Finite-time Motion Planning of Multi-agent Systems with Collision Avoidance

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Abstract—Finite-time motion planning with collision avoidance is a challenging issue in multi-agent systems. This paper proposes a novel distributed controller based on a new Lyapunov barrier function which guarantees finite-time stability for multi-agent systems without collisions. First, the problem of finite-time motion planning of multi-agent systems is formulated. Then, a novel finite-time distributed controller is developed based on a Lyapunov barrier function. Finally, numerical simulations demonstrate the effectiveness of proposed method.

Index Terms—Motion planning, finite-time stability, distributed controller, multi-agent systems

I. INTRODUCTION

Multi-agent systems have received much interest in recent years, thanks to its applicability to a wide range of study fields and real-world applications. Various distributed coordination and control problems, such as consensus (also known as agreement/synchronization/rendezvous), formation, distributed observation, estimation and optimization, have been introduced and implemented in many practical industrial applications [1]–[3].

When it comes to multi-vehicle systems in particular, a common ground may be the requirement for several agents to collaborate in order to achieve one or more cooperative goals. In such circumstances, the available patterns on sensing and information exchange, as well as physical/environmental limits and intrinsic limitations, inherently dictate coordination and control. As a result, motion planning, coordination, and control has been, and continues to be, a hot focus of research in the robotics and control societies.

Inter-agent collision avoidance, convergence to spatial destinations/regions or tracking of reference signals/trajectories, maintenance of information exchange among agents, and avoidance of physical obstacles are the main concerns when coordinating the motions of multi-vehicle or multi-robot (the terms are used interchangeably) teams. [6]–[11] deal with flocking and [12]–[17] works on consensus, rendezvous, and/or formation control. Collision avoidance is a non-negotiable condition in such problems, and potential function approaches and Lyapunov-based analysis provide useful and effective ways to handle it. Recently, a method based on Lyapunov barrier function is proposed and shows its soundness in coping with multi-task multi-agent problems [4], [5]. Please check the survey of formation control by using artificial potential functions with various different setting-ups [18].

Despite these objectives, finite time stability is, however, seldom achieved. Finite Time Stability (FTS) is of essential importance because in real cases multiple agents need to one or multiple common goals within limited time. [14] focuses on continuous autonomous systems and provides necessary and sufficient conditions based on the Lyapunov stability theory. In [20]–[23], consensus and formation control problems within finite time framework are well investigated by several classes of protocols. An inspiring work studied the problem of FTS for a single system by using barrier functions in [24].

Different from the work [24], we propose a finite-time distributed controller without collision for multi-agent system, and the contribution of this paper is threefold. First, a new type of Lyapunov barrier function is constructed, whose derivative satisfy specific properties such that its gradient is non-zero everywhere except the equilibrium point. Second, by using the developed Lyapunov barrier function, we provide the distributed controller for multi-agent systems with considering collision avoidance. Finally, distinct with the work in literature, the benefits of proposed method can also be found in the situations where unsafe region or boundaries of safe region exists, which allows a more practical implementation environment for multi-agent systems.

II. MODELING AND PROBLEM STATEMENT

Consider a network of $N$ mobile agents deployed in a known workspace $W$. Each agent $i \in \{1, \ldots, N\}$ is modeled as a circular disk of radius $r$ and circular agents are centered at known positions $x_i$, $i \in \{1, \ldots, N\}$. Moreover, we assume all kinetic agents are located within a bounded circle of radius $R$. To achieve this, we place $M$ static agents with radius of $r$ along the circular boundary. With collision avoidance of agents on the boundary, kinetic agents remain in the circle according to
designed barrier function. As a result, network connectivity is preserved and one agent can receive the position and velocity information of other agents. Now consider one kinetic agent \(i\). Its motion under single integrator dynamics is

\[
\dot{x}_i = v_i
\]

where \(x_i, v_i \in \mathbb{R}^n\)

. The problem of reaching to a specified goal position in finite time can be formulated mathematically as follows:

\[
\exists t^* < \infty, \forall t > t^* \|x_i(t) - \tau_i\| = 0
\]

where \(\tau_i\) is the desired goal location. Collision avoidance between agents can be written as:

\[
\forall t > t_0 \|x_i(t) - x_j(t)\| > d_c
\]

where \(x_i(t)\) and \(x_j(t)\) represent the location of the agents and \(t_0\) is the starting time. Here, we assume that \(d_c\) is bigger than \(2r\) and the distances between goal positions \(\|\tau_i - \tau_j\| > d_c\). We also assume that one agent starts sufficiently far away from the other agents so that \(\|x_i(t_0) - x_j(t_0)\| > d_c\). What we should do is to design a feedback-law \(v_i\) such that all agents reach their own goal positions within finite time while maintaining safe distance between each other.

### III. Motion Coordination

All the agents initiate in the closed region where the wireless communication links can be established. As the agents are moving inside the region during the process, their wireless communication can remain stable, thus resulting in connectivity maintenance. For any one of the agents, for example the \(i^{th}\) agent \(x_i\), can not only know the location of the boundary but also perceive the position and velocity of its nearest agent, for example \(x_j\). We assume that: (1) The goal positions \(\tau_i, i \in \{1, 2, \ldots, n\}\) are static; (2) There are no physical obstacles in the region; (3) The distance between any two agents’ goal positions \(\|\tau_i - \tau_j\| > d_c\); (4) All agents are equal, i.e., there is no leader among them. Since all the agents are equal, it suffices to focus on one agent. We seek a continuous feedback law \(u_i\) for the \(i^{th}\) agent to achieve multi-agent coordination with obstacle avoidance. More specifically, we seek a barrier-function based controller for the system. We define the barrier function for the \(i^{th}\) agent as follows:

\[
B_i(x_i, x_j) = \frac{\|x_i - \tau_i\|^2}{\|x_i - x_j\|^2 - d_c + \frac{1}{\varepsilon}}
\]

where \(\varepsilon \gg 1\) is a very large number. We define the controller as follows:

\[
v_i = \begin{cases} 
-k_1 \|\nabla B_{ix_i}\|^{\alpha-1} \nabla B_{ix_i} + \left(1 - \frac{2(x_i - \tau_i)^T v_j}{x_0(\nabla B_{ix_i})^T v_j}\right) v_j & \text{if } x \neq \tau_i \\
0 & \text{if } x = \tau_i 
\end{cases}
\]

where \(k_1 > 0\) and \(0 < \alpha < 1\) With this controller, we have the following result:

**Theorem 1**: Under the control law, the point

\[
x = \tau_i
\]

is FTS equilibrium for the system and the agent will remain collision avoidance w.r.t any other agents.

Before presenting the proof, we present some useful Lemmas:

**Lemma 1**: Under the control law, the point \(x = \tau_i\) is an equilibrium for the system, i.e.,

\[
\lim_{x_i \to \tau_i} -k_1 \|\nabla B_{ix_i}\|^{\alpha-1} \nabla B_{ix_i} + \left(1 - \frac{2(x_i - \tau_i)^T v_j}{x_0(\nabla B_{ix_i})^T v_j}\right) v_j = 0
\]

**Proof**: Consider:

\[
\begin{align*}
\left(\nabla B_{ix_i}\right)^T v_j &= \frac{2(x_i - \tau_i)^T - \frac{\|x_i - \tau_i\|^2}{x_0(x_i - x_j)^T}}{(x_i - \tau_i) \cdot v_j} v_j \\
&= 2 - \|x_i - \tau_i\|^2 \frac{(x_i - x_j) \cdot v_j}{x_0(x_i - \tau_i) \cdot v_j} v_j \\
&= 2 - \|x_i - \tau_i\|^2 \frac{(x_i - x_j) \cdot v_j}{x_0\|v_j\|\cos \theta} v_j
\end{align*}
\]

Therefore,

\[
\lim_{x_i \to \tau_i} \left(\nabla B_{ix_i}\right)^T v_j = 2
\]

\[
\Rightarrow \lim_{x_i \to \tau_i} \frac{(x_i - \tau_i)^T \cdot v_j}{\nabla B_{ix_i}^T \cdot v_j} = \frac{1}{2}
\]

\[
\Rightarrow \lim_{x_i \to \tau_i} \left(1 - \frac{2(x_i - \tau_i)^T \cdot v_j}{x_0(\nabla B_{ix_i})^T \cdot v_j}\right) v_j = 0
\]

Since

\[
\nabla B_{ix_i}(\tau_i) = 0, \lim_{x_i \to \tau_i} -k_1 \|\nabla B_{ix_i}\|^{\alpha-1} \nabla B_{ix_i} = 0
\]

which leads to

\[
\lim_{x_i \to \tau_i} -k_1 \|\nabla B_{ix_i}\|^{\alpha-1} \nabla B_{ix_i} + \left(1 - \frac{2(x_i - \tau_i)^T \cdot v_j}{x_0(\nabla B_{ix_i})^T \cdot v_j}\right) v_j = 0
\]

**Lemma 2**: Time derivative of the barrier function \(\dot{B}_i\) satisfies:

\[
\dot{B}_i = -k_1 \|\nabla B_{ix_i}\|^{\alpha+1}
\]
Proof:

\[ \bar{B}_i = (\nabla B_{ix_i})^T \cdot v_i + (\nabla B_{ix_i})^T \cdot v_j \]
\[ = (\nabla B_{ix_i})^T \cdot [-k_1 \| \nabla B_{ix_i} \|^a + \nabla B_{ix_i}] \]
\[ + \left( 1 - \frac{2(\alpha - \tau_i)^T \cdot v_j}{x_0 (\nabla B_{ix_i})^T \cdot v_j} \right) v_j \]
\[ = -k_1 \| \nabla B_{ix_i} \|^a + 2 \cdot \frac{(\alpha - \tau_i)^T \cdot v_j}{x_0 (\nabla B_{ix_i})^T \cdot v_j} \]
\[ - (\nabla B_{ix_i})^T \cdot v_j \cdot \frac{2(\alpha - \tau_i)^T \cdot v_j}{x_0 (\nabla B_{ix_i})^T \cdot v_j} \]
\[ = -k_1 \| \nabla B_{ix_i} \|^a \]

Lemma 3: In the domain \( D_0 = \{ x \mid \| x - x_j \| > d_c \} \),
\[ B(x) \leq \epsilon \| x - \tau_i \|^2 \]  \hspace{1cm} (23)

Proof:
\[ \| x - x_j \| \geq d_c \Rightarrow \| x - x_j \| - d_c \geq 0 \]
\[ \Rightarrow \| x - x_j \| - d_c + \frac{1}{\epsilon} \geq \frac{1}{\epsilon} \]
\[ \Rightarrow \frac{1}{\epsilon} \leq \| x - x_j \| - d_c + \frac{1}{\epsilon} \]
\[ \Rightarrow B_i = \frac{\| x - \tau_i \|^2}{\| x - x_j \| - d_c + \frac{1}{\epsilon}} \leq \epsilon \| x - \tau_i \|^2 \]  \hspace{1cm} (27)

Lemma 4: \( \nabla B_{ix_i} \) is non-zero everywhere except the equilibrium point \( \tau_i \), the point \( x = \tau_i + 2 \frac{\| x_j - \tau_i \| + d_c - \frac{1}{\epsilon}}{\| x_j - \tau_i \|}(x_j - \tau_i) \)  \hspace{1cm} (28)

and the point \( x = x_j + 2 \frac{\| \alpha - x_j \| + d_c - \frac{1}{\epsilon}}{\| \alpha - x_j \|}(\alpha - x_j) \)  \hspace{1cm} (29)

Proof: Denote \( \nabla B_{ix_i} = \frac{\partial B_{ix_i}}{\partial x_i}, \nabla B_{ix_j} = \frac{\partial B_{ix_j}}{\partial x_j} \), and \( x_0 = \| x_i - x_j \| - d_c + \frac{1}{\epsilon} \)
\[ \nabla B_{ix_i} = 2 \frac{x - \tau_i}{x_0} - \frac{\| x - \tau_i \|^2}{x_0^2} \frac{x - o}{\| x - o \|} \]  \hspace{1cm} (30)
solve the equation gives:
\[ x = \tau_i + 2 \frac{\| x_j - \tau_i \| + d_c - \frac{1}{\epsilon}}{\| x_j - \tau_i \|}(x_j - \tau_i) \]  \hspace{1cm} (31)
or
\[ x = x_j + 2 \frac{\| \alpha - x_j \| + d_c - \frac{1}{\epsilon}}{\| \alpha - x_j \|}(\alpha - x_j) \]  \hspace{1cm} (32)

Lemma 5: In any closed, compact domain \( D \subset \mathbb{R}^n \) containing point \( \tau_i \) and excluding the region \( D = \{ x \mid \| x - (x_j + 2 \frac{\| \alpha - x_j \| + d_c - \frac{1}{\epsilon}}{\| \alpha - x_j \|}(\alpha - x_j) \| < r \} \)
\[ \| x - [x_j + 2 \frac{\| \alpha - x_j \| + d_c - \frac{1}{\epsilon}}{\| \alpha - x_j \|}(\alpha - x_j)] \| < r \] where \( r \) is an arbitrary small positive number, the gradient of the barrier function satisfies:
\[ \| \nabla B_{ix_i} \| \geq c \| x - \tau_i \| \]  \hspace{1cm} (33)

where \( c > 0 \)

Proof: It can be easily verified that \( \nabla B_{ix_i}(\tau_i) = 0 \). Select \( D_1 = \{ x \mid \| x - \tau_i \| < \Delta \} \), where \( \Delta \) is a very small positive number. Choose domain \( D = D \setminus D_1 \). Since \( D \) doesn’t include \( \bar{D} \), \( D \) does not include the point as in Lemma 4. Hence, from Lemma 4, at any point \( x \in \bar{D}, \nabla B_{ix_i} \neq 0 \) and since \( \bar{D} \) is a closed domain, we can find \( c_1 = \min_{x \in \bar{D}} \| \nabla B_{ix_i} \| > 0 \).

Therefore, we have that \( \forall x \in \bar{D}, \| \nabla B_{ix_i} \| \geq c_1 \| x - \tau_i \| \)
Next, consider \( D_2 = \{ x \mid \| x - \tau_i \| \leq \Delta \} \). In a very small neighborhood of \( \tau_i \), the Hessian matrix \( \nabla^2 B_{ix_i} > 0 \). Hence, using First-order condition for convexity, we have that \( \forall x \in D_2 \):
\[ B_i(\tau_i) \geq B_i + \nabla B_{ix_i}^T (\tau_i - x) \]
\[ \Rightarrow 0 \geq B_i - B_i + \nabla B_{ix_i}^T (\tau_i - x) \]
\[ \Rightarrow \nabla B_{ix_i}(x - \tau_i) \geq B_i \]  \hspace{1cm} (36)

It is obvious that \( B_i \) can be bounded as \( B_i \geq c_2 \| x - \tau_i \|^2 \). Also, using Cauchy-Schwartz inequality, we have that \( \| \nabla B_{ix_i}(x - \tau_i) \| \leq \| \nabla B_{ix_i} \| \| x - \tau_i \| \)
Therefore,
\[ \| \nabla B_{ix_i} \| \| x - \tau_i \| \geq \| \nabla B_{ix_i} \| \| x - \tau_i \| \]
\[ \geq B_i \geq c_2 \| x - \tau_i \|^2 \]  \hspace{1cm} (38)
\[ \Rightarrow \nabla B_{ix_i} \geq c_2 \| x - \tau_i \|. \]  \hspace{1cm} (39)

Now we can give the proof of Theorem 1:

Proof: From Lemma 4, we have that \( \nabla B_{ix_i} = 0 \) at the equilibrium point \( \tau_i \) and at the point \( x = \tau_i + \mu (o - \tau_i) \) where \( \mu \) takes the value as per Lemma 3. Let’s assume that the initial condition is such that \( x(t_0) \) doesn’t lie in \( \bar{D} \) defined as per Lemma 4. Consider the open domain around the desired location \( D_0 \) as defined in Lemma 2. Define \( D = D_0 \setminus D \). Since \( D \) is a closed domain and \( D_0 \) is open, domain \( D \) is an open domain around the equilibrium \( \tau_i \). Choose the candidate Lyapunov function
\[ V_i = B_i \]  \hspace{1cm} (40)

From Lemma 2 we have:
\[ V_i = -k_1 \| \nabla B_{ix_i} \|^a \]  \hspace{1cm} (41)

From Lemma 5 we have:
\[ \| \nabla B_{ix_i} \| \geq c_0 \| x - \tau_i \| \Rightarrow \| x - \tau_i \| \leq \| \nabla B_{ix_i} \| \]  \hspace{1cm} (42)

From Lemma 3 we have:
\[ V_i = B_i \leq \epsilon \| x - \tau_i \|^2 \leq \epsilon \| \nabla B_{ix_i} \|^2 \]  \hspace{1cm} (43)
Therefore, we have:

\[
\dot{V}_i = -k_1 \|\nabla B_{ix_i}\|^\alpha + 1
\]

\[
\leq -k_1 \frac{c\alpha + 1}{\epsilon^{\alpha + 1}} (B_i)^{\frac{\alpha + 1}{\epsilon^{\alpha + 1}}} = -k_1 \frac{c\alpha + 1}{\epsilon^{\alpha + 1}} (V_i)^{\frac{\alpha + 1}{\epsilon^{\alpha + 1}}}
\]

If we set \(k_1 \frac{c\alpha + 1}{\epsilon^{\alpha + 1}} = c > 0\) and \(0 < \frac{\alpha + 1}{\epsilon^{\alpha + 1}} = \beta < 1\), then

\[
\dot{V}_i \leq -cV_i^\beta
\]

which satisfies the condition of FTS.

IV. NUMERICAL SIMULATIONS

A. Example 1: Cases of 4 agents

To verify the efficacy of the proposed controller, we demonstrated the whole process using MATLAB. The first case involves 4 agents, where the radius of every agent is set to be \(r = 0.99\). The radius of total move area \(R = 98\). We choose \(d_c = 2\), \(\epsilon = 10000\) in the barrier function, and \(\alpha = \frac{1}{3}\) in the controller. The time interval for the simulation is \(10^{-3}s\). The evolution of the inter-agent distances during the entire simulation time is depicted in Fig. 1. The motion of the followers towards their destinations is depicted in Fig. 3. All agents arrive at their destinations within finite time.

B. Example 2: Cases of 20 agents

The second case involves 20 agents and the time interval is set to be \(2 \times 10^{-3}s\). Other settings are the same as case 1. The evolution of the inter-agent distances during the entire simulation time is depicted in Fig. 2. The motion of the followers towards their destinations is depicted in Fig. 4. All agents arrive at their destinations within finite time.

V. CONCLUSION

In this paper, we propose a finite-time distributed controller with considering collision avoidance for multi-agent systems. Our main contribution lies in three aspects: First, a new type of Lyapunov barrier function is proposed, whose derivative satisfies specific properties such that its gradient is non-zero everywhere except the equilibrium point; Then, a distributed controller for multi-agent systems with considering collision avoidance is provided by using the developed Lyapunov barrier function; Finally, the proposed method can deal with the situations where unsafe region or boundaries of safe region exists, providing a more practical implementation environment for multi-agent systems.
Fig. 3. States of four agents

Fig. 4. States of twenty agents
This paper considers a simple first-order dynamical multi-agent system, while sophisticated models could be investigated with a more practical setting up. Therefore, our future efforts will be devoted to the follows: A second-order dynamic model of nonholonomic mobile agents will be considered with applications in real wheeled swarm robots. We will consider more practical constraints in the agents’ sensing and communication, like the problem of connectivity maintainance and preservation. Design of robust distributed controller is a promising way to cope with the disturbance in dynamics and communications, and parametric Lyapunov-like barrier function might be exploited. Learning-based methods by using Lyapunov-barrier functions will also be attempted.

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