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Controlling a triangular flexible formation of autonomous agents. [★]

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Abstract: In formation control, triangular formations consisting of three autonomous agents serve as a class of benchmarks that can be used to test and compare the performances of different controllers. We present an algorithm that combines the advantages of both position- and distance-based gradient descent control laws. For example, only two pairs of neighboring agents need to be controlled, agents can work in their own local frame of coordinates and the orientation of the formation with respect to a global frame of coordinates is not prescribed. We first present a novel technique based on adding artificial biases to neighboring agents' range sensors such that their eventual positions correspond to a collinear configuration. Right after, a small modification in the bias terms by introducing a prescribed rotation matrix will allow the control of the bearing of the neighboring agents.

Keywords: Formation control, Distributed control, Multi-agent system.

1. INTRODUCTION

The theory of rigidity and in particular the concepts of *infinitesimal and minimal rigidity* have been proven to be very useful in formation control with the goal to define and achieve prescribed shapes by just controlling the distances between neighboring agents (see Anderson et al. (2008) and Krick et al. (2009)). The popular distance-based gradient descent algorithm for rigid formations has appealing properties. For example, the agents can work in their own local frame of coordinates, the system can be made robust against biased sensors, and since the orientation of the desired shape is not prescribed, one can induce rotational motions to the formation as presented in Garcia de Marina et al. (2015, 2016). In comparison, position-based formation control needs a common frame of coordinates and the steady-state orientation of the shape is restricted, which implies that a free rotational motion is not allowed; however, the same formation shape can be achieved by fewer pairs of neighboring agents in position-based than in distance-based control. This prompts a

search for compromise among the advantages of both formation control techniques.

The minimum total number of pairs of neighboring agents in a distance-based rigid formation control system is given by the necessary conditions for infinitesimal and minimal rigidity, which is equal to $2n - 3$ for 2D scenarios, where n is the total number of agents. This is directly related to the necessary number of inter-agent distances to be controlled such that the desired shape is (at least locally) uniquely defined. For example, the triangle is the simplest rigid shape in the 2D case and the necessary number of desired distances to define such a shape is three. On the other hand, one can achieve a desired triangular shape by just controlling (the relative positions rather than distances of) two neighboring agent pairs in the position-based setup. We illustrate these basic concepts in Figure 1.

This paper focuses on triangular formations consisting of three agents. This apparently simple setup has been considered as a benchmark in formation control (see Cao et al. (2007); Anderson et al. (2007); Cao et al. (2008); Liu et al. (2014); Mou et al. (2014)), since it allows detailed rigorous analysis for novel techniques as we aim at in this work. This provides a starting point in order to achieve more general formations. The goal of this paper is to propose an algorithm that combines the advantages of both distance- and position-based control, i.e., agents employ their own local frames of coordinates, no prescribed orientation for the desired shape (so rotational motions are allowed) and

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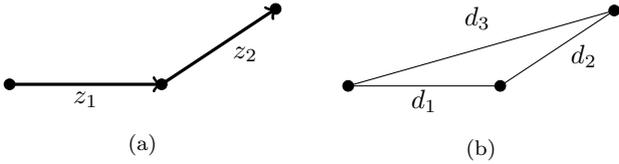


Fig. 1. The same triangular shape depicted by the three agents (dots) can be achieved by controlling the two vectors z_1 and z_2 (position-based control) or the three distances d_1 , d_2 and d_3 (distance-based control). Note that in a) the orientation of the triangle is prescribed while it is not in b).

smaller number of controlled pairs of neighboring agents than in distance-based minimally rigid formations.

The novelty of this algorithm lies in exploiting the sensing-error-induced collective motion, e.g., those caused by biased range sensors among neighboring agents. For example, it has been reported in Mou et al. (2016) that small constant biases in range sensors for a rigid formation cause the whole team of agents to converge to a distorted version of the desired shape and eventually it will exhibit some steady-state motion. On the other hand, it has been shown in Garcia de Marina et al. (2016) that one can introduce artificially such biases in order to steer the desired rigid formation in a controlled way.

This paper will first study the effect of these biases in a *flexible* (or non-rigid) distance-based formation of three agents, i.e., we control only two distances instead of three for a rigid triangular formation. We prove that if the range measurements between agents are not perfectly accurate then the three agents converge to collinear positions. This is somewhat counter-intuitive since for a non-rigid formation the steady-state shape is, depending on the initial conditions, arbitrary within a constraint set. Furthermore, we will show that depending on the value of these biases, the eventual collinear formation will move with a constant velocity or remain stationary. In our first step, we take advantage of this effect in order to align the three agents with the desired distances between them. Then, we will show how a small modification in this *biased* control law can achieve the control of the angle between the two relative vectors of the flexible formation, leading to the control of a stationary desired triangular shape with no restriction on its orientation.

In the following section we introduce some notations, briefly review the (exponential) stability of formation systems under the *flexible* shape and introduce the biases in the range measurements. We continue in Section 3 analyzing the consequences of such biases in the control of a non-rigid formation. We introduce a technique motivated by Mou et al. (2016), where the authors study non-minimally but infinitesimally rigid formations by *reducing* the problem to minimally rigid ones. In this paper since we are dealing with fewer pairs of neighboring agents than in minimally rigid setups, we instead *augment* our formation to a minimally rigid one in order to study the stability of a biased non-rigid formation. In Section 4 we introduce and analyze a modified version of the *biased* algorithm in order to control triangular shapes. We finish the paper with simulations in Section 5 and some conclusions in Section 6.

2. NON-RIGID FORMATION OF THREE AGENTS

2.1 Gradient descent distance-based formation control

We consider a team of three agents governed by the first-order kinematic model

$$\dot{p}_i = u_i, \quad (1)$$

where $p_i \in \mathbb{R}^2$ is the position of the agent $i = \{1, 2, 3\}$, and $u_i \in \mathbb{R}^2$ is the control action over agent i .

Let us define the two vectors

$$z_1 \triangleq p_1 - p_2, \quad z_2 \triangleq p_2 - p_3, \quad (2)$$

which are the relative positions corresponding to the two links available to the agents. For each link one can construct a potential function $V_k, k \in \{1, 2\}$ with its minimum at the desired distance d_k , so that the gradient of such functions can be used to control inter-agent distances distributively. We consider the following *shape potential function*

$$V_k(z_k) = \frac{1}{2}(\|z_k\| - d_k)^2, \quad (3)$$

with the following gradient along z_k

$$\nabla_{z_k} V_k(z_k) = \hat{z}_k(\|z_k\| - d_k), \quad (4)$$

where $\hat{z}_k \triangleq \frac{z_k}{\|z_k\|}$. We then apply to each agent i in (1) the following gradient descent control

$$u_i = -\nabla_{p_i} \sum_{k=1}^2 V_k(z_k), \quad (5)$$

and by denoting the distance error for the k th link by

$$e_k = \|z_k\| - d_k, \quad (6)$$

we arrive at the following dynamics

$$\begin{cases} \dot{p}_1 &= -\hat{z}_1 e_1 \\ \dot{p}_2 &= \hat{z}_1 e_1 - \hat{z}_2 e_2 \\ \dot{p}_3 &= \hat{z}_2 e_2. \end{cases} \quad (7)$$

System (7) has some interesting properties. For example, the agents can employ their own local systems of coordinates, i.e., a global or common frame of coordinates is not necessary, and collision avoidance is guaranteed among pairs of neighboring agents (Oh et al. (2015)). For *tree graph* topology, one can prove the (almost global) exponential convergence of the error system, i.e., the signals $e_1(t)$ and $e_2(t)$ both converge to zero exponentially fast if the agents do not start at the same positions (Dimarogonas and Johansson (2008); Sun et al. (2016)).

It is clear that although the error signals $e_1(t)$ and $e_2(t)$ converge to zero, the final relative positions may not guarantee any prescribed shapes as one would like to have for minimally rigid formations. In particular, agents p_1 and p_3 will lie somewhere on a circumference with the center at p_2 and the radius as d_1 and d_2 respectively. More precisely, the relative positions converge to the set

$$\mathcal{Z} \triangleq \{z : \|z_k\| = d_k, \forall k \in \{1, 2\}\}. \quad (8)$$

Furthermore, the exponential convergence of $e(t)$ and $z(t)$ to the origin and \mathcal{Z} , respectively, implies that $\dot{p}_1(t)$ and $\dot{p}_2(t)$ converge to zero exponentially fast, therefore the agents converge to only one stationary shape among all the ones defined by \mathcal{Z} , i.e., all the agents eventually stop.

2.2 Biases in distance-based formation control

It is clear from (5) with V_k as in (3) that neighboring agents i and j share the same potential V_k in the implementation of the gradient formation control. Let us focus on agent i whose control is

$$u_i = - \sum_{k=1}^2 u_i^k,$$

where u_i^k is the corresponding gradient $\nabla_{p_i} V_k$. More precisely, for the link k where i and j are neighbors we have that

$$u_i^k = \nabla_{p_i} V_k = \hat{z}_k(\|z_k\| - d_k) \quad (9)$$

$$u_j^k = \nabla_{p_j} V_k = -\hat{z}_k(\|z_k\| - d_k). \quad (10)$$

However, when the range sensors of the two agents are biased with respect to each other by a constant μ_k , without loss of generality, the pair (9)-(10) should be modified as

$$u_i^k = \hat{z}_k(\|z_k\| - d_k + \mu_k) \quad (11)$$

$$u_j^k = -\hat{z}_k(\|z_k\| - d_k). \quad (12)$$

The effects of these biases on undirected rigid formations have been studied in Mou et al. (2016) and how to remove or take advantage of them have been shown in Garcia de Marina et al. (2015, 2016). Flexible formations have not received much attention for formation control problems as the rigid ones since the sets like (8) do not define (locally) uniquely any shape. Nevertheless, we will show that the intentional introduction of biases (following the spirit of Garcia de Marina et al. (2016)) can make these apparently flexible formations appealing for formation control.

3. BIASED NON-RIGID FORMATION OF THREE AGENTS: CONVERGING TO A COLLINEAR STEADY-STATE

Let us consider biases in the range sensors of the agents in system (7). In particular, as it has been shown in (11)-(12), we can always write these biases from the point of view of agent 2, namely

$$\hat{p}_2 = \hat{z}_1(e_1 + \mu_1) - \hat{z}_2(e_2 - \mu_2), \quad (13)$$

which leads to the following *biased* system

$$\begin{cases} \dot{p}_1 &= -\hat{z}_1 e_1 \\ \dot{p}_2 &= \hat{z}_1 e_1 - \hat{z}_2 e_2 + \mu_1 \hat{z}_1 + \mu_2 \hat{z}_2 \\ \dot{p}_3 &= \hat{z}_2 e_2. \end{cases} \quad (14)$$

We shall now study the effect of imposing a special condition on the biases. This condition will be achievable if we are free to introduce biases into the controls, namely

$$\mu_1 = -\mu_2 = c, \quad (15)$$

for some constant $c \in \mathbb{R} \setminus \{0\}$. These biases can be considered as a parametric disturbance to the following closed-loop system involving the dynamics of z and e derived from (14):

$$\begin{cases} \dot{z}_1 &= -2\hat{z}_1 e_1 + \hat{z}_2 e_2 - c\hat{z}_1 + c\hat{z}_2 \\ \dot{z}_2 &= -2\hat{z}_2 e_2 + \hat{z}_1 e_1 + c\hat{z}_1 - c\hat{z}_2 \\ \dot{e}_1 &= -2e_1 + \hat{z}_1^T \hat{z}_2 e_2 - c + c\hat{z}_1^T \hat{z}_2 \\ \dot{e}_2 &= -2e_2 + \hat{z}_2^T \hat{z}_1 e_1 - c + c\hat{z}_2^T \hat{z}_1. \end{cases} \quad (16)$$

As we have discussed, one can derive the exponential stability of the closed-loop system (16) with $c = 0$ (Dimarogonas and Johansson (2008)). Therefore the stability

of (16) for small values of c is not compromised but at the cost of (probably) shifting its equilibrium from the one described in (8).

One realizes by a quick inspection in (14) together with (15)¹ that $e_1 = e_2 = 0$ with $\hat{z}_1 = \hat{z}_2 = \hat{z}^*$ is an equilibrium for the system (16), implying that all the agents are stationary. Note that the agents are collinear with agent 2 in the middle for this configuration. We define this equilibrium for (16) by

$$\mathcal{U}_d \triangleq \{z, e : (e_1 = e_2 = 0) \wedge (\hat{z}_1 = \hat{z}_2)\}. \quad (17)$$

However, a more careful inspection of (16) reveals that there is another equilibrium corresponding to $e_1 = e_2 = -\frac{2c}{3}$ with $\hat{z}_1 = -\hat{z}_2 = \hat{z}^*$, i.e., agents 1 and 3 converge to the same side of agent 2. This implies by inspecting (14) that the whole formation is travelling with velocity $\dot{p}_i = \frac{2c}{3}\hat{z}^*, \forall i = \{1, 2, 3\}$. We define this equilibrium for (16) by

$$\mathcal{U}_u \triangleq \{z, e : (e_1 = e_2 = -\frac{2c}{3}) \wedge (\hat{z}_1 = -\hat{z}_2)\}. \quad (18)$$

Proposition 1. Consider the closed-loop system (16) derived from the biased flexible formation control (14). If $\mu_1 = -\mu_2 = c$ with $c \in \mathbb{R} \setminus \{0\}$, then the only two equilibrium sets for (16) are \mathcal{U}_d and \mathcal{U}_u .

Proof. By using the last two equations of (16) one has that the following condition holds at the equilibrium

$$(2 + \hat{z}_1^T \hat{z}_2)(e_2 - e_1) = 0, \quad (19)$$

from which one can derive that regardless of \hat{z}_1 and \hat{z}_2 , one has that $e_1 = e_2 = e^*$ at an equilibrium. Note that this condition implies that the norms of z_1 and z_2 at the equilibrium are constant. We are going to check that such a situation can only occur at the equilibrium. For example, it might be possible that both vectors could be rotating while keeping their norms constant. Let us write the first two equations of system (16) and require that the error signals are constant

$$\frac{d}{dt} \begin{bmatrix} z_1 \\ z_2 \end{bmatrix} = \left(\begin{bmatrix} -\frac{2e^* - c}{\|z_1\|^*} & \frac{e^* + c}{\|z_2\|^*} \\ \frac{e^* + c}{\|z_1\|^*} & -\frac{2e^* - c}{\|z_2\|^*} \end{bmatrix} \otimes I_2 \right) \begin{bmatrix} z_1 \\ z_2 \end{bmatrix}, \quad (20)$$

where I_2 is the 2×2 identity matrix, the symbol \otimes denotes the Kronecker product, and we have that $\|z_1\|^*$ and $\|z_2\|^*$ are constant since $e_1 = e_2 = e^*$. Note that we have split $\hat{z}_{\{1,2\}} = \frac{1}{\|z_{\{1,2\}}\|^*} z_{\{1,2\}}$. It is clear that the matrix in (20) cannot be skew-symmetric since the off-diagonal terms always have the same sign, therefore we discard the possibility of rotations for z_1 and z_2 when the error signals are constant. Therefore $\dot{e}_1 = \dot{e}_2 = 0$ implies that the time derivatives $\dot{z}_1 = \dot{z}_2 = 0$ as well. By using the first two equations in (16) we can derive the following condition at an equilibrium of (16)

$$(\hat{z}_2^* - \hat{z}_1^*)(3e^* + 2c) = 0, \quad (21)$$

where z_1^* and z_2^* are two fixed vectors for z_1 and z_2 such that $e_1^* = e_2^* = e^*$.

¹ For arbitrary values of μ_1 and μ_2 the results of this paper apply as well but with different steady-state values for e_1 and e_2 . For the sake of simplicity and clarity we only consider the case where (15) holds. For more details, we refer to Mou et al. (2016) and Garcia de Marina et al. (2016).

It is clear from (21) that we only have two possible equilibrium sets. Let us check the case $\hat{z}_1^* = \hat{z}_2^*$. By checking the equilibrium of the third equation in (16) one can quickly derive that $e^* = 0$. So we have checked the equilibrium \mathcal{U}_d . For the second case $e^* = -\frac{2c}{3}$, one can check in the first equation in (16) that for its equilibrium the relation $\hat{z}_1^* = -\hat{z}_2^*$ must be satisfied. So we have checked the equilibrium \mathcal{U}_u . ■

One may wonder about the stability of \mathcal{U}_d and \mathcal{U}_u . In fact, if one is interested in having the three agents in a collinear fashion while they are stationary it would be desirable to know that at least \mathcal{U}_d is locally stable. We will see this fact in more detail in the following subsection.

3.1 Stability analysis of the error system

By inspecting the system (16) one realizes that the error system for the flexible formation is not autonomous. This is due to the fact that the term

$$\hat{z}_1^T \hat{z}_2 = \frac{1}{2\|z_1\|\|z_2\|} (\|p_3 - p_1\|^2 - \|z_1\|^2 - \|z_2\|^2), \quad (22)$$

depends on $\|p_3 - p_1\|^2$ and it cannot be written as a function of e_1 and e_2 . It would be desirable to handle a system involving only scalar variables than a mixture of scalar and vectorial ones as in (16). The key idea for starting the stability analysis of (16) is by including the following states

$$e_3 \triangleq \|z_3\| - d_3, \quad z_3 \triangleq p_3 - p_1,$$

where one can choose the value for d_3 depending on the equilibrium to be studied. For example, for the equilibrium \mathcal{U}_d we set $d_3 = d_1 + d_2$. The inclusion of these states will make the *augmented* system for the error vector $e \triangleq [e_1 \ e_2 \ e_3]^T$ autonomous and therefore easier to analyze.

Let us first write the dynamics of e derived from (14)

$$\begin{cases} \dot{e}_1 &= -2e_1 + \hat{z}_1^T \hat{z}_2 e_2 - c + c \hat{z}_1^T \hat{z}_2 \\ \dot{e}_2 &= -2e_2 + \hat{z}_2^T \hat{z}_1 e_1 - c + c \hat{z}_2^T \hat{z}_1 \\ \dot{e}_3 &= \hat{z}_3^T \hat{z}_1 e_1 + \hat{z}_3^T \hat{z}_2 e_2. \end{cases} \quad (23)$$

By employing the relation (22) and similarly for $\hat{z}_3^T \hat{z}_2$ and $\hat{z}_3^T \hat{z}_1$ one can derive the following self-contained dynamics for the error system

$$\begin{cases} \dot{e}_1 &= -2e_1 - c \\ &+ \frac{(e_2 + c) ((e_3 + d_3)^2 - (e_1 + d_1)^2 - (e_2 + d_2)^2)}{2(e_1 + d_1)(e_2 + d_2)} \\ \dot{e}_2 &= -2e_2 - c \\ &+ \frac{(e_1 + c) ((e_3 + d_3)^2 - (e_1 + d_1)^2 - (e_2 + d_2)^2)}{2(e_1 + d_1)(e_2 + d_2)} \\ \dot{e}_3 &= \frac{e_1 ((e_2 + d_2)^2 - (e_1 + d_1)^2 - (e_3 + d_3)^2)}{2(e_1 + d_1)(e_3 + d_3)} \\ &+ \frac{e_2 ((e_1 + d_1)^2 - (e_2 + d_2)^2 - (e_3 + d_3)^2)}{2(e_2 + d_2)(e_3 + d_3)}. \end{cases} \quad (24)$$

Now we linearize the system (24) at the equilibrium \mathcal{U}_d or equivalently $e = 0$ for $d_3 = d_1 + d_2$. We first compute the partial derivatives of

$$\begin{aligned} f(e) &= \frac{(e_3 + d_3)^2 - (e_1 + d_1)^2 - (e_2 + d_2)^2}{2(e_1 + d_1)(e_2 + d_2)} \quad (25) \\ \frac{\partial f(e)}{\partial e_1} \Big|_{e=0} &= -\frac{((e_3 + d_3)^2 - (e_1 + d_1)^2 - (e_2 + d_2)^2)}{2(e_1 + d_1)^2(e_2 + d_2)} \Big|_{e=0} \\ &= -\frac{2(e_1 + d_1)}{2(e_1 + d_1)(e_2 + d_2)} \Big|_{e=0} \\ &= -\frac{d_3^2 - d_1^2 - d_2^2}{2d_1^2 d_2} - \frac{1}{d_2} = -\frac{d_1 + d_2}{d_1 d_2} \\ \frac{\partial f(e)}{\partial e_2} \Big|_{e=0} &= -\frac{d_3^2 - d_1^2 - d_2^2}{2d_1 d_2^2} - \frac{1}{d_1} = -\frac{d_1 + d_2}{d_1 d_2} \\ \frac{\partial f(e)}{\partial e_3} \Big|_{e=0} &= \frac{d_3}{d_1 d_2} = \frac{d_1 + d_2}{d_1 d_2}, \end{aligned}$$

that in combination with

$$f(0) = \frac{d_3^2 - d_1^2 - d_2^2}{2d_1 d_2} = 1, \quad (26)$$

helps us to arrive at the Jacobian matrix defining the linearization of the autonomous system (24) at $e = 0$ as below

$$J_e = \begin{bmatrix} -2 - ca & 1 - ca & ca \\ 1 - ca & -2 - ca & ca \\ -1 & -1 & 0 \end{bmatrix}, \quad (27)$$

where $a = \frac{d_1 + d_2}{d_1 d_2}$ is always positive. Note that for the particular case $c = 0$ we have that in (27) the error signals e_1 and e_2 do not depend on e_3 and the points $e_1 = e_2 = 0$ are stable regardless of e_3 . In fact, in such a case we have a *flexible shape* where the steady state of e_3 will depend on the initial conditions $p(0)$, i.e., the agents converge to the set (8).

Remark 1. Note that by just adding a bias in a range sensor, no matter how small it is, we are linking the dynamics of the error signals e_1 and e_2 with e_3 . This has an important implication since in practice it is not common (or arguably possible) to have perfect measurements in the agents' sensors. Therefore, to have a *truly* flexible formation in practice would be quite challenging without any further action in (7). A technique employing estimators in order to remove the biases has been introduced in Garcia de Marina et al. (2015).

In order to check the stability of the origin of system (24) we compute the characteristic polynomial of (27)

$$\begin{aligned} P(\lambda) &= -\lambda(2 + ca + \lambda)^2 - 2ca(1 - ca) - 2ca(2 + ca + \lambda) \\ &+ \lambda(1 - ca)^2 \\ &= (\lambda + 1)(\lambda + 3)(\lambda + 2ca), \end{aligned} \quad (28)$$

therefore we can conclude that for any positive (negative) c the equilibrium \mathcal{U}_d is locally exponentially stable (unstable). The same analysis can be done for checking the stability of \mathcal{U}_u which is unstable (stable) for positive (negative) values of c . We summarize these findings with the following proposition.

Proposition 2. Consider system (14) for positive values of $c = \mu_1 = -\mu_2$, then the equilibria \mathcal{U}_d and \mathcal{U}_u for system (16) are (exponentially) stable and unstable respectively. The converse holds for negative values of c .

Remark 2. For negative values of c we have that the agents not only converge to a collinear formation but by inspecting (14) we have that $\dot{p}_i(t) \rightarrow z^* \frac{3c}{2}$, as $t \rightarrow \infty, \forall i \in \{1, 2, 3\}$, i.e., the agents will travel with a constant velocity.

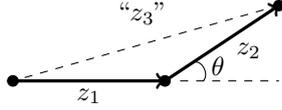


Fig. 2. The triangular shape can be defined by setting the norms $\|z_1\| = d_1$, $\|z_2\| = d_2$ and the angle θ between z_1 and z_2 .

Remark 3. Checking the eigenvalues given by $P(\lambda)$ show that one cannot increase the convergence speed by increasing c . Conversely it can be made arbitrarily slow by choosing a small positive c . In fact, the effect of a small bias in the range sensors of system (7) will be practically noticed after a sufficiently long time.

4. FLEXIBLE TRIANGULAR SHAPE

In this section we are going to show that the previous setup for collinear formations is just a particular case of a general setup where we can design controllers to achieve a prescribed angle between z_1 and z_2 .

Consider the following system

$$\begin{cases} \dot{p}_1 &= -\hat{z}_1 e_1 \\ \dot{p}_2 &= \hat{z}_1 e_1 - \hat{z}_2 e_2 + W(\theta)\hat{z}_1 - \hat{z}_2, \\ \dot{p}_3 &= \hat{z}_2 e_2. \end{cases} \quad (29)$$

where

$$W(\theta) = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}, \quad (30)$$

is a rotation matrix. One can check that the two cases $\mu_1 = -\mu_2 = 1$ and $\mu_1 = \mu_2 = -1$ in (14) are in fact particular cases of (29) with $\theta = 0$ and $\pm\pi$ radians respectively. In fact, the angle θ defines the prescribed angle between \hat{z}_1 and \hat{z}_2 in a target formation and note that the agents are not required to measure the actual angle between z_1 and z_2 . We proceed to show that the agents indeed can employ their own local frames.

Lemma 3. The agents in system (29) can employ their own local frames of coordinates.

Proof. Consider y_{ij} for the position of agent j with respect to a fixed local frame of coordinates for agent i with $j, i \in \{1, 2, 3\}$. Therefore, there exists a rotation matrix R_i and a translation vector τ_i such that $y_{ij} = R_i p_j + \tau_i$. Note that the vectors τ_i with $i \in \{1, 2, 3\}$ vanish for the calculation of z_1 and z_2 and the rotation matrices do not affect the value of e_1 and e_2 . Hence, agent 2 measures the signals $R_2 z_1$, $R_2 z_2$ and we have that $W(\theta)R_2 z_2 = R_2 W(\theta)z_2$ since rotation matrices commute in 2D. Thus, for agent 2 in global coordinates we have that

$$\begin{aligned} \dot{p}_2 &= R_2^T (R_2 \hat{z}_1 e_1 - R_2 \hat{z}_2 e_2 + R_2 W(\theta)\hat{z}_1 - R_2 \hat{z}_2) \\ &= \hat{z}_1 e_1 - \hat{z}_2 e_2 + W(\theta)\hat{z}_1 - \hat{z}_2. \end{aligned} \quad (31)$$

Therefore, one can conclude that the control action of agent 2 does not need any global information, e.g., a common frame of coordinates. The same reasoning can be applied to agents 1 and 3. ■

For the sake of convenience for the stability analysis, let us reformulate (29) as

$$\begin{cases} \dot{p}_1 &= -\hat{z}_1 e_1 \\ \dot{p}_2 &= \hat{z}_1 e_1 - \hat{z}_2 e_2 + W(\frac{\theta}{2})\hat{z}_1 - W(\frac{\theta}{2})^T \hat{z}_2. \\ \dot{p}_3 &= \hat{z}_2 e_2, \end{cases} \quad (32)$$

Note that we are still defining the same desired steady-state angle between \hat{z}_1 and \hat{z}_2 . The corresponding augmented error system derived from (32), with $d_3^2 = d_1^2 + d_2^2 + 2d_1 d_2 \cos \theta$, is given by

$$\begin{cases} \dot{e}_1 &= -2e_1 + \hat{z}_1^T \hat{z}_2 e_2 - \hat{z}_1^T W(\frac{\theta}{2})\hat{z}_1 + \hat{z}_1^T W(\frac{\theta}{2})^T \hat{z}_2 \\ \dot{e}_2 &= -2e_2 + \hat{z}_2^T \hat{z}_1 e_1 - \hat{z}_2^T W(\frac{\theta}{2})^T \hat{z}_2 + \hat{z}_2^T W(\frac{\theta}{2})\hat{z}_1 \\ \dot{e}_3 &= \hat{z}_3^T \hat{z}_2 e_2 + \hat{z}_3^T \hat{z}_1 e_1. \end{cases} \quad (33)$$

In order to analyze the stability of the linearization of (33) around $e = 0$ we need to work out some technical results first.

Lemma 4. Consider the matrix

$$M = \begin{bmatrix} p_1 & p_2 & a \\ p_2 & p_1 & a \\ b & c & 0 \end{bmatrix},$$

where $p_1, p_2, a, b, c \in \mathbb{R}$ with $p_1 > 0$, $p_1 > p_2$ and $p_1^2 > p_2^2$, so the second principal minor of M is positive definite. Consider a small perturbation $a < 0$. If $(b + c) > 0$, then $-M$ is Hurwitz.

Proof. First we note that if $a = 0$ the three eigenvalues of M are defined by the two positive eigenvalues λ_1 and λ_2 of its 2x2 leading principal submatrix and $\lambda_3 = 0$. Therefore a small perturbation a will not change the sign of the positive eigenvalues of M but we do not know about the sensitivity of the zero eigenvalue λ_3 with respect to the disturbance a . In order to check such sensitivity we derive the characteristic polynomial of M

$$\begin{aligned} P(\lambda) &= -\lambda(p_1 - \lambda)^2 + abp_2 + acp_2 - ab(p_1 - \lambda) + \lambda p_2^2 \\ &\quad - ac(p_1 - \lambda) \\ &= \lambda(p_2^2 - (p_1 - \lambda)^2 + a(b + c)) + a(b + c)(p_2 - p_1). \end{aligned} \quad (34)$$

Note that a can be chosen arbitrarily small. Since λ_3 is a continuous function of a , it can also be made arbitrarily small, so $|p_1| \gg |\lambda_3|$ and $|(p_2^2 - p_1^2)| \gg |a(b + c)|$. Then, without affecting the analysis of the sign of the perturbed $\tilde{\lambda}_3$, we can approximate the first bracket in (34) by $(p_2^2 - p_1^2)$. Therefore we can look at

$$P(\tilde{\lambda}_3) \approx \tilde{\lambda}_3(p_2^2 - p_1^2) + a(b + c)(p_2 - p_1) = 0, \quad (35)$$

in order to check the sign of the perturbed $\tilde{\lambda}_3$. We note that under the assumptions in the statement of the lemma one concludes that for a small negative a the eigenvalue $\tilde{\lambda}_3$ becomes positive. Hence $-M$ is Hurwitz. ■

Remark 4. The Jacobian in (27) fits the description of the matrix M in Lemma 4. Note that from Proposition 2 the sign of $\tilde{\lambda}_3$ from (27) only depends on the sign of a and not on its magnitude.

We now work out a series of calculations that will be needed for the calculation of the Jacobian matrix of (33) at $e = 0$. Consider the expression

² Note z_1 points at agent 2 and z_2 starts at agent 2, therefore we will have $\cos(\pi - \theta) = -\cos(\theta)$ in the law of cosines for Figure 2.

$$\hat{z}_2^T W(\alpha) \hat{z}_1 = \cos(\gamma - \alpha), \quad (36)$$

where γ is the actual angle between z_1 and z_2 and $\alpha \in \mathbb{R}$ is a constant angle. We also have that

$$\frac{\partial \hat{z}_2^T W(\alpha) \hat{z}_1}{\partial e} = \frac{\partial \cos(\gamma(e) - \alpha)}{\partial e}, \quad (37)$$

with

$$\gamma(e) = \arccos \left(\frac{(e_3 + d_3)^2 - (e_1 + d_1)^2 - (e_2 + d_2)^2}{2(e_1 + d_1)(e_2 + d_2)} \right), \quad (38)$$

and note that $\gamma(0) = \theta$. Therefore by applying the chain rule

$$\begin{aligned} \frac{\partial \cos(\gamma(e) - \alpha)}{\partial e_1} \Big|_{e=0} &= -\sin(\gamma(e) - \alpha) \frac{\partial \gamma(e)}{\partial e_1} \Big|_{e=0} \\ &= \sin(\theta - \alpha) \frac{1}{\sqrt{1 - \cos^2 \theta}} \left(-\frac{2d_1}{2d_1 d_2} - \frac{d_3^2 - d_1^2 - d_2^2}{d_1^2 d_2} \right) \\ &= -\frac{\sin(\theta - \alpha)}{\sin(\theta)} \left(\frac{d_1 + d_2 \cos \theta}{d_1 d_2} \right) =: a_1 \end{aligned} \quad (39)$$

$$\frac{\partial \cos(\gamma(e) - \alpha)}{\partial e_2} \Big|_{e=0} = -\frac{\sin(\theta - \alpha)}{\sin(\theta)} \left(\frac{d_2 + d_1 \cos \theta}{d_1 d_2} \right) =: a_2 \quad (40)$$

$$\frac{\partial \cos(\gamma(e) - \alpha)}{\partial e_3} \Big|_{e=0} = \frac{\sin(\theta - \alpha)}{\sin \theta} \frac{d_3}{d_1 d_2} =: a_3. \quad (41)$$

Remark 5. We highlight again that α is a constant parameter that can be chosen arbitrarily. For example, the computation of a_3 for $\theta = 0$, i.e., $d_3 = d_1 + d_2$, and $\alpha = 0$ becomes

$$\lim_{\theta \rightarrow 0} a_3(\theta) = \lim_{\theta \rightarrow 0} \frac{\sin(\theta - 0)}{\sin \theta} \frac{d_1 + d_2}{d_1 d_2} = \frac{d_1 + d_2}{d_1 d_2}, \quad (42)$$

which is consistent with the values obtained for the Jacobian in (27).

Similarly as in (14) with (15), we also introduce a gain $c > 0$ in (32) multiplying the terms $W(\frac{\theta}{2})\hat{z}_1$ and $W(\frac{\theta}{2})^T \hat{z}_2$. Now we are ready to present our main result.

Theorem 5. For the system

$$\begin{cases} \dot{p}_1 &= -\hat{z}_1 e_1 \\ \dot{p}_2 &= \hat{z}_1 e_1 - \hat{z}_2 e_2 + c \left(W(\frac{\theta}{2})\hat{z}_1 - W(\frac{\theta}{2})^T \hat{z}_2 \right) \\ \dot{p}_3 &= \hat{z}_2 e_2, \end{cases} \quad (43)$$

its equilibrium defined by the set $\mathcal{U} \triangleq \{p_1, p_2, p_3 : W(\theta)\hat{z}_1 - \hat{z}_2 = 0, \|z_1\| = d_1, \|z_2\| = d_2\}$, for every $\theta \in (-\pi, \pi)$ is locally exponentially stable for a sufficiently small $c > 0$.

Proof. We first note that the study of the stability of the equilibrium \mathcal{U}_d corresponds to checking the eigenvalues of the Jacobian of the autonomous *augmented* error system (33), with the introduced gain c , at $e = 0$. The Jacobian is given by

$$J_e = \begin{bmatrix} -2 + ca_1 & \cos(\theta) + ca_2 & ca_3 \\ \cos(\theta) + ca_1 & -2 + ca_2 & ca_3 \\ -\frac{d_1 + d_2 \cos \theta}{d_3} & -\frac{d_2 + d_1 \cos \theta}{d_3} & 0 \end{bmatrix}, \quad (44)$$

where a_1, a_2 and a_3 are as in (39)-(41) respectively. We notice that the following expression

$$\begin{aligned} & - \left(\frac{d_1 + d_2 \cos \theta}{d_3} + \frac{d_2 + d_1 \cos \theta}{d_3} \right) = \\ & - \frac{d_1(1 + \cos \theta) + d_2(1 + \cos \theta)}{d_3}, \end{aligned} \quad (45)$$

is always negative for $\theta \in (-\pi, \pi)$. Since $\alpha = \frac{\theta}{2}$ one can also check that (41) is always positive. Indeed, the second principal minor is not symmetric (negative definite) anymore because of the added small perturbations ca_1 and ca_2 , but its eigenvalues are still with real negative part and bounded from zero for all θ . Furthermore, the diagonal elements and off-diagonal elements of the second principal minor are very similar respectively for a very small $c > 0$. Therefore from the eigenvalue sensitivity analysis in Lemma 4 one can derive that indeed (44) is Hurwitz for a very small gain $c > 0$. Note that the local exponential convergence of the augmented $e(t)$ to the origin implies that the agents' velocities in (32) converge to zero as well, so the agents converge to fixed positions. ■

Remark 6. Note that in system (43), the set of collinear positions for all the agents is not invariant. Therefore, it outperforms in that sense the distance-based control of rigid formations, where if the agents start collinear, they remain collinear.

5. SIMULATIONS

In this section we validate the results in Proposition 2 and Theorem 5. We first show in Figure 3 that for $\mu_1 = -\mu_2 = 1$ in system (14) the equilibrium \mathcal{U}_d is stable, where all the agents converge to collinear positions with $\hat{z}_1 = \hat{z}_2$. The desired distances are $d_1 = 30$ and $d_2 = 10$, and the agents 1, 2 and 3 are marked with red, green and blue colors respectively. We then show in Figure 4 that the equilibrium \mathcal{U}_u is stable for $\mu_2 = -\mu_1 = 1$. Note that the agents do not stop moving and the distance errors converge both to $\frac{2}{3}$.

We move on to control a triangular shape by employing the results from Theorem 5. The desired angle between \hat{z}_1 and \hat{z}_2 has been set to 60 degrees. The evolution of the agents can be seen in Figure 5, where the left inner plot shows the evolution of the *inter-angle* and the right inner plot the evolution of the distance errors.

6. CONCLUSIONS

We have presented a formation control algorithm for achieving triangular shapes. The algorithm is based on a new technique derived from the biased distance-based gradient descent control of two links. The algorithm enjoys the advantages of both position-based and distance-based formation controls. Current research is aimed at providing a systematic approach for achieving stable arbitrary shapes consisting of more than three agents in non-rigid formation. We have already made some progress for chain topologies Garcia de Marina et al. (2017).

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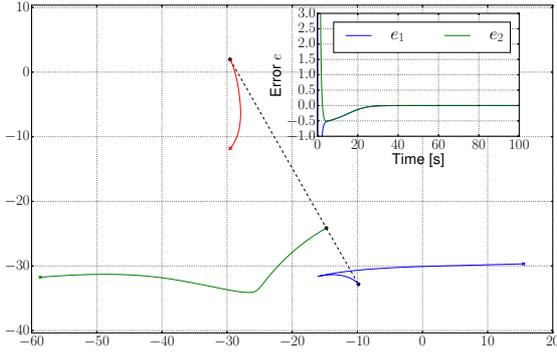


Fig. 3. Simulation result from Proposition 2. If $\mu_1 = -\mu_2 = 1$, then the equilibrium set \mathcal{U}_d is stable and all the agents stop eventually once the distance errors converge to zero (shown in the inner plot). In this equilibrium, we recall that $\hat{z}_1 = \hat{z}_2$, therefore we have that the agent 2 marked in green will finish between the other two agents.

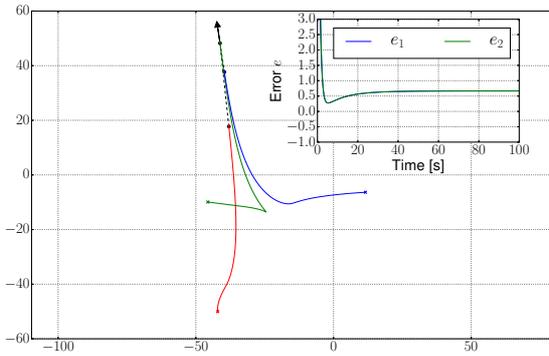


Fig. 4. Simulation result from Proposition 2. If $\mu_2 = -\mu_1 = 1$, then the equilibrium set \mathcal{U}_u is stable and the formation describes a steady-state collective motion (denoted by the black arrow) in the direction of \hat{z}_2 . Note that $\hat{z}_1 = -\hat{z}_2$, therefore the agent 2 marked in green is leading the motion. Both distance errors e_1 and e_2 (in the inner plot) converge to $\frac{2}{3}$, regardless of the prescribed distances.

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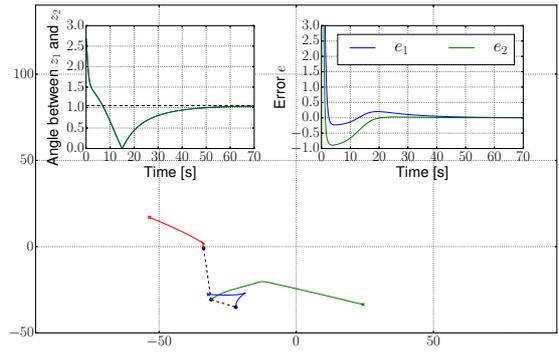


Fig. 5. Simulation result from Theorem 5. In addition to the distance control between agents 1, 2 and 2, 3 (right inner plot), the agent 2 marked in green color is controlling the angle between \hat{z}_1 and \hat{z}_2 (left inner plot). Note that the final orientation of the shape is not controlled and depends on the initial conditions.

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