Constructing tensegrity frameworks and related applications in multi-agent formation control
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Document Version
Publisher's PDF, also known as Version of record

Publication date:
2018

Link to publication in University of Groningen/UMCG research database

Citation for published version (APA):

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Chapter 5

Constructing universally rigid tensegrity frameworks with application in multi-agent formation control

Rigidity graph theory has found broad applications in engineering, architecture, biology, and chemistry, while systematic and computationally tractable construction of rigid frameworks is still a challenging task. In this chapter, starting from any given configuration in general positions, we show how to construct a universally rigid tensegrity framework by looking into the kernel of the tensegrity framework’s stress matrix. As one application, we show how to stabilize a formation of mobile agents by assigning a universally rigid virtual tensegrity framework for the formation and then design distributed controllers based on the forces determined by the stresses of the edges. Such formation controllers are especially useful when one needs to satisfy formation constraints in the form of strict upper or lower bounds on inter-agent distances arising from tethered robots.

5.1 Introduction

Rigidity graph theory has always been playing a key role in solving topology related problems in various fields, e.g. the formation control problem of multi-agent systems [15, 35, 78], the geometric analysis of molecular models in bio-chemistry [128], and the localization of wireless sensor networks [139]. A challenging problem concerned with rigidity is to determine whether a given framework is rigid (resp. globally rigid and universally rigid). It is already well known that the local rigidity can be guaranteed if the rank of the corresponding rigidity matrix exceeds some bound. However, for global rigidity and universal rigidity, the problem is shown to be NP-hard even in the lower dimensional space $\mathbb{E}^1$ [108]. Then, to make the problem more tractable, researchers concentrate on the frameworks with a generic configuration, where the nodal coordinates are algebraically independent over the rationals. Some exploring efforts along this line have been made in [1, 20, 22], where the sufficient conditions for a framework of generic configurations to be globally/universally rigid have been given, and the necessity of these conditions is later proved in [49]. Recently, a new kind of universal rigidity called ‘iterative...
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universal rigidity' has been addressed in [24], where the analysis of the rigidity of the framework is decomposed into a sequence of affine sets in the configuration space. One weaker condition compared with being generic for the configuration of the framework is general, which is preferable in practice since it is checkable in polynomial time, while not for the generic configuration [4]. It has been verified that the genericity assumption of the configuration to ensure universal rigidity can be relaxed to the situation of general positions in [3]. In addition, the universal rigidity of one-dimensional framework in general positions with a complete bipartite underlying graph is fully discussed in [63].

Of particular theoretical and practical interest is tensegrity frameworks. In recent years, they have found broad applications in engineering, architecture, biology, and arts because of their superior features, such as deployability, deformability and robustness [76, 123]. However, little research has been conducted to construct universally rigid tensegrity frameworks from some given geometric shape described by the overall configuration. One of the well-established findings is that for a framework in the shape of convex polygons in the plane, the universal rigidity can be obtained if the boundary and the interior members are respectively set to be cables and struts [19]. For a class of so-called Grünbaum frameworks, the construction for universally rigid tensegrity frameworks in two and three dimensions is studied in [66], in which the approach strongly relies on the computation of the convex hull. If the underlying graph of the framework is given beforehand, a purification based algorithm is designed to compute the corresponding stresses for the \((d + 1)\)-lateration frameworks in general positions [4]. In the case that only the configuration is known, to create a universally rigid tensegrity framework, the underlying graph together with the type of members are determined through the algorithm in [88], while it is highly likely to result in a complete graph.

For control engineers, the issue of identifying and designing rigid frameworks are particularly relevant for the formation control problem of teams of mobile agents. The stable tensegrity frameworks, due to their strong robustness and clear physical interpretation, have been used as the virtual multi-agent framework to understand better the behavior of the agents around the equilibrium corresponding to the prescribed formation shape. Tensegrity frameworks, due to their robustness and scalability, have been employed as the virtual framework to solve the formation control problem of multi-agent systems. Starting from one-dimensional space, i.e., a line, [93] introduced a tensegrity-based control law that can exponentially stabilize the agents with prescribed distances. Then the same idea was used to deal with the problem in higher-dimensional space by collinear projections. In [72], the model of an unmanned aerial vehicle was integrated with a virtual cross-tensegrity framework, based on which a decentralized control strategy was designed such that a scalable formation was achieved. As a direct application, a stress-based formation control scheme is proposed to stabilize “affine” formations in [78]. In contrast
to a rigid formation, an affine formation allows more transformations besides translation and rotation, such as scaling, shearing and reflection. But most of the existing results are restricted to special class of frameworks, e.g., one-dimensional tensegrity framework [93] or cross-tensgrities that are rectangles with four boarder cables and two crossing struts [71, 72].

Motivated by the recent advances in rigidity graph theory, it is the goal of this chapter to first design an algorithm to construct a universally rigid tensegrity framework for any given configurations in general positions. Taking into account the engineering and theoretical concerns that sparse frameworks are more desirable because of extendibility, complexity and computation, we will focus on building tensegrity frameworks with fewer members through constructing stress matrices with close to maximal allowed number of zeros. We then use the constructed universally rigid tensegrity framework for multi-agent formation control. Distributed control laws are designed to stabilize formations when their topologies are implemented using the virtual tensegrity framework. So the main contribution of this chapter is two-fold. First, we develop a systematic algorithm to construct universally rigid tensegrity frameworks. Such algorithms have not been reported in the literature. Second, we apply the virtual tensegrity frameworks to distributed formation control, which therefore enables us to have a clear intuitive estimate of the domain of attraction around the system’s equilibrium. This application is particularly useful for the emerging cooperative control of tethered robots, where the challenging formation constraints, such as the strict maximum or minimum inter-agent distances, could be incorporated in the virtual tensegrity framework.

The rest of this chapter is organized as follows. In Section 5.2, we propose our algorithm to construct universally rigid tensegrity frameworks for any given configuration in general position. The application for formation control is discussed in Section 5.3. Some illustrative examples are presented in Section 5.4. Section 5.5 concludes this chapter.

### 5.2 Constructing universally rigid tensegrity frameworks

We first provide the steps of the algorithm in detail and then show the constructed tensegrity frameworks are in general close to minimal by giving an upper bound of the numbers of their members.

#### 5.2.1 Algorithm

We assume the given configuration \( q^* \in \mathbb{R}^{d \times n} \) is in general position. Define the extended configuration matrix \( Q^* \triangleq [(q^*)^T, 1_n]^T \in \mathbb{R}^{(d+1) \times n} \). Then, it follows that
every \((d + 1) \times (d + 1)\) submatrix of \(Q^*\) has full rank. From the definition of the stress matrix \(\Omega\) and (2.5), one can check that matrix \(\Omega\) always lives in the null space of \(Q^*\), i.e.

\[
Q^*\Omega = 0_{(d+1) \times n}.
\] (5.1)

Given \(q^*\), the key component of the algorithm is to determine matrix \(\Omega\), which determines in turn which two nodes are connected together with their nonzero stress. Obviously, such \(\Omega\) is in general not unique and naturally we want to obtain an \(\Omega\) with more zeros which leads to fewer members and thus lower complexity. Towards this end, we convert our problem into the sparse null space problem first considered in [96], namely, given a matrix \(A\), to find a sparse matrix \(B\) such that

\[
B\text{ is full rank and its column span is null (}A\text{)}.
\] (5.2)

In view of Lemmas 2.9, to obtain a universally rigid tensegrity framework, matrix \(\Omega\) is required to be positive semi-definite with rank \(n - d - 1\). However, since \(\Omega\) in (5.1) is not full rank, we cannot directly solve the sparse null space problem. Instead, we try to construct a column full-rank matrix \(D \in \mathbb{R}^{n \times (n - d - 1)}\) such that

\[
Q^*D = 0_{(d+1) \times (n-d-1)},
\] (5.2)

where \(D\) is a Gale matrix of \(q^*\) [3]. If indeed such a \(D\) can be constructed, it must be true that

\[
Q^*DD^T = 0_{(d+1) \times (n-d-1)}D^T = 0_{(d+1) \times n},
\] (5.3)

and hence the matrix \(DDD^T\) can serve as the stress matrix \(\Omega\). So the construction of an \(\Omega\) is equivalent to the design of such a sparse \(D\). In addition, for computational efficiency, we make an even stronger requirement that \(\Omega\) is in its band form, whose non-zero entries are confined to be in a diagonal band containing the main diagonal. Now we present our 5-step algorithm to construct the universally rigid framework with the stress matrix \(\Omega\), which is inspired by the classical “turning back” method for computing the sparse null space basis [48].

Step 1: Arrange matrix

\[
Q^* = [(q^*)^T, 1_n]^T \in \mathbb{R}^{(d+1) \times n}.
\]

Step 2: Locate the nonzero elements of \(D\). We first find the smallest \(k_1 > 0\) such that \(Q^*\)'s columns with the indices \(d+2, d+1, \ldots, d+2-k_1\) are linearly dependent. We then set the nonzero elements of \(D\)'s first column to be located at the positions \(d+2-k_1\) through \(d+2\). Then, in order to record the positions of the nonzero elements of the second column of \(D\), we find the smallest \(k_2 > 0\) such that those columns with the indices \(d+3, d+2, \ldots, d+3-k_2\), excluding \(d+2-k_1\), of \(Q^*\) are linearly dependent. Again, the indices correspond to the nonzero elements’ positions of \(D\)'s second column. We repeat this procedure until we have determined the positions of the nonzero elements of the last column of \(D\). Note that the configuration \(q^*\) is in general position, there are
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exactly \( d + 2 \) nonzero elements in each column of \( D \), yielding

\[
D = \begin{pmatrix}
1 & 2 & \cdots & n - d - 1 \\
* & 0 & \cdots & 0 \\
* & * & \cdots & 0 \\
\vdots & \vdots & \ddots & \vdots \\
* & * & \cdots & 0 \\
0 & * & \cdots & * \\
\vdots & \vdots & \ddots & \vdots \\
0 & 0 & \cdots & * \\
\end{pmatrix}
\]

(5.4)

where ‘*’ indicates a nonzero element.

Step 3: Now we compute the values of the nonzero elements of \( D \) by solving the following equation

\[
Q^* D = 0_{(d+1) \times (n-d-1)},
\]

(5.5)

which is underdetermined since it is a set of \( d + 1 \) linear equations with \( d + 2 \) unknowns. Hence, we can always find a set of nonzero elements of \( D \) and thus fully determine \( D \). In addition, it is easy to check that the constructed \( D \) is always column full-rank.

Step 4: The stress matrix is then the positive semidefinite matrix \( \Omega = DD^T \) whose rank is \( n - d - 1 \). The obtained \( \Omega \) is a square, symmetric, band matrix with one upper triangular and one lower triangular \( (n - d - 2) \)-dimensional submatrix in its upper right and lower left corners respectively. We write \( \Omega \) in the notation below where all the determined zero elements are denoted by zero and all the other elements that may or may not be zero are denoted by ‘×’:

\[
\Omega = \begin{pmatrix}
\times & \cdots & \times & 0 & \cdots & \cdots & 0 \\
\times & \cdots & \times & \times & \ddots & \ddots & \vdots \\
\vdots & \vdots & \ddots & \ddots & \ddots & \ddots & \vdots \\
\times & \cdots & \times & \cdots & \cdots & \times & \times \\
\vdots & \vdots & \vdots & \ddots & \ddots & \ddots & \vdots \\
0 & \times & \cdots & \times & \cdots & \cdots & \times \\
\vdots & \ddots & \ddots & \ddots & \ddots & \ddots & \vdots \\
\vdots & \ddots & \ddots & \ddots & \ddots & \ddots & \vdots \\
0 & \cdots & 0 & \times & \cdots & \times & \times \\
\end{pmatrix}
\]

(5.6)

\[ n \times n \]
Step 5: According to $\Omega$, assign cables and struts to the $n$-node framework: For each $\Omega_{ij} < 0$, we assign a cable between nodes $i$ and $j$ and for each $\Omega_{ii} > 0$, we assign a strut between nodes $i$ and $j$. The stresses of the assigned cables and struts are $\omega_{ij} = -\Omega_{ii}$. The desired tensegrity framework is then obtained.

Now we prove that the constructed tensegrity framework is indeed universally rigid.

**Theorem 5.1.** Given a configuration $q^*$ in general position, the proposed 5-step algorithm returns a universally rigid tensegrity framework with the geometric shape prescribed by $q^*$.

**Proof.** The constructed tensegrity framework has the stress matrix $\Omega$ which, as shown in step 4, is positive semi-definite and has rank $n - d - 1$. So the conditions (1) in Lemma 2.9 is satisfied. Recalling that the given configuration $q^*$ is in general position, so the obtained tensegrity framework is again universally rigid because of Lemma 2.9.

**Remark 5.2.** In step 4, the stress matrix $\Omega$ can be more generally determined by $\Omega = D\Psi D^T$, where $\Psi$ is a nonsingular, $(n - d - 1) \times (n - d - 1)$, symmetric matrix, used to adjust the magnitudes of the stresses of the members of the constructed tensegrity framework. In this chapter, we set $\Psi = I_{n-d-1}$ for simplicity.

In practice, one usually prefers fewer edges to reduce the complexity, if possible, in a rigid framework. Hence, a natural question to ask is whether the tensegrity framework obtained by the proposed algorithm is indeed structurally simple. To address this question, we construct an upper bound of the number of members of the obtained tensegrity framework, namely $|E|$ of $(G, q^*)$, which is roughly $nd$ when $n$ is big.

### 5.2.2 Upper bound of $|E|$}

We look into the number of nonzero elements of $\Omega$ in (5.6). The densest $\Omega$, namely that contains the largest possible number of nonzero elements, appears when all the elements denoted by ‘$\times$’ are nonzero. Then the number of off-diagonal zero elements in $\Omega = DD^T$ is

$$2((n - d - 2) + (n - d - 3) + \cdots + 1) = (n - d - 1)(n - d - 2),$$

or equivalently the number of off-diagonal nonzero elements in $\Omega$ is

$$n(n - 1) - (n - d - 1)(n - d - 2) = (d + 1)(2n - d - 2),$$
which, when divided by 2, is exactly the number of members to be inserted into the tensegrity framework in view of Step 5 of the algorithm. Therefore, this densest $\Omega$ gives an upper bound $(d + 1) \left( n - \frac{d + 2}{2} \right)$.

So we have proved the following theorem.

**Theorem 5.3.** The number of members of the constructed tensegrity framework is bounded from above by

$$|E| \leq (d + 1) \left( n - \frac{d + 2}{2} \right).$$

(5.7)

**Remark 5.4.** One necessary condition for a tensegrity framework in generic configurations (with $n$ larger than $d + 1$) to be globally rigid or universally rigid is that it has to be rigid with at least one self-stress state, which in turn implies that it needs to have at least $nd - d(d + 1)/2 + 1$ members. This is the lower bound of the number of members required to construct a globally rigid or universally rigid tensegrity framework in generic configurations. The lower bound differs from the constructed upper bounded roughly by $n$ when $n$ is big.

**Remark 5.5.** The upper bound of the number of members presented in (5.7) corresponds to the number of members of universally rigid tensegrity frameworks constructed on $(d+1)$-tree graphs [4]. The construction of universally rigid Grünbaum frameworks in generic configurations with the minimal number of edges in 2D and 3D are investigated in [66], where it is shown that the two-dimensional Grünbaum frameworks in nongeneric configurations are also universally rigid. The problem on how to compute the stress matrix has also been considered in [88], which, however, most likely yields stress matrices without any zero elements. In comparison, our algorithm in general always returns a stress matrix with close to maximal allowed number of zeros.

In the next section, we show how universally rigid tensegrity frameworks can be used as virtual structures to help the design of distributed formation controllers for teams of mobile agents.

### 5.3 Formation stabilization

#### 5.3.1 Formation Control Problem

We consider a group of $n$ mobile agents, each of which is modeled by a kinematic point

$$\dot{q}_i = u_i, \quad i = 1, 2, \cdots, n,$$

(5.8)

where $q_i \in \mathbb{R}^d$ represents agent $i$’s position and $u_i \in \mathbb{R}^d$ is its control input. The neighbor relationships will be designed and characterized by an undirected graph
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$G$ with the vertex set $V$ and the edge set $E$.

The formation stabilization problem is that given a desired configuration $q^* = [q_1^*, \ldots, q_n^*] \in \mathbb{R}^{d \times n}$ for this team of $n$ agents (5.8), design the neighbor relationship graph $G$ for the team and correspondingly, for each agent $i = 1, \ldots, n$, design distributed control laws $u_i(q_i - q_j, q_i^* - q_j^*)$, $j \in \mathcal{N}_i$, such that the agents’ positions are driven to the target set

$$\mathcal{T} = \{q \in \mathbb{R}^{dn} | q_i - q_j = q_i^* - q_j^*, \quad \forall (i, j) \in \mathcal{E}\}. \quad (5.9)$$

Note that here $q^*$ is given in an arbitrary coordinate system of choice. When each agent has its own coordinate system that may differ from each other, $q^*$ is then given to each agent in its own coordinate system, in which case different agents’ given $q^*$ differ up to some congruent transformation of translation and rotation.

Ample previous work has discussed how to solve this problem locally when $(G, q^*)$ is rigid and the controllers are derived by attaching virtual springs to the agents. To make the controllers simpler, $G$ is usually required to contain as few edges as possible. In what follows, we use the virtual tensegrity framework constructed in Section 5.2 to describe the necessary sensings between the agents and the resulted distributed controllers whose gains are derived from the stresses.

### 5.3.2 Controller design and stability analysis

We use the universally rigid tensegrity framework constructed previously to determine which agents need to sense which other agents. To be specific, from the given $q^*$, we run the 5-step algorithm and obtain a framework $(G, q^*)$. Then the underlying graph $G$ is used to represent the sensing graph of the $n$-agent team.

To calculate the virtual forces utilizing the stresses, we set the rest length of each edge to be

$$l_{ij} = \gamma_{ij}\|r_{ij}^*\| = \begin{cases} \gamma_{ij}^c\|r_{ij}^*\| & \text{if } \omega_{ij} > 0, \\ \gamma_{ij}^s\|r_{ij}^*\| & \text{if } \omega_{ij} < 0, \end{cases} \quad (5.10)$$

where $\gamma_{ij}^c \in (0, 1)$ and $\gamma_{ij}^s \in (1, +\infty)$ are constants as design parameters, and $r_{ij}^*$ is the prescribed relative position of agent $j$ with respect to agent $i$, i.e., $r_{ij}^* = q_j^* - q_i^*$. Then the formation control gain $k_{ij}$ for agent $i$ with respect to its neighbor $j$ is given by

$$k_{ij} = \frac{\omega_{ij}}{1 - \gamma_{ij}}, \quad \forall (i, j) \in \mathcal{E}. \quad (5.11)$$

For convenience in notation, we define the auxiliary variable

$$z_i = q_i - q_i^*. \quad (5.12)$$

Let $z \in \mathbb{R}^{dn}$ be the vector obtained by stacking all the $z_i$ together.
5.3. Formation stabilization

Now, we define the set $\mathcal{D}(0)$ as

$$
\mathcal{D}(0) \triangleq \{z(0) \in \mathbb{R}^n \mid \|z_i(0) - z_j(0)\| < \text{sign}(\omega_{ij})(\|r_{ij}^*\| - l_{ij}), \\
\forall (i, j) \in \mathcal{E}\}.
$$

(5.13)

To proceed, we assume all the agents are initially located in the set $\mathcal{D}(\epsilon)(0)$ defined as

$$
\mathcal{D}(\epsilon)(0) \triangleq \{z(0) \in \mathbb{R}^n \mid \|z_i(0) - z_j(0)\| \leq \text{sign}(\omega_{ij})(\|r_{ij}^*\| - l_{ij}) - \epsilon, \\
\forall (i, j) \in \mathcal{E}\},
$$

(5.14)

where $\epsilon$ is a small positive number.

The potential function for agent $i$ is defined as

$$
P_i(z(t)) \triangleq \sum_{\omega_{ij} < 0} k_{ij} \left[ (l_{ij} - \|r_{ij}^*\|)^\alpha - \rho_{ij}^\alpha \right]^{\beta} \\
+ \sum_{\omega_{ij} > 0} k_{ij} \left[ (\|r_{ij}^*\| - l_{ij})^\alpha - \varrho_{ij}^\alpha \right]^{\beta},
$$

(5.15)

where

$$
\rho_{ij} = l_{ij} - \|r_{ij}^*\| - \|r_{ij}(t) - r_{ij}^*\|,
$$

$$
\varrho_{ij} = \|r_{ij}^*\| - l_{ij} - \|r_{ij}(t) - r_{ij}^*\|,
$$

(5.16)

with $r_{ij} = q_j - q_i$ being the relative position between agents $i$ and $j$. $\alpha > 2$ and $0 < \beta < 1$ are the positive exponents. Note that in the set $\mathcal{D}(\epsilon)(0)$, $\rho_{ij}$ and $\varrho_{ij}$ are both small positive numbers. The parameters $k_{ij}$ are chosen such that

$$
P_0 = \begin{cases} 
  k_{ij} (\|r_{ij}^*\| - l_{ij})^{\alpha\beta}, & \text{if } \omega_{ij} > 0, \\
  k_{ij} (l_{ij} - \|r_{ij}^*\|)^{\alpha\beta}, & \text{if } \omega_{ij} < 0,
\end{cases}
$$

(5.17)

where $P_0$ is an arbitrary positive pre-defined constant. All the parameters $k_{ij}$ would be determined if any one of them is decided with given $\alpha$ and $\beta$.

Correspondingly, the potential function for the whole system is given by

$$
P(z(t)) = \frac{1}{2} \sum_{i=1}^n P_i(z(t))
$$

(5.18)

The control input of agent $i$ is

$$
\dot{q}_i = \dot{z}_i = u_i = -\nabla_z P_i(z(t))
$$

(5.19)
Proposition 5.6. For any given initial position \( q(0) \in D^e(0) \), the set \( D(0) \) is invariant for the system (5.8) under the control law (5.19).

Before presenting the proof for Proposition 5.6, we first analyse the properties of the potential function (5.15). Consider the numerators of (5.15), where \( \delta_s \) (resp. \( \delta_c \)) are regarded as independent variables. Define function \( f(x) \) as

\[
f(x) = k_f(c_f^\alpha - x^\alpha)^\beta > 0, \quad k_f > 0, x \in (0, c_f],
\]

which coincides with the form of the numerators of the potential function. The first order derivative of \( f(x) \) satisfies

\[
f'(x) = -\alpha \beta k_f x^{\alpha-1} (c_f^\alpha - x^\alpha)^\beta < 0, \quad x \in (0, c_f],
\]

which implies that the function \( f(x) \) is monotonically decreasing with respect to \( x \). Besides this, for sufficiently small \( x_s \), \( f'(x_s) \to 0 \), due to \( x_s^{\alpha-1} \to 0 \), when \( \alpha > 2 \).

Look at the second order derivative of \( f(x) \),

\[
f''(x) = -\alpha(\alpha-1)\beta k_f x^{\alpha-2} (c_f^\alpha - x^\alpha)^{\beta-1}
+ \alpha^2 \beta(\beta-1)k_f x^{2(\alpha-1)}(c_f^\alpha - x^\alpha)^{\beta-2}.
\]

It follows from \( \alpha > 2 \) and \( 0 < \beta < 1 \) that \( f'' < 0, \forall x \in (0, c_f] \). Together with the fact that \( f'(x) < 0 \), it implies that the decreasing rate of \( f(x) \) increases as \( x \) grows.

Consider

\[
f(x) = f(0) + \int_0^x f'(s)ds = f(0) - \int_0^x |f'(s)|ds, \quad x \in (0, c_f].
\]

It can be seen that \( f'(x) \) drops rapidly, when the value of \( x \) increases.

Based on \( f(x) \), together with (5.15), we introduce the function \( P_f(x) \) as

\[
P_f(x) = \frac{f(x)}{x}, \quad x \in (0, c_f].
\]

The first order derivatives of \( P_f(x) \) satisfies

\[
P'_f(x) < 0.
\]

It can be seen that \( P_f(x) \) has the same monotonicity with the potential function in (5.15) for a single edge. Hence, the potential energy (5.15) is monotonically decreasing with respect to \( \delta_s \) and \( \delta_c \). This property implies that the closer the edges approach their rest lengths, the higher their potential energy will become.

Now, we give the proof of Proposition 5.6, which is partially motivated by [62].
Proof of Proposition 5.6. Recall the control input (5.19) for agent $i$

$$u_i = -\nabla z_i P_i(z(t)).$$

(5.26)

Hence the time derivative of $P_i(t)$ satisfies,

$$\dot{P}_i(z(t)) = \nabla z_i P_i(z(t))^T \dot{z}_i(t)$$

$$= - [\nabla z_i P_i(z(t))]^T [\nabla z_i P_i(z(t))] \leq 0,$$

(5.27)

which implies that the potential energy associated with agent $i$ is not increasing. However, the potential energy stored in some edge, say $(i, \nu)$ with negative stress, might increase, even though $P_i(z(t))$ is not increasing. This means that the length of $(i, \nu)$ tends to its rest length $l_{i\nu}$ and the other edges are driven far away from their rest lengths. Since the potential function for every edge defined in (5.15) is monotonically decreasing, the potential energy of edge $(i, \nu)$ will increase when $\|r_{i\nu}\|$ approaches to its rest length, i.e., $l_{i\nu}$. Meanwhile, since the energy generated by the single edge $(i, \nu)$ is less than that generated by all the edges that connecting agent $i$, by letting $P_i(z(t)) = P_{i\nu}(t)$, we can get the maximum length for edge $(i, \nu)$. Then, it can be checked whether $\|r_{i\nu}\|$ reaches its rest length $l_{i\nu}$.

Now, we consider the length changes of the edges. Since the potential energy $P_i(z(t))$ is not increasing, we have

$$P_{i\nu}(t) + [P_i(z(t)) - P_{i\nu}(t)] \leq P_i(z(0)), \quad t \geq 0,$$

(5.28)

where $P_{i\nu}(t)$ denotes the potential energy of edge $(i, \nu)$ stored at time $t$. $P_i(z(t))$ is the potential energy of all the neighbor edges of agent $i$. Obviously, we have $P_i(z(t)) > P_{i\nu}(t)$. Then, from (5.28), it is straightforward to check

$$P_{i\nu}(t) \leq P_i(z(0)) - [P_i(z(t)) - P_{i\nu}(t)] < P_i(z(0)).$$

(5.29)

Hence, to obtain the upper bound of $\|r_{i\nu}\|$, it is assumed that

$$P_{i\nu}(t) = P_i(z(0)).$$

(5.30)

According to (5.25), the potential energy $P_i(z(t))$ would increase if the length of the edges reach closer to their rest lengths. Hence, the maximum of $P_i(z(0))$ is obtained when the initial positions of the edges satisfy

$$\|z_i(0) - z_j(0)\| = \text{sign}(\omega_{ij})(\|r_{ij}^*\| - l_{ij}) - \epsilon, \quad \forall (i, j) \in \mathcal{E}.$$  

(5.31)
Consequently,

\[
P_i(z(0))_{\text{max}} = \sum_{\omega_{ij} < 0} \frac{k_{ij} [(l_{ij} - \|r_{ij}^*\|)^{\alpha} - \epsilon^{\alpha}]^\beta}{\epsilon} + \sum_{\omega_{ij} > 0} \frac{k_{ij} [(\|r_{ij}^*\| - l_{ij})^{\alpha} - \epsilon^{\alpha}]^\beta}{\epsilon}. \tag{5.32}
\]

Then, we let

\[
P_{i\nu} = \frac{k_{i\nu} [(l_{i\nu} - \|r_{i\nu}^*\|)^{\alpha} - \rho_{i\nu}^{\alpha}]^\beta}{l_{i\nu} - \|r_{i\nu}^*\| - \|r_{i\nu} - r_{i\nu}^*\|} = P_i(z(0))_{\text{max}}. \tag{5.33}
\]

So far, the \(\|r_{i\nu}\|\) can be derived from (5.32)-(5.33) theoretically. However, it can be seen that the resulting \(\|r_{i\nu}\|\) depends on the other edges’ rest lengths and desired lengths, which is difficult to determine whether edge \((i, \nu)\) reaches its rest length or not. Therefore, we further look for another relationship on the potential energy between the edge \((i, \nu)\) and the other edges.

According to the statement above, to check whether \(\|r_{i\nu}\|\) reaches the rest length, only the potential energy of edge \((i, \nu)\) increases, i.e., only the edge \((i, \nu)\) changes towards its rest length, and the other edges move away from their rest lengths. This implies

\[
0 < \rho_{i\nu}(t) \triangleq \epsilon_1 < \epsilon, \tag{5.34}
\]

where \(\epsilon_1\) is a sufficiently small number. Now, we only take the numerators of (5.15) into consideration.

For two different edges, consider two functions

\[
f_1(x) = k_{f1}(c_{f1}^{\alpha} - x^{\alpha})^\beta, \quad k_{f1} > 0, \quad x \in (0, \epsilon], \tag{5.35}
\]

and

\[
f_2(x) = k_{f2}(c_{f2}^{\alpha} - x^{\alpha})^\beta, \quad k_{f2} > 0, \quad x \in (0, \epsilon]. \tag{5.36}
\]

Given sufficiently small number \(\epsilon_1\), we have

\[
f_1(\epsilon_1) = f_1(0) + \int_0^{\epsilon_1} f_1'(x)dx. \tag{5.37}
\]

Since the function \(f_1(x)\) is continuous and differentiable in \((0, \epsilon)\), it follows from mean value theorem that

\[
f_1(\epsilon_1) = f_1(0) + \epsilon_1 f_1'(s_1), \quad s_1 \in (0, \epsilon_1). \tag{5.38}
\]
Similarly, for $\epsilon > \epsilon_1$, it follows
\begin{align*}
f_2(\epsilon) &= f_2(0) + \int_0^{\epsilon} f'_2(x)dx \\
&= f_2(0) + \int_0^{\epsilon_1} f'_2(x)dx + \int_{\epsilon_1}^{\epsilon} f'_2(x)dx \\
&= f_2(0) + \epsilon_1 f'_2(s_2) + (\epsilon - \epsilon_1) f'_2(s_3),
\end{align*}
where $s_2 \in (0, \epsilon_1)$ and $s_3 \in (\epsilon_1, \epsilon)$.

From (5.11), the initial values satisfy
\[ f_1(0) = f_2(0) = P_0. \] (5.40)

And, from (5.21), for a sufficiently small $\epsilon_1$, under the condition that $\alpha > 2$, we have
\[ f'_1(s_1) = f'_2(s_2) \to 0. \] (5.41)

Considering the fact that $f''(x) < 0$ in (5.22), it yields
\[ f'_2(s_3) < f'_2(s_2) < 0. \] (5.42)

Combining (5.38)-(5.42), we obtain
\[ f_2(\epsilon) < f_1(\epsilon_1), \] (5.43)
which implies
\[ f_{ij}(\epsilon) < f_{i\nu}(\rho_{i\nu}), \quad j \in \mathcal{N}_i. \] (5.44)

In view of (5.32)-(5.34), we know that the edge $(i, \nu)$ satisfies
\begin{align*}
k_{ij} \left[ (\text{sign}(\omega_{ij})(||r^*_{ij}|| - l_{ij}))^\alpha - \epsilon^\alpha \right]^\beta \\
< k_{i\nu} \left[ (l_{i\nu} - ||r^*_{i\nu}||)^\alpha - \rho_{i\nu}(t)^\alpha \right]^\beta, \quad j \in \mathcal{N}_i.
\end{align*}
(5.45)

Taking (5.32)-(5.33) and (5.45) into consideration, we have
\[ \frac{1}{l_{i\nu} - ||r^*_{i\nu}|| - ||r_{i\nu}(t) - r^*_{i\nu}||} < \frac{|\mathcal{N}_i|}{\epsilon}, \] (5.46)
where $|\mathcal{N}_i|$ denotes the cardinality of the set $\mathcal{N}_i$, i.e., the number of agent $i$’s neighbors. Then, through simple calculation, we get
\[ ||r_{i\nu}(t) - r^*_{i\nu}|| \leq l_{i\nu} - ||r^*_{i\nu}|| - \frac{\epsilon}{|\mathcal{N}_i|}. \] (5.47)
Note that $z_{i\nu}(t)$ satisfies
\[ \|z_{i\nu}(t)\| = \|r_{i\nu}(t) - r^*_{i\nu}\| \geq \|r_{i\nu}(t)\| - \|r^*_{i\nu}\|. \] (5.48)

Then, the upper bound for $\|r_{i\nu}(t)\|$ can be derived from (5.47)-(5.48)
\[ \|r_{i\nu}(t)\| \leq l_{i\nu} - \frac{\epsilon}{|N_i|}, \] (5.49)
which implies the length of edge $(i, \nu)$ will not reach its rest length if it starts from $D^\epsilon(0)$.

For the edge $(i, \zeta)$ with positive stress, we can prove similarly that
\[ \|r_{i\zeta}(t)\| \geq l_{i\zeta} + \frac{\epsilon}{|N_i|}. \] (5.50)

Therefore, we can draw the conclusion from (5.49) and (5.50) that the edges will never escape from the set $D(0)$ during the evolution, if the edges are initially located in $D^\epsilon(0)$, namely, the set $D(0)$ is an invariant set. This completes the proof. 

Remark 5.7. Proposition 5.6 indicates that using the control law (5.19) the agents will not escape the set $D(0)$ if they start from $D^\epsilon(0)$, which implies there will always be virtual forces along the edges. This further implies no sensing breakdown will happen since the virtual edges represent the sensing relationships between the agents.

Now, we are ready to present our main theorem.

Theorem 5.8. For any given initial position $q(0) \in D^\epsilon(0)$, the formation stabilization of the networked single-integrator systems modeled by (5.8) is achieved by the controller (5.19).

Proof. From the control law (5.19) and potential function (5.15), we have
\[ u_i = -\left( \sum_{\omega_{ij} < 0} \frac{\phi_{ij}}{\rho_{ij}^2\|r_{ij} - r^*_{ij}\|} + \sum_{\omega_{ij} > 0} \frac{\varphi_{ij}}{\varrho_{ij}^2\|r_{ij} - r^*_{ij}\|} \right) (z_i - z_j), \] (5.51)
where
\[ \phi_{ij} = k_{ij} \left( \alpha \beta \rho_{ij}^\alpha \left[ (l_{ij} - \|r_{ij}\|)^\alpha - \rho_{ij}^\alpha \right]^{\beta-1} + \left[ (l_{ij} - \|r^*_{ij}\|)^\alpha - \rho_{ij}^\alpha \right]^{\beta} \right), \]
\[ \varphi_{ij} = k_{ij} \left( \alpha \beta \varrho_{ij}^\alpha \left[ (\|r_{ij}\| - l_{ij})^\alpha - \varrho_{ij}^\alpha \right]^{\beta-1} + \left[ (\|r^*_{ij}\| - l_{ij})^\alpha - \varrho_{ij}^\alpha \right]^{\beta} \right), \]
with $\rho_{ij}$ and $\varrho_{ij}$ being defined in (6.4). We further write the closed-loop system
(5.8) and (5.15) into their compact form
\[ \dot{\vec{q}} = \dot{z} = - (L_w(l, \vec{q}, \vec{q}^*) \otimes I_d) z, \] (5.52)
where \( \vec{q} = [q_1^T, \ldots, q_n^T]^T \in \mathbb{R}^{dn}, \ \vec{q}^* = [(q_1^*)^T, \ldots, (q_n^*)^T]^T \in \mathbb{R}^{dn}, \) and \( L_w = H^T W(l, \vec{q}, \vec{q}^*) H \) is \( G \)'s weighted Laplacian matrix. \( H \) is \( G \)'s incidence matrix defined in (2.2.1), and the diagonal weight matrix \( W(l, \vec{q}, \vec{q}^*) \) is
\[ W(l, \vec{q}, \vec{q}^*) = \text{diag}(\{w_\iota\}), \ \iota = 1, 2, \ldots, |E|, \] with
\[ w_\iota = \begin{cases} \phi_\iota \rho_\iota^2 \|r_\iota - r_\iota^*\|, & \text{if } \omega_\iota < 0, \\ \varphi_\iota \rho_\iota^2 \|r_\iota - r_\iota^*\|, & \text{if } \omega_\iota > 0. \end{cases} \] (5.53)

Given \( q(0) \in D^c(0) \), it follows from Proposition 5.6 that \( L_w \) in (5.52) is well defined and positive semi-definite [62]. Also, the interaction graph of the agents is connected due to the fact that the edges are always in tension or in compression. Then, it can be concluded from Lemma 2.1 that
\[ \lim_{t \to \infty} z_1(t) = \lim_{t \to \infty} z_2(t) = \cdots = \lim_{t \to \infty} z_n(t), \] (5.54)
which implies that
\[ \lim_{t \to \infty} r_{ij}(t) = r_{ij}^*, \ \forall (i, j) \in \mathcal{E}. \] (5.55)

Thus, the prescribed formation is achieved. This completes the proof. \( \square \)

Remark 5.9. Even though the result is in the sense of local stability, we can enlarge the stability region by choosing small \( \gamma_{ij}^c \) for edges with positive stress and large \( \gamma_{ij}^s \) for those with negative stress. The definition of \( D^c(0) \) also describes clearly an estimate of the domain of attraction of the desired equilibrium of the closed-loop system.

Remark 5.10. Note that the configuration of the desired formation coincides with the given configuration \( q^* \) in (5.1) for designing universally rigid tensegrity framework, we thus can stabilize the proper formation consisting of any arbitrary number of agents theoretically. When one or more agents encounter mechanical failures, to stabilize the rest, the weights \( k_{ij} \) in control law (5.51) are required to be updated by recalculating the stresses \( \omega_{ij} \) based on the proposed algorithm, where the matrix \( Q^* \) needs to be altered via removing the failure agents' configuration. The resultant formation will remain the same as the original one associated with the agents without failure.
Remark 5.11. From (5.55), we know the equilibrium configuration is
\[ \lim_{t \to \infty} q(t) = q^* + \gamma \text{span}(1_d), \]
where \(\gamma\) is an arbitrary real number. Indeed, the stabilized configuration \(q\) is a class of affine transformation of \(q^*\) [23], which is still in equilibrium with the stress \(\omega\), i.e.,
\[ \sum_{j \in \mathcal{N}_i} \omega_{ij}(q_i - q_j) = \sum_{j \in \mathcal{N}_i} \omega_{ij}(q^*_i - q^*_j) = 0, \quad i = 1, \ldots, n. \]

Remark 5.12. In this chapter, the single integrator model is taken into account under the assumption that the velocities of the agents can be controlled directly. In practice, the signals generated by the system can serve as the commanded 3D-velocity for the tracking controllers of the quadrotors [59, 81] or vessels [44]. This technique has been employed in experimental setups in formation and motion control [42].

5.4 Simulation results

This section gives the simulation results to validate the effectiveness of the theoretical results derived in the preceding sections. For the construction of universally rigid tensegrity frameworks, we consider a general as well as generic configuration. Then, the formation stabilization algorithm is simulated on the resultant universally rigid tensegrity framework.

5.4.1 Construction of universally rigid tensegrity frameworks

The generic configuration is given as follows
\[
q^* = \begin{bmatrix}
-\sqrt{3}/2 & \sqrt{3}/2 & 0 & -1/2 & 1 & -2/3 \\
-1/2 & -1/2 & 1 & -1/2 & 0 & 8/3 - \sqrt{3} \\
0 & 0 & 0 & 3 & 3 & 3\sqrt{3} - 2
\end{bmatrix}.
\] (5.56)

With this configuration, the corresponding geometric shape of the six nodes is shown in Fig. 5.1. Based on our proposed algorithm, one solution of the Gale matrix \(D\) of the Gale matrix \(D\) derived from (5.5) is given as follows
\[
D = \frac{1}{3} \begin{bmatrix}
2\sqrt{3} & -\sqrt{3} & -\sqrt{3} & -3 & 3 & 0 \\
0 & 3 & 3\sqrt{3} - 8 & -\sqrt{3} & 2 - 2\sqrt{3} & 3
\end{bmatrix}^T.
\]
Then the stress matrix $\Omega$ can be calculated via $\Omega = DD^T$:

$$\Omega = \frac{1}{9} \begin{bmatrix} 12 & -6 & -6 & -6\sqrt{3} & 6\sqrt{3} & 0 \\ -6 & 12 & 9\sqrt{3} - 21 & 0 & 6 - 9\sqrt{3} & 9 \\ -6 & 9\sqrt{3} - 21 & 94 - 48\sqrt{3} & 11\sqrt{3} - 9 & 19\sqrt{3} - 34 & 9\sqrt{3} - 24 \\ -6\sqrt{3} & 0 & 11\sqrt{3} - 9 & 12 & -3 - 2\sqrt{3} & -3\sqrt{3} \\ 6\sqrt{3} & 6 - 9\sqrt{3} & 19\sqrt{3} - 34 & -3 - 2\sqrt{3} & 25 - 8\sqrt{3} & 6 - 6\sqrt{3} \\ 0 & 9 & 9\sqrt{3} - 24 & -3\sqrt{3} & 6 - 6\sqrt{3} & 9 \end{bmatrix}. \quad (5.57)$$

It can be seen from $\Omega$ that 13 members are required to construct a universally rigid tensegrity framework, 10 of which are cables and 3 struts. The corresponding tensegrity framework is shown in Fig. 5.2, where the thin black and thick blue lines are cables and struts, respectively. It is worth noting that 13 is exactly the minimal number of members required to build a universally rigid tensegrity framework in generic configurations with 6 nodes in 3D.

**Figure 5.1:** Desired geometric shape associated with (5.56).

**Figure 5.2:** Resultant framework based on $\Omega$ in (5.57).

### 5.4.2 Formation stabilization

Based on the obtained tensegrity framework, where the nodes represent the agents modeled by (5.8), the control input (5.51) for each agent can be specified. For simplification, we set the parameters $\gamma^s_{ij}$ to be the same value $\gamma^s = 2$, and $\gamma^c_{ij}$ is chosen to be $\gamma^c = 0.5$. The initial states of $q(t)$ are chosen as $q_i(0) = q^*_i + 0.3 \times \text{rands}(3,1)$, $i = 1, 2, \cdots, 6$.

It can be seen from Fig. 5.3 that the length of each member converges to its desired one. The upper panel of Fig. 5.3 shows the stabilization errors of the cables, and the corresponding stabilization errors of the struts are shown in the lower panel. The length evolution intervals of the cables and struts together with their rest lengths are presented respectively in Table 5.1 and 5.2, from which we can observe that the length of any cable (resp. strut) is always longer (resp. shorter) than its rest length during the evolution, verifying Proposition 5.6 numerically. As
can be seen from Fig. 5.3, all the stabilization errors converge to zero within 0.1s. Then the formation shape variations during $[0, 0.1]$s are depicted sequentially in Fig. 5.4, where the motion of the straight line of the whole formation results from an additional control input $u_e = [140, 0, 0]^T$ for each agent. This input is independent of the system state, and thus can be separated from the control input (5.51). The intention of designing $u_e$ is to clearly show the variation of the geometric shape of the formation during the system evolution. Overall, these numerical results indicate that the prescribed formation can be achieved using the virtual framework and our proposed control algorithm (5.51).

![Graph of stabilization errors](image)

**Figure 5.3:** The stabilization errors of the members.

### 5.5 Concluding remarks

In this chapter, we have proposed a numerical algorithm, based on which universally rigid tensegrity frameworks can be built for any given configuration in general position. Furthermore, the upper bound of the members inserted in the framework has also been given. Then, we have investigated the formation stabilization problem as one of the applications, where distributed control strategies have been designed, such that the prescribed formation can be realized. During the stabilization evolution, the lengths of the members are shown to vary over or below some bounds, which is of great interest for tethered robot control.
5.5. Concluding remarks

Table 5.1: Length evolution intervals of the cables and their rest lengths.

<table>
<thead>
<tr>
<th>Cable</th>
<th>Length interval</th>
<th>Rest length</th>
</tr>
</thead>
<tbody>
<tr>
<td>(1, 2)</td>
<td>[1.5238, 1.7328]</td>
<td>0.8660</td>
</tr>
<tr>
<td>(1, 3)</td>
<td>[1.0683, 1.7340]</td>
<td>0.8660</td>
</tr>
<tr>
<td>(1, 4)</td>
<td>[2.8182, 3.0474]</td>
<td>1.5222</td>
</tr>
<tr>
<td>(2, 3)</td>
<td>[1.7211, 1.8152]</td>
<td>0.8600</td>
</tr>
<tr>
<td>(2, 5)</td>
<td>[2.8860, 3.0471]</td>
<td>1.5222</td>
</tr>
<tr>
<td>(3, 5)</td>
<td>[2.8011, 3.3172]</td>
<td>1.6583</td>
</tr>
<tr>
<td>(3, 6)</td>
<td>[3.1644, 3.2659]</td>
<td>1.6328</td>
</tr>
<tr>
<td>(4, 5)</td>
<td>[1.7305, 1.9374]</td>
<td>0.8600</td>
</tr>
<tr>
<td>(4, 6)</td>
<td>[1.8143, 2.0299]</td>
<td>0.9095</td>
</tr>
<tr>
<td>(5, 6)</td>
<td>[1.8263, 1.9219]</td>
<td>0.9604</td>
</tr>
</tbody>
</table>

Table 5.2: Length evolution intervals of the struts and their rest lengths.

<table>
<thead>
<tr>
<th>Strut</th>
<th>Length interval</th>
<th>Rest length</th>
</tr>
</thead>
<tbody>
<tr>
<td>(1, 5)</td>
<td>[3.0050, 3.5694]</td>
<td>7.1364</td>
</tr>
<tr>
<td>(2, 6)</td>
<td>[3.8210, 4.0727]</td>
<td>7.6479</td>
</tr>
<tr>
<td>(3, 4)</td>
<td>[3.1661, 3.5706]</td>
<td>7.1364</td>
</tr>
</tbody>
</table>

Figure 5.4: Formation evolution between $t = 0 - 0.1s$. 