Active Constraints using Vector Field Inequalities for Surgical Robots

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Abstract—Robotic assistance allows surgeons to perform dexterous and tremor-free procedures, but is still underappreciated in deep brain neurosurgery and endonasal surgery where the workspace is constrained. In these conditions, the vision of surgeons is restricted to areas near the surgical tool tips, which increases the risk of unexpected collisions between the shafts of the instruments and their surroundings, in particular in areas outside the surgical field-of-view. Active constraints can be used to prevent the tools from entering restricted zones and thus avoid collisions. In this paper, a vector field inequality is proposed that guarantees that tools do not enter restricted zones. Moreover, in contrast with early techniques, the proposed method limits the tool approach velocity in the direction of the forbidden zone boundary, guaranteeing a smooth behavior and that tangential velocities will not be disturbed. The proposed method is evaluated in simulations featuring two eight degrees-of-freedom manipulators that were custom-designed for deep neurosurgery. The results show that both manipulator-manipulator and manipulator-boundary collisions can be avoided using the vector field inequalities.

I. INTRODUCTION

Surgical robots have received considerable attention in the context of minimally invasive surgery, as aids in procedures performed through small incisions, in systems such as the da Vinci, the RAVEN [1], and the SteadyHand [2]. Their use has been extended to procedures in restricted workspaces such as deep brain microsurgery [3], [4] and endonasal surgery [5].

In minimally invasive surgery and microsurgery, the surgeon operates with long thin tools, and views the workspace through an endoscope or microscope. Their vision is frequently limited to a region near the surgical tool tips. As the surgeon operates the tools, the tool shaft may inadvertently collide with neighboring structures. In particular in microsurgery, the amplitude of hand tremor can be higher than size of the structures being treated. In this context, surgical robots are used as equipment to assist the surgeon in order to increase accuracy and safety, and reduce surgeon’s mental load and the effects of their hand tremor.

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To increase accuracy and attenuate hand tremor, surgical robots are commanded in task space coordinates either through teleoperation or comanipulation. Joint space control inputs are generally obtained by using a kinematic control law, which is usually based on the robot differential kinematics [6]. Kinematic control laws are valid when low accelerations are imposed in the joint space, and are ubiquitous in control algorithms designed for surgical robotics [5], [7]–[10], since low accelerations are expected in such scenarios.

An increase in safety and a reduction in the surgeon’s mental load has been achieved through the generation of active constraints (virtual fixtures) that can, for instance, act as a layer of safety to prevent the surgical tool from entering a restricted region, even if this contradicts the commands given to the robot [5], [7], or transparently generate constrained motion [8]–[10]. An in-depth survey of active constraints was presented by Bowyer et al. [11]. More recent papers published after this survey addressed the use of guidance virtual fixtures to assist in knot tying in robotic laparoscopy [12], and to allow surgeons to feel the projected force feedback from the distal end of a tool to the proximal end of a tool in a comanipulation context [13].

The generation of active constraints using kinematic control laws requires obtaining a Jacobian that relates the constraint to the task at hand. For instance, for the generation of the remote center of motion in minimally invasive surgery, some groups [8], [9] developed Jacobians to maintain a low remote center of motion error. Aghakhani et al. [8] constrained lateral motion in the pivoting point by using the pivoting point dynamics, and Pham et al. [9] used the so-called constrained Jacobian to project the desired end effector velocity into a constrained velocity in the pivoting point frame. Additional requirements, such as the desired tool tip pose following and obstacle avoidance are usually projected in the null space of the constrained task or stacked to form an augmented Jacobian. The main problem from which both approaches suffer is the difficulty in adding inequality constraints, which are useful in the design of virtual fixtures [14].

II. RELATED WORKS

A framework for manipulator control in surgical robotics that can take inequalities into account was developed by Funda et al. [14] using quadratic programming. Their framework was extended by Kapoor et al. [7], who developed a library of virtual fixtures, including five task primitives that can be combined into customized active constraints. Li et al. [5] used the extended framework to aid operators to
move a surgical tool in a highly constrained space in sinus surgery without collisions. In all these studies, non-linear constraints were used. Solving for non-linear constraints requires an initial guess and takes longer than solving for linear constraints, and therefore, the authors also considered the use of linear approximations, which reduce computation time but may result in errors [7]. This raises the question of whether the constraints can be written directly in linear form, avoiding approximations. An additional consideration is that the closed-loop stability of their approach has never been formally proven.

Gonçalves et al. [15] developed a Lyapunov-stable kinematic controller that takes into account both equality and inequality constraints, which are linear with respect to the joint velocities. They experimentally validated their approach using a humanoid robot. Their framework was extended by Quiroz-Omana and Adorno [16] to the control of a mobile manipulator by adding a unilateral equality constraint that pushes the mobile manipulator out of a forbidden zone. This, however, was a reactive constraint and did not impede the mobile robot from entering the restricted zone in the first place. Such reactive behavior is particularly undesirable in medical robotic applications, in which the patient may be harmed if a tool enters a forbidden zone.

This allowed the system to smoothly avoid collisions, but proportional gain in an admittance control law proportionally increases the attraction force of attractive virtual fixtures.

A. Quaternions and dual Quaternions

Quaternions can be regarded as an extension of complex numbers. The quaternion set is defined as

\[ q = h + i\hat{h}_2 + j\hat{h}_3 + k\hat{h}_4 \quad (h_1, h_2, h_3, h_4 \in \mathbb{R}) \]

in which the imaginary units \(i, j, k\) have the properties \(i^2 = j^2 = k^2 = ijk = -1\). The dual quaternion set is defined as

\[ \mathcal{H} \triangleq \{ h + \varepsilon h' : h, h' \in \mathbb{H}, \varepsilon^2 = 0, \varepsilon \neq 0 \} \]

where \(\varepsilon\) is the dual (or Clifford) unit [23]. Addition and multiplication are defined for dual quaternions analogously to complex numbers, and hence we need only to respect the properties of the imaginary and dual units.

Given \(h \in \mathcal{H}\) such that \(h = h_1 + i\hat{h}_2 + j\hat{h}_3 + k\hat{h}_4 + \varepsilon (h'_1 + i\hat{h}'_2 + j\hat{h}'_3 + k\hat{h}'_4)\), we define the operators

\[ \mathcal{P}(h) \triangleq h_1 + i\hat{h}_2 + j\hat{h}_3 + k\hat{h}_4, \quad \mathcal{D}(h) \triangleq h'_1 + i\hat{h}'_2 + j\hat{h}'_3 + k\hat{h}'_4, \]

and

\[ \begin{aligned} \text{Re}(h) &\triangleq h_1 + \varepsilon h'_1, \\ \text{Im}(h) &\triangleq \varepsilon (i\hat{h}_2 + j\hat{h}_3 + k\hat{h}_4 + \varepsilon (i\hat{h}'_2 + j\hat{h}'_3 + k\hat{h}'_4)). \end{aligned} \]

The conjugate of \(h\) is defined as \(h^* \triangleq \text{Re}(h) - \text{Im}(h)\), and its norm is given by \(\|h\| = \sqrt{\text{Re}(h) \cdot \text{Re}(h)} = \sqrt{\|h\|^2} \).

The set \(\mathbb{H}_p \triangleq \{ h \in \mathbb{H} : \text{Re}(h) = 0 \}\) is isomorphic to \(\mathbb{R}^3\) under the addition operation. Thus, the quaternion \(\hat{x} + y\hat{j} + z\hat{k}\) in \(\mathbb{H}_p\) represents the point \((x, y, z) \in \mathbb{R}^3\).

The set of quaternions with unit norm is defined as \(S^3 \triangleq \{ h \in \mathcal{H} : \|h\| = 1 \}\), and \(S^3\) can always be written as \(r = \cos(\phi/2) + v \sin(\phi/2)\), where \(\phi \in \mathbb{R}\) is the rotation angle around the rotation axis \(v \in S^3 \cap \mathbb{H}_p\) [22]. Elements of the set \(S^3\) are called unit dual quaternions and represent tridimensional poses (i.e.,
combined position and orientation) of rigid bodies. Given \( \mathbf{x} \in \mathbf{S}_r \), it can always be written as \( \mathbf{x} = \mathbf{r} + \varepsilon (1/2) \mathbf{r} \mathbf{t} \), where \( \mathbf{r} \in \mathbb{S}^3 \) and \( \mathbf{t} \in \mathbb{H}_p \) represent the orientation and position, respectively [23]. The set \( \mathbf{S}_r \) equipped with the multiplication operation forms the group \( \text{Spin}(3) \times \mathbb{R}^3 \), which double covers \( \mathbf{S}(3) \).

Elements of the set \( \mathcal{H}_p \triangleq \{ \mathbf{h} \in \mathbb{H} : \text{Re}(\mathbf{h}) = 0 \} \) are called pure dual quaternions and are useful for representing Plücker lines. More specifically, a Plücker line belongs to the set \( \mathcal{H}_p \cap \mathbf{S}_r \) and thus is represented by a pure unit dual quaternion such as \([20],[22]\)

\[
\mathbf{l} = \mathbf{l} + \varepsilon \mathbf{m},
\]

where \( \mathbf{l} \in \mathbb{H}_p \cap \mathbb{S}^3 \) is a pure quaternion with unit norm that represents the line direction, and the line moment is given by \( \mathbf{m} = \mathbf{p} \times \mathbf{l} \), in which \( \mathbf{p} \in \mathbb{H}_p \) is a point in the line and \( \times \) is the cross product. Given \( \mathbf{a}, \mathbf{b} \in \mathcal{H}_p \), the inner product and the cross product are respectively defined as \([20],[22]\)

\[
\mathbf{a} \cdot \mathbf{b} \triangleq -\frac{ab + ba}{2}, \quad \mathbf{a} \times \mathbf{b} \triangleq \frac{ab - ba}{2}.
\]

The operator vec\(_4\) maps quaternions into \( \mathbb{R}^4 \) and vec\(_8\) maps dual quaternions into \( \mathbb{R}^8 \). For instance, \( \text{vec}_4 \mathbf{h} = [h_1 h_2 h_3 h_4]^T \) and vec\(_8\) establishes a match between quaternions and Hamilton operators are matrices that satisfy vec\(_4\) \( (h_1 h_2) = H_4(h_1) \) vec\(_4\) \( h_2 = H_4(h_2) \) vec\(_4\) \( h_1 \). Analogously, given \( h_1, h_2 \in \mathcal{H} \), the Hamilton operators satisfy vec\(_8\) \( (h_1 h_2) = H_8(h_1) \) vec\(_8\) \( h_2 = H_8(h_2) \) vec\(_8\) \( h_1 \). Finally, it can be shown by direct calculation that, when \( \mathbf{a}, \mathbf{b} \in \mathbb{H}_p \), the inner product can be written as

\[
\mathbf{a} \cdot \mathbf{b} = (\text{vec}_4 \mathbf{a})^T \text{vec}_4 \mathbf{b} = (\text{vec}_4 \mathbf{b})^T \text{vec}_4 \mathbf{a}
\]

and the cross product between \( \mathbf{a} \) and \( \mathbf{b} \), which is analogous to the case of vectors in \( \mathbb{R}^3 \), can be mapped into \( \mathbb{R}^4 \) by using the operator \( \mathbf{S}(\cdot) \) as

\[
\text{vec}_4(\mathbf{a} \times \mathbf{b}) = \mathbf{S}(\mathbf{a}) \text{vec}_4 \mathbf{b} = \mathbf{S}(\mathbf{b})^T \text{vec}_4 \mathbf{a}.
\]

Finally, the time-derivative of a quaternion \( \mathbf{a} \in \mathbb{H}_p \) is given by

\[
d(\|\mathbf{a}\|) = \frac{\dot{\mathbf{a}} \mathbf{a}^* + \mathbf{a} \dot{\mathbf{a}}^*}{2\|\mathbf{a}\|} = \frac{\langle \mathbf{a}, \dot{\mathbf{a}} \rangle}{\|\mathbf{a}\|} \mathbf{a} \text{vec}_4 \mathbf{a}^T \text{vec}_4 \mathbf{a},
\]

for any \( \mathbf{a} \in \mathbb{H}_p \) we have \( \mathbf{a}^* = -\mathbf{a} \).

### B. Differential kinematics

Differential kinematics is the relation between task space velocities and joint space velocities, in the general form \( \dot{x} = J \dot{q} \), in which \( \dot{x} \in \mathbb{R}^m \) is the vector of task space velocities, \( \dot{q} \in \mathbb{R}^n \) is the vector of manipulator joint velocities, and \( J \in \mathbb{R}^{m \times n} \) is the Jacobian matrix.

For instance, given the robot end effector pose \( \mathbf{x} \triangleq \mathbf{x}(t) \in \text{Spin}(3) \times \mathbb{R}^3 \), the differential kinematics is given by vec\(_8\) \( \mathbf{x} = J_{\mathbf{x}} \dot{q} \), where \( J_{\mathbf{x}} \in \mathbb{R}^{8 \times n} \) is the dual quaternion analytical Jacobian, which can be found by using dual quaternion algebra \([19]\). Similarly, given the end effector position \( \mathbf{r} \triangleq \mathbf{r}(t) \in \mathbb{H}_p \) and the orientation \( \mathbf{r} \triangleq \mathbf{r}(t) \in \mathbb{S}_3 \), we have

\[
\text{vec}_4 \dot{t} = J_t \dot{q},
\]

\[
\text{vec}_4 \dot{r} = J_r \dot{q},
\]

where \( J_t, J_r \in \mathbb{R}^{4 \times n} \) are calculated from \( J_{\mathbf{x}} \) also by using dual quaternion algebra \([24]\). It is important to note that, in general applications, \( \dot{x}, \dot{t}, \) and \( \dot{r} \) may refer to any relevant coordinate system related to the robot, and not only to the end effector.

### C. Linear programming for differential inverse kinematics

In closed-loop differential inverse kinematics, first a desired task space target, \( x_d \), and the task error \( \dot{x} = x - x_d \) are defined. For \( \eta \in (0, \infty) \), the minimum-norm analytical solution of the optimization problem

\[
\min_{\dot{q}} \| J \dot{q} + \eta \ddot{x} \|^2
\]

is \( J^\dagger \eta \ddot{x} = \dot{q} \), in which \( J^\dagger \) is the generalized Moore-Penrose pseudoinverse of \( J \).

Adding inequality and equality constraints in Problem 6 turns it into a quadratic programming problem requiring a numerical solver \([25]\). The 1-norm analogue of Problem 6 is

\[
\min_{\dot{q}} \| J \dot{q} + \eta \ddot{x} \|_1.
\]

A Lyapunov solution stable for Problem 7 that considers both equality and inequality constraints is given by the following linear program in canonical form \([15]\).

\[
\min_{\dot{q}} \left[ -1^T J^T J 2 \cdot 1^T 0 0 \right] g - \eta 1^T \ddot{x}
\]

subject to

\[
\begin{bmatrix}
J - J - I & I & 0 & 0 \\
1^T & 1^T & 0^T & 1^T & 0^T & 0^T & 0^T & 0^T
\end{bmatrix} g = \begin{bmatrix}
-\eta \ddot{x} \\
\beta \| \ddot{x} \|_1
\end{bmatrix},
\]

\[
g \geq 0
\]

where \( g = [\dot{q}_P^T \dot{q}_N^T \dot{q}_I^T \dot{y}_A^T \dot{y}_B^T \dot{y}_C^T]^T \) is the vector of decision variables. Moreover, \( 0^T \) and \( 1^T \) are row vectors of ones and zeros of appropriate dimensions, respectively. In the canonical form, all decision variables should be non-negative, and hence the joint velocities are split into the non-negative \( \dot{q}_P \) and \( \dot{q}_N \) such that

\[
\dot{q} = \dot{q}_P - \dot{q}_N.
\]

Moreover, \( \dot{y} \) is a vector of residuals for the objective function, and the slack variables \( \dot{y}_A, \dot{y}_B, \dot{y}_C \) are used to transform inequality constraints into equality ones.

The matrix of constraints has three horizontal blocks from top to bottom. The first block concerns error convergence, the second block guarantees that the joints will stop moving when \( \ddot{x} \to 0 \) with \( \beta \in (0, \infty) \), and the third block is a generic block of \( r \) constraints, in which \( W \in \mathbb{R}^{r \times n} \) and \( w \)
The vector field inequalities proposed in this work concern an appropriate choice of $W$ and $w$ to generate active constraints, as described in the following section.

### IV. Vector field inequality

![Fig. 1. Proposed vector field inequality method. The vector field is given by an inequality constraint, and to each point in space is assigned a maximum approach velocity (the red vector in each vector pair), and a maximum separating velocity (the green vector in each vector pair) in the direction perpendicular to the restricted zone boundary. Tangential velocities are unconstrained.]

Using the distance function and the Jacobian, complex restrictions can be generated, by either maintaining the distance above a desired level or keeping the distance below a certain level. The vector field inequality is illustrated in Fig. 1.

#### A. Preventing the robot from entering a restricted region

We first define a minimum safe distance, $d_{\text{safe}} \in [0, \infty)$, that delineates the boundary of the restricted zone. We then define a distance error as

$$\tilde{d}(t) \triangleq d(t) - d_{\text{safe}}, \quad (11)$$

which will be positive when in the safe-zone, zero along the boundary, and negative within the restricted zone.

Assuming a constant safe distance, the distance error dynamics is given by $\dot{\tilde{d}}(t) = \tilde{d}(t)$. A positive $\dot{\tilde{d}}(t)$ means that the system is moving away from the restricted space boundary and a negative $\dot{\tilde{d}}(t)$ means that the system is moving closer to the restricted space. The goal is to constrain the distance dynamics by the inequality

$$\dot{\tilde{d}}(t) \geq -\eta_d \tilde{d}(t), \quad (12)$$

where $\eta_d \in (0, \infty)$.

To understand Constraint (12) let us suppose that $\tilde{d}(t) > 0$, which means that the system is outside the restricted zone. In this situation, any increase in distance is always allowed, which implies $\dot{d} \geq 0 \geq -\eta_d \tilde{d}$. However, when the system is approaching the boundary from the safe region (i.e., $0 \geq d \geq -\eta_d d$), it can approach with a maximum velocity that decreases, in the worst case, exponentially with the distance $\dot{\tilde{d}}(t)$, because the maximum decreasing rate is achieved when $\dot{\tilde{d}}(t) = -\eta_d \tilde{d}(t)$. Thus, the closer the object is to the boundary, the slower it can move in its direction. Any slower motion toward the boundary is also allowed, and hence $\dot{\tilde{d}}(t) \geq -\eta_d \tilde{d}(t)$. As soon as $\tilde{d}(t) = 0$, the restriction becomes $\dot{\tilde{d}}(t) \geq 0$ and therefore the system will not enter the restricted zone.

Considering (10), Constraint (12) is written explicitly in terms of joint velocities as

$$J_d \dot{q} \geq -\eta_d \tilde{d}(t) \iff -J_d \dot{q} \leq \eta_d \tilde{d}(t). \quad (13)$$

To turn Constraint (13) into an equality and fit it into the linear programming formalism, we introduce the slack variable $z_D$ and decompose $\dot{q}$ as in (9) to find

$$-J_d \dot{q}_P + J_d \dot{q}_N + z_D = \eta_d \tilde{d}(t). \quad (14)$$

Notice that any number of constraints in the form of Constraint (14) can be found for different interactions within the robot workspace.

**Remark 1:** Consider that the system is initially inside the restricted zone; that is, $\tilde{d}(t) < 0$. In this case, Constraint (12) will be fulfilled only if $\tilde{d}(t)$ is greater than zero, which means that the system will, at least, be pushed back to the boundary exponentially.

#### B. Maintaining the robot inside a safe region

Using the same reasoning as in Section IV-A, if we want to maintain the robot inside a safe region, we must redefine $d_{\text{safe}}$; that is,

$$\tilde{d}(t) \triangleq d_{\text{safe}} - d(t),$$

with the final solution, assuming the desired error dynamics (12), given by

$$J_d \dot{q}_P - J_d \dot{q}_N + z_D = \eta_d \tilde{d}(t).$$

### V. Distance functions and Jacobians for active constraints

The proposed vector field inequality discussed in the previous section requires a distance function $d(t)$ and the corresponding distance Jacobian $J_d$ for each collidable pair. In this section, we find four distance functions and Jacobians for pairs in which one element is static and the second element is dynamic.

#### A. Point–static-plane distance Jacobian, $J_{t,\pi}$

One of the primitives for virtual fixtures is the constraint of a point such that it is above a static plane. Given a reference frame $F$, a plane $\pi$ in dual quaternion space is given by [22]

$$\pi \triangleq n_\pi + \varepsilon d_\pi,$$
in which \( n_\pi \in \mathbb{H}_p \cap S^3 \) is the normal to the plane and \( d_\pi \in \mathbb{R} \) is the signed perpendicular distance between the plane and the origin of the reference frame. Moreover, the distance \( d_\pi \) is given by \( d_\pi = \langle p_\pi, n_\pi \rangle \), where \( p_\pi \) is an arbitrary point in the plane. If \( F_\pi \) is a frame attached to the plane, the signed distance between \( t \), an arbitrary point in the robot kinematic chain, and the plane, from the point of view of the plane, is given by

\[
d_{t,\pi}^2 \triangleq \langle t, n_\pi \rangle - d_\pi.
\] (15)

The time derivative of (15) is

\[
d \left( d_{t,\pi}^2 \right) = (t, n_\pi)^T J_{t,\pi} \dot{q}.
\] (16)

**B. Line Jacobian, \( J_{l,t} \)**

The line Jacobian is a pre-requisite for the more complex Jacobians described in the following sections.

The Plücker line [22] \( \mathbf{l}_z \in \mathbb{H}_p \cap \mathbf{S} \) collinear to the \( z \)-axis of a frame represented by \( \mathbf{g} = \mathbf{r} + \varepsilon (1/2) \mathbf{tr} \) is given by

\[
\mathbf{l}_z = \mathbf{r} + \varepsilon \mathbf{z}_z,
\] (17)

where \( \mathbf{r} = \hat{\mathbf{r}} \mathbf{r}^* \) and \( \mathbf{z}_z = \mathbf{t} \times \mathbf{l}_z \). The derivative of (17) is

\[
\dot{\mathbf{l}}_z = \dot{\mathbf{r}} + \varepsilon \dot{\mathbf{z}}_z.
\] (18)

Hence,

\[
\vec{c}_z \dot{\mathbf{l}}_z = \left( \begin{array}{c} \mathbf{H}_4 (\mathbf{r}^*) + \mathbf{H}_4 (\dot{\mathbf{r}}) \mathbf{C}_4 \end{array} \right) J_{t} \dot{q},
\] (19)

in which \( \mathbf{C}_4 = \text{diag}(1, -1, -1, -1) \). In addition,

\[
\vec{c}_z \dot{\mathbf{m}}_z = \left( \begin{array}{c} \mathbf{H}_4 (\mathbf{l}_z) - \mathbf{H}_4 (\dot{\mathbf{r}}) \mathbf{C}_4 \end{array} \right) J_{t} \dot{q}.
\]

Finally, (18) can be re-written in term of joint velocities as

\[
\vec{c}_z \dot{\mathbf{l}}_z = \left[ \begin{array}{c} J_{t,z} \\ J_{t,m_3} \end{array} \right] \dot{q} \triangleq J_{t,z} \dot{q}.
\] (20)

**C. Line–static-point distance Jacobian, \( J_{l,z,p} \)**

The line–static-point Jacobian can be used to generate a remote center of motion. First, we notice that the distance between \( \mathbf{l}_z \in \mathbb{H}_p \cap \mathbf{S} \) and an arbitrary static point \( \mathbf{p} \in \mathbb{H}_p \) is given by

\[
d_{l,z,p} = \| \mathbf{p} \times \mathbf{l}_z - \mathbf{m}_z \|.
\] (21)

The derivative of (21) is

\[
d_{l,z,p} = \frac{1}{d_{l,z,p}} \vec{c}_4 (\mathbf{p} \times \mathbf{l}_z - \mathbf{m}_z)^T \vec{c}_4 \left( \frac{\text{d}(\mathbf{p} \times \mathbf{l}_z - \mathbf{m}_z)}{\text{dt}} \right) \]

\[
= \frac{1}{d_{l,z,p}} \vec{c}_4 (\mathbf{p} \times \mathbf{l}_z - \mathbf{m}_z)^T \left( \mathbf{S}(\mathbf{p}) J_{t,z} - J_{t,m_3} \right) \dot{q}.
\] (22)

**D. Point–static-line distance Jacobian, \( J_{t,l} \)**

The point–static-line Jacobian can be used to keep a point inside/outside a cylinder. First, we notice that the distance between an arbitrary line \( \mathbf{l} \in \mathbb{H}_p \cap \mathbf{S} \), such that \( \mathbf{l} = \mathbf{l} + \varepsilon \mathbf{m} \), and a point \( \mathbf{t} \in \mathbb{H}_p \) in the robot kinematic chain is given by

\[
d_{t,l} = \| \mathbf{t} \times \mathbf{l} - \mathbf{m} \|.
\] (23)

The derivative of (23) is given by

\[
\dot{d}_{t,l} = \frac{1}{d_{t,l}} \vec{c}_4 (\mathbf{t} \times \mathbf{l} - \mathbf{m})^T \vec{c}_4 \left( \frac{d((\mathbf{t} \times \mathbf{l} - \mathbf{m}))}{dt} \right) \]

\[
= \frac{1}{d_{t,l}} \vec{c}_4 (\mathbf{t} \times \mathbf{l} - \mathbf{m})^T \mathbf{S}(\mathbf{l})^T J_{t} \dot{q}.
\] (24)

**E. Line–static-line distance Jacobian**

The line–static-line Jacobian is particularly useful for avoiding collisions between a moving cylinder and a static cylinder. In order to obtain the line–static-line Jacobian, we use the concept of dual angle between Plücker lines.

1) **Inner product Jacobian, \( J_{\langle \mathbf{l}, \mathbf{L} \rangle} \)**: The dual cosine between Plücker lines \( \mathbf{l}_z, \mathbf{L}_z \in \mathbb{H}_p \cap \mathbf{S} \) is obtained by using the inner product [20]

\[
\langle \mathbf{l}_z, \mathbf{L}_z \rangle = \| \mathbf{l}_z \| \| \mathbf{L}_z \| \cos (\phi + \varepsilon d_1) = \cos \phi - \varepsilon d_1 \sin \phi,
\] (25)

where \( d_1 \in [0, \infty) \) and \( \phi \in [0, 2\pi) \) are the distance and the angle between the lines, respectively.

Since the line \( \mathbf{l} \) is static, \( \mathbf{l} = 0 \) and thus the derivative of (25) is given by

\[
\vec{c}_8 \frac{d (\langle \mathbf{l}_z, \mathbf{L} \rangle)}{dt} = - \left( \frac{\mathbf{H}_8 (\mathbf{L}) + \mathbf{H}_8 (\mathbf{l})}{2} \right) J_{l} \dot{q}.
\]

\[
\left[ \begin{array}{c} \vec{c}_4 \dot{\mathbf{P}} (\langle \mathbf{l}_z, \mathbf{L} \rangle) \\
\vec{c}_4 \dot{\mathbf{D}} (\langle \mathbf{l}_z, \mathbf{L} \rangle) \end{array} \right] = \left[ \begin{array}{c} J_{l} \mathbf{r} (\langle \mathbf{l}_z, \mathbf{L} \rangle) \\
J_{l} \sigma (\langle \mathbf{l}_z, \mathbf{L} \rangle) \end{array} \right] \dot{q}.
\] (26)

2) **Cross product Jacobian, \( J_{\mathbf{l}_z \times \mathbf{L}_z} \)**: Given the dual angle \( \phi = \phi + \varepsilon d_1 \), where \( d_1 \in [0, \infty) \) and \( \phi \in [0, 2\pi) \), and the Plücker line \( \mathbf{g} \in \mathbb{H}_p \cap \mathbf{S} \), with \( \mathbf{g} = \varepsilon \mathbf{m}_s \), the cross product between \( \mathbf{l}_z \) and \( \mathbf{l} \) is given by [20]

\[
\mathbf{l}_z \times \mathbf{L} = \| \mathbf{l}_z \| \| \mathbf{L} \| \varepsilon \sin \phi
\]

\[
= \varepsilon \sin \phi \varepsilon \mathbf{m}_s \sin \phi + \varepsilon s d_1 \cos \phi,
\] (27)

in which \( \mathbf{g} \) is the line perpendicular to both \( \mathbf{l}_z \) and \( \mathbf{L} \). The time derivative of the cross product between \( \mathbf{l}_z \) and the static \( \mathbf{g} \)

\[
\mathbf{g} = (a + \varepsilon b) = f(a) + \varepsilon f'(a)\right). \text{ For more details, see [22].}
\]
Finally, we substitute (31) and (32) in (30) to find

\[
\dot{d}_{t,z,t} \frac{1}{P(L_z \times L)} \frac{d}{dt} \left( \frac{\|D(L_z, L)\|}{P(L_z \times L)} \right) = \frac{d}{dt} \left( \frac{\|D(L_z, L)\|}{P(L_z \times L)^2} \right) \dot{q}.
\]  

and similarly we obtain the derivative of the norm of \( P(L_z \times L) \) using (3) and (28) as

\[
\frac{d}{dt} \left( \frac{\|P(L_z \times L)\|}{\|P(L_z \times L)^2\|} \right) = \frac{1}{\|P(L_z \times L)\|} \text{vec}_4 D(L_z, L)^T J_P(L_z, L) \dot{q}.
\]  

Finally, we substitute (31) and (32) in (30) to find

\[
\dot{d}_{t,z,t} = \left( a J\|D(L_z, L)\| + b J\|P(L_z \times L)\| \right) \dot{q}.
\]  

4) Parallel distance Jacobian, \( J_{t,z,t} \): In the degenerate case in which \( L_z \) and \( L \) are parallel (i.e., \( \phi \in \{0, \pi\} \)), the distance between them can be retrieved as

\[
\dot{d}_{t,z,t} = \|D(L_z, L)\| = \|sd_z\| = d_t.
\]  

Resorting once more to (3), we find the derivative of (34) as

\[
\dot{d}_{t,z,t} = \frac{1}{\|D(L_z \times L)\|} \text{vec}_4 D(L_z, L)^T J_D(L_z \times L) \dot{q}.
\]  

3) Distance Jacobian between non-parallel lines: The distance \( d_{t,z,t} \) between \( L_z \) and \( L \) when they are not parallel (i.e., \( \phi \in (0, 2\pi) \setminus \{0, \pi\} \)) can be obtained as

\[
d_{t,z,t} = \|D(L_z, L)\| = \|d_z \sin \phi\| = d_t.
\]  

Noting that both the numerator and denominator of (29) are real numbers, we can find the derivative of (29) as

\[
\dot{d}_{t,z,t} = \frac{1}{\|P(L_z \times L)\|} \frac{d}{dt} \left( \|D(L_z, L)\| / \|P(L_z \times L)\| \right) \dot{q}.
\]  

We obtain the derivative of the norm of \( D(L_z, L) \) using (3) and (26) as

\[
\frac{d}{dt} \left( \frac{\|D(L_z, L)\|}{\|P(L_z \times L)\|} \right) = \frac{1}{\|P(L_z \times L)\|} \text{vec}_4 D(L_z, L)^T J_D(L_z, L) \dot{q}.
\]  

The distance between lines can be given by the composition of the distance (29) and (34) as

\[
\begin{cases}
J_{t,z,t} \phi \in (0, 2\pi) \setminus \{0, \pi\}, \\
J_{t,z,t} \phi \in \{0, \pi\}.
\end{cases}
\]  

5) Manipulator line-to-line distance Jacobian, \( J_{t,z,t} \). The line-to-line distance Jacobian is given by combining (33) and (35) as

\[
\begin{cases}
J_{t,z,t} \phi \in (0, 2\pi) \setminus \{0, \pi\}, \\
J_{t,z,t} \phi \in \{0, \pi\}.
\end{cases}
\]  

VI. SIMULATIONS

In order to validate our approach, we developed a realistic simulated version, using V-REP\(^2\) and DQ Robotics for MATLAB\(^4\) of a deep neurosurgical robotic system that we previously developed in close cooperation with neurosurgeons [4] (see Fig. 2). In deep neurosurgery, the workspace is highly constrained in the microscopic view. The diameter of the tools is 3.5 mm, and the radius and depth of the workspace cylinder are 2.8 cm and 8 cm, respectively.

In this preliminary evaluation, one of the tools was kept fixed while the second tool moved. Five simulations (S1-S5) with increasing complexity were run, as follows. S1: the task consisted of moving the end effector, \( \mathbf{z}(t) \), of one of the tools in a given trajectory \( \mathbf{z}_g(t) \). The constraints are: C1, a line-static-line distance constraint, to avoid collisions between tool shafts, where the shaft of the moving tool has a center-line given by \( L_z \) and the static shaft has a center-line \( L_2 \); C2, a line-static-point distance constraint to generate the entry sphere with the center given by \( p_{cen} \); C3, a point-static-distance constraint to prevent the lower part of the moving tool given by \( t_0(t) \) from colliding with the cylindrical workspace with center-line \( L_z \); and C4, a point-static-plane distance constraint to prevent the end effector from crossing the lower bound of the cylindrical workspace given by the plane \( \pi \).

\(^2\)See accompanying video.
\(^3\)http://www.coppeliarobotics.com/
\(^4\)http://dqrobotics.sourceforge.net
to avoid collisions of the lower part of the shaft with the workspace boundary, which is represented by a cylinder, using $J_{t,l}$ from (24) and $d_{t,l}$ as in (25); and finally, S5: in addition to constraints C1, C2, and C3, a point–static-plane distance constraint C4 was added to prevent the end effector from crossing the plane $\pi_t$, using $J_{t,\pi}$ from (16) and $d_{t,\pi}$ as in (15).

The robot has eight degrees-of-freedom in which the first, second, and fifth joints were prismatic and the remaining joints were revolute. The robots’ initial configurations were such that the shafts were at a long distance from each other. The trajectory, which was the same for all simulations, was the screwline interpolation of four poses. The final form of the linear program, when all constraints were used, was

$$\min \left[ -1^T J_{\pi} J_{\pi} 2 \cdot 1^T 0^T 0^T 0^T 0^T \right] g$$

subject to

$$\begin{bmatrix} J_{t,l}^T & -J_{t,p}^T & -I & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} W_1 - W_2 = 0,$$

$$\begin{bmatrix} J_{t,\pi}^T & -J_{t,l}^T \end{bmatrix} W_3 - W_4 = 0,$$

$$g = \left[ -\eta \begin{bmatrix} x \end{bmatrix}, \beta \begin{bmatrix} x \end{bmatrix} \right],$$

in which $W_1$ and $W_1$ are constraints for the joint limits [16], $z^I_1, z^I_2, z^I_3, z^I_4$ slack variables, $d_{C1} = d_{safe,C1} - d_{t,l}$, $d_{C2} = d_{t,p} - d_{safe,C2}$, $d_{C3} = d_{t,l} - d_{safe,C3}$, and $d_{C4} = d_{safe,C4} - d_{t,\pi}$.

For all simulations, the trajectory tracking gain was $\eta = 50$, and $\beta = 40$. Whenever present in a given simulation, vector field inequality gains were set at the same value $\eta_{t,l} = \eta_{t,p} = \eta_{t,l} = \eta_{t,\pi} = 0.5$. The safety distances for each constraint were $d_{safe,C1} = 5 \text{mm}$, $d_{safe,C2} = d_{safe,C3} = 14 \text{mm}$, and $d_{safe,C4} = 0$.

VII. RESULTS AND DISCUSSION

The results of the simulation are shown in Fig.3 in terms of the distance between shafts, the distance between the entry point and tool shaft, the distance between the lower point of the shaft and the cylinder centerline, and the distance between the tool tip and the lower plane.

Simulation S1, in which no virtual fixtures were applied caused collisions between the shafts and between the moving tool and the workspace boundaries. Constraining only the shafts in S2 was sufficient to avoid collisions between the shafts, but tool-boundary collisions still occurred. Constraining both the shafts and the entry point in S3 provided a collision-free shaft motion and prevented collisions with the uppermost part of the cylinder. Simulation S4 violated only the plane constraint, and S5 provided a collision-free path. From the point of view of trajectory error, as shown in Fig.4, naturally S1 was able to follow the trajectory closely since no active constraints were applied. All other simulations presented tracking errors given the gradual imposition of restrictions. The trajectories for all the experiments are shown in Fig.5.

In S3, the robot showed a low amplitude vibration near the end of the trajectory, which may occur when the trajectory tracking error is unable to stabilize at the origin. It can be somewhat reduced by tuning $\beta$; however, other solutions, such as regularizing the objective function, are being analyzed in ongoing research.

VIII. CONCLUSIONS

In this paper a novel method for applying active constraints using vector field inequalities was proposed. The method can be used to prevent a robot from entering a restricted zone or for maintaining its location inside a safe area, and is based on a Lyapunov stable solution that can handle both equality and inequality constraints. The vector field inequalities limit the velocity of the robot in the direction of the forbidden zone’s boundary while leaving tangential velocities undisturbed. To use the method, a Jacobian describing the relation between the time derivative of the distance and joint velocities is required. In this work, four of the Jacobians are obtained for relations in which one of the entities is static. Simulation results for two eight-DOF neurosurgical arms showed that, by using the vector field inequalities and relevant primitives, all collisions could be avoided and the robot safely navigated through a restricted workspace.

The proposed method does not require a high-accuracy model of the environment, but the environment must be decomposed in sufficient primitives. However, it still requires high-accuracy calibration between the primitives and the robot. A practical methodology for achieving this is the topic of ongoing work. An additional topic of ongoing work is the tracking of a user generated trajectory, which raises points of discussion such as gain tuning for minimum trajectory tracking error and latency analysis. The implementation on the real platform is currently in progress.

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Fig. 3. Distance between collidable entities. Only S5 provided a collision-safe trajectory, since S4 violated the plane, S3 violated the plane and the lower part of the cylinder, S2 violated the plane and both the upper and lower parts of the cylinder, and S1 violated all constraints.

Fig. 4. Trajectory tracking error (bottom) and 2-norm of joint velocities (top). Only S1 could closely follow the prescribed trajectory as it had no active constraints. In all other simulations, some trajectory tracking error was allowed in order to avoid collisions.

Fig. 5. Tool tip trajectories for all experiments. Notice that S1 is effectively the same as the desired trajectory, and S5 was the only collision-free trajectory.

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