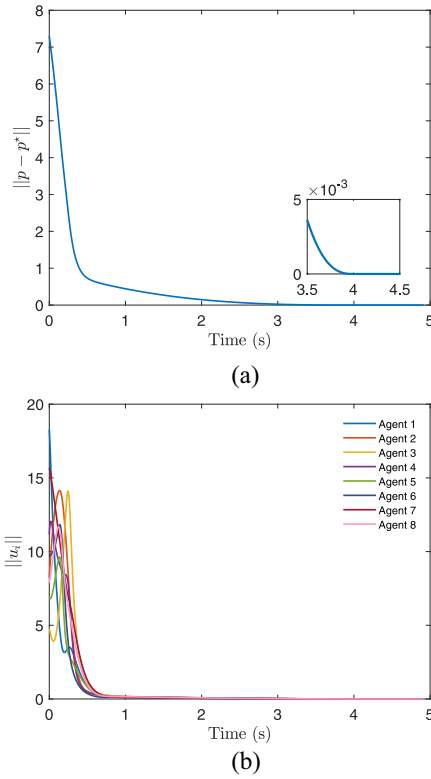


Fig. 4. Simulation results of control law (2). (a) Position error $\|p - p^*\|$. (b) Norm of control input $\|u_i\| \forall i = 1, \dots, 8$.

Remark 7: Any positive a and b can guarantee $\dot{V} \leq 0$. Conditions (9) and (20) are required to guarantee the boundedness and the smoothness of the control input.

Remark 8: In the leader-follower case, the initial requirements $\bar{p}(t_0) = \bar{p}^*$ and $s(t_0) = s^*$ are removed. Intuitively, due to $n_l \geq 2$, at least two points and the edge between these two points are fixed. Since these two points and the edge can determine the translation and scaling of the target formation, together with the fact that the target formation is infinitesimally bearing rigid, the target formation is uniquely determined. Furthermore, since we have $\delta_l = \mathbf{0}$, the system will not start from the initial condition $\delta(t_0) = -2r^*$. Hence, the global stability can be achieved.

V. SIMULATION EXAMPLE

To validate the effectiveness of control law (2), we show an example of eight agents with a cubic target formation. The initial positions are chosen to satisfy the conditions in Theorem 2 and the parameters are set as follows: $a = 0.2$, $b = 5$, $h = 5$, and $T = 4$ s.

The initial positions and the positions at 4 s are shown in Fig. 3(a) and (b). The vertices in different color are the agents and the solid lines are trajectories of the agents. The dashed lines in gray and the plus sign in black represent the relative bearing and the centroid \bar{p} , respectively. We can observe that the centroid is invariant. Fig. 3(c) shows that the minimum distance between agents is larger than 0.5. Hence, there is no collision between agents. Together with Fig. 4(a), we can see that the target formation is achieved at 4 s. Furthermore, Fig. 4(b) shows that the control inputs $u_i \forall i = 1, \dots, 8$ are bounded and smooth.

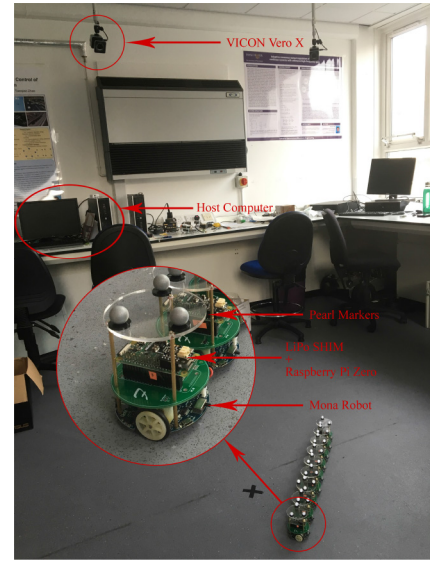


Fig. 5. Experimental platform.

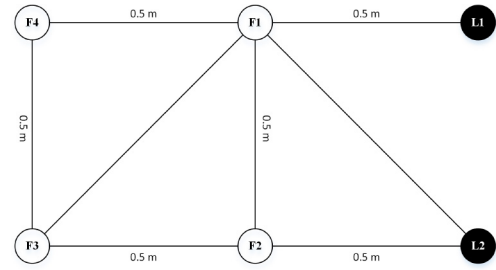


Fig. 6. Target formation with two leaders L1 and L2.

VI. EXPERIMENTAL VALIDATION

To demonstrate the performance of control law (18), we design an experimental platform with self-fabricated mobile robots shown in Fig. 5. In this platform, a VICON motion capture system with 6 Vero X cameras is used to obtain the position of mobile robots. A Linux-based host computer (CPU 2.7-GHz, 4-GB RAM) is used to transfer the position data into the relative bearings, package the relative bearings into robot operating system (ROS) topics, and broadcast the topics through Wi-Fi. To simulate a distributed sensor network, each robot only subscribes the topics of neighboring robots. The mobile robot is mainly composed of three levels.

- 1) Mona robot [33] based on Arduino Pro Mini (designed by the University of Manchester).
- 2) LiPo SHIM + Raspberry Pi Zero [running control law (18) and subscribing ROS topics at 80 Hz].
- 3) 14-mm pearl markers (forming unique patterns for motion capture).

In this experiment, the target formation of six robots is given in Fig. 6 and the parameters are set as $a = 0.12$, $b = 0.3$, $h = 2$, and $T = 35$ s. It is worth noting that the noise introduced by the motion capture system is inevitable and may result in an unbounded $(\dot{\mu}/\mu) \sum_{j \in \mathcal{N}_i} P_{gij} g_{ij}^*$ [with $(\dot{\mu}/\mu)$ growing unbounded, while $\sum_{j \in \mathcal{N}_i} P_{gij} g_{ij}^*$ not decaying to zero]. To address this issue, inspired by [27], we set T in μ on $[t_0, t_0 + T)$.

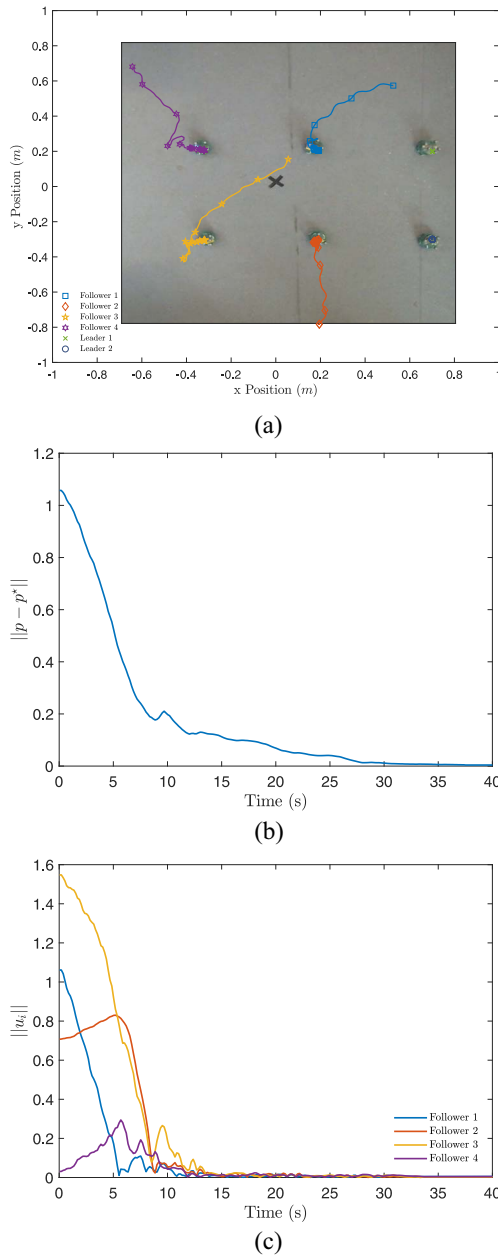


Fig. 7. Experimental results of control law (18). (a) Trajectories and final configuration. (b) Position error $\|p - p^*\|$. (c) Norm of control input $\|u_i\| \forall i \in \mathcal{V}_f$.

to a value \bar{T} slightly larger than the user prespecified settling time, that is

$$\mu(t) = \frac{\bar{T}^h}{(t_0 + \bar{T} - t)^h}, \quad t \in [t_0, t_0 + T] \quad (23)$$

where $\bar{T} = 35.4 \text{ s} > T$. The dynamics of robots are described as unicycle [8]. To implement control law (18), we linearize the dynamics of robots by following the steps in [8]. (Please refer to [8] for details.)

The experimental results are shown in Fig. 7. Fig. 7(a) is plotted against an image taken by the downward-looking camera on the ceiling. Together with Fig. 7(b), we can see that the target configuration is achieved at 35 s and there is no collision between robots. Furthermore, the control inputs u_i of followers are bounded as shown in Fig. 7(c).

VII. CONCLUSION

This article proposes new bearing-only control laws to achieve target formations in finite time. The almost global convergence is guaranteed. Furthermore, the convergence time is not related to initial conditions and can be arbitrarily chosen by users. Sufficient conditions for collision avoidance are also given. Then, the almost global convergence is extended to global convergence by using a leader–follower control structure. Since no signum function or fractional power feedback is used, the control action of the proposed control laws is C^1 smooth. The simulation and experimental results both demonstrate the effectiveness of our design.

REFERENCES

- [1] B. D. O. Anderson, C. Yu, B. Fidan, and J. M. Hendrickx, “Rigid graph control architectures for autonomous formations,” *IEEE Control Syst. Mag.*, vol. 28, no. 6, pp. 48–63, Dec. 2008.
- [2] X. Dong, B. Yu, Z. Shi, and Y. Zhong, “Time-varying formation control for unmanned aerial vehicles: Theories and applications,” *IEEE Trans. Control Syst. Technol.*, vol. 23, no. 1, pp. 340–348, Jan. 2015.
- [3] Z. Sun, S. Mou, M. Deghat, and B. D. O. Anderson, “Finite time distributed distance-constrained shape stabilization and flocking control for d -dimensional undirected rigid formations,” *Int. J. Robust Nonlinear Control*, vol. 26, no. 13, pp. 2824–2844, 2016.
- [4] Z. Lin, L. Wang, Z. Han, and M. Fu, “A graph Laplacian approach to coordinate-free formation stabilization for directed networks,” *IEEE Trans. Autom. Control*, vol. 61, no. 5, pp. 1269–1280, May 2016.
- [5] Z. Sun, S. Mou, B. D. Anderson, and M. Cao, “Exponential stability for formation control systems with generalized controllers: A unified approach,” *Syst. Control Lett.*, vol. 93, pp. 50–57, Jul. 2016.
- [6] F. Xiao, L. Wang, J. Chen, and Y. Gao, “Finite-time formation control for multi-agent systems,” *Automatica*, vol. 45, no. 11, pp. 2605–2611, Nov. 2009.
- [7] X. Dong, Y. Zhou, Z. Ren, and Y. Zhong, “Time-varying formation tracking for second-order multi-agent systems subjected to switching topologies with application to quadrotor formation flying,” *IEEE Trans. Ind. Electron.*, vol. 64, no. 6, pp. 5014–5024, Jun. 2017.
- [8] C. Wang, H. Tnunay, Z. Zuo, B. Lennox, and Z. Ding, “Fixed-time formation control of multirobot systems: Design and experiments,” *IEEE Trans. Ind. Electron.*, vol. 66, no. 8, pp. 6292–6301, Aug. 2019, doi: [10.1109/TIE.2018.2870409](https://doi.org/10.1109/TIE.2018.2870409).
- [9] M. Ou, H. Du, and S. Li, “Finite-time formation control of multiple nonholonomic mobile robots,” *Int. J. Robust Nonlinear Control*, vol. 24, no. 1, pp. 140–165, 2014.
- [10] E. Montijano, E. Cristofalo, D. Zhou, M. Schwager, and C. Sagués, “Vision-based distributed formation control without an external positioning system,” *IEEE Trans. Robot.*, vol. 32, no. 2, pp. 339–351, Apr. 2016.
- [11] A. N. Bishop, I. Shames, and B. D. O. Anderson, “Stabilization of rigid formations with direction-only constraints,” in *Proc. 50th IEEE Conf. Decis. Control Eur. Control Conf.*, Orlando, FL, USA, Dec. 2011, pp. 746–752.
- [12] A. Franchi and P. R. Giordano, “Decentralized control of parallel rigid formations with direction constraints and bearing measurements,” in *Proc. 51th IEEE Conf. Decis. Control*, Maui, HI, USA, Dec. 2012, pp. 5310–5317.
- [13] S. Zhao and D. Zelazo, “Bearing rigidity and almost global bearing-only formation stabilization,” *IEEE Trans. Autom. Control*, vol. 61, no. 5, pp. 1255–1268, May 2016.
- [14] S. Zhao and D. Zelazo, “Localizability and distributed protocols for bearing-based network localization in arbitrary dimensions,” *Automatica*, vol. 69, pp. 334–341, Jul. 2016.
- [15] S. Zhao, Z. Li, and Z. Ding, “Bearing-only formation tracking control of multiagent systems,” *IEEE Trans. Autom. Control*, vol. 64, no. 11, pp. 4541–4554, Nov. 2019, doi: [10.1109/tac.2019.2903290](https://doi.org/10.1109/tac.2019.2903290).
- [16] S. Zhao, F. Lin, K. Peng, B. M. Chen, and T. H. Lee, “Finite-time stabilisation of cyclic formations using bearing-only measurements,” *Int. J. Control*, vol. 87, no. 4, pp. 715–727, 2014.
- [17] M. H. Trinh, D. Mukherjee, D. Zelazo, and H. Ahn, “Finite-time bearing-only formation control,” in *Proc. 56th IEEE Conf. Decis. Control*, Melbourne, VIC, Australia, Dec. 2017, pp. 1578–1583.

- [18] Q. Van Tran, M. H. Trinh, D. Zelazo, D. Mukherjee, and H. Ahn, "Finite-time bearing-only formation control via distributed global orientation estimation," *IEEE Trans. Control Netw. Syst.*, vol. 6, no. 2, pp. 702–712, Jun. 2019.
- [19] R. Tron, J. Thomas, G. Loianno, K. Daniilidis, and V. Kumar, "A distributed optimization framework for localization and formation control: Applications to vision-based measurements," *IEEE Control Syst. Mag.*, vol. 36, no. 4, pp. 22–44, Aug. 2016.
- [20] G. Mao, B. Fidan, and B. D. O. Anderson, "Wireless sensor network localization techniques," *Comput. Netw.*, vol. 51, no. 10, pp. 2529–2553, 2007.
- [21] Z. Zuo, "Nonsingular fixed-time consensus tracking for second-order multi-agent networks," *Automatica*, vol. 54, pp. 305–309, Apr. 2015.
- [22] J. Cortés, "Finite-time convergent gradient flows with applications to network consensus," *Automatica*, vol. 42, no. 11, pp. 1993–2000, 2006.
- [23] X. Wang, S. Li, and P. Shi, "Distributed finite-time containment control for double-integrator multiagent systems," *IEEE Trans. Cybern.*, vol. 44, no. 9, pp. 1518–1528, Sep. 2014.
- [24] Y. Zhao, Y. Liu, G. Wen, W. Ren, and G. Chen, "Edge-based finite-time protocol analysis with final consensus value and settling time estimations," *IEEE Trans. Cybern.*, vol. 50, no. 4, pp. 1450–1459, Apr. 2020, doi: [10.1109/TCYB.2018.2872806](https://doi.org/10.1109/TCYB.2018.2872806).
- [25] H. Wang, W. Yu, W. Ren, and J. Lü, "Distributed adaptive finite-time consensus for second-order multiagent systems with mismatched disturbances under directed networks," *IEEE Trans. Cybern.*, early access, doi: [10.1109/TCYB.2019.2903218](https://doi.org/10.1109/TCYB.2019.2903218).
- [26] B. Tian, H. Lu, Z. Zuo, and W. Yang, "Fixed-time leader-follower output feedback consensus for second-order multiagent systems," *IEEE Trans. Cybern.*, vol. 49, no. 4, pp. 1545–1550, Apr. 2019.
- [27] Y. Song, Y. Wang, J. Holloway, and M. Krstic, "Time-varying feedback for regulation of normal-form nonlinear systems in prescribed finite time," *Automatica*, vol. 83, pp. 243–251, Sep. 2017.
- [28] C. Godsil and G. Royle, *Algebraic Graph Theory*. New York, NY, USA: Springer, 2001.
- [29] M. H. Trinh, S. Zhao, Z. Sun, D. Zelazo, B. D. O. Anderson, and H. Ahn, "Bearing-based formation control of a group of agents with leader-first follower structure," *IEEE Trans. Autom. Control*, vol. 64, no. 2, pp. 598–613, Feb. 2019.
- [30] K.-K. Oh, M.-C. Park, and H.-S. Ahn, "A survey of multi-agent formation control," *Automatica*, vol. 53, pp. 424–440, Mar. 2015.
- [31] B. Zhu, L. Xie, D. Han, X. Meng, and R. Teo, "A survey on recent progress in control of swarm systems," *Sci. China Inf. Sci.*, vol. 60, no. 7, 2017, Art. no. 070201.
- [32] S. Li, H. Du, and X. Lin, "Finite-time consensus algorithm for multi-agent systems with double-integrator dynamics," *Automatica*, vol. 47, no. 8, pp. 1706–1712, 2011.
- [33] F. Arvin, J. Espinosa, B. Bird, A. West, S. Watson, and B. Lennox, "Mona: An affordable open-source mobile robot for education and research," *J. Intell. Robot. Syst.*, vol. 94, no. 3, pp. 761–775, 2019.



Zhenzhong Li received the B.Eng. degree in electrical engineering from the Huazhong University of Science and Technology, Hubei, China, in 2013, and the M.Sc. and Ph.D. degrees in control systems from the University of Manchester, Manchester, U.K., in 2014 and 2019, respectively.

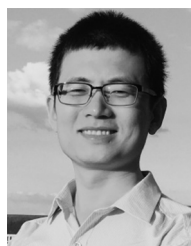
He is currently a Research Fellow with the School of Electronics and Electrical, University of Leeds, Leeds, U.K. From 2018 to 2019, he was a Research Associate with the Department of Electrical and Electronic Engineering, University of Manchester.

His research interests include distributed optimization, cooperative control of multiagent system, and human-robot interaction.



Hilton Tnunay received the B.Eng. degree in electrical engineering from Universitas Gadjah Mada, Yogyakarta, Indonesia, in 2015. He is currently pursuing the Ph.D. degree in control systems with the School of Electrical and Electronic Engineering, University of Manchester, Manchester, U.K.

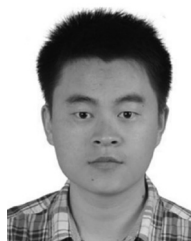
His research areas include distributed coordination and estimation of robotic sensor networks.



Shiyu Zhao received the B.Eng. and M.Eng. degrees from the Beijing University of Aeronautics and Astronautics, Beijing, China, in 2006 and 2009, respectively, and the Ph.D. degree in electrical engineering from the National University of Singapore, Singapore, in 2014.

He was a Postdoctoral Researcher with the Technion—Israel Institute of Technology, Haifa, Israel, and the University of California at Riverside, Riverside, CA, USA, from 2014 to 2016. He was a Lecturer with the Department of Automatic Control and Systems Engineering, University of Sheffield, Sheffield, U.K., from 2016 to 2018. He is currently an Assistant Professor with the School of Engineering, Westlake University, Hangzhou, China. His research interests lie in theories and applications of aerial robotic systems.

Dr. Zhao was a co-recipient of the Best Paper Award (Guan Zhao-Zhi Award) in the 33rd Chinese Control Conference, Nanjing, China, in 2014.



Wei Meng received the Ph.D. degree in information and mechatronics engineering jointly trained by the Wuhan University of Technology, Wuhan, China, and the University of Auckland, Auckland, New Zealand, in 2016.

He is currently with the School of Information Engineering, Wuhan University of Technology and a Research Fellow with the School of Electronic and Electrical Engineering, University of Leeds, Leeds, U.K. He has authored/coauthored three books and over 50 peer-reviewed papers in rehabilitation robotics and control.

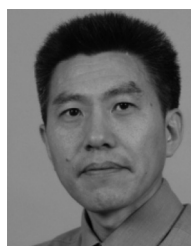


Sheng Q. Xie (Senior Member, IEEE) received the Ph.D. degree in mechanical engineering from the University of Canterbury, Christchurch, New Zealand, in 2002.

In 2003, he joined the University of Auckland, Auckland, New Zealand, where he became a Chair Professor of (bio)mechatronics in 2011. Since 2017, he has been the Chair of robotics and autonomous systems with the University of Leeds, Leeds, U.K. He has authored or coauthored 8 books, 15 book chapters, and over 400 international journal and conference papers.

His current research interests are medical and rehabilitation robots and advanced robot control.

Prof. Xie is an Elected Fellow of the Institution of Professional Engineers New Zealand.



Zhengtao Ding (Senior Member, IEEE) received the B.Eng. degree from Tsinghua University, Beijing, China, and the M.Sc. degree in systems and control and the Ph.D. degree in control systems from the University of Manchester Institute of Science and Technology, Manchester, U.K.

He was a Lecturer with Ngee Ann Polytechnic, Singapore, for ten years. In 2003, he joined the University of Manchester, Manchester, where he is currently a Professor of control systems with the Department of Electrical and Electronic Engineering.

He has authored the book *Nonlinear and Adaptive Control Systems* (IET, 2013) and has published over 200 research articles. His research interests include nonlinear and adaptive control theory and their applications, and more recently network-based control, distributed optimization, and distributed learning, with applications to power systems and robotics.

Prof. Ding has served as an Associate Editor for the IEEE TRANSACTIONS ON AUTOMATIC CONTROL, IEEE CONTROL SYSTEMS LETTERS, and several other journals. He is a member of the IEEE Technical Committee on Nonlinear Systems and Control, the IEEE Technical Committee on Intelligent Control, and the IFAC Technical Committee on Adaptive and Learning Systems.