Safe-by-Design Control for Euler-Lagrange Systems

Wenceslao Shaw Cortez, Dimos V. Dimarogonas

Abstract

Safety-critical control is characterized as ensuring constraint satisfaction for a given dynamical system. Recent developments in zeroing control barrier functions (ZCBFs) have provided a framework for ensuring safety of a superlevel set of a single constraint function. Euler-Lagrange systems represent many real-world systems including robots and vehicles, which must abide by safety-regulations, especially for use in human-occupied environments. These safety regulations include state constraints (position and velocity) and input constraints that must be respected at all times. ZCBFs are valuable for satisfying system constraints for general nonlinear systems, however their construction to satisfy state and input constraints is not straightforward. Furthermore, the existing barrier function methods do not address the multiple state constraints that are required for safety of Euler-Lagrange systems. In this paper, we propose a methodology to construct multiple, non-conflicting control barrier functions for Euler-Lagrange systems subject to input constraints to satisfy safety regulations, while concurrently taking into account robustness margins and sampling-time effects. The proposed approach consists of a sampled-data controller and an algorithm for barrier function construction to enforce safety (i.e. satisfy position and velocity constraints). The proposed method is validated in simulation on a 2-DOF planar manipulator.

1 Introduction

Recent technological advancements have increased the presence of autonomous systems in human settings. The push for self-driving cars, drone delivery systems, and automated warehouses are a few examples of how autonomous systems are being exploited to improve efficiency and productivity. However, safety is key to properly incorporate these systems, particularly in human settings. The control of these autonomous systems must be able to guarantee safety of both the device and humans.

Here we are motivated by safety in terms of the regulations provided by the International Standards Organization (ISO), which aim to ensure that machines respect position, velocity, and input constraints, e.g., do not leave a pre-defined region, exceed this speed, or apply excessive force, which must be respected at all times [1]. Simultaneous satisfaction of these position, velocity, and input constraints for Euler-Lagrange systems renders the system safe. Furthermore, these systems are almost always controlled digitally in a sampled-data fashion and are prone to model uncertainties or external disturbances that must be accounted for. The problem addressed here is how to simultaneously satisfy input and state constraints for Euler-Lagrange systems to ensure safety.

Satisfaction of state and input constraints can be accomplished by model predictive control (MPC) or reference governors (RGs). A common setback of MPC is the high computational cost required to repeatedly simulate the system dynamics into the future [2]. RGs can also be applied to handle state and input constraints, but can be conservative as they require the Lyapunov function level sets to be completely contained in the constraint set [3,4]. This prevents the system from touching the constraint boundary. We should note that level set methods, i.e., keeping the system inside a level set of a Lyapunov function, can be developed using, for example, backstepping for an Euler-Lagrange system. However requiring the constraint sets to be defined by level sets of a Lyapunov function can be highly restrictive. Furthermore, conventional MPC and RG methods are tailored towards stability and safety. However in many real-world applications such as human-robot cooperation, the aim is to safely let a human control the system and not just to stabilize to some equilibrium point. Conventional MPC and RGs do not allow arbitrary control inputs to be applied and so may be inappropriate for such settings. We aim to develop a methodology that allows any control input to be implemented in the safe set while guaranteeing safety conditions are met.

Email addresses: w.shawcortez@pnnl.gov (Wenceslao Shaw Cortez), dimos@kth.se (Dimos V. Dimarogonas).

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Control barrier functions have attracted attention for constraint satisfaction of nonlinear systems. Existing barrier function methods have been applied to general nonlinear continuous/hybrid systems [5] and used in control to satisfy constraints while providing stability [6]. Those methods have been extended to less restrictive barrier function definitions and have been applied to bi-pedal walking, adaptive cruise control, and robotics [7–10]. Similar approaches have also addressed high relative degree systems [11] and systems evolving on manifolds [12]. Recently, the distinction between reciprocal control barrier functions (RCBFs) and zeroing control barrier functions (ZCBFs) has been established [13], in which RCBFs are undefined at the constraint boundary while ZCBFs are zero at the boundary and well-defined outside of the constraint set. Aside from practical implementations, ZCBFs are advantageous in that they hold robustness properties in the form of input-to-state stability [14] and are usually implemented as a point-wise minimum-norm control law that allows for any control input to be applied subject to the barrier constraint. A review of existing approaches can be found in [15].

Despite the novel developments in safety-critical methods using ZCBFs, there is no existing constructive procedure to design ZCBFs to satisfy the safety requirements for Euler-Lagrange systems. Recall that here we consider safety to incorporate state (position and velocity) and input constraints. One existing method to handle state and input constraints includes sum-of-squares programming [5,15,16], however that approach is only applicable to polynomial systems, and not to the Euler-Lagrange systems considered here. Another approach for addressing general state and input constraints requires a pre-defined function (referred to as an evasive maneuver) to then construct the ZCBF [15,17]. However the design of the evasive maneuver is not straightforward in general, particularly with dynamically coupled systems such as Euler-Lagrange systems.

Furthermore, a significant setback in existing ZCBF methods is the lack of a systematic design to simultaneously satisfy position and velocity constraints, which is crucial for ensuring safety of the overall system. More specifically, the system should have bounded velocities and slow down as it approaches the boundary of the position constraint set. The existing approaches that address high relative degree [11,12] are prone to singularities in which the velocity is allowed to go unbounded inside the position workspace. These singularities occur when the gradient of the position constraint function is zero inside of the safe set, which prevents bounding the velocity even in standard norm-ball type position constraints (see [18]). Recently, we developed methods to “remove the singularity” in the high order barrier construction [19], however that approach only addresses the control input and not the velocity requirement.

To address the singularity issue, here we consider multiple position constraints (e.g., box constraints), which collectively bounds the position without suffering from singularities and generalizes existing methods to handle multiple constraints simultaneously. The initial idea was presented in [20] where box constraints in the form of multiple ZCBFs were shown to naturally bound the velocity of the system and for which no such singularities occur. However addressing multiple ZCBFs while simultaneously handling input constraints requires ensuring that the multiple constraints are non-conflicting and has received little attention in the literature. In [20], multiple ZCBFs were handled with input constraints in a sampled-data control law, however that method assumed that the controller was feasible. Other existing work has addressed multiple ZCBFs, but cannot handle input constraints [16]. Recently, integral control barrier functions have been proposed as a means to satisfy input constraints [21], however for multiple ZCBFs, there is no guarantee that such a feasible control exists (see Remark 4 of [21]). Finally, recent methods using energy-based barrier functions are in fact able to simultaneously bound the position and velocity using a single barrier function, but insofar those methods cannot handle multiple barrier function nor input constraints [22,23]. Thus despite the advances in safety-critical control, existing methods have yet to provide truly safe controllers for Euler-Lagrange systems.

In this paper, we present a methodology to construct ZCBFs for Euler-Lagrange systems. The proposed approach satisfies multiple workspace constraints (position and velocity), while simultaneously handling input constraints to ensure safety of real-world systems. A correct-by-design algorithm is presented for the ZCBF construction that ensures forward invariance of the safe set, which can be computed off-line. The method considers robustness margins and sampling time effects. The main results are proposed for handling constraints in the form of box constraints, however we demonstrate how the approach can be extended to more general constraint types. The proposed approach is validated in numerical simulation on the 2-DOF planar manipulator. All of the code, including the algorithm to construct the the ZCBFs, is provided in [24]. A preliminary version of this work can be found in [25]. The approach presented here is less conservative than that of [25] and also relaxes the assumptions of [25]. Furthermore, the approach presented here also addresses robustness and sampling terms in the ZCBF construction, which are not considered in [25].

Notation: Throughout this paper, the term $e_j \in \mathbb{R}^r$ denotes the $j$th column of the identity matrix $I_{r \times r}$. The Lie derivatives of a function $h(x)$ for the system $\dot{x} = f(x) + g(x)u$ are denoted by $L_h = \frac{\partial}{\partial x} f(x)$ and $L_h = \frac{\partial}{\partial x} g(x)$, respectively. The terms $\leq$ and $\geq$ are used to denote element-wise vector inequalities. The matrix inequality $A < B$ for square matrices $A$ and $B$ means...
that the matrix $B - A$ is positive-definite. The interior and boundary of a set $A$ are denoted $\bar{A}$ and $\partial A$, respectively. The notation $f \circ g$ for a function $f$ represents the composition of $\beta$. We use the notation $x \prec a$ and $x \succ a$, for some $a \in \mathbb{R}$, to denote the limit as $x$ approaches $a$ from above and below, respectively. A set $\mathcal{N}_p$ for $p \in \mathbb{N}$ is $\mathcal{N}_p = \{1, \ldots, p\}$.

2 Background

2.1 Control Barrier Functions

Here we introduce the existing work regarding ZCBFs for nonlinear affine systems: $\dot{x} = f(x) + g(x)u$, where $x(t) \in \mathbb{R}^n$ is the state, $u \in \mathbb{R}^m$ is the control input, $f : \mathbb{R}^n \to \mathbb{R}^n$ and $g : \mathbb{R}^n \to \mathbb{R}^{n \times m}$ are locally Lipschitz continuous. We denote $\mathcal{I} \subset \mathbb{R}_{\geq 0}$, where $0 \in \mathcal{I}$, as the maximal interval of existence of $x(t)$. A set $S \subset \mathbb{R}^n$ is forward invariant if $x(0) \in S$ implies $x(t) \in S$ for all $t \in \mathcal{I}$.

Let $h(x) : \mathbb{R}^n \to \mathbb{R}$ be a continuously differentiable function, and let the associated constraint set be defined by:

$$
\mathcal{C} = \{x \in \mathbb{R}^n : h(x) \geq 0\}
$$

(1)

The function $h$ is considered the zeroing control barrier function and formerly defined as:

**Definition 1 ([15])** Let $\mathcal{C} \subset \mathcal{E} \subset \mathbb{R}^n$ be the superlevel set of a continuously differentiable function $h : \mathcal{E} \to \mathbb{R}$, then $h$ is a zeroing control barrier function if there exists an extended class-$\mathcal{K}_\infty$ function $\alpha$ such that for the control system $\dot{x} = f(x) + g(x)u$, the following holds: $\sup_{u \in \mathcal{U}} [L_1h(x) + L_2h(x)u] \geq -\alpha(h(x)), \forall x \in \mathcal{E}$

If $h$ is a zeroing control barrier function, the condition $h(x) \geq -\alpha(h(x))$ is then enforced in the control by rewriting it as: $L_1h + L_2h(x)u \geq -\alpha(h(x))$, which is linear with respect to $u$, and ensures forward invariance of $\mathcal{C}$ [15]. We further note that the ZCBF conditions can be extended to sampled-data systems for which $u$ is piece-wise continuous. That is, for a ZCBF $h$ where $h(x) \geq -\alpha(h(x))$ holds for almost all $t \in [0, T]$ (see [20, 26]).

2.2 System Dynamics

Consider the following dynamical system for the generalized coordinates $q, v \in \mathbb{R}^n$:

$$
\begin{align*}
\dot{q} &= v \\
\dot{v} &= G(q) \left( f_1(q,v) + f_2(q,v) + f_3(q) + u \right)
\end{align*}
$$

(2)

where $G(q) \in \mathbb{R}^{n \times m}$ and $f_1(q,v), f_2(q,v), f_3(q) \in \mathbb{R}^m$ are globally Lipschitz continuous functions, and $u \in \mathcal{U} \subset \mathbb{R}^m$ is the control input. For conventional Euler-Lagrange systems, $G$ is the inertia matrix, $f_1$ is the Coriolis term, $f_2$ is the damping term, and $f_3$ is the gravity torque. This more general form of the Euler-Lagrange system allows for application to over-actuated systems, task-space constraints, and nonlinear constraints.

Here we consider the following well-known properties for Euler-Lagrange systems [27]:

**Property 1** : $G(q)$ is full row rank such that there exists a $G^+(q) \in \mathbb{R}^{m \times n}$ for which $GG^+ = I_{n \times n}$.

**Property 2** : There exists $h_\gamma \in \mathbb{R}_{>0}$ such that $\|f_1(q,v)\| \leq h_\gamma \|v\|^2$, $\forall (q,v) \in \mathbb{R}^n \times \mathbb{R}^n$.

**Property 3** : There exist constants $f_j \in \mathbb{R}_{>0}$ such that $\|f_2(q,v)\| \leq f_j \|v\|$, $\forall (q,v) \in \mathbb{R}^n \times \mathbb{R}^n$, $\forall j \in \mathcal{N}_m$.

**Remark 1** Note that Properties 1-3 and global Lipschitz continuity of the dynamics can be relaxed and are only taken for simplicity here. We consider safety of the system for $(q,v) \in \mathcal{H}$ for a compact set $\mathcal{H}$ such that these properties can be checked in a compact set containing $\mathcal{H}$ instead of over the entire space $\mathbb{R}^n \times \mathbb{R}^n$.

2.3 Problem Formulation

The goal of constraint satisfaction is to ensure the states $q, v$ stay within a set of constraint-admissible states. Here we focus on workspace constraints reminiscent of real-world systems which are defined by:

$$
\mathcal{Q} = \{q \in \mathbb{R}^n : q_{\min} \preceq q \preceq q_{\max}\}
$$

(3)

where $q_{\min}, q_{\max} \in \mathbb{R}^n$ and $q_{\max} \succ q_{\min}$. These types of constraints are highly applicable in robotics and general automated systems.

We further address the velocity constraints that the system must satisfy as:

$$
\mathcal{V} = \{v \in \mathbb{R}^n : v_{\min} \preceq v \preceq v_{\max}\}
$$

(4)

where $v_{\min}, v_{\max} \in \mathbb{R}^n$, $v_{\max} \succ 0$, and for simplicity of the presentation let $v_{\min} = -v_{\max}$.

In addition to state constraints, real-world systems have limited actuation capabilities. Thus the aforementioned state constraints must be realizable with the available control inputs. Let $\mathcal{U}$ be the available control inputs:

$$
\mathcal{U} = \{u \in \mathbb{R}^m : u_{\min} \preceq u \preceq u_{\max}\}
$$

(5)

where $u_{\min}, u_{\max} \in \mathbb{R}^m$, $u_{\max} \succ 0$, and for simplicity of the presentation let $u_{\min} = -u_{\max}$.
The problem addressed here is to design a control law that renders the set of state constraints forward invariant. We formally define a safe system as follows:

**Definition 2** Consider the constraint sets (3), (4), and (5). Suppose for a given $u$, (2) with initial condition $(q(0), v(0)) \in \mathbb{R}^2 \times \mathbb{R}^{n}$ admits an absolutely continuous solution $(q(t), v(t))$ for all $t \in I \subset \mathbb{R}_{\geq 0}$. The system (2) is considered safe if for any $(q(0), v(0)) \in Q \times V$, $I \subseteq \mathbb{R}_{\geq 0}$ and $(q(t), v(t)) \in Q \times V$ for all $t \geq 0$.

We note that this definition of safety is stronger than forward invariance of the constraint set as we require forward invariance for all $t \geq 0$. The problem addressed here is formally stated as follows:

**Problem 1** Consider the system (2) with position, velocity, and input constraints (3), (4), (5). Design a control law $u \in \mathcal{U}$ that renders (2) safe.

### 3 Proposed Solution

In this section, we present the candidate ZCBFs and the proposed control laws to ensure safety. We first construct the candidate ZCBFs with design parameters. We proceed to construct bounds on the design parameters such that system safety is ensured under the condition that $u \in \mathcal{U}$. The construction of the design parameters yields an algorithm for constructing ZCBFs. Finally, we design continuous-time and sampled-data control laws to guarantee system safety.

#### 3.1 ZCBF Construction

In this section, we construct the ZCBFs for system safety. We note that the construction is motivated by the approaches from [11, 12], which were later adapted in [19, 20]. To define the ZCBFs, we re-write the constraint set $Q$ into individual constraints with respect to functions $h_i, b_i : \mathbb{R} \rightarrow \mathbb{R}$, which are defined as:

$$
\bar{h}_i(q_i) = \max_{q_{\max}} - q_i, \ b_i = q_i - q_{\min}, \ i \in \mathcal{N}_n
$$

where $q_{\max}, q_{\min} \in \mathbb{R}$ are the $i$th elements of $q_{\max}, q_{\min}$, respectively, from (3). We define the superlevel set of $h_i$ and $b_i$ as:

$$
\mathcal{Q}_i = \{q_i \in \mathbb{R} : \bar{h}_i(q_i) \geq 0, b_i(q_i) \geq 0\}, \ i \in \mathcal{N}_n
$$

It follows that: $\mathcal{Q}_1 \times \ldots \times \mathcal{Q}_n = \mathcal{Q}$.

In order to define a superset of $Q$ over which the ZCBF conditions hold, we introduce the following functions:

$$
\tilde{h}_i(q_i) = h_i(q_i) + \delta, \ \tilde{b}_i(q_i) = b_i(q_i) + \delta, \ i \in \mathcal{N}_n
$$

where $\delta \in \mathbb{R}_{\geq 0}$ is a design parameter. We similarly define a superlevel set for $\tilde{h}_i$ and $\tilde{b}_i$ as:

$$
\mathcal{Q}^\delta_i = \{q_i \in \mathbb{R} : \tilde{h}_i^\delta(q_i) \geq 0, \tilde{b}_i^\delta(q_i) \geq 0\}, \ i \in \mathcal{N}_n
$$

Note that $\mathcal{Q}_i \subset \mathcal{Q}^\delta_i$ for $\delta > 0$ and $\mathcal{Q}_i = \mathcal{Q}^0_i$ if $\delta = 0$. Let $\mathcal{Q}^\delta = \mathcal{Q}^\delta_1 \times \ldots \times \mathcal{Q}^\delta_n$, and we note that $\mathcal{Q} = \mathcal{Q}^0$. Moreover, consideration of $\mathcal{Q}^\delta$ for $\delta > 0$ allows for consideration of robustness to perturbations in the proposed formulation. We refer to [14] for a discussion on robustness of ZCBFs.

We now introduce new functions to address the relative-degree of the system: $b_i, \ b_i : \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R}$, and we treat these functions as the candidate ZCBFs for Euler-Lagrange systems defined as follows:

$$
\bar{b}_i(q_i, v_i) = -v_i + \gamma \alpha(h_i(q_i)), \ b_i(q_i, v_i) = v_i + \gamma \alpha(b_i(q_i)), \ i \in \mathcal{N}_n
$$

where $\alpha$ is a continuously differentiable, extended class-$\mathcal{K}$ function, and $\gamma \in \mathbb{R}_{\geq 0}$ is a design parameter. We see that when $b_i \geq 0$ and $\bar{b}_i \geq 0$ it follows that $\dot{\bar{h}}_i \geq -\gamma \alpha(h_i)$, $\dot{b}_i \geq -\gamma \alpha(b_i)$ as required from Definition 1 for forward invariance of $\mathcal{Q}_i$.

We treat $\bar{b}_i \geq 0$ and $b_i \geq 0, \forall i \in \mathcal{N}_n$ as new constraints to be satisfied. To properly address the set of states where $\bar{b}_i \geq 0$ and $b_i \geq 0$, we define the following set:

$$
\mathcal{B}_i = \{(q_i, v_i) \in \mathbb{R} \times \mathbb{R} : \bar{b}_i(q_i, v_i) \geq 0, b_i(q_i, v_i) \geq 0\}
$$

with $\mathcal{B} = \mathcal{B}_1 \times \ldots \times \mathcal{B}_n$.

We define the following functions to define supersets of $\mathcal{B}$:

$$
\bar{b}_i^\delta(q_i, v_i) = -v_i + \gamma \alpha(\tilde{b}_i^\delta(q_i)), \ b_i^\delta(q_i, v_i) = v_i + \gamma \alpha(\tilde{b}_i^\delta(q_i)), \ i \in \mathcal{N}_n
$$

with the following superlevel sets:

$$
\mathcal{B}^\delta = \{(q_i, v_i) \in \mathbb{R} \times \mathbb{R} : \bar{b}_i^\delta(q_i, v_i) \geq 0, b_i^\delta(q_i, v_i) \geq 0\}
$$

and $\mathcal{B}^\delta = \mathcal{B}_1^\delta \times \ldots \times \mathcal{B}_n^\delta$. By construction, it follows that $\mathcal{B}^\delta \supset \mathcal{B}$ for $\delta > 0$ and $\mathcal{B}^0 = \mathcal{B}$.

Next we define the intersection of $\mathcal{Q}_i$ and $\mathcal{B}_i$ and the respective superset as:

$$
\mathcal{H}_i := (\mathcal{Q}_i \times \mathbb{R}) \cap \mathcal{B}_i, \ i \in \mathcal{N}_n
$$

$$
\mathcal{H}_i^\delta := (\mathcal{Q}_i^\delta \times \mathbb{R}) \cap \mathcal{B}_i^\delta, \ i \in \mathcal{N}_n
$$

with $\mathcal{H} = \mathcal{H}_1 \times \ldots \times \mathcal{H}_n$ and $\mathcal{H}^\delta = \mathcal{H}_1^\delta \times \ldots \times \mathcal{H}_n^\delta$ such that $\mathcal{H}^\delta \supset \mathcal{H}$ if $\delta > 0$ and $\mathcal{H}^0 = \mathcal{H}$. We denote $\mathcal{H}$ as the safe set. Note that since $\mathcal{Q}$ is compact, so are $\mathcal{H}_i$ and $\mathcal{H}_i^\delta$ for $i \in \mathcal{N}_n$. 

4
In order to ensure forward invariance of $B$ (and thus $Q$), we repeat the ZCBF conditions as per Definition 1 with respect to the ZCBF candidates $b_i, b_i^*$:

\[
\begin{align*}
\dot{b}_i(q_i, v_i) &\geq -\nu \beta(b_i(q_i, v_i)) + \tilde{\eta}, \\
\dot{b}_i(q_i, v_i) &\geq -\nu \beta(b_i(q_i, v_i)) + \tilde{\eta}
\end{align*}
\]

for all $(q_i, v_i) \in \mathcal{H}_i^\delta, i \in \mathcal{N}_n$, where $\beta$ is an extended class-$K_\infty$ function, $\nu \in \mathbb{R}_{>0}$ is an additional barrier function design parameter, and $\tilde{\eta} \in \mathbb{R}_{\geq 0}$ is an added term motivated by [20] to incorporate sampling-time effects into the proposed ZCBF construction. We note that for $\tilde{\eta} := 0$, (16) follows the conventional requirements for ZCBF's [15].

Substitution of (2) into (16) and concatenation over all $i \in \mathcal{N}_n$ yields:

\[
SG(q)(f_1(q, v) + f_2(q, v) + f_3(q, u) + u) + \gamma \Lambda(q) S v 
\geq -\nu p(q, v) + \tilde{\eta} \mathbf{1}_{2n}
\]

where $S = [-I_{n \times n}, I_{n \times n}]^T$, $\Lambda(q) = \text{diag} \{ \partial \tilde{h}_i / \partial q_i \}$, $\partial \tilde{h}_i / \partial q_i = [\beta \circ b_1(q_1, v_1), ..., \beta \circ b_n(q_n, v_n)]$ and $p(q, v) := \beta \circ b_1(q_1, v_1), ..., \beta \circ b_n(q_n, v_n)$.\(^{17}\)

To summarize, satisfaction of (17) for all $(q, v) \in \mathcal{H}_i^\delta \supseteq \mathcal{H}$ for some $\delta > 0$ ensures (16) holds for all $(q_i, v_i) \in \mathcal{H}_i^\delta, i \in \mathcal{N}_n$, which in turn ensures $q_i \in Q_i$ for all $i \in \mathcal{N}_n$.

We note that (17) is linear with respect to $u$, and define the proposed quadratic program-based control law:

\[
\begin{align*}
u^\ast(q, v, t) &= \text{arg} \min_{u \in U} \|u - u_{\text{nom}}(q, v, t)\|_2^2 \\
& \text{s.t. (17)}
\end{align*}
\]

where $u_{\text{nom}} : \mathbb{R}^n \times \mathbb{R}^n \times \mathbb{R} \to \mathbb{R}^n$ is some nominal control law which can represent, for example, a pre-defined stabilizing controller or possibly a human input to the system. Implementation of (18), assuming a solution exists, can be used to ensure forward invariance of $\mathcal{H}$. We emphasize that the construction of the barriers leads to linearly dependent input constraints as can be seen in (18) due to $S$ and the input constraints. This is in contrast with the QP-based control formulations, typical of ZCBF controllers, which require linear independence of the constraints on $u$ [28, 29]. For continuous-time implementations, linear independence of the QP constraints is important for ensuring local Lipschitz continuity of $u^\ast$. Here, we allow for discontinuous, sampled-data control, which is impartial to linear dependency in the control constraints since the control is inherently non-Lipschitz.

Before we present the main theorem, we must state two assumptions to be satisfied. First, we make the following realistic assumption that the system has sufficient control authority in the set $Q_3^\delta$:

**Assumption 1** There is sufficient control authority such that for given $\delta, \tilde{\eta} \in \mathbb{R}_{\geq 0}$, there exists some $\epsilon \in \mathbb{R}_{>0}$ such that $u_{\text{max}} > |e_j^T f_i(q)| + |(\epsilon + \tilde{\eta}) \|e_j^T G(q)\|_\infty |$ for all $q \in Q_3^\delta, j \in \mathcal{N}_m$.

This is a common assumption to ensure that in fact the system can be held statically and has the capability to move from any configuration over $Q_3^\delta$. From a pragmatic perspective, we note that this assumption is always satisfied in practice for the system to perform a desired task. Furthermore, this assumption is much less conservative than that of [25], which effectively requires each $u_i$ to satisfy (16) independently while all other inputs are at their respective maximum values (i.e., $|u_j| = u_{\text{max}}$, for all $j \neq i$).

Second, we require the extended class-$K_\infty$ functions $\alpha$ and $\beta$ to satisfy the following properties:

**Assumption 2** Given a $\delta \in \mathbb{R}_{>0}$, the extended class-$K_\infty$ functions $\alpha$ and $\beta$ satisfy the following conditions:

1. There exists a $d \in \mathbb{R}_{>0}$ such that $\alpha(-c) + \alpha(q_{\text{max}} - q_{\text{min}} + c) \geq d$ holds for all $i \in \mathcal{N}_n, c \in [0, \delta]$.
2. For any $a, c \in \mathbb{R}_{>0}$ and $b \in \mathbb{R}_{>0}$ such that $a - b = 2c$, then $\beta$ satisfies: $\beta(a) + \beta(-b) \geq \beta(c)$.

Assumption 2 requires that the slope of $\alpha$ and $\beta$ on the negative real-axis is sufficiently small with respect to that of the positive real-axis. This condition is required to consider how the system behaves in $\mathcal{H}_i^\delta \setminus \mathcal{H}$. For example, if a disturbance exists that pushes the system into $\mathcal{H}_i^\delta \setminus \mathcal{H}$ (where $b_i < 0$ or $b_i < 0$), the restoring “force” that keeps the system ultimately bounded [20] must not exceed the capabilities of the actuators.

**Remark 2** Assumption 2 is not restricted to linear functions used in “exponential barrier functions” [11], nor polynomial functions used in sum-of-squares programming techniques [5]. Assumption 2 only restricts the slope of the two extended class-$K_\infty$ functions over the negative real-axis. As a result of this generality, both linear functions and (odd) polynomial functions are subclasses of functions that satisfy Assumption 2.

In the following theorem, we ensure a solution to (18) always exists for all $(q, v) \in \mathcal{H}_i^\delta$ by appropriately computing $\gamma$ and $\nu$.

**Theorem 1** Consider the system (2) with the state and input constraints defined by (3), (4), and (5). Let the set $\mathcal{H}_i^\delta$ be defined by (15) for $i \in \mathcal{N}_n$ with the continuously differentiable extended class-$K_\infty$ function $\alpha$ and extended class-$K_\infty$ function $\beta$. Suppose Assumptions 1 and 2 hold
for sufficiently small $\delta^2$, $\bar{\eta} \in \mathbb{R}_{\geq 0}$. Then there exist $\gamma^*_1$, $\gamma^*_2$, $\gamma^*_3$, $\nu^*_1$, $\nu^*_2 \in \mathbb{R}_{>0}$ (with $\nu^*_1 < \nu^*_2$) such that the choice of $\gamma \in (0, \min\{\gamma^*_1, \gamma^*_2, \gamma^*_3\}) \in [\nu^*_1, \nu^*_2]$ if $\delta > 0$ otherwise $\nu \geq \nu^*_1$ if $\delta = 0$, ensures that $u^*$ defined by (18) exists and is unique for all $(q, v) \in \mathcal{H}^i$. Furthermore, for any $(q, v) \in \mathcal{H}^i$, $v \in \mathcal{V}$.

The proof of Theorem 1 is constructive. In the following section, we analyze the properties of $b_i$ and $\bar{b}_i$ for Euler-Lagrange systems to construct $\gamma^*_1$, $\gamma^*_2$, $\gamma^*_3$, $\nu^*_1$, and $\nu^*_2$, as well as bounds on $\delta$, $\bar{\eta}$, and a valid control $\tilde{u} \in \mathcal{U}$ such that there always exists a solution to (18).

3.2 Analysis

In this section, we present properties of $\mathcal{H}^i$ in relation to the candidate ZCBFs $b_i$, and $\bar{b}_i$ to design $\gamma$ and $\nu$.

3.2.1 Velocity Relations

First, we state the following Lemma to relate the system velocity with $\mathcal{H}^i$:

**Lemma 1** Consider the functions and sets $h^i$, $\bar{b}^i$, $\bar{b}_i$, $Q^i$, $\mathcal{B}^i$, and $\mathcal{H}^i$ defined respectively by (8), (12), (9), (13), (15) with extended class $\mathcal{K}_\infty$ function $\alpha$. Then for $\delta > 0$, $\|v\|_{\infty} \leq \bar{v}$ for all $(q, v) \in \mathcal{H}^i$, where

$$\bar{v} = \gamma a := \gamma a(2\delta + \|q_{\min} - q_{\max}\|_{\infty})$$

**PROOF.** From (9), (12), and (13) it follows that $-\gamma a(h_i(q_i)) \leq v_i \leq \gamma a(h_i(q_i))$ for $(q_i, v_i) \in \mathcal{H}^i, i \in \mathcal{N}_n$. Thus $v$ is bounded in $\mathcal{H}^i$. Furthermore, the maximum value of $\alpha(h_i(q_i))$ and $\alpha(h_i(q_i))$ in $Q^i$ for $i \in \mathcal{N}_n$ is $a = \alpha(2\delta + \|q_{\min} - q_{\max}\|_{\infty})$, which yields $\gamma a$ as the maximum value of $v$ and completes the proof.

Lemma 1 provides insight into how the ZCBF construction affects the system behaviour. First, by appropriately tuning $\gamma$, the velocity bounds from (19) can be adjusted to satisfy the state constraint $v \in \mathcal{V}$. Second, the relation $-\gamma a(h_i(q_i)) \leq v_i \leq \gamma a(h_i(q_i))$ shows that as $q$ approaches the boundary $\partial Q$, the velocity approaches zero. This is an important property because it restricts the system’s inertia relative to the constraint boundary. This aligns with intuition in that if the velocity is too high near the boundary, exceedingly large control effort would be required to ensure forward invariance. While $\gamma$ dictates the system’s velocity, $\nu$ dictates the behaviour of $u$ as the system approaches the constraint boundary.

From (16), $v$ will dictate how soon the control acts to keep the system in the constraint set.

From Lemma 1, we define the following upper bound on $\gamma$ such that the maximum velocity will be contained in $\mathcal{V}$ to ensure safety:

$$\gamma^*_1 := \frac{1}{\bar{q} \in \mathcal{N}_n} \min_{q_i \in Q^i} v_{\max}$$

where $a \in \mathbb{R}_{>0}$ is defined in (19).

**Lemma 2** Consider the functions and sets $h^i$, $\bar{b}^i$, $\bar{b}_i$, $Q^i$, $\mathcal{B}^i$, $\mathcal{H}^i$, and $\mathcal{V}$ defined respectively by (8), (12), (9), (13), (15), (4) with extended class $\mathcal{K}_\infty$ function $\alpha$. If $\delta \in \mathbb{R}_{\geq 0}$, then $\gamma^*_1$ defined by (20) is strictly positive, and if $\gamma \in (0, \gamma^*_1]$, then $v \in \mathcal{V}$ for all $(q, v) \in \mathcal{H}^i$.

**PROOF.** Strict positivity of $\gamma^*_1$ follows since $v_{\max} > 0$ and for $\delta > 0$, $a > 0$ due to $q_{\max} > q_{\min}$. From Lemma 1 it follows that $\|v\|_{\infty} \leq \gamma a$. To ensure $v \in \mathcal{V}$, i.e., $v_{\min} \leq v \leq v_{\max}$, we must ensure $\gamma$ is sufficiently small such that $\bar{v}$ from (19) is smaller than the minimum component of $v_{\max}$. Note that we are only concerned with $v_{\max}$ since $v_{\min} = -v_{\max}$. More precisely, for $\gamma \in (0, \gamma^*_1]$, $\|v\|_{\infty} \leq \bar{v} \leq \gamma a \leq \min_{i \in \mathcal{N}_n} v_{\max}$, which implies that $\max_i |v_i| \leq \min_{i \in \mathcal{N}_n} v_{\max}$. Thus $v \leq v_{\max}$. In similar fashion, it follows that $v_{\min} \leq v$. Since this holds for all $(q, v) \in \mathcal{H}^i$, the proof is complete.

3.2.2 Satisfaction of Input Constraints

Next, we will construct a $\tilde{u} \in \mathcal{U}$ to show that there always exists a solution to (17). However to do so, we must introduce some notation and additional terms. First we present $\rho : Q^i \rightarrow \mathbb{R}$:

$$\rho(q_i) := \frac{\gamma}{2} \left( \alpha(h_i(q_i)) + \alpha(h_i(q_i)) \right)$$

The function $\rho(q_i)$ defines the level set that divides $\mathcal{B}_i$ (see Figure 1). More specifically, the manifold defined by $\{q_i, v_i \in \mathcal{H}^i : b_i - \rho = 0\} = \{q_i, v_i \in \mathcal{H}^i : b_i - \rho = 0\}$ is the level set for which $b_i = b_i$. Furthermore if $b_i \leq \rho(q_i)$ then $\rho^2$ and vice versa. We denote the lower bound of $\rho(q_i)$ over $Q^i$ as

$$\rho^0 := \min_{i \in \mathcal{N}_n} \left\{ \min_{q_i \in Q^i} \rho(q_i) \right\}$$

**Lemma 3** Suppose the conditions of Theorem 1 hold. Consider $\rho(q_i)$ and $\rho^0$ defined by (21) and (22), respectively, for a given $\gamma > 0$, $\delta \geq 0$ for $q_i \in Q^i, i \in \mathcal{N}_n$.  

---

2 We note that Definition 1 requires $\mathcal{E} \supset \mathcal{C}$ which, equivalently stated, requires $\delta > 0$. For the sake of generality we show that the results of Theorem 1 hold for $\delta = 0$, although in Section 3.3 we also require $\delta > 0$.  

---
Then $\rho(q_i)$ is strictly positive, and there exists a $c \in \mathbb{R}_{>0}$ such that $\rho^3 \geq c$.

**Proof.** First, we show $\rho(q_i)$ is always strictly positive in $Q_i \subseteq Q_i^3$. From $b_i \geq 0$, $b_i \geq 0$, and $\gamma > 0$, then $\alpha(h_i(q_i))$ only equals $0$ at the boundary when $q_i = q_{\text{max}}$, and $\alpha(h_i(q_i))$ only equals $0$ at the boundary when $q_i = q_{\text{min}}$. Evaluation at both boundaries yields $\rho(q_{\text{max}}) = \rho(q_{\text{min}}) = \frac{1}{2} \alpha(e_i)$, for $e_i = q_{\text{max}} - q_{\text{min}}$. Since $q_{\text{max}} > q_{\text{min}}$, $e_i > 0$. Now in the interior of $Q_i$ (i.e. $q_{\text{min}} < q_i < q_{\text{max}}$), $\alpha(h_i)$ and $\alpha(b_i)$ are strictly positive. Thus there exists no such $q_i \in Q_i$ such that $\rho(q_i) = 0$. Since $\rho$ is a continuous function on the compact set $Q_i$, and is strictly positive, there exists some $q_i \in \mathbb{R}_{>0}$ such that $\rho(q_i) \geq q_i$ in $Q_i$. We note that $q_i$ is independent of $\delta$.

Next, for when $\delta > 0$ and $Q_i \subset Q_i^3$, we divide $Q_i^3$ into two sections: a) when $q_i \geq q_{\text{max}}$, and b) when $q_i < q_{\text{min}}$. For $Q_i^3 \setminus Q_i$ where $q_i > q_{\text{max}}$, let $q_i = q_{\text{max}} + e$ for $e \in [0, \delta]$ such that $\rho(e) = \frac{1}{2} (\alpha(-e) + \alpha(q_{\text{max}} - q_{\text{min}} - e))$. Then from Assumption 2 it follows that $\rho(q_i) \geq d$. Finally, for $Q_i \setminus \{q_i \mid q_i < q_{\text{min}}\}$, let $q_i = q_{\text{min}} + e$ for $e \in [0, \delta]$ similarly, it follows that $\rho(e) = \frac{1}{2} (\alpha(e) + \alpha(q_{\text{max}} - q_{\text{min}} - e))$, and again from Assumption 2, $\rho(q_i) \geq d$. Thus there exists some $\bar{d}_i = \min\{d, g_i\}$, $\bar{d}_i \in \mathbb{R}_{>0}$ such that $\rho(q_i) \geq \bar{d}_i$ on $Q_i^3$. Let $c$ be the minimum of $\bar{d}_i$ for $i \in \mathbb{N}_n$. By definition of $\rho^3$, it follows that $\rho^3 \geq c$.

The following Lemma ensures that the sum of $b_i$ and $\bar{b}_i$ is always positive on $H^\delta$.

**Lemma 4** Suppose the conditions of Theorem 1 hold, $\gamma > 0$, $\delta > 0$, and consider $\rho(q_i)$ from (21). Then $b_i(q_i, v_i) + \bar{b}_i(q_i, v_i) = 2\rho(q_i) > 0$ for all $(q_i, v_i) \in H^\delta$. Furthermore, if $b_i(q_i, v_i) < \rho(q_i)$, then $b_i(q_i, v_i) < \rho(q_i)$, and if $b_i(q_i, v_i) > \rho(q_i)$, then $b_i(q_i, v_i) > \rho(q_i)$.

**Proof.** Substitution of (10) into $b_i + \bar{b}_i$ yields $b_i + \bar{b}_i = 2\rho(q_i)$. From Lemma 3, $\rho$ is strictly positive. Thus it follows that $b_i = 2\rho - \bar{b}_i > 0$ if $\bar{b}_i < \rho$, and $\bar{b}_i = 2\rho - b_i > \rho$ if $b_i < \rho$.

Second, we introduce $\zeta^i_\delta \in \mathbb{R}$:

$$\zeta^i_\delta := \min\{ \min_{(q_i, v_i) \in H^\delta} b_i, \min_{(q_i, v_i) \in H^\delta} \bar{b}_i \} \tag{23}$$

The term $\zeta^i_\delta$ is the lower bound of $b_i$ and $\bar{b}_i$ on $H^\delta$. We denote the lower bound of $\zeta^i_\delta$ over $i \in \mathbb{N}_n$ as:

$$\zeta^\delta = \min_{i \in \mathbb{N}_n} \zeta^i_\delta \tag{24}$$

**Lemma 5** Suppose the conditions of Theorem 1 hold, and consider $\zeta^i_\delta, \zeta^\delta$ defined by (23) and (24), respectively, for $\gamma > 0$ and $\delta > 0$. Then $\zeta^\delta$ always exists, is non-positive, and as $\delta \searrow 0$, $\zeta^\delta \nearrow 0$.

**Proof.** A solution for $\zeta_i$ always exists since $b_i$ and $\bar{b}_i$ are continuous functions over the compact set $H^\delta$. Furthermore, with $\gamma > 0$, $\delta > 0$, there exists a coordinate $(q_{\text{max}}, \delta, 0) \in H^\delta$ for which $b_i = -\gamma \alpha(\delta) \leq 0$. Similarly the coordinate $((q_{\text{min}}, -\delta, 0)) \in H^\delta$ ensures $b_i = -\gamma \alpha(\delta) \leq 0$. Since by definition (23), $\zeta^i_\delta$ is the minimum value of the minimum of $b_i$ and $\bar{b}_i$ and we have specified coordinates in $H^\delta$ for which $b_i$ and $\bar{b}_i$ are non-positive, it follows that $\zeta^\delta$ must also be non-positive.

Next, from the proof of Lemma 1, it follows that $-\nu_i \geq -\gamma \alpha(q_{\text{max}}, -\delta)$ and $\nu_i \geq -\gamma \alpha(q_i, q_{\text{min}}, +\delta)$. Thus from (10), it follows that $b_i(q_i, v_i) \geq \bar{f}_i(q_i) := -\gamma \alpha(q_{\text{max}}, -\delta) + \gamma \alpha(q_{\text{max}}, q_i) + \bar{b}_i(q_i, v_i) \geq f_i(q_i) := -\gamma \alpha(q_i, q_{\text{min}, 0}) + \gamma \alpha(q_i, q_{\text{min}})$. Thus we can re-write (23) as:

$$\zeta^i_\delta := \min\{ \min_{q_i \in Q_i^3} \bar{f}_i(q_i), \min_{q_i \in Q_i^3} f_i(q_i) \} \tag{25}$$

By inspection of $f_i$ and $\bar{f}_i$, it follows that $\zeta^i_\delta = 0$ when $\delta = 0$. Furthermore, $f_i$ and $\bar{f}_i$ are non-positive, continuous, and strictly decreasing functions of $\delta$ since $\alpha$ is an extended class-$K^\infty$ function and $\delta > 0$. Thus as $\delta \searrow 0$, $f_i \nearrow 0$ and $\bar{f}_i \searrow 0$. Since $\zeta^\delta$ is the minimum of $f_i$ and $\bar{f}_i$ over $Q_i^3$, it follows that as $\delta \searrow 0$, $\zeta^\delta \nearrow 0$. Finally, since this property holds for all $i \in \mathbb{N}_n$, it also holds for $\zeta^\delta$, which completes the proof.

**Remark 3** The computation of $\zeta^\delta_i$ can be done offline as it is purely a function of the choice of $\alpha$. We explicitly define $\zeta^\delta_i$ for the following commonly used choices for $\alpha$ for $\alpha(h) = h$, $\zeta^\delta_i = -\gamma \delta$, for $\alpha(h) = \tan^{-1}(h)$, $\zeta^\delta_i = -2\gamma (\sqrt{2}) \frac{1}{\sqrt{1 + h^2}}$, and for $\alpha(h) = h^3$, $\zeta^\delta_i = \gamma (\alpha(q_{\text{max}}, q_{\text{min}}, +\delta) - \alpha(q_{\text{max}}, q_{\text{min}}, +\delta))$.

Finally, we divide $H^\delta_i$ into eight regions which are outlined in Table 1, and depicted in Figure 1. We note that Figure 1 shows the desired property that the velocity $v_i$ approaches zero as the position $q_i$ approaches the boundary of $Q_i$, and the velocities are bounded for all $q_i \in Q_i$. We are now ready to present a candidate $u \in U$ that satisfies (17) and $\bar{u} \in U$:

$$u(q, v) := G^+(q) (\mu(q, v) + \chi(q, v) + \psi(q, v)) - f_i(q, v) - f_2(q, v) - f_3(q) \tag{26}$$
Fig. 1. Depiction of $\mathcal{H}_i$ (outlined by dashed lines) and $\mathcal{H}_i^\delta$ (outlined by solid black lines). The subsets of $\mathcal{H}_i^\delta$ are: I (grey), II (blue), III (light blue), IV (orange), V (green), VI (yellow), VII (red dash-dotted line), and VIII (yellow dash-dotted line). ZCBF parameters used in this example: $VI$ (yellow), VII (red dash-dotted line), and VIII (yellow).

Table 1

| Decomposition of $\mathcal{H}_i^\delta$ |
|----------------------------------------|
| I = \{(q, v_i) \in \mathcal{H}_i^\delta : \bar{b}_i(q, v_i) \in [0, \rho(q_i)) \land v_i \geq 0\} |
| II = \{(q, v_i) \in \mathcal{H}_i^\delta : \bar{b}_i(q, v_i) \in [0, \rho(q_i)) \land v_i < 0\} |
| III = \{(q, v_i) \in \mathcal{H}_i^\delta : \bar{b}_i(q, v_i) \in [0, \rho(q_i)) \land v_i \leq 0\} |
| IV = \{(q, v_i) \in \mathcal{H}_i^\delta : \bar{b}_i(q, v_i) \in [0, \rho(q_i)) \land v_i \geq 0\} |
| V = \{(q, v_i) \in \mathcal{H}_i^\delta : \bar{b}_i(q, v_i) < 0\} |
| VI = \{(q, v_i) \in \mathcal{H}_i^\delta : \bar{b}_i(q, v_i) > 0\} |
| VII = \{(q, v_i) \in \mathcal{H}_i^\delta : \bar{b}_i(q, v_i) = \rho(q_i) \land v_i \geq 0\} |
| VIII = \{(q, v_i) \in \mathcal{H}_i^\delta : \bar{b}_i(q, v_i) = \rho(q_i) \land v_i \leq 0\} |

where

\[
\mu_i(q, v_i) := \begin{cases} 
-\frac{\partial \alpha}{\partial h_i}(q_i) v_i, & \text{if } (q_i, v_i) \in I \cup IV \cup VII, \\
0, & \text{if } (q_i, v_i) \in II \cup III, \\
-\frac{\partial \alpha}{\partial h_i}(q_i) v_i, & \text{if } (q_i, v_i) \in IV \cup VI \cup VII \cup VIII 
\end{cases}
\]  

\[
\chi_i(q, v_i) := \begin{cases} 
0, & \text{if } (q_i, v_i) \in I \cup II \cup III \cup IV \cup VII \cup VIII, \\
\nu \beta(b_i(q_i, v_i)), & \text{if } (q_i, v_i) \in V, \\
-\nu \beta(b_i(q_i, v_i)), & \text{if } (q_i, v_i) \in VI 
\end{cases}
\]

\[
\psi_i(q, v_i) := \begin{cases} 
-\bar{\eta}, & \text{if } (q_i, v_i) \in I \cup II \cup V, \\
\eta, & \text{if } (q_i, v_i) \in III \cup IV \cup VI, \\
0, & \text{if } (q_i, v_i) \in VII \cup VIII
\end{cases}
\]

\[
\chi(q, v) = [\chi_1(q_1, v_1), ..., \chi_n(q_n, v_n)]^T, \quad \mu := [\mu_1(q_1, v_1), ..., \mu_n(q_n, v_n)]^T, \quad \text{and } \psi := [\psi_1(q_1, v_1), ..., \psi_n(q_n, v_n)]^T.
\]

We note that $\tilde{\mathbf{u}}$ is well-defined over all of $\mathcal{H}^\delta$. Furthermore, $\tilde{\mathbf{u}}$ is discontinuous over $\mathcal{H}^\delta$. We address discontinuities in a sampled-data fashion as will be discussed later.

Our first task is to ensure that $\tilde{\mathbf{u}} \in \mathcal{U}$ for all $(q, v) \in \mathcal{H}^\delta$. We do this by bounding $\gamma$ using:

\[
\gamma^2 = \min_{q \in Q^\delta} \min_{j \in N_m} \frac{-d_j(q) + \sqrt{d_j^2 - 4c_j(q)}}{2}
\]

where

\[
d_j(q) = \frac{f_j}{\| \epsilon_j^T G(q)^+ \|_{\infty}^2 y(q) + k_c a^2},
\]

\[
c_j(q) = \frac{\| \epsilon_j^T f_j(q) + (\epsilon + \bar{\eta}) \| \epsilon_j^T G(q)^+ \|_{\infty}^2 - u_{max,j}}{\| \epsilon_j^T G(q)^+ \|_{\infty}^2 y(q) a + k_c a^2},
\]

\[
y(q) = \max_{i \in N_n} \left\{ \frac{\partial \alpha}{\partial h_i}(q_i), \frac{\partial \alpha}{\partial h_i}(q_i) \right\} \quad \text{and } f_j \in \mathbb{R} \text{ is from Property 3.}
\]

Similarly, we define the upper bound $\nu^*_2$ to ensure $\| \mathbf{x} \|_{\infty} \leq \varepsilon$ to respect actuator constraints in $\mathcal{H}^\delta \setminus \mathcal{H}$:

\[
\nu^*_2 := \frac{\varepsilon}{|\beta'(\zeta^2)|}
\]

In the event that $\delta = 0$, then clearly $\nu^*_2 = \infty$, which implies that the choice of $\nu$ is not upper bounded.

Satisfaction of $\tilde{\mathbf{u}} \in \mathcal{U}$ is formally stated in the following Lemma:

**Lemma 6** Suppose the conditions of Theorem 1 hold. Consider $\tilde{\mathbf{u}} : \mathcal{H}^\delta \rightarrow \mathbb{R}^n$ defined by (26), $\gamma^2$ defined by (30) and $\nu^*_2$ defined by (33) with $\varepsilon$ from Assumption 1. Then $\gamma^2$ always exists and is strictly positive, and $\nu^*_2$ is always strictly positive and bounded if $\delta > 0$, otherwise $\nu^*_2 = \infty$ if $\delta = 0$. Furthermore, if $\gamma \in (0, \gamma^*_2), \nu \in (0, \nu^*_2)$ for $\delta > 0$ then $\nu > 0$ if $\delta = 0$, then $\tilde{\mathbf{u}} \in \mathcal{U}$ for all $(q, v) \in \mathcal{H}^\delta$.

**PROOF.** We start with ensuring existence of strictly positive $\gamma^2$ and $\nu^*_2$. Existence and positivity of $\nu^*_2$ follows trivially from (33) and Assumption 1 for $\delta > 0$. If $\delta = 0,$
then $\nu_2^* = +\infty$ follows trivially from (33). Since we chose $\varepsilon$ from Assumption 1, it follows that $c_j(q) < 0$ in (30), and so $\gamma_2^*$ is real and positive.

Now we ensure the satisfaction of the actuator constraints $\bar{u} \in \mathcal{U}$. Since $u_{\max} = -u_{\min}$, we write the actuator constraint condition as $|u_j| - u_{\max,j} \leq 0$ for all $j \in N_m$. Substitution of $\bar{u}$ into $|u_j| - u_{\max,j} \leq 0$ yields:

$$|e_j^T (G^* \mu(q, v) + \chi(q, v) + \psi(q, v)) - f_1(q, v) - f_2(q, v) + f_3(q)| - u_{\max,j} \leq 0$$

First we consider the case $\delta > 0$ such that $\nu_2^* < \infty$. By choice of $\nu \in (0, \nu_2^*)$, $|\beta(\xi^*)| \leq \varepsilon$. It straightforward to see that the lower bound on $b_\eta$ is reached in $V_\varepsilon$ when $b_\eta < 0$, and similarly $b_\eta$ reaches its lower bound in VI when $b_\eta < 0$, for $i \in N_\varepsilon$. From (23) and (24) it follows that $|\beta(b_\eta)| \leq |\beta(\xi^*)|$, $|\beta(b_\eta)| \leq |\beta(\xi^*)|$ in V and VI, respectively. From (28), in I–IV, VII, and VIII, $x_i = 0$. In V, $|\beta(b_\eta)| \leq |\beta(\xi^*)| \leq \varepsilon$. In VI, $|\beta(b_\eta)| \leq |\beta(\xi^*)| \leq \varepsilon$. Thus $|x_i| \leq \varepsilon$ on $H_\varepsilon$. It is also straightforward to see that $|\nu| \leq \eta$.

From Properties 2, 3, and Lemma 1, it follows that for all $(q, v) \in H_\varepsilon^\delta$, $f_i(q, v) \leq k_\varepsilon |v| \leq k_\varepsilon \varepsilon = k_\varepsilon \varepsilon a^2$ and $|e_j^T f_2(q, v)| \leq f_j(v) \leq f_j(v) a^2$, for $j \in N_m$. By definition of $y(q)$, it follows that $|\mu| \leq \gamma^2 y(q) a$. Substitution of $f_i(q, v) \leq f_i(v) \leq f_i(v) a^2$, for $i \in N_m$, $|\mu| \leq \gamma^2 y(q) a$, $|e_j^T f_2(q, v)| \leq f_j(v) a^2$, and application of the triangle inequality yields the following sufficient condition for guaranteeing that $\bar{u} \in \mathcal{U}$:

$$\gamma^2 \left(\|e_j^T G(q)^\delta\|_\infty y(q) a + k_\varepsilon \varepsilon a^2 \right) + \gamma f_j a + |e_j^T f_3(q)| \leq 0$$

Application of the standard quadratic formula to solve for $\gamma$ (at equality) for all $j \in N_m$ yields (30). Thus if $\gamma = \gamma_2^*$, then $\bar{u} \in \mathcal{U}$. Furthermore, it is trivial to see that any $\gamma \in (0, \gamma_2^*)$ also ensures $\bar{u} \in \mathcal{U}$. In the event that $\delta = 0$, then the sets V and VI are in fact empty. Thus $\chi_0 = 0$ on $H_\varepsilon^\delta$, which satisfies $|x_i| \leq \varepsilon$ and the previous existence ensures that if $\gamma \in (0, \gamma_2^*)$, $\nu > 0$, then $\bar{u} \in \mathcal{U}$.

### 3.2.3 Non-Conflicting ZCBFs

Next, we design $\gamma_3^*$, $\delta^*$, $\nu_1^*$, and $\eta^*$ to ensure non-conflicting ZCBF conditions. The candidate ZCBFs require the $2n_c$ conditions from (16) to be satisfied at all times on $H_\varepsilon^\delta$. We substitute (26) into (16), which yields:

$$\mu_i + \chi_i + \psi_i + \frac{\partial c_i}{\partial b_i} b_i - \nu \beta(b_i) + \eta \leq 0$$

$$\nu = \gamma_3^* \frac{2L_a}{\beta(\rho^*)}$$

for $i \in N_c$. Thus satisfaction of (34) and (35) over all $i \in N_c$ ensures (16) holds. We must now ensure there are no conflicting conditions such that $\bar{u}$ can satisfy (34) and (35) simultaneously for all $i \in N_c$.

We now define the following upper bound $\gamma_3^*$ to prevent conflict in (17):

$$\gamma_3^* := \frac{\varepsilon}{La}$$

where $a$ is defined in (19) and $L \in \mathbb{R}_{>0}$ is the Lipschitz constant of $a$ for all $h_i(q_i), b_i(q_i)$ for all $q \in \Omega^\delta$.

Next we design the lower bound $\nu_1^*$ to ensure there always exists a control in $H_\varepsilon \setminus \mathcal{H}$ to satisfy the ZCBF conditions:

$$\nu_1^* := \frac{\gamma_3^* L_a}{\beta(\rho^*)}$$

In the following Lemma we show that for a sufficiently small $\delta$ and choice of $\gamma \in (0, \gamma_3^*)$, the previous designs of $\nu_1^*, \nu_2^*$ are well-defined such that $\nu_1^* < \nu_2^*$:

**Lemma 7** Suppose the conditions of Theorem 1 hold and consider $\gamma_3^*$, $\nu_1^*$, $\nu_2^*$ defined, respectively, by (36), (37), (33), for $\delta \geq 0$. Then $\gamma_3^*$ always exists and is strictly positive. Furthermore, for $\gamma \in (0, \gamma_3^*)$, there exists a $\delta^* \in \mathbb{R}_{>0}$ that satisfies the following conditions:

$$|\beta(\xi^*)| < \beta(\rho^*) \quad \forall \delta \in [0, \delta^*)$$

and for $\delta \in [0, \delta^*)$, $\nu_1^* > 0, \nu_2^* > 0, \nu_1^* < \nu_2^*$.

**PROOF.** First, we ensure $\gamma_3^*$ is strictly positive. Since $\alpha$ is continuously differentiable there always exists a Lipschitz constant $L > 0$ and with $\alpha > 0$ it is straightforward to see that $\frac{\varepsilon}{La}$, and thus $\gamma_3^*$, is strictly positive.

Existence of (38) follows from Lemmas 5 and 3 and the fact that $\beta$ is an extended class-$K_{\infty}$ function such that as $\delta \to 0$, $|\beta(\rho^*)| \to 0$. Furthermore, since $\rho^\delta \geq c$ from Lemma 3, there exists a sufficiently small $\delta^* \in \mathbb{R}_{>0}$ such that $|\beta(\rho^*)| < \beta(c) \leq \beta(\rho^*)$. Let $\delta^* = \delta$. Since $\rho^\delta$ is lowered bounded by $c$ and $\xi^*$ will continue to approach 0, it follows that the choice of $\delta^*$ satisfies (38).

Next, we show $\nu_1^*$ is well-defined such that $\nu_1^* < \nu_2^*$. Since $\rho(q_i)$ (and thus $\rho^\delta$) is strictly positive from Lemma 3, $\nu_1^*$ is strictly positive. For $\gamma \in (0, \gamma_3^*)$, it follows that $\nu_1^* = \frac{\gamma_3^* L_a}{\beta(\rho^*)} \leq \frac{\varepsilon}{\beta(\rho^*)}$. Now for $\delta \in [0, \delta^*)$, it follows that $|\beta(\xi^*)| < \beta(\rho^*)$ such that $\nu_1^* \leq \frac{\varepsilon}{\beta(\rho^*)} < \frac{\varepsilon}{|\beta(\xi^*)|} := \nu_2^*$. 

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The final component to the proper design of $\gamma$ and $\nu$ is the design of $\bar{\eta}$. Recall that $\bar{\eta}$ is an added robustness margin to handle sampling time effects. In this respect, $\bar{\eta}$ must be sufficiently small (i.e., the sampling frequency must be sufficiently fast) such that no conflict occurs when attempting to simultaneously satisfy (34) and (35). We define the upper bound on $\bar{\eta}$ as:

$$\eta^* := \frac{\nu \beta(\rho^2) - \gamma^2 La}{2}$$

(39)

**Lemma 8** Suppose the conditions of Theorem 1 hold and consider $\gamma^*_1, \nu^*_1, \nu^*_2$ defined, respectively, by (36), (37), (33), for $\gamma > 0$, $\delta \geq 0$. If $\delta \in [0, \delta^*]$, $\gamma \in (0, \gamma^*_1]$, $\nu \in [\nu^*_1, \nu^*_2]$ for $\delta > 0$ otherwise $\nu \geq \nu^*_1$ if $\delta = 0$, then $\eta^*$ is non-negative. Furthermore, if $\nu > \nu^*_1$ then $\eta^*$ is strictly positive.

**PROOF.** By Lemma 7, it follows that $\nu^*_1 < \nu^*_2$. For $\nu \geq \nu^*_1$, then $\nu \geq \frac{\gamma^2 La}{\beta(\rho^2)}$ and it follows that $\nu \beta(\rho^2) - \gamma^2 La \geq 0$. Thus $\eta^*$ from (39) must be non-negative. Similarly if $\nu > \nu^*_1$ then $\nu > \frac{\gamma^2 La}{\beta(\rho^2)}$ and so $\nu \beta(\rho^2) - \gamma^2 La > 0$, and so $\eta^*$ is strictly positive.

The following Lemma shows that the choice of $\gamma \in (0, \gamma^*_1], \nu \in [\nu^*_1, \nu^*_2]$, and $\bar{\eta} \in [0, \eta^*]$ prevents conflict between the ZCBF conditions:

**Lemma 9** Suppose the conditions of Theorem 1 hold and consider $\gamma^*_1, \nu^*_1, \nu^*_2, \delta^*$, and $\eta^*$ defined, respectively, by (36), (37), (33), (38), (39). For $\delta \in [0, \delta^*]$, $\gamma \in (0, \gamma^*_1]$, $\nu \in [\nu^*_1, \nu^*_2]$ for $\delta > 0$ otherwise $\nu \geq \nu^*_1$ if $\delta = 0$, and $\bar{\eta} \in [0, \eta^*]$, then the following conditions are always satisfied:

$$-\gamma \left( \frac{\partial \alpha}{\partial h_i}(q_i) - \frac{\partial \alpha}{\partial h_j}(q_i) \right) v_i - 2\bar{\eta} + \nu \beta(\rho^2) > 0, \quad \forall (q_i, v_i) \in \mathcal{H}^\delta_i$$

(40)

$$\gamma \frac{\partial \alpha}{\partial h_i}(q_i)v_i - 2\bar{\eta} + \nu \beta(\rho^2) \geq 0, \quad \forall (q_i, v_i) \in \mathcal{H}^\delta_i$$

(41)

$$\gamma \frac{\partial \alpha}{\partial h_i}(q_i)v_i + 2\bar{\eta} - \nu \beta(\rho^2) \leq 0 \quad \forall (q_i, v_i) \in \mathcal{H}^\delta_i$$

(42)

Note that the requirements of Lemma 9 are the main components to avoid conflict such that (40) and (41) always hold simultaneously. The formal guarantees of non-conflicting conditions are found in the following proof of Theorem 1.

We are now ready to present the proof of Theorem 1:

**PROOF.** [Proof of Theorem 1] We must show that there exists a $u \in U$ such that (17) holds for all $(q, v)$ in $\mathcal{H}^\delta$. The proof is composed of four parts. First, we ensure the existence of $\gamma^*_1$, $\gamma^*_2$, $\gamma^*_3$, $\nu^*_1$, $\nu^*_2$, and define the upper bounds on $\delta$ and $\eta$. Second, we show that a candidate $u \in U$ is well-defined in $\mathcal{H}^\delta$. Third, we ensure that $v \in \mathcal{V}$. Finally, we show that $u$ satisfies (17) on $\mathcal{H}^\delta$.

1) Let $\gamma^*_1$, $\gamma^*_2$, $\gamma^*_3$ be defined by (20), (30), and (36), respectively. For $\delta$, $\eta \geq 0$ satisfying Assumption 1, it follows that $\gamma^*_1$ exists and is strictly positive from (20). Lemmas 6 and 7 ensure $\gamma^*_2$ and $\gamma^*_3$ always exist and are strictly positive. Lemma 6 also ensures $\nu^*_2$ exists and is strictly positive for $\delta > 0$, and otherwise $\nu^*_2 = +\infty$ if $\delta = 0$. For $\gamma \in (0, \min(\gamma^*_1, \gamma^*_2, \gamma^*_3)]$, Lemma 7 ensures that $\delta^*$ is well-defined and strictly positive. We restrict $\delta$ such that $\delta \in [0, \delta^*]$. Now Lemma 7 ensures $\nu^*_1$ is strictly positive and $\nu^*_1 < \nu^*_2$. Finally, Lemma 8 ensures that for
\( \nu \in [\nu^*_1, \nu^*_2] \) if \( \delta > 0 \) otherwise \( \nu \geq \nu^*_1 \) if \( \delta = 0 \), \( \eta^* \) is non-negative. We restrict \( \tilde{\eta} \) such that \( \tilde{\eta} \in [0, \eta^*] \). We restrict \( \bar{\eta} \) such that \( \bar{\eta} \in [0, \eta^*] \), it follows that \( \bar{\eta} \leq 2\bar{\eta} + \gamma_2 La \leq \nu \beta(\bar{\rho} \delta) \). Thus 
\[-\bar{\eta} + \nu \beta(\bar{\rho} \delta) \geq 0 \]
and (35) holds.

We note that the above inequality holds due to Assumption 2 since \( \bar{b}_i + \bar{b}_i = 2p(q_i) \) (via Lemma 4), \( \bar{b}_i < 0 \) in V, and thus \( \beta(\bar{b}_i) + \beta(\bar{b}_i) \geq \beta(p(q_i)) \). Since \( \rho \geq \rho^p \), (35) is satisfied from Lemma 9.

Again, the above inequality holds due to Lemma 4 and Assumption 2 such that \( -\beta(\bar{b}_i) - \beta(\bar{b}_i) \leq -\beta(p(q_i)) \). Thus (35) holds from Lemma 9. The left-hand-side of (35) equals 0 and so (35) is satisfied.

The left-hand-side of (35) with the substitution of \( -\bar{\eta} \geq -2\bar{\eta} \) and \( \bar{b}_i = \rho \geq \rho^p \) is greater than or equal to the left-hand-side of (40), and thus Lemma 9 ensures (35) holds.

The left-hand-side of (35) with the substitution of \( \bar{\eta} \leq 2\bar{\eta} \) and \( \bar{b}_i = \rho \geq \rho^p \) (see Lemma 4) is less than or equal to the negative of the left-hand-side of (40), such that (34) holds via Lemma 9.

The left-hand-side of (35) yields \( -\bar{\eta} + \nu \beta(\bar{b}_i) = -\bar{\eta} + \nu \beta(\rho^p) \). Again, from (39) and since \( \bar{\eta} \in [0, \eta^*] \), it follows that \( \bar{\eta} \leq 2\bar{\eta} + \gamma^2 La \leq \nu \beta(\rho^p) \). Thus 
\[-\bar{\eta} + \nu \beta(\rho^p) \geq 0 \] and (35) holds.
Finally, since (34) and (35) hold for all $i \in \mathbb{N}_n$, $\tilde{u} \in \mathcal{U}$ is a valid control law to enforce (17) over $H^\delta$. This implies that there always exists at least one point-wise solution to $u^*$ from (18), namely $\tilde{u}$. Due to the linearity in the constraints and positive-definiteness of the cost function in (18), the solution to $u^*$ is uniquely defined [31]. Thus for any $(q, v) \in H^\delta$, there always exists a unique, point-wise solution to (18), and $v \in \mathcal{V}$.

**Remark 4** Theorem 1 ensures each $b_i$ satisfies the conditions of Definition 1 on the set $H^\delta$ and explicitly uses $\delta$ in the derivation of $\gamma$ and $\nu$. The use of $\delta$ shows how robustness can be incorporated into the control design while respecting input constraints. In the set $H^\delta \setminus H$, the system (2) with (18) is asymptotically stable to the safe set $H$ [14]. In other words, for a sufficiently small, bounded perturbation (e.g. from model uncertainty) the system will be contained in $H^\delta$.

The proof of Theorem 1 is constructive and provides insight into designing $\gamma$, $\nu$ to ensure there always exists a solution to (18). As discussed in Remark 4, the proposed design considers both constraints on the available control input and robustness with respect to bounded perturbations and sampling time effects. The full ZCBF design is outlined in Algorithm 1.

**Algorithm 1** Control Barrier Function Design

1: procedure ZCBF DESIGN($\alpha$, $\beta$, $\delta_0 \geq 0$, $\eta_0 \geq 0$)
2: Determine $\varepsilon$ satisfying Assumption 1 for $\delta_0$, $\eta_0$
3: Compute $\gamma_1^\delta$, $\gamma_2^\delta$, $\gamma_3^\delta$ from (30), (36), and (20).
4: Choose $\gamma \in (0, \min\{\gamma_1^\delta, \gamma_2^\delta, \gamma_3^\delta\})$.
5: Compute $\xi^\delta$, $\rho^\delta$, $\bar{v}$ from (24), (22), (19).
6: if $\delta_0 > 0$ then
7: \hspace{1em} if (38) holds for all $\delta \in [0, \delta_0]$ then
8: \hspace{2em} Let $\delta^* = \delta_0$
9: \hspace{1em} else
10: \hspace{2em} Find $\delta^* \in (0, \delta_0]$ satisfying (38)
11: \hspace{2em} end if
12: \hspace{1em} end if
13: else if $\delta_0 = 0$ then
14: \hspace{1em} Find $\delta^* > 0$ satisfying (38)
15: \hspace{1em} Set $\delta = \delta_0 = 0$
16: \hspace{1em} end if
17: Compute $\nu^\delta_1$, $\nu^\delta_2$ from (37), (33) respectively.
18: if $\delta > 0$ then
19: \hspace{1em} Choose $\nu \in [\nu^\delta_1, \nu^\delta_2]$
20: else if $\delta = 0$ then
21: \hspace{1em} Choose $\nu \geq \nu^\delta_1$
22: end if
23: Compute $\eta^*$ from (39)
24: if $\eta_0 = 0$ then
25: \hspace{1em} Set $\eta = 0$
26: else
27: \hspace{1em} Choose $\eta \in (0, \min\{\eta_0, \eta^*\})$
28: end if
29: end procedure

**Remark 5** Algorithm 1 presents a guaranteed method of designing ZCBFs for Euler-Lagrange systems with input constraints. The most computationally expensive component involves the computation of $\gamma_3^\delta$ which requires searching over all $q \in \mathcal{Q}^\delta$. We note however that the proposed approach requires significantly less computation compared to searching over the entire set $H^\delta$. An alternative, albeit more conservative, approach is to bound the terms $G(q)^+$ and $f_3(q)$ by their respective bounds on $H^\delta$, as done in [25].

The following corollary ensures the use of Algorithm 1 always ensures a solution to (18) exists:

**Corollary 1** Consider the system (2) with the state and input constraints defined by (3), (4), and (5). Given a continuously differentiable extended class-$\mathcal{K}_\infty$ function $\alpha$, extended class-$\mathcal{K}_\infty$ function $\beta$, $\delta_0 \in \mathbb{R}_{>0}$, and $\eta_0 \in \mathbb{R}_{>0}$ that satisfy Assumptions 1 and 2, Algorithm 1 will always output a $\gamma, \nu \in \mathbb{R}_{>0}, \delta, \eta \in \mathbb{R}_{>0}$. Additionally if $\delta_0 > 0$, then $\delta$ from Algorithm 1 is strictly positive, and if $\eta_0 > 0$, then $\eta$ from Algorithm 1 is strictly positive. Furthermore, for this choice of $\alpha, \beta, \gamma, \nu, \delta, \eta$, let $H^\delta$ be defined by (15). Then there always exists a solution to (18) for any $(q, v) \in H^\delta$.

**PROOF.** The proof follows directly from the construction of the ZCBF parameters from Theorem 1.

### 3.3 Control Implementation

Theorem 1 ensures the proposed control (18) is well-posed in that there always exists a unique solution to $u^*$ over $H^\delta$. In this section, we present a sampled-data form of $u^*$ and ensure forward invariance of $H$ of the system (2).

To introduce the sampled-data formulation, we denote $q_k = q(t = t_k)$ and $v_k = v(t = t_k)$ as the sampled states at time $t_k \in \mathbb{R}_{>0}$ for $k \in \mathbb{N}$ and sampling period $T \in \mathbb{R}_{>0}$. To ensure satisfaction of (16) between sampling times, we formally define $\eta(T)$ as [20]:

$$
\eta(T) := \frac{(c_1 + c_2 + c_3 c_4) c_5}{c_1 + c_2 c_4} \left( e^{(c_1 + c_2 c_4) T} - 1 \right)
$$

where $c_1 \in \mathbb{R}_{>0}$ is the Lipschitz constant associated with the Lipschitz continuous function: $G(q)(f_1(q, v) + f_2(q, v) + f_3(q))$, $c_2 \in \mathbb{R}_{>0}$ is the Lipschitz constant for the extended class-$\mathcal{K}_\infty$ function $\beta$, $c_3 \in \mathbb{R}_{>0}$ is the Lipschitz constant for the Lipschitz continuous function $G(q)$, $c_4 := \max_{x \in \mathcal{X}, u \in \mathcal{U}_{\text{max}}}$, and $c_5 := k_{\text{max}}(k_c V^2 + \cdots)$.

Less conservative bounds can be substituted for $\eta(T)$ in this framework so long as the new $\eta(T)$ is of class-$\mathcal{K}$. 


In regards to the analysis in Section 3.2, \( \eta(T) \) is substituted for \( \tilde{\eta} \). In this context, the sampling time \( T \) is considered a design parameter and the chosen \( \tilde{\eta} \in (0, \eta^*] \) defines the maximum allowable sampling frequency for the control law. The use of \( \eta(T) \), as explained in [20], is to keep the solution \( (q(t), v(t)) \) “close enough” to \( (q_k, v_k) \) for \( t \in [t_k, t_{k+1}] \). This then prevents unsafe behaviour during sampling times. We note that \( \eta(T) \) is a class-\( \mathcal{K} \) function, which fits with intuition in that as \( T \) increases, a larger robustness margin \( \eta \) is required to keep the system safe.

The proposed sampled-data control law is:

\[
\begin{aligned}
    u^*_k(q_k, v_k, kT) &= \arg\min_{u \in \mathcal{U}} \| u - u_{\text{nom}}(q_k, v_k, kT) \|_2^2 \\
    \text{s.t. } & S\mathcal{G}(q_k)(f_1(q_k, v_k) + f_2(q_k, v_k) \\
    &+ f_3(q_k) + \bar{u}) + \gamma \Lambda(q_k)Sv_k \geq -\nu p(q_k, v_k) + \eta(T)1_{2n}
\end{aligned}
\]

(44)

Here \( u^*_k \) is the ZCBF-based control law which satisfies a sampled-order hold condition between sampling times.

In the following theorem, we ensure safety of the system (2) under (44):

**Theorem 2** Consider the system (2) with the state and input constraint sets defined by (3), (4), and (5). Let the sets \( Q^\delta_1, B^\delta_1 \), and \( H^\delta_1 \) be defined by (9), (13), and (15), respectively, for \( i \in \mathcal{N}_n \) with the continuously differentiable extended class-\( \mathcal{K}_\infty \) function \( \gamma \) and extended class-\( \mathcal{K}_\infty \) function \( \nu \). Consider \( \gamma_1^*, \gamma_2^*, \gamma_3^*, \nu_1^*, \nu_2^*, \delta^*, \eta^* \) defined, respectively, by (20), (30), (36), (37), (33), (38), (39). Let \( \eta(T) \) be defined by (43) for a given sampling time \( T \in \mathbb{R}_{>0} \). Suppose Assumptions 1 and 2 hold for a sufficiently small \( \delta, \eta \in \mathbb{R}_{>0} \) and let \( u_{\text{nom}} : H^\delta_1 \times \mathbb{R} \rightarrow \mathbb{R}^m \) be a given nominal control law. Let \( \delta \in (0, \delta^*], \gamma \in (0, \min\{\gamma_1^*, \gamma_2^*, \gamma_3^*\}) \), \( \nu \in (\nu_1^*, \nu_2^*) \), and further suppose \( T \) is small enough such that \( \eta(T) \in (0, \eta^* \leq \eta^* \). Then \( u^*_k \) defined by (44) exists and is uniquely defined in \( H^\delta \). Furthermore, if \( \beta \circ b_1, \beta \circ b_2 \) are locally Lipschitz continuous on \( H^\delta_1 \) for all \( i \in \mathcal{N}_n \) and \( (q(0), v(0)) \in H \), then (2) under (44) is safe.

**Proof.** By Theorem 1, \( u^*_k \) always exists and is uniquely defined on \( H^\delta \). Due to the Lipschitz properties of the closed-loop system (2) under \( u^* \) on \( H^\delta \), for \( (q(t), v(t)) \in H \) Theorem 3.1 of [33] ensures there exists a time \( \delta t_1 \in \mathbb{R}_{>0} \) such that \( (q(t), v(t)) \) is uniquely defined on \( [0, \delta t_1] \). Since the controller \( u^* \) enforces the ZCBF conditions of (34), Brezis Theorem (Theorem 4 of [34]) ensures that \( (q(t), v(t)) \) is defined for all \( t \in [0, \delta t_1] \). Repeated application of Brezis Theorem ensures then that \( q(t) \in Q \) on \( [0, \delta t_1] \) and so \( (q(t), v(t)) \) is safe for all \( t \in [0, \delta t_1] \). Since \( H \) is a compact set we can extend the forward invariance interval to \( [0, \infty) \) as in Theorem 2 and the proof is complete.

**Remark 6 (Extension to general constraints)**

The generality in the proposed methodology allows for...
other systems that can be formulated as (2) and other nonlinear constraints, which are typical in robot applications (e.g., task-space constraints). To extend to multiple nonlinear constraints, consider a system defined by (2) for the generalized coordinates $\tilde{q}, \tilde{v}$ and a twice-continuously differentiable constraint function $c(\tilde{q}) : \mathbb{R}^n \rightarrow \mathbb{R}^n$ with full rank gradient $\nabla c \in \mathbb{R}^{n \times n}$ and locally Lipschitz Hessian. Now let $q = c(\tilde{q})$, $v = \nabla c^T \tilde{v}$ such that $\dot{v} = \frac{\partial}{\partial \tilde{q}} \nabla c^T \tilde{v} + \nabla c^T \tilde{v}$. We can now address nonlinear constraints of the form (3), (4), and (5) for the transformed system. It is straightforward to see that with the full rank assumption that allows for an invertible $\nabla c$ and the bounded Hessian of $c$, the dynamics of the transformed system can be written as (2) and still satisfy Properties 1-3. An example of our method for multiple nonlinear constraints will be provided in the next section. In the context of task-space constraint satisfaction for robotics, this transformation requires singular configurations to not be elements of the safe set $H^0$. However, unlike many existing methods, our approach enforces the condition that singular configurations can never be reached instead of simply assuming this to be true.

4 Numerical Examples

Here we demonstrate the proposed technique in simulation on a 2-DOF planar manipulator. The simulations were performed in Python and the code used for these results along with Algorithm 1 is available at [24]. We note that the results presented here are accompanied with the corresponding simulation file to recreate the results.

4.1 Scenario 1

The manipulator consists of two identical links with a length of 1 m and mass of 1 kg, which are parallel to the ground such that $g = 0$. The system is equipped with motors capable of $u_{max1} = u_{min1} = 18$ Nm, and $u_{max2} = u_{min2} = 10$ Nm of torque. The system damping is $F = 0.001 I_{2\times2}$ kg/s. Let the position/velocity safety constraints be defined by $q_{max1} = q_{min1} = \pi/2$ rad, $q_{max2} = 5\pi/6$ rad, $q_{min2} = \pi/2$ rad, and $v_{max1,2} = v_{min1,2} = 1.5$ rad/s. We choose the following extended class-$K$ functions for the ZCBFs: $\alpha_1(h) = \tan(h)^{-1}$, $\alpha_2(h) = h^3$. The nominal control is the computed torque control law: $u_{nom} = M(q_2)(\ddot{q} - \dot{e} - e) + Cv$ [35] where $e = q - \dot{r}$ and $r = [3.1708\sin(1.3t), 2.6236\sin(1.3t) + 2.0944]^T$ is the reference that attempts to move the system outside of $Q \times \dot{V}$ and $U$. This nominal control is used to represent a pre-defined control law or equivalently a human that is incorrectly operating the system. The reader is directed to [24] for all simulation parameters used.

First, we compare the proposed technique presented with the preliminary, more conservative method from [25] in continuous time. Figure 2 shows three system trajectories. The first, depicted in orange, is the system (2) subject to the nominal control law, $u_{nom}$, alone. As shown, the nominal control results in violation of all system and input constraints. The second trajectory, depicted in green, shows the result of the system (2) subject to the proposed control (18) (in continuous time) using the ZCBF parameters constructed from [25] (“ZCBF_control_exp1.yaml” from [24]). The resulting trajectories show satisfaction of all state and input constraints, while attempting to track the nominal control law. This implementation ensures safety, however significant conservativeness is seen by the distances between the trajectories and state/input constraints. The third trajectory, depicted in blue, shows the system (2) subject to the proposed control (18) using the ZCBF parameters constructed from Algorithm 1 (“ZCBF_control_exp2.yaml” from [24]). The output of Algorithm 1 for the simulations in Figure 2 is: $\gamma = 1.17$, $\delta = 0.1$, $\nu = 2473.70$, $\eta = 0.0$ for the input parameters: $\alpha(h) = \tan^{-1}(h)$, $\beta(h) = h^2$, $\delta_0 = 0.1$. As shown, the controller ensures safety of the overall system, but is also less conservative than the approach from [25]. One
The difference between the ZCBF parameter construction between [25] and Algorithm 1 lies in computation. The method in [25] only requires the associated bounds from Properties 1-3 and a bound on $f_3$ and scales well with the number of degrees of freedom. Algorithm 1 on the other hand is dependent on searching over some dynamic terms of (2) over $Q^k$. This results in larger computational effort, but yields less conservative behaviour as seen in Figure 2. By less conservative behaviour, we mean that the state trajectories more closely approach the state constraints for a more aggressive system response.

Next, we note that the results shown in Figure 2 were developed using the continuous time control law (18). However, this is dependent on the assumption of local Lipschitz continuity of $u^*$, which is not guaranteed in general. Indeed, under certain parameter configurations (see “ZCBF_control_exp2_discrete.yaml”) the system leaves the safe set as a result of discontinuities in the control. When discontinuities occur, $\dot{q} > 0$ is required to account for jumps in the control law to ensure forward invariance of the safe set. However, the sampled-data control law (44) is able to ensure forward invariance of the safe set for $T = 0.001$ s (see “ZCBF_control_exp2_discrete.yaml”). The results of the system trajectory subject to the sampled-data controller and ZCBF parameters from Algorithm 1 are shown in Figure 3. The output of Algorithm 1 for the simulations in Figure 3 is: $\gamma = 0.52$, $\delta = 0.01$, $\nu = 4.57 \times 10^6$, $\eta = 6.26$ for the input parameters: $\alpha(h) = \tan^{-1}(h)$, $\beta(h) = h^3$, $\delta_0 = 0.01$, $\eta_0 = 7.0$.

Figure 3 shows the proposed, sampled-data control $u_k^*$ enforcing state constraints, while always respecting input constraints. The effect of incorporating $\dot{\eta} > 0$ into the control design does impose some conservativeness in the system behaviour. This can be seen by comparing the blue curves between Figures 2 and 3. The state trajectories resulting from the sampled-data control do not approach the state constraints as closely as that of the continuous-time controller.

4.2 Scenario 2: Nonlinear Constraints

We next provide an example of how the proposed methodology can be applied to nonlinear constraints. All model parameters are the same as from Scenario 1 except now the position is bounded by the intersection of ellipsoids and planes, which are presented as follows. We re-define the original joint angles of the 2-DOF manipulator as $q \in \mathbb{R}^2$. Let $c(q) = [c_1(q), c_2(q)]^T$ for $c_1(q) = -1+(q-q_1)^T P(q-q_1), c_2(q) = q_r^T q$ with $q_1 = [5,0]^T$, $q_r = [0.1, 1.0]^T$, and $P = \text{diag}([-1,0,0,-1])$. Now we define the (transformed) system state as $\bar{q} = c(q)$ and define the constraints via (3) with $q_{min} = [8,1.7]^T$ and $q_{max} = [12,2.5]^T$. A picture of the constraint set $Q$ in the joint space, i.e. $Q = \{q \in \mathbb{R}^2 : c(q) \in \mathbb{Q}\}$ can be seen in Figure 4a. We leave it to the reader to derive the system dynamics using the proposed transformation, but note that $\nabla c$ is full rank for all $\bar{q} \in Q$.

The same velocity and input bounds are used as in Scenario 1. The reference signal for this scenario is: $r = [0.5 \sin(0.5t) + 0.8, 0.5 \sin(1.0t) + 2]^T$, which yields a figure-eight trajectory (see Figure 4a), and all other parameters associated with Scenario 2 can be found under “ZCBF_control_exp5_nonlinear.yaml” in [24].

Algorithm 1 was used to derive the following correct-by-design ZCBF parameters: $\gamma = 1.13$, $\delta = 0.01$, $\nu = 12.7 \times 10^6$, $\eta = 7.32$ for the input parameters: $\alpha(h) = \tan^{-1}(h)$, $\beta(h) = h^3$, $\delta_0 = 0.01$, $\eta_0 = 8.0$. Figure 4 shows a comparison between the nominal tracking controller and the proposed sampled-data controller. As shown in the plots, the proposed control enforces the multiple nonlinear position constraints, while respecting bounds on the (transformed) velocity, and the input constraints simultaneously. The plots show that the proposed control attempts to implement the nominal controller as much as possible, but deviates as necessary to satisfy all the safety constraints.

Finally, we note some caveats associated with Algorithm...
5 Conclusion

In this paper, we designed multiple, non-conflicting ZCBFs to ensure safety of Euler-Lagrange systems. The design takes into account actuator limitations, robustness margins, and sampling time effects. The proposed design yielded an algorithm to compute safe-by-design ZCBF parameters. A sampled-data controller was presented to enforce safety of the Euler-Lagrange system. The proposed approach was demonstrated in simulation on a 2 DOF planar manipulator. Future work will consider simultaneous safety and stability as well as the use of data-based methods to further improve system performance.

References

[1] Robots and robotic devices – Safety requirements for industrial robots. Part 1: Robots; Part 2: Robot systems and integration, ISO 10218-1-2011 Std., 2011.
[2] L. Grüne and J. Pannek, Nonlinear Model Predictive Control, ser. Communications and Control Engineering. London: Springer, 2017.
[3] M. Hosseinzadeh and E. Garone, “An explicit reference governor for the intersection of concave constraints,” IEEE Trans. Autom. Control, vol. 65, no. 1, pp. 1–11, 2020.
[4] E. Garone, S. Di Cairano, and I. Kolmanovsky, “Reference and command governors for systems with constraints: A survey on theory and applications,” Automatica, vol. 75, pp. 306–328, 2017.
[5] L. Grüne and J. Pannek, Nonlinear Model Predictive Control, ser. Communications and Control Engineering. London: Springer, 2017.
[6] K. P. Tee, S. S. Ge, and E. H. Tay, “Barrier lyapunov functions for the control of output-constrained nonlinear systems,” Automatica, vol. 45, no. 4, pp. 918–927, 2009.
[7] A. D. Ames, J. W. Grizzle, and P. Tabuada, “Control barrier function based quadratic programs with application to adaptive cruise control,” in Proc. IEEE Conf. on Decision and Control, New York, 2014, pp. 6271–6278.
[8] S. C. Hsu, X. Xu, and A. D. Ames, “Control barrier function based quadratic programs with application to bipedal robotic walking,” in Proc. American Control Conf., 2015, pp. 4542–4548.
[9] M. Rauscher, M. Kimmel, and S. Hirche, “Constrained robot control using control barrier functions,” in Proc. IEEE/RSJ Int. Conf. Intel. Robot. Syst., 2016, pp. 279–285.
[10] W. Shaw Cortez, D. Oetomo, C. Manzie, and P. Choong, “Grasp constraint satisfaction for object manipulation using robotic hands,” in Proc. IEEE Conf. on Decision and Control, 2018, pp. 415–420.
[11] Q. Nguyen and K. Sreenath, “Exponential control barrier functions for enforcing high relative-degree safety-critical constraints,” in Proc. American Control Conf., 2016, pp. 322–328.
[12] G. Wu and K. Sreenath, “Safety-critical and constrained geometric control synthesis using control lyapunov and control barrier functions for systems evolving on manifolds,” in Proc. American Control Conf., 2015, pp. 2038–2044.
[13] A. Ames, X. Xu, J. Grizzle, and P. Tabuada, “Control barrier function based quadratic programs for safety critical systems,” *IEEE Trans. Autom. Control*, vol. 62, no. 8, pp. 3861–3876, 2017.

[14] X. Xu, P. Tabuada, J. W. Grizzle, and A. D. Ames, “Robustness of control barrier functions for safety critical control,” in *Proc. IFAC Conf. Anal. Design Hybrid Syst.*, vol. 48, 2015, pp. 54–61.

[15] A. D. Ames, S. Coogan, M. Egerstedt, G. Notomista, K. Sreenath, and P. Tabuada, “Control barrier functions: Theory and applications,” in *Proc. European Control Conf.*, 2019, pp. 3420–3431.

[16] X. Xu, J. W. Grizzle, P. Tabuada, and A. D. Ames, “Correctness guarantees for the composition of lane keeping and adaptive cruise control,” *IEEE Trans. Autom. Sci. Eng.*, vol. 15, no. 3, pp. 1216–1229, 2018.

[17] E. Squires, P. Pierpaoli, and M. Egerstedt, “Constructive barrier certificates with applications to fixed-wing aircraft collision avoidance,” in *Proc. IEEE Conf. Control Tech. and Applications*, 2018, pp. 1656–1661.

[18] F. Barbosa, L. Lindemann, D. V. Dimarogonas, and J. Tumova, “Provably safe control of lagrangian systems in obstacle-scattered environments,” in *IEEE Conference on Decision and Control*, 2020, pp. 2056–2061.

[19] X. Tan, W. Shaw Cortez, and D. V. Dimarogonas, “High-order barrier functions: Robustness, safety and performance-critical control,” *IEEE Transactions on Automatic Control*, pp. 1–8, 2021, early access.

[20] W. Shaw Cortez, D. Oetomo, C. Manzie, and P. Choong, “Control barrier functions for mechanical systems: theory and application to robotic grasping,” *IEEE Trans. Control Syst. Technol.*, pp. 1–16, 2019.

[21] A. D. Ames, G. Notomista, Y. Wardi, and M. Egerstedt, “Integral control barrier functions for dynamically defined control laws,” *IEEE Control Systems Letters*, vol. 5, no. 3, pp. 887–892, 2021.

[22] W. Shaw Cortez, C. Verginis, and D. V. Dimarogonas, “Safe, passive control for mechanical systems with application to physical human-robot interactions,” in *IEEE Conference on Robotics and Automation*, 2021, to appear. [Online]. Available: https://arxiv.org/abs/2011.01810

[23] A. Singletary, S. Kolathaya, and A. Ames. (2020) Safety-critical kinematic control of robotic systems. [Online]. Available: https://arxiv.org/abs/2009.09100

[24] W. Shaw Cortez. (2020) Zcbf algorithm design code. [Online]. Available: https://github.com/shawcortez/safe-control-euler-lagrange

[25] W. Shaw Cortez and D. V. Dimarogonas, “Correct-by-design control barrier functions for Euler-Lagrange systems with input constraints,” in *Proc. American Control Conf.*, 2020, pp. 950–955.

[26] P. Glotfelter, J. Cortés, and M. Egerstedt, “Nonsmooth barrier functions with applications to multi-robot systems,” *IEEE Control Systems Letters*, vol. 1, no. 2, pp. 310–315, 2017.

[27] R. Ortega, A. Loria, P. J. Nicklasson, and H. Sira-Ramirez, *Passivity-Based Control of Euler-Lagrange Systems: Mechanical, Electrical, and Electromechanical Applications*, ser. Communications and Control Engineering. Springer, 1998.

[28] B. Morris, M. Powell, and A. Ames, “Continuity and smoothness properties of nonlinear optimization-based feedback controllers,” in *IEEE Conf. Decision and Control*, 2015, pp. 151–158.

[29] W. W. Hager, “Lipschitz continuity for constrained processes,” *SIAM Journal on Control and Optimization*, vol. 17, no. 3, pp. 321–338, 1979.

[30] L. Lindemann and D. V. Dimarogonas, “Control barrier functions for signal temporal logic tasks,” *IEEE Control Systems Letters*, vol. 3, no. 1, pp. 96–101, 2019.

[31] J. Nocedal and S. Wright, *Numerical Optimization*, 2nd ed. Springer-Verlag, 2006.

[32] E. D. Sontag, *Mathematical Control Theory*, 2nd ed., ser. Texts in Applied Mathematics. Springer, 1998.

[33] H. K. Khalil, *Nonlinear Systems*. Upper Saddle River, N.J.: Prentice Hall, c2002., 2002.

[34] R. M. Redheffer, “The theorems of bony and brezis on flow-invariant sets,” *The American Mathematical Monthly*, vol. 79, no. 7, pp. 740–747, 1972.

[35] M. Spong, *Robot Dynamics and Control*. New York: Wiley, 1989.