Observer Design for Unilaterally Constrained Lagrangian Systems: A Passivity-Based Approach

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Abstract

This paper addresses the problem of estimating the velocity variables, using the position measurement as output, in nonlinear Lagrangian dynamical systems with perfect unilateral constraints. Using the class of bounded variation functions to model the velocity variables (so that Zeno phenomenon is not ruled out), we represent the derivative of such functions with the Lebesgue-Stieltjes measure, and use the framework of measure differential inclusion (MDI) to describe the dynamics. A class of observers is proposed, which also uses the framework of MDIs, and is shown to generate asymptotically converging state estimates. The existence and uniqueness of solutions for the proposed estimators is rigorously proven. The global stability of error dynamics is analyzed using the generalized Lyapunov methods for functions of bounded variation.

I. INTRODUCTION

In this paper, we consider Lagrangian mechanical systems with unilateral constraints (without friction) on the position of a moving point. The position and velocity of this point is denoted by \( q \) and \( \dot{q} \), respectively. Assuming the mass matrix \( M(q) \) to be symmetric and positive definite, the unconstrained motion of the system satisfies the equation

\[
M(q)\ddot{q} + F(t, q, \dot{q}) = 0,
\]

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and the position $q$ is constrained by

$$h_i(q) \geq 0, \quad i \in \{1, \ldots, m\}$$  \hspace{1cm} (1b)$$

where $F : \mathbb{R} \times \mathbb{R}^n \times \mathbb{R}^n \to \mathbb{R}^n$ denotes a vector field of generalized forces, and $h_i : \mathbb{R}^n \to \mathbb{R}$ represent the unilateral constraints imposed on the system’s motion. Mechanical systems with impacts, such as robots and colliding rigid bodies could be seen as systems with unilateral constraints. In general, the trajectories of such systems are algebraically constrained and exhibit continuous as well as discrete dynamics; hence, forming an important class of nonsmooth systems. When none of the constraints $h_i(\cdot)$ are active, that is, $h_i(q) > 0$ for every $1 \leq i \leq m$, then $q$ and $\dot{q}$ are obtained simply by integrating (1a) and are absolutely continuous. The discontinuity in the velocity $\dot{q}$ may appear in such systems when any one of the constraint is active, that is, $h_i(q) = 0$, and the velocity points outside the admissible domain, that is, $\nabla h_i^T(q)\dot{q} < 0$, where we use the notation $\nabla h_i(\cdot)$ to denote the gradient of the function $h_i(\cdot)$. This is because the velocity must change its direction instantaneously to keep the moving point inside the admissible set. In case $\nabla h_i^T(q)\dot{q} \geq 0$, and $h_i(q) = 0$, there are no discontinuities and one only observes continuous motion on the constraint surface (of reduced dimension) defined by $h_i(q) = 0$. There are several modeling frameworks for such nonsmooth systems; one such modeling framework, which is used to model the motion of state-constrained trajectories is the so-called sweeping process [19], [22], [24], [25]. The term so-coined because it represents the motion of a point inside a closed set. As the the set moves, the point is swept across by the moving set. If for such processes, the constraint set is parameterized by time only, then we call it the first-order sweeping process. However, for system (1), we first define an admissible set for velocity $\dot{q}(\cdot)$ which is parameterized by the position $q(\cdot)$, and this formulation leads to a second-order sweeping process.

This paper is concerned with the design of observers for estimating the velocity $\dot{q}(\cdot)$ using the position $q(\cdot)$ as the output, while using the sweeping process formulation to describe the dynamics of the system and the observer. The construction of observers, or state estimators, is a classical problem in the design of control systems and has found many useful applications such as output feedback control, and fault diagnosis. For smooth systems, there are several standard techniques such as Kalman filter, Luenberger observer, or high-gain approach. Such techniques have also been applied to design of observers for certain classes of smooth and unconstrained Lagrangian systems with application to output feedback control, see for example, [6], [7], [27].
A common element of these designs is to assume that the velocity \( \dot{q}(\cdot) \) is uniformly bounded (in time) which is primarily because \( F(t, q, \cdot) \) is quadratic in general for mechanical systems.

Lately, the researchers have started looking at the state-estimation problem in nonsmooth systems. In this regard, we mention the recent work on observer design of switched systems with ordinary differential equations [34], [38], [39], switched differential-algebraic equations [40], [41], certain classes of differential inclusions [10], [29], [36], complementarity systems [17], and the references therein for more details. Classical approaches for observer design are based on constructing an auxiliary dynamical system driven by the error between the measured output and the estimated output, where it is shown that the resulting dynamics of the state estimation error converge to the origin. However, for nonsmooth systems subjected to impacts, such schemes are not easily implementable since the impacts, or discrete dynamics, are not influenced by error injection and hence destroy the integration effect.

For nonsmooth Lagrangian systems with impacts, the problem of state estimation has been considered in [21] under certain restrictive assumptions, and state estimation with tracking control in [16] for motions restricted within a convex polyhedral domain. The work of [5] also deals with the problem of tracking control (without estimation) for similar kind of systems. This article, however, deals with a more general class of nonsmooth Lagrangian dynamics, that allow more general admissible domains for position variable \( q \) using the formalism of differential inclusions. Some examples that can be treated within our setup are given in Figure 1. The interesting aspects

![Fig. 1: Examples of systems with unilateral constraints.](image-url)
of these examples are

- In Fig. 1a, when the ball $q_1$ collides with $q_2$ and $q_3$ stacked together at rest, then $q_2$ and $q_3$ may remain glued after the impact, and hence one of the constraints causes discontinuities in the velocities of $q_2$ and $q_3$, whereas another constraint only allows continuous motion on its boundary. The same happens when $q_3$ collides with the wall.
- In Fig. 1b, the point mass is subjected to downward gravitational force only. After multiple impacts initially with the two boundaries of the constraints, one sees an accumulation of jumps, followed by a continuous motion on the constraint parabolic surface, which is a surface of reduced dimension than the state space.
- In Fig. 1c, there are many impacts in short time-intervals, the domain is nonconvex, and there is possibly a chaotic behavior due to increased frequency of impacts.\(^1\)

Our goal is to design velocity estimators for the particles subjected to unilateral constraints of the form mentioned in Figure 1, which in general may depict all the above complexities. The approach adopted is closely related to the observer design presented in [10] that is itself strongly inspired from the material in [9]. The authors in [10] work with differential inclusions that represent the first-order sweeping process and the resulting trajectories are absolutely continuous. Since the constraint set is a function of time, this information is passed to the state estimator. The state estimator replicates the system dynamics, and hence is a sweeping process of the first-order. The convexity of the constraint set then generates the passivity relation between the error dynamics and the output estimation error, which is the key component in proving the error convergence.

In this paper, however, the system under consideration is a second-order sweeping process. This way, the state-trajectories of the system are allowed to be of locally bounded variation (BV), and hence discontinuous, which introduces the major difference. Since locally BV functions may admit an infinite number of discontinuities in finite time, the Zeno phenomenon is not excluded in our setup. The proposed observer only describes the dynamics for the velocity estimate (and not the position) in the form of a differential inclusion driven by the measured output (position). It is proved that, for each measured output, there exists a unique solution to such differential inclusion, and that this solution converges asymptotically to the actual velocity of the system.

\(^1\)This example was pointed to the authors by the associate editor responsible for handling this paper, L. Menini.
The article is organized as follows: in Section II some useful mathematical definitions are recalled. The Moreau’s sweeping process in which we embed Lagrangian nonsmooth mechanical systems, and the definition of its solutions are described in Section III. The proposed class of estimators is described in Section IV, and rigorous analysis is carried out in Sections V and VI, for existence of solutions, and convergence of estimation error, respectively. In section VII, numerical algorithm for implementing the proposed estimators, and simulation results obtained with the INRIA software package SICONOS are presented for the three systems shown in Figure 1. Section VIII is dedicated to one of the main results of this article, i.e. the proof of the well-posedness (existence and uniqueness of solutions) of the observer. Conclusions end the paper in Section IX.

II. Preliminaries

In this section, we collect some basic definitions and notation that will be used later on.

*Functions of bounded variation:* For an interval $I \subseteq \mathbb{R}$, and a function $f : I \to \mathbb{R}^n$, the variation of $f(\cdot)$ over the interval $I$ is the supremum of $\sum_{i=1}^{k} |f(s_i) - f(s_{i-1})|$ over the set of all finite sets of points $s_0 < s_1 < \cdots < s_k$ (called partitions) of $I$. When this supremum is finite, the mapping $f(\cdot)$ is said to be of *bounded variation* on $I$. We say that $f(\cdot)$ is of *locally bounded variation* if it is of bounded variation on each compact subinterval of $I$. The variation of $f(\cdot)$ over an interval $[0,t]$ is denoted by $\text{var}_f(t)$. If $f(\cdot)$ is right-continuous and of (locally) bounded variation, we call it (locally) *rcbv*. A function of locally bounded variation on $I$ has at most a countable number of jump discontinuities in $I$. Moreover, it has right and left limits everywhere. The right and left limits of the function $f(\cdot)$ at $t \in I$ are denoted by $f(t^+) := \lim_{s \uparrow t} f(s)$ and $f(t^-) := \lim_{s \downarrow t} f(s)$, respectively, provided they exist. In this notation, right continuity of $f(\cdot)$ in $t$, means that $f(t^+) = f(t)$.

*Locally integrable functions:* We denote by $L_1(I, \mathbb{R}^n; d\mu)$ and $L_1^{\text{loc}}(I, \mathbb{R}^n; d\mu)$ the space of integrable and locally integrable functions, respectively, from $I$ to $\mathbb{R}^n$ with respect to the measure $d\mu$. If the measure is not specified then the integration is with respect to the Lebesgue measure. An absolutely continuous (AC) function $f : I \to \mathbb{R}^n$ is a function that can be written as $f(t) - f(t_0) = \int_{t_0}^{t} \dot{f}(s) ds$ for any $t_0, t \in I$, $t_0 \leq t$, and some $\dot{f} \in L_1(I, \mathbb{R}^n)$, which is considered as its derivative. The space of continuously differentiable functions from $\mathbb{R}^n$ to $\mathbb{R}^m$ is denoted by $C^1(\mathbb{R}^n, \mathbb{R}^m)$, for $m, n \in \mathbb{N}$. 
**Lebesgue-Stieltjes measure associated with BV functions:** If \( v : I \to \mathbb{R}^n \) is a function of bounded variation, then one can associate with it a Lebesgue-Stieltjes measure or the so-called differential measure \( dv \) on \( I \). Also, if \( v(\cdot) \) is \( rcbv \) on \([a,b]\), then we have the relation that \( v(t) = v(a) + \int_{[a,t]} dv \).

The density of the measure \( dv \) with respect to a positive Radon measure \( d\mu \) over an interval \( I \) is defined as:

\[
\frac{dv}{d\mu}(t) := \lim_{\varepsilon \to 0} \frac{dv(I(t,\varepsilon))}{d\mu(I(t,\varepsilon))},
\]

where \( I(t,\varepsilon) := I \cap [t-\varepsilon, t+\varepsilon] \). Similarly, one can define the density of the Lebesgue measure \( dt \) with respect to the Radon measure \( d\mu \). A Radon measure \( d\nu \) is absolutely continuous with respect to \( d\mu \) if for every measurable set \( A \), \( d\mu(A) = 0 \) implies that \( d\nu(A) = 0 \). Further, the measure \( d\nu \) is absolutely continuous with respect to \( d\mu \) if and only if the density function \( \frac{d\nu}{d\mu}(\cdot) \) is well-defined (finite \( \mu \)-almost everywhere) and is \( d\mu \) integrable.

**Convex analysis:** For a set \( V \subset \mathbb{R}^n \), we will denote its interior by \( \text{int} \ V \), and the boundary of this set is denoted by \( \text{bd}(V) \). If \( V \) is closed convex, then \( N_V(v) \) denotes the normal cone to \( V \) at \( v \in V \) and is defined as:

\[
N_V(v) := \{ w \in \mathbb{R}^n \mid \langle w, x - v \rangle \leq 0 \ \forall x \in V \}.
\]

where \( \langle \cdot, \cdot \rangle \) denotes the usual inner product in \( \mathbb{R}^n \). We adopt the convention that \( N_V(v) = \emptyset \) if \( v \not\in V \). Obvious from the definition, the normal cone to a closed convex set is a monotone operator, that is, if \( w_i \in N_V(v_i) \), \( i = 1, 2 \), then

\[
\langle w_1 - w_2, v_1 - v_2 \rangle = \langle w_1, v_1 - v_2 \rangle - \langle w_2, v_1 - v_2 \rangle \geq 0.
\]

(3) \( \Rightarrow \geq 0 \)

(3) \( \Rightarrow \leq 0 \)

When \( V \) is a closed convex cone, we denote by \( V^\circ(q) \) the closed convex polyhedral cone polar to \( V(q) \) with respect to usual inner product on \( \mathbb{R}^n \), which is defined as:

\[
V^\circ(q) := \{ w \in \mathbb{R}^n \mid w^\top v \leq 0, \ \forall v \in V(q) \}.
\]

**III. Dynamic Model for Constrained Lagrangian Systems**

In this section, we will describe the dynamics of nonsmooth Lagrangian systems using differential inclusions and briefly talk about their solutions. The observer will then be designed using this formalism.
A. Mathematical description

We consider mechanical systems with a finite number of degrees of freedom that are subjected to the unilateral constraints described in (1b). The position variable \( q \in \mathbb{R}^n \) is thus assumed to evolve in a set that admits the following form:

\[
\Phi := \{ q \in \mathbb{R}^n \mid h_i(q) \geq 0, \ i = 1, 2, \ldots, m \}.
\]  

The geometry of the set \( \Phi \) is determined by the functions \( h_i(\cdot) \), and the only condition we will impose on the functions \( h_i(\cdot) \) is that they are continuously differentiable so that \( \nabla h_i(\cdot) \) is continuous for each \( i \). This allows us to model a large number of closed domains which may even be nonconvex.

The convex polyhedral tangent cone \( V(q) \) to the region \( \Phi \) at a point \( q \) is given by:

\[
V(q) := \{ v \in \mathbb{R}^n \mid v^T \nabla h_i(q) \geq 0, \ \forall i \in J(q) \}
\]  

where the set \( J(q) \) denotes the set of active constraints at \( q \), i.e.,

\[
J(q) := \{ i \in \{1, \ldots, m\} \mid h_i(q) = 0 \}.
\]

One can think of the set \( V(q(t)) \) as the set of admissible velocities that keep the position variable \( q(t) \) inside the set \( \Phi \). In what follows, the notion of normal cone to the set \( V(q) \), denoted \( N_{V(q)} \), is instrumental. For \( v \in V(q) \), we have

\[
N_{V(q)}(v) = \left\{ w \in \mathbb{R}^n \mid w = - \sum_{i \in J(q)} \lambda_i \nabla h_i(q), \lambda_i \geq 0 \right\}
\]  

\[
= \left\{ w \in \mathbb{R}^n \mid w = - \sum_{i=1}^{m} \lambda_i \nabla h_i(q), \ 0 \leq \lambda_i \perp h_i(q) \geq 0 \right\}
\]

where the later equation, at once, shows the link with complementarity framework.

We now formulate the dynamics of system (1) as a measure differential inclusion:

\[
dq = vdt
\]  

\[
M(q)dv + F(t, q, v)dt \in -N_{V(q)}(v_e)
\]

where

\[
v_e(t) := \frac{v(t^+) + ev(t^-)}{1 + e},
\]
and $e \in [0, 1]$ is the coefficient of restitution. The initial condition is assumed to satisfy $q_0 := q(0) \in \Phi$, and $v_0 := v(0)$ is such that $v_e(0) \in V(q_0)$. As a graphical illustration of the normal cone $N_{V(q)}$, we consider the example given in Figure 1b, where the two constraint functions are $h_1(q_x, q_y) = -q_y + 4 \geq 0$, and $h_2(q_x, q_y) = q_y - q_x^2 \geq 0$. Three different scenarios are depicted in Figure 2 for this example corresponding to $J(q) = \{1\}$ in Fig. 2a, $J(q) = \{2\}$ in Fig. 2b, and $J(q) = \{1, 2\}$ in Fig. 2c.

B. Interpreting MDI (8)

The motivation for working with the MDI is that we are seeking a solution to the evolution problem in the space of locally $rcbv$ functions to deal with possible collisions with the boundary of the admissible set. Functions which are locally $rcbv$ possess generalized derivatives that can be identified with Stieltjes measure and equation (8b) precisely describes the inclusion of the measure $dv$, associated with $v(\cdot)$, into a normal cone described by the constraint set $\Phi$.

a) When no constraints are active: It is noted that, if $q \in \text{int} \Phi$, that is, $h_i(q) > 0$, for each $1 \leq i \leq m$, so that $J(q) = \emptyset$, then $V(q) = \mathbb{R}^n$ and consequently $N_{V(q)}(\cdot) = \{0\}$. This reduces (8) to ordinary differential equations described by $\dot{q} = v$ and $M(q)\dot{v} + F(t, q, v) = 0$.

b) Post-impact velocities: It is also noted that the post-impact velocity determined according to Moreau’s collision rule (or Newton’s impact law) is directly encoded in the MDI (8). To derive

(a) Cones for the case $h_1(q) \leq 0$. (b) Cones for the case $h_2(q) \leq 0$. (c) Corner case: $h_1(q) \leq 0 \land h_2(q) \leq 0$.

Fig. 2: Polyhedral cones for two constraints: $h_1(q) = -q_y + 4 \geq 0$, $h_2(q) = q_y - q_x^2 \geq 0$. 

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an explicit expression for the post-impact velocity, we proceed as follows:

\[ M(q(t_k))[v(t_k^+) - v(t_k^-)] \in -\mathcal{N}_{V(q(t_k))}(v_e(t_k)) \]  \hspace{1cm} (9a)

\[ \Leftrightarrow v(t_k^+) - v(t_k^-) \in -M(q(t_k))^{-1}\mathcal{N}_{V(q(t_k))}(v_e(t_k)) \]  \hspace{1cm} (9b)

\[ \Leftrightarrow \frac{1}{1+e} [v(t_k^+) - v(t_k^-)] \in -M(q(t_k))^{-1}\mathcal{N}_{V(q(t_k))}(v_e(t_k)) \]  \hspace{1cm} (9c)

\[ \Leftrightarrow v_e(t_k) - v(t_k^-) \in -M(q(t_k))^{-1}\mathcal{N}_{V(q(t_k))}(v_e(t_k)) \]  \hspace{1cm} (9d)

\[ \Leftrightarrow v_e(t_k) = \text{proj}_{M(q(t_k))}(V(q(t_k)); v(t_k^-)) \]  \hspace{1cm} (9e)

\[ \Leftrightarrow v(t_k^+) = -ev(t_k^-) + (1+e) \text{proj}_{M(q(t_k))}(V(q(t_k)); v(t_k^-)) \]  \hspace{1cm} (9f)

where \( \text{proj}_{M(q)}(V(q); v) \) denotes the projection of \( v \) on the set \( V(q) \) according to the kinetic metric at \( q \), which is defined by the inner product \( \langle v, w \rangle_{M(q)} = \langle v, M(q)w \rangle = \langle M(q)v, w \rangle \). In the above expression, it is used that a normal cone is invariant under multiplication by a nonnegative scalar in (9c). Expression (9e) is obtained using a well-known result from convex analysis that relates the projection of a point on a convex set with the normal cone to the convex set at the projected point.

One can also interpret MDE (8) at impact times in the sense that, we want to compute \( v(t_k^+) \) such that \( v_e(t_k) \) belongs to the set \( V(q(t_k)) \) while minimizing \( |v_e(t_k) - v(t_k^-)|_{M(q(t_k))} \). Thus, there is an optimization problem to be solved in order to compute \( v(t_k^+) \). One typically reformulates this optimization problem using the framework of complementarity program [13] for which there are already some efficient solvers. More precisely, we let

\[ M(q_k)[v(t_k^+) - v(t_k^-)] = -\sum_{\alpha \in \mathcal{J}(q_k)} \lambda_\alpha \nabla h_\alpha(q_k), \quad \lambda_\alpha \geq 0, \alpha \in \mathcal{J}(q_k) \]

where \( q_k := q(t_k) \), and \( \lambda_\alpha \) is computed from

\[ 0 \leq \lambda_\alpha \perp \langle \nabla h_\alpha(q_k), M(q_k)[v(t_k^+) + ev(t_k^-)] \rangle \geq 0, \quad \alpha \in \mathcal{J}(q_k). \]  \hspace{1cm} (10)

The above complementarity relation is an equivalent way of writing

\[ v_e(t_k) \in V(q_k) \quad \text{and} \quad \langle v(t_k^+) - v(t_k^-), M(q_k)(v(t_k^+) + ev(t_k^-)) \rangle = 0. \]

which are the relations encoded in MDI (8).
c) Continuous motion on constraint surfaces: Contact with the surface will not always result in the discontinuities of the velocity variable $v$. From (9f), it is seen that, if $v(t^-_k) \in V(q(t_k))$, then we have $v(t^+_k) = v(t^-_k)$. The MDE (8) in case of continuous motion along the boundary of the constraint is written as

$$M(q)\dot{v} + F(t, q, v) = -\sum_{\alpha \in \mathcal{J}(q)} \lambda_\alpha \nabla h_\alpha(q), \quad \lambda_\alpha \geq 0, \alpha \in \mathcal{J}(q)$$

where $\lambda_\alpha$ are again obtained through the relation (10). We see that (8) encapsulates switches to lower dimensional systems, thanks to the the existence of suitable multipliers (i.e. contact forces) calculated from a complementarity problem.

The formulation for constrained mechanical systems, as in (8), was pioneered by J. J. Moreau [24], and the MDI (8) is called a second order sweeping process (because the constraint set for velocities appearing in (8b) depends on the state variable $q(\cdot)$). Further details on inclusions of type (8) and comparisons with other modeling frameworks could be found in [8, Section 5.3]. For our purpose, it is seen that the observer design given in Section IV is partially aided by this compact formulation. It is noteworthy that there is a close link between the sweeping process in (8) and so-called complementarity Lagrangian systems, see e.g. [2, Section 3.6], and the examples in Section VII.

C. Assumptions on System Data and Solutions

The solution of MDI (8) is considered in the following sense:

**Definition 1.** A solution to the Cauchy problem (8) with initial data $(q_0, v_0) \in \Phi \times V(q_0)$, over an interval $I = [0, T]$, is a pair $(q, v)$ such that $v(\cdot)$ is rcbv on $I$; $q(t) = q_0 + \int_0^t v(s)ds$; $q(t) \in \Phi$ and $v_e(t) \in V(q(t))$ for all $t \geq 0$; and furthermore, there exists a positive measure (represented by) $d\mu$ such that both $dt$ and $dv$ possess densities with respect to $d\mu$, denoted by $dt/d\mu$, and $dv/d\mu$ respectively, such that

$$M(q)\frac{dv}{d\mu}(t) + F(t, q, v)\frac{dt}{d\mu}(t) \in -\mathcal{N}_{V(q(t))}(v_e), \quad d\mu-a.e. \text{ on } I. \quad (11)$$

The choice of the measure $d\mu$ is not unique since the right-hand side of (8) is a cone. However, by Lebesgue-Radon-Nikodym theorem, the functions $dt/d\mu(\cdot) \in \mathcal{L}_1(I, R; d\mu)$ and $dv/d\mu(\cdot) \in \mathcal{L}_1(I, \mathbb{R}^n; d\mu)$ are uniquely determined for a given $d\mu$. 

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The problem of existence of solutions for evolution problems (1) has been studied for a long time. Earlier results on this problem dealt with the single constraint case \((m = 1)\) and one may refer to [22, Chapter 3], [32] for results in this direction. The basic idea in these works is to introduce a time discretization scheme, either at position level [32] or velocity level [22] to construct a sequence of approximate solutions which is shown to converge as the step size converges to zero. For several unilateral constraints \((m \geq 2)\), the existence and uniqueness has been proved in [3] under analytic assumptions on the data using the solution theory for differential equations and variational inequalities. Building on the results derived in [20], the most relaxed conditions, under which the existence of solutions has been proved using discretization at velocity level, have appeared recently in [14] for the inelastic case \((e = 0)\), and in [15], [31] for general values of \(e \in [0, 1]\). Based on the work of [31], the following regularity assumptions are required on the system data for the existence of solution, and are also needed for the observer design:

\((H1)\) The function \(F(\cdot, \cdot, \cdot)\) is continuous and is continuously differentiable \((C^1)\) with respect to its second and third arguments.

\((H2)\) The mapping \(M(\cdot)\), from \(\mathbb{R}^n\) to the set of symmetric positive definite matrices, belongs to class \(C^1\) and there exists \(0 \leq \lambda_M \leq \overline{\lambda}_M\) such that

\[
\lambda_M |v|^2 \leq v^\top M(q)v \leq \overline{\lambda}_M |v|^2 \quad \forall (q, v) \in \Phi \times \mathbb{R}^n.
\]  \(12\)

\((H3)\) For each \(i \in \{1, \ldots, m\}\), the function \(h_i \in C^1(\mathbb{R}^n, \mathbb{R})\), its Euclidean gradient \(\nabla h_i(q)\) is locally Lipschitz continuous and does not vanish in a neighborhood of \(\{q \in \mathbb{R}^n \mid h_i(q) = 0\}\).

\((H4)\) The active constraints are functionally independent, i.e., \(\{\nabla h_i(q)\}_{i \in J(q)}\) is linearly independent for all \(q \in \Phi\).

Without recalling the formal result on existence and assuming that a solution exists in the sense of Definition 1 under hypotheses \((H1) – (H4)\), we only collect the properties of the solutions to system (8) which provide more insight.

\(D.\) Solution Characteristics

1) Regularity of state trajectories: The function \(q\) is absolutely continuous, but not necessarily everywhere differentiable. The velocity \(v(\cdot)\) is a locally \(rcbv\) function, for which the left and right limits are defined everywhere. The acceleration is represented by the measure \(dv\) and can
be decomposed as a sum of three measures: an atomic measure $d\mu_a$, Lebesgue measure $dt$, and a measure associated with singularly continuous function $d\mu_{sc}$ i.e., $dv = d\mu_a + \dot{v}dt + d\mu_e + d\mu_{sc}$.

2) **Countably many impacts:** The set of impact times, at which $v(\cdot)$ is discontinuous, is at most countable. One may simply take $d\mu_a = \sum_{k \geq 0} [v(t_k^+) - v(t_k^-)] \delta_{t_k}$, where $\delta_{t_k}$ is the Dirac impulse at time $t_k$ and $\{t_k\}_{k \geq 0}$ is an ordered sequence of impact times. Thus, the formulation (8) does not exclude the Zeno phenomenon (with a finite or infinite number of left accumulation points). However, if $e = 1$, then it is shown in [4] that there exists a constant $\rho_T(q(0), v(0)) > 0$ such that $t_{k+1} - t_k > \rho_T(q(0), v(0))$, for each $t_k, t_{k+1}$ belonging to a compact interval $[0, T]$.

3) **Non-uniqueness and Continuity of solutions:** The solution of system (8) is unique if the system data is analytic [3], but in general, it may not be the case. Even under the analyticity assumption, the solutions may not vary continuously with respect to initial conditions under the hypotheses (H1) - (H4). For this to hold, there is an additional condition on the set $\{\nabla h_i(q)\}_{i \in J(q)}$ given in [30], which states that for $e = 0$, the active constraints must satisfy $\langle \nabla h_i(q), M^{-1}(q)\nabla h_j(q) \rangle \leq 0$, and for $e \in (0, 1]$, $\langle \nabla h_i(q), M^{-1}(q)\nabla h_j(q) \rangle = 0$. The work of [16] assumes this condition because their design is based on closeness of solutions (in graphical sense) with respect to initial conditions. However, our observer design doesn’t require this property and hence no such condition is imposed in our results.

### IV. Observer Design

We now address the problem of designing observers for the systems considered in Section III. It will be assumed that the position $q(\cdot)$ is the measured variable, and the objective is then to design an estimator which either estimates the full state $(q, v)$, or only the unknown velocity $v$ of the moving particle. The class of state estimators that we propose for this purpose comprises a differential inclusion with state $z := (z_1^T, z_2^T)^T \in \mathbb{R}^{d_z}$ satisfying

\begin{align}
\dot{z}_1 &= F_1(t, q, z) \\
M(q) \dot{z}_2 + F_2(t, q, z) dt &\in -N_{V(q)}(\hat{v}_e)
\end{align}

where $\hat{v}_e(t) \in \mathbb{R}^n$ is given by

\begin{equation}
\hat{v}_e(t) = \frac{\hat{v}(t^+) + e\hat{v}(t^-)}{1 + e}.
\end{equation}
The state estimate \((\hat{q}(t), \hat{v}(t)) \in \mathbb{R}^{2n}\) is defined as:

\[
\hat{q} = f_1(z_1, q) \quad (15a)
\]
\[
\hat{v} = z_2 + f_2(z_1, q) \quad (15b)
\]

and the function \(\begin{pmatrix} z_1 \\ z_2 \end{pmatrix} \mapsto f(q, z) := \begin{pmatrix} f_1(z_1, q) \\ z_2 + f_2(z_1, q) \end{pmatrix}\) is assumed to be a diffeomorphism for each \(q \in \mathbb{R}^n\), so that the function \(f^{-1}(q, \cdot)\) is well-defined and continuously differentiable.

Choosing the functions \(F_1, F_2\) and \(f_1, f_2\) is a part of the design procedure and we will give two possible ways of choosing these functions so that the estimate \((\hat{q}, \hat{v})\) converges asymptotically to the actual state \((q, v)\). Moreover, it will be shown that, under certain regularity assumptions on the functions \(F_1, F_2\) and \(f_1, f_2\), there exists a unique solution to the proposed observer (13).

Before proceeding towards these main results, we choose to rewrite the observer dynamics in \((\hat{q}, \hat{v})\) coordinates

\[
\dot{\hat{q}}(t) = \widehat{F}_1(t, x, \hat{x}) \quad (16a)
\]
\[
M(q) d\hat{v} + \widehat{F}_2(t, x, \hat{x}) \in -N_{V(q)}(\hat{v}), \quad (16b)
\]

where, for brevity, we let

\[
\begin{align*}
\widehat{F}_1(t, x, \hat{x}) &:= \frac{\partial f_1}{\partial z_1} F_1(t, q, f^{-1}(q, x)) + \frac{\partial f_1}{\partial q} v \\
\widehat{F}_2(t, x, \hat{x}) &:= F_2(t, q, f^{-1}(q, x)) + M(q) \frac{\partial f_2}{\partial z_1} F_1(t, q, f^{-1}(q, x)) - M(q) \frac{\partial f_2}{\partial q} v.
\end{align*} \quad (16c)
\]

This new description of the observer dynamics also provides an insight about its mechanism. Equation (16b) basically tells us that the estimate \(\hat{v}\) is constrained in the same way as the actual velocity \(v\). The nonsmooth behavior in the velocity variable is due to the forces that belong to the set \(N_{V(q)}(v)\). By measuring the position variable, the set \(V(q)\) can be computed at each time. One then uses the monotonicity property of the normal cone operator \(N_{V(q)}(\cdot)\) in analyzing the error dynamics to show convergence.

In Sections V and VI, we will show that the proposed observer (16) has the following two properties, respectively:

- **Well-posedness**: For each absolutely continuous function \(q(\cdot)\), there exists a unique locally \(rcbv\) function \(\hat{v}(\cdot)\) obtained from (13)–(15).

- **Error convergence**: The estimates \(\hat{q}(\cdot), \hat{v}(\cdot)\) converge to \(q(\cdot), v(\cdot)\) asymptotically.
Before proceeding with these technical results, note that the original system may not have unique solutions, but the observer has the property that it generates a unique trajectory corresponding to each function \( q(\cdot) \) observed as an output of system (8); see [10, Remark 3.3] for further explanation along these lines.

V. OBSERVER WELL-POSEDNESS

The estimator (13) is actually an evolution inclusion in which the multi-valued function \( N_{V(q)}(\cdot) \) is closed and convex valued. It is noted that the function \( q(\cdot) \) is seen as an external “input” by the observer and hence \( V(q(\cdot)) \) is seen as a time-parameterized multi-valued function that does not depend on any of the internal states of the estimator. This makes the observer (13) a sweeping process of first order. We will basically prove the well-posedness result for the differential inclusion

\[
M(q(t))d\hat{v} + g(t, \hat{v})dt \in -N_{V(q(t))}(\hat{v}_e)
\]

(17)

under certain regularity assumption on the function \( g(t, \hat{v}) \). Using this result, it will be shown that the observer (13) can be transformed into a system of form (17). The solution to system (13) is interpreted in a sense similar to Definition 1.

Let us now state the following result on existence and uniqueness of solution to (17). This is a fundamental step since existence of solutions secures that the error stability analysis is meaningful, while uniqueness property secures that the observer output is unique for a given plant trajectory, as stated earlier.

**Theorem 1.** Consider the differential inclusion (17) under the hypotheses (H2)–(H4), and \( V(q) \) defined in (6). Assume that the function \( q : [0, T] \rightarrow \mathbb{R}^n \) is absolutely continuous, and that the function \( g : [0, T] \times \mathbb{R}^n \rightarrow \mathbb{R}^n \) satisfies

\[
|g(t, \hat{v}_1) - g(t, \hat{v}_2)| \leq C_{g,l}|\hat{v}_1 - \hat{v}_2|, \quad \forall \hat{v}_1, \hat{v}_2 \in \mathbb{R}^n, \; \forall t \in [0, T]
\]

(18)

\[
|g(t, \hat{v})| \leq C_{g,b}(1 + |\hat{v}|), \quad \forall t \in [0, T]
\]

(19)

for some constants \( C_{g,l}, C_{g,b} > 0 \). Then the system (17) is well-posed, that is, there exists a unique solution \( \hat{v} \in BV([0, T]; \mathbb{R}^n) \) for any initial condition \( \hat{v}(0) \in V(q(0)) \). Moreover, it holds that

\[
\hat{v}_e(t) \in V(q(t)), \quad \forall t \in [0, T].
\]

(20)
The result on existence and uniqueness of solutions for MDI (17), stated in Theorem 1, is important in several respects:

(a) The multivalued operator on the right-hand side is non-compact, time-varying and the variation of this set-valued map (measured using Hausdorff-distance) is not of bounded variation.

(b) Even though the interior of \( V(q) \) for each \( q \) is nonempty, in general, there does not exist any common open ball of which is contained in \( V(q) \), for each \( q \in \Phi \).

(c) The argument of \( N_{V(q)}(\cdot) \) is not simply the state \( \hat{v} \) but rather a weighted sum of pre- and post-impact values of \( \hat{v}(\cdot) \).

(d) The mapping \( t \mapsto V(q(t)) \) is lower semicontinuous (because \( t \mapsto q(t) \) is absolutely continuous and \( q \mapsto V(q) \) is lower semicontinuous).

Because of these reasons, we cannot use the existing results on solutions of first-order sweeping processes, for example [22, Chapter 2], in a straightforward manner. Moreover, the numerical implementation of the examples considered in this paper (see Section VII) is based on a time-discretization procedure and the proof of Theorem 1 shows that the proposed sequence of discretized solutions indeed converges to a unique solution of system (17). With this motivation, we work out a formal proof of Theorem 1 in this paper. In this section, we will only develop an outline which shows all the steps involved in the proof and for some of these steps, detailed calculations are given in Section VIII.

A. Applying Theorem 1 to Observer (13)

Our goal is to show that the proposed observer (13) can be written in the form of (17), and that the hypotheses of Theorem 1 hold in this case. To see this, one can rewrite the description of the observer in (16) as follows:

\[
\begin{bmatrix}
I & 0 \\
0 & M(q(t))
\end{bmatrix}
\begin{bmatrix}
d\hat{q} \\
d\hat{\hat{v}}
\end{bmatrix} + \begin{bmatrix}
\hat{F}_1(t, q(t), v(t), \hat{q}, \hat{v}) \\
\hat{F}_2(t, q(t), v(t), \hat{q}, \hat{v})
\end{bmatrix} \in -N_{\mathbb{R}^n \times V(q)} \begin{bmatrix}
\hat{q} \\
\hat{\hat{v}}
\end{bmatrix}.
\]

(21)

The underlying reasoning between this transformation is that \( N_{S_1 \times S_2}(\hat{q}, \hat{\hat{v}}) = N_{S_1}(\hat{q}) \times N_{S_2}(\hat{\hat{v}}) \), for \( \hat{q} \in S_1, \hat{\hat{v}} \in S_2 \), and \( S_1, S_2 \) being closed, convex subsets of \( \mathbb{R}^n \). Let \( \hat{F}(t, \hat{x}) := \begin{bmatrix}
\hat{F}_1(t, q(t), v(t), \hat{q}, \hat{v}) \\
-\hat{F}_2(t, q(t), v(t), \hat{q}, \hat{v})
\end{bmatrix} \), where we see \( q, v \) as functions of time, and use the notation \( \hat{x} := (\hat{q}^T, \hat{\hat{v}}^T)^T \). We now have the following corollary:
Corollary 1. Consider the differential inclusion (21) under the hypotheses (H2)–(H4), and \( V(q) \) defined in (6). Assume that the function \( q : [0, T] \to \mathbb{R}^n \) is absolutely continuous, \( v : [0, T] \to \mathbb{R}^n \) is a function of bounded variation, and that the function \( \hat{F} : [0, T] \times \mathbb{R}^{2n} \to \mathbb{R}^{2n} \) satisfies

\[
|\hat{F}(t, \hat{x}_1) - \hat{F}(t, \hat{x}_2)| \leq C_{\hat{F},l}|\hat{x}_1 - \hat{x}_2| \quad \forall \hat{x}_1, \hat{x}_2 \in \mathbb{R}^{2n}, \forall t \in [0, T]
\]
\[
|\hat{F}(t, \hat{x})| \leq C_{\hat{F},b}(1 + |\hat{x}|) \quad \forall \hat{x} \in \mathbb{R}^{2n}, \forall t \in [0, T].
\]

for some constants \( C_{\hat{F},l}, C_{\hat{F},b} > 0 \). Then the system (21) is well-posed, that is, there exists a unique solution \( \hat{q} \in AC([0, T], \mathbb{R}^n) \), \( \hat{v} \in BV([0, T]; \mathbb{R}^n) \) for any initial condition \( (\hat{q}(0), \hat{v}(0)) \in \mathbb{R}^n \times V(q(0)) \). Moreover, it holds that

\[
\hat{v}_e(t) \in V(q(t)) \quad \forall t \in [0, T].
\]

The proof of this corollary is a direct application of Theorem 1 where we work with the augmented variable \( (\hat{q}^T, \hat{v}^T)^T \). The hypotheses stated in Theorem 1 for (17) also hold for (21). However, Corollary 1 claims that \( \hat{q} \) is absolutely continuous, whereas Theorem 1 only guarantees that \( \hat{q} \) is of bounded variation. This extra regularity on \( \hat{q} \) follows due to the fact that \( \hat{q} \) dynamics are basically unconstrained and are obtained by integrating \( \hat{F}_1(t, q(t), v(t), \hat{q}, \hat{v}) \).

B. Proof Outline for Theorem 1

Consider a partition \( \mathcal{P} \) of the interval \([0, T]\) given by:

\[
\mathcal{P} := \{t_{\mathcal{P},i}, 0 \leq i \leq N_{\mathcal{P}}\}, \quad 0 = t_{\mathcal{P},0} < t_{\mathcal{P},1} < t_{\mathcal{P},2} < \ldots < t_{\mathcal{P},N_{\mathcal{P}}} = T
\]

and let

\[
\hat{v}_{\mathcal{P},0} = \hat{v}_0 \quad (23a)
\]
\[
\hat{v}_{\mathcal{P},i} = -e\hat{v}_{\mathcal{P},i-1} + (1 + e) \text{proj}_{M_{\mathcal{P},i}} \left[ \hat{v}_{\mathcal{P},i-1} - \frac{1}{1 + e} M_{\mathcal{P},i}^{-1} G_{\mathcal{P},i} \right] V_{\mathcal{P},i} \quad (23b)
\]

where \( G_{\mathcal{P},i} := \int_{t_{\mathcal{P},i-1}}^{t_{\mathcal{P},i}} g(s, \hat{v}_{\mathcal{P},i-1}) \, ds \) and \( M_{\mathcal{P},i} := M(q(t_{\mathcal{P},i})) \). One can then define a piecewise constant solution \( \hat{v}_{\mathcal{P}}(\cdot) \) for each partition \( \mathcal{P} \) as follows:

\[
\hat{v}_{\mathcal{P}}(t) := \begin{cases} 
\hat{v}_{\mathcal{P},i} & t \in [t_{\mathcal{P},i}, t_{\mathcal{P},i+1}) \\
\hat{v}_{\mathcal{P},N_{\mathcal{P}}} & t = t_{N_{\mathcal{P}}}. 
\end{cases}
\]
The motivation behind defining the successive elements of a piecewise constant solution using (23b) is that:\(^2\)

\[
\frac{\dot{v}_{P,i} + e\dot{v}_{P,i-1}}{1 + e} = \text{proj}_{M_{P,i}} \left[ \dot{v}_{P,i-1} - \frac{1}{1 + e} M_{P,i}^{-1} G_{P,i}, V_{P,i} \right]
\]

\[\iff M_{P,i}(\dot{v}_{P,i} - \dot{v}_{P,i-1}) + G_{P,i} \in -N_{V_{P,i}} \left( \dot{v}_{P,i} + e\dot{v}_{P,i-1} \right) \]

which is a quite natural discretization of (17).

In the sequel,

- a uniform bound (with respect to \(P\)) is derived on \(|\dot{v}_P|\) in Section VIII-A, and
- an estimate of the total variation of \(\dot{v}_P\) over a compact interval is computed in Section VIII-B.

Using these bounds to invoke a generalized version of Helly’s first theorem (see Theorem A.2 in Appendix A), there exists a filter \(F\) finer than the filter of sections of \(P\), and a function of bounded variation \(\hat{v}_P\) \(\colon [0, T] \to \mathbb{R}^n\), which is the weak pointwise limit of \(\dot{v}_P(\cdot)\) with respect to \(F\). Since we are working in the finite-dimensional setup, \(\hat{v}(\cdot)\) is a strong pointwise generalized sublimit of \(\dot{v}_P\):

\[
\lim_{F} |\hat{v}(t) - \dot{v}_P(t)| = 0 \quad \forall t \in [0, T].
\] (25)

The next step is to show that \(\hat{v}(\cdot)\) obtained above is indeed a solution to system (17). We demonstrate it by showing that

- the differential inclusion (17) holds at continuity points of \(\hat{v}\) (Section VIII-C), and
- the inclusion (17) is satisfied at discontinuity points of \(\hat{v}\) (Section VIII-D).

The fact that \(\hat{v}_e(t) \in V(q(t))\) follows due to closedness of \(V(q(t))\). To complete the proof, it remains to show that the solution to (17) is unique, which basically follows due to convexity of \(V(q(t))\) and Lipschitz continuity of \(g(t, \cdot)\). To see that, let \(\hat{v}^1(\cdot), \hat{v}^2(\cdot)\) be two solutions to (17) with \(\hat{v}^1(0) = \hat{v}^2(0)\), then there exists a measure \(d\hat{\mu}\) such that

\[
M(q(t)) \frac{d\hat{v}^i}{d\hat{\mu}}(t) + g(t, \hat{v}^i) \frac{dt}{d\hat{\mu}}(t) \in -N_{V(q(t))}(\hat{v}^i), \quad i = 1, 2.
\]

\(^2\)We use the fact that for a convex set \(V\), it holds that \(x = \arg \min_{y \in V} |z - y|_M\), that is, \(x\) is the projection of \(z\) onto \(V\) with respect to the norm induced by a symmetric positive definite matrix \(M\), if and only if \(\langle M(z - x), y - x \rangle \leq 0, \forall y \in V \iff M(z - x) \in N_V(x)\).
Using the monotonicity property of the normal cone, we get
\[
\left\langle M(q(t)) \left( \frac{d\hat{v}_1}{d\mu}(t) - \frac{d\hat{v}_2}{d\mu}(t) \right), \hat{v}_1(t) - \hat{v}_2(t) \right\rangle \leq |g(t, \hat{v}_2(t)) - g(t, \hat{v}_1(t))| \cdot |\hat{v}_1(t) - \hat{v}_2(t)|.
\]

Since \( g(t, \cdot) \) is Lipschitz, and \( \hat{v}_1(0) = \hat{v}_2(0) \), the above inequality becomes
\[
|\hat{v}_1 - \hat{v}_2|^2 \leq \frac{2C_{g,t}}{\Lambda_M} \int_{0}^{t} |\hat{v}_1(s) - \hat{v}_2(s)|^2 d\hat{\mu}(s).
\]

One can now invoke the Gronwall-Bellman like lemma for functions of bounded variation [18, Lemma 4], to get
\[
|\hat{v}_1 - \hat{v}_2|^2 \leq 0,
\]
whence it follows that \( \hat{v}_1(t) = \hat{v}_2(t) \), for \( t \in [0, T] \).

VI. Error Stability Analysis

In this section, we address the convergence of the estimation error to zero. In what follows, let \( x := (q^T, v^T)^T, \hat{x} := (\hat{q}^T, \hat{v}^T)^T \), and let the state estimation error be denoted by \( \tilde{x} := (\tilde{q}^T, \tilde{v}^T)^T := x - \hat{x} \). The main result on convergence of error now follows:

**Theorem 2.** Assume that there exists a symmetric positive definite matrix-valued function \( R : \mathbb{R}^n \to \mathbb{R}^{2n \times 2n}, q \mapsto R(q) \),
\[
R(q) = \begin{bmatrix} R_{11} & 0 \\ 0 & M(q) \end{bmatrix}
\]
and a constant \( \beta > 0 \) such that
\[
\tilde{x}^T R(q) \left( \begin{array}{c} v - \hat{F}_1(t, x, \hat{x}) \\ -F(t, x) + \hat{F}_2(t, x, \hat{x}) \end{array} \right) + \tilde{x}^T \hat{R}(q, v) \tilde{x} \leq -\beta \tilde{x}^T R(q) \tilde{x} \tag{26}
\]
then the state estimation error decays exponentially, that is,
\[
|\tilde{x}(t)| \leq e^{-\beta t} |\tilde{x}(0)|.
\]

The statement of Theorem 2 basically requires us to choose a state estimator where the unconstrained ODEs result in error dynamics which are dissipative with respect to a quadratic Lyapunov function. The matrix that determines this quadratic form has some structure described by \( R(q) \). We will show in Sections VI-A and VI-B that two possible observer design techniques for unconstrained Lagrangian systems could be tailored into the framework of (13), and satisfy
the conditions for well-posedness listed in Theorem 1 and the stability requirements given in Theorem 2.

Proof of Theorem 2: The error dynamics are defined as

\[ \dot{q} = v - \hat{F}(t, x, \hat{x}) \]  

(M(q)\dot{v} + F(t, x) - \hat{F}(t, x, \hat{x}) \in -(\eta - \dot{\eta}) \)  

where

\[ \eta \in \mathcal{N}_{V(q)}(v_e) \quad \text{and} \quad \dot{\eta} \in \mathcal{N}_{V(q)}(\dot{v}_e) \]

and \(v_e, \dot{v}_e\) are defined as in (8c) and (14), respectively.

In what follows, we fix \(d\mu = dt + d\mu_a + d\tilde{\mu}_a\), where the atomic measure \(d\mu_a\) (respectively \(d\tilde{\mu}_a\)) is supported by the time instants at which \(v(\cdot)\) (respectively \(\dot{v}(\cdot)\)) is discontinuous. It is seen that \(dt + d\mu_a\) and \(dt + d\tilde{\mu}_a\) are absolutely continuous with respect to \(d\mu\) and hence the densities \(\frac{d\mu}{dt}\) and \(\frac{d\mu}{d\mu}(\cdot)\) are well-defined on the complement of a \(d\mu\)-null set.

Pick \(W(q, \tilde{x}) = \tilde{x}^TR(q)\tilde{x}\), then \(W(\cdot)\) is locally rcbv using the chain rule [26, Theorem 3], and its differential is computed as follows:

\[ \frac{dW}{d\mu}(t) = (\tilde{x}(t^+) + \tilde{x}(t^-))^\top R(q) \frac{d\tilde{x}}{d\mu}(t) + \frac{\partial}{\partial q}(\tilde{x}(t^-))^\top R(q(t)) \tilde{x}(t^+) \frac{dq}{d\mu}(t). \]

At time \(t_k\), if there is a jump in \(v(\cdot)\) or \(\dot{v}(\cdot)\), then \(\frac{dt}{d\mu}(t_k) = 0\), which also implies that \(\frac{dq}{d\mu}(t_k) = 0\). We thus obtain:

\[ \frac{dW}{d\mu}(t_k) = (\tilde{v}_k^+ + \tilde{v}_k^-)^\top M_k \frac{d\tilde{v}_k}{d\mu}(t_k) = (\tilde{v}_k^+ + \tilde{v}_k^-)^\top M_k (\tilde{v}_k^+ - \tilde{v}_k^-), \]

where we used the notation \(\tilde{v}_k^+ := \tilde{v}(t_k^+), \tilde{v}_k^- := \tilde{v}(t_k^-), \) and \(M_k := M(q(t_k))\). We can rewrite the above expression as:

\[ \frac{dW}{d\mu}(t_k) = \tilde{v}_k^{+\top} M_k \tilde{v}_k^+ - \tilde{v}_k^{-\top} M_k \tilde{v}_k^- + \frac{(1-e)}{(1+e)}(\tilde{v}_k^+ - \tilde{v}_k^-)^\top M_k (\tilde{v}_k^+ - \tilde{v}_k^-) - \frac{(1-e)}{(1+e)}(\tilde{v}_k^+ - \tilde{v}_k^-)^\top M_k (\tilde{v}_k^+ - \tilde{v}_k^-) \]

\[ = \frac{1}{(1+e)} [(1+e) \left( \tilde{v}_k^{+\top} M_k \tilde{v}_k^+ - \tilde{v}_k^{-\top} M_k \tilde{v}_k^- \right) + (1-e)(\tilde{v}_k^+ - \tilde{v}_k^-)^\top M_k (\tilde{v}_k^+ - \tilde{v}_k^-)] \]

\[ = 2 \frac{(1-e)}{(1+e)}(\tilde{v}_k^+ - \tilde{v}_k^-)^\top M_k (\tilde{v}_k^+ - \tilde{v}_k^-) \]

\[ = 2 \frac{(1-e)}{(1+e)} \langle M_k(\tilde{v}_k^+ - \tilde{v}_k^-), \tilde{v}_k^+ + e\tilde{v}_k^- \rangle - \frac{(1-e)}{(1+e)}(\tilde{v}_k^+ - \tilde{v}_k^-)^\top M_k (\tilde{v}_k^+ - \tilde{v}_k^-). \]

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Substituting $M(q(t_k))(\tilde{v}^+ - \tilde{v}^-) = - (\eta_k - \hat{\eta}_k)$, the above equation becomes

$$
\frac{dW}{d\mu}(t_k) = -2\left\langle \eta_k - \hat{\eta}_k, \frac{\tilde{v}_k^+ + e\tilde{v}_k^-}{1 + e} \right\rangle - \frac{(1 - e)}{(1 + e)}(\tilde{v}_k^+ - \tilde{v}_k^-)^\top M_k(\tilde{v}_k^+ - \tilde{v}_k^-).
$$

(29)

By definition (see (3)), it follows that

$$
\eta_k \in \mathcal{N}_{V^c(q(t_k))}(v_e(t_k)) \iff \left\langle \eta, v_e - \hat{v}_e \right\rangle \geq 0
$$

(30)

$$
\hat{\eta}_k \in \mathcal{N}_{V^c(q(t_k))}(\hat{v}_e(t_k)) \iff \left\langle \hat{\eta}, v_e - \hat{v}_e \right\rangle \leq 0
$$

(31)

which in turn implies that

$$
\left\langle \eta_k - \hat{\eta}_k, v_e(t_k) - \hat{v}_e(t_k) \right\rangle \geq 0
$$

or equivalently,

$$
\left\langle \eta_k - \hat{\eta}_k, \frac{\tilde{v}_k^+ + e\tilde{v}_k^-}{1 + e} \right\rangle \geq 0.
$$

(32)

Using the inequality (32) in equation (29), we get

$$
\frac{dW}{d\mu}(t_k) \leq - \frac{(1 - e)}{(1 + e)}(\tilde{v}_k^+ - \tilde{v}_k^-)^\top M_k(\tilde{v}_k^+ - \tilde{v}_k^-) \leq 0.
$$

(33)

Thus, when $0 \leq e < 1$, we have a strict decrease in the value of Lyapunov function $W(\cdot)$ at jump instants, and $W(\cdot)$ at most remains constant for the case $e = 1$.

If $t \neq t_k$, then we are interested in computing $\frac{dW}{d\mu_c}(t)$, where $\mu_c$ denotes the continuous part of the measure $\mu$, that is, $\mu_c$ is the sum of the absolutely continuous component of $\mu$ and its singularly continuous component. In particular, $\frac{d\mu_c}{d\mu}(t) = 1$ when $t \neq t_k$, and

$$
\frac{dW}{d\mu}(t) = \left\langle R(q)t, t, \frac{dx}{d\mu}(t) - \frac{d\hat{x}}{d\mu}(t) \right\rangle + \frac{\partial}{\partial q}(\hat{x}(t)R(q(t))\hat{x}(t)) \frac{dq}{d\mu}(t)
$$

$$
= 2\tilde{x}^\top(t)R(q) \begin{pmatrix} v - \hat{F}_1(t, x) \\ -F(t, x) + \hat{F}_2(t, x, \hat{x}) \end{pmatrix} \frac{dt}{d\mu}(t)
$$

$$
- \tilde{v}^\top(t)(\eta - \hat{\eta}) + \tilde{x}(t)\hat{R}(q(t), v(t))\tilde{x}(t) \frac{dt}{d\mu}(t).
$$

In the above expression, $\tilde{v}^\top(\eta - \hat{\eta}) \geq 0$ because $v(t), \hat{v}(t) \in V(q(t))$ for all $t$ at which $v, \hat{v}$ are continuous, due to which $\left\langle \eta, v - \hat{v} \right\rangle \geq 0$, and $\left\langle \hat{\eta}, v - \hat{v} \right\rangle \leq 0$. It now follows under condition (26) that

$$
\frac{dW}{d\mu}(t) \leq -\beta W(t) \frac{dt}{d\mu}(t), \quad t \neq t_k.
$$

(34)
Since we had fixed \( d\mu = d\mu_c + d\mu_a \), and \( W \) is non-increasing at the atoms of \( d\mu_a \) because of (33), and decreasing exponentially with respect to continuous measure due to (34). One can now invoke the chain rule for differential of bounded variation functions [26] to arrive at the following inequality (the formal arguments can also be found in our recent work [37, Proof of Theorem 1]):

\[
W(t) \leq e^{-\beta t} W(0) \Rightarrow |\tilde{\nu}(t)| \leq \sqrt{\frac{\lambda_M}{\lambda_M}} e^{-\beta t} |\tilde{\nu}(0)|.
\]

for all \( t \geq 0 \).

It is worth mentioning that, in order to deal with discontinuities of \( v_e \), we don’t just consider the classical derivative of the storage function, but instead compute the density of \( dW \) with respect to \( d\mu \). We also remark that the condition (26) was introduced explicitly to obtain dissipation of smooth part of the error dynamics with respect to kinetic metric. It basically highlights the fact that if there is any observer available in the literature for smooth Lagrangian systems for which the continuous error dynamics admit \( \tilde{x}^T R(q) \tilde{x} \) as the Lyapunov function, then those designs could be embedded into the formalism of (13) to arrive at a different criteria for error convergence. We now show two particular instances of how observers in the literature for unconstrained Lagrangian systems can be modified to fit in the framework of (13), and satisfy the conditions required for well-posedness and convergence of state estimation error.

A. Full-Order Observer

In order to arrive at a result on convergence of velocity estimation error, we introduce additional structure on the nonlinear term \( F(t, q, v) \) which is natural for Lagrangian dynamical systems. We suppose that the following assumption holds:

**Assumption 1.** The velocity \( v(\cdot) \) obtained as a solution to (8) stays bounded, that is:

\[
v(t) \in B_v := \{ v \in \mathbb{R}^n \mid |v| \leq C_v \} \quad \forall t \geq 0.
\]

(35)

The following properties are satisfied by such systems [28]:

**P1** If \( \dot{M}(q, v) \) denotes the derivative of the mass matrix, then \( \dot{M}(q, v) - 2C(q, v) \) is a skew-symmetric operator, that is, \( \tilde{\nu}^T (\dot{M}(q, v) - 2C(q, v)) \tilde{\nu} = 0, \forall \tilde{\nu} \in \mathbb{R}^n \). Here, \( C(q, v)v \) is defined using Christoffel symbols and denotes the Coriolis and centrifugal torques.
(P2) There exists a constant $C_M(q) > 0$ s.t.

$$||C(q, v)|| \leq C_M(q)|v|, \quad \forall v \in B_v.$$  \hspace{1cm} (36)

Before describing the observer dynamics, we let $\overline{F}(t, q, \cdot)$ denote the Lipschitz extension\(^3\) of $F(t, q, \cdot)$ from $B_v$ such that there exists $C_F(t, q)$ satisfying

$$|\overline{F}(t, q, v_1) - \overline{F}(t, q, v_2)| \leq C_F(t, q) \cdot |v_1 - v_2|, \quad \forall v_1, v_2 \in \mathbb{R}^n$$

and it is understood by definition that $\overline{F}(t, q, v) = F(t, q, v)$ for $v \in B_v$. The idea of using Lipschitz extension of the system vector fields for state estimators appeared in [33].

The following full-state observer is an adaptation of the design presented in [6] for unconstrained Lagrangian systems:

$$\dot{z}_1 = z_2 + L_d(q - z_1) \quad \hspace{1cm} (37a)$$

$$M(q) \dot{z}_2 + \overline{F}(t, q, \hat{v}) dt - (L_{\mu_1} + M(q)L_{\mu_2})(q - z_1) dt \in -\mathcal{N}_V(q)(\hat{v}) \quad \hspace{1cm} (37b)$$

where we let the estimates to be\(^4\)

$$\hat{q} = z_1 \quad \hspace{1cm} (37c)$$

$$\hat{v} = z_2 + l_d \hat{q}. \quad \hspace{1cm} (37d)$$

It is seen that the observer (37) indeed fits within the general framework proposed in (13) and satisfies the assumptions required for well-posedness.

The matrices $L_{\mu_1}$ and $\Lambda$ are symmetric, positive definite, and the matrices $L_d$ and $L_{\mu_2}$ are defined as follows:

$$L_d := l_d I + \Lambda, \quad L_{\mu_2} := l_d \Lambda$$

\(^3\)For a locally Lipschitz function $F(t, q, \cdot) : B \to \mathbb{R}^n$, the function $\overline{F}(t, q, \cdot) : \mathbb{R}^n \to \mathbb{R}^n$ is called the Lipschitz extension of $F(t, q, \cdot)$ from $B \subseteq \mathbb{R}^n$ if $\overline{F}(t, q, \cdot)$ is globally Lipschitz over $\mathbb{R}^n$ and $\overline{F}(t, q, v) = F(t, q, v)$ for all $(t, q) \in \mathbb{R} \times \mathbb{R}^n$ and $v \in B$.

\(^4\)The definition of $\hat{v}$ considered in [6] is different than the definition of $\hat{v}$ considered here. In [6], the authors take $\hat{v} = z_2 + L_d \tilde{q}$, whereas in our definition $\hat{v} = z_2 + l_d \tilde{q}$. Due to this difference, the error variable $(\tilde{q}^T, \tilde{v}^T)^T$ in our calculations is a linear transformation of the error variable considered in [6]. The reason for introducing this linear transformation is that it allows us to work with a quadratic Lyapunov function $\tilde{x}^T R(q) \tilde{x}$ where $R(q)$ is block-diagonal as required in the statement of Theorem 2.
for some scalar $l_d > 0$. One can equally write
\[
\dot{q} = \dot{\hat{q}} - \Lambda (q - \hat{q}) \tag{38}
\]
\[
M(q) \dot{\hat{v}} + \overline{F}(t, q, \dot{\hat{v}}) dt - L_{\rho 1}(q - \hat{q}) dt - l_d M(q)(v - \dot{\hat{v}}) dt \in - (\eta - \hat{\eta}) \tag{39}
\]

**Corollary 2.** Consider system (8) under hypotheses (H1) - (H4) and assume that the properties (P1), (P2), and Assumption 1 hold. For the estimator (37), if $l_d > 0$ is chosen such that the condition
\[
\lambda M l_d > C_F(t, q) + C_M(q) C_v + \beta
\]
is satisfied for all $(t, q) \in \mathbb{R}_+ \times \Phi$, and some constant $\beta > 0$, then the estimates $\hat{q}(\cdot), \hat{v}(\cdot)$ given by (37c) and (37d) respectively, converge to $q(\cdot), v(\cdot)$ exponentially, that is, for some $c > 0$,
\[
|v(t) - \hat{v}(t)| \leq c e^{-\beta t} |v(0) - \hat{v}(0)|. \tag{40}
\]

**Proof.** To show that (26) holds, we let $R_{11} := L_{\rho 1}, \overline{F}_1(t, x, \hat{x}) = \dot{\hat{q}} - \Lambda (q - \hat{q})$, and $\overline{F}_2(t, x, \hat{x}) = \overline{F}(t, q, \hat{v}) - L_{\rho 1} (q - \hat{q}) - l_d M(q)(v - \dot{\hat{v}})$, and observe that
\[
\begin{align*}
\tilde{q}^T L_{\rho 1} (v - \Lambda \tilde{q}) + \tilde{v}^T (\dot{F}(t, q, v) + \overline{F}(t, q, \dot{\hat{v}}) - L_{\rho 1} \tilde{q} - l_d M(q) \tilde{v}) + \tilde{v}^T \dot{M}(q, v) \tilde{v} &
\leq -\tilde{q}^T L_{\rho 1} \Lambda \tilde{q} - l_d \tilde{v}^T M(q) \tilde{v} + \tilde{v}^T (\overline{F}(t, q, \dot{\hat{v}}) - \dot{F}(t, q, v) + C(q, v) \tilde{v}) \\
&
\leq -\tilde{q}^T L_{\rho 1} \Lambda \tilde{q} - l_d \tilde{v}^T \tilde{v} + \tilde{v}^T (C_F(t, q)(\tilde{v}) + C_M(q) C_v |\tilde{v}|) \\
&
\leq -\tilde{q}^T L_{\rho 1} \Lambda \tilde{q} - \beta |\tilde{v}|^2
\end{align*}
\]
and the exponential decay of the state estimation error now follows from Theorem 2. \qed

**B. Partial-Order Observer**

One can also design a reduced order observer to show that the conditions of Theorem 2 hold in such case. Consider the following state estimator
\[
M(q) d\hat{z} + \overline{F}(t, q, \hat{z}) dt - l_d M(q) \dot{\hat{v}} dt \in - N_{V(q)}(\hat{v}_e) \tag{41a}
\]
where we let
\[
\begin{align*}
\hat{q}(t) &= q(t) \tag{41b} \\
\hat{v}(t) &:= z(t) - l_d q(t), \tag{41c} \\
\hat{v}_e(t) &:= \frac{\dot{\hat{v}}(t^+) + e\dot{\hat{v}}(t^-)}{1 + e} \tag{41d}
\end{align*}
\]
and $q(\cdot)$ in (41a) and (41c) is an absolutely continuous function of time which is obtained from (8) as the measured output. The initial condition $\hat{v}(0) \in V(q(0))$. Once again, it is seen that the observer (41) falls under the class of estimators proposed in (13), and satisfies the regularity conditions stated in Theorem 1 for well-posedness. For error convergence, we have the following result similar to Corollary 2.

**Corollary 3.** Consider system (8) under hypotheses (H1) - (H4) and assume that the properties (P1), (P2), and Assumption 1 hold. For the estimator (41), if the constant $l_d > 0$ is chosen such that the condition

$$l_d \Delta M \geq 2C_v C_M(q) + 2C_F(t, q) + \beta,$$

(42)

for all $(t, q) \in \mathbb{R}_+ \times \Phi$, and some constant $\beta > 0$, then the velocity estimate $\hat{v}(\cdot)$ given by (41c) converges to $v(\cdot)$ exponentially, that is, for some $c > 0$,

$$|v(t) - \hat{v}(t)| \leq c e^{-\beta t} |v(0) - \hat{v}(0)|.$$

(43)

Since we have chosen $\hat{q}(t) = q(t)$, we have $\hat{q}(t) \equiv 0$. We let $\hat{F}_1(t) = v(t)$, and observe that

$$\begin{align*}
\tilde{v}^T (-F(t, q, v) + \overline{F}(t, q, \hat{v}) - l_d M(q) \tilde{v}) + \tilde{v}^T \overline{M}(q, v) \tilde{v} \\
\leq -l_d \tilde{v}^T M(q) \tilde{v} + \tilde{v}^T (\overline{F}(t, q, \hat{v}) - F(t, q, v) + C(q, v) \tilde{v}) \\
\leq -l_d \Delta M |\tilde{v}|^2 + |\tilde{v}| (|C_F(t, q)(q)| |\varepsilon| + C_M(q) C_v |\varepsilon|) \\
\leq -\beta |\tilde{v}|^2
\end{align*}$$

from where (26) follows and Theorem 2 can now be invoked to show the asymptotic convergence of the state estimate.

**C. Passivity Interpretation**

Lagrangian systems are basically modeled such that the total energy, that is, the sum of kinetic and potential energy, of the system decreases with the passage of time. The kinetic energy is obtained by the quadratic form of $v$ induced by the symmetric positive definite mass matrix $M(q)$. When dealing with impacts, the kinetic energy actually dissipates at each impact. This allows one to state the dissipativity of Lagrangian systems subjected to unilateral constraints and impacts, see [8, Sections 6.8.2 and 7.2.4].
Inspired by these preliminary results, the basic idea behind the observer design is to realize an interconnection of three passive blocks as shown in Fig. 3. To analyze the passivity of each one of these blocks, we introduce the variables

\[ \chi := R(q)\dot{x} + \dot{R}(q,v)\ddot{x}, \quad \nu := \begin{pmatrix} 0_{1 \times n} \\ \eta - \hat{\eta} \end{pmatrix}. \]

It is then seen that

- We have a passive interconnection from \( \chi + \nu \) to \( \ddot{x} \):

\[ \langle \ddot{x}, \chi + \nu \rangle \geq \int_{0}^{T} \ddot{x}^\top(s)\chi(s)ds = \int_{0}^{T} \frac{dW}{d\mu}d\mu \geq -W(\ddot{q}(0), \ddot{v}(0)). \]

- Also, \( \chi = R(q) \begin{pmatrix} v - \hat{F}_1(t,x,\dot{x}) \\ -F(t,x) + \hat{F}_2(t,x,\dot{x}) \end{pmatrix} + \ddot{x}^\top \dot{R}(q,v)\ddot{x} \) and the output injection gain is chosen such that the condition (26) holds, so that

\[ \langle -\chi, \ddot{x} \rangle_{L_2} = -\int_{0}^{T} \chi(s)^\top \ddot{x}(s)ds \geq \int_{0}^{T} \ddot{x}^\top(s)R(q(s))\ddot{x}(s)ds \geq c\|\ddot{x}\|_2^2 \]

where \( c > 0 \) is some constant, and \( \|\ddot{x}\|_2 \) denotes the \( L_2 \) norm.

- Having chosen \( W(q, \ddot{x}) = \ddot{x}^\top R(q)\ddot{x} \), the above two items guarantee that the error is actually decreasing during the continuous motion of the error dynamics, so that

\[ \frac{dW}{d\mu}(t) \leq -\langle \eta - \hat{\eta}, \ddot{v}_e(t) \rangle. \]

---

Fig. 3: Interpretation of error dynamics (27) in terms of passivity.
In equation (29), it is shown that at any instant $t_k$ where $\tilde{v}$ jumps, we have

$$\frac{dW}{d\mu}(t_k) \leq -\frac{2}{(1 + e)} \langle \eta - \hat{\eta}, \tilde{v}(t_k) \rangle.$$

Next, we invoke the monotone property of the normal cone to convex sets which results in

$$\langle \eta - \hat{\eta}, \tilde{v}(t) \rangle \geq 0.$$

This last argument could also be interpreted as saying that we have passivity from $\tilde{v}$ to $-(\eta - \hat{\eta})$ with respect to the storage function $W(\cdot)$.

\textbf{D. Sensitivity to the Coefficient of Restitution}

The coefficient of restitution $e$ depends on various physical factors, and in general it is hard to get an accurate value of this coefficient. For designing estimators, it was assumed that this coefficient $e$ is known exactly and is taken to be the same as the one that governs the real plant dynamics. A natural question to ask now is to what extent the results presented in this manuscript are robust with respect to variations in the parameter $e$. In order to analyze this situation, we work with some value of coefficient of restitution $\hat{e}$ for our estimator, and it is assumed that $\hat{e} = e + \delta_e$ where $\delta_e$ is sufficiently small. The observer we will simulate in this case is exactly the same, however, $\hat{v}_e$ is now defined as

$$\hat{v}_e(t) = \frac{\hat{v}(t^+) + \hat{v}(t^-)}{1 + \hat{e}}.$$

It can be assumed without loss of generality that $\hat{e} \in [0, 1]$. The results on well-posedness of the observer dynamics hold in exactly the same manner. In the analysis of stability of error dynamics at jump instants, however, we see that the error after the jump may not decrease due to this uncertainty in the value of $e$. The exact expression comes out to be

$$\frac{dW}{d\mu}(t_k) \leq -\frac{1 - e\hat{e}}{(1 + e\hat{e} + 2e)} (\hat{v}_e^+ - \hat{v}_e^-)^\top M_k (\hat{v}_e^+ - \hat{v}_e^-)$$

$$- 2\delta_e \langle M_k (\hat{v}_e^+ - \hat{v}_e^-), v_k^+ - v_k^- \rangle,$$

where the last term can be bounded by a term which is linear in $|\hat{v}_e^+ - \hat{v}_e^-|$ because $|v| \leq C_v$ by assumption. Thus, the error at jump instants may increase due to uncertainty in the value of $e$, but due to the strictly negative first term, the discrete dynamics are input-to-state stable with $\delta_e$ as the input. Through standard calculations one can find a ball around the origin, parameterized by $\delta_e$, where the state estimation error converges as $k$ tends to infinity. Since, we allow for Zeno phenomenon, this may happen in finite time. However, between two discrete events, if one can
guarantee a certain bound on flow time, then the effect of $\delta_e$ will diminish during that time. Note that the time between two Zeno phenomena is actually controlled by the external forces. Thus, if certain restrictions are imposed on the external forces which do not allow accumulations of jumps to occur too often, then error dynamics will still be convergent despite the small uncertainties in the knowledge of $e$.

VII. NUMERICAL IMPLEMENTATION – NEEDS TO BE REVISED

We now give some remarks on how to execute the measure differential inclusion of type (13). Since the solutions are allowed to accumulate in finite time, the classical event-driven schemes are not suitable for simulating such systems. We therefore propose a time-stepping scheme to simulate (13), inspired by the Moreau-Jean algorithm [2]. It works as follows: Consider the time interval $[0,T]$ and a sampling time $\tau_s$ small enough. With $z(0)$ arbitrarily chosen, we want to solve for $z(t_{k+1})$, $t_k = k\tau_s$, $k \in \mathbb{N}$, and on the interval $[t_k,t_{k+1})$, $z$ is obtained through some interpolation techniques. In the sequel, we discuss how to compute $z(t_{k+1})$ by addressing two cases.

**Sampling interval without impact:** The easier case is when it is known that the measured position $q$ does not have any contact with the boundary of the set $\Phi$ over the entire interval $(t_k,t_{k+1}]$. In that case, $V(q(t)) = \mathbb{R}^n$, for each $t \in (t_k,t_{k+1}]$ and the multivalued part on the right-hand side of (13b) is reduced to $\{0\}$. We then obtain $z(t_{k+1})$ by classical numerical integration algorithms. For example, using a semi-explicit Euler’s method, we obtain:

$$z_1(t_{k+1}) = z_1(t_k) + \tau_s F_1(t_{k+1}, q(t_{k+1}), z(t_k))$$
$$z_2(t_{k+1}) = z_2(t_k) - \tau_s \left( M(q(t_{k+1}))^{-1} F_2(t_{k+1}, q(t_{k+1}), z(t_k)) \right)$$

and the discretized state estimates are defined as:

$$\hat{q}(t_{k+1}) = f_1(z_1(t_{k+1}), q(t_{k+1}))$$
$$\hat{v}(t_{k+1}) = z_2(t_{k+1}) - f_2(q_{k+1}, z_1(t_{k+1})).$$

**Sampling intervals with impact:** If, however, the contact between $q$ and $\Phi$ is detected during the interval $(t_k,t_{k+1}]$ then we basically reformulate the inclusion (13b) as a complementarity relation to determine the post impact value of $\hat{v}$. In order to state the desired complementarity relation, suppose that the contact happens at some time $\bar{t} \in (t_k,t_{k+1}]$, that is, $h_{\alpha}(\bar{q}) \leq 0$, for
some $\alpha = 1, \ldots, m$, where $\bar{q} := q(\bar{t})$. It follows from the inclusion (13) that:

$$z(\bar{t}^+) - z(\bar{t}^-) \in -M(\bar{q})^{-1}N_{V(\bar{q})}(\hat{v}_e(\bar{t}))$$

and from the definition of $V(\bar{q})$, it follows that

$$z(\bar{t}^+) - z(\bar{t}^-) = \sum_{\alpha \in J(\bar{q})} \lambda_\alpha \frac{M^{-1}(\bar{q})\nabla h_\alpha(\bar{q})}{|M^{-1/2}(\bar{q})\nabla h_\alpha(\bar{q})|}, \quad \lambda_\alpha \geq 0, \alpha \in J(\bar{q}).$$

We are thus interested in finding the value of $\lambda_\alpha, \alpha \in J(\bar{q})$. Also, it follows by definition that

$$\hat{v}_e(\bar{t}) \in V(\bar{q}) \quad \text{and} \quad \langle \hat{v}(\bar{t}^+) - \hat{v}(\bar{t}^-), \hat{v}(\bar{t}^+) + e\hat{v}(\bar{t}^-) \rangle_{M(\bar{q})} = 0.$$

The preceding relations are in turn equivalent to solving the following complementarity problem:

$$0 \leq \lambda_\alpha \perp \left\langle \frac{M^{-1}(\bar{q})\nabla h_\alpha(\bar{q})}{|M^{-1/2}(\bar{q})\nabla h_\alpha(\bar{q})|}, M(\bar{q})(\hat{v}(\bar{t}^+) + e\hat{v}(\bar{t}^-)) \right\rangle \geq 0, \quad \alpha \in J(\bar{q}). \quad (44)$$

In theory, solving for $\lambda_\alpha$ allows us to compute $\hat{v}(\bar{t}^+)$ whenever $h_\alpha(q(\bar{t})) = 0$. Knowing that there might be infinitely many such instances in finite interval, running such an algorithm in practice is not feasible on machines with finite precisions. However, note that the second term in (44) is continuous with respect to the variable $q$ and since $q$ is an absolutely continuous function of time, the value of $\lambda_\alpha$ will not vary a lot with respect to small variations in the value of $q(\bar{t})$. Hence, when implementing this algorithm on the computer, we replace $q(\bar{t})$ with $q(t_{k+1})$ and compute $\hat{v}(t_{k+1})$. With small enough sampling time, the resulting values of $\hat{v}$ are still accurate enough (see [2, Chapter 14] for numerical tests with various step sizes, in particular [2, Table 14.2], see also [23]). The important thing is to know which constraints were active during a certain sampling interval. Thus, we don’t count the number of times $q$ makes contact with the boundary of $\Phi$ during a particular sampling interval, we just count the number of constraints that are active during a particular sampling interval. Since, there are at most $m$ constraints, the above algorithm is feasible for implementation as shown in the following examples.

A. Simulation Results

We now apply our results to the three systems described in Figure 1 given in the introduction. Since the velocity variable is the only quantity of interest that needs to be estimated, we will only implement the observer (41) for these systems. This observer has been implemented in the software platform SICONOS [1]. The animations for the examples appearing in the sequel, and several different cases, can be found online at [35].
Example 1. For the 3-ball chain with wall constraints given in Fig. 1a, let the position and velocity of the center of the ball be denoted by \( q_i \) and \( v_i \), respectively, \( i = 1, 2, 3 \). Also, suppose that each ball has a unit mass, and have a fixed radius \( r = 0.5 \). The distance between the two walls is denoted by \( l = 15 \). The constraint relations for this systems are:

\[
\begin{align*}
    h_1(q) &= q_2 - q_1 - 2r \geq 0 \\
    h_2(q) &= q_3 - q_2 - 2r \geq 0 \\
    h_3(q) &= q_1 - r \geq 0 \\
    h_4(q) &= l - q_3 - r \geq 0.
\end{align*}
\]

The simulation results are given in Fig. 4. The key aspect of our observer design is that the estimated velocity respects the same constraints as imposed on the actual velocity of the system. It is also important to note that not every impact of the ball with the ground causes a discontinuity in \( \hat{v}(\cdot) \), as can be seen in the transient response of the estimated velocity components.

Example 2. As a second example, we consider a particle of unit mass bouncing in two-dimensional plane with a parabolic and linear constraint, see Fig. 1b. We denote the position variable by \( (q_x, q_y) \) and the corresponding velocity vector by \( (v_x, v_y) \). The constraint relations are
Fig. 5: Velocity estimation for a 2-degree of freedom mechanical system: A ball bouncing inside a parabola with ceiling. The top plot shows the evolution of two gap functions $h_1(q)$ and $h_2(q)$ with time. The middle and bottom plot show the velocity components $v_x$ and $v_y$, along with their estimates, respectively. The boxes in the center of these plots provides a magnified image of the quantity of interest around the accumulation point $t = 4.5s$.

defined as:

$$h_1(q) = q_y - q_x^2 \geq 0$$
$$h_2(q) = c - q_y \geq 0,$$

where we choose $c = 8$ for the sake of simulations. Using the notation of (8), we choose $F(t, q, v) = 9.81$, that is, the point mass is subjected to the gravitational force, and the coefficient of restitution at impacts is $e = 0.9$. This choice eventually leads to accumulation of impacts in finite time, as one sees from the plots in Fig. 5. Initially, when either of the constraints $h_1$ or $h_2$ becomes zero, a jump in at least one of the velocity components is observed. The accumulation of impacts is observed around $t = 4.5s$, because the mass is being pulled downward continuously by gravity, and the dissipative reaction force (that acts on the particle to maintain the constraint $h_1(q) \geq 0$) reduces the norm of the velocity at each impact. Eventually, after the accumulation point, we see that $h_1(q)$ remains identically zero and sliding of the particle on the reduced-order
Fig. 6: Simulation of velocity estimator for a particle moving in a nonconvex hyperbolic billiard.

Surface \( \{ q \mid h(q) = 0 \} \) is observed. Our velocity estimator replicates this phenomenon, and after the initial transients, the estimates converge to the actual velocity of the particle.

**Example 3.** We now address the system depicted in Fig. 1c, which comprises a particle of unit mass moving in a two-dimensional hyperbolic billiard. Using the notation used in Example 2, the constraint relations that we consider are:

\[
\begin{align*}
    h_1(q) &= 1 - q_x q_y \geq 0 \\
    h_2(q) &= 1 + q_x q_y \geq 0.
\end{align*}
\]

One can see that the admissible configuration set for the position \( q \) of the particle is nonconvex and unbounded. Because of the particular shape of this admissible set, the trajectories of the system show a chaotic behavior. In the center of the billiard, there are relatively few impacts, but as the particle enters in one of the narrow tubes, the frequency of the impacts is very rapid to the extent that it is not possible to put a lower bound on the time between two impacts. When the particle comes out, there is again a phase where the impacts are relatively few. Such a behavior makes the system also very sensitive to numerical errors. The animations on the author’s webpage [35] may help understand this discussion better. However, if the position of
the particle is exactly known, as it is assumed for our observer, the methods proposed in this paper, also allow us estimate this kind of chaotic velocity trajectories, as illustrated in Fig. 6.

VIII. CALCULATIONS FOR THEOREM 1

In this section, we show calculations for the claims made in the proof of Theorem 1.

A. Estimate of a uniform bound on \( \hat{v}_p \)

It is assumed that \( |v(t)| \leq C_v \) for all \( t \in [0, T] \), so that \( q(t) \in B(q_0, C_v T) \). Let \( C_M^{1/2} \) be the Lipschitz constant associated with the mapping \( q \mapsto M^{1/2}(q) \) on \( B(q_0, C_v T) \). The projection, with respect to the norm induced by \( M_{p,i} := M(q(t_{p,i})) \), on the set \( V(q(t_{p,i})) \) is denoted by \( P_{p,i} \) and on the set \( M^{-1}(q(t_{p,i}))V^0(q(t_{p,i})) \) by \( Q_{p,i} \). We denote by \( \bar{\lambda}_M, \Delta_M \) the constants introduced in (12) for the compact set \( B(q_0, C_v T) \). Let \( u_{p,i} \) be defined as:

\[
    u_{p,i} := \hat{v}_{p,i-1} - \frac{1}{1 + e} M_{p,i}^{-1} G_{p,i}.
\]

These notations are now used in deriving a bound on \( \hat{v}_p \). Using Moreau’s two-cone lemma (see Lemma A.1 in Appendix A), we first get:

\[
    |\hat{v}_{p,i}|_{M_{p,i}} = |P_{p,i}(u_{p,i}) - eQ_{p,i}(u_{p,i}) - \frac{e}{1 + e} M_{p,i}^{-1} G_{p,i}|_{M_{p,i}} \\
    \leq |u_{p,i}|_{M_{p,i}} + \frac{e}{1 + e} \| M_{p,i}^{-1/2} \| \cdot |G_{p,i}| \\
    \leq |\hat{v}_{p,i-1}|_{M_{p,i}} + \frac{1}{\sqrt{\Delta_M}} |G_{p,i}| \\
    \leq |\hat{v}_{p,i-1}|_{M_{p,i-1}} + \| M_{p,i-1}^{1/2} - M_{p,i}^{1/2} \| \cdot |\hat{v}_{p,i-1}| + \frac{1}{\sqrt{\Delta_M}} |G_{p,i}| \\
    \leq |\hat{v}_{p,i-1}|_{M_{p,i-1}} + C_M^{1/2} \int_{t_{p,i-1}}^{t_{p,i}} |v(s)| ds \cdot |\hat{v}_{p,i-1}| + \frac{1}{\sqrt{\Delta_M}} |G_{p,i}|.
\]

Since \( \hat{v}_p(\cdot) \) has the constant value \( \hat{v}_{p,i} \) on the interval \( [t_{p,i}, t_{p,i+1}] \), the above inequality results in:

\[
    \sqrt{\Delta_M} |\hat{v}_p(t_{p,i})| - \sqrt{\Delta_M} |\hat{v}_p(0)| \leq \sum_{j=1}^{i} |\hat{v}_{p,j}|_{M_{p,j}} - |\hat{v}_{p,j-1}|_{M_{p,j-1}} \\
    \leq C_M^{1/2} C_v \int_{0}^{t_{p,i}} |\hat{v}_p(s)| ds + \frac{C_{g,b}}{\sqrt{\Delta_M}} \int_{0}^{t_{p,i}} (1 + |\hat{v}_p(s)| ds
\]

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that is, for each \( t \in [0, T] \),
\[
|\dot{v}_P(t)| \leq \sqrt{\frac{\lambda_M}{\Delta_M}} |\dot{v}_0| + \frac{C_{g,b}}{\Delta_M} t + \left( \frac{C_{M^{1/2} C_v}}{\sqrt{\Delta_M}} + \frac{C_{g,b}}{\Delta_M} \right) \int_0^t |\dot{v}_P(s)| \, ds.
\]

Applying the Gronwall-Bellman inequality for discontinuous functions [18, Lemma 1] to (24), the following bound on \(|\dot{v}_P(t)|\) is obtained for each \( t \in [0, T] \), which does not depend on the partition \( \mathcal{P} \):
\[
|\dot{v}_P(t)| \leq \sqrt{\frac{\lambda_M}{\Delta_M}} |\dot{v}_0| + \frac{C_{g,b}}{\Delta_M} t + \left( \frac{C_{M^{1/2} C_v}}{\sqrt{\Delta_M}} + \frac{C_{g,b}}{\Delta_M} \right) \exp \left( \left( \frac{C_{M^{1/2} C_v}}{\sqrt{\Delta_M}} + \frac{C_{g,b}}{\Delta_M} \right) t \right) \leq C_{sup}
\]
where \( C_{sup} \) is obtained by evaluating the right-hand side of the first inequality at \( t = T \).

B. Estimates on the variation

For a fixed partition \( \mathcal{P} \) of the interval \([0, T]\), we now compute the total variation of \( \dot{v}_P(\cdot) \). For conciseness, we drop the subscript \( \mathcal{P} \) in the quantities appearing in (23) and (24). By definition, we have
\[
\dot{v}_i = -e\dot{v}_{i-1} + (1 + e) P_i(u_i).
\]
Using Moreau’s two-cone lemma, we can write \( u_i := P_i(u_i) + Q_i(u_i) \), so that
\[
\dot{v}_i - \dot{v}_{i-1} = -(1 + e) u_i + (1 + e) P_i(u_i) - M_i^{-1} G_i
\]
\[
= -(1 + e) Q_i(u_i) - M_i^{-1} G_i. \tag{47}
\]

Since \( Q_i(\cdot) \) denotes the projection on \( V_i^* := M_i^{-1} V^*(q_i) \) with respect to the kinetic metric, we take \( Q_i(u) = 0 \) if \( \mathcal{J}(q_i) = \emptyset \), in which case
\[
|\dot{v}_i - \dot{v}_{i-1}| = |M_i^{-1} G_i| \leq \frac{|G_i|}{\Delta_M}.
\]
Otherwise, if \( \mathcal{J}(q_i) \neq \emptyset \), we have
\[
Q_i(u) := \sum_{\alpha \in \mathcal{J}(q_i)} \langle u, \nabla h_\alpha(q_i) \rangle M_i^{-1} H_\alpha(q_i) \quad \text{where} \quad H_\alpha(q_i) := \frac{\nabla h_\alpha(q_i)}{\nabla h_\alpha^\top(q_i) M_i^{-1} \nabla h_\alpha(q_i)}
\]
and \( \langle u, \nabla h_\alpha(q) \rangle := \min\{\langle u, \nabla h_\alpha(q) \rangle, 0\} \). One may rewrite (47) as
\[
\dot{v}_i - \dot{v}_{i-1} = -(1 + e) (Q_i(u_i) - Q_i(\tilde{v}_{i-1})) - (1 + e) (Q_i(\tilde{v}_{i-1}) - Q_i(\tilde{v}_{i-1})) - (1 + e) Q_{i-1}(\tilde{v}_{i-1}) - M_i^{-1} G_i \tag{48}
\]
where
\[ \hat{Q}_{i-1}(u) := \sum_{\alpha \in \mathcal{J}(q_i)} \langle u, \nabla h_\alpha(q_{i-1}) \rangle M_{i-1}^{-1} H_\alpha(q_{i-1}). \]

We now compute an upper bound on the norm of the right-hand side of (48).

**First term:** It is noted using the contraction property of the projection map and the bound derived in (46) that
\[ |Q_i(u_t) - Q_i(\hat{v}_{i-1})| \leq \frac{1}{\sqrt{\Delta M}} |u_t - \hat{v}_{i-1}| \leq \frac{1}{(1 + e)^{1/2} \Delta M} |M_{i-1}^{-1} G_i| \leq \frac{1}{1 + e} \frac{|G_i|}{\Delta M} \]
where \( | \cdot |_i \) is used as a short-hand for \( | \cdot |_{M \cdot P_i} \) when the partition \( P \) is considered to be fixed.

**Second term:** Under the hypothesis that \( \nabla h_\alpha(\cdot) \) is locally Lipschitz continuous, for each \( \alpha = 1, \cdots, m \), and that \( q(\cdot) \) evolves within a compact set over the interval \([0, T]\), there exists a constant \( C_h \) such that
\[ |\nabla h_\alpha(q_t) - \nabla h_\alpha(q_{i-1})| \leq C_h |q_t - q_{i-1}| \leq C_h C_v |t_i - t_{i-1}|, \quad \forall t_i, t_{i-1} \in [0, T]. \]
Similarly, since \( M^{-1/2}(\cdot) \) is locally Lipschitz continuous, there exists \( C_H > 0 \) such that
\[ |H_\alpha(q_t) - H_\alpha(q_{i-1})| \leq C_H |q_t - q_{i-1}| \leq C_H C_v (t_i - t_{i-1}), \quad \forall t_i, t_{i-1} \in [0, T]. \]

Thus, we get
\[ Q_i(\hat{v}_{i-1}) - \hat{Q}_{i-1}(\hat{v}_{i-1}) = \sum_{\alpha \in \mathcal{J}(q_i)} \langle \hat{v}_{i-1}, \nabla h_\alpha(q_t) \rangle M_{i-1}^{-1} H_\alpha(q_t) - \sum_{\alpha \in \mathcal{J}(q_i)} \langle \hat{v}_{i-1}, \nabla h_\alpha(q_{i-1}) \rangle M_{i-1}^{-1} H_\alpha(q_{i-1}) \]
\[ = \sum_{\alpha \in \mathcal{J}(q_i)} \left[ \langle \hat{v}_{i-1}, \nabla h_\alpha(q_t) \rangle - \langle \hat{v}_{i-1}, \nabla h_\alpha(q_{i-1}) \rangle \right] M_{i-1}^{-1} H_\alpha(q_t) \]
\[ + \sum_{\alpha \in \mathcal{J}(q_i)} \langle \hat{v}_{i-1}, \nabla h_\alpha(q_{i-1}) \rangle \left[ (M_{i-1}^{-1} M_{i-1}^{-1} H_\alpha(q_t) + M_{i-1}^{-1} H_\alpha(q_t) - H_\alpha(q_{i-1})) \right]. \]

This further leads to
\[ |Q_i(\hat{v}_{i-1}) - \hat{Q}_{i-1}(\hat{v}_{i-1})| \leq \sum_{\alpha \in \mathcal{J}(q_i)} | \langle \hat{v}_{i-1}, \nabla h_\alpha(q_t) \rangle - \langle \hat{v}_{i-1}, \nabla h_\alpha(q_{i-1}) \rangle | \cdot |M_{i-1}^{-1} H_\alpha(q_t)| \]
\[ + \sum_{\alpha \in \mathcal{J}(q_i)} | \langle \hat{v}_{i-1}, \nabla h_\alpha(q_{i-1}) \rangle | \cdot \left( \|M_{i-1}^{-1} M_{i-1}^{-1}\| \cdot |H_\alpha(q_t)| + \|M_{i-1}^{-1}\| \cdot |H_\alpha(q_t) - H_\alpha(q_{i-1})| \right) \]
\[ \leq \frac{m}{\Delta M} C_{\sup} C_h C_v \hat{C}_{H_\alpha}(t_i - t_{i-1}) + m C_{\sup} \hat{C}_{h_\alpha} (C_{M^{-1}} \hat{C}_{H_\alpha} + C_H) C_v (t_i - t_{i-1}) \]
\[ =: C_{\text{proj}}(t_i - t_{i-1}) \]
where \( \tilde{C}_{h_{\alpha}} = \sup_{q \in B(q_0, C_v T)} |\nabla h_{\alpha}(q)| \) and \( \tilde{C}_{H_{\alpha}} := \sqrt{\frac{x}{2}} C_{h_{\alpha}} \geq \sup_{q \in B(q_0, C_v T)} |H_{\alpha}| \). The constants \( C_v, C_{sup}, C_h, \) and \( C_H \) were introduced in (35), (46), (50), and (51), respectively. The constant \( C_{M-1} \) is chosen such that \( \|M_i - M_{i-1}\| \leq C_{M-1} \), for each \( i \in \mathbb{N} \).

**Third term:** We have

\[
\hat{v}_i = -e \hat{v}_{i-1} + (1 + e) P_i(u_i) = P_i(u_i) - e Q_i(u_i) - \frac{e}{1 + e} M_i^{-1} G_i
\]

which gives

\[
\hat{v}_i + \frac{e}{1 + e} M_i^{-1} G_i = P_i(u_i) - e Q_i(u_i) \in V(q_i).
\]

This further leads to

\[
|\tilde{Q}_{i-1}(\hat{v}_{i-1})| \leq \sum_{\alpha \in J(q_i)} \left| \left\langle \hat{v}_{i-1} + \frac{e}{1 + e} M_i^{-1} G_i - \frac{e}{1 + e} M_i^{-1} G_i, \nabla h_{\alpha}(q_{i-1}) \right\rangle \right|^2 |\tilde{Q}_{i-1} H_{\alpha}(q_{i-1})|.
\]

From (53), we have

\[
\left\langle \hat{v}_{i-1} + \frac{e}{1 + e} M_i^{-1} G_i, \nabla h_{\alpha}(q_{i-1}) \right\rangle \geq 0, \quad \forall j = 1, \ldots, m,
\]

and hence\(^5\)

\[
|\tilde{Q}_{i-1}(\hat{v}_{i-1})| \leq \sum_{\alpha \in J(q_i)} \frac{e}{1 + e} \left| \left\langle M_i^{-1} G_i, \nabla h_{\alpha}(q_{i-1}) \right\rangle \right| \cdot \frac{|M_i^{-1} \nabla h_{\alpha}(q_{i-1})|_{i-1}}{\sqrt{\Delta M \nabla h_{\alpha}(q_{i-1}) M_i^{-1} \nabla h_{\alpha}(q_{i-1})}} \cdot \frac{1}{\sqrt{\Delta M}} \frac{|M_i^{-1} \nabla h_{\alpha}(q_{i-1})|_{i-1}}{\cdot |M_i^{-1} \nabla h_{\alpha}(q_{i-1})|_{i-1}} \leq \frac{me}{1 + e} \frac{|G_{i-1}|}{\Delta M}.
\]

Plugging the bounds from (49), (52), and (54) into (48), we obtain

\[
|\hat{v}_i - \hat{v}_{i-1}| \leq \frac{1}{\Delta M} |G_i| + 2C_{proj}(t_i - t_{i-1}) + \frac{1}{\Delta M} |G_i| + \frac{m}{\Delta M} |G_{i-1}|.
\]

Using the norm estimate on \( \hat{v}_{\mathcal{P}}(\cdot) \), we have \( |G_i| \leq (1 + C_{sup})(t_i - t_{i-1}) \) and thus for \( 0 \leq s < t \leq T \), it follows that

\[
\text{Var}(\hat{v}_{\mathcal{P}}(s, t]) \leq C_{var}(t - s)
\]

where \( C_{var} := \frac{1}{\Delta M} ((m + 2)(1 + C_{sup}) + 2C_{proj}) \).

\(^5\)We use the fact that for \( a, b, c \in \mathbb{R}^n \), satisfying \( \langle a, c \rangle \geq 0 \), we have \( |(a + b, c)| = \min\{0, \langle a+b, c \rangle \} \leq |\min\{0, a, c \}| + |\min\{0, b, c \}| = |(b, c)| \).
C. Continuity points of the limit function

Assume that \( y \in V(q(\tau)) \) for all \( \tau \in [s, t] \subseteq [0, T] \), then it is claimed that

\[
\int_s^t \langle -g(\sigma, \hat{v}), (y - \hat{v}) \rangle + \left\langle \hat{M}(\sigma) \hat{v}, \left( y - \frac{\hat{v}}{2} \right) \right\rangle \, d\sigma \leq \langle M(q(t))\hat{v}(t) - M(q(s))\hat{v}(s), y \rangle - \frac{1}{2} \left( |\hat{v}(t)|^2_{\hat{M}(q(t))} - |\hat{v}(s)|^2_{\hat{M}(q(s))} \right). \tag{56}
\]

In the sequel, we proceed to prove this claim:

Consider a partition \( \mathcal{P} \) of the interval \([0, T]\) that contains the nodes \( t_{\mathcal{P}, j} = s \) and \( t_{\mathcal{P}, k} = t \) for some \( j, k \in \mathbb{N} \). From the discretization scheme (23), for \( j + 1 \leq i \leq k \), we have

\[
\frac{\hat{v}_{\mathcal{P}, i} + e\hat{v}_{\mathcal{P}, i-1}}{1 + e} = P_{\mathcal{P}, i} \left( \hat{v}_{\mathcal{P}, i-1} - \frac{1}{1 + e} M_{\mathcal{P}, i}^{-1} G_{\mathcal{P}, i} \right).
\]

Using the definition of the projection operator, we get

\[
\langle -M_{\mathcal{P}, i}^{-1} G_{\mathcal{P}, i} + \hat{v}_{\mathcal{P}, i-1} - \hat{v}_{\mathcal{P}, i}, (1 + e)y - (\hat{v}_{\mathcal{P}, i} + e\hat{v}_{\mathcal{P}, i-1}) \rangle_{M_{\mathcal{P}, i}} \leq 0, \quad \forall y \in V_{\mathcal{P}, i}.
\]

The above inequality is equivalently written as:

\[
\langle -M_{\mathcal{P}, i}^{-1} G_{\mathcal{P}, i}, (1 + e)y \rangle_{M_{\mathcal{P}, i}} + \langle M_{\mathcal{P}, i}^{-1} G_{\mathcal{P}, i}, \hat{v}_{\mathcal{P}, i} + e\hat{v}_{\mathcal{P}, i-1} \rangle_{M_{\mathcal{P}, i}} \\
\leq (1 + e) \langle \hat{v}_{\mathcal{P}, i} - \hat{v}_{\mathcal{P}, i-1}, y \rangle_{M_{\mathcal{P}, i}} - \langle \hat{v}_{\mathcal{P}, i} - \hat{v}_{\mathcal{P}, i-1}, \hat{v}_{\mathcal{P}, i} + e\hat{v}_{\mathcal{P}, i-1} \rangle_{M_{\mathcal{P}, i}} \tag{57a}
\]

\[
\leq (1 + e) \left[ \langle M_{\mathcal{P}, i} \hat{v}_{\mathcal{P}, i} - M_{\mathcal{P}, i-1} \hat{v}_{\mathcal{P}, i-1}, y \rangle - \frac{1}{2} (|\hat{v}_{\mathcal{P}, i}|^2_{M_{\mathcal{P}, i}} - |\hat{v}_{\mathcal{P}, i-1}|^2_{M_{\mathcal{P}, i-1}}) \right] \\
- (1 + e) \left( \langle M_{\mathcal{P}, i} - M_{\mathcal{P}, i-1}, \hat{v}_{\mathcal{P}, i-1} \rangle - \frac{\hat{v}_{\mathcal{P}, i-1}}{2} \right). \tag{57b}
\]

To arrive at (57b), the last term in (57a) is rewritten as:

\[
\langle \hat{v}_{\mathcal{P}, i} - \hat{v}_{\mathcal{P}, i-1}, \hat{v}_{\mathcal{P}, i} \rangle_{M_{\mathcal{P}, i}} = \frac{1}{2} \left[ |\hat{v}_{\mathcal{P}, i} - \hat{v}_{\mathcal{P}, i-1}|^2_{M_{\mathcal{P}, i}} + |\hat{v}_{\mathcal{P}, i}|^2_{M_{\mathcal{P}, i}} - |\hat{v}_{\mathcal{P}, i-1}|^2_{M_{\mathcal{P}, i-1}} - \hat{v}_{\mathcal{P}, i-1}^\top (M_{\mathcal{P}, i} - M_{\mathcal{P}, i-1}) \hat{v}_{\mathcal{P}, i-1} \right] \tag{58}
\]

and

\[
\langle \hat{v}_{\mathcal{P}, i} - \hat{v}_{\mathcal{P}, i-1}, e\hat{v}_{\mathcal{P}, i-1} \rangle_{M_{\mathcal{P}, i}} = \frac{e}{2} \left[ -|\hat{v}_{\mathcal{P}, i} - \hat{v}_{\mathcal{P}, i-1}|^2_{M_{\mathcal{P}, i}} + |\hat{v}_{\mathcal{P}, i}|^2_{M_{\mathcal{P}, i}} - |\hat{v}_{\mathcal{P}, i-1}|^2_{M_{\mathcal{P}, i-1}} - \hat{v}_{\mathcal{P}, i-1}^\top (M_{\mathcal{P}, i} - M_{\mathcal{P}, i-1}) \hat{v}_{\mathcal{P}, i-1} \right].
\]
Inequality (57b) now leads to:
\[
\left\langle -G_{p,i}, y - \frac{\hat{v}_{p,i} + e\hat{v}_{p,i-1}}{1 + e} \right\rangle + \left\langle (M_{p,i} - M_{p,i-1})\hat{v}_{p,i-1}, y - \frac{\hat{v}_{p,i-1}}{2} \right\rangle \leq \\
\left\langle M_{p,i}\hat{v}_{p,i} - M_{p,i-1}\hat{v}_{p,i-1}, y - \frac{1}{2}(|\hat{v}_{p,i}|_{M_{p,i}}^2 - |\hat{v}_{p,i-1}|_{M_{p,i-1}}^2) \right\rangle
\]
which further yields
\[
\sum_{i=j+1}^{k} \left\langle -G_{p,i}, y - \frac{\hat{v}_{p,i} + e\hat{v}_{p,i-1}}{1 + e} \right\rangle + \left\langle (M_{p,i} - M_{p,i-1})\hat{v}_{p,i-1}, y - \frac{\hat{v}_{p,i-1}}{2} \right\rangle \leq \\
\left\langle M_{p,k}\hat{v}_{p,k} - M_{p,j}\hat{v}_{p,j}, y - \frac{1}{2}(|\hat{v}_{p,k}|_{M_{p,k}}^2 - |\hat{v}_{p,j}|_{M_{p,j}}^2) \right\rangle. \tag{59}
\]
Using the fact that \(\frac{\hat{v}_{p}(\tau) + e\hat{v}_{p}(\tau)}{1 + e}\) converges to \(\hat{v}(\tau)\) for Lebesgue almost all \(\tau \in [0, T]\), it follows that
\[
\sum_{i=j+1}^{k} \left\langle G_{p,i}, y - \frac{\hat{v}_{p,i} + e\hat{v}_{p,i-1}}{1 + e} \right\rangle \to \int_{s}^{t} \left\langle g(\tau, \hat{v}(\tau)), y - \hat{v}(\tau) \right\rangle \, d\tau. \tag{60}
\]
Since \(t \mapsto M(q(t))\) is an absolutely continuous function, \(\dot{M}(\tau) := \frac{d}{dt} M(q(t))\big|_{t=\tau}\) exists for Lebesgue almost-all \(\tau\), and we have
\[
\left\langle (M_{p,i} - M_{p,i-1})\hat{v}_{p,i-1}, y - \frac{\hat{v}_{p,i-1}}{2} \right\rangle = \int_{t_{p,i-1}}^{t_{p,i}} \left\langle \dot{M}(\tau)\hat{v}_{p}(\tau), y - \frac{\hat{v}_{p}(\tau)}{2} \right\rangle \, d\tau,
\]
and
\[
\sum_{i=j+1}^{k} \left\langle (M_{p,i} - M_{p,i-1})\hat{v}_{p,i-1}, y - \frac{\hat{v}_{p,i-1}}{2} \right\rangle = \int_{t_{p,j}}^{t_{p,k}} \left\langle \dot{M}(\tau)\hat{v}_{p}(\tau), y - \frac{\hat{v}_{p}(\tau)}{2} \right\rangle \, d\tau. \tag{61}
\]
For the terms on the right-hand side, we have the following convergence:
\[
\left\langle M_{p,k}\hat{v}_{p,k} - M_{p,j}\hat{v}_{p,j}, y - \frac{1}{2}(|\hat{v}_{p,k}|_{M_{p,k}}^2 - |\hat{v}_{p,j}|_{M_{p,j}}^2) \right\rangle \to \left\langle M(q(t))\hat{v}(t) - M(q(s))\hat{v}(s), y \right\rangle \\
- \frac{1}{2}(|\hat{v}(t)|_{M(q(t))}^2 - |\hat{v}(s)|_{M(q(s))}^2). \tag{62}
\]
The desired inequality (56) now follows by taking the limit in (59) along all partitions finer than \(\mathcal{P}\) and using (60), (61) and (62).

Let \(\mu\) be the measure defined by \(d\mu = |d\hat{v}| + dt\). Since \(d\hat{v}\) and \(dt\) are absolutely continuous with respect to \(d\mu\) there exists a \(d\mu\) negligible set \(A\) such that, for all \(t \in [0, T] \setminus A\):
\[
\frac{dt}{d\mu}(t) = \lim_{\epsilon \to 0^+} \frac{dt([t, t+\epsilon])}{d\mu([t, t+\epsilon])} \quad \text{and} \quad \frac{d\hat{v}}{d\mu}(t) = \lim_{\epsilon \to 0^+} \frac{d\hat{v}([t, t+\epsilon])}{d\mu([t, t+\epsilon])}.
\]
Assume that \( \dot{v} \) is continuous at \( t \) and let \( y \in \text{int} \, V(q(t)) \). Then due to lower semicontinuity of \( q \mapsto V(q) \) and absolute continuity of \( t \mapsto q(t) \), \( y \in V(q(\tau)) \) for all \( \tau \in I_\varepsilon := [t, t + \varepsilon] \). Due to the variational inequality (56), we get

\[
\int_t^{t+\varepsilon} \langle -g(\sigma, \dot{v}), (y - \dot{v}) \rangle + \langle \dot{M}(\sigma) \dot{v}(\sigma), \left(y - \frac{\dot{v}}{2}\right) \rangle \, d\sigma \leq \langle M(q(t + \varepsilon)) \dot{v}(t + \varepsilon) - M(q(t)) \dot{v}(t), y \rangle - \frac{1}{2} \left( |\dot{v}(t + \varepsilon)|_{\dot{M}(q(t + \varepsilon))}^2 - |\dot{v}(t)|_{\dot{M}(q(t))}^2 \right). \quad (63)
\]

Divide both sides by \( d\mu([t, t + \varepsilon]) \). When \( \varepsilon \to 0 \), the left-hand side of (63) converges to

\[
\frac{dt}{d\mu} \left[ \langle -g(t, \dot{v}(t)), y - \dot{v}(t) \rangle + \langle \dot{M}(t) \dot{v}(t), y - \frac{1}{2} \dot{v}(t) \rangle \right].
\]

The first term on the right-hand side of (63) becomes

\[
\frac{1}{d\mu(I_\varepsilon)} \langle M(q(t + \varepsilon)) \dot{v}(t + \varepsilon) - M(q(t)) \dot{v}(t), y \rangle = \langle M(q(t)) \frac{d\dot{v}(I_\varepsilon)}{d\mu}, \dot{v}(t), y \rangle + \int_t^{t+\varepsilon} \dot{M}(s) ds \, \dot{v}(t + \varepsilon), y \rangle
\]

\[
\quad \xrightarrow{\varepsilon \to 0} \langle M(q(t)) \frac{d\dot{v}}{d\mu}(t), \dot{v}(t), y \rangle + \langle \dot{M}(t) \dot{v}(t), y \rangle \frac{dt}{d\mu},
\]

and the second term on the right-hand side of (63) becomes

\[
\frac{1}{2d\mu(I_\varepsilon)} \left[ \langle M(q(t + \varepsilon)) \dot{v}(t + \varepsilon), d\dot{v}(I_\varepsilon) \rangle + \langle M(q(t)) d\dot{v}(I_\varepsilon), \dot{v}(t) \rangle + \int_t^{t+\varepsilon} \dot{M}(s) ds \, \dot{v}(t + \varepsilon), \dot{v}(t) \rangle \right]
\]

\[
\quad \xrightarrow{\varepsilon \to 0} \langle M(q(t)) \frac{d\dot{v}}{d\mu}(t), \dot{v}(t) \rangle + \langle \dot{M}(t) \dot{v}(t), \frac{\dot{v}(t)}{2} \rangle \frac{dt}{d\mu}.
\]

Thus, in the limit (63) leads to

\[
\langle M(q(t)) \frac{d\dot{v}}{d\mu}(t) + g(t, \dot{v}(t), \frac{dt}{d\mu}(t), y - \dot{v}(t) \rangle \geq 0, \quad \forall y \in \text{int} \, V(q(t)).
\]

From the definition of \( \mathcal{N}_{V(q(t))}(\cdot) \), and using the density argument, it follows that \( \dot{v}(\cdot) \) satisfies the differential inclusion (17) at the continuity points of \( \dot{v}(\cdot) \).

**D. Impact characterization of the limit solution**

Let \( t_k \in [0, T] \) be the time instant at which \( V(q(t_k)) \neq \mathbb{R}^n \). If \( \dot{v}(t_k^-) \in V(q(t_k)) \), then \( \dot{v}(t_k^+) = \dot{v}(t_k^-) \). This is a straightforward consequence of the variational inequality (56). Indeed, let \( y \in \text{int} \, V(q(t_k)) \) then \( y \in V(q(t)) \) for all \( \tau \in [t_k - \delta, t_k + \delta] \) and some \( \delta > 0 \). Applying the inequality (56) with \( s = t_k - \delta, t = t_k + \delta \), and letting \( \delta \to 0 \), we get

\[
\langle M(q(t_k)) (\dot{v}(t_k^+) - \dot{v}(t_k^-)), y \rangle - \frac{1}{2} \left( |\dot{v}(t_k^+)|_{M(q(t_k))} - |\dot{v}(t_k^-)|_{M(q(t_k))} \right) \geq 0.
\]
By density, the same inequality holds for all \( y \in \text{int} V(q(t_k)) \). Picking \( y = \hat{v}(t_k^-) \) gives

\[
|\hat{v}(t_k^+) - \hat{v}(t_k^-)|^2 \leq 0 \quad \Rightarrow \quad \hat{v}(t_k^+) = \hat{v}(t_k^-).
\]

Next, consider the case where \( \hat{v}(t_k^-) \not\in V(q(t_k)) \). We use the shorthand notation \( \hat{v}^+ := \hat{v}(t_k^+) \) and \( \hat{v}^- := \hat{v}(t_k^-) \), and show that the following impact law holds:

\[
\hat{v}^+ = -e \hat{v}^- + (1 + e) \text{proj}_{M(q(t_k))}(\hat{v}^-, V(q(t_k))).
\]

Define \( \bar{u}_k := -e \hat{v}(t_k^-) + (1 + e) \text{proj}_{M(q(t_k))}(\hat{v}(t_k^-), V(q(t_k))) \). Consider the partition \( \mathcal{P} \) that contains the node \( t_{\mathcal{P},d} = t_k \) for some \( d \in \mathbb{R} \). By definition,

\[
|\hat{v}_{\mathcal{P}}(t_k) - \bar{u}| = e|\hat{v}_{\mathcal{P},d-1} - \hat{v}^-| + (1 + e)\left|\text{proj}_{M_{\mathcal{P},d}}\left[\hat{v}_{\mathcal{P},d-1} - \frac{1}{1+e} M_{\mathcal{P},d}^{-1} G_{\mathcal{P},d} \hat{v}(t_k)\right] - \text{proj}_{M_{\mathcal{P},d}}[\hat{v}^-; V(q(t_k))]\right|
\leq e|\hat{v}_{\mathcal{P},d-1} - \hat{v}^-| + (1 + e)\sqrt{\lambda_M} \left|\hat{v}_{\mathcal{P},d-1} - \frac{1}{1+e} M_{\mathcal{P},d}^{-1} G_{\mathcal{P},d} \hat{v}^-\right|
\leq \left(1 + e\right)\sqrt{\lambda_M + e}|\hat{v}_{\mathcal{P},d-1} - \hat{v}^-| + \frac{\sqrt{\lambda_M}}{\Delta_M} |G_{\mathcal{P},d}|. \tag{64}
\]

The pointwise convergence (25) implies the existence of some filter \( \mathcal{F} \) such that, for all \( \mathcal{P} \in \mathcal{F} \), we have \( t', t_k \in \mathcal{P} \) where \( t' \) is such that \( |\hat{v}(t') - \hat{v}(t_k^-)| < \frac{\varepsilon}{3} \), \( |\hat{v}_{\mathcal{P}}(t') - \hat{v}(t')| < \frac{\varepsilon}{3} \) and \( t_k - t' < \frac{\varepsilon}{3C_{\text{var}}} \); then

\[
|\hat{v}_{\mathcal{P},d-1} - \hat{v}(t_k^-)| \leq |\hat{v}_{\mathcal{P},d-1} - \hat{v}(t')| + |\hat{v}_{\mathcal{P}}(t') - \hat{v}(t')| + |\hat{v}(t') - \hat{v}(t_k^-)|
\leq \text{Var}(\hat{v}_{\mathcal{P}}; [t', t_k]) + \frac{2}{3} \varepsilon \leq C_{\text{var}}(t_k - t') + \frac{2}{3} \varepsilon < \varepsilon. \tag{65}
\]

Substituting (65) in (64), and taking the limit, we obtain

\[
|\hat{v}(t_k) - \bar{u}_k| < \varepsilon
\]

for every \( \varepsilon > 0 \), whence the desired result follows.

**IX. CONCLUSIONS**

The problem of designing asymptotically convergent state estimators for nonsmooth mechanical systems with frictionless unilateral constraints and impacts was considered in this paper. The formalism of differential inclusions was used for both the system and observer dynamics. For the latter, we proved the existence and uniqueness of solutions. The error analysis (for the convergence
of velocity estimate) was based on generalizing the Lyapunov techniques to functions of locally bounded variation, which also allow for accumulations of impacts (Zeno phenomenon).

Several directions of research could stem from the current work. As a first extension, we would like to generalize our result for more complex domains which are not necessarily characterized by the sublevel sets of continuously differentiable functions. Afterwards, we would also like to study the performance of the proposed state estimate in designing controllers that only use the information of the output and not the full state.

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APPENDIX

**Lemma A.1** (Moreau’s Two-Cone Lemma). If $V$ and $V^o$ denote a pair of mutually polar closed convex cones of a Euclidean linear space $\mathbb{R}^n$, then the following statements are equivalent for $x, y, z \in \mathbb{R}^n$,

- $x = \text{proj}(z; V)$ and $y = \text{proj}(z, V^o)$
- $z = x + y$, $x \in V$, $y \in V^o$, and $\langle x, y \rangle = 0$.

**Theorem A.2** (Generalization of Helly’s first theorem [22, Theorem 0.2.2]). Let $(u_\alpha)$ be a generalized sequence or net of functions of bounded variation from the interval $[0, T]$ to a Hilbert space $H$. Assume that the norm and the variation of $u_\alpha$ are uniformly bounded, that is, there exist $C_{\text{max}}$ and $C_{\text{var}}$ such that

$$\|u_\alpha\| \leq C_{\text{max}} \quad \text{and} \quad \text{Var}(u_\alpha; [0, T]) \leq C_{\text{var}}$$

then there is a filter $\mathcal{F}$ finer than the filter of the sections of the index set (that is, there exists a subnet extracted from the given net) and there exists a function of bounded variation $u: [0, T] \to H$ that satisfies

$$\text{weak-lim}_\mathcal{F} u_\alpha = u \quad \text{and} \quad \text{Var}(u; [0, T]) \leq C_{\text{var}}.$$

In the foregoing result if the Hilbert space $H$ into consideration is finite dimensional then the convergence is uniform, that is,

$$\lim_{\mathcal{F}} \|u_\alpha - u\| = 0.$$
REFERENCES


