Safe Cooperative Robot Dynamics on Graphs*  

Robert W. Ghrist†  Daniel E. Koditschek‡  
February 1, 2008  

Abstract  
This paper initiates the use of vector fields to design, optimize, and implement reactive schedules for safe cooperative robot patterns on planar graphs. We consider Automated Guided Vehicles (AGV’s) operating upon a predefined network of pathways. In contrast to the case of locally Euclidean configuration spaces, regularization of collisions is no longer a local procedure, and issues concerning the global topology of configuration spaces must be addressed. The focus of the present inquiry is the achievement of safe, efficient, cooperative patterns in the simplest nontrivial example (a pair of robots on a Y-network) by means of a state-event hierarchical controller.

1 Introduction  
Recent literature suggests the growing awareness of a need for “reactive” scheduling wherein one desires not merely a single deployment of resources but a plan for successive re-deployments against a changing environment [19]. But scheduling problems have been traditionally solved by appeal to a discrete representation of the domain at hand. Thus the need for “tracking” changing goals introduces a conceptual dilemma: there is no obvious topology by which proximity to the target of a given deployment can be measured. In contrast to problems entailing the management of information alone, problems in many robotics and automation settings involve the management of work — the exchange of energy in the presence of geometric constraints. In these settings, it may be desirable to postpone the imposition of a discrete representation long enough to gain the benefit of the natural topology that accompanies the original domain.

This paper explores the use of vector fields for reactive scheduling of safe cooperative robot patterns on graphs. The word “safe” means that obstacles — designated illegal portions of the configuration space — are avoided. The word “cooperative” connotes situations wherein physically distributed agents are collectively responsible for executing the schedule. The word “pattern” refers to tasks that cannot be encoded simply in terms of a point goal in the configuration space. The word “reactive” will be interpreted as requiring that the desired pattern reject perturbations: conditions close but slightly removed from those desired remain close and, indeed, converge toward the exactly desired pattern.

1.1 Setting: AGV’s on a Guidepath Network of Wires  
An automated guided vehicle (AGV) is an unmanned powered cart “capable of following an external guidance signal to deliver a unit load from destination to destination” where, in most common applications, the guidepath signal is buried in the floor [5]. Thus, the AGV’s workspace is a network of wires — a graph. The motivation to choose AGV based materials handling systems over more conventional fixed conveyors rests not simply in their ease of reconfigurability but in the potential they offer for graceful response to perturbations in normal plant operation. In real production facilities, the flow of work in process fluctuates constantly in the

*Supported in part by National Science Foundation Grant IRI-9510673 [DK] and by National Science Foundation Grant DMS-9971629 [RG]. A sketch of these ideas appeared as a conference proceedings [11].  
†School of Mathematics, Georgia Institute of Technology, Atlanta, GA 30017, USA. ghrist@math.gatech.edu  
‡Department of Electrical Engineering and Computer Science, The University of Michigan, Ann Arbor, MI 48109-2110, USA. kod@eecs.umich.edu
face of unanticipated workstation downtime, variations in process rate, and, indeed, variations in materials
transport and delivery rates [7]. Of course, realizing their potential robustness against these fluctuations in
work flow remains an only partially fulfilled goal of contemporary AGV systems.

Choreographing the interacting routes of multiple AGVs in a non-conflicting manner presents a novel,
complicated, and necessarily on-line planning problem. Nominal routes might be designed offline but they
can never truly be traversed with the nominal timing, for all the reasons described above. Even under normal
operating conditions, no single nominal schedule can suffice to coordinate the workflow as the production
volume or product mix changes over time: new vehicles need to be added or deleted and the routing scheme
adapted. In any case, abnormal conditions — unscheduled process down times; blocked work stations; failed
vehicles — continually arise, demanding altered routes.

The traffic control schemes deployed in contemporary AGV systems are designed to simplify the real-time
route planning and adaptation process by “blocking zone control” strategies. The workspace is partitioned
into a small number of cells and, regardless of the details of their source and destination tasks, no two AGVs
are ever allowed into the same cell at the same time [5]. Clearly, this simplification results in significant loss
of a network’s traffic capacity.

In this paper, we will consider a centralized approach that employs dynamical systems theory to focus
on real-time responsiveness and efficiency as opposed to computational complexity or average throughput.
No doubt, beyond a certain maximum number of vehicles, the necessity to compute in the high dimensional
configuration space will limit the applicability of any algorithms that arise. However, this point of view
seems not to have been carefully explored in the literature. Indeed, we will sketch some ideas about how
an approach that starts from the coupled version of the problem may lend sufficient insight to move back
and forth between the individuals’ and the group’s configuration spaces even in real time. For the sake
of concreteness we will work in the so-called “pickup and delivery” (as opposed to the “stop and go” [2])
paradigm of assembly or fabrication, and we will not be concerned with warehousing style AGV applications.

1.2 Contributions of the Paper

The paper is organized as follows. In §2, we review fundamental facts about the topology of graphs, with
which we define the class of edge point fields — locally defined dynamics that realize single letter patter ns.
These act collectively as a toolbox from which to build a hybrid controller for achieving arbitrary patterns
with a single AGV. This represents a slight generalization of the scheme the second author and colleagues
have proposed in [4].

The problem of dynamics and control on non-trivial graphs is then considered in §3, beginning with a
detailed discussion of a natural intrinsic coordinate system in which to frame the configuration space. We
present a topological analysis of the configuration space for a pair of AGV’s on a Y-shaped graph — the
simplest nontrivial situation. Here, a clarification of the configuration space presentation leads easily to a
vector field construction that brings all initial conditions of two robots on the graph to any desired pair of
goal points while guaranteeing safety (i.e., no collisions along the way). The desire for a more decoupled
controller — the hope of an “interleaving” of otherwise independent individual patterns — impels a revised
approach to safe navigation leading to the construction of a vector field that enables the AGV’s to “dance”
about one other at a vertex.

The dynamical features of this circulating field are suggestive of a hybrid construction that would allow
multiple independent patterns to be safely interleaved. We proceed in §4 to construct a 12-symbol grammar
of so-called “monotone” cycles: those patterns which exclude multiple robots on a single edge of the graph.
The goal of this excursion is to tune limit cycles visiting various docking stations in such a way as to be
optimal with respect to certain notions of distance or performance.

We complete our treatment of this fundamental example by synthesizing the grammar results into a
state-event controlled hybrid system for achieving cooperative patterns. Appendix A is included to place on
a rigorous foundation the use of vector fields on graphs and configuration spaces thereof.

2 Notation and Background
2.1 Graph Topology

A graph, \( \Gamma \), consists of a finite collection of 0-dimensional vertices \( V := \{ v_i \}_1^N \), and 1-dimensional edges \( E := \{ e_j \}_1^M \) assembled as follows. Each edge is homeomorphic to the closed interval \([0, 1]\) attached to \( \Gamma \) along its boundary points \( \{0\} \) and \( \{1\} \). We place upon \( \Gamma \) the quotient topology given by the endpoint identifications: Neighborhoods of an edge in the interior of \( e_j \) are homeomorphic images of interval neighborhoods of the corresponding point in \([0, 1]\), and neighborhoods of a vertex \( v_i \) consist of the union of homeomorphic images of half-open neighborhoods of the endpoints for all incident edges.

The configuration spaces we consider in \( \mathbb{R}^2 \) and following are self-products of graphs. The topology of \( \Gamma \times \Gamma \) is easily understood in terms of the topology of \( \Gamma \) as follows. Let \( (x, y) \in \Gamma \times \Gamma \) denote an ordered pair in the product. Then any small neighborhood of \((x, y)\) within \( \Gamma \times \Gamma \) is the union of neighborhoods of the form \( N(u) \times N(v) \), where \( N(\cdot) \) denotes neighborhood within \( \Gamma \). In other words, the products of neighborhoods form a basis of neighborhoods in the product space.

Given a graph, \( \Gamma \), outfitted with a finite number \( N \) of non-colliding AGV’s constrained to move on \( \Gamma \), the (labeled) configuration space of safe motions is defined as

\[
C := (\Gamma \times \ldots \times \Gamma) - \bigcup_{\Delta} N(\Delta),
\]

where \( \Delta := \{(x_i) \in \Gamma \times \ldots \times \Gamma : x_j = x_k \text{ for some } j \neq k\} \) denotes the pairwise diagonal and \( N(\cdot) \) denotes (small) neighborhood.

For general graphs, the topological features of \( C \) can be extremely complicated. We do not treat the general aspects of this problem comprehensively in this paper; rather, we restrict attention to the simplest nontrivial example which illustrates nicely the relevant features present in the more general situation. The topological characteristics of general configuration spaces on graphs is treated in [1, 2]. Mathematically, it is usually most interesting to pass to the quotient of \( C \) by the action of the permutation group on \( N \) elements, thus forgetting the identities of the AGV elements; however, as such spaces are almost completely divorced from any applications involving coordinated transport, we work on the “full” configuration space \( C \).

In order to proceed, it is necessary to clarify what we mean by a vector field on a simplicial complex that fails to be a manifold. This is a nontrivial issue: for example, in the case of a graph, the tangent space to a vertex with incidence number greater than two is not well-defined. We defer a more detailed discussion of these statements to Appendix A. The essential difference is that we construct semiflows: flows which possess unique forward orbits.

2.2 Edge Point Fields for Single AGV Control

In the context of describing and executing patterns or periodic motions on a graph, one desires a set of building blocks for moving from one goal to the next. We introduce the terminology and philosophy for constructing patterns by way of the simplest possible examples: a single AGV on a graph. This avoids the additional topological complications present in the context of cooperative motion.

We thus introduce the class of edge point fields as a dynamical toolbox for a hybrid controller. Given a specified goal point \( g \in e_j \) within an edge of \( \Gamma \), an edge point field is a locally defined vector field \( X_g \) on \( \Gamma \) with the following properties:

**Locally Defined:** \( X_g \) is defined on a neighborhood \( N(e_j) \) of the goal-edge \( e_j \) within the graph topology.

Furthermore, forward orbits under \( X_g \) are uniquely defined.

**Point Attractor:** every forward orbit of \( X_g \) asymptotically approaches the unique fixed point \( g \in e_j \).

**Navigation-Like:** \( X_g \) admits a \( C^0 \) Lyapunov function, \( \Phi_g : \Gamma \rightarrow R \).

The following existence lemma (whose trivial proof we omit) holds.

\[1\] We will assume away in the sequel the possibility of “homoclinic” edges whose boundary points are attached to the same vertex.

\[2\] When it is not clear from the context, we shall denote the goal point achieved by an edge point flow as \( g(X_g) = \{g\} \).
Lemma 1 Given any edge \( e_j \subset \Gamma \) which is contractible within \( \Gamma \), there exists an edge point field \( X_g \) for any desired goal \( g \in e_j \).

The only occasion for which an edge \( e_j \) is not contractible in \( \Gamma \) is in the “homoclinic case” when both endpoints of \( e_j \) are attached to the same vertex, forming a loop. In such instances, one may avoid the problem by subdividing the edge to include more vertices, which is very natural in the setting of this paper, since vertices correspond to workstations along a path.

2.3 Discrete Regulation of Patterns

We adopt the standard framework of symbolic dynamics. By an excursion on a graph is meant a (possibly infinite) sequence of edges from the graph, \( E = e_i, \ldots, e_{iN} \in E^k \), having the property that each pair of contiguous edges, \( e_i \) and \( e_{i+1} \), share a vertex in common. The set of excursions forms a language, \( L \): the so-called subshift on the alphabet defined by the named edges (we assume each name is unique). The shift operator, \( \sigma \), defines a discrete dynamical system on the set of excursions, mapping the set of infinite sequences into itself by decrementing the time index. An \( M \)-block extension of the original language arises in the obvious way from grouping together each successive block of \( M \) contiguous letters from an original sequence, and it is clear how \( \sigma \) induces a shift operator, \( \sigma^M \) on this derived set of sequences.

Given a legal block, \( B = e_i, \ldots, e_{iN} \in L \), we will say that an excursion realizes that pattern if its \( M \)-block extension eventually reaches the “goal” \( BBBBB \ldots \) under the iterates of \( \sigma^M \). In other words, after some finite number of applications of \( \sigma \), the excursion consists of repetitions of the block \( B \) (terminating possibly with the empty edge).

In a previous paper, the second author and colleagues introduced a very simple but effective discrete event controller for regulating patterns on graphs from all reachable initial edges by pruning the graph back to a tree (imposing an ordering). Of course, this simple idea has a much longer history. In robotics it was introduced in [4] as “pre-image backchaining;” pursued in [5] as a method for building verifiable hardened automation via the metaphor of a funneling, and in [6] as a means of prescribing sensor specifications from goals and action sets. In the discrete event systems literature an optimal version of this procedure has been introduced in [7] and a generalization recently has been proposed in [8].

Let \( E^0 := B \subset E \) denote the edges of \( \Gamma \) that appear in the block of letters specifying the desired pattern. Denote by

\[
E^{n+1} \subset E - \bigcup_{k \leq n} E^k
\]

those edges that share a vertex with an edge in \( E^n \) but are not in any of the previously defined subsets. This yields a finite partition of \( E \) into “levels,” \( \{ E^p \}_{p=0} \), such that for each edge, \( e_i^p \in E^p \), there can be found a legal successor edge, \( e_j^{p-1} \in E^{p-1} \), such that \( e_i^p e_j^{p-1} \in L \) is a legal block in the language. Note that we have implicitly assumed \( E^0 \) is reachable from the entire graph — otherwise, there will be some “leftover” component of \( E \) forming the last cell in the partition starting within which it is not possible to achieve the pattern. Note as well that we impose some ordering of each cell \( E^p = \{ e_i^p \}_{i=1}^{M_p} \); the edges of \( E^0 = B \) are ordered by their appearance in the block; the ordering of edges in higher level cells is arbitrary.

We may now define a “graph controller” law, \( G: E \rightarrow E \) as follows. From the nature of the partition \( \{ E^p \} \) above, it is clear that the least legal successor function,

\[
L(e_i^p) := \begin{cases} 
  i + 1 \mod M & : p = 0 \\
  \min \{ j \leq M_p . : e_i^p e_j^{p-1} \in L \} & : p > 0
\end{cases}
\]  
(2)

is well-defined. From this, we construct the graph controller:

\[
G(e_i^p) := e_i^{p-1} L_{(p,i)}.
\]  
(3)

It follows almost directly from the definition of this function that its successive application to any edge leads eventually to a repetition of the desired pattern:

Proposition 2 The iterates of \( G \) on \( E \) achieve the pattern \( B \).
2.4 Hybrid Edge Point Fields

A semiflow, \((X)^t\), on the graph induces excursions in \(L\) parametrized by an initial condition as follows. The first letter corresponds to the edge in which the initial condition is located (initial conditions at vertices are assigned to the incident edge along which the semiflow points). The next letter is added to the sequence by motion through a vertex from one edge to the next.

We will say of two edge point fields, \(X_1, X_2\) on a graph, \(\Gamma\), that \(X_1\) prepares \(X_2\), denoted \(X_1 \succ X_2\), if the goal of the first is in the domain of attraction of the second, \(g(X_1) \subset \mathcal{N}(X_2)\). Given any finite collection of edge point fields on \(\Gamma\), we will choose some \(0 < \alpha < 1\) and assume that their associated Lyapunov functions have been scaled in such a fashion that \(X_1 \succ X_2\), implies \((\Phi_1)^{-1}[0, \alpha] \subset \mathcal{N}(X_2)\). In other words, an \(\alpha\) crossing of the trajectory \(\Phi_1 \circ (X_1)^t\) signals arrival in \(\mathcal{N}(X_2)\).

Suppose now that for every edge in some pattern block, \(e^0_i \in E^0\), there has been designated a goal point, \(g^0_i\), along with an edge point field \(X^0_i\) taking that goal, \(g(X^0_i) = g^0_i\). Assume as well that the edge point field associated with each previous edge in the pattern prepares the flow associated with the next edge, in other words, using the successor function (2) we have,
\[ g(X^0_j) \subset \mathcal{N}(X^0_{L(j)}) \]

Now construct edge point fields on all the edges of \(\Gamma\) such that the tree representation of their \(\succ\) relations is exactly the tree pruned from the original graph above — namely we have
\[ g(X^p_j) \subset \mathcal{N}(X^p_{L(j)}) \]

We are finally in a position to construct a hybrid semi-flow on \(\Gamma\). This feedback controller will run the piece-wise smooth vector field, \(\dot{x} = X\), as follows
\[
X := \left\{ \begin{array}{ll}
X^p_j : x \in e^p_j \text{ and } \Phi^p_j > \alpha \\
X^{p-1}_{L(j)} : x \in e^{p-1}_{L(j)} \text{ or } \Phi^p_j \leq \alpha
\end{array} \right.
\]
(4)

It is clear from the construction that progress from edge to edge of the state of this flow echoes the graph transition rule \(G\), constructed above.

**Proposition 3** The edge transitions induced by the hybrid controller (4) are precisely the iterates of the graph map, \(G, \Gamma\) in the language, \(L\).

3 The Y-Graph

We now turn our attention to the safe control of multiple AGV’s on a graph work-space via vector fields. For the remainder of this work, we consider the simplest example of a non-trivial configuration space: that associated to the Y-graph, \(\Upsilon\), having four vertices \(\{v_i\}_0^3\) and three edges \(\{e_i\}_1^3\), as illustrated in Figure 1.

Although this is a simple scenario compared to what one finds in a typical setting, there are several reasons why this example is in many respects canonical.

1. **Simplicity:** Any graph may be constructed by gluing \(K\)-prong graphs together for various \(K\). The \(K = 3\) model we consider is the simplest nontrivial case and is instructive for understanding the richness and challenges of local cooperative dynamics on graphs.

2. **Genericity with respect to graphs:** Graphs which consist of copies of \(\Upsilon\) glued together, the trivalent graphs, are generic in the sense that any nontrivial graph may be perturbed in a neighborhood of the vertex set so as to be trivalent. For example, the 4-valent graph resembling the letter ‘X’ may be perturbed slightly to resemble the letter ‘H’ — a trivalent graph. An induction argument shows that this is true for all graphs. Hence, the dynamics on an arbitrary graph are approximated by patching together dynamics on copies of \(\Upsilon\).
3. **Genericity with respect to local dynamics:** Finally, pairwise local AGV interactions on an arbitrary graph restrict precisely to the dynamics of two agents on $\mathcal{Y}$ as follows. Given a vertex $v$ of a graph $\Gamma$, assume that two AGV’s, $x$ and $y$, are on different edges $e_1$ and $e_2$ incident to $v$ and moving towards $v$ with the goal of switching positions. A collision is imminent unless one AGV “moves out of the way” onto some other edge $e_3$ incident to $v$. The local interactions thus restrict to dynamics of a pair of AGV’s on the subgraph defined by $\{v; e_1, e_2, e_3\}$. Hence, the case we treat in this paper is the generic scenario for the local resolution of collision singularities in cooperative dynamics on graphs.

3.1 **Intrinsic Coordinates**

The configuration space of $N$ points on $\mathcal{Y}$ is a subset of the $N$-fold cartesian product $\mathcal{Y} \times \mathcal{Y} \times \cdots \times \mathcal{Y}$. Since each graph which is physically relevant to the setting of this paper is embedded in a factory floor or ceiling and thus planar, the configuration space $\mathcal{C}^N(\mathcal{Y})$ embeds naturally in $\mathbb{R}^{2N}$. We wish to modify this embedding to facilitate both analysis on and visualization of the configuration space. We will present alternate embeddings in both higher and lower dimensional Euclidean spaces for these purposes.

We begin with representing the configuration space within a higher-dimensional Euclidean space via a coordinate system which is *intrinsic*: it is independent of how the graph is embedded in space. We illustrate this coordinate system with the $\mathcal{Y}$-graph $\mathcal{Y}$, noting that a few simple modifications yields a coordinate scheme for more general graphs.

Let $\{e_i\}_{i=1}^3$ denote the three edges in $\mathcal{Y}$, parametrized so that $e_i$ is identified with $[0, 1]$ with each $\{0\}$ at the center $v_0$ of $\mathcal{Y}$. Any point in $\mathcal{Y}$ is thus given by a vector $x$ in the $\{e_i\}$ basis whose magnitude $|x| \in [0, 1]$ determines the position of the point in the $e_i$ direction. Denote by $\iota(x)$ the value of $i$ so that $x = |x|e_{\iota(x)}$.

This parametrization embeds $\mathcal{Y}$ as the positive unit axis frame in $\mathbb{R}^3$. Likewise, a point in $\mathcal{C}$ is given as a pair of distinct vectors $(x, y)$, i.e., as the positive unit axis frame in $\mathbb{R}^3$ cross itself sitting inside of $\mathbb{R}^3 \times \mathbb{R}^3 \cong \mathbb{R}^6$.

We have thus embedded the configuration space of two distinct points on $\mathcal{Y}$ in the positive orthant of $\mathbb{R}^6$. It is clear that one can embed the more general configuration space of $N$ points on $\mathcal{Y}$ in $\mathbb{R}^{3N}$ in this manner.

This coordinate system is particularly well-suited to describing vector fields on $\mathcal{C}$ and in implementing numerical simulations of dynamics, as the coordinates explicitly keep track of the physical position of each point on the graph.

3.2 **A Topological Analysis**

More useful for visualization purposes, however, is the following construction which embeds $\mathcal{C}$ within $\mathbb{R}^3$.

![Diagram of Y-graph and configuration space](image)

**Figure 1:** [left] The $\mathcal{Y}$-graph $\mathcal{Y}$; [right] the configuration space $\mathcal{C}$ embedded in $\mathbb{R}^3$.

**Theorem 1** The configuration space $\mathcal{C}$ associated to a pair of AGV’s restricted to the $\mathcal{Y}$-graph $\mathcal{Y}$ is homeomorphic to a punctured disc with six 2-simplices attached as in Figure [7].
Corollary 4

This follows from the cellular structure of theorems of Koditschek and Rimon [12] are met, since an annullus is a measure-zero set of initial conditions to \( g \) may then extend \( \Phi \) to the remainder of the navigation function \( \Phi \) on this subspace exist, but an explicit procedure for determining \( \Phi \) is given [12]. One to an annulus \( S \) are attached respectively to \( D_{k,j} \) and \( D_{k,k} \), where \( k \) is the unique index not equal to \( i \) or \( j \).

Furthermore, each 2-cell \( D_{i,j} \) has a product structure as follows: decompose \( D_{i,j} \) along lines of constant \( \theta := \tan^{-1} \left( \frac{|y|}{|x|} \right) \). It is clear that \( \theta \) is the angle in the unit first quadrant in which \( D_{i,j} \) sits. Hence, each \( D_{i,j} \) is decomposed into a product of a closed interval \( S_{i,j} := \theta \in [0, \pi/2] \) an ‘angular’ coordinate) with the half-open interval \((0,1]\) (a ‘radial’ coordinate). As this product decomposition is respected along the gluing edges, we have a decomposition of all of \( D \) into the product of \((0,1] \times S \), where \( S \) is a cellular complex given by gluing the six segments \( S_{i,j} \) end-to-end cyclically along their endpoints. The set \( S \) is a 1-manifold without boundary since each \( S_{i,j} \) is a closed interval, each of whose endpoints is glued to precisely one other \( S_{l,m} \). Hence, by the classification of 1-manifolds, \( S \) is homeomorphic to a circle. We have thus decomposed \( D \) as the cross product of a circle with \((0,1] \) — a punctured unit disc.

The complement of \( D \) in \( C \) consists of those regions where \( \iota(x) = \iota(y) \). For each \( i = 1, 3 \), the subset of \( C \) where \( \iota(x) = \iota(y) = i \) is homeomorphic to \((0,1] \times (0,1]\) (an ‘angular’ coordinate) with the product decomposition is respected along the gluing edges, we have a decomposition of all of \( D \) into the product of \((0,1] \times S \), where \( S \) is a cellular complex given by gluing the six segments \( S_{i,j} \) end-to-end cyclically along their endpoints. The set \( S \) is a 1-manifold without boundary since each \( S_{i,j} \) is a closed interval, each of whose endpoints is glued to precisely one other \( S_{l,m} \). Hence, by the classification of 1-manifolds, \( S \) is homeomorphic to a circle. We have thus decomposed \( D \) as the cross product of a circle with \((0,1] \) — a punctured unit disc.

The precise drawing of Figure 4 represents this punctured disc \( D \) as a hexagon-shaped complex with a punctured center: this follows from the cellular structure of \( D \) as being built from six squares sewn together together.

Corollary 4

**Proof:**

Recall that \( C \) consists of pairs of distinct vectors \((x, y)\) in intrinsic coordinates. Restrict attention to the subspace \( D \subset C \) defined by

\[
D := \{ (x, y) \in C : \iota(x) \neq \iota(y) \},
\]

where an undefined index is considered to be different than one which is defined. Thus, \( D \) consists of those configurations for which both AGV’s do not occupy the same edge interior.

The set \( D \) has a natural cellular decomposition as follows. There are 2 AGV’s and 3 edges in \( \Upsilon \); hence, there are \( 2 \cdot 3 = 6 \) cells \( D_{i,j} \subset D \) where \( i := \iota(x) \neq \iota(y) = j \). Since (the closure of) each edge in \( \Upsilon \) is homeomorphic to \([0, 1] \) (determined by \( |x|, |y| \)), the cell \( D_{i,j} \) is homeomorphic to \(([0, 1] \times [0, 1]) - \{(0, 0)\}\), where, of course, the origin \((0,0)\) is removed as it belongs to the diagonal \( \Delta \). A path in \( D \) can move from cell to cell only along the subsets where the index of one AGV changes: e.g., \( |x| = 0 \) or \( |y| = 0 \). Thus, the edges \( \{0\} \times (0,1] \) and \((0,1] \times \{0\} \) of the punctured square \( S_{i,j} \) are attached respectively to \( D_{k,j} \) and \( D_{k,k} \), where \( k \) is the unique index not equal to \( i \) or \( j \).

The configuration space of a graph turns out to be aspherical: there are no essential closed spheres of dimension larger than one.\(^3\) Hence, we consider an alternate solution to the problem of

\[\begin{align*}
\Phi(x, y) := \begin{cases} 
\frac{1}{1 - |x|} \Phi(0, y) & \text{if } |x| < |y| \\
\frac{1}{1 - |y|} \Phi(x, 0) & \text{if } |y| < |x|
\end{cases}
\end{align*}\]
realizing compatible goals by means of a vector field on the configuration space. This method is adaptable to more complicated settings.

### 3.3 Example: a Circulating Flow

Before proceeding with a general scheme for controlling two agents on the Y-graph, we present a simple example of a vector field on the configuration space which can be used to regularize collisions about a generic trivalent vertex. Theorem 1 and Figure 1 suggest a natural circulating flow on the configuration space $C$ which has the effect of inducing a “dance” between the pair of AGV’s which cycles through all combinations of distinct point goals.

**Theorem 2** There exists a piecewise-smooth vector field $X$ on $C$ which has the following properties:

1. $X$ defines a nonsingular semiflow on $C$;
2. The diagonal $\Delta$ is repelling with respect to $X$; and
3. Every orbit of $X$ approaches a unique attracting limit cycle on $C$ which cycles through all possible ordered pairs of distinct edge-states.

**Proof:** Denote by $D$ that portion of the configuration space which corresponds to the AGV’s being on distinct edges of the graph: as proven earlier, $D$ is homeomorphic to a punctured disc. The intrinsic coordinates on the configuration space $C$ is illustrated in Figure 2, where only $D$ is shown for illustration purposes. The reader should think of this as a collection of six square coordinate planes, attached together pairwise along axes with the origin removed (this is actually an isometry for the natural product metric). The six triangular fins are then attached as per Figure 1.

![Figure 2](image)

Figure 2: [left] The coordinate system on the unfinned region $D$ of $C$; [right] The circulating flow with a typical orbit.

Recall that any point in the graph is represented as a vector $x = |x|e_i$ for some $i$. Denote by $\hat{e}_i$ the unit tangent vector in each tangent space $T_{e_i}e_i$ pointing in the positive (outward) direction towards the endpoint $v_i$. The vector field we propose is the following: given $(x, y) \in C$,

1. If $\iota(x) = \iota(y)$ then

\[
\begin{align*}
\dot{x} &= -|y|\hat{e}_{\iota(x)} \\
\dot{y} &= |y|(1-|y|)\hat{e}_{\iota(y)} \\
\dot{x} &= |x|(1-|x|)\hat{e}_{\iota(x)} \\
\dot{y} &= -|x|\hat{e}_{\iota(y)}
\end{align*}
\]

(7)
SAFE ROBOT DYNAMICS

2. If \( \iota(x) = \iota(y) + 1 \) or \(|x| = 0\) then
   \[
   \begin{cases}
   \dot{x} = |y| \hat{e}_{\iota(y) + 1} \\
   \dot{y} = |y|(1 - |y|) \hat{e}_{\iota(y)} \\
   \dot{x} = |x|(1 - |x|) \hat{e}_{\iota(x)} \\
   \dot{y} = -|x| \hat{e}_{\iota(y)}
   \end{cases}
   \]  
   \begin{align}
   0 \leq |x| < |y| \\
   0 < |y| \leq |x|
   \end{align} \tag{8}

3. If \( \iota(y) = \iota(x) + 1 \) or \(|y| = 0\) then
   \[
   \begin{cases}
   \dot{x} = -|y| \hat{e}_{\iota(x)} \\
   \dot{y} = |y|(1 - |y|) \hat{e}_{\iota(y)} \\
   \dot{x} = |x|(1 - |x|) \hat{e}_{\iota(x)} \\
   \dot{y} = |x| \hat{e}_{\iota(x) + 1}
   \end{cases}
   \]  
   \begin{align}
   0 < |x| \leq |y| \\
   0 \leq |y| < |x|
   \end{align} \tag{9}

Note that all addition operations on \( \iota(x) \) and \( \iota(y) \) are performed mod three.

The vector field is nonsingular as follows: if \(|x| |y| \neq 0\), then the vector field is by inspection nonsingular. If \(|x| = 0\), then \(|y| > 0 = |x|\) since the points are distinct. It then follows from Equation (8) that the vector field on this region has \( d|x|/dt = |y| \neq 0 \). A similar argument holds for the case where \(|y| = 0\).

The vector field defines a semiflow as follows: on those regions where \( 0 \neq |x| \neq |y| \neq 0 \), the vector field is smooth and hence defines a true flow. Along the lines where \(|x| = |y|\), the vector field is only \( C^0 \), but nevertheless is constructed so as to define unique solution curves; hence the region \( D \), where \( \iota(x) \neq \iota(y) \), is invariant under the flow. Finally, along the branch line curves where \(|x| = 0 \) or \(|y| = 0\), the vector field points into the the branch lines from the fins, implying that the dynamics is a semiflow (see the remarks in Appendix A).

This vector field admits a \( C^0 \) Lyapunov function \( \Phi : \mathcal{C} \to [0, 1) \) of the form

\[
\Phi(x, y) := \begin{cases}
1 - |(|x| - |y|)| : \iota(x) = \iota(y) \\
1 - \max\{|x|, |y|\} : \iota(x) \neq \iota(y)
\end{cases}.
\tag{10}
\]

From Equation (7), one computes that on the fins \( (\iota(x) = \iota(y)) \),

\[
\frac{d\Phi}{dt} = -\left| \frac{d|x|}{dt} - \frac{d|y|}{dt} \right| < 0,
\]

since here \(|x| \neq |y|\). Furthermore, on the disc \( D \) \( (\iota(x) \neq \iota(y)) \), \( \Phi \) changes as

\[
\frac{d\Phi}{dt} = \Phi(\Phi - 1).
\tag{12}
\]

Hence, \( \Phi \) strictly decreases off of the boundary of the disc

\[
\partial D := \{(x, y) : |x| = 1 \text{ or } |y| = 1\} = \Phi^{-1}(0).
\tag{13}
\]

It follows from the computation of \( d\Phi/dt \) that the diagonal set \( \Delta \) of \( \Upsilon \times \Upsilon \) is repelling, and that the boundary cycle \( \partial D \) is an attracting limit cycle.

The action of the vector field is to descend off of the “fins” of \( \mathcal{C} \) onto the region \( D \), and then to circulate about while pushing out to the boundary cycle \( \partial D \), as in Figure 2.

This example illustrates how one can use a relatively simple vector field on the configuration space to construct a pattern which is free from collisions. In fact, one could use this circulating flow to regularize potential collisions between AGV’s in a general graph setting by localizing the dynamics near a pairwise collision to those on a trivalent subgraph.
4 Patterns and Vector Fields for Monotone Cycles

Optimization of patterns in the workspace is deeply entwined with the geometry of the configuration space: in [10], it is shown that various Finsler structures on $C$ can be chosen to optimize total distance traveled or net time elapsed. The net result of this inquiry is that minimizing Euclidean distance (in the product of the graph metric) on the cells of the configuration space yields locally optimal configuration sequences with respect to both distance traversed and elapsed time. In this section, we consider the problem of constructing vector fields which are tuned to trace out specific patterns of cooperative dynamics. We begin with a specification of a suitable language for describing patterns.

4.1 A Grammar for Patterns

The setting we envisage is as follows: the three ends of the graph $\Upsilon$ are stations at which an AGV can perform some function. The AGV pair is required to execute an ordered sequence of functions, requiring an interleaved sequence of visitations. In order to proceed with vector field controls for cooperative patterns, it is helpful to construct the appropriate symbolic language, as done in §2 for single AGV systems. Denote the pair of AGV states as $x$ and $y$ respectively. Also, denote the three docking stations as vertices $v_1$ through $v_3$ as in Figure 1. The grammar $G$ we use is defined as follows:

- $(x_i)$: These represent configurations for which the AGV $x$ is docked at the vertex $v_i$, $i = 1..3$. The AGV $y$ is at an unspecified undocked position.

- $(y_i)$: These represent configurations for which the AGV $y$ is docked at the vertex $v_i$, $i = 1..3$. The AGV $x$ is at an unspecified undocked position.

- $(x_iy_j)$: These represent configurations for which the AGV $x$ is docked at vertex $v_i$ while the AGV $y$ is simultaneously docked at the vertex $v_j$, $j \neq i$.

For example, the word $(x1)(y2)(x3y2)$ executes a sequence in which the first AGV docks at Station $v_1$, then undocks while the second AGV docks at Station $v_2$. Finally, the AGV’s simultaneously dock at Stations $v_3$ and $v_2$ respectively.

As we have assumed from the beginning, the one-dimensional nature of the graph-constraints precludes the presence of multiple agents at a single docking station; hence, there are exactly twelve symbols in the grammar $G$. From this assumption, it follows that particular attention is to be paid to those trajectories which do not make excursions onto the “fins” of the configuration space. It is obvious from the physical nature of the problem that planning paths which involve traveling on the fins is not a locally optimal trajectory with respect to minimizing distance or elapsed time. Suffice to say that we restrict attention for the moment to trajectories, and limit cycles for patterns in particular, which are constrained to the region $D \subset C$.

We identify each symbol with a region of the boundary of the unbranched portion of $C$; namely, $\partial D$ is partitioned into twelve docking zones as in Figure 3. Note further that there is a cyclic ordering, $<_{\partial}$, on $G$ induced by the orientation on the boundary of the disc along which the zones lie. By a cyclic ordering, we mean a way of determining whether a point $q$ lies between any ordered pair of points $(p_1,p_2)$.

We proceed with the analysis of limit cycles on $C$. Consider the class of pattern vector fields, $X_P$, on $C$; for every $X \in X_P$,

1. $X$ defines a semiflow on $C$ and a genuine flow off the non-manifold set of $C$;

2. There is a unique limit cycle $\gamma$ which is attracting and which traces out a nonempty word in the grammar $G$;

3. The diagonal set $\Delta$ is a repellor with respect to $X$;

4. There are no fixed invariant sets of $X$ which attract a subset of positive measure save $\gamma$.

Then, the class of monotone vector fields, $X_M$, is that subset of $X_P$ for which the limit cycle, $\gamma$, lies in $D$. A word $w$ composed of elements $w = w_1w_2...w_n$ in the grammar $G$ said to be monotone with respect to the
Figure 3: The six pairs of contiguous edges in $\partial D$ each corresponds to a configuration where one AGV is docked at an extreme vertex of the graph. Outermost vertices of $\partial D$ are points where both AGV’s are docked. Labelling the edges $(x_i)$ and $(y_j)$ and the vertices $(x_i y_j)$ yields the cyclically ordered grammar $\mathcal{G}$.

cyclic ordering $<_{\partial}$ if $w_{i-1} < w_i < w_{i+1}$ for every $i$ (index operations all mod $n$). The following result justifies our use of the term monotone in describing those limit cycles which lie on the disc.

**Theorem 3** Within the class of vector fields $\mathcal{X}_M$, the limit cycles trace out monotone words in the cyclically ordered grammar $(\mathcal{G}, <_{\partial})$.

**Proof:** Any limit cycle of the flow must be embedded (the curve does not intersect itself). After a small perturbation, one may assume that the boundary zone $\partial D$ is visited by $\gamma$ in a finite number of points, $Q := \gamma \cap \partial D$. There is a cyclic order $<_t$, defined via time with respect to the dynamics of the limit cycle: i.e., the order in which points are visited by $\gamma$. This is contrasted with the induced cyclic ordering $<_{\partial}$ on the set $Q$ given by the orientation of the boundary curve $\partial D$, up to a choice between clockwise and counterclockwise. The theorem follows from showing that $<_t = <_{\partial}$ up to a cyclic permutation and a choice of orientation of $\partial D$.

Induct on $J$ the number of points in $Q$. For $J = 1$, the theorem is trivially true, so assume that the orderings are equivalent for all embedded curves on a disc with less than $J$ boundary-intersections. Consider two points $p, q \in Q$ which are consecutive in the $<_t$ ordering. There is an embedded sub-arc $\alpha \subset \gamma$ which connects $p$ to $q$ within the interior of $D$. By the Jordan Curve Theorem \[17\], $\alpha$ separates $D$ into two topological discs; hence, $\gamma$ must lie entirely within the closure of one of these discs. Consider the open subdisc whose intersection with $\gamma$ is empty and collapse the closure of this disc to a point: this yields a modified curve $\tilde{\gamma}$ which is still embedded in a disc having $J - 1$ intersections with the boundary, as illustrated in Figure 4. By induction, the ordering $<_t$ equals $<_{\partial}$ up to orientation on this subdisc. Reinserting the distinct points $p$ and $q$ by “blowing up” the crushed disc does not change the ordering properties, since these were chosen to be adjacent.

Hence, the only admissible words in the grammar $\mathcal{G}$ are those which are monotone. It is, however, possible to realize many if not all of the non-monotone cycles as limit cycles for a semiflow on the full configuration space $\mathcal{C}$; one must design the semiflow so as to utilize the fins for “jumping” over regions of $D$ cut off by the limit cycle. Such vector fields quickly become very convoluted, even for relatively simple non-monotone limit cycles.
Figure 4: The embedded arc $\alpha$ divides $D$ (pictured as a smooth disc) into two discs, one of which is collapsed to a point $x$.

4.2 Isotopy Classes of Limit Cycles

Given a limit cycle $\gamma$ which traces out a pattern by visiting the boundary zone $\partial D$ in the ordered set $Q \subset \partial D$, one wants to know which other limit cycles minimize a given performance functional while still visiting $Q$ in the proper sequence. The mathematical framework for dealing with this problem is the notion of isotopy classes of curves.

Two subsets $A_0$ and $A_1$ of a set $B$ are said to be (ambiently) isotopic rel $C$ (where $C \subset B$) if there exists a continuous 1-parameter family of homeomorphisms, $f_t : B \to B$ such that

1. $f_0$ is the identity map on $B$;
2. $f_1(A_0) = A_1$; and
3. $f_t|_C$ is the identity map on $C$ for all $t$.

As $t$ increases, $f_t$ deforms $B$, pushing $A_0$ to $A_1$ without cutting or tearing the spaces and without disturbing $C$.

There are two ways in which optimization questions relate to isotopy classes of limit cycles: (1) Given an element of the grammar $G$, in which isotopy class (rel the docking zones) of curves does an optimal limit cycle reside? (2) Within a given isotopy class of cycles rel $Q$, which particular cycle is optimal?

For a monotone limit cycle on $D$, the above question (1) focuses on the location of the cycle with respect to the central point $(0,0)$, which is deleted from the disc $D$. It is a standard fact from planar topology that every curve in the punctured disc has a well-defined winding number, which measures how many times the cycle goes about the origin, and, furthermore, that this number is either -1, 0, or 1 if the cycle is an embedded curve. This winding number determines the isotopy class of the curve in $D$. Hence, the problem presents itself: given an element of the grammar $G$, which isotopy class rel the docking zones is optimal (with respect to any/all of the functionals defined)? Is the winding number zero or nonzero?\footnote{The difference between $+1$ and $-1$ is the orientation of time.}

To briefly address this question, we define the gap angles associated to a limit cycle. Given the docking zones $Q = \{q_1, q_2, \ldots, q_J\}$ ordered with respect to time, we define the gap angles to be the successive differences in the angular coordinates of the $q_j$: thus $ga_j := P(q_{j+1}) - P(q_j)$, where $P$ denotes projection of points in $D$ onto their angular coordinates and subtraction is performed with respect to the orientation on $\partial D$.

For simplicity, we consider the optimization–isotopy problem in the case of a discrete cost functional $W_d$, defined to be the intersection number of the path with the branch locus of $C$ — i.e., the number of times an AGV occupies the central vertex (the shared resource in the problem). Similar arguments are possible for other natural performance metrics \cite{11}.

\textbf{Proposition 5} There is a $W_d$-minimizing embedded monotone cycle on $D$ (rel a given docking zone $Q$) having winding number zero with respect to the origin if there is a gap angle greater than $\pi$. Conversely, if there are no gap angles greater than $\pi$, then there is a $W_d$-minimizing embedded cycle of index $\pm 1$. 
Lemma 6 The following is a piecewise-linear homeomorphism from the punctured unit disc in $R^2$ to the

Proof: Let $Q$ consist of the points $\{q_j\}_{i=1}^3$ on the boundary circle. The gap angles $\{ga_j\}_{i=1}^3$ are the differences of the angles between the points $q_j$ and $q_{j+1}$ (indices mod $J$). Since $\sum_j ga_j = 2\pi$, there can be at most one gap angle greater than $\pi$. To simplify the problem, use a 1-parameter family $P_t$ of maps from the identity $P_0$ to the projection $P = P_1$ which deforms $D$ to the boundary circle $S := \partial D$ by projecting continuously along radial lines. The index of a curve on $D$ is invariant under this deformation, as is the function $W_d$.

Denote by $\gamma_j$, the subarc of $\gamma$ between points $q_j$ and $q_{j+1}$ (all indices mod $J$). Denote by $\alpha_j$ the subarc of the boundary $S$ between points $q_j$ and $q_{j+1}$, where the arc is chosen to subdivide the gap angle $ga_j$. Since the boundary curve $S = \cup_j \alpha_j$ is a curve of index $\pm 1$, the arcs $\gamma_j$ and $\alpha_j$ are isotopic in $D$ rel their endpoints for all $j$ if and only if $\gamma$ is a curve of index $\pm 1$.

Assume first that there is a gap angle $ga_j > \pi$ with $\gamma$ an index $\pm 1$ curve on $S$ which intersects the branch angles $\Theta = \{n\pi/3 : n \in \mathbb{Z}\}$ in a minimal number of points among all other closed curves on $S$ which visit the points $Q$ in the specified order. It follows that the arc $P(\gamma_j)$ subtends an angle greater than $\pi$ and thus increments $W_d$ by at least three. One may replace $\gamma_j$ by a curve $\gamma'_j$ which substitutes for the arc $\gamma_j$ one which wraps around the other way monotonically. This changes the index of $\gamma$ from nonzero to zero, since the arc $\gamma'_j$ is no longer isotopic to $\alpha_j$. Also, it is clear that this either decreases the number of intersections with $\Theta$ or leaves this number unchanged.

We must show that the replacement arc $\gamma'_j$ can be chosen in such a way that it does not intersect the remainder of $\gamma$. However, since $\gamma$ is a curve of index $\pm 1$, we may isotope each arc $\gamma_i$ to the boundary curve $\alpha_i$ without changing the value of $W_d$. Thus, we may remove $\gamma_j$ and replace it with the curve which is, say, a geodesic (in the natural metric geometry) from $q_j$ to $q_{j+1}$. As this curve does not approach the boundary $S$ apart from its ends, the new curve $\gamma'$ which intersects the branch set in at most three components, since, otherwise, the subtended arc would be in excess of $4\pi/3$. In the case where there exists an arc with exactly three intersections with the branch set, this arc may be replaced by an arc which goes around the singularity in the other direction without changing the number of intersections with the branch set (since there are a total of six branch lines); however, the index of the curve is toggled between zero and nonzero.

The final case is that in which each arc intersects the branch set in at most two places. However, since $\gamma$ is a curve of index zero, some arc $\gamma_j$ must not be isotopic to $\alpha_j$. Hence, the projection deformations $P_t$ must push $\gamma_j$ to a curve in the boundary which $S$ whose subtended gap angle is $2\pi - ga_j > \pi$. Thus, $\gamma_j$ intersects the branch set in at least three places, yielding a contradiction. Replacing $\gamma_j$ by the appropriate arc which is isotopic to $\alpha_j$ yields a $W_d$-minimal cycle of nonzero index. \hfill \qed

4.3 Tuning Cycles

In order to proceed with the construction of vector fields which realize monotone cycles, we work with vector fields on the smooth unit disc in $R^2$ and map these to the annular region $D$ of the configuration space via the push-forward induced by the natural homeomorphism. It will be convenient to keep track of which “wedge” of the annular region a point $(r, \theta)$ is. To do so, we introduce a parity function

$$P(\theta) := (-1)^{\lfloor 3\theta/\pi \rfloor + \lfloor 6\theta/\pi \rfloor},$$

where $[t]$ is the integer-valued floor function. Recall the notation for the intrinsic coordinates for a point $x$ on the graph $T$: $x = |x|e_\theta(x)$, where $|x| \in [0, 1]$ is the distance from $x$ to the central vertex, and $e_\theta(x)$ is the unit tangent vector pointing along the direction of the $e_\theta(x)$-edge. Here the index, $e(x)$ is an integer (defined modulo 3) and will be undefined in the case when $|x| = 0$, i.e., $x$ is at the central vertex.

Lemma 6 The following is a piecewise-linear homeomorphism from the punctured unit disc in $R^2$ to the
SAFE ROBOT DYNAMICS

subset $\mathcal{D}$: $F(r, \theta) = (x, y)$ where,

$$
|x| = \begin{cases} 
  r & \mathcal{P}(\theta) = +1 \\
  r \cot \frac{3}{2} \theta & \mathcal{P}(\theta) = -1 \\
  r \tan \frac{3}{2} \theta & \mathcal{P}(\theta) = +1 \\
  r & \mathcal{P}(\theta) = -1
\end{cases}
$$

$$
|y| = \begin{cases} 
  \frac{3}{2 \pi} (\theta - \pi) & \mathcal{P}(\theta) = +1 \\
  \frac{3}{2 \pi} (\theta) + 1 & \mathcal{P}(\theta) = -1
\end{cases}
$$

The inverse of this homeomorphism is given by $F^{-1}(x, y) = (r, \theta)$, where,

$$
\theta = \begin{cases} 
  \frac{2}{3} \tan^{-1} \frac{|y|}{|x|} - \frac{2 \pi}{3} (\iota(y) + 1) & \iota(y) = \iota(x) + 1 \\
  -\frac{2}{3} \tan^{-1} \frac{|y|}{|x|} - \frac{2 \pi}{3} (\iota(x) - 1) & \iota(x) = \iota(y) + 1
\end{cases}
$$

$$
r = \begin{cases} 
  |x| & \mathcal{P}(\theta) = +1 \\
  |y| & \mathcal{P}(\theta) = -1
\end{cases}
$$

Note that all $\theta$ values are defined modulo $2\pi$ and all index values are integers defined modulo 3.

**Proof:** Begin by working on the region $D_{1,2} \subset \mathcal{D}$ where $\iota(x) = 1$ and $\iota(y) = 2$. As noted earlier, this subspace is isometric to the positive unit square in $R^2$ with the origin removed. We need to map this to the subset $\{(r, \theta) : r \in (0, 1], \theta \in [0, \pi/3]\}$. The simplest such homeomorphism is to first shrink along radial lines, leaving the angle invariant; hence

$$
r = \begin{cases} 
  |x| & |x| \leq |y| \\
  |y| & |y| \leq |x|
\end{cases}
$$

Next, we squeeze the quarter-circle into a sixth of a circle by multiplying the angle by $2/3$, leaving the radial coordinate invariant:

$$
\theta = \frac{2}{3} \tan^{-1} \frac{|y|}{|x|}
$$

This gives the basic form of $F^{-1}$ as per Equation (16). To extend this to the remainder of $\mathcal{D}$, it is necessary to carefully keep track of $\iota(x)$ and $\iota(y)$ and subtract off the appropriate angle from the computation of $\theta$. Also, the condition of $|x| \leq |y|$, etc., in Equation (17) is incorrect on other domains of $\mathcal{D}$, since the inequalities flip as one traverses from square to square: the parity function $\mathcal{P}(\theta)$ keeps track of which “wedge” one is working on.

To determine $F$ from $F^{-1}$ is a tedious but unenlightening calculation, made more unpleasant by the various indices to be kept track of. Briefly, given $r$ and $\theta$ on the first sixth of the unit disc, one knows from Equation (17) that either $|x| = r$ or $|y| = r$, depending on whether $\theta$ is above or below $\pi/4$. To solve for the other magnitude, one inverts Equation (16) to obtain $|y| = r |\tan \frac{3}{2} \theta|$ or $|x| = r |\cot \frac{3}{2} \theta|$ respectively. To generalize this to the other $D_{i,j}$ domains of $\mathcal{D}$, it is necessary to take absolute values and to use the parity function $\mathcal{P}(\theta)$ as before. Finally, the computation of the index is obtainable from the combinatorics of the coordinate system as illustrated in Figure 4.

Hence, by taking the push-forward of a vector field $X = (\dot{r}, \dot{\theta})$ with respect to $F$, one obtains the piecewise smooth vector field,

$$
\dot{x} = \dot{r} \left\{ \begin{array}{ll} 
  r & \mathcal{P}(\theta) = +1 \\
  r \cot \frac{3}{2} \theta & \mathcal{P}(\theta) = +1 \\
  r \tan \frac{3}{2} \theta & \mathcal{P}(\theta) = +1
\end{array} \right.
$$

$$
\dot{y} = \dot{r} \left\{ \begin{array}{ll} 
  \frac{3}{2 \pi} (\theta - \pi) & \mathcal{P}(\theta) = +1 \\
  \frac{3}{2 \pi} (\theta) + 1 & \mathcal{P}(\theta) = -1
\end{array} \right.
$$

$$
\dot{\theta} = \dot{r} \left\{ \begin{array}{ll} 
  \frac{2}{3} \tan^{-1} \frac{|y|}{|x|} - \frac{2 \pi}{3} (\iota(y) + 1) & \iota(y) = \iota(x) + 1 \\
  -\frac{2}{3} \tan^{-1} \frac{|y|}{|x|} - \frac{2 \pi}{3} (\iota(x) - 1) & \iota(x) = \iota(y) + 1
\end{array} \right.
$$

$$
\dot{r} \left\{ \begin{array}{ll} 
  |x| & \mathcal{P}(\theta) = +1 \\
  |y| & \mathcal{P}(\theta) = -1
\end{array} \right.
$$
in addition to the proximity to β

The Lyapunov functions φβ take the push-forward of the vector field \( \dot{r} \) on the vector field of Equation (21) and, if desired, takes the image of this vector field under Equation (20).

\( \Phi_j \) Lyapunov function Ψβ\( \mathcal{D}_\alpha \) intrinsic coordinates on α arcs via the homeomorphism of Equation (16). Then, representing the limit cycle chords can be used to generate all monotone limit cycles on specifications. There is nothing in this construction which relies on the index-zero property and thus these which vary discretely in time and which may be pieced together so as to tune a limit cycle to the desired

\( \{ \) limit cycles, or chords \( \} \)

To design optimal cycles with winding number zero, then, we turn to constructing customized portions of 4.4 Optimal Chords within a Hybrid Controller and (16) to map it into intrinsic coordinates.

which simplifies to:

\[
\begin{align*}
\begin{cases}
|\dot{x}| = \dot{r} \\
|\dot{y}| = \dot{r} |\frac{y}{x}| + \frac{3}{2} \dot{\theta} \left( 1 + \frac{|x|}{|y|} \right)
\end{cases}
\quad \mathcal{P}(\theta) = +1
\end{align*}
\]

Given a simple closed curve γ in \( \mathbb{R}^2 \) which has nonzero winding number with respect to the origin, γ may be parametrized as \( \{(r, \theta) : r = f(\theta)\} \) for some periodic positive function f. To construct a vector field on \( \mathbb{R}^2 \) whose limit sets consist of the origin as a source and γ as an attracting limit cycle, it suffices to take the push-forward of the vector field \( \dot{r} = r(1-r) \quad \dot{\theta} = \omega \) under the planar homeomorphism \( \phi : (r, \theta) \mapsto (f(\theta)r, \theta) \), which rescales linearly in the angular component. The calculations follow:

\[
\phi_\ast \left( \begin{array}{c} \dot{r} \\ \dot{\theta} \end{array} \right) = D\phi \left( \begin{array}{c} \dot{r} \\ \dot{\theta} \end{array} \right) \bigg|_{r=r_f} = \left[ \begin{array}{cc} f & rf' \\ 0 & 1 \end{array} \right] \left( \begin{array}{c} r(1-r) \\ \omega \end{array} \right) \bigg|_{r=r_f}.
\]

Hence, given f(θ), we may tune a vector field to trace out the desired limit cycle and then use Equations (13) and (14) to map it into intrinsic coordinates.

4.4 Optimal Chords within a Hybrid Controller

To design optimal cycles with winding number zero, then, we turn to constructing customized portions of limit cycles, or chords which can be pieced together via a state-actuated hybrid controller, much as in [2]. In other words, instead of building a simple fixed vector field with a limit cycle, we will use a set of vector fields which vary discretely in time and which may be pieced together so as to tune a limit cycle to the desired specifications. There is nothing in this construction which relies on the index-zero property and thus these chords can be used to generate all monotone limit cycles on \( \mathcal{C} \).

Let \( \mathcal{G} \) denote a word representing a desired monotone limit cycle on the configurations space \( \mathcal{C} \). Choose points \( \{q_i\} \) on the boundary of \( \mathcal{D} \) which correspond to the docking zones for the cycle given by \( \mathcal{G} \). Choose arcs \( \alpha_i \) on \( \mathcal{D} \) which connect \( q_i \) to \( q_{i+1} \) (using cyclic index notation). The arcs \( \alpha_i \) are assumed given in the intrinsic coordinates on \( \mathcal{D} \), as would be the case if one were determining a length-minimizing curve.

In the case where the limit cycle \( \alpha := \cup_i \alpha_i \) is an embedded curve of nonzero index, the procedure of the previous subsection determines a vector field \( X_\alpha \) on \( \mathcal{C} \) which realizes \( \alpha \) as an attracting limit cycle with the appropriate dynamics on the complementary region. Recall: one translates \( \alpha \) to a curve on the disc model via the homeomorphism of Equation (11). Then, representing the limit cycle \( \alpha \) as a function \( f_\alpha(\theta) \), one takes the vector field of Equation (21) and, if desired, takes the image of this vector field under Equation (20).

If, however, this is not the case, consider the arc \( \alpha_j \) for a fixed \( j \) and construct an index ±1 cycle \( \beta^j = \cup_i \beta^j_i \) which has docking zones \( \{q_i\} \) such that \( \beta^j_i = \alpha_j \). Then the vector field \( X^j \) as constructed above has \( \beta \) as an attracting limit cycle. Denote by \( \Phi^j \) the Lyapunov function which measures proximity to \( \beta \): \( \Phi^j(p) := \|p - \beta^j\| \) (with distance measured in say the product metric on \( \mathcal{C} \)). Then, consider the modified Lyapunov function \( \Psi^j(p) := \Phi^j(p) + \|p - q_{j+1}\| \), which measures the distance to the endpoint of the arc \( \beta^j_j \) in addition to the proximity to \( \beta^j \).

Repeat this procedure for each \( j \), yielding the vector fields \( \{X^j\} \) which attract respectively to limit cycles \( \beta^j \). It follows that \( X^j \) prepares \( X^{j+1} \) since the goal point of \( X^j \), \( q_{j+1} \) lies on the attracting set of \( X^{j+1} \). The Lyapunov functions \( \{\Psi^j\} \) serve as a set of funnels which channel the orbit into the sequence of arcs \( \alpha_j \),
forming $\alpha$. One scales the $\Psi^j$ so that a $\Psi^j < \epsilon$ event triggers the switching in the hybrid controller from $X^j$ to $X^{j+1}$:

$$X := \begin{cases} 
X^1 & : \Phi^j > \epsilon \forall j \\
X^j & : \Phi^j < \epsilon \text{ and } \Psi^j > \epsilon
\end{cases}$$

(22)

By construction, the hybrid controller (22) realizes a limit cycle within $\epsilon$ of $\alpha$ as the attracting set.

5 Future Directions

A point of primary concern is the adaptability of the global topological approach to systems which increase in complexity, either through more intricate graphs or through increased numbers of AGV’s. The latter is of greater difficulty than the former, since the dimension of the resulting configuration space is equal to the number of AGV’s. Hence, no matter how simple the underlying graph is, a system with ten independent AGV’s will require a dynamical controller on a (topologically complicated) ten-dimensional space: a formidable problem both from the topological, dynamical, and computational viewpoints.

However, there are some approaches which may facilitate working with such spaces. Consider the model space $\mathcal{C}$ with which this paper is concerned: although a two-dimensional space, $\mathcal{C}$ can be realized as the product of a graph (a circle with six radial edges attached) with the interval $(0, 1]$. In fact, if we consider the circulating flow of Equations (7)-(9), one can view this as a product field of a semiflow on the graph (which “circulates”) with a vector field on the factor $(0, 1]$ (which “pushes out” to the boundary).

A similar approach is feasible for arbitrary graphs. The following result has recently been proven [8]:

**Theorem 4** Given a graph $\Gamma$, the configuration space of $N$ distinct points on $\Gamma$ can be deformation retracted to a subcomplex whose dimension is bounded above by the number of vertices of $\Gamma$ of valency greater than two.

This theorem implies the existence of low-dimensional spines which carry all of the topology of the configuration space. For example, the above theorem implies that the configuration space of $N$ points on the Y-graph can be continuously deformed to a one-dimensional graph, regardless of the size of $N$. Since the full space can be deformation retracted onto the spine, a vector field defined on the spine can be pulled back continuously to the full configuration space, thus opening up the possibility of reducing the control problem to that on a simpler space. Additional results about the topology of configuration spaces on graphs may yield computationally tractible means of dealing with complex path planning: for example, having a presentation for the fundamental group of a configuration space of a graph in terms of a suitably simple set of cycles would be extremely well-suited to a hybrid control algorithm based on “localized” vector fields supported on small portions of the full configuration space.

The optimization problem is another avenue for inquiry. The fact that a dynamical approach allows for increased density of AGV’s on a graph (as compared with blocking-zone strategies) would indicate an increased efficiency with respect to, say, elapsed time-of-flight. However, a more careful investigation of the tuning of optimal cycles is warranted.

We believe that the benefits associated with using the full configuration space to tune optimal dynamical cycles justifies a careful exploration of these challenging spaces.

A The Topology and Dynamics of Graphs

In this appendix, we provide a careful basis for the use of vector fields on configuration spaces of graphs. In the setting of manifolds, all of the constructions used in this paper are entirely natural and well-defined. However, on spaces like $\mathcal{C}$, the most fundamental of notions (like the Existence and Uniqueness Theorems for ODEs) are not in general valid.

We begin by defining vector fields on graphs. For present purposes, we find it convenient to work with an intrinsic formulation (i.e., directly in the graph rather than via an embedding in a Euclidean space) of

---

5 We have since learned of two others who have independently proved this result: [11], having learned of these spaces from [1], and [1], who discovered these spaces while working on the topology of Brownian motion on graphs.
these objects. To this end, denote by \( v \) a vertex with \( K \) incident edges \( \{e_i\}_{i=1}^K \), and by \( \{X_i\}_{i=1}^K \) a collection of nonsingular vector fields locally defined on a neighborhood of the endpoint of each \( e_i \) (homeomorphic to \( [0,1) \)).

**Lemma 7** A set of nonsingular vector fields \( \{X_i\} \) on the local edge set of a graph \( \Gamma \) generates a well-defined semiflow on \( \Gamma \) if

1. Each edge field \( X_i \) generates a well-defined local semiflow on \((0,1)\); and

2. The magnitude of the endpoint vectors \( \|X_i(0)\| \) (taken with respect to the attaching homeomorphisms) are all identical; and

3. Among the signs of the endpoint vectors \( X_i(0) \) (either positive if pointing into \([0,\epsilon)\) or negative if pointing out) there is a single positive sign.

**Proof:** Since the vector field is well-defined away from the vertex, it is only necessary to have the magnitudes \( \|X_i(0)\| \) agree in order to have a well-defined function \( \|X\| \) on \( \Gamma \). In order to make this a well-defined field of directions, we must also consider in which direction the vector is pointing. Again, this is determined off of the vertex by (1). Condition (3) means that at the vertex, there is a unique direction along which the vector field is pointing out: all other edges point in. Hence, the direction field, as well as the magnitude field, is well-defined.

The semiflow property follows naturally from this. Assume that the \( N^{th} \) edge of \( \Gamma \) has the positive sign. Then, given an initial point \( x \in \Gamma \), if \( x \in e_N \), then the orbit of \( x \) under the local field \( X_N \) remains in \( e_N \) and is well-defined. If \( x \in e_j \) for some \( j \neq N \), then the union of the edges \( e_j \cup e_N \) is a manifold homeomorphic to \( R \) on which the vector fields \( X_j \) and \( X_N \) combine to yield a well-defined vector field, since the directions are “opposite.” As we are now on a manifold, the standard Existence Theorem implies that \( x \) has a forward orbit (which passes through the vertex and continues into \( e_N \)). Thus every point on \( \Gamma \) has a well-defined forward orbit. \( \diamond \)

In the case where the vector fields have singularities, it is a simpler matter. If the singularities are not at the vertex, then there is no difference. If there is a singularity at the vertex, then condition (3) in Lemma 7 is void — all such vector fields are well-defined.

In order to extend these results to the configuration space of this paper, consider the space \( C = \Upsilon \times \Upsilon - \Delta \) and let \((x,y) \in C\) denote a point on the branch set of \( C \). Because of the structure of \( \Upsilon \) and the fact that the diagonal points are deleted, it follows that at most one AGV may occupy a non-manifold point of \( \Upsilon \). Hence, a neighborhood of \((x,y)\) in \( C \) has a natural product structure \( N \cong \T \times R \). Let \( P : N \rightarrow \Upsilon \) denote projection onto the first factor.

**Lemma 8** A nonsingular vector field \( X \) on the individual cells of \( C \) generates a well-defined semiflow if (1) the projection of the local vector fields onto the graph factor, \( P_*(X|_{\{x\} \times \Upsilon}) \), satisfies Lemma 7 for each point \( x \) in the branch set of \( C \); and (2) the projections of the vector fields on the branch set to the \( R \)-factor are equal up to the attaching maps.

**Proof:** Off of the branch set, the space is a manifold and hence the vector field gives a well-defined flow. If \( p \) is a point on the branch line, condition (2) implies that the vector field is well-defined with respect to the attaching maps and the net effect in the \( R \)-factor is a drift in this direction. In the graph factor, condition (1) and the proof of Lemma 7 implies that there is a unique forward orbit through \( p \). \( \diamond \)

Heuristically, this condition means that, as in the case of a graph, the vector field must point “in” on all but one sheet of the configuration space in order to have well-defined orbits. We may thus lift the criteria of Lemma 7 to the product configuration space. All of the vector fields in this paper are so constructed.

**References**

[1] A. Abrams. *Configuration spaces of graphs and Brownian motion* (?). PhD thesis, UC Berkeley, 2000. In preparation.
[2] Y. A. Bozer and M. M. Srinivasan. Tandem configurations for automated guided vehicle systems and the analysis of single vehicle loops. *IIE Transactions*, 23(1):72–82, 1991.

[3] Y. Brave and M. Heymann. On optimal attraction of discrete-event processes. *Information Science*, 67(3):245–276, 1993.

[4] R. R. Burridge, A. A. Rizzi, and D. E. Koditschek. Sequential composition of dynamically dexterous robot behaviors. *Int. J. Rob. Res.*, (to appear).

[5] G. A. Castleberry. *The AGV Handbook*. Braun-Brumfield, Ann Arbor, MI, 1991.

[6] M. Erdmann. Understanding action and sensing by designing actions-based sensors. *Int. J. Rob. Res.*, 14(5):483–509, 1995.

[7] S. B. Gershwin. *Manufacturing Systems Engineering*. Prentice Hall, Englewood Cliffs, NJ, 1994.

[8] R. Ghrist. Configuration spaces and braid groups on graphs in robotics. To appear, *Braids, Links, and Mapping Class Groups: the Proceedings of Joan Birman’s 70th Birthday*, 2000.

[9] R. Ghrist. Configuration spaces of graphs. Preprint, 2000.

[10] R. Ghrist. Geometric optimization on configuration spaces of networks. Preprint, 2000.

[11] R. Ghrist and D. E. Koditschek. Safe cooperative robot dynamics on graphs. In Y. Nakayama, editor, *Eighth Intl. Symp. on Robotic Research*, pages 81–92. Springer-Verlag, 1998.

[12] D. E. Koditschek and E. Rimon. Robot navigation functions on manifolds with boundary. *Advances in Applied Mathematics*, 11:412–442, 1990.

[13] D. Lind and B. Marcus. *Introduction to Symbolic Dynamics and Coding Theory*. Cambridge, 1995.

[14] T. Lozano-Perez, M. T. Mason, and R. H. Taylor. Automatic synthesis of fine-motion strategies for robots. *Int. J. Rob. Res.*, 3(1):3–23, 1984.

[15] M. T. Mason. The mechanics of manipulation. In *Proc. International Conference on Robotics and Automation*, pages 544–548, March 1985.

[16] R. J. Milgram and S. Kaufman. Topological characterization of safe coordinated vehicle motion. In preparation, November 1999.

[17] J. R. Munkres. *Topology, A First Course*. Prentice Hall, 1975.

[18] R. Sengupta. An optimal control theory for discrete event control systems. *SIAM J. Control and Optimization*, 36(2):488–541, 1998.

[19] S. F. Smith. Reactive scheduling systems. In D. E. Brown and W. T. Schering, editors, *Intelligent Scheduling Systems*, pages 155–192. Kluwer Academic Publishers, Boston, MA, 1995.