Optimizing Tube Precurvature to Enhance the Elastic Stability of Concentric Tube Robots

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Abstract—Robotic instruments based on concentric tube technology are well suited to minimally invasive surgery since they are slender, can navigate inside small cavities, and can reach around sensitive tissues by taking on shapes of varying curvatures. Elastic instabilities can arise, however, when rotating one precurved tube inside another. In contrast to prior work that considered only tubes of piecewise constant precurvature, we allow precurvature to vary along the tube’s arc length. Stability conditions for a planar tube pair are derived and used to formulate an optimal design problem. An analytic formulation of the optimal precurvature function is derived, which achieves a desired tip orientation range while maximizing stability and respecting bending strain limits. This formulation also includes straight transmission segments at the proximal ends of the tubes. The result, confirmed by both numerical and physical experiment, enables designs with enhanced stability in comparison to designs of constant precurvature.

Index Terms—Concentric tube robot, elastic stability, optimal design, precurvature.

I. INTRODUCTION

CONCENTRIC tube robots are a type of continuum robot that are comprised of nested combinations of precurved superelastic tubes [1]–[3]. The shape of these robots is determined by the bending and torsional elastic interaction of the tubes. Since they can assume curves of complex shapes in three-dimensional space, and also possess sufficient stiffness to both steer through tissue and manipulate tools in body cavities, they are particularly well suited to minimally invasive surgery [4]–[7].

One limitation, however, is that instabilities can arise when torsional elastic energy is suddenly released through rapid untwisting of one or more tubes [1], [2]. One approach to mitigating this problem is to plan paths that avoid such unstable configurations, as done in, e.g., [8]. Another approach is to design or manufacture the robot to be globally stable, in the sense that there are no elastic instabilities anywhere in its operational workspace [9]–[11]. Nevertheless, elastic instability imposes significant constraints on robot and workspace design. An important example, shown in Fig. 1, is comprised of a pair of tubes of equal stiffness and equal constant precurvature. By rotating these tubes at their base, the combined curvature varies from the maximum precurvature value to zero. The latter configuration corresponds to a base rotation angle of $\pi$. The tube pair is stable for all relative rotation angles if and only if the following condition is met [1], [12]:

$$L\sqrt{(1+\nu)||\hat{u}_1||||\hat{u}_2||\beta_i\beta_2(1+\nu)} < \pi/2$$  \hspace{1cm} (1)

Here $L$ is the tube length, $\nu$ is the Poisson’s ratio of the tubes, and $\hat{u}_i$ and $\beta_i$ are the precurvature and central angle of tube $i$, respectively. Xu et al. [13] extended this condition to include straight transmission and proposed an implicit method to design a stable robot with more than two tubes. A bifurcation and energy-based analysis to measure local elastic stability are presented in [14] and more general mathematical conditions for local elastic stability of tubes with arclength-varying curvatures in the presence of external loads are derived in [15].

Viewed from different design perspectives, this stability condition places bounds on either the tube length $L$, the precurvatures $\hat{u}_i$, or the central angles $\beta_i$. In particular, two tubes of equal precurvature and Poisson’s ratio $\nu = 0.3$ are globally stable for maximum tip orientation angles, $\beta_1 = \beta_2$, which are...
less than 79°. There are many clinical applications, however, for which it is desirable for tip orientations to vary smoothly between ±90° or even a wider range. Consequently, the development of techniques to increase the stability of concentric tube robots is important for expanding their clinical utility.

Based on (1), two possible approaches to enhancing stability are to vary the tube properties so as to modify Poisson’s ratio, or to consider precurvatures that vary with the arc length. Studies based on the former approach include [10] and [11], where tubes are modified by, e.g., laser cutting, to reduce their bending stiffness without substantially affecting their torsional stiffness. The latter approach of precurvature optimization was first proposed in [16], where a linear ordinary differential equation (ODE)-based stability condition for varying precurvature is presented, and used to numerically optimize a planar tube pair design for maximum elastic stability. It is shown there that, by employing tubes with precurvatures that vary with arc length, the tip orientation angle \( \beta \) can exceed the stability limit given by (1).

In this paper, we expand on [16] by including the following additional contributions. First, we provide a generalization of the stability condition (1) to include straight transmission segments of different length for each tube. Second, we eliminate the constraint of our prior stability condition that both tubes be of equal precurvature and stiffness and instead require that only the planar precurvature of both tubes be in the same direction along their lengths. Also, each tube must be of constant stiffness along its length, but the stiffnesses and Poisson’s ratios of the tubes can be different. Third, we introduce a nondimensionalization of the optimal control problem formulation that allows us to treat the product of maximum curvature and tube length as a single variable. Fourth, we derive an analytic formulation to the maximally stable precurvature problem that satisfies the necessary conditions for optimality and shows that it agrees with the numerical solution. The analytic solution also provides a new upper bound on stable tip orientation angle for tube pairs of given total length, transmission length, and stiffness. Finally, we demonstrate the enhanced stability of tube pairs with our optimized precurvature in comparison to tube pairs of constant precurvature using physical experiments.

The remainder of this paper is organized as follows. Section II reviews our mechanics-based model for the tubes, while the subsequent section presents the kinematics for the special case of planar tube pairs. A necessary and sufficient condition for evaluating the stability of a tube pair with nonconstant precurvatures is derived in Section III. Examples of stability evaluation for two analytically prescribed precurvature functions are presented in Section IV. Section V formulates the selection of precurvatures as an optimal design problem, and provides both analytic and numerical solutions. This section also provides a comparison of the stability limits of constant- and optimal-precurvature tube pairs. An experimental comparison of tube pair stability appears in Section VI. The conclusion is provided in Section VII, while details of the stability condition, derivation of the analytic solution, and experimental calculations appear in three appendixes.

II. KINEMATICS OF TUBE PAIRS WITH PLANAR PRECURVATURE

To analyze elastic stability, it is possible to use a reduced kinematic model that considers only the relative rotation between tubes along their length. These equations are a subset of the general robot kinematics that describe the pose of the robot, which can be found in [1]. Consider a pair of tubes shown in Fig. 2(a).

Let tube 1 and tube 2 denote the outer and inner tubes, respectively. Suppose that tube \( i \) exists over the interval \( s \in [a_i, b_i] \), as shown in Fig. 2(a), where \( s \) is the arc length parameter. Note that we do not explicitly define where the origin \( s = 0 \) is since the kinematics does not depend on the location of the origin. Without loss of generality, the tubes are assumed to be aligned at the tip when combined. If one tube extends beyond the tip of the other, the extended portion does not interact with the other tube, and can thus be ignored in the context of stability.

Let \( R_t(s) \in SO(3) \) denote the material coordinate frame of tube \( i \) at \( s \), which is a body frame of tube \( i \) at \( s \). The frame \( R_t(s) \) is chosen to have its \( z \)-axis tangentially aligned to the tube, and \( y \)-axis perpendicular to the plane in which the tube lies. Now \( \alpha(s) \), the relative rotation between tubes at \( s \), is defined by the angle from \( R_1(s) \) to \( R_2(s) \) about their (common) \( z \)-axis, i.e.,

\[
R_z(s) = R_1(s)R_2(\alpha(s))
\]

(2)

where \( R_z(\alpha) \) is rotation matrix for rotation of \( \alpha \) about the \( z \)-axis.

Let \( \hat{u}_i(s) \in \mathbb{R}^3 \) and \( K_i \in \mathbb{R}^{3 \times 3} \) denote the three-dimensional precurvature vector at \( s \) and the stiffness tensor, respectively, where

\[
\hat{u}_i(s) = \begin{bmatrix} 0 \\ \hat{u}_y(s) \\ 0 \end{bmatrix}, \quad i = 1, 2
\]

\[
K_i = \begin{bmatrix} k_x & 0 & 0 \\ 0 & k_x & 0 \\ 0 & 0 & k_z \end{bmatrix}, \quad i = 1, 2.
\]

Note that the first and second components of \( K_i \) are the same since the cross sections of the tubes are assumed to be circular.
Using the notation of Fig. 2, the kinematics of the tube pair can be derived from the two-tube case of the concentric tube robot kinematics given in [1], i.e.,
\[ \dot{\alpha} = u_{2z} - u_{1z} \]  
\[ u_{2z} = \frac{k_{1z}}{k_{1z} + k_{2z}} \dot{a}_{1y}(s) \sin \alpha \]  
\[ \dot{u}_{2z} = \frac{k_{2z}}{k_{2z}} u_{2x} \dot{a}_{2y}(s) \]  
\[ u_{1z} = -\frac{k_{2z}}{k_{1z}} u_{2z} \]  
where \( u_{ix}, u_{iy} \) are \( x \)- and \( y \)-bending curvatures of the combined tubes expressed in the material coordinate frame of tube \( i \), and \( u_{iz} \) is mechanical twist rate of tube \( i \). Substituting (4) into (5) yields
\[ \dot{u}_{2z} = \frac{k_{1z} k_{2z}}{k_{2z} (k_{1z} + k_{2z})} \dot{a}_{1y}(s) \dot{a}_{2y}(s) \sin \alpha. \]  
From (3) and (6), \( \dot{\alpha} \) is given by
\[ \dot{\alpha} = \frac{k_{1z} + k_{2z}}{k_{1z}} u_{2z}, \]  
and combining (7) and (8) yields an ODE
\[ \dot{\alpha} = \tilde{k} \dot{a}_{1y}(s) \dot{a}_{2y}(s) \sin \alpha \]  
with the boundary condition
\[ \dot{\alpha}(b) = 0 \]  
where \( \tilde{k} \) is defined by
\[ \tilde{k} = \frac{k_{1z} k_{2z} (k_{1z} + k_{2z})}{k_{1z} k_{2z} (k_{1z} + k_{2z})} \]  
When the tubes are made of the same material with Poisson’s ration \( \nu \), then \( \tilde{k} \) reduces to \( \tilde{k} = 1 + \nu \) by substituting \( k_{ix}/k_{iz} = 1 + \nu \) into (11).

Note that (9) applies only for \( s \in [a_1, b] \) where both tubes exist. For \( s \in [a_2, a_1] \), tube 2 is subject to the base torque only, and experiences a constant twist rate. Let \( \Delta \theta \) denote the twist angle accumulated in this interval. Then, \( \Delta \theta \) is given by
\[ \Delta \theta = (a_1 - a_2) u_{2z}(a_1). \]  
Combining (8) and (12) yields
\[ \Delta \theta = -\frac{k_{1z}}{k_{1z} + k_{2z}} (a_1 - a_2) \dot{\alpha}(a_1). \]  
Defining the base rotation \( \alpha_0 \) as the relative rotation of the base frames of the tubes, i.e., \( R_1(a_1) \) and \( R_2(a_2) \), about their (common) \( z \)-axis, \( \alpha(a_1) \) is then sum of \( \alpha_0 \) and \( \Delta \theta \), i.e.,
\[ \alpha(a_1) = \alpha_0 + \frac{k_{1z}}{k_{1z} + k_{2z}} (a_1 - a_2) \dot{\alpha}(a_1). \]  
This equation is another boundary condition for the ODE (9) when the base rotation \( \alpha_0 \) is given as the kinematic input.

Now, consider new tubes obtained by straightening the outer tube and shortening the inner tube, as depicted in Fig. 2(b), so that they both have the same base location at \( s = \alpha \). Since both tubes do not interact from \( s = a \) to \( s = a_1 \), they are constantly twisted. Then, the boundary condition for these new tubes is given by
\[ \alpha(a_1) = \alpha_0 + (a_1 - a) \dot{\alpha}(a_1). \]  
When \( \alpha \) is chosen to be
\[ a = \frac{k_{2z} a_1 + k_{1z} a_2}{k_{1z} + k_{2z}} \]  
the boundary conditions for the original tube pair and the new tube pair in (14) and (15), respectively, become identical. Note that (16) can also be derived from the elastic energy equivalence between the two tube pairs in Fig. 2. Then, given the same base rotation \( \alpha_0 \), the sets of kinematic solutions \( \alpha(s) \) for both tube pairs are identical over \( s \in [a_1, b] \) based on the kinematics and boundary conditions. This equivalence allows us to derive a stability condition and to formulate the optimal design problem for the tube pair in Fig. 2(b) to obtain the solution for the tubes of Fig. 2(a).

### III. Stability Condition for Planar Tube Pairs

The stability condition of (1) applies for tube pairs of constant precurvature and was derived in [1] based on the uniqueness (stable) or nonuniqueness (unstable) of the solution to the kinematic equations. Stability can be presented graphically by plotting relative tube rotation at the base versus the tip for tubes of constant precurvature. The solid curve possesses a unique tip rotation for each base rotation and so is stable. The dashed curve, comprised of tubes of higher precurvature, possesses multiple tip rotations for a range of base rotations and so is unstable.

![Fig. 3. Example plots of relative tube rotation at the base versus the tip for tubes of constant precurvature. The solid curve possesses a unique tip rotation for each base rotation and so is stable. The dashed curve, comprised of tubes of higher precurvature, possesses multiple tip rotations for a range of base rotations and so is unstable.](image-url)
In this section, we present a generalized stability condition. Instead of requiring both tubes to have equal stiffness and precurvature, we only assume that the precurvatures satisfy an inequality
\[
\dot{u}_1(s) \dot{u}_2(s) \geq 0
\] (17)
over the entire length, which is required for the proof of the stability condition presented in Appendix A. This condition only requires both tubes to be curved in the same direction at every point on the length.

Using the notation of Fig. 2, the condition for the curve of Fig. 3 to be monotonically increasing can be written as
\[
\frac{\partial \alpha(b)}{\partial \alpha(a)} > 0 \text{ for } \alpha(a) \in [0, 2\pi].
\] (18)
Since evaluation of this expression involves solving a boundary value problem using the split boundary conditions
\[
\alpha(a) \in [0, 2\pi] \quad \dot{\alpha}(b) = 0
\] (19)
we use the equivalent reciprocal expression and due to symmetry, we need only consider half the interval
\[
\frac{\partial \alpha(a)}{\partial \alpha(b)} > 0, \text{ for } \alpha(b) \in [0, \pi].
\] (20)
In this form, we can evaluate it by integrating backward in arc length from \(b\) to \(a\) using the boundary conditions
\[
\alpha(b) \in [0, \pi] \quad \dot{\alpha}(b) = 0.
\] (21)
Our stability condition based on (20) and derived in Appendix A is given by the following proposition.

**Proposition 1:** Let \(\alpha(s)\) denote the solution to (9) and (21). If \(\dot{u}_1(s) \dot{u}_2(s) \geq 0\), the condition
\[
\frac{\partial \alpha(a)}{\partial \alpha(b)} > 0, \text{ for } \alpha(b) \in [0, \pi]
\] (22)
is equivalent to
\[
\frac{\partial \alpha(s)}{\partial \alpha(b)} > 0, \text{ for } \alpha(b) = \pi, \ s \in [a, b].
\] (23)
This proposition substitutes evaluation of the entire curve in Fig. 3 with a single evaluation at the tip rotation angle of \(\alpha(b) = \pi\). Stability evaluation can be further simplified by considering the linear differential equation
\[
\frac{d^2}{ds^2} \frac{\partial \alpha(s)}{\partial \alpha(b)} \Big|_{\alpha(b) = \pi} = -\tilde{k} \dot{u}_1(s) \dot{u}_2(s) \frac{\partial \alpha(s)}{\partial \alpha(b)} \Big|_{\alpha(b) = \pi}
\] (24)
with boundary conditions
\[
\frac{\partial \alpha(b)}{\partial \alpha(b)} \Big|_{\alpha(b) = \pi} = 1, \quad \frac{d}{ds} \frac{\partial \alpha(b)}{\partial \alpha(b)} \Big|_{\alpha(b) = \pi} = 0.
\] (25)
These equations are obtained by differentiating (9) and (21) with respect to \(\alpha(b)\) at \(\alpha(s) = \pi\). If the solution to (24) and (25) is always positive over \(s \in [a, b]\), the tube pair is stable. This allows us to determine the stability of a given tube pair by examining the solution of a linear differential equation for a set of specific boundary conditions.

### IV. Evaluating Stability for Specific Precurvature Functions

The stability condition presented above can be used to evaluate the stability of particular precurvature functions. Two examples are provided here. The first consists of a pair of tubes with distal sections of constant curvature and straight proximal transmission sections. The second example considers precurvatures of the form \(\dot{u}_y = \frac{q}{s + p}\) with \(p > -a\) and \(q > 0\). The analytic solution to (24)

#### A. Constant Precurvature With Proximal Straight Transmission Lengths

Tube pairs often have straight transmission lengths at their base, as depicted in Fig. 2. While they do not bend, these straight sections do twist and contribute to instability. Consider the case when their precurvature functions are given by

\[
\dot{u}_1(s) = \begin{cases} 0, & \text{if } a \leq s < l_1 \\ \dot{a}_1, & \text{if } l_1 \leq s \leq b \end{cases}
\] (26)
and
\[
\dot{u}_2(s) = \begin{cases} 0, & \text{if } a \leq s < l_2 \\ \dot{a}_2, & \text{if } l_2 \leq s \leq b \end{cases}
\] (27)
Defining \(\tilde{l}\) as \(\tilde{l} = \max(l_1, l_2)\), the solution to (24) is given by
\[
\frac{\partial \alpha(s)}{\partial \alpha(b)} \Big|_{\alpha(b) = \pi} = \begin{cases} \cos(c(b - \tilde{l})) + c(s - \tilde{l}) \sin(c(b - \tilde{l})) & \text{if } s < \tilde{l} \\ \cos(c(b - s)) & \text{if } s \geq \tilde{l} \end{cases}
\] (28)
where \(c\) is defined by
\[
c = \sqrt{\tilde{k} \dot{u}_1 \dot{u}_2}
\] (29)
and \(\tilde{k}\) given in (11). Then, stability condition (23) reduces to
\[
c(L - \tilde{l}) < \tan^{-1} \frac{1}{cL}
\] (30)
where \(L\) is the tube length, i.e., \(L = b - a\).

This equation is a generalized form of (1), the stability condition for constant precurvature tubes without transmission lengths. To recover this simpler form, we can set \(\tilde{l} \to 0\) and assume the same material for both tubes resulting in the right side of (30) converging to \(\pi/2\) and \(\tilde{k}\) in \(c\) becoming \(k = 1 + \nu\), where \(\nu\) is Poisson’s ratio. Then, the condition (30) reduces to
\[
L \sqrt{(1 + \nu) \dot{u}_1 \dot{u}_2} < \pi/2
\] (31)
which is equivalent to (1) under assumption (17).

#### B. Precurvature Function \(\dot{u}_y = q/(s + p)\)

Consider a precurvature function \(\dot{u}_y(s) = \dot{u}_2(s) = \frac{q}{s + p}\) with scalars \(p > -a\) and \(q > 0\). The analytic solution to (24)
is given by

\[
\frac{\partial \alpha}{\partial \beta}(s, \pi) = \begin{cases} 
-\frac{c_2}{c_1} \left( \frac{s + p}{p + p} \right)^{c_1} + \frac{c_2}{2} \left( \frac{s + p}{p + p} \right)^{c_1}, & \text{if } c_0 < \frac{1}{4} \\
-\frac{c_2}{c_1} \sqrt{\frac{s + p}{p + p}} \left( 1 - \frac{1}{2} \ln \frac{s + p}{p + p} \right), & \text{if } c_0 = \frac{1}{4} \\
-\frac{c_2}{c_1} \sqrt{\frac{s + p}{p + p}} \sin \left( c_5 \ln \frac{s + p}{p + p} - c_6 \right), & \text{if } c_0 > \frac{1}{4}
\end{cases}
\]

where

\[
c_0 = \frac{k q^2}{\pi}, \quad c_2 = \sqrt{1 - 4c_0}, \quad c_1 = \frac{1}{2} - \sqrt{\frac{1}{4} - c_0} \\
c_2 = \frac{1}{2} + \sqrt{\frac{1}{4} - c_0}, \quad c_4 = \sqrt{\frac{1}{4} + c_0}, \quad c_5 = \sqrt{c_0 - \frac{1}{4}} \\
c_6 = \tan^{-1} 2c_2.
\]

When \( c_0 \leq \frac{1}{4} \), it is easily shown that \( \frac{\partial \alpha(s)}{\partial \alpha(b)} > 0 \) for \( \alpha(b) = \pi \) and \( s \in [a, b] \). In other words, it can be verified that the tube pair is stable simply by checking \( \frac{\partial \alpha(0)}{\partial \alpha(b)} > 0 \) and \( \frac{d}{ds} \frac{\partial \alpha(s)}{\partial \alpha(b)} \geq 0 \) for \( \alpha(b) = \pi \). This implies that any precurvature with \( (p, q) \) satisfying \( c_0 \leq \frac{1}{4} \) can be used for a stable tube pair. Since \( c_0 \) is invariant to \( p \), the choice of \( p \) is unbounded as long as \( p > -a \).

Defining the central angle \( \theta_{des} \) swept out by the initial curvature of the tubes as

\[
\theta_{des} = \int_0^L \dot{\theta}_y \, ds = q \ln \frac{b + p}{a + p}
\]

an arbitrarily large \( \theta_{des} \) can be achieved by selecting \( p \) to be larger than, but very close to, \( -a \). From a theoretical perspective, this result is quite meaningful, since the swept angle \( \theta_{des} \) was bounded by the inequality (31) for the constant precurvature case. In practice, it is hard to achieve \( c_0 \leq \frac{1}{4} \) due to the yield strain of the material, which bounds the magnitude of precurvature.

On the other hand, when \( c_0 > \frac{1}{4} \), the following inequality must hold in order to satisfy \( \frac{\partial \alpha(s)}{\partial \alpha(b)} > 0 \):

\[
\sqrt{c_0 - \frac{1}{4}} \ln \frac{s + p}{b + p} - \tan^{-1} 2 \sqrt{c_0 - \frac{1}{4}} > -\pi.
\]

Noting that the left side attains a minimum at \( s = a \), the inequality reduces to

\[
al + p > \left( \left( \frac{\pi - \tan^{-1} 2 \sqrt{c_0 - \frac{1}{4}} - 1}{\sqrt{c_0 - \frac{1}{4}}} \right)^{-1} \right. (33)
\]

where \( L = b - a \) is tube length.

Inequality (33) serves as the general stability condition for precurvatures of the form \( \dot{\theta}_y = \frac{\dot{u}_y}{s + p} \). This condition can be validated by plotting \( \alpha(a) \) versus \( \alpha(b) \) curves for various \( (p, q) \) pairs. Some examples are given in Fig. 4. To plot each curve in the figure, \( \alpha(b) \) is discretized into 21 values between 0 and \( 2\pi \) and given as boundary conditions for the general kinematic equations of [1]. As anticipated, pairs \( (p, q) \) that satisfy the stability condition (33) produce monotonically increasing curves. Pairs \( (p, q) \) on the boundary of the inequality (33) produce a curve of infinite slope at \( (\alpha(a), \alpha(b)) = (\pi, \pi) \). Unstable pairs produce negative slopes over some portions of their curves.

\[\dot{u}_y(s) = \frac{k_{1x} \alpha_1(s) + k_{2x} \alpha_2(s)}{k_{1x} + k_{2x}}. \] (35)
From (34) and (35), $\hat{u}_y(s)$, $\hat{u}_1(s)$, and $\hat{u}_2(s)$ satisfy

$$\hat{u}_y(s) = \frac{\eta k_1 + k_2}{\eta (k_1 + k_2)} \hat{u}_1(s) = \frac{\eta k_1 + k_2}{k_1 + k_2} \hat{u}_2(s).$$

(36)

We choose $\hat{u}_y(s)$ as the optimization variable, while any of them could be chosen as they are dependent. Then, the maximum tip angle is given by

$$\int_a^b \hat{u}_y(s)ds = \theta_{\text{des}}.$$  

(37)

In practice, arbitrarily high precurvatures cause tubes to buckle during shape setting or during the relative rotation of the combined tubes. Thus, precurvature must be bounded to a physically realizable value. For example, the shape setting involves bending straight tubes in desired shapes. In this procedure, a precurvature bound may be specified based on the maximum allowable bending strain $\varepsilon$

$$\hat{u}_y(s) \leq 2\varepsilon/d_i, \quad \text{for } i = 1, 2.$$  

(38)

in which $d_i$ is the outer diameter of tube $i$. This induces an upper bound of $\hat{u}_y$ based on (36) of the form

$$\hat{u}_y(s) \leq \hat{u}_{\text{max}}.$$  

(39)

A tube pair is stable if and only if the solution to (9), (10), and (21) satisfies (22). Since the stability boundary corresponds to equality in this expression, maximizing stability equates to maximizing the minimum value of $\frac{\partial \alpha(a)}{\partial \alpha(b)}$ over $\alpha(b) \in [0, \pi]$

$$\max_{\alpha(b) \in [0, \pi]} \left( \min_{\alpha(b) \in [0, \pi]} \frac{\partial \alpha(a)}{\partial \alpha(b)} \right).$$

Noting that $\frac{\partial \alpha(a)}{\partial \alpha(b)}$ is the inverse of the slope of the curves in Fig. 3, the above optimization straightens the curves between $(0, 0)$ and $(\pi, \pi)$ by maximizing the minimum inverse slope.

If the design is stable, then $\frac{\partial \alpha(a)}{\partial \alpha(b)} > 0$ for $\alpha(b) \in [0, \pi]$ and, based on Proposition 3 in Appendix A, its minimum value is given by

$$\min_{\alpha(b) \in [0, \pi]} \frac{\partial \alpha(a)}{\partial \alpha(b)} = \frac{\partial \alpha(s)}{\partial \alpha(b)} \bigg|_{\alpha(b) = \pi} \forall s \in [a, b].$$  

(40)

Based on this equation, the design problem can be stated as follows.

**Design problem:** Solve for

$$\max_{\hat{u}_y(s)} \frac{\partial \alpha(a)}{\partial \alpha(b)} \bigg|_{\alpha(b) = \pi}$$  

(41)

subject to the constraints

$$\int_a^b \hat{u}_y(s)ds = \theta_{\text{des}}$$  

(42)

$$\hat{u}_y(s) = 0 \quad \text{for } s \in [a, t]$$  

(43)

$$\hat{u}_y(s) \leq \hat{u}_{\text{max}} \quad \text{for } s \in [t, b]$$  

(44)

$$\frac{\partial \alpha(s)}{\partial \alpha(b)} \bigg|_{\alpha(b) = \pi} > 0$$  

(45)

where $s = t$ is the end of the straight transmission, and the last constraint ensures stability by Proposition 1.

**B. Nondimensionalization**

As stated above, the design problem depends on four tube parameters: their total length, their straight transmission length, the desired maximum tip tangent angle $\theta_{\text{des}}$, and the maximum curvature $\hat{u}_{\text{max}}$. Noting that $\hat{u}_{\text{max}}$ has units of inverse length, the length parameters and curvature parameters in the problem can be nondimensionalized by multiplying or dividing by $\hat{u}_{\text{max}}$, respectively. The value of nondimensionalization is that the optimal solution, as shown graphically in Section V-E, can be expressed as a function of three rather than four parameters.

Dimensionless variables for arc length $s$, total tube length $\tilde{L}$, straight transmission length $\tilde{T}$, and precurvature $\tilde{u}_y(s)$ are defined as

$$\tilde{s} = \hat{u}_{\text{max}}(s - a) \in [0, \tilde{L}]$$  

(46)

$$\tilde{L} = \hat{u}_{\text{max}}(b - a)$$  

(47)

$$\tilde{T} = \hat{u}_{\text{max}}(t - a)$$  

(48)

$$\tilde{u}_y(s) = \hat{u}_y(s)/\hat{u}_{\text{max}}.$$  

(49)

The dimensionless form of the stability equation (24) and boundary conditions (25) are given by

$$\frac{d^2 \frac{\partial \alpha(\tilde{s})}{\partial \alpha(\tilde{L})}}{d\tilde{s}^2} \bigg|_{\alpha(\tilde{L}) = \pi} = -\kappa \tilde{u}_y^2(\tilde{s}) \frac{\partial \alpha(\tilde{s})}{\partial \alpha(\tilde{L})} \bigg|_{\alpha(\tilde{L}) = \pi}$$  

(50)

$$\frac{\partial \alpha(\tilde{s})}{\partial \alpha(\tilde{L})} \bigg|_{\alpha(\tilde{L}) = \pi} = 1, \quad \frac{d}{d\tilde{s}} \frac{\partial \alpha(\tilde{s})}{\partial \alpha(\tilde{L})} \bigg|_{\alpha(\tilde{L}) = \pi} = 0$$  

(51)

where $\kappa$ is given from (36) by

$$\kappa = \tilde{\kappa} \left( \frac{k_1 + k_2}{k_1 + k_2} \right)^2.$$  

(52)

The dimensionless design problem can be stated as follows.

**Dimensionless design problem:** Solve for

$$\max_{\tilde{u}_y(s)} \frac{\partial \alpha(0)}{\partial \alpha(\tilde{L})} \bigg|_{\alpha(\tilde{L}) = \pi}$$  

(53)

subject to

$$\int_0^L \tilde{u}_y(s)d\tilde{s} = \theta_{\text{des}}$$  

(54)

$$\tilde{u}_y(s) = 0, \quad \text{for } \tilde{s} \in [0, \tilde{T}]$$  

(55)

$$\tilde{u}_y(s) \leq 1$$  

(56)

$$\frac{\partial \alpha(\tilde{s})}{\partial \alpha(\tilde{L})} \bigg|_{\alpha(\tilde{L}) = \pi} > 0.$$  

(57)

**C. Analytic Solution to Optimal Design Problem**

It is sometimes possible to solve the design problem (53) analytically for a specified precurvature function. For example, an analytic solution to the design problem is presented in [16] for the precurvature function $\tilde{u} = \frac{a}{\pi + \alpha}$.

Alternatively, we can attempt to solve for the optimal functional form by considering the dimensionless design problem of Section V-B as an optimal control problem with control variable $\tilde{u}_y(\tilde{s})$. This problem is solved analytically in Appendix B,
where the first-order optimality condition is applied based on [17]. The solution is given in two forms depending on whether $\theta_{\text{des}}$ is greater than $\theta_{\text{ref}}$ or not, where $\theta_{\text{ref}}$ is defined in (129).

1) If $\theta_{\text{des}} > \theta_{\text{ref}}$,

$$
\tilde{u}_y(\hat{s}) = \begin{cases} 
0, & \text{for } 0 \leq \hat{s} < \hat{T} \\
1, & \text{for } \hat{T} \leq \hat{s} < \hat{l} \\
c_1 + \kappa - \sqrt{(s-w)^2}, & \text{for } \hat{l} \leq \hat{s} \leq \hat{L}
\end{cases}
$$

(58)

where $c_1$, $v$, and $w$ are given in (122) and (123), and the saturation point $l$ is computed by root finding of (125) for a given $\theta_{\text{des}}$.

2) If $\theta_{\text{des}} \leq \theta_{\text{ref}}$,

$$
\tilde{u}_y(\hat{s}) = \begin{cases} 
0, & \text{for } 0 \leq \hat{s} < \hat{T} \\
c_1 + \kappa - \sqrt{(s-w)^2}, & \text{for } \hat{T} \leq \hat{s} \leq \hat{L}
\end{cases}
$$

(59)

where $c_1$, $v$, and $w$ are given in (126) and (127). The unknown $\tilde{u}_0$ is the optimal precurevature value at $\hat{s} = \hat{T}$, i.e., $\tilde{u}_y(\hat{T}) = \tilde{u}_0$. This unknown is computed by root finding of (128) for a given $\theta_{\text{des}}$.

Since this solution satisfies the first-order necessary condition for optimality, it could be either a minimizer, a maximizer, or just a saddle point. To answer this question, the analytical solution is compared with the numerically obtained optimal solution below.

D. Numerical Solution to Optimal Design Problem

In this section, the steepest descent method from optimal control, described in [18], is applied to generate numerical solutions, and the numerical solutions are compared with the analytic solution (58), (59). For notational simplicity, we define

$$
x_1(\hat{s}) = \frac{\partial \alpha(\hat{s})}{\partial \alpha(L)} \bigg|_{\alpha(L) = \pi}, \quad x_2(\hat{s}) = \frac{d}{d\hat{s}} \frac{\partial \alpha(\hat{s})}{\partial \alpha(L)} \bigg|_{\alpha(L) = \pi}.
$$

(60)

The Hamiltonian is given by

$$
\mathcal{H} = x_2p_1 - \kappa \tilde{u}_y^2 x_1 p_2
$$

(61)

in which $x(\hat{s}) = (x_1(\hat{s}), x_2(\hat{s}))$ and $p(\hat{s}) = (p_1(\hat{s}), p_2(\hat{s}))$ are the state and costate variables, respectively, that satisfy

$$
\frac{dx}{d\hat{s}} = \frac{d\mathcal{H}}{dp}, \quad x(\hat{L}) = (1, 0),
$$

$$
\frac{dp}{d\hat{s}} = -\frac{d\mathcal{H}}{dx}, \quad p(0) = (0, 0).
$$

(62)

The update direction in the steepest descent method is then given by

$$
\frac{d\mathcal{H}}{d\tilde{u}_y} = -2\kappa \tilde{u}_y(\hat{s}) p_2(\hat{s}) x_1(\hat{s})
$$

(63)

where $x_1(\hat{s})$ and $p_2(\hat{s})$ are computed from the state and costate equations for a given $\tilde{u}(\hat{s})$. In the numerical implementation, the constraints (54)–(56) are given as linear constraints of the form

$$
\sum_{i=1}^{N} \tilde{u}_y^i \Delta \hat{s} = \theta_{\text{des}}
$$

(64)

$$
\tilde{u}_y^i = 0, \quad i = 1, \ldots, M
$$

(65)

$$
\tilde{u}_y^i \leq 1, \quad i = 1, \ldots, N
$$

(66)

where $N$ is the dimension of $\tilde{u}_y$ discretized by the step size $\Delta \hat{s}$, $\tilde{u}_y^i$ is its $i$th component, and $M$ is the largest step index in the transmission length. Then, $d\mathcal{H}/d\tilde{u}_y$ is the $N$-dimensional gradient vector of the cost (53), which can be computed by solving the initial value problems (62) and then substituting $x_1$ and $p_2$ into (63).

As an example, the numerical and analytical solutions are compared for parameter values $\hat{L} = 2$, $\hat{T} = 0$, $\kappa = 1.3$, and $\theta_{\text{des}} = 90^\circ$. The tubes are assumed to be equally stiff and equally precurved.
To investigate the relative stability of the optimal solution, consider a constant curvature tube pair with the same tip rotation angle $\theta_{\text{des}} = 90^\circ$. The value of constant curvature equates to the average curvature over the length of the optimal tube pair and is shown in Fig. 5. A comparison of base and tip rotation angles for the two designs is shown in Fig. 7. As described in Section I, the constant curvature pair will be unstable for any $\theta_{\text{des}} > 79^\circ$ at this value of Poisson’s ratio. The optimal design is stable, however, as seen from its nonnegative slope.

Based on our observations, the numerical solution always converges to the analytic solution. As an illustration, the error between the analytic and numerical solutions is plotted as a function of tube length and tip angle in Fig. 8. Error is defined in (124).

$$E = \frac{1}{L} \int_0^L \left| \tilde{u}_{y,a}(\tilde{s}) - \tilde{u}_{y,n}(\tilde{s}) \right| d\tilde{s} \quad (67)$$

where $\tilde{u}_{y,a}$ and $\tilde{u}_{y,n}$ denote the analytic and numerical solutions, respectively. This figure shows that the average difference in curvature is less than 0.4% of the maximum precurvature.

The empty region in the upper-left corner of parameter space in Fig. 8 is where no analytical solution exists and where the numerical method fails to find a stable solution. The curve bounding this region represents the stability limit and is investigated below.

### E. Stability Limit

The stability limit of the swept angle has been presented for constant precurvatures in [1]. This section presents a new stability limit based on the optimal precurvature presented in Section V-C. This gives a new upper bound on the swept angle for stable tube pairs, given the tube length $L$, straight transmission length $l$, and maximum precurvature $u_{\text{max}}$.

Suppose that $\tilde{L}$ is small enough that the maximum constant precurvature solution is stable, i.e., the dimensionless form of

$$\sqrt{K} (\tilde{L} - \tilde{T}) < \tan^{-1} \frac{1}{T \sqrt{K}} \quad (68)$$

When this inequality holds, the maximum stable tip angle $\theta_{\text{limit}}$ is given by

$$\theta_{\text{limit}} = \hat{u}_{\text{max}} (b - t) = \tilde{L} - \tilde{T}. \quad (69)$$

If, however, $\tilde{L}$ is too large to satisfy (68), then the stability limit must be derived based on the optimal precurvature solution. Since we are interested in the upper bound of $\theta_{\text{des}}$, we will focus on the solution (58), which is for $\theta_{\text{des}}$ larger than $\theta_{\text{ref}}$. Using the notation of (60) and substituting (58) into (24) yields

$$x_1(\tilde{s}) = \begin{cases} x_1(\tilde{T}) + \frac{d^2 x_1}{d\tilde{T}^2} (\tilde{T}) \tilde{s}, & \text{if } 0 \leq \tilde{s} < \tilde{T} \\ \sqrt{\frac{c_1}{c_1 + \kappa}} \tan^{-1} \frac{1}{\sqrt{\kappa}} \left( \frac{u_{\text{ref}} - \tilde{L}}{\sqrt{\kappa}} \right), & \text{if } \tilde{T} \leq \tilde{s} < l \\ \frac{x_1(l) \cos \sqrt{\kappa} (\tilde{s} - l) + x_{\text{ref}}(l) \sin \sqrt{\kappa} (\tilde{s} - l)}{\sqrt{\frac{c_1}{c_1 + \kappa}} \tan^{-1} \left( \frac{u_{\text{ref}} - \tilde{L}}{\sqrt{\kappa}} \right)}, & \text{if } l \leq \tilde{s} \leq \tilde{L} \end{cases} \quad (70)$$

where a scalar $c_2$ can be computed straightforwardly from the boundary condition $x_1(\tilde{L}) = 1$, while the explicit form is unnecessary here. From (50) and (51), $x_1(\tilde{s})$ is a nondecreasing function if the stability condition

$$x_1(\tilde{s}) > 0 \quad (71)$$

holds. Thus, $x_1(\tilde{s})$ is minimum at $\tilde{s} = 0$ and the stability boundary corresponds to

$$x_1(0) = 0 \quad (72)$$

$$x_1(\tilde{s}) \geq 0. \quad (73)$$

After algebraic manipulations with $x_1(\tilde{s})$ in (70), the conditions (72) and (73) reduce to

$$\frac{\kappa}{\sqrt{K}} (l - L) - 1 = 0 \quad (74)$$

$$\tilde{T} \leq l \leq \frac{1}{\sqrt{K}} \tan^{-1} \frac{1}{T \sqrt{K}} + \tilde{T} \quad (75)$$

where $\tau$ is defined in (124).

Defining $\theta_{\text{limit}}$ as the stability limit of swept angle, it can be calculated by root finding of (74) as presented in Table I.
\[ \theta_{\text{lim}} = \text{STABILITY \_LIMIT}(L, \hat{T}, \kappa) \]

1. If (68) holds, return \( \theta_{\text{lim}} = \hat{L} - \hat{T} \).
   Else, go to 2.
2. Find a root \( l \) of (74) in the interval (75).
3. Compute \( \theta_{\text{des}} \) by substituting the root \( l \) into (125).
4. Return \( \theta_{\text{lim}} = \theta_{\text{des}} \).

**TABLE I**

<table>
<thead>
<tr>
<th>COMPUTATION OF STABILITY LIMIT</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \theta_{\text{lim}} = \text{STABILITY _LIMIT}(L, \hat{T}, \kappa) )</td>
</tr>
<tr>
<td>1. If (68) holds, return ( \theta_{\text{lim}} = \hat{L} - \hat{T} ).</td>
</tr>
<tr>
<td>Else, go to 2.</td>
</tr>
<tr>
<td>2. Find a root ( l ) of (74) in the interval (75).</td>
</tr>
<tr>
<td>3. Compute ( \theta_{\text{des}} ) by substituting the root ( l ) into (125).</td>
</tr>
<tr>
<td>4. Return ( \theta_{\text{lim}} = \theta_{\text{des}} ).</td>
</tr>
</tbody>
</table>

Fig. 9. Stability limits of optimal precurvature and constant precurvature varying with \( \hat{L} \) when \( \hat{T} = 0 \) and \( \kappa = 1.3 \).

Fig. 9 provides a comparison of stable tip angles for optimal- and constant-precurvature tube pairs by plotting \( \theta_{\text{des}} \) versus normalized tube length \( \hat{L} \). Note that dimensionless tube length \( \hat{L} = \hat{u}_{\text{max}} \hat{L} \) equates to the tip angle, in radians, of a tube pair designed with maximum precurvature \( \hat{u}_{\text{max}} \). Thus, tubes of precurvature \( \hat{u}_{\text{max}} \) are described by a line of unit slope when tip angle is expressed in radians. (See the right \( y \)-axis in Fig. 9.) For \( \hat{L} \leq 1.38 \) rad (79°), this line separates unstable designs (above) from stable designs (below). For larger values of \( \hat{L} \), however, tube pairs with maximum precurvature along their entire length will be unstable. Consequently, the stability boundary for \( \hat{L} > 1.38 \) lies below this line and is defined by the stability limit obtained, as described in Table I.

If one wishes to make stable tube pairs of constant pre-curvature longer than \( \hat{L} > 1.38 \), they must lie on or below the horizontal dashed line shown in the figure. Designs on this line will produce the maximum stable tip angle \( \theta_{\text{des}} = 1.38 \) rad (79°), but their constant precurvature will be less than the maximum \( \hat{u}_{\text{max}}(s) = \theta_{\text{des}} / \hat{L} \).

The optimal solution enables the design of tube pairs with significantly larger tip angles. Furthermore, the region between the horizontal dashed line and the solid curve corresponds to the new stable designs revealed by this paper.

Fig. 10 shows the effect on the stability limit of adding straight transmission lengths \( \hat{T} \) to the tube pairs. The effect of increasing \( \hat{T} \) is to reduce the maximum stable length of maximum constant curvature tube pairs and to shift the curved portion of the stability limit downward. The transition point can be computed from (68).

**VI. EXPERIMENTS**

To validate our results experimentally, an optimal design tube pair was compared with a constant precurvature tube pair for a desired tip angle of \( \theta_{\text{des}} = 97.4^\circ \). This tip angle exceeds the stability limit for constant precurvature and is close to the stability limit of the optimal design for the tube parameters given in Table II. The tubes were shape set in a grooved template fabricated using an numerical control milling machine. The tube pairs are shown in Fig. 11. Since they relax during use, the shapes of the tube pairs do not exactly match the design parameters in Table II. However, they closely match the optimal shape associated with the relaxed parameters, also given in Table II, \( \theta_{\text{des}} = 94^\circ \) and \( \hat{u}_{\text{max}} = 1/70 \text{ mm}^{-1} \), as shown in Fig. 11.

The relative stiffness is given in the table since it is required to compute \( \kappa \) in (16). \( \eta \) in (34) is chosen to be 1 so that the tubes share the same precurvature. Poisson’s ratio \( \nu \) is also used to compute \( \kappa \) in (11) based on the relation \( k_{xz}/k_{zz} = 1 + \nu \), where \( \nu = 0.3 \) given in the table is equivalent to \( \kappa = 1.3 \). The optimal precurvature is computed using (58). Two types of experiments were conducted. The first consisted of performing relative rotations of the tube pairs at their bases while measuring the relative rotation at their tips. Rotations comprised one complete clockwise revolution followed by one complete counterclockwise revolution. The goal of this experiment was to produce an experimental version of Figs. 3, 4, and 7 showing base rotation versus tip rotation. The second set of experiments was designed to perform sequences of five motions in the neighborhood of any instabilities in order to more precisely map their boundaries. Each set of experiments used the experimental methods described below.

Tube pairs were mounted in a computer-controlled motorized drive system, as shown in Fig. 12(a). The drive system was...
Fig. 11. Constant (left) and optimal (right) precurvature tube pairs. Inset shows the transmission length of the inner tube that passes through collar of outer tube. Location $s$ can be computed by (16) using parameters of Table II. The theoretical shapes (semitransparent green) for the relaxed parameters are superimposed on the top of the tube pairs.

Fig. 12. (a) Tube actuation system. Each tube collar is rotated independently by the system. (b) EM sensors attached to tube tips using plastic brackets. Inner tube is slightly longer to allow for mounting.

Programmed to independently rotate each tube and record base angles. Electromagnetic (EM) trackers (Model 180, Ascension Technologies) were attached to the tip of each tube, as shown in Fig. 12(b), and used to measure orientation. The measurement frequency was 52 Hz and the base was rotated at $5 \, ^\circ / s$. These measurements are converted offline to relative rotation angles at the tube tips following the method presented in Appendix C.

Plots of measured base rotation versus tip rotation for constant precurvature and optimal precurvature are shown, respectively, in Figs. 13(a) and 14(a). The analytical solutions are also plotted in these figures. Recall that base motor rotation and measured tip rotation correspond in the dimensionless equivalent tube model to $\alpha(\tilde{s} = 0)$ and $\alpha(\tilde{L})$, respectively. The constant precurvature tube pair exhibits instability as expected. High resolution plots of the transients that follow each unstable jump are given in Fig. 15. In contrast, the optimal precurvature design is stable.

To investigate whether or not the configurations inside the hysteresis loop are reachable and stable, a second set of experiments was performed. In these experiments, the relative rotation angle at the base of the tubes was commanded as of decreasing
amplitude generated by a sequence of set points
\[ \alpha_{\text{base}, k} = \pi + (-1)^{k+1} \left( 1 - \frac{k}{40} \right) \pi, \quad \text{for } k = 0, \ldots, 40. \] (76)

Tip angle in response to this input is plotted for the constant precurvature and optimal precurvature tube pairs in Figs. 13(b) and 14(b), respectively. For the unstable constant precurvature tube pair, it is possible to move inside a small portion of the hysteresis loop by reversing direction before the instability occurs. This can be seen at the top of the loop in Fig. 13(b). A second region is present at the bottom of the loop. In a given trial, the location of the region depends on when the amplitude of the set point commands (76) becomes small enough to reverse the base rotation without causing an unstable jump. However, most of the interior of the loop cannot be reached. In contrast, twist direction can be reversed at any point along the stable loop of the optimal precurvature pair of Fig. 14(b). As a result, all configurations inside the hysteresis loop are reachable and stable.

VII. CONCLUSION

Prior concentric tube robot designs have considered tubes of piecewise-constant precurvature for reasons of simplicity and also since the combined tube shape is approximately piecewise constant. The results of this paper demonstrate, however, that stability is enhanced for curvatures that decrease with increasing arc length. In particular, the optimal precurvature that maximizes the stability and includes the effects of proximal transmission sections has been derived. In comparison with prior stability results for constant tube precurvatures, this approach removes the limits on both tip orientation range and tube length. The price paid for enhanced stability, though, is a larger average robot radius of curvature. In the concentric tube robot design approach proposed in [1] and further developed in [19], robots are designed as telescoping concatenations of variable and fixed curvature sections. Variable curvature sections correspond directly to the planar tube pairs considered in this paper. Consequently, the new stability results can be directly incorporated into this design framework to create designs with larger stable workspaces.

APPENDIX A

PROOF OF PROPOSITION 1

We first prove the following two propositions.

Proposition 2: Let \( \alpha(s) \) denote the solution to (9) and (21). Then
\[ \frac{\partial \alpha(a)}{\partial \alpha(b)} > 0, \quad \text{for } (a, b) \in [0, \pi] \] (77)
if and only if
\[ \frac{\partial \alpha(s)}{\partial \alpha(b)} > 0, \quad \text{for } s \in [a, b] \text{ and } (a, b) \in [0, \pi]. \]

Proof: The backward direction is obvious, requiring only proof of the forward direction. For notational simplicity, \( \frac{\partial \alpha(s)}{\partial \alpha(b)} \) is replaced by \( x(s) = \frac{\partial \alpha(s)}{\partial \alpha(b)} \). Differentiating (9) and (21) with respect to \( \alpha(b) \) yields a linear ODE
\[ \ddot{x}(s) = \dddot{k} \hat{u}_1 y(s) \hat{u}_2 y(s) \cos(\alpha) x(s) \] (78)
with boundary conditions
\[ x(b) = 1, \quad \dot{x}(b) = 0. \] (79)

When \( \alpha(b) = 0 \), the solution to (9) and (21) is \( \alpha(s) = 0 \). In this case, (78) reduces to
\[ \ddot{x}(s)|_{\alpha(b)=0} = \dddot{k} \hat{u}_1 y(s) \hat{u}_2 y(s) x(s)|_{\alpha(b)=0}. \]

Given the boundary conditions (79) together with a positive value of \( \dddot{k} \hat{u}_1 y(s), \) \( x(s)|_{\alpha(b)=0} \) results in a decreasing function over \( s \in [a, b] \), which satisfies \( x(s)|_{\alpha(b)=0} > 0 \). Suppose there exists a solution to (78) \( x(s)|_{\alpha(b)=0} \) which is not always positive in \( s \in [a, b] \). Differentiating (78) again with respect to \( \alpha(b) \) yields
\[ \frac{d^2 \partial \alpha(s)}{ds^2} \frac{\partial x(s)}{\partial \alpha(b)} = \dddot{k} \hat{u}_1 y(s) \hat{u}_2 y(s) \cos(\alpha) \frac{\partial x(s)}{\partial \alpha(b)} \]
\[ -\dddot{k} \hat{u}_1 y(s) \hat{u}_2 y(s) \sin(\alpha) x(s). \]

Since this is also a linear ODE in \( \frac{\partial x(s)}{\partial \alpha(b)} \) for which the system input is associated with \( x(s) \), it leads to a finite value of \( \frac{\partial x(s)}{\partial \alpha(b)} \) for a finite \( x(s) \). Consequently, \( x(s) \) as well as its minimum value over \( s \in [a, b] \) is continuously varying over \( \alpha(b) \in [0, \pi] \). Since the minimum value of \( x(s)|_{\alpha(b)=0} \) is not positive while that of \( x(s)|_{\alpha(b)=0} \) is 1, there exists at least one \( \alpha(b) \) between 0 and \( \gamma \) for which the minimum value of \( x(s) \) is zero. Let \( \alpha(b) = \beta \) and \( s = c \) denote this \( \alpha(b) \) and corresponding minimizer, respectively. Then
\[ x(c)|_{\alpha(b)=\beta} = 0. \] (80)

By (77) and (79), \( c \) is neither \( a \) nor \( b \). The first-order necessary condition for the minimizer \( c \) which is not on the boundary of the domain \( s \in [a, b] \) is given by
\[ \ddot{x}(c)|_{\alpha(b)=\beta} = 0. \] (81)

However, this is not possible since the only solution to (78) given (80) and (81) is a constant function \( x(s)|_{\alpha(b)=\beta} = 0 \), which does not satisfy the boundary condition (79). Thus, there does not exist any solution to (78) that is not always positive in \( s \in [a, b] \).

Proposition 3: Let \( \alpha(s) \) denote the solution to (9) and (21). If
\[ \frac{\partial \alpha(s)}{\partial \alpha(b)} > 0 \quad \text{for } \alpha(b) = \pi, \quad s \in [a, b] \] (82)
then it satisfies
\[ \frac{\partial \alpha(s)}{\partial \alpha(b)} \geq \frac{\partial \alpha(s)}{\partial \alpha(b)}|_{\alpha(b)=\pi} \]
for any \( s \in [a, b] \) and \( \alpha(b) \in [0, \pi] \).

Proof: Let \( x(s) \) denote \( \frac{\partial \alpha(s)}{\partial \alpha(b)} \) again. By differentiating (9) and (21) with respect to \( \alpha(b) \), the same equations (78), (79) are obtained. When \( \alpha(b) = \pi \), the solution to (9) and (21) is \( \alpha(s)|_{\alpha(b)=\pi} = \pi \). In this case, (78) reduces to
\[ \ddot{x}(s)|_{\alpha(b)=\pi} = -\dddot{k} \hat{u}_1 y(s) \hat{u}_2 y(s) x(s)|_{\alpha(b)=\pi}. \]

The proposition is clearly satisfied when \( s = b \) or \( \alpha(b) = \pi \). Suppose there exists \( s = c \in [a, b] \) and \( \alpha(b) = \beta \in [0, \pi] \) that
does not satisfy the proposition, i.e.,
\[ x(c)|_{\alpha(b) = \beta} < x(c)|_{\alpha(b) = \pi}. \] (83)

Since \( x(s)|_{\alpha(b) = \pi} \) is positive over \( s \in [a, b] \) by (82), the following inequality holds:
\[ \ddot{x}(s)|_{\alpha(b) = \pi} \leq \tilde{k}\dot{u}_y(s)\dot{u}_y(s)\cos(\alpha(s)|_{\alpha(b) = \pi}) x(s)|_{\alpha(b) = \pi} \] (84)
for \( s \in [a, b] \). By (79) and (83), the following inequalities hold as well:
\[
\begin{align*}
\psi(c)|_{\alpha(b) = \pi} &\geq x(c)|_{\alpha(b) = \beta} \\
\psi(b)|_{\alpha(b) = \pi} &\geq x(b)|_{\alpha(b) = \beta}.
\end{align*}
\] (85)

The inequalities (84) and (85) are the conditions for \( x(s)|_{\alpha(b) = \pi} \) to be an upper solution to \( x(s)|_{\alpha(b) = \beta} \) over \( s \in [c, b] \) [20]. It has been proven that any solution to an ODE lies below the upper solution. However, because of the same boundary conditions \( x(b)|_{\alpha(b) = \beta} = x(\bar{b})|_{\alpha(b) = \pi} = 1 \), \( \dot{x}(b)|_{\alpha(b) = \beta} = \dot{x}(\bar{b})|_{\alpha(b) = \pi} = 0 \), and the smaller value of the second derivative at \( s = b \), i.e.,
\[
\begin{cases}
\dot{x}(b)|_{\alpha(b) = \pi} = 0 \\
\dot{x}(b)|_{\alpha(b) = \beta} = -\tilde{k}\dot{u}_y(b)\dot{u}_y(b) \\
\end{cases}
\]
it follows that \( x(s)|_{\alpha(b) = \pi} \) cannot be the upper solution near \( s = b \). This can be shown via a Taylor expansion for a small positive scalar \( \epsilon \) around \( s = b \)
\[
\dot{x}(b - \epsilon, \pi) - x(b - \epsilon, \beta) \approx \frac{1}{2} \epsilon^2 (\dot{x}(b, \pi) - \dot{x}(b, \beta)) < 0.
\]

Since there is a contradiction between (82) and (83), there does not exist any \( c \in [a, b] \) and \( \beta \in [0, \pi] \) satisfying (83) if (82) holds.

The forward direction of Proposition 1 is satisfied straightforwardly by Proposition 2. The backward direction is also satisfied by Proposition 3, i.e.,
\[
\frac{\partial \alpha(s)}{\partial \alpha(b)} \geq \frac{\partial \alpha(s)}{\partial \alpha(b)}|_{\alpha(s) = \pi} > 0.
\]

**APPENDIX B**

**DERIVATION OF THE SOLUTION TO THE DIMENSIONLESS DESIGN PROBLEM**

This appendix presents the derivation of the analytical solution from the first-order optimality conditions of the dimensionless design problem (53)–(57). Again, for notational simplicity, we define
\[
x_1(\tilde{s}) = \frac{\partial \alpha(\tilde{s})}{\partial \alpha(L)}|_{\alpha(L) = \pi}, \quad x_2(\tilde{s}) = \frac{d}{d\tilde{s}} \frac{\partial \alpha(\tilde{s})}{\partial \alpha(L)}|_{\alpha(L) = \pi}.
\]

Let the prime notation denote the derivative with respect to \( \tilde{s} \), for instance, \( x'_1 = dx_1/d\tilde{s} \). Then, (24) and (25) are written as
\[
\begin{pmatrix} x_1' \cr x_2' \cr \end{pmatrix} = \begin{pmatrix} x_2 \\
-\kappa \tilde{u}_y x_1 \end{pmatrix}, \quad x(\tilde{L}) = \begin{pmatrix} 1 \\
0 \end{pmatrix}
\]
and the dimensionless design problem (53)–(57) reduces to
\[
\min_{\tilde{u}_y} \int_0^{\tilde{L}} x_2 d\tilde{s}
\]
subject to
\[
\int_0^L \tilde{u}_y d\tilde{s} = \theta_{des} \quad \text{(88)}
\]
\[
f_1(\tilde{u}_y) = \tilde{u}_y = 0 \quad \text{for} \quad \tilde{s} \in [0, \tilde{T}] \quad \text{(89)}
\]
\[
f_2(\tilde{u}_y) = \tilde{u}_y - 1 \leq 0 \quad \text{(90)}
\]
\[
f_3(\epsilon) = \epsilon - x_1 \leq 0 \quad \text{(91)}
\]

A small positive scalar \( \epsilon \), such that \( 0 < \epsilon \ll 1 \), is introduced to cast the open inequality (57) to a closed inequality (91) to include it in an optimal control problem. The above problem is then an optimal control problem with the state variable \( x = (x_1, x_2) \) and the control variable \( \tilde{u}_y \).

The integral and control constraints (88)–(90) are directly augmented with the Hamiltonian with Lagrange multipliers, while the state constraint (91) should be decomposed into a control and point constraints by differentiating it with respect to \( \tilde{s} \) until \( \tilde{u}_y \) is explicitly shown [17]. The decomposed constraints are given by
\[
f_3'' = \kappa \tilde{u}_y^2 x_1 = 0, \quad \text{for} \quad \tilde{s} \in [\tilde{s}_1, \tilde{s}_2] \quad \text{(92)}
\]
\[
f_3' = -x_2 = 0, \quad \text{at} \quad \tilde{s} = \tilde{s}_p \in [\tilde{s}_1, \tilde{s}_2] \quad \text{(93)}
\]
\[
f_3 = \epsilon - x_1 = 0 \quad \text{at} \quad \tilde{s} = \tilde{s}_p \in [\tilde{s}_1, \tilde{s}_2] \quad \text{(94)}
\]

where \( [\tilde{s}_1, \tilde{s}_2] \) is an unknown interval where the optimal path is on the boundary of (91), and \( \tilde{s}_p \) is any point in \( [\tilde{s}_1, \tilde{s}_2] \). The next step in general is to augment them with the Hamiltonian and end point function using Lagrange multipliers. While the first-order optimality condition outside the interval \( [\tilde{s}_1, \tilde{s}_2] \) is identical to the case without the constraint, a set of specific optimality conditions should be applied to this interval to find the constrained optimal path as well as the unknowns \( \tilde{s}_1 \) and \( \tilde{s}_2 \) [17]. In our case, however, we could skip this step since there does not exist the interval \( [\tilde{s}_1, \tilde{s}_2] \) that satisfies (93) and (94) together. This can be proven as follows: from (86) and (91), \( x_1 \) is a nondecreasing concave function and \( x_2 \) is a nonincreasing function such that
\[
x_1 \geq \epsilon, \quad x_2 = x'_2 \geq 0, \quad x'_2 = x''_2 \leq 0.
\]
If there exists \( \tilde{s}_1, \tilde{s}_2 \) that satisfies (93) and (94), then there exists \( \tilde{s}_p \) such that
\[
x_1(\tilde{s}_p) = \epsilon, \quad x_2(\tilde{s}_p) = 0.
\]

The only \( x_1 \) and \( x_2 \) that satisfy (95) and (96) are the constant functions, \( x_1 = \epsilon \) and \( x_2 = 0 \). However, they do not satisfy the boundary condition in (86). Thus, there does not exist \( [\tilde{s}_1, \tilde{s}_2] \) that satisfies (93) and (94) together.

Consequently, we need to consider the Hamiltonian adjoined with the constraints (88)–(90) only of the form
\[
\mathcal{H} = \begin{cases}
\mathcal{H} + \mu \tilde{u}_y + \sum_{i=1}^{2} \lambda_i f_i, & \text{if} \quad \tilde{s} \in [0, \tilde{T}) \\
\mathcal{H} + \mu \tilde{u}_y + \lambda_2 f_2, & \text{otherwise}
\end{cases}
\]
where the regular Hamiltonian is
\[
\mathcal{H} = x_2 p_1 - \kappa \tilde{u}_y^2 x_1 p_2.
\]
Here, \( \mu \) is a constant, \( \lambda_i \) are functions of \( \tilde{s} \) such that \( \lambda_1 f_1 = 0 \) and \( \lambda_2 \geq 0 \), and \( p = (p_1, p_2) \) is a costate variable. The optimality
conditions are then given by
\[
\frac{d\tilde{H}}{d\tilde{u}_y} = 0, \quad p' = -\frac{d\tilde{H}}{dx}, \quad p(0) = \begin{pmatrix} 0 \\ 0 \end{pmatrix}
\] (99)
with (86). Substituting (97) into (99) yields
\[
2\kappa\tilde{u}_y p_2 x_1 - \lambda_2 = \mu \text{ for } \tilde{s} \in [\tilde{T}, \tilde{L}]
\] (100)
\[
\begin{pmatrix} p_1 \\ p_2 \end{pmatrix}' = \begin{pmatrix} \kappa\tilde{u}_y^2 p_2 \\ -1 - p_1 \end{pmatrix}, \quad p(\tilde{T}) = \begin{pmatrix} 0 \\ 0 \end{pmatrix}.
\] (101)
We omitted the equation for \( \tilde{s} \in [0, \tilde{T}) \) from (100) since it is obvious from (89) that the optimal control is given by
\[
\tilde{u}_y = 0 \text{ for } \tilde{s} \in [0, \tilde{T}).
\] (102)
Now we would like to show that \( p_2 \) is a negative decreasing function over \( \tilde{s} \in [T, \tilde{L}] \). Consider the following second-order ODEs for \( x_1 \) and \( p_2 \) derived from (86) and (101):
\[
x''_1 = -\kappa\tilde{u}_y^2 x_1, \quad x_1(\tilde{L}) = 1, \quad x'_1(\tilde{L}) = 0
\] (103)
\[
p''_2 = -\kappa\tilde{u}_y^2 p_2, \quad p_2(0) = 0, \quad p'_2(0) = -1.
\] (104)
Defining \( y_1 = x'_1/x_1 \) and \( y_2 = p'_2/p_2 \), (103) and (104) reduce to
\[
y'_1 + y''_1 = -\kappa\tilde{u}_y^2, \quad i = 1, 2.
\] (105)
Suppose that \( y_1 \) and \( y_2 \) intersect at \( s = \sigma \), i.e., \( y_1(\sigma) = y_2(\sigma) \). Then, \( y_1 \) and \( y_2 \) are an identical function since (105) is of first order. The contraposition yields that if they are not identical, they never intersect. Now, let us show that they are not identical. First, note that \( y_1 \) is a nonnegative finite function over \( \tilde{s} \in [0, \tilde{L}] \) by (95). For \( \tilde{s} = \delta \in [0, T) \), \( p_2, p'_2 \), and \( y_2 \) are given from (102) and (104) by
\[
p_2(\delta) = -\delta, \quad p'_2(\delta) = -1, \quad y_2(\delta) = 1/\delta.
\] (106)
Since \( y_1(\delta) \) is finite, \( y_2(\delta) > y_1(\delta) \) when \( \delta \) is sufficiently small. Thus, \( y_1 \) and \( y_2 \) are not identical, never intersect, and thus maintain the order relation \( y_2 > y_1 \geq 0 \) over \( \tilde{s} \in [\delta, \tilde{L}] \). Finally, \( p_2 \) is given as a negative decreasing function of the form
\[
p_2(\tilde{s}) = -\delta \exp \int_\delta^\tilde{s} y_2(\sigma) d\sigma, \quad \text{for } \tilde{s} \in [\delta, \tilde{L}].
\] (107)
Now, let us see the sign of \( \mu \). From (88), there exists at least one \( \tilde{s} \in [T, \tilde{L}] \) such that \( \tilde{u}_y(\tilde{s}) > 0 \). Thus, \( \mu \) is negative by (100) since \( 2\kappa\tilde{u}_y p_2 x_1 < 0 \) by positive \( x_1 \) and negative \( p_2 \), and \( \lambda_2 \geq 0 \) by definition. Moreover, \( \tilde{u}_y \) is always positive over \([T, \tilde{L}] \) since, otherwise, \( \mu \) could not be negative by (100).

Until now, we have shown that
1) \( x_1 \) is a positive increasing function over \( \tilde{s} \in [T, \tilde{L}] \);
2) \( p_2 \) is a negative decreasing function over \( \tilde{s} \in [T, \tilde{L}] \);
3) \( \mu \) is a negative constant, i.e., \( \mu < 0 \);
4) \( \tilde{u}_y \) is a positive function over \( \tilde{s} \in [T, \tilde{L}] \).

Keeping these in mind, let us derive the optimal control \( \tilde{u}_y \) for \( \tilde{s} \in [T, \tilde{L}] \). Consider the following two cases.
1) \( \lambda_2 > 0 \): The optimal control is given from \( \lambda_2 f_2 = 0 \) by
\[
\tilde{u}_y = 1.
\] (108)
2) \( \lambda_2 = 0 \): The optimal control is given from (100) by
\[
\tilde{u}_y = \frac{\mu}{2\kappa p_2 x_1}
\] (109)
which is a decreasing function by negative \( \mu \), positive increasing \( x_1 \), and negative decreasing \( p_2 \).

Note that \( \tilde{u}_y \) is continuous at the boundary points between case 1) and case 2). It is shown as follows: since \( \mu \) is the same constant for both cases, the evaluations of (100) for the left and right limits of a bound point are equal
\[
-2\kappa \tilde{u}_y^+ p_2 x_1^+ = -2\kappa p_2 x_1 + \lambda_2
\] (110)
where case 1) and case 2) are assumed to be on the left(−) and right(+) sides of the boundary point, respectively, while this assumption does not affect the generality since the opposite case is given simply by swapping the signs. Noting that \( x_1 \) and \( p_2 \) are continuous as they are solutions of second-order ODEs, the above equation reduces to
\[
2\kappa (1 - \tilde{u}_y^+) p_2 x_1 = \lambda_2.
\] (111)
Since the left side is nonpositive and the right side is nonnegative, both sides should be zero, i.e., \( \tilde{u}_y^+ = 1 \) and \( \lambda_2 = 0 \). Thus, \( \tilde{u}_y \) is continuous at the boundary point by \( \tilde{u}_y = \tilde{u}_y^- = 1 \).

Consequently, the optimal control \( \tilde{u}_y \) over \( \tilde{s} \in [T, \tilde{L}] \) is a set of continuous (and possibly multiple) connections of (108) and (109). Recalling that (109) is a decreasing function, there exist only two possible combinations of (108) and (109).

Solution 1) Equation (108) for \( \tilde{s} \in [T, l] \), and (109) for \( \tilde{s} \in [l, \tilde{L}] \), where \( l \) is an unknown saturation point.

Solution 2) Equation (109) for \( \tilde{s} \in [T, \tilde{L}] \).

Now, let us derive explicit from of \( \tilde{u}_y \) from (109) by eliminating \( x_1 \) and \( p_2 \). Without loss of generality, \( x_1 \) can be expressed by
\[
x_1 = \sqrt{\tilde{u}_y} e^f
\] (112)
where \( f \) is an unknown function of \( \tilde{s} \). Note that this expression does not bound the candidates of \( \tilde{u}_y \) and \( x_1 \) since \( f \) always exists for any combination of \( \tilde{u}_y \) and \( x_1 \) as long as \( \tilde{u}_y \) and \( x_1 \) are positive functions. Then, \( p_2 \) is given from (109) and (112) by
\[
p_2 = \frac{\mu}{2\kappa \sqrt{\tilde{u}_y}} e^{-f}.
\] (113)
Substituting (112) and (113) into (103) and (104), respectively, yields
\[
f = c_1 \int \tilde{u}_y ds + c_2
\] (114)
\[
\frac{3}{4} \tilde{u}_y^2 - \frac{1}{2} \tilde{u}_y \tilde{u}_y'' + (c_1^2 + \kappa) \tilde{u}_y^4 = 0
\] (115)
where \( c_1 \) and \( c_2 \) are constants of integration. Introducing a new function \( g = 1/\tilde{u}_y \), the ODE (115) becomes
\[
-\frac{1}{4} g'^2 + \frac{1}{2} g g'' + (c_1^2 + \kappa) g^4 = 0.
\] (116)
By differentiating with respect to \( \tilde{s} \), (116) reduces to
\[
g'' = 0.
\] (117)
Thus, $g$ is a second-order polynomial that satisfies (116). From straightforward calculations, $g$ is given by

$$g = \frac{c_i^2 + \kappa - v^2(\hat{s} - w)^2}{v}$$  \hspace{1cm} (118)

where $v$ and $w$ are new constants of integration. Then, $\hat{u}_y$ becomes

$$\hat{u}_y = \frac{v}{c_i^2 + \kappa - v^2(\hat{s} - w)^2}. \hspace{1cm} (119)$$

For Solution 1, the five constants $\mu, c_1, c_2, v,$ and $w$ can be computed by substituting (112)–(114) and (119) into four boundary conditions in (103) and (104). The fifth equation is given by continuity of $\hat{u}_y$ at $s = l$, i.e.,

$$\frac{v}{c_i^2 + \kappa - v^2(l - w)^2} = 1. \hspace{1cm} (120)$$

For Solution 2, (120) cannot be used. Instead, one can introduce an unknown value $\hat{u}_0 \leq 1$ such that $\hat{u}_y(T) = \hat{u}_0$

$$\frac{v}{c_i^2 + \kappa - v^2(T - w)^2} = \hat{u}_0. \hspace{1cm} (121)$$

Note that $l$ in Solution 1 or $\hat{u}_0$ in Solution 2 is still unknown. They should be determined by $\theta_{des}$ from (88). Finally, the constants for both cases are given as follows.

**Constants for Solution 1**

$$c_i = v(L - w), \hspace{1cm} v = \frac{\sqrt{\kappa} \tau}{L - l + \tau / \sqrt{\kappa}} \hspace{1cm} (122)$$

$$w = \frac{\hat{L} - l + (l + \hat{L}) \tau}{2 \tau^2} \hspace{1cm} (123)$$

where

$$\tau = \frac{\tan \left( \sqrt{\kappa} (l - \hat{T}) \right) + \hat{T} \sqrt{\kappa}}{1 - \hat{T} \sqrt{\kappa} \tan \left( \sqrt{\kappa} (l - \hat{T}) \right)}. \hspace{1cm} (124)$$

Given $\theta_{des}$, $l$ is determined by

$$\theta_{des} = \frac{\tanh^{-1} \left( \frac{v(L - w)}{\sqrt{c_i^2 + \kappa}} \right) - \tanh^{-1} \left( \frac{v(l - w)}{\sqrt{c_i^2 + \kappa}} \right)}{\sqrt{c_i^2 + \kappa}} + l - \hat{T}. \hspace{1cm} (125)$$

**Constants for Solution 2**

$$c_i = v(\hat{L} - w), \hspace{1cm} v = \hat{u}_0 \kappa \hat{T} / \hat{L} \hspace{1cm} (126)$$

$$w = \frac{\hat{L}}{2 \hat{u}_0 \kappa \hat{L}^2} + (\hat{T} + \hat{L}) / 2. \hspace{1cm} (127)$$

Given $\theta_{des}$, $\hat{u}_0$ is determined by

$$\theta_{des} = \frac{\tanh^{-1} \left( \frac{v(\hat{L} - w)}{\sqrt{c_i^2 + \kappa}} \right) - \tanh^{-1} \left( \frac{v(\hat{L} - w)}{\sqrt{c_i^2 + \kappa}} \right)}{\sqrt{c_i^2 + \kappa}}. \hspace{1cm} (128)$$

Here, $c_2$ and $\mu$ are not presented since they are unnecessary to compute $\hat{u}_y$ in (119). Note that Solution 1 with $l = \hat{T}$ and Solution 2 with $\hat{u}_0 = 1$ have the same $\theta_{des} = \theta_{ref}$:

$$\theta_{ref} = \frac{\tanh^{-1} \left( \frac{\hat{u}_0 \kappa \hat{T} (1 - l)}{2 \sqrt{c_i^2 + \kappa}} \right) - \tanh^{-1} \left( \frac{\hat{u}_0 \kappa \hat{T} (l - 1)}{2 \sqrt{c_i^2 + \kappa}} \right)}{\sqrt{c_i^2 + \kappa}}. \hspace{1cm} (129)$$

If a given $\theta_{des}$ is larger than this value, then Solution 1 is optimal. Otherwise, the solution is given by Solution 2.

**APPENDIX C**

**RELATIVE ROTATIONS FROM EM TRACKER MEASUREMENTS**

The relative rotation of the EM trackers mounted on the robot tips can be described by (see [21])

$$R_i = R_0 e^{[\hat{w}] \alpha_i} \in SO(3)$$

where $R_0 \in SO(3), \hat{w} \in \mathbb{R}^3,$ and $\alpha_i \in \mathbb{R}$ are the initial relative orientation, the screw axis, and the tip rotation angles, respectively, such that $\Vert \hat{w} \Vert = 1$ and $[\cdot]$ denotes the skew-symmetric representation of a vector in $\mathbb{R}^3$. For $r = [r_1 \ r_2 \ r_3]^T \in \mathbb{R}^3,$ the skew-symmetric representation is given by

$$[r] = \begin{bmatrix} 0 & -r_3 & r_2 \\ r_3 & 0 & -r_1 \\ -r_2 & r_1 & 0 \end{bmatrix} \in \mathbb{R}^{3 \times 3}. \hspace{1cm} (130)$$

Also, $\{R_i\}_{i=1, \ldots, m}$ are the rotation matrices corresponding to the sequence of measurements.

Then, $R_0, \hat{w},$ and $\alpha_i$ can be computed sequentially by a series of least squares problems. Suppose that there is no noise in the measurements $\{R_i\}_{i=1, \ldots, m}$. Then, they satisfy

$$R_i \hat{w} = R_0 \hat{w} \hspace{1cm} (131)$$

for any $i = 1, \ldots, m$. When there exists measurement noise, one can consider the following least squares criterion:

$$\frac{1}{2} \sum_{i=1}^m \Vert R_i \hat{w} - R_0 \hat{w} \Vert^2 = m - \hat{w}^T \left( \sum_{i=1}^m R_i \right) R_0 \hat{w}. \hspace{1cm} (132)$$

Note that $R_i$ is known, and that $R_0$ and $\hat{w}$ are unknowns. One can consider $R_0$ to be known as well by assuming that it is equal to the first measurement of $R_i,$ but we prefer to consider it as unknown so as to avoid the dependence of the optimization result concentrated on the noise of the first measurement. Ignoring the constant term $m$ and switching the overall sign, the least squares problem becomes a maximization of the form

$$\max_{\hat{w}, \hat{\phi}} \hat{w}^T \left( \sum_{i=1}^m R_i^T \right) R_0 \hat{w} \hspace{1cm} (133)$$

where $R_0 \in SO(3), \hat{w} \in \mathbb{R}^3$ and $\Vert \hat{w} \Vert = 1,$ or equivalently

$$\max_{\hat{w}, \hat{\phi}} \hat{w}^T \left( \sum_{i=1}^m R_i^T \right) \hat{\hat{w}} \hspace{1cm} (134)$$

where $\hat{w}, \hat{\phi} \in \mathbb{R}^3$ such that $\Vert \hat{w} \Vert = \Vert \hat{\phi} \Vert = 1$ and $\hat{\hat{w}} = R_0 \hat{w}.$

Applying the standard approach to solve a vector space optimization using Lagrangian multipliers and the first-order necessary conditions, the solutions for $\hat{w}$ and $\hat{\phi}$ are given by the first column vectors of $U \in \mathbb{R}^{3 \times 3}$ and $V \in \mathbb{R}^{3 \times 3},$ respectively, where $U$ and $V$ are the matrices from the singular value
decomposition of $\sum_i R_i^T$, i.e.,
\[
\sum_{i=1}^m R_i^T = U S V^T.
\] (135)

Here $U$, $V$, and $S$ are sorted by descending singular values.

Now, let us compute $R_0$. At this point, there are multiple choices of $R_0$ since any $R_0$ that satisfies $\tilde{v} = R_0 \hat{\omega}$ is a minimizer of (133). Note that the equation $\tilde{v} = R_0 \hat{\omega}$ has one-parameter family of solutions of the form
\[
R_0 = R_{0,p} e^{[\hat{w}][t]}
\] (136)
where $t \in \mathbb{R}$ is the solution parameter and $R_{0,p}$ is a particular solution [22]. One particular solution is given by
\[
R_{0,p} = V \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & \det UV \end{bmatrix} U^T
\] (137)
which can be easily checked to satisfy $R_{0,p} \hat{\omega} = \hat{v}$ and to be a rotation matrix. Any other particular solution can be used if available. Now, $R_0$ can be chosen from the solution family to minimize the difference from $R_1$. The minimization is given by
\[
\min_{t \in \mathbb{R}} \| R_1 - R_{0,p} e^{[\hat{w}][t]} \|_F^2.
\] (138)

Here, $\| \cdot \|_F$ denotes the Frobenius norm defined as $\| A \|_F = \sqrt{\text{Tr}(A^T A)}$, where $\text{Tr}(\cdot)$ is the trace operator. Substituting the exponential formula presented in [22]
\[
e^{[\hat{w}][t]} = I + \sin t \, [\hat{w}] + (1 - \cos t) \, [\hat{w}]^2
\] (139)
into (138) yields
\[
\min_{t \in \mathbb{R}} \sqrt{g^2 + h^2} \sin(t + \arctan(2h/g)) + \text{constant}
\] (140)
where
\[
g = -\text{Tr}(R_1^TR_{0,p}[\hat{w}]), \quad h = \text{Tr}(R_1^TR_{0,p}[\hat{w}]^2).
\] (141)
The minimizer of (140) is given by
\[
t^* = 3\pi/2 - \arctan(2h/g)
\] (142)
and $R_0$ is then given by
\[
R_0 = R_{0,p} e^{[\hat{w}][t^*]}
\] (143)
Once $\hat{\omega}$ and $R_0$ are computed, $\alpha_i$ can also be computed by a minimization
\[
\min_{t \in \mathbb{R}} \| R_i - R_0 e^{[\hat{w}][\alpha_i]} \|_F
\] (144)
for each $i = 2, \ldots, m$. This can be solved in the same way as applied to (138).

Note that all the solutions are in closed form; they are fast and reliable to compute without worrying about local minima.

REFERENCES


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