# ON ROBUST RENDEZVOUS FOR MOBILE AUTONOMOUS AGENTS 

Sonia Martínez* Jorge Cortés ** Francesco Bullo*

\author{

* Department of Mechanical and Environmental Engineering, University of California, Santa Barbara, California 93106, USA <br> \{smartine, bullo\}@engineering.ucsb.edu <br> ** Department of Applied Mathematics and Statistics, University of California, Santa Cruz, California 95064, USA <br> jcortes@soe.ucsc.edu
}


#### Abstract

This paper presents coordination algorithms for networks of mobile autonomous agents. The objective of the proposed algorithms is to achieve rendezvous, that is, agreement over the location of the agents in the network. We provide analysis and design results for multi-agent networks in arbitrary dimensions under weak requirements on the switching and failing communication topology. The correctness proof relies on proximity graphs and their properties and on a LaSalle Invariance Principle for nondeterministic discrete-time systems.


Keywords: Networked control systems, rendezvous, proximity graphs, non-deterministic discrete-time dynamical systems, LaSalle Invariance Principle

## 1. INTRODUCTION

This work is a contribution to the emerging discipline of motion coordination for ad-hoc networks of mobile autonomous agents. With this loose terminology we refer to groups of robotic agents with limited mobility and communication capabilities. In the future these groups of coordinated devices will perform a variety of challenging tasks including search and recovery operations, surveillance, exploration and environmental monitoring. The potential advantages of employing arrays of agents have recently motivated vast interest in this topic. From a control viewpoint, a group of agents inherently provides robustness to failures of single agents or of communication links.

The motion coordination problem for groups of autonomous agents is a control problem in the presence of communication constraints. Typically,
each agents makes decisions based only on partial information about the state of the entire network that is obtained via communication with its immediate neighbors. An important difficulty is that the topology of the communication network depends on the agents' locations and, therefore, changes with the evolution of the network.

The "multi-agent rendezvous" problem and a first "circumcenter algorithm" have been introduced by Ando and coworkers in (Ando et al., 1999). This algorithm has been extended to various (a)synchronous stop-and-go strategies in (Lin et al., 2003; Lin et al., 2004a). A related algorithm, in which connectivity constraints are not imposed, is proposed in (Lin et al., 2004b). These schemes are memoryless (static feedback), anonymous (all agents are indistinguishable), and spatially distributed (only local information is required). An incomplete list of recent works on
motion coordination algorithms includes (Suzuki and Yamashita, 1999; Justh and Krishnaprasad, 2004) on pattern formation, (Jadbabaie et al., 2003) on flocking, (Klavins et al., 2004) on self-assembly, (Liu and Passino, 2004) on foraging, (Ögren et al., 2004) on gradient climbing, and (Cortés et al., 2004b) on deployment. Consensus and distributed decision making protocols are discussed in (Olfati-Saber and Murray, 2004; Moreau, 2003).
In this paper we provide novel analysis and design results on a class of rendezvous algorithms. First, we define and analyze a class of "circumcenter algorithms" defined over switching communication topologies. We classify communication topologies for our algorithms via the notion of "proximity graphs," see (Jaromczyk and Toussaint, 1992) and (Cortés et al., 2004b). Admissible communication topologies are proximity graphs with the property of being "spatially distributed" over the disk graph (i.e., they can be computed with only the local information encoded in the disk graph) and such that their connected components have the same vertices as the disk graph. This is a more general class of communication topologies than the ones adopted in most works on motion coordination including (Ando et al., 1999; Lin et al., 2004a; Lin et al., 2004b). The ability to rely on general communication topologies is advantageous in the design of wireless communication strategies and is referred to as "topology control", see (Li, 2003) and references therein.
Second, we consider networks of agents whose state space is $\mathbb{R}^{d}$, where $d \in \mathbb{N}$. We prove that our proposed class of circumcenter algorithms is correct in arbitrary dimensions and include simulations in two and three dimensions. As a natural outcome, we prove that the original circumcenter algorithm in (Ando et al., 1999) can be adapted to work in higher dimensions, and that it is guaranteed to converge in finite time.

Third, we establish a general theorem on the robustness of the proposed class of circumcenter algorithms with respect to communication link failures. Rendezvous is guaranteed even if each agent experiences link failures, provided the resulting directed communication graph is strongly connected at least once every finite number of time instants. Our results provide the first contribution to the theoretical explanation of the robustness properties of the circumcenter algorithm observed in computer simulations in (Ando et al., 1999).

Because of length constraints, we refer the interested reader to (Cortés et al., 2004a) for all the proofs. We only mention here that the (novel) method of proof is based on a recently-developed LaSalle Invariance Principle for nondeterministic discrete-time systems, see (Cortés et al., 2004b).

## 2. PRELIMINARY DEVELOPMENTS

### 2.1 LaSalle Invariance Principle for nondeterministic discrete-time systems

We review some concepts regarding the stability of discrete-time dynamical systems and set-valued maps following (Luenberger, 1984; Cortés et al., $2004 b$ ). For $d \in \mathbb{N}$, an algorithm on $\mathbb{R}^{d}$ is a setvalued map $T: \mathbb{R}^{d} \rightarrow 2^{\left(\mathbb{R}^{d}\right)}$ with the property that $T(p) \neq \emptyset$ for all $p \in \mathbb{R}^{d}$. A map from $\mathbb{R}^{d}$ to $\mathbb{R}^{d}$ can be interpreted as a singleton-valued map. A trajectory of an algorithm $T$ is a sequence $\left\{p_{m}\right\}_{m \in \mathbb{N} \cup\{0\}} \subset \mathbb{R}^{d}$ such that

$$
p_{m+1} \in T\left(p_{m}\right), \quad m \in \mathbb{N} \cup\{0\} .
$$

In other words, given any initial $p_{0} \in \mathbb{R}^{d}$, a trajectory of $T$ is computed by recursively setting $p_{m+1}$ equal to an arbitrary element in $T\left(p_{m}\right)$. An algorithm $T$ is closed at $p \in \mathbb{R}^{d}$ if for all pairs of convergent sequences $p_{k} \rightarrow p$ and $p_{k}^{\prime} \rightarrow p^{\prime}$ such that $p_{k}^{\prime} \in T\left(p_{k}\right)$, one has that $p^{\prime} \in T(p)$. An algorithm is closed on $W \subset \mathbb{R}^{d}$ if it is closed at $p$, for all $p \in W$. In particular, every continuous map $T: \mathbb{R}^{d} \rightarrow \mathbb{R}^{d}$ is closed on $\mathbb{R}^{d}$. A set $C$ is weakly positively invariant with respect to $T$ if, for any $p_{0} \in C$, there exists $p \in T\left(p_{0}\right)$ such that $p \in C$. A point $p_{0}$ is said to be a fixed point of $T$ if $p_{0} \in$ $T\left(p_{0}\right)$. The function $V: \mathbb{R}^{d} \rightarrow \mathbb{R}$ is non-increasing along $T$ on $W \subset \mathbb{R}^{d}$ if $V\left(p^{\prime}\right) \leq V(p)$ for all $p \in W$ and $p^{\prime} \in T(p)$. The proof of the following result is provided in (Cortés et al., 2004b).

Theorem 1. (LaSalle Invariance Principle for closed algorithms) Let $T$ be a closed algorithm on $W \subset$ $\mathbb{R}^{d}$ and let $V: \mathbb{R}^{d} \rightarrow \mathbb{R}$ be a continuous function non-increasing along $T$ on $W$. Assume the trajectory $\left\{p_{m}\right\}_{m \in \mathbb{N} \cup\{0\}}$ of $T$ takes values in $W$ and is bounded. Then there exists $c \in \mathbb{R}$ such that

$$
p_{m} \longrightarrow M \cap V^{-1}(c),
$$

where $M$ is the largest weakly positively invariant set contained in

$$
\left\{p \in \bar{W} \mid \exists p^{\prime} \in T(p) \text { such that } V\left(p^{\prime}\right)=V(p)\right\}
$$

### 2.2 Basic geometric notions

We review some notation for standard geometric objects; for additional information we refer the reader to (de Berg et al., 1997) and references therein. For a bounded set $S \subset \mathbb{R}^{d}, d \in \mathbb{N}$, we let $\operatorname{co}(S)$ denote the convex hull of $S$. For $p, q \in \mathbb{R}^{d}$, we let $] p, q[=\{\lambda p+(1-\lambda) q \mid \lambda \in$ $] 0,1[ \}$ and $[p, q]=\operatorname{co}(\{p, q\})$ denote the open and closed segment with extreme points $p$ and $q$, respectively. For a bounded set $S \subset \mathbb{R}^{d}$, we let $\mathrm{CC}(S)$ and $\mathrm{CR}(S)$ denote the circumcenter and circumradius of $S$, respectively, that is, the center and radius of the smallest-radius $d$-sphere
enclosing $S$. The computation of the circumcenter and circumradius of a bounded set is a strictly convex problem and in particular a quadratically constrained linear program. For $p \in \mathbb{R}^{d}$, we let $B(p, r)$ and $\bar{B}(p, r)$ denote the open and closed ball of radius $r \in \mathbb{R}_{+}$centered at $p$, respectively. Here, we let $\mathbb{R}_{+}$and $\overline{\mathbb{R}}_{+}$denote the positive and the nonnegative real numbers, respectively. A polytope is the convex hull of a finite point set. We let $\operatorname{Ve}(Q)$ denote the set of vertices of a polytope $Q$, and we emphasize that any vertex of $Q$ is strictly convex, i.e., $v \in \operatorname{Ve}(Q)$ if and only if there exists $u \in \mathbb{R}^{d}$ such that $(s-v) \cdot u>0$ for all $s \in Q \backslash\{v\}$.

Proposition 2. Let $S$ be a bounded set in $\mathbb{R}^{d}$. The following statements hold:
(i) $\mathrm{CC}(S) \in \operatorname{co}(S) \backslash \mathrm{Ve}(\operatorname{co}(S))$;
(ii) if $p \in S \backslash \mathrm{CC}(S)$ and $r \in \mathbb{R}_{+}$satisfy $S \subset \bar{B}(p, r)$, then $] p, \mathrm{CC}(S)[$ has nonempty intersection with $\bar{B}\left(\frac{p+q}{2}, \frac{r}{2}\right)$ for all $q \in S$.

### 2.3 Proximity graphs and their properties

We introduce some concepts regarding proximity graphs for point sets in $\mathbb{R}^{d}$. We assume the reader is familiar with the standard notions of graph theory as defined in (Diestel, 2000, Chapter 1). Given a vector space $\mathbb{V}$, let $\mathbb{F}(\mathbb{V})$ be the collection of finite subsets of $\mathbb{V}$. We shall denote an element of $\mathbb{F}\left(\mathbb{R}^{d}\right)$ by $\mathcal{P}=\left\{p_{1}, \ldots, p_{n}\right\} \subset \mathbb{R}^{d}$, where $p_{1}, \ldots, p_{n}$ are distinct points in $\mathbb{R}^{d}$. Let $\mathbb{G}\left(\mathbb{R}^{d}\right)$ be the set of undirected graphs whose vertex set is an element of $\mathbb{F}\left(\mathbb{R}^{d}\right)$. A proximity graph function $\mathcal{G}: \mathbb{F}\left(\mathbb{R}^{d}\right) \rightarrow \mathbb{G}\left(\mathbb{R}^{d}\right)$ associates to a point set $\mathcal{P}$ an undirected graph with vertex set $\mathcal{P}$ and edge set $\mathcal{E}_{\mathcal{G}}(\mathcal{P})$, where $\mathcal{E}_{\mathcal{G}}: \mathbb{F}\left(\mathbb{R}^{d}\right) \rightarrow \mathbb{F}\left(\mathbb{R}^{d} \times \mathbb{R}^{d}\right)$ has the property that $\mathcal{E}_{\mathcal{G}}(\mathcal{P}) \subseteq \mathcal{P} \times \mathcal{P} \backslash \operatorname{diag}(\mathcal{P} \times \mathcal{P})$ for any $\mathcal{P}$. Here, $\operatorname{diag}(\mathcal{P} \times \mathcal{P})=\{(p, p) \in \mathcal{P} \times \mathcal{P} \mid p \in$ $\mathcal{P}\}$. In other words, the edge set of a proximity graph depends on the location of its vertices. General properties of proximity graphs and the following examples are defined in (de Berg et al., 1997; Jaromczyk and Toussaint, 1992; Cortés et al., 2004b):
(i) the $r$-disk graph $\mathcal{G}_{\text {disk }}(r)$, for $r \in \mathbb{R}_{+}$, with $\left(p_{i}, p_{j}\right) \in \mathcal{E}_{\mathcal{G}_{\text {disk }}(r)}(\mathcal{P})$ if $\left\|p_{i}-p_{j}\right\| \leq r ;$
(ii) the Delaunay graph $\mathcal{G}_{\mathrm{D}}$, with $\left(p_{i}, p_{j}\right) \in$ $\mathcal{E}_{\mathcal{G}_{\mathrm{D}}}(\mathcal{P})$ if the Voronoi regions of $p_{i}$ and $p_{j}$ have non-empty intersection;
(iii) the Relative Neighborhood graph $\mathcal{G}_{\text {RN }}$, with $\left(p_{i}, p_{j}\right) \in \mathcal{E}_{\mathcal{G}_{\mathrm{RN}}}(\mathcal{P})$ if, for all $p_{k} \in \mathcal{P} \backslash\left\{p_{i}, p_{j}\right\}$, $p_{k} \notin B\left(p_{i},\left\|p_{i}-p_{j}\right\|\right) \cap B\left(p_{j},\left\|p_{i}-p_{j}\right\|\right)$;
(iv) the Gabriel graph $\mathcal{G}_{\mathrm{G}}$, with $\left(p_{i}, p_{j}\right) \in \mathcal{E}_{\mathcal{G}_{\mathrm{G}}}(\mathcal{P})$ if, for all $p_{k} \in \mathcal{P} \backslash\left\{p_{i}, p_{j}\right\}, p_{k} \notin B\left(\frac{p_{i}+p_{j}}{2}, \frac{\left\|p_{i}-p_{j}\right\|}{2}\right)$;
(v) the Euclidean Minimum Spanning Tree $\mathcal{G}_{\text {EMST }}$, which for each $\mathcal{P}$, is a minimum-weight spanning tree of the complete graph $(\mathcal{P}, \mathcal{P} \times \mathcal{P} \backslash$ $\operatorname{diag}(\mathcal{P} \times \mathcal{P}))$ whose edge $\left(p_{i}, p_{j}\right)$ has weight $\left\|p_{i}-p_{j}\right\|$.

If needed, we write $\mathcal{G}_{\text {disk }}(\mathcal{P}, r)$ to denote $\mathcal{G}_{\text {disk }}(r)$ at $\mathcal{P}$. In what follows, we will consider the proximity graphs $\mathcal{G}_{\text {RN } \cap \operatorname{disk}}(r)$ and $\mathcal{G}_{\mathrm{G} \cap \text { disk }}(r)$ defined by the intersection of $\mathcal{G}_{\mathrm{RN}}$ and $\mathcal{G}_{\mathrm{G}}$ with $\mathcal{G}_{\text {disk }}(r)$, $r \in \mathbb{R}_{+}$, respectively. A different proximity graph related to, but different from, the intersection $\mathcal{G}_{\mathrm{D}} \cap$ disk $(r)$ of $\mathcal{G}_{\mathrm{D}}$ with $\mathcal{G}_{\text {disk }}(r)$ is the $r$-limited Delaunay graph $\mathcal{G}_{\mathrm{LD}}(r)$ (see (Cortés et al., 2004b)).

To each proximity graph function $\mathcal{G}$, one can associate the set of neighbors map $\mathcal{N}_{\mathcal{G}}: \mathbb{R}^{d} \times$ $\mathbb{F}\left(\mathbb{R}^{d}\right) \rightarrow \mathbb{F}\left(\mathbb{R}^{d}\right)$, defined by

$$
\mathcal{N}_{\mathcal{G}}(p, \mathcal{P})=\left\{q \in \mathcal{P} \mid(p, q) \in \mathcal{E}_{\mathcal{G}}(\mathcal{P} \cup\{p\})\right\}
$$

Typically, $p$ is a point in $\mathcal{P}$, but the definition is well-posed for any $p \in \mathbb{R}^{d}$. Given $p \in \mathbb{R}^{d}$, it is convenient to define the map $\mathcal{N}_{\mathcal{G}, p}: \mathbb{F}\left(\mathbb{R}^{d}\right) \rightarrow$ $\mathbb{F}\left(\mathbb{R}^{d}\right)$ by $\mathcal{N}_{\mathcal{G}, p}(\mathcal{P})=\mathcal{N}_{\mathcal{G}}(p, \mathcal{P})$. Let $\mathcal{G}_{1}$ and $\mathcal{G}_{2}$ be two proximity graph functions. We say that $\mathcal{G}_{1}$ is spatially distributed over $\mathcal{G}_{2}$ if, for all $p \in \mathcal{P}$,

$$
\mathcal{N}_{\mathcal{G}_{1}, p}(\mathcal{P})=\mathcal{N}_{\mathcal{G}_{1}, p}\left(\mathcal{N}_{\mathcal{G}_{2}, p}(\mathcal{P})\right) .
$$

It is clear that if $\mathcal{G}_{1}$ is spatially distributed over $\mathcal{G}_{2}$, then $\mathcal{G}_{1}$ is a subgraph of $\mathcal{G}_{2}$, that is, $\mathcal{G}_{1}(\mathcal{P}) \subset$ $\mathcal{G}_{2}(\mathcal{P})$ for all $\mathcal{P} \in \mathbb{F}\left(\mathbb{R}^{d}\right)$. The converse is in general not true (for instance, the graph $\mathcal{G}_{\mathrm{D}}$ กdisk is a subgraph of $\mathcal{G}_{\text {disk }}$, but it is not spatially distributed over it, see (Cortés et al., 2004b)). Finally, we say that two proximity graph functions $\mathcal{G}_{1}$ and $\mathcal{G}_{2}$ have the same connected components if, for all point sets $\mathcal{P}$, the graphs $\mathcal{G}_{1}(\mathcal{P})$ and $\mathcal{G}_{2}(\mathcal{P})$ have the same number of connected components consisting of the same vertices.

We conclude this section with some examples of proximity graphs in $\mathbb{R}^{2}$ and $\mathbb{R}^{3}$; see Figs 1 and 2.


Fig. 1. From left to right, $r$-disk, $r$-limited Delaunay, and Euclidean Minimum Spanning Tree graphs in $\mathbb{R}^{2}$ for a configuration of 25 agents with coordinates uniformly randomly generated within the square $[-7,7] \times$ $[-7,7]$. The parameter $r$ is taken equal to 4 .


Fig. 2. From left to right, $r$-disk, Gabriel, and Relative Neighborhood graphs in $\mathbb{R}^{3}$ for a configuration of 25 agents with coordinates uniformly randomly generated within the square $[-7,7] \times[-7,7] \times[-7,7]$. The parameter $r$ is taken equal to 4 .

### 2.4 Proximity graphs over point tuples and spatially distributed maps

The notion of proximity graph is defined for sets of distinct points $\mathcal{P}=\left\{p_{1}, \ldots, p_{n}\right\}$. However, we will often consider tuples of elements of $\mathbb{R}^{d}$ of the form $P=\left(p_{1}, \ldots, p_{n}\right)$, i.e., ordered sets of possibly coincident points. Let $i_{\mathbb{F}}:\left(\mathbb{R}^{d}\right)^{n} \rightarrow$ $\mathbb{F}\left(\mathbb{R}^{d}\right)$ be the natural immersion, i.e., $i_{\mathbb{F}}(P)$ is the point set that contains only the distinct points in $P=\left(p_{1}, \ldots, p_{n}\right)$. Note that $i_{\mathbb{F}}$ is invariant under permutations of its arguments and that the cardinality of $i_{\mathbb{F}}\left(p_{1}, \ldots, p_{n}\right)$ is in general less than or equal to $n$. In what follows, $\mathcal{P}=i_{\mathbb{F}}(P)$ always denotes the point set associated to $P \in\left(\mathbb{R}^{d}\right)^{n}$.
We can now extend the notion of proximity graphs to this setting. Given a proximity graph function $\mathcal{G}$ with edge set function $\mathcal{E}_{\mathcal{G}}$, we define (with a slight abuse of notation)

$$
\begin{aligned}
& \mathcal{G}=\mathcal{G} \circ i_{\mathbb{F}}:\left(\mathbb{R}^{d}\right)^{n} \rightarrow \mathbb{G}\left(\mathbb{R}^{d}\right), \\
& \mathcal{E}_{\mathcal{G}}=\mathcal{E}_{\mathcal{G}} \circ i_{\mathbb{F}}:\left(\mathbb{R}^{d}\right)^{n} \rightarrow \mathbb{F}\left(\mathbb{R}^{d} \times \mathbb{R}^{d}\right) .
\end{aligned}
$$

We define the set of neighbors map $\mathcal{N}_{\mathcal{G}}:\left(\mathbb{R}^{d}\right)^{n} \rightarrow$ $\left(\mathbb{F}\left(\mathbb{R}^{d}\right)\right)^{n}$ as the function whose $j$ th component is

$$
\mathcal{N}_{\mathcal{G}, j}\left(p_{1}, \ldots, p_{n}\right)=\mathcal{N}_{\mathcal{G}}\left(p_{j}, i_{\mathbb{F}}\left(p_{1}, \ldots, p_{n}\right)\right) .
$$

Note that coincident points in the tuple $\left(p_{1}, \ldots, p_{n}\right)$ will have the same set of neighbors.

Given a set $Y$ and a proximity graph function $\mathcal{G}$, a map $T:\left(\mathbb{R}^{d}\right)^{n} \rightarrow Y^{n}$ is spatially distributed over $\mathcal{G}$ if there exists $\tilde{T}: \mathbb{R}^{d} \times \mathbb{F}\left(\mathbb{R}^{d}\right) \rightarrow Y$, such that, for all $\left(p_{1}, \ldots, p_{n}\right) \in\left(\mathbb{R}^{d}\right)^{n}$ and for all $j \in\{1, \ldots, n\}$,

$$
T_{j}\left(p_{1}, \ldots, p_{n}\right)=\tilde{T}\left(p_{j}, \mathcal{N}_{\mathcal{G}, j}\left(p_{1}, \ldots, p_{n}\right)\right),
$$

where $T_{j}$ denotes the $j$ th-component of $T$. In other words, the $j$ th component of a spatially distributed map at $\left(p_{1}, \ldots, p_{n}\right)$ can be computed with only the knowledge of the vertex $p_{j}$ and the neighboring vertices in the graph $\mathcal{G}\left(\left\{p_{1}, \ldots, p_{n}\right\}\right)$.

## 3. RENDEZVOUS VIA PROXIMITY GRAPHS

In this section we state the model, the control objective, the motion coordination algorithm, and the properties of the resulting closed-loop system.

### 3.1 Modeling a network of robotic agents

We introduce the notions of robotic agent and of network of robotic agents. Let $n$ be the number of agents in the network. The $i$ th agent has a processor with the ability of allocating and operating on continuous and discrete states. The $i$ th agent occupies a location $p_{i} \in \mathbb{R}^{d}, d \in \mathbb{N}$, and it is capable of moving at any time $m \in \mathbb{N}$, for any unit period of time, according to

$$
\begin{equation*}
p_{i}(m+1)=p_{i}(m)+u_{i} . \tag{1}
\end{equation*}
$$

The control $u_{i}$ takes values in a bounded subset of $\mathbb{R}^{d}$. We assume that there is a maximum step size $s_{\mathrm{m}} \in \mathbb{R}_{+}$common to all agents, i.e., $\left\|u_{i}\right\| \leq s_{\mathrm{m}}$, for all $i \in\{1, \ldots, n\}$. The sensing and communication model is the following. The processor of each agent has access to its location, and transmits this information to any other agent within a closed disk of radius $r \in \mathbb{R}_{+}$. The communication radius is the same for all agents.

### 3.2 The rendezvous motion coordination problem

We now state the control design problem for the network of robotic agents. The rendezvous objective is to achieve agreement over the location of the agents in the network, that is, to steer each agent to a common location. This objective is to be achieved with the limited information flow described in the model above. Typically, it will be impossible to solve the rendezvous problem if the agents are placed in such a way that they do not form a connected communication graph. Arguably, a good property of any algorithm for rendezvous is that of maintaining some form of connectivity between agents.

### 3.3 The Circumcenter Algorithm

Here is an informal description of the Circumcenter Algorithm over a proximity graph $\mathcal{G}$ :

Each agent performs the following tasks: (i) it detects its neighbors according to $\mathcal{G}$; (ii) it computes the circumcenter of the point set comprised of its neighbors and of itself, and
(iii) it moves toward this circumcenter while maintaining connectivity with its neighbors.
This algorithm is an extension of the one introduced in (Ando et al., 1999). Let us clarify two which proximity graphs are allowable and how connectivity is maintained. First, we are allowed to design motion coordination algorithms that are spatially distributed over the $r$-disk graph $\mathcal{G}_{\text {disk }}(r)$, or more generally, over any proximity graph $\mathcal{G}$ that is spatially distributed over $\mathcal{G}_{\text {disk }}(r)$. This is a consequence of our modeling assumption that each agent can acquire the location of each other agent within distance less than or equal to $r$. Second, we maintain connectivity by restricting the allowable motion of each agent. In particular, it suffices to restrict the motion of each agent as follows. If agents $p_{i}$ and $p_{j}$ are neighbors in $\mathcal{G}$, then their subsequent positions are required to belong to $\bar{B}\left(\frac{p_{i}+p_{j}}{2}, \frac{r}{2}\right)$. If agent $p_{i}$ has its neighbors at locations $\left\{q_{1}, \ldots, q_{l}\right\}$, then its constraint set is

$$
C_{p_{i}, r}\left(\left\{q_{1}, \ldots, q_{l}\right\}\right)=\bigcap_{q \in\left\{q_{1}, \ldots, q_{l}\right\}} \bar{B}\left(\frac{p_{i}+q}{2}, \frac{r}{2}\right) .
$$

Finally, for $q_{0}$ and $q_{1}$ in $\mathbb{R}^{d}$, and for a convex closed set $Q \subset \mathbb{R}^{d}$ with $q_{0} \in Q$, let $\lambda\left(q_{0}, q_{1}, Q\right)$ denote the solution of the strictly convex problem:

$$
\begin{align*}
& \operatorname{maximize} \lambda \\
& \text { subject to } \lambda \leq 1,(1-\lambda) q_{0}+\lambda q_{1} \in Q \tag{2}
\end{align*}
$$

This convex optimization problem has the following interpretation: move along the segment from $q_{0}$ to $q_{1}$ the maximum possible distance while remaining in $Q$. Under the stated assumptions the solution exists and is unique. We are now ready to formally describe the algorithm.

| Name: | Circumcenter Algorithm over $\mathcal{G}$ |
| :--- | :--- |
| Goal: | Solve the rendezvous problem |

Assumes: (i) $s_{\mathrm{m}} \in \mathbb{R}_{+}$maximum step size
(ii) $r \in \mathbb{R}_{+}$communication radius (iii) $\mathcal{G}$ spatially distributed proximity graph over $\mathcal{G}_{\text {disk }}(r)$

Agent $i \in\{1, \ldots, n\}$ executes at each time instant in $\mathbb{N}$ :

```
1: acquire \(\left\{q_{1}, \ldots, q_{k}\right\}:=\mathcal{N}_{\mathcal{G}_{\text {disk }}(r), p_{i}}(\mathcal{P})\)
: compute \(\mathcal{M}_{i}:=\mathcal{N}_{\mathcal{G}, p_{i}}\left(\left\{q_{1}, \ldots, q_{k}\right\}\right) \cup\left\{p_{i}\right\}\)
: compute \(Q_{i}:=C_{p_{i}, r}\left(\mathcal{M}_{i} \backslash\left\{p_{i}\right\}\right) \cap \bar{B}\left(p_{i}, s_{\mathrm{m}}\right)\)
compute \(\lambda_{i}^{*}:=\lambda\left(p_{i}, \mathrm{CC}\left(\mathcal{M}_{i}\right), Q_{i}\right)\)
set \(u_{i}:=\lambda_{i}^{*}\left(\operatorname{CC}\left(\mathcal{M}_{i}\right)-p_{i}\right)\), i.e.,
    move from \(p_{i}\) to \(\left(1-\lambda_{i}^{*}\right) p_{i}+\lambda_{i}^{*} \mathrm{CC}\left(\mathcal{M}_{i}\right)\)
```

In what follows we refer to the Circumcenter Algorithm over $\mathcal{G}$ as the map $T_{\mathcal{G}}:\left(\mathbb{R}^{d}\right)^{n} \rightarrow\left(\mathbb{R}^{d}\right)^{n}$.

### 3.4 Correctness of the Circumcenter Algorithm

We now state the main convergence result, whose proof is provided in (Cortés et al., 2004a).

Theorem 3. Let $p_{1}, \ldots, p_{n}$ be a network of robotic agents in $\mathbb{R}^{d}$, for $d \in \mathbb{N}$, with maximum step size $s_{\mathrm{m}} \in \mathbb{R}_{+}$and communication radius $r \in \mathbb{R}_{+}$. Let the proximity graph $\mathcal{G}$ be spatially distributed over $\mathcal{G}_{\text {disk }}(r)$ and have the same connected components as $\mathcal{G}_{\text {disk }}(r)$. Any trajectory $\left\{P_{m}\right\}_{m \in \mathbb{N} \cup\{0\}}$ of $T_{\mathcal{G}}$ has the following properties:
(i) if the locations of two agents belong to the same connected component of $\mathcal{G}_{\text {disk }}\left(P_{k}, r\right)$ for some $k \in \mathbb{N} \cup\{0\}$, then they remain in the same connected component of $\mathcal{G}_{\text {disk }}\left(P_{m}, r\right)$ for all $m \geq k$;
(ii) there exists $P^{*}=\left(p_{1}^{*}, \ldots, p_{n}^{*}\right) \in\left(\mathbb{R}^{d}\right)^{n}$ with the following properties: $P_{m} \rightarrow P^{*}$ as $m \rightarrow$ $+\infty$, and $p_{i}^{*}=p_{j}^{*}$ or $\left\|p_{i}^{*}-p_{j}^{*}\right\|>r$ for each $i, j \in\{1, \ldots, n\}$;
(iii) if $\mathcal{G}=\mathcal{G}_{\text {disk }}(r)$, then there exists $k \in \mathbb{N}$ such that $P_{m}=P^{*}$ for all $m \geq k$, that is, convergence is achieved in finite time.

A consequence of Theorem 3(i) and (ii) is that, if the locations of two agents belong to the same connected component of $\mathcal{G}$ at some time, then they converge to the same point in $\mathbb{R}^{d}$. The statements Theorem 3(i) and (ii) were originally proved in (Ando et al., 1999) for the Circumcenter Algorithm over $\mathcal{G}_{\text {disk }}$ and for $d=2$. This result was extended to other control policies by (Lin et al., 2003; Lin et al., 2004a) (still on the plane and with $\mathcal{G}_{\text {disk }}$ communication topology).

### 3.5 Robustness of the Circumcenter Algorithm

Here we characterize the robustness of the Circumcenter Algorithm with respect to link failures. We provide no physical model to motivate the occurrence for link failures; rather we analyze the resulting closed-loop network.

Definition 4. A link failure in $\mathcal{G}_{\text {disk }}(r)$ at $P \in$ $\left(\mathbb{R}^{d}\right)^{n}$ is said to occur at agent $p_{i}$ if $\left(p_{i}, p_{j}\right)$ is an edge in $\mathcal{G}_{\text {disk }}(P, r)$ and the agent $p_{i}$ does not detect agent $p_{j}$. For $\mathcal{P}=i_{\mathbb{F}}(P)$, we denote this link failure by the directed edge $\left(p_{i}, p_{j}\right) \in \mathcal{P} \times \mathcal{P}$.

Remark 5. Consider an application of the Circumcenter Algorithm over a proximity graph $\mathcal{G}$ as described in the steps 1-5 above. If the link failure $\left(p_{i}, p_{j}\right)$ takes place at step 1 , then the following two events will ensue:
(i) if $p_{j}$ is a neighbor of $p_{i}$ according to $\mathcal{G}$, then $p_{i}$ looses the neighbor $p_{j}$ at step 2,
(ii) if $p_{k}$ is not a neighbor of $p_{i}$ according to $\mathcal{G}$ because of the presence of $p_{j}$, then $p_{i}$ gains the neighbor $p_{k}$ at step 2.

After steps 1 and 2, the collection of neighbors has been computed inaccurately. Nevertheless the execution of steps 3 through 5 can continue.

Definition 6. For $P \in\left(\mathbb{R}^{d}\right)^{n}$, let $\mathcal{P}=i_{\mathbb{F}}(P)$. Let $\mathcal{G}$ be a proximity graph spatially distributed over $\mathcal{G}_{\text {disk }}(r)$ and $F \subset \mathcal{P} \times \mathcal{P}$ be a set of link failures. Let
(i) $\mathcal{G}_{\text {disk }}(\mathcal{P}, r) \nleftarrow F$ be the directed graph with vertex set $\mathcal{P}$ and with edge set $\mathcal{E}_{\text {disk }}(\mathcal{P}, r) \backslash F$;
(ii) $\mathcal{G}(\mathcal{P}) \nleftarrow F$ be the directed graph with vertex set $\mathcal{P}$ and with edges determined as follows; the neighbors of $p \in \mathcal{P}$ are

$$
\mathcal{N}_{\mathcal{G}, p}\left(\left\{q \mid(p, q) \in \mathcal{E}_{\text {disk }}(\mathcal{P}, r) \backslash F\right\}\right)
$$

that is, the edges of $\mathcal{G}(\mathcal{P}) \nleftarrow F$ arise from the computation of $\mathcal{G}(\mathcal{P})$ with the link failures $F$, as described in Remark 5;
(iii) $T_{\mathcal{G} \nleftarrow F}(P)$ is the configuration obtained from applying the Circumcenter Algorithm over $\mathcal{G}$ (steps 1-5) at configuration $P$ with the link failures $F$ at step 1.

Note that only a finite number of possible link failures can occur at any configuration. Consequently,
the set of possible directed graphs arising from link failures is finite. We are now ready to state the main robust convergence result, whose proof is provided in (Cortés et al., 2004a).

Theorem 7. Let the network $p_{1}, \ldots, p_{n}$ and the proximity graph $\mathcal{G}$ have the same properties as in Theorem 3. Given $P_{0} \in\left(\mathbb{R}^{d}\right)^{n}$, consider the two sequences $\left\{P_{m}\right\}_{m \in \mathbb{N} \cup\{0\}}$ and $\left\{F_{m}\right\}_{m \in \mathbb{N} \cup\{0\}}$ defined recursively by
(i) $F_{m}$ is a set of link failures in $\mathcal{G}_{\text {disk }}(r)$ at $P_{m}$, (ii) $P_{m+1}=T_{\mathcal{G} \nleftarrow F_{m}}\left(P_{m}\right)$.

If there is $\ell \in \mathbb{N}$ such that at least one graph of any $\ell$ consecutive elements of $\left\{\mathcal{G}\left(P_{m}\right) \nleftarrow F_{m}\right\}_{m \in \mathbb{N} \cup\{0\}}$ is strongly connected, then there exists $p^{*} \in \mathbb{R}^{d}$ such that $P_{m} \rightarrow\left(p^{*}, \ldots, p^{*}\right)$ as $m \rightarrow+\infty$.

This theorem provides the first theoretical explanation for the robustness behavior against sensor and control errors of the Circumcenter Algorithm over $\mathcal{G}_{\text {disk }}(r)$ observed in (Ando et al., 1999).

Corollary 8. With the same notation as in Theorem 7 , if at each step $m \in \mathbb{N}$, the proximity graph $\mathcal{G}\left(P_{m}\right)$ is $k_{m}$-connected and if $F_{m}$ contains at most $k_{m}-1$ link failures, then there exists $p^{*} \in \mathbb{R}^{d}$ such that $P_{m} \rightarrow\left(p^{*}, \ldots, p^{*}\right)$ as $m \rightarrow+\infty$.

Next, we analyze the performance of the Circumcenter Algorithm when each agent of the mobile network at each time step is allowed to use a different proximity graph to compute its neighbors.

Definition 9. Let $\mathcal{S}$ be a set of proximity graph functions that are spatially distributed over $\mathcal{G}_{\text {disk }}(r)$. The Circumcenter Algorithm over $\mathcal{S}$ is the Circumcenter Algorithm where step 2 is replaced by

$$
\begin{aligned}
& 2(a): \text { choose any } \mathcal{G} \in \mathcal{S} \\
& 2(b): \text { compute } \mathcal{M}_{i}:=\mathcal{N}_{\mathcal{G}, p_{i}}\left(\left\{q_{1}, \ldots, q_{k}\right\}\right) \cup\left\{p_{i}\right\} .
\end{aligned}
$$

The selection algorithm for each agent at each execution of step $2(a)$ is left unspecified.

Corollary 10. Let the network $p_{1}, \ldots, p_{n}$ be as in Theorem 3. Let $\mathcal{S}$ be a set of proximity graph functions that are spatially distributed over $\mathcal{G}_{\text {disk }}(r)$. Assume there exists a proximity graph $\mathcal{F}$ with the same connected components as $\mathcal{G}_{\text {disk }}(r)$ such that $\mathcal{F} \subset \mathcal{G}$, for all $\mathcal{G} \in \mathcal{S}$. Then any trajectory $\left\{P_{m}\right\}_{m \in \mathbb{N} \cup\{0\}}$ of the Circumcenter Algorithm over $\mathcal{S}$ has properties (i) and (ii) in Theorem 3.

For $r \in \mathbb{R}_{+}, \mathcal{G}_{\mathrm{RN} \cap \operatorname{disk}}(r), \mathcal{G}_{\mathrm{G} \cap \operatorname{disk}}(r)$ and $\mathcal{G}_{\mathrm{LD}}(r)$ are spatially distributed over $\mathcal{G}_{\text {disk }}(r)$ and contain $\mathcal{G}_{\text {EMST }}$ @disk $(r)$, which has the same connected components as $\mathcal{G}_{\text {disk }}(r)$ (cf. (Cortés et al., 2004a)). As a consequence, any subset of
$\left\{\mathcal{G}_{\mathrm{RN} \cap \operatorname{disk}}(r), \mathcal{G}_{\mathrm{G}} \cap \operatorname{disk}(r), \mathcal{G}_{\mathrm{LD}}(r)\right\}$ satisfies the hypothesis of Corollary 10.

## 4. SIMULATIONS

In order to illustrate the performance of our rendezvous algorithms, we developed a library of basic geometric routines. The resulting Mathematica ${ }^{\circledR}$ packages PlanGeom.m (containing the 2-dimensional routines) and SpatialGeom.m (containing the 3 -dimensional routines) are freely available at http://motion.csl.uiuc.edu.

The simulation run for the Circumcenter Algorithm in the plane, $d=2$, over $\mathcal{G}_{\text {LD }}(r)$ with link failures is illustrated in Figure 3. The 25 vehicles have a maximum step size $s_{\mathrm{m}}=.15$, and a communication radius $r=4$. At each time step, a


Fig. 3. Evolution (in light gray) of the Circumcenter Algorithm over the $r$-limited Delaunay graph $\mathcal{G}_{\mathrm{LD}}(r)$ with link failures. The initial configuration of the network is as in Figure 1.
set consisting of 18 numbers between 1 and 25 is randomly selected, corresponding to the identities of the agents where link failures occur. For each of them, a randomly selected link failure in $\mathcal{G}_{\text {disk }}(r)$ is chosen. Since the identity of an agent might appear more than once in the random set, more than one link failure may occur at the same agent. However, rendezvous is asymptotically achieved according to Theorem 7 (usually after 80 steps).
The simulation run for the Circumcenter Algorithm in space, $d=3$, over the set of proximity graphs $\left\{\mathcal{G}_{\text {disk }}(r), \mathcal{G}_{\mathrm{G}}(r) \cap \mathcal{G}_{\text {disk }}(r), \mathcal{G}_{\mathrm{RN}}(r) \cap\right.$ $\left.\mathcal{G}_{\text {disk }}(r)\right\}$ is illustrated in Figure 4. The 25 vehicles have, as before, a maximum step size $s_{\mathrm{m}}=.15$, and a communication radius $r=4$. At each time step, each agent randomly selects one of the proximity graphs in $\left\{\mathcal{G}_{\text {disk }}(r), \mathcal{G}_{\text {RN } \cap \text { disk }}(r), \mathcal{G}_{\text {G } \cap \text { disk }}(r)\right\}$ and computes its corresponding set of neighbors according to it. Then, it executes steps 3 through 5 of the Circumcenter Algorithm. Rendezvous is


Fig. 4. Evolution (in light gray) of the Circumcenter Algorithm over $\left\{\mathcal{G}_{\text {disk }}(r), \mathcal{G}_{G}(r) \cap \mathcal{G}_{\text {disk }}(r), \mathcal{G}_{\text {RN }}(r) \cap \mathcal{G}_{\text {disk }}(r)\right\}$. The initial configuration of the network is as in Figure 2. The right figure is a rotated view of the left figure by 45 degrees.
achieved in a finite number of steps (usually after 100 steps).

## 5. CONCLUSIONS

We have designed and analyzed a class of circumcenter algorithms over proximity graphs for multi-agent rendezvous. Also, we have provided a set of novel tools that we believe are important in the design and analysis of general motion coordination algorithms. Future directions of research in motion coordination include the study of increasingly realistic communication settings (asynchronicity (Lin et al., 2004a), quantization, media access and power control issues), the analysis of the performance and complexity of the algorithms, and the formal design of other spatially distributed coordination primitives.

## REFERENCES

Ando, H., Y. Oasa, I. Suzuki and M. Yamashita (1999). Distributed memoryless point convergence algorithm for mobile robots with limited visibility. IEEE Transactions on Robotics and Automation 15(5), 818-828.
Cortés, J., S. Martínez and F. Bullo (2004a). Robust rendezvous for mobile autonomous agents via proximity graphs in $d$ dimensions. IEEE Transactions on $A u-$ tomatic Control. Submitted. Electronic version available at http://www.me.ucsb.edu/bullo.
Cortés, J., S. Martínez and F. Bullo (2004b). Spatiallydistributed coverage optimization and control with limited-range interactions. ESAIM. Control, Optimisation 8 Calculus of Variations. Submitted. Electronic version available at http://www.me.ucsb.edu/ bullo.
de Berg, M., M. van Kreveld and M. Overmars (1997). Computational Geometry: Algorithms and Applications. Springer Verlag. New York.
Diestel, R. (2000). Graph Theory. Vol. 173 of Graduate Texts in Mathematics. second ed.. Springer Verlag. New York.
Jadbabaie, A., J. Lin and A. S. Morse (2003). Coordination of groups of mobile autonomous agents using nearest
neighbor rules. IEEE Transactions on Automatic Control 48(6), 988-1001.
Jaromczyk, J. W. and G. T. Toussaint (1992). Relative neighborhood graphs and their relatives. Proceedings of the IEEE 80(9), 1502-1517.
Justh, E. W. and P. S. Krishnaprasad (2004). Equilibria and steering laws for planar formations. Systems $\mathcal{B}$ Control Letters 52(1), 25-38.
Klavins, E., R. Ghrist and D. Lipsky (2004). A grammatical approach to self-organizing robotic systems. International Journal of Robotics Research. Submitted.
Li, X.-Y. (2003). Algorithmic, geometric and graphs issues in wireless networks. Wireless Communications and Mobile Computing 3(2), 119-140.
Lin, J., A. S. Morse and B. D. O. Anderson (2003). The multi-agent rendezvous problem. In: IEEE Conf. on Decision and Control. Maui, Hawaii. pp. 1508-1513.
Lin, J., A. S. Morse and B. D. O. Anderson (2004a). The multi-agent rendezvous problem: an extended summary. In: Proceedings of the 2003 Block Island Workshop on Cooperative Control (N. E. Leonard, S. Morse and V. Kumar, Eds.). Lecture Notes in Control and Information Sciences. Springer Verlag. New York. To appear.
Lin, Z., M. Broucke and B. Francis (2004b). Local control strategies for groups of mobile autonomous agents. IEEE Transactions on Automatic Control 49(4), 622-629.
Liu, Y. and K. M. Passino (2004). Stable social foraging swarms in a noisy environment. IEEE Transactions on Automatic Control 49(1), 30-44.
Luenberger, D. G. (1984). Linear and Nonlinear Programming. second ed.. Addison-Wesley. Reading, MA.
Moreau, L. (2003). Time-dependent unidirectional communication in multi-agent systems. Preprint. Electronically available at http://xxx.arxiv.org/math.OC/ 0306426.

Ögren, P., E. Fiorelli and N. E. Leonard (2004). Cooperative control of mobile sensor networks: adaptive gradient climbing in a distributed environment. IEEE Transactions on Automatic Control 49(8), 12921302.

Olfati-Saber, R. and R. M. Murray (2004). Consensus problems in networks of agents with switching topology and time-delays. IEEE Transactions on Automatic Control 49(9), 1520-1533.
Suzuki, I. and M. Yamashita (1999). Distributed anonymous mobile robots: Formation of geometric patterns. SIAM Journal on Computing 28(4), 1347-1363.

