The Curious Case of Integrator Reach Sets, Part I: Basic Theory

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Abstract—This is the first of a two part paper investigating the geometry of the integrator reach sets, and the applications thereof. In this Part I, assuming box-valued input uncertainties, we establish that this compact convex reach set is semialgebraic, translated zonoid, and not a spectrahedron. We derive the parametric as well as the implicit representation of the boundary of this reach set. We also deduce the closed-form formula for the volume and diameter of this set, and discuss their scaling with state dimension and time. We point out that these results may be utilized in benchmarking the performance of the reach set overapproximation algorithms.

Index Terms—Convex geometry, integrator, reach set, set-valued uncertainty.

I. INTRODUCTION

INTEGRATORS with bounded controls are ubiquitous in systems-control. They appear as Brunovsky normal forms for the feedback linearizable nonlinear systems. They also appear frequently as benchmark problems to demonstrate the performance of the reach set computation algorithms. Despite their prominence, specific results on the geometry of the integrator reach sets are not available in the systems-control literature. Broadly speaking, the existing results come in two flavors. On the one hand, very generic statements are known, e.g., these reach sets are compact convex sets wherever the set of initial conditions is compact convex, and the controls take values from a compact (not necessarily convex) set [1]. On the other hand, several numerical toolboxes [2], [3] are available for tight outer approximation of the reach sets over computationally benign geometric families, such as ellipsoids and zonotopes. The lack of concrete geometric results imply the absence of ground truth when comparing the efficacy of different algorithms, and one has to content with graphical or statistical (e.g., Monte Carlo) comparisons.

Building on the preliminary results in [4], this article undertakes a systematic study of the integrator reach sets. In particular, we answer the following basic questions.

Q1. What kind of compact convex sets are these (Section IV)?
Q2. How big are these sets (Section V)?
Q3. How these results on the geometry of integrator reach sets can be applied in practice (Section VI)?

We consider the integrator dynamics having states and inputs with relative degree vector \( r = (r_1, r_2, \ldots, r_m) \in \mathbb{Z}_+^m \) (vector of positive integers). In other words, we consider a Brunovsky normal form with \( m \) integrators where the \( j \)th integrator has degree \( r_j \) for \( j \in \{1, \ldots, m\} \). The dynamics is

\[
\dot{x} = Ax + Bu, \quad x \in \mathbb{R}^d, \quad u(\cdot) \in \mathcal{U} \subset \mathbb{R}^m
\]

where \( r_1 + r_2 + \cdots + r_m = d \), the set \( \mathcal{U} \) is compact, and

\[
A := \text{blkdiag}(A_1, \ldots, A_m), \quad B := \text{blkdiag}(b_1, \ldots, b_m) \quad \text{(2a)}
\]

\[
A_j := \begin{pmatrix} 0_{r_j \times 1} & e_1^{r_j} & e_2^{r_j} & \cdots & e_{r_j-1}^{r_j} \end{pmatrix}, \quad b_j := e_r^{r_j}, \quad \text{(2b)}
\]

In (2a), the symbol \( \text{blkdiag}() \) denotes a block diagonal matrix whose arguments constitute its diagonal blocks. In (2b), the notation \( 0_{r_j \times 1} \) stands for the \( r_j \times 1 \) column vector of zeros, and \( e_k^{r_j} \) denotes the \( k \)th basis (column) vector in \( \mathbb{R}^{r_j} \) for \( k \leq r_j \). The symbol \( (\ldots) \) denotes horizontal concatenation.

Notice that an integrator that replaces the ones appearing in the system matrices with arbitrary nonzero reals, is always reducible to the normal form (1) and (2) by renaming the variables. For instance, for any \( a, b \in \mathbb{R} \setminus \{0\} \), the system \( \dot{x}_1 = ax_2, \dot{x}_2 = bx_1 \), is equivalent to \( \dot{x}_1 = \ddot{x}_2, \dot{x}_2 = \ddot{u}_1 \), where \( \ddot{x}_2 := ax_2, \ddot{u}_1 := abu_1 \). Let \( \mathcal{R}(x_0, t) \) denote the forward reach set of (1) at time \( t > 0 \), starting from a given compact convex set of initial conditions \( x_0 \subset \mathbb{R}^d \), i.e.,

\[
\mathcal{R}(x_0, t) := \bigcup_{\text{measurable } u(\cdot) \in \mathcal{U} \subset \mathbb{R}^m} \{ x(t) \in \mathbb{R}^d \mid (1) \text{ and (2) hold,} \}
\]

\[
x(t = 0) \in x_0 \text{ compact convex, } \mathcal{U} \text{ compact} \}.
\]

In words, \( \mathcal{R}(x_0, t) \) is the set of all states that the controlled dynamics (1) and (2) can reach at time \( t > 0 \), starting from the set \( x_0 \) at \( t = 0 \), with measurable control \( u(\cdot) \in \mathcal{U} \). Compact Formally

\[
\mathcal{R}(x_0, t) = \exp(tA)x_0 + \int_0^t \exp((t - \tau)A)Bu \, d\tau
\]

\[
= \exp(tA)x_0 + \int_0^t \exp(sA)Bu \, ds
\]

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where $+$ denotes the Minkowski sum. The set-valued integral [5] in (4) is defined for any point-to-set function $F(\cdot)$, as

$$\int_0^t F(s)ds := \lim_{\Delta \to 0} \sum_{i=0}^{\lfloor t/\Delta \rfloor} \Delta F(i\Delta)$$  \hspace{1cm} (5)

where the summation symbol $\Sigma$ denotes the Minkowski sum, and $\lfloor \cdot \rfloor$ is the floor operator; see, e.g., [1]. Our objective is to study the geometry of (4) in detail.

This article significantly expands our preliminary works [4], [6], here we consider multi-input integrators as opposed to the single-input case considered in [4]. Even for the single-input case, while Haddad and Halder [4, Th. 1] derived an exact formula for the volume of the reach set, that formula involved limit and nested sums, and in that sense, was not really a closed-form formula for the volume of the reach set, that formula involved limit and nested sums, and in that sense, was not really a closed-form formula for numerical computation. In this article, we derive closed-form formula for the general multi-input case when the input uncertainty is box-valued, i.e., $\mathcal{U}$ is hyperrectangle. In the same setting, this article addresses previously unexplored directions, the scaling laws for the volume and diameter of integrator reach sets, exact parametric and implicit equations for the boundary, and the classification of these sets.

The rest of this article is organized as follows. After reviewing some preliminary concepts in Section II, we consider the integrator reach set resulting from box-valued input set uncertainty in Section III. The results on taxonomy and the boundary of the corresponding reach set are provided in Section IV. The results on the size of this set are collected in Section V. The application of these results for benchmarking the reach set overapproximation algorithms are discussed in Section VI. Section VII concludes this article and outlines the directions pursued in its Part II. All proofs are deferred to the Appendix.

II. PRELIMINARIES

A. State Transition Matrix

For $0 \leq s < t$, the state transition matrix $\Phi(t, s)$ associated with (2) is

$$\Phi(t, s) = \exp(A(t-s)) = \text{blkdiag}(\exp(A_1(t-s)), \ldots, \exp(A_m(t-s)))$$

with each diagonal block is upper triangular. Specifically, the $j$th diagonal block of size $r_j \times r_j$ is written element-wise as

$$\exp(A_j(t-s)) = \begin{cases} (t-s)^{\ell-k} / (\ell-k)! & \text{for } k \leq \ell \\ 0 & \text{otherwise} \end{cases}$$  \hspace{1cm} (6)

where $k$ is the row index, $\ell$ is the column index, and $k, \ell \in [r_j]$ for each $j \in [m]$. The diagonal entries in (6) are unity.

1) Support Function: The support function $h_K(\cdot)$ of a compact convex set $K \subset \mathbb{R}^d$, is given by

$$h_K(y) := \sup_{x \in K} \langle y, x \rangle, \quad y \in \mathbb{R}^d$$  \hspace{1cm} (7)

where $\langle \cdot, \cdot \rangle$ denotes the standard Euclidean inner product. Geometrically, $h_K(y)$ gives the signed distance of the supporting hyperplane of $K$ with outer normal vector $y$, measured from the origin. Furthermore, the supporting hyperplane at $x^{bdy} \in \partial K$ is

$$\langle y, x^{bdy} \rangle = h_K(y),$$

and we can write

$$K = \{ x \in \mathbb{R}^d | \langle y, x \rangle \leq h_K(y) \text{ for all } y \in \mathbb{R}^d \}.$$

For compact $K_1, K_2 \subset \mathbb{R}^d$,

$$K_1 \subseteq K_2 \quad \text{if and only if} \quad h_{K_1}(\cdot) \leq h_{K_2}(\cdot).$$  \hspace{1cm} (8)

The support function $h_K(y)$ is convex in $y$. For more details on the support function, we refer the readers to [8, Ch. VI].

The support function $h_K(y)$ uniquely determines the set $K$. Given matrix-vector pair $(\Gamma, \gamma) \in \mathbb{R}^{d \times d} \times \mathbb{R}^d$, the support function of the affine transform $\Gamma K + \gamma$ is

$$h_{\Gamma K + \gamma}(y) = \langle y, \gamma \rangle + h_K(\Gamma^* y).$$  \hspace{1cm} (9)

Given a function $f: \mathbb{R}^d \mapsto \mathbb{R} \cup \{+\infty\}$, its Legendre–Fenchel conjugate is

$$f^*(y) := \sup_{x \in \text{domain}(f)} \{ \langle y, x \rangle - f(x) \}, \quad y \in \mathbb{R}^d.$$  \hspace{1cm} (10)

From (7) to (10), it follows that $h_K(y)$ is the Legendre–Fenchel conjugate of the indicator function

$$1_K(x) := \begin{cases} 0 & \text{if } x \in K \\ +\infty & \text{otherwise} \end{cases}.$$  \hspace{1cm} (11)

Since the indicator function of a convex set is a convex function, the biconjugate $1_K^*(\cdot) = h_K(\cdot) = 1_K(\cdot)$. This will be useful in Section IV.

To proceed further, we introduce some notations. Since $\mathcal{U}$ is compact, let

$$\alpha_j := \min_{u \in \mathcal{U}} u_j, \quad \beta_j := \max_{u \in \mathcal{U}} u_j, \quad j \in [m]$$  \hspace{1cm} (12)

that is, $\alpha_j$ and $\beta_j$ are the component-wise minimum and maximum, respectively, of the input vector. Furthermore, let

$$\mu_j := \frac{\beta_j - \alpha_j}{2}, \quad \nu_j := \frac{\beta_j + \alpha_j}{2}$$

and introduce

$$\xi(s) := \begin{pmatrix} \mu_1 \xi_1(s) \\ \vdots \\ \mu_m \xi_m(s) \end{pmatrix}, \quad \zeta_j(s) := \begin{pmatrix} s^{r_j-1} / (r_j - 1)! \\ \vdots \\ s^{r_j-2} / (r_j - 2)! \end{pmatrix}$$  \hspace{1cm} (13)

for $j \in [m]$. Also, let

$$\zeta(t_0, t) := \begin{pmatrix} \mu_1 \zeta_1(t_0, t) \\ \mu_2 \zeta_2(t_0, t) \\ \vdots \\ \mu_m \zeta_m(t_0, t) \end{pmatrix}, \quad \zeta_j(t_0, t) := \int_{t_0}^t \xi_j(s) ds \in \mathbb{R}^{r_j}$$  \hspace{1cm} (14)

for $j \in [m]$. When $t_0 = 0$, we simplify the notations as

$$\zeta(t) := \zeta(0, t), \quad \zeta_j(t) := \zeta_j(0, t) \text{ for all } j \in [m].$$  \hspace{1cm} (15)

Using (13) and following (7), we deduce Proposition 1 stated next (proof in Appendix A).

Proposition 1 (Support function for compact $\mathcal{U}$): For compact convex $\mathcal{X}_0 \subset \mathbb{R}^d$, and compact $\mathcal{U} \subset \mathbb{R}^m$, the support function
of the reach set (4) is
\[
    h_{R(x_0,t)}(y) = \sup_{x \in A} \sum_{j=1}^{m} \langle y_j, \exp(tA_j)x_{j0} \rangle + \int_0^t \sup_{u \in \text{closure}(\text{conv}(U)))} \sum_{j=1}^{m} \{ \langle y_j, \xi_j(s) \rangle u_j \} \, ds
\]
(16)
where \( \text{conv}(\cdot) \) denotes the convex hull.

2) Polar Dual: The polar dual \( K^\circ \) of any nonempty set \( K \subset \mathbb{R}^d \) is given by
\[
    K^\circ := \{ y \in \mathbb{R}^d \mid \langle y, x \rangle \leq 1 \text{ for all } x \in K \}.
\]
(17)
From this definition, it is immediate that \( K^\circ \) contains the origin, and is a closed convex set. The bipolar \( (K^\circ)^\circ = \text{closure}(\text{conv}(K \cup \{0\})) \). Thus, if \( K \) is compact convex and contains the origin, then we have the involution \( K^\circ = \mathcal{K} \). From (7) and (17), notice that \( K^\circ \) is the unit support function ball, i.e., \( K^\circ = \{ y \in \mathbb{R}^d \mid h_K(y) \leq 1 \} \). In Section IV-D, we will mention some properties of the polar dual of the reachable integrator reach set.

3) Vector Measure: Let \( F \) be a \( \sigma \)-field of the subsets of a set. A countably additive mapping \( \bar{\mu} : F \rightarrow \mathbb{R}^d \) is termed a vector measure. Here, "countably additive" means that for any sequence \( \{ \Omega_i \}_{i=1}^\infty \) of disjoint sets in \( F \) such that their union is in \( F \), we have \( \bar{\mu}(\bigcup_{i=1}^\infty \Omega_i) = \sum_{i=1}^\infty \bar{\mu}(\Omega_i) < \infty \). Some of the early investigations of vector measures were due to Liapounoff [9] and Halmos [10]; relatively recent references are [11] and [12].

4) Zonotope: A zonotope \( Z \subset \mathbb{R}^d \) is a finite Minkowski sum of closed line segments or intervals \( \{I_j\}_{j=1}^m \) where these intervals are imbedded in the ambient Euclidean space \( \mathbb{R}^d \).

Explicitly, for some positive integer \( n \), we write
\[
    Z := I_1 + \cdots + I_n
    = \left\{ x \in \mathbb{R}^d \mid x = \sum_{i=1}^n x_i, \ x_i \in I_i, \ i \in [m] \right\}.
\]
Thus, a zonotope is the range of an atomic vector measure. Alternatively, a zonotope can be viewed as the affine image of the unit cube. A compact convex polytope is a zonotope if and only if all its two dimensional faces are centrally symmetric [13, p. 182]. For instance, the cross polytope \( \{ x \in \mathbb{R}^d \mid \|x\|_1 \leq 1 \} \), is not a zonotope. Standard references on zonotope include [14], [15], [16, Ch. 2.7].

The set of zonotopes is closed under affine image and Minkowski sum, but not under intersection. In the systems-control literature, a significant body of work exists on computationally efficient overapproximation of reach sets via zonotopes [17], [18], and [19], and their variants, such as zonotope bundles [20], constrained zonotopes [21], complex zonotopes [22], and polynomial zonotopes [23], [24].

5) Variety and Ideal: Let \( p_1, \ldots, p_n \in \mathbb{R}[x_1, \ldots, x_d] \), the vector space of real-valued \( d \)-variate polynomials. The (affine) variety \( V_{R[x_1, \ldots, x_d]}(p_1, \ldots, p_n) \) is the set of all solutions of the system \( p_1(x_1, x_2, \ldots, x_d) = \cdots = p_n(x_1, x_2, \ldots, x_d) = 0 \). Given \( p_1, \ldots, p_n \in \mathbb{R}[x_1, \ldots, x_d] \), the set
\[
    I := \left\{ \sum_{i=1}^n \alpha_i p_i \mid \alpha_1, \ldots, \alpha_n \in \mathbb{R}[x_1, \ldots, x_d] \right\}
\]
is called the ideal generated by \( p_1, \ldots, p_n \). We write this symbolically as \( I = \langle p_1, \ldots, p_n \rangle \). Roughly speaking, \( \langle p_1, \ldots, p_n \rangle \) is the set of all polynomial consequences of the given system of \( n \) polynomial equations in \( d \) indeterminates. We refer the readers to [25, Ch. 1] for detailed exposition of these concepts.

III. BOX-VALUED INPUT UNCERTAINTY

In the rest of this article, we characterize the exact reach set (3) when input set \( U \subset \mathbb{R}^m \) is box-valued, and remark on the quality of approximation for the same when \( U \) is arbitrary compact.

When \( U \subset \mathbb{R}^m \) is box-valued, denote the reach set (3) as \( \mathcal{R}^\square \), i.e., with a box superscript.\(^1\) In this case, each of the \( m \) single input integrator dynamics with \( r_j \) dimensional state subvectors for \( j \in [m] \), are decoupled from each other. Then, \( \mathcal{R}^\square(X_0, t) \subset \mathbb{R}^d \) is the Cartesian product of these single input integral reach sets: \( \mathcal{R}_j(X_0, t) \subset \mathbb{R}^{r_j} \) for \( j \in [m] \), i.e.,
\[
    \mathcal{R}^\square = \mathcal{R}_1 \times \mathcal{R}_2 \times \cdots \times \mathcal{R}_m.
\]
In what follows, we will sometimes exploit that (18) may also be written as\(^2\) a Minkowski sum \( \mathcal{R}_1 + \cdots + \mathcal{R}_m \). Notice that the decoupled dynamics also allows us to write a Minkowski sum decomposition for the set of initial conditions
\[
    X_0 = X_{10} + \cdots + X_{m0}
\]
and accordingly, the initial condition subvectors \( x_{j0} \in X_{j0} \subset \mathbb{R}^{r_j} \) for \( j \in [m] \).

Since the support function of the Minkowski sum is equal to the sum of the support functions, we have
\[
    h_{\mathcal{R}^\square(X_0,t)}(y) = \sum_{j=1}^{m} h_{\mathcal{R}_j(X_0,t)}(y_j).
\]
(19)
This leads to the following result (proof in Appendix B), which will come in handy in the ensuing development.\(^3\)

Theorem I (Support function for box-valued \( U \)): For compact convex \( X_0 \subset \mathbb{R}^d \), and box-valued input uncertainty set given by
\[
    U := [\alpha_1, \beta_1] \times [\alpha_2, \beta_2] \times \cdots \times [\alpha_m, \beta_m] \subset \mathbb{R}^m
\]
the support function of the reach set (4) is
\[
    h_{\mathcal{R}(X_0,t)}(y) = \sum_{j=1}^{m} \left\{ \sup_{x_{j0} \in X_{j0}} \langle y_j, \exp(tA_j)x_{j0} \rangle + \nu_j \right\} \mu_j \int_0^t |\langle y_j, \xi_j(s) \rangle| \, ds.
\]
(20)

The formula (21) upper bounds (16) resulting from the same initial condition and arbitrary compact \( U \subset \mathbb{R}^m \) with \( [\alpha_j, \beta_j]_{j=1}^m \) related to \( U \) via (11). Thus, from (8), the reach set \( \mathcal{R}^\square \) with box-valued input uncertainty will overapproximate the reach set \( \mathcal{R} \) associated with arbitrary compact \( U \), at any given \( t > 0 \), provided \( [\alpha_j, \beta_j]_{j=1}^m \) are defined as (11).

When \( U \) is compact but not box-valued, then we can quantify the quality of the aforesaid overapproximation in terms of the

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\(^1\)For the single input (\( m = 1 \)) case, we drop the box superscript.

\(^2\)In general, the Minkowski sum of a given collection of compact convex sets is not equal to their Cartesian product. However, the "factor sets" in (18) belong to disjoint mutually orthogonal \( r_j \) dimensional subspaces, \( j = 1, \ldots, m \), which allows writing this Cartesian product as a Minkowski sum.

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two-sided Hausdorff distance metric \( \text{dist} (\mathcal{R}, \mathcal{R}') \) between the convex compact sets \( \mathcal{R}, \mathcal{R}' \subset \mathbb{R}^d \), expressible [13, Th. 1.8.11] in terms of their support functions \( h_{\mathcal{R}} (\cdot), h_{\mathcal{R}'} (\cdot) \) as

\[
\text{dist} (\mathcal{R}, \mathcal{R}') = \sup_{\|y\|_2 = 1} \left| h_{\mathcal{R}} (y) - h_{\mathcal{R}'} (y) \right|.
\]  

(22)

Thanks to (8), the absolute value in (22) can be dispensed since \( \mathcal{R} \subseteq \mathcal{R}' \) with set equality if \( \mathcal{U} \) is box, in which case \( h_{\mathcal{R}} (\cdot) = h_{\mathcal{R}'} (\cdot) \) and dist = 0.

It is known that [26, Prop. 6.1] the set \( \mathcal{R}(X_0, t) \) resulting from a linear time invariant dynamics such as (1) and (2) remains invariant under the closure of convexification of the input set \( \mathcal{U} \). Therefore, it is possible that \( \mathcal{R} = \mathcal{R}' \) and dist = 0 even when the compact set \( \mathcal{U} \) is nonconvex. For instance, the reach set \( \mathcal{R}(X_0, t) \) resulting from some compact convex \( X_0 \subset \mathbb{R}^d \) and dynamics (1) and (2) with the nonconvex input uncertainty set \( \{-1, 1\}^m \), is identical to \( \mathcal{R}'(X_0, t) \) resulting from the same \( X_0 \), same dynamics, and the box-valued uncertainty input set (20) with \( \alpha_j = -1, \beta_j = 1 \) for all \( j \in [m] \).

Likewise, for the same compact convex \( X_0 \subset \mathbb{R}^d \), the reach set \( \mathcal{R}(X_0, t) \) resulting from (1) to (2) with the nonconvex input set \( \{u \in \mathbb{R}^m \mid \|u\|_1 \leq 1\} \), \( 0 < p \leq 1 \), is the same as that resulting from the cross-polytope \( \{u \in \mathbb{R}^m \mid \|u\|_1 \leq 1\} \). More generally, for \( 0 < p < \infty \), suppose \( \mathcal{R}_p (X_0, t) \) results from the unit \( p \) norm ball input uncertainty set \( \{u \in \mathbb{R}^m \mid \|u\|_p \leq 1\} \). Let \( M^1 (\tau) := \|\exp (\tau A) B = \text{blkdiag} (\xi_1, \ldots, \xi_m) \| \). If \( \mathcal{R}'(X_0, t) \) results from the same \( X_0 \), same dynamics, and input uncertainty set (20) with \( \alpha_j = -1, \beta_j = 1 \) for all \( j \in [m] \), then using [27, Th. 1], (22) simplifies to

\[
\text{dist} (\mathcal{R}, \mathcal{R}') = \sup_{\|y\|_2 = 1} \int_0^t \|M (\tau) y_1 - \|M (\tau) y_2\|_q d\tau
\]  

(23)

where \( q \) is the Hölder conjugate of \( \max \{1, p\} \), i.e., \( \frac{1}{q} = \frac{1}{p} + \frac{1}{\frac{1}{2} + 1} \), and \( 0 < p < \infty \). The objective in (23) being positive homogeneous, admits lossless constraint convexification \( \|y\|_2 \leq 1 \), and the corresponding maximal value\(^3\) for moderate dimensions \( d \), can be found by direct numerical search.

IV. TAXONOMY AND BOUNDARY

For \( X_0 \subset \mathbb{R}^d \) compact convex, it is well-known [1, Sec. 2] that the reach set \( \mathcal{R} \) given by (3) is compact convex for all \( t > 0 \) provided \( \mathcal{U} \) is compact. However, it is not immediate what kind of convex set \( \mathcal{R} \) is, even for singleton \( X_0 = \{x_0\} \).

In this section, we examine the question "what type of compact convex set \( \mathcal{R} (\{x_0\}, t) \) is" when \( \mathcal{U} \) is box-valued uncertainty set of the form (20). In the same setting, we also derive the equations for the boundary \( \partial \mathcal{R} (\{x_0\}, t) \).

Notice that for nonsingleton \( X_0 \), the taxonomy question is not well-posed since the classification then will depend on \( X_0 \). Also, setting \( X_0 = \{x_0\} \) in (4), it is apparent that \( \mathcal{R} (\{x_0\}, t) \) is a translation of the set-valued integral in (4). Thus, classifying \( \mathcal{R} (\{x_0\}, t) \) amounts to classifying the second summand in (4).

\(^3\) As such, (23) has a difference of convex objective, and by the Weierstrass extreme value theorem, the maximum is achieved.

A. \( \mathcal{R} (\{x_0\}, t) \) is a Zonoid

A zonoid is a compact convex set that is defined as the range of an atom free vector measure (see Section II-A3). Affine image of a zonoid is a zonoid. Minkowski sum of zonoids is also a zonoid. We refer the readers to [28, 29, 30, 31, Sec. 1] for more details on the properties of a zonoid. By slight abuse of nomenclature, in this article, we use the term zonoid up to translation, i.e., we refer to the translation of zonoids as zonoids (instead of using another term such as “zonoidal translates”).

Let us mention a few examples. Any compact convex symmetric set in \( \mathbb{R}^2 \) is a zonoid. In dimensions three or more, all \( \ell_p \) norm balls for \( p \geq 2 \) are zonoids.

An alternative way to think about the zonoid is to view it as the limiting set (convergence with respect to the two-sided Hausdorff distance, see, e.g., [4, Appendix B]) of the Minkowski sum of line segments, i.e., the limit of a sequence of zonotopes [14, 15, 28]. Formally, given a Hausdorff convergent sequence of zonotopes \( \{Z_n\} \), the zonoid \( Z_{\infty} \) is

\[
Z_{\infty} := \lim_{n \to \infty} Z_n, \text{ where } Z_n := \sum_{i=1}^{n(j)} [a_{ij}, b_{ij}], \quad a_{ij}, b_{ij} \in \mathbb{R}^d
\]  

for some \( a_{ij} \leq b_{ij} \) (element-wise vector inequality), and a suitable mapping \( n : \mathbb{Z}_+ \to \mathbb{Z}_+ \). Our analysis will make use of this viewpoint in Section V-A. Our main result in this section is the following.

Theorem 2: The reach set \( \mathcal{R} (\{x_0\}) \) given by (3) with \( X_0 = \{x_0\} \) and \( \mathcal{U} \) given by (20), is a zonoid.

To appreciate Theorem 2 via the limiting viewpoint mentioned before, let us write

\[
\mathcal{R} (\{x_0\}, t) = \exp (tA) x_0 + \sum_{j=1}^{m} \nu_j \xi_j (t)
\]  

(24)

where all summation symbols denote Minkowski sums. The first term in (24) denotes a translation. In the second term, the outer summation over index \( j \) arises by writing the Cartesian product (18) as the Minkowski sum \( \mathcal{R}_1 + \ldots + \mathcal{R}_m \). Furthermore, uniformly discretizing \([0, t]\) into \( n \) subintervals \([i - 1]t/n, it/n\), \( i = 1, \ldots, n \), we write \( \int_0^t \exp (sA) b_j [-\mu_j, \mu_j] ds \) as the limit of the Minkowski sum over index \( i \). Geometrically, the innermost summands in the second term denote nonuniformly rotated and scaled line intervals in \( \mathbb{R}^d \). In other words, the second term in (24) is a Minkowski sum of \( m \) sets, each of these sets being the limit of a sequence of sets \( \{Z_n\} \) comprising of zonotopes

\[
Z_n := \sum_{i=0}^{n} t \mu_j \xi_j (t_i) [-1, 1]
\]  

(24)

where \( \lim_{n \to \infty} Z_n \) is a zonoid, the second term in (24) is a Minkowski sum of \( m \) zonoids, and is, therefore, a zonoid [28, Th. 1.5].
entire right hand side of (24), then, is translation of a zonoid, and hence a zonoid.

Remark 1: If \( A_0 \subset \mathbb{R}^d \) is not singleton, but instead a zonoid, then \( \mathcal{R}(A_0, t) \) is still a (translated) zonoid. To see this, notice from (4) and (21) that

\[
\mathcal{R}(A_0, t) = \exp(tA)x_0 + \mathcal{R}((0), t)
\]

and that \( \exp(tA)x_0 \), being linear image of a zonoid, is a zonoid [28, Lemma 1.4]. Thus, (25) being Minkowski sum of zonoids, is a zonoid too [28, Th. 1.5], up to translation.

In the following, we derive formulae for the boundary (Proposition 2 and Section IV-C) and volume (Theorem 5) of the integrator reach set with \( A_0 = \{ x_0 \} \) (singleton set). From (25), it is clear that one cannot expect similar closed-form formulae for arbitrary compact (or even arbitrary compact convex) \( A_0 \). In this sense, our closed-form formulae are as general as one might hope for. For a specific nonsingleton \( A_0 \), one can use these formulae to first derive the boundary (resp. volume) of \( \mathcal{R}(A_0, t) \), and then use (25) to get numerical estimates for the boundary (resp. volume) of \( \mathcal{R}(A_0, t) \) (cf. Remark 2).

**B. **\( \mathcal{R}(\{ x_0 \}, t) \) is Semialgebraic

A set in \( \mathbb{R}^d \) is called basic semialgebraic if it can be written as a finite union of polynomial inequalities and equalities, the polynomials being in \( \mathbb{R}[x_1, \ldots, x_d] \). Finite union of basic semialgebraic sets is called a semialgebraic set. A semialgebraic set need not be basic semialgebraic; see, e.g., [32, Example 2.2].

Semialgebraic sets are closed under finitely many unions and intersections, complement, topological closure, polynomial mapping including projection [33], [34], and Cartesian product. For details on semialgebraic sets, we refer the readers to [35, Ch. 2]; see [36, Appendix A.4.4] for a short summary.

In Proposition 2 as follows, we derive a parametric representation of \( x^{bdy} \in \partial \mathcal{R}(\{ x_0 \}, t) \), the boundary of the reach set. Then, we use this representation to establish semialgebraicity of \( \mathcal{R}(\{ x_0 \}, t) \) in Theorem 4 that follows.

**Proposition 2:** For relative degree vector \( r = (r_1, \ldots, r_m)^\top \), and fixed \( x_0 \in \mathbb{R}^d \) comprising of subvectors \( x_j \in \mathbb{R}^{r_j} \) where \( j \in [m] \), consider the reach set (4) with singleton \( \{ x_0 \} \) and \( \mathcal{U} \) given by (20). For \( j \in [m] \), define \( \mu_1, \ldots, \mu_m \) and \( \nu_1, \ldots, \nu_m \) as in (11) and (12). Let the indicator function \( 1_{k \leq t} := 1 \) for \( k \leq t \), and := 0 otherwise. Then, the components of

\[
x^{bdy} = \begin{pmatrix} x_1^{bdy} \\ x_2^{bdy} \\ \vdots \\ x_m^{bdy} \end{pmatrix} \in \partial \mathcal{R}(\{ x_0 \}, t), \quad x_j^{bdy} \in \mathbb{R}^{r_j}, \quad j \in [m]
\]

admit parametric representation in terms of the parameters \( s_1, s_2, \ldots, s_{r_j-1} \) satisfying \( 0 \leq s_1 \leq s_2 \leq \ldots \leq s_{r_j-1} \leq t \). This parameterization is given by

\[
x^{bdy}_j(k) = \sum_{\ell=1}^{r_j} 1_{k \leq \ell} \frac{\ell^{r_j-k}}{(r_j-k)!} x_j(\ell) + \frac{\nu_j t^{r_j-k+1}}{(r_j-k+1)!} x_j(0)
\]

\[\pm \frac{\mu_j}{(r_j-k+1)!} \left\{ (-1)^{r_j-1} \ell^{r_j-k+1} + 2 \sum_{q=1}^{r_j-1} (-1)^{q+1} s_q^{r_j-k+1} \right\}
\]

where \( x^{bdy}_j(k) \) denotes the \( k \)th component of the \( j \)th subvector of \( x^{bdy}_j \) for \( k \in [r_j] \).

The following is a consequence of the \( \pm \) appearing in (26).

**Corollary 3:** The single input integrator reach set \( \mathcal{R}_j(\{ x_0 \}, t) \subset \mathbb{R}^r \) has two bounding surfaces for each \( j \in [m] \). In other words, there exist \( p^{upper}_j, p^{lower}_j : \mathbb{R}^r \to \mathbb{R} \) such that

\[
\mathcal{R}_j(\{ x_0 \}, t) = \{ x \in \mathbb{R}^r \mid p^{upper}_j(x) \leq 0, p^{lower}_j(x) \leq 0 \}
\]

with boundary \( \partial \mathcal{R}_j(\{ x_0 \}, t) = \{ x \in \mathbb{R}^r \mid p^{upper}_j(x) = 0 \} \cup \{ x \in \mathbb{R}^r \mid p^{lower}_j(x) = 0 \} \).

During the proof of Theorem 4 as follows, it will turn out that in fact \( p^{upper}_j, p^{lower}_j \in \mathbb{R}[x_1, \ldots, x_j] \) for all \( j \in [m] \). In words, \( p^{upper}_j, p^{lower}_j \) are real algebraic hypersurfaces for all \( j \in [m] \).

Let us exemplify the parameterization (26) for the case \( r = (r_1, r_2)^\top = (2, 3)^\top \).

In (27), taking plus (resp. minus) signs in each of component gives the parametric representation of the curve \( p^{upper}_j = 0 \) (resp. \( p^{lower}_j = 0 \)). These curves are as in [4, Fig. 1(a)], and their union defines \( \partial \mathcal{R}_j \). We note that the parameterization (27) appeared in [37, p. 111].

Likewise, in (28), taking plus (resp. minus) signs in each of component gives the parametric representation of the surface \( p^{upper}_j(x) = 0 \) (resp. \( p^{lower}_j = 0 \)). The resulting set \( \mathcal{R}_j \) is the triple integrator reach set, and is shown in Fig. 1.

Now we come to the main result of this section.

**Theorem 4:** The reach set \( \mathcal{R}(\{ x_0 \}, t) \) given by (3) with \( A_0 = \{ x_0 \} \) and \( \mathcal{U} \) as in (20), is semialgebraic.

Let us illustrate the bounding curves and surfaces for (27) and (28), respectively, in the implicit form. Eliminating the parameter \( s_1 \) from (27) reveals that \( p^{upper}_1, p^{lower}_1 \) are parabolas. In particular

\[
p^{upper}_1(x^{bdy}_1(1), x^{bdy}_1(2)) = \frac{1}{4} \left( \frac{x^{bdy}_1(2) - x^{bdy}_1(2) - \nu_1 t}{\mu_1} + t \right)^2 - \frac{x^{bdy}_1(1) - x^{bdy}_1(1) - t x^{bdy}_1(2) - \nu_1 t^2}{\mu_1} - \frac{t^2}{2} \]

(29)
Thus, the notationally simplified version of (30) that suffices for implicitization, is

$$\rho_k = \sum_{q=1}^{d-1} (-1)^{q+1} s_q^{d-k+1}, \quad k = 1, \ldots, d$$

which is a system of $d$ homogeneous polynomials in variables $(s_1, s_2, \ldots, s_{d-1})$. The objective is to derive the implicitized polynomial $\varphi(\rho_1, \rho_2, \ldots, \rho_d)$ associated with (32).

When $d=2$, the parameterization (32) becomes

$$\rho_1 = s_1^2, \quad \rho_2 = s_1$$

and we get degree 2 implicitized polynomial

$$\varphi(\rho_1, \rho_2) = \rho_2^2 - \rho_1 = 0.$$  

For $k = 1, 2$, substituting for the $\rho_1, \rho_2$ in (33) from (31) with appropriate plus-minus signs recovers (29).

When $d = 3$, the parameterization (32) becomes

$$\rho_1 = s_1^3 - s_2, \quad \rho_2 = s_1^2 - s_2^2, \quad \rho_3 = s_1 - s_2$$

elementary algebra gives degree 4 implicitized polynomial

$$\varphi(\rho_1, \rho_2, \rho_3) = \rho_1^4 - 4\rho_3\rho_1 + 3\rho_2^2 = 0.$$  

As before, for $k = 1, 2, 3$, substituting for the $\rho_1, \rho_2, \rho_3$ in (35) from (31) with appropriate plus-minus signs recovers (30). However, for $d = 4$ or higher, it is practically impossible to derive the implicitization via brute force algebra.

A principled way to implicitize (32) is due to Zaimi [38], and starts with defining $\lambda_k := \rho_{d-k+1}$ for $k = 1, \ldots, d$. Introduce the sequence $A_k(s_1, s_2, \ldots, s_{d-1})$ via the generating function (see, e.g., [39, Ch. 1])

$$F(\tau) = \sum_{k \geq 0} A_k \tau^k = \frac{(1 - s_1 \tau)(1 - s_2 \tau)\cdots(1 - s_{d-1} \tau)}{(1 - s_2 \tau)\cdots(1 - s_{d-1} \tau)}.$$  

Taking the logarithmic derivative of (35), and then using the generating functions $(1 - s_q \tau)^{-1} = \sum_{k \geq 0} (s_q \tau)^k$ for all $q = 1, \ldots, d - 1$, yields

$$\frac{F'(\tau)}{F(\tau)} = -s_1 \sum_{k \geq 0} (s_1 \tau)^k + s_2 \sum_{k \geq 0} (s_2 \tau)^k - s_3 \sum_{k \geq 0} (s_3 \tau)^k + \cdots.$$  

Integrating (36) with respect to $\tau$, we obtain

$$F(\tau) = \exp \left( -\sum_{k=1}^{d} \frac{\lambda_k \tau^k}{k} \right).$$  

Equating (35) and (37) allows us to compute $A_k$ as a degree $k$ polynomial of the $\lambda$s.

On the other hand, since the generating function (35) is a rational function with denominator polynomial of degree $\delta := \left[ \frac{d}{2} \right]$, the following Hankel determinant vanishes:

$$\det[A_{d-2\delta+1+i}, i = 0, 1, \ldots, \delta] = 0.$$  

Substituting the $A_{k\delta}$ obtained as degree $k$ polynomials of the $\lambda$s into (38) gives an implicit polynomial in indeterminate $(\lambda_1, \ldots, \lambda_d)$ of degree $(\delta + 1)(d - \delta)$. Finally, reverting back the $\lambda$s to the $\rho$s result in the desired implicit polynomial $\varphi(\rho_1, \rho_2, \ldots, \rho_d)$, which is also of degree $(\delta + 1)(d - \delta)$.

This result goes back to Kronecker [40]. See also [41, p. 5, Lemma III].
Double integrator reach set \( \mathcal{R}(\{x_0\}, t) \) and its polar dual \( \mathcal{R}^o(\{x_0\}, t) \) at \( t = 2 \), \( \mu = 0.5 \). The curves \( u^{\text{upper}}, u^{\text{lower}} \) defining the reach set boundary (see Corollary 3 and the discussion thereafter) are shown too. (a) \( \mathcal{R} \) with \( x_0 = (0.05, 0.05)^T \). (a) \( \mathcal{R}^o \) with \( x_0 = (0.05, 0.05)^T \).

For instance, when \( d = 3 \), the relation (38) becomes

\[
\det \left( \begin{bmatrix} A_1 & A_2 \\ A_2 & A_3 \end{bmatrix} \right) = 0.
\]  

(39)

In this case, equating (35) and (37) gives

\[
A_1 = -\lambda_1, \quad A_2 = \frac{1}{2} \lambda_1^2 - \frac{1}{2} \lambda_2, \quad A_3 = -\frac{1}{6} \lambda_3^2 + \frac{1}{2} \lambda_1 \lambda_2 - \frac{1}{3} \lambda_3.
\]

Substituting these back in (39) yields the quartic polynomial \( \lambda_1^4 + 3 \lambda_2^2 - 4 \lambda_3 \lambda_1 = 0 \), which under the mapping \( (\lambda_1, \lambda_2, \lambda_3) \mapsto (\rho_3, \rho_2, \rho_1) \) recovers (35), and thus, (30).

In summary, (38) is the desired implicitization of the bounding hypersurfaces of the single input integrator reach set (up to the change of variables). The Cartesian product of these implicit hypersurfaces gives the implicitization in the multi input case.

**D. Dual of \( \mathcal{R}(\{x_0\}, t) \)**

From convex geometry standpoint, it is natural to ask what kind of characterization is possible for the polar dual (see Section II-A2) of the integrator reach set \( \mathcal{R} \) or \( \mathcal{R}^o \). We know in general that \( \mathcal{R}^o \) will be a closed convex set. Depending on the choice of \( x_0, \mu \) and \( t \), the set \( \mathcal{R}(\{x_0\}, t) \) may not contain the origin, and thus, the bipolar

\[
(\mathcal{R}(\{x_0\}, t))^\circ = \text{closure}(\text{conv}(\mathcal{R}(\{x_0\}, t) \cup \{0\}))
\]

that is, we do not have the involution in general.

Furthermore, since \( \mathcal{R}^o(\{x_0\}, t) \) is semialgebraic from Section IV-B, so must be its polar dual \( (\mathcal{R}^o(\{x_0\}, t))^\circ \); see, e.g., [36, Ch. 5, Sec. 5.2.2].

We also know from Section IV-A that \( \mathcal{R}^o(\{x_0\}, t) \) is a zonoid. However, the polar of a zonoid is not a zonoid in general [42], [43], and we should not expect \( (\mathcal{R}^o(\{x_0\}, t))^\circ \) to be one. Fig. 2 shows \( \mathcal{R}(\{x_0\}, t) \) and \( (\mathcal{R}(\{x_0\}, t))^\circ \) for the double integrator (\( d = 2, m = 1 \)).

**E. Summary of Taxonomy**

So far we explained that the compact convex set \( \mathcal{R}^o(\{x_0\}, t) \) is semialgebraic, and a translated zonoid. Two well-known subclasses of convex semialgebraic sets are the spectrahedra and the spectrahedral shadows. The spectrahedra, a.k.a. linear matrix inequality (LMI) representable sets are affine slices of the symmetric positive semidefinite cone. The spectrahedral shadows, a.k.a. lifted LMI or semidefinite representable sets are the projections of spectrahedra. The spectrahedral shadows subsume the class of spectrahedra; e.g., the set \( \{(x_1, x_2) \in \mathbb{R}^2 \mid x_1^2 + x_2^4 \leq 1\} \) is a spectrahedral shadow but not a spectrahedron. The polar duals of spectrahedra are spectrahedral shadows [36, Ch. 5, Sec. 5.5].

We note that \( \mathcal{R}^o \) is not a spectrahedron. To see this, we resort to the contrapositive of [44, Th. 3.1]. Specifically, the number of intersections made by a generic line passing through an interior point of the \( d \)-dimensional reach set \( \mathcal{R}^o \) with its real algebraic boundary is not equal to the degree of the bounding algebraic hypersurfaces, the latter we know from Section IV-C to be \( (\frac{d+1}{2}) + 1 \) \((d - \frac{d+1}{2}) \). In other words, the \( \mathcal{R}^o \) is not rigidly convex, see [44, Sec. 3.1 and 3.2]. Fig. 3 helps visualize this for \( m = 1 \). From Fig. 3(a), we observe that a generic line for \( d = 2 \) has 4 intersections with the bounding real algebraic curves whereas from (29), we know that \( u^{\text{upper}}, u^{\text{lower}} \) are degree 2 polynomials. Likewise, Fig. 3(b) reveals that a generic line for \( d = 3 \) has 6 intersections with the bounding real algebraic surfaces whereas from (30), we know that the polynomials \( u^{\text{upper}}, u^{\text{lower}} \) in this case, are of degree 4.

Could the reach set \( \mathcal{R}^o \) be spectrahedral shadow? Some calculations show that sufficient conditions as in [45] do not seem to hold. However, this remains far from conclusive. We summarize our taxonomy results in Fig. 4; the highlighted region shows where the integrator reach set belongs. To answer whether...

---

**Fig. 2.** Double integrator reach set \( \mathcal{R}(\{x_0\}, t) \) and its polar dual \( \mathcal{R}^o(\{x_0\}, t) \) at \( t = 2 \), \( \mu = 0.5 \). The curves \( u^{\text{upper}}, u^{\text{lower}} \) defining the reach set boundary (see Corollary 3 and the discussion thereafter) are shown too. (a) \( \mathcal{R} \) with \( x_0 = (0.05, 0.05)^T \). (a) \( \mathcal{R}^o \) with \( x_0 = (0.05, 0.05)^T \).

**Fig. 3.** The bounding polynomials for the double and triple integrator reach sets at \( t = 0.5 \) with \( x_0 = 0 \) and \( \mu = 1 \). (a) Real algebraic curves \( u^{\text{upper}}, u^{\text{lower}} \) for the double integrator. (b) Real algebraic surfaces \( u^{\text{upper}}, u^{\text{lower}} \) for the triple integrator.

**Fig. 4.** Summary of taxonomy for the integrator reach set \( \mathcal{R}^o \).
Integrator reach set \( R \) with \( r = (2,1)^\top \), \( x_0 = (1,1,0)^\top \), \( [a_1,a_2] = [-5,5], [a_2,b_2] = [-3,3] \).

this highlighted region can be further narrowed down, seems significantly more challenging.

V. SIZE

We next quantify the “size” of the reach set \( R \) by computing two functionals: its \( d \)-dimensional volume (see Section V-A), and its diameter or maximum width (see Section V-B). In Section V-C, we discuss how these functionals scale with the state dimension \( d \).

A. Volume

The following result gives the volume formula for the integrator reach set \( R \).

**Theorem 5:** Fix \( x_0 \in \mathbb{R}^d \), let \( \mathcal{X}_0 = \{ x_0 \} \) and \( \mathcal{U} \) given by (20). Consider the integrator dynamics (1) and (2) with \( d \) states, \( m \) inputs, and relative degree vector \( r = (r_1, r_2, \ldots, r_m)^\top \). Define \( \mu_1, \ldots, \mu_m \) as in (11) and (12). Then, the \( d \)-dimensional volume of the integrator reach set (3) at time \( t > 0 \) is

\[
\text{vol}(R\square\{(x_0), t\}) = \frac{2^d}{\prod_{j=1}^m \left( \mu_j^{r_j} \frac{r_j+1}{2} \frac{\prod_{k=1}^{r_j-1} k!}{(2k+1)!} \right)^{1/d}}.
\]

(40)

For a simple illustration of Theorem 5, consider \( d = 3, m = 2 \) with \( r = (2,1)^\top \). The corresponding reach set \( R\square\{(x_0), t\} \) at \( t = 4 \) is shown in Fig. 5 for \( x_0 = (1,1,0)^\top \), \( U = [-5 \times 5] \times [-3,3] \). Here, \( \mu_1 = 5 \) and \( \mu_2 = 3 \).

This reach set, being a direct product of the double integrator reach set \( R_1 \) (cf. Fig. 2) and the single integrator reach set \( R_2 = \{ x_0(3) \} + [-\mu_2t, \mu_2t] \), is a cylinder.\(^5\) In [4], we explicitly derived that \( \text{vol}(R_1) = \frac{2}{3} \mu_2^2 \mu_1^3 \), and therefore, the volume of this cylindrical set must be equal to “height of the cylinder \( \times \) cross sectional area,” i.e.,

\[
2\mu_2 t \times \frac{2}{3} \mu_2^2 \mu_1^3 = \frac{4}{3} \mu_1^2 \mu_2 t^2.
\]

\(^5\) Here, the notation \( x_0(3) \) stands for the third component of vector \( x_0 \).

Indeed, a direct application of the formula (40) recovers the abovementioned expression.

**Remark 2:** If the initial set \( X_0 \) is not singleton, then computing the volume of the \( R \) requires us to compute the volume of a Minkowski sum. Notice that

\[
\text{vol}(\exp(tA)\lambda_0) = |\text{det}(\exp(tA))| \text{vol}(\lambda_0) = \exp(\text{trace}(tA)) \text{vol}(\lambda_0)
\]

\[
= \exp\left( \sum_{j=1}^m \text{trace}(tA_j) \right) \text{vol}(\lambda_0) = \text{vol}(\lambda_0)
\]

since from (2b), \( \text{trace}(A_j) = 0 \) for all \( j = 1, \ldots, m \). Therefore, combining (4), (40) with the classical Brunn–Minkowski inequality, we obtain a lower bound for \( \text{vol}(R\square) \) as

\[
\left( \text{vol}(R\square(X_0, t)) \right)^{1/d} \geq \left( \text{vol}(\lambda_0) \right)^{1/d} + 2\left( \prod_{j=1}^m \mu_j^{r_j} \frac{r_j+1}{2} \frac{\prod_{k=1}^{r_j-1} k!}{(2k+1)!} \right)^{1/d}.
\]

The abovementioned bound holds for any compact \( X_0 \subset \mathbb{R}^d \), not necessarily convex.

B. Diameter

We now focus on another measure of “size” for the integrator reach set \( R \), namely its diameter, or maximal width.

By definition, the width [13, p. 42] of \( R\square(X_0, t) \), is

\[
w_{R\square(X_0, t)}(\eta) := h_{R\square(X_0, t)}(\eta) + h_{R\square(X_0, t)}(-\eta)
\]

where \( \eta \in \mathbb{S}^{d-1} \) (the unit sphere imbedded in \( \mathbb{R}^d \)), and the support function \( h_{R\square(X_0, t)}(\cdot) \) is given by (21). In other words, (41) gives the width of \( R \) in the direction \( \eta \).

For singleton \( X_0 = \{ x_0 \} \), combining (21) and (41), we have

\[
w_{R\square(x_0, t)}(\eta) = \int_0^t \left| \langle \eta, \xi(s) \rangle \right| + \left| \langle -\eta, \xi(s) \rangle \right| \, ds
\]

\[
= 2 \int_0^t \left| \langle \eta, \xi(s) \rangle \right| \, ds
\]

(42)

where the last equality follows from the fact that \( \xi(s) \) in (13) is component-wise nonnegative for all \( 0 \leq s \leq t \).

The diameter of the reach set \( R\square \) is its maximal width

\[
diam \left( R\square(X_0, t) \right) := \max_{\eta \in \mathbb{S}^{d-1}} w_{R\square(X_0, t)}(\eta).
\]

(43)

Notice that (42) is a convex function of \( \eta \); see, e.g., [46, p. 79]. Thus, computing (43) amounts to maximizing a convex function over the unit sphere. We next derive a closed-form expression for (43).

**Theorem 6:** Fix \( x_0 \in \mathbb{R}^d \), let \( X_0 = \{ x_0 \} \) and \( \mathcal{U} \) given by (20). Consider the integrator dynamics (1) and (2) with \( d \) states, \( m \) inputs, and relative degree vector \( r = (r_1, r_2, \ldots, r_m)^\top \). Define \( \mu_1, \ldots, \mu_m \) as in (11) and (12). The diameter of the integrator...
reach set (3) at time $t > 0$ is
\[
diam(R^\square(\{x_0\}, t)) = 2 \|\zeta(t)\|_2 = 2 \left(\sum_{j=1}^{m} \mu_j^2 \|\zeta_j\|^2\right)^{\frac{1}{2}} \tag{44}\]

wherein $\zeta(t)$ is defined as in Section II-A1, and the $i$th component of the subvector $\zeta_j(t) \in \mathbb{R}^r$ is
\[
\int_{0}^{t} s^{(r_j-i)}(r_j-i)! \, ds = \frac{r_j-i+1}{(r_j-i+1)!}, \quad i = 1, 2, \ldots, r_j. \tag{45}
\]

To illustrate Theorem 6, consider the triple integrator with $d = 3$ and $m = 1$. In this case, $U = [\alpha, \beta], \mu := (\beta - \alpha)/2$, and we can parameterize the unit vector $\eta \in S^2$ as
\[
\eta = \left(\frac{\sin \theta \cos \phi}{\cos \theta}, \frac{\sin \theta \sin \phi}{\cos \theta}\right), \quad \theta \in [0, \pi], \quad \phi \in [0, 2\pi).
\]

Thus, (43) reduces to
\[
2\mu \max_{\theta \in [0, \pi], \phi \in [0, 2\pi]} \int_{0}^{t} s \left(\sin \theta \cos \phi\right)/2 + s \sin \theta \sin \phi + \cos \theta \right) \, ds.
\]

Furthermore, $\zeta(t) = (t^3/6, t^2/2, t)^\top$, and we obtain
\[
\eta_{\text{max}}^\text{max} = \left(\frac{\sin \theta_{\text{max}} \sin \phi_{\text{max}}}{\cos \theta_{\text{max}}}, \frac{\sin \theta_{\text{max}} \cos \phi_{\text{max}}}{\cos \theta_{\text{max}}}\right) = \frac{\pm 1}{\sqrt{t^4 + 9t^2 + 36}} \left(\frac{t^2}{6}\right)
\]

where $\pm$ means that either all components are plus or all minus. Thus, the maximizing tuples $(\phi_{\text{max}}, \theta_{\text{max}}^\text{max}) \in [0, \pi] \times [0, 2\pi)$ are given by
\[
(\phi_{\text{max}}, \theta_{\text{max}}) = \{(\arctan(3/t), \arcsin(6/\sqrt{t^4 + 9t^2 + 36})) \}
\]

Hence, the diameter of the triple integrator reach set at time $t$ is equal to $(\mu t/3)\sqrt{t^4 + 9t^2 + 36}$.

Fig. 6 shows how the width of the integrator reach set for $d = 3$, $m = 1$ varies over $(\phi, \theta) \in [0, \pi] \times [0, 2\pi)$, which parameterize the unit sphere $S^2$. The location of the maximizers are given by (46), and are depicted in Fig. 6 via filled black circle and filled black square.

For a visualization of the width and diameter for the double integrator, see [4, Fig. 2].

### C. Scaling Laws

We now turn to investigate how the volume and the diameter of the integrator reach set scale with time and the state dimension. While scaling laws reveal limits of performance of engineered systems (see, e.g., [47], [48]), they have not been formally investigated in the context of reach sets even though empirical studies are common [20], [49, Table 1], [50, Fig. 4].

For clarity, we focus on the single-input case and hence do not notationally distinguish between $R$ and $R^\square$.

1) Scaling of the Volume: Fig. 7 plots the volume (40) for the single input $(m = 1)$ case against time $t$ for varying state space dimension $d$. In this case, $U = [\alpha, \beta]$, and therefore, $\mu := (\beta - \alpha)/2$. As expected, the volume of the reach set increases with time for any fixed $d$.

Let us now focus on the scaling of the volume with respect to the state dimension $d$. For $m = 1$, using the known asymptotic [51] for $\prod_{k=1}^{d-1}(2k + 1)/k!$, we find the $d \to \infty$...
asymptotic for the volume
\[
\text{vol} (\mathcal{R}_d (\{x_0\}, t)) \sim (2\mu)^d t^{d(d+1)/2} \exp \left( \frac{\mu^2 t^2 + 1}{e} \right) \frac{\Gamma \left( \frac{d+1}{2} \right)}{2^{(d+1)/2}} \text{vol} \left( \mathcal{R}_d (\{x_0\}, t_{cr}) \right)
\]
where \( c \approx 1.2824 \ldots \) is the Glaisher–Kinkelin constant \([52, \text{Sec. 2.15}]\).

Fig. 7 shows that when \( t \) is small, the volume of the larger dimensional reach set stays lower than its smaller dimensional counterpart. In particular, given two state space dimensions \( d, d' \) with \( d > d' \), and all other parameters kept fixed, there exists a critical time \( t_{cr} \) when the volume of the \( d \) dimensional reach set overtakes that of the \( d' \) dimensional reach set.

For any \( d > d' \), the critical time \( t_{cr} \) satisfies
\[
\frac{\text{vol} (\mathcal{R}_d (\{x_0\}, t_{cr}))}{\text{vol} (\mathcal{R}_{d'} (\{x_0\}, t_{cr}))} = \frac{d_{d'}}{d_{d'}}
\]
which together with (40) yields
\[
t_{cr} = (2\mu)^{-d} \frac{2}{d+1} \left( \prod_{k=d'}^{d-1} \frac{2k+1}{k!} \right)^{1/(d+1)}.
\]

In particular, for \( d' = d - 1 \), we get
\[
t_{cr} = \left( \frac{2}{2\mu} \right)^{1/d} (d-1)!^{1/d}, \quad d = 2, 3, \ldots
\]

For instance, when \( \mu = 1, \) \( d = 3, \) \( d' = 2, \) we have \( t_{cr} = (30)^{1/3} \approx 3.1072. \) When \( \mu = 1, d = 4, d' = 3, \) we have \( t_{cr} = (420)^{1/4} \approx 4.5270. \) The dashed vertical lines in Fig. 7 show the critical times given by (48).

Applying Stirling’s approximation \( n! \sim \sqrt{2\pi n} (n/e)^n \), we obtain the \( d \to \infty \) asymptotic for (48)
\[
t_{cr} \sim \frac{4}{e} \frac{1}{d} \mu^{-\frac{1}{2}} 2^{-\frac{d}{2}}
\]
where \( \sim \) denotes asymptotic equivalence \([53, \text{Ch. 1.4}]\), and \( e \) is the Euler number.

2) Scaling of the Diameter: Fig. 8 plots the diameter of (44) for the single input \((m = 1)\) case against time \( t \) for varying state space dimension \( d \). As earlier, \( t_{cr} = (\alpha, \beta), \mu := (\beta - \alpha)/2 \). As expected, the diameter of the reach set increases with time for any fixed \( d \).

As \( d \to \infty \), the diameter approaches a limiting curve shown by the dotted line in Fig. 8. To derive this limiting curve, notice that for \( m = 1 \), the formula (44) gives
\[
\lim_{d \to \infty} \text{diam} (\mathcal{R} (\{x_0\}, t)) = \lim_{d \to \infty} 2\mu \left( \sum_{j=1}^{d} \frac{t_j^2}{j!} \right)^{1/2}.
\]

We write the partial sum
\[
\sum_{j=1}^{d} \left( \frac{t_j^2}{j!} \right)^2 = \sum_{j=1}^{d} \left( \frac{t_j^2}{j!} \right)^2 - \sum_{j=d+1}^{\infty} \left( \frac{t_j^2}{j!} \right)^2 =: S_1 - S_2
\]
and by ratio test, note that both the sums \( S_1, S_2 \) converge. In particular, \( S_1 \) converges to \( I_0(2t) - 1 \), where \( I_0(\cdot) \) is the zeroth order modified Bessel function of the first kind. This follows from the very definition of the zeroth order modified Bessel function of the first kind, given by
\[
I_0(z) := (z/2)^n \sum_{j=0}^{\infty} \frac{(z^2/4)^j}{j! \Gamma(n+j+1)}, \quad z \in \mathbb{R}
\]
where \( \Gamma(\cdot) \) denotes the Gamma function.

On the other hand, using the definition of the generalized hypergeometric function\(^6\)
\[
1_F(1; b_1, b_2; z) := \sum_{n=0}^{\infty} \frac{(a_1)_n}{(b_1)_n} \frac{z^n}{n!}
\]
we find that
\[
S_2 = \frac{t^{2(d+1)} \cdot 1_F(1; d+2, d+2; t^2)}{((d+1)!)^2}.
\]
Therefore, (50) evaluates to
\[
S_1 - S_2 = I_0(2t) - 1 - \frac{t^{2(d+1)} \cdot 1_F(1; d+2, d+2; t^2)}{((d+1)!)^2}.
\]

Combining (49), (50), (51), and using the continuity of the square root function on \([0, \infty)\), we deduce that
\[
\lim_{d \to \infty} \text{diam} (\mathcal{R} (\{x_0\}, t)) = 2\mu \sqrt{I_0(2t) - 1}.
\]

That \( \lim_{d \to \infty} S_2 \) exists and equals to zero, follows from (50) and the continuity of the square
\[
\lim_{d \to \infty} S_2 = \lim_{j \to \infty} \left( \frac{t_j^2}{j!} \right)^2 = \left( \lim_{j \to \infty} \frac{t_j^2}{j!} \right)^2 = 0.
\]

\(^6\)Here, \((\cdot)_n\) denotes the Pochhammer symbol \([54, \text{p. 256}]\) or rising factorial.
To see the last equality, let \( a_j := n_j/j! \). By the ratio test, \( \lim \sup (a_{j+1}/a_j) = \lim t/j = 0 < 1 \), hence, \( \{a_j\} \) is a Cauchy sequence and \( \lim j \to \infty a_j = 0 \).

The dotted line in Fig. 8 is the curve (52).

VI. BENCHMARKING OVERAPPROXIMATIONS OF INTEGRATOR REACH SETS

In practice, a standard approach for safety and performance verification is to compute “tight” overapproximation of the reach sets of the underlying controlled dynamical system. Several numerical toolboxes such as [2], [3] are available which overapproximate the reach sets using simple geometric shapes such as zonotopes and ellipsoids. Depending on the interpretation of the qualifier “tight,” different optimization problems ensue, e.g., minimum volume outer-approximation [55], [56], [57], [58], [59], [60], [61], [62].

One potential application of our results in Section V is to help quantify the conservatism in different overapproximation algorithms by taking the integrator reach set as a benchmark case. For instance, Fig. 9 shows the conservatism in zonotopic overapproximations \( \mathcal{R}_{\text{approx}}(t) \) of the single input integrator reach sets \( \mathcal{R}(\{0\}, t) \subset \mathcal{R}_{\text{approx}}(\{0\}, t) \) for \( d = 2, 3, 4 \) with \( 0 \leq t \leq 1 \) and \( \mu = 1 \), computed using the CORA toolbox [2], [63]. To quantify the conservatism, we used the volume formula (40) for computing the ratio of the volumes \( \text{vol}(\mathcal{R})/\text{vol}(\mathcal{R}_{\text{approx}}) \in [0, 1] \). The results shown in Fig. 9 were obtained by setting the zonotope order 50 in the CORA toolbox, which means that the number of zonotopic segments used by CORA for overapproximation was \( \leq 50d \). As expected, increasing the zonotope order improves the accuracy at the expense of computational speed, but among the different dimensional volume ratio curves, trends similar to Fig. 9 remain.

For the numerical results shown in Fig. 9, we found the diameters of the overapproximating zonotopes for \( d = 2, 3, 4 \), to be the same as that of the true diameters given by (44) for all times.

Fig. 10 depicts the conservatism in ellipsoidal overapproximations \( \mathcal{R}_{\text{approx}}(t) \) of the single input integrator reach sets \( \mathcal{R}(\{0\}, t) \subset \mathcal{R}_{\text{approx}}(\{0\}, t) \) for \( d = 2, 3, 4 \) with \( 0 \leq t \leq 1 \) and \( \mu = 1 \), following the algorithms in ellipsoidal toolbox [3]. Specifically, the reach set at time \( t \), is overapproximated by the intersection of a carefully constructed parameterized family of ellipsoids \( E(q(t), Q_{\ell_i}(t)) \) defined as

\[
\{ x \in \mathbb{R}^d \mid (x - q(t)) (Q_{\ell_i}(t))^{-1} (x - q(t))^\top \leq 1 \}
\]

for unit vectors \( \ell_i(0) \in \mathbb{R}^d \), \( i = 1, \ldots, N \). The choice of \( \ell_i(0) \) determines \( \ell_i(t) := \exp(-A^\top t)\ell_i(0) \), which in turn parameterizes the \( d \times d \) symmetric positive definite shape matrix \( Q_{\ell_i}(t) \); we refer the readers to [64, Ch. 3.2], [37, Ch. 3] where the corresponding evolution equations were derived using optimal control. The center vectors \( q(t) \in \mathbb{R}^d \), and the shape matrices \( Q_{\ell_i}(t) \) for this parameterized family of ellipsoids are constructed such that \( \bigcap_{i=1}^{\infty} E(q(t), Q_{\ell_i}(t)) \) is guaranteed to be a superset of the reach set at time \( t \) for any finite \( N \), and for \( N \to \infty \), recovers the reach set at that time.

For the results shown in Fig. 10, we used \( N = 20 \) randomly chosen unit vectors \( \ell_i(0) \in \mathbb{R}^d \). Ideally, one would like to compute the (unique) minimum volume outer ellipsoid (MVOE), a.k.a. the Löwner–John ellipsoid [65], [66] of the convex set \( \bigcap_{i=1}^{\infty} E(q(t), Q_{\ell_i}(t)) \), which is a semi-infinite programming problem [46, Ch. 8.4.1], and has no known exact semidefinite programming (SDP) reformulation. We computed two different relaxations of this problem: one based on the \( S \) procedure [67, Ch. 3.7.2], and the other by homothetic scaling of the maximum volume inner ellipsoid (MVIE) [65, Th. III] of the set \( \bigcap_{i=1}^{\infty} E(q(t), Q_{\ell_i}(t)) \). Both of these lead to solving SDP problems, and both are guaranteed to contain the Löwner–John ellipsoid of the intersection of the parameterized family of ellipsoids. These suboptimal (w.r.t. the MVOE criterion) solutions, computed using cvx [68], are shown in Fig. 10.

Fig. 10 shows that the \( S \) procedure entails less conservatism compared to the MVIE scaling, in terms of volume. While the volume ratio trends in Fig. 10 are similar to that observed in Fig. 9, the approximation quality is lower. In light of the results in Section IV-A, this is not surprising: the integrator reach sets being zonoids (i.e., Hausdorff limit of zonotopes), the zonotopic outer approximations are expected to perform better than other overapproximating shape primitives.

The main point here is that our results in Section V provide the ground truth for the size of the integrator reach set, thereby help benchmarking the performance of reach set approximation algorithms.

VII. EPILOGUE

A. Recap

This article initiates a systematic study of integrator reach set. When the input uncertainty set is hyperrectangle, we showed that the corresponding compact convex reach set \( \mathcal{R}^{\square} \) is in fact semialgebraic (Section IV-B) as well as a zonoid (range of an atom free vector measure) up to translation (Section IV-A). We derived the equation of its boundary in both parametric (Proposition 2) and implicit form (Section IV-C). We obtained the closed-form formula for the volume (Section V-A) and
Fig. 10. (Top) Ellipsoidal overapproximations of the double integrator reach sets; (bottom) the ratio of the volume (left) and diameter (right) of the single input integrator reach set $R(t)$ and that of its ellipsoidal overapproximation $R_{\text{approx}}(t)$ for $d = 2, 3, 4$, plotted against time $t \in [0, 1]$. Two different ellipsoidal overapproximations are shown: one (in red) based on the $S$ procedure, and the other (in blue) obtained by scaling the MVIE of the intersection of a parameterized family of ellipsoids. The results are computed for $\mu = 1$, $X_0 = \{0\}$.

The block diagonal structure of the matrix $A$ in (2) implies

$$\sup_{x_0 \in X_0} \langle y, \exp(tA)x_0 \rangle = \sup_{x_0 \in X_0} \sum_{j=1}^m \langle y_j, \exp(tA_j)x_{j0} \rangle. \quad (54)$$

Following the definition of support function and [4, Proposition 1], we then have

$$h_{f_{\mu}^t \exp(sA)Bu} ds(y) = \int_0^t h_{\exp(sA)Bu} \langle y \rangle ds \quad (13)$$

$$= \int_0^t \sup_{u \in U} \langle y, \exp(sA)Bu \rangle ds \quad (13)$$

$$= \int_0^t \sup_{u \in \text{closure}(\text{conv}(U))} \sum_{j=1}^m \langle y_j, \xi_j(s)u_j \rangle ds. \quad (55)$$

The last equality in (55) follows from (13), and from the fact [26, Prop. 6.1] that the reach set remains invariant under the closure of convexification of the input set $U$. Substituting (54) and (55) in (53) yields (16).

B. Proof of Theorem 1

Since the uncertainties in (20) along different input coordinate axes are mutually independent, the support function of the reach set is of the form (19). Therefore, in this case, (16)

APPENDIX

A. Proof of Proposition 1

Since support function is distributive over sum, we have

$$h_{R(x_0,t)}(y) = \sup_{x_0 \in X_0} \langle y, \exp(tA)x_0 \rangle + h_{f_{\mu}^t \exp(sA)Bu} ds(y). \quad (53)$$

The block diagonal structure of the matrix $A$ in (2) implies

$$\sup_{x_0 \in X_0} \langle y, \exp(tA)x_0 \rangle = \sup_{x_0 \in X_0} \sum_{j=1}^m \langle y_j, \exp(tA_j)x_{j0} \rangle. \quad (54)$$

Following the definition of support function and [4, Proposition 1], we then have

$$h_{f_{\mu}^t \exp(sA)Bu} ds(y) = \int_0^t h_{\exp(sA)Bu} \langle y \rangle ds \quad (13)$$

$$= \int_0^t \sup_{u \in U} \langle y, \exp(sA)Bu \rangle ds \quad (13)$$

$$= \int_0^t \sup_{u \in \text{closure}(\text{conv}(U))} \sum_{j=1}^m \langle y_j, \xi_j(s)u_j \rangle ds. \quad (55)$$

The last equality in (55) follows from (13), and from the fact [26, Prop. 6.1] that the reach set remains invariant under the closure of convexification of the input set $U$. Substituting (54) and (55) in (53) yields (16).

B. Proof of Theorem 1

Since the uncertainties in (20) along different input coordinate axes are mutually independent, the support function of the reach set is of the form (19). Therefore, in this case, (16)
takes the form
\[
\begin{align*}
    h_{R,\{x_0,t\}}(y) &= \sum_{j=1}^{m} \left\{ \sup_{x_j \in R_{\{x_0\}}} \langle y_j, \exp(tA_j)x_j \rangle + \int_0^t \sup_{u_j \in [\alpha_j,\beta_j]} \langle y_j, \xi_j(s) \rangle u_j \, ds \right\}, \\
    &\text{The optimizer } u_j^{\text{opt}} \text{ of the integrand in the RHS of (56), for } j \in [m], \text{ can be written in terms of the Heaviside unit step function } H(\cdot) \text{ as}
    \end{align*}
\]
for \(0 \leq s \leq t\). Substituting (57) back in (56) and integrating over \(s\) completes the proof. □

C. Proof of Theorem 2

For \(s \in [0,t]\), let the vector measure \(\mu\) be defined as \(d\mu(s) := \xi(s)ds\) where \(\xi(s)\) is given by (13). Then, \(\int_0^t |\langle y, \xi(s) \rangle|ds\) is exactly in the form of a support function of a zonoid (see, e.g., [28, Sec. 2]). Using the one-to-one correspondence between a compact convex set and its support function, the corresponding set is a zonoid.

From (9) and (21), \(R^\square(\{x_0\},t)\) is the translation of a set with support function \(\int_0^t |\langle y, \xi(s) \rangle|ds\), i.e., the translation of a zonoid. Thus, \(R^\square(\{x_0\},t)\) is a zonoid. □

D. Proof of Proposition 2

From Section II-A1, the supporting hyperplane at any \(x_{b,b,y} \in \partial R^\square(\{x_0\},t)\) with outward normal \(y \in \mathbb{R}^d\) is \(\langle y, x_{b,b,y} \rangle = h_{R^\square(\{x_0\},t)}(y)\), and the Legendre–Fenchel conjugate
\[
    h_{R^\square(\{x_0\},t)}(x_{b,b,y}) = 0.
\]

For \(j \in [m]\), let \(y\) comprise of subvectors \(y_j \in \mathbb{R}^{r_j}\). Since the Cartesian product (18) is equivalent to the Minkowski sum \(R_1 + \cdots + R_m\), and the support function of Minkowski sum is the sum of support functions of the summand sets [13, p. 48], we have
\[
\begin{align*}
    h_{R^\square(\{x_0\},t)}(y) &= \sum_{j=1}^{m} h_{R_j(\{x_0\},t)}(y_j) \\
    \Rightarrow h_{R^\square(\{x_0\},t)}(x_{b,b,y}) &= \sum_{j=1}^{m} h_{R_j(\{x_0\},t)}(x_{b,b,y})
\end{align*}
\]

wherein the last line follows from the property that the Legendre–Fenchel conjugate of a separable sum equals to the sum of the Legendre–Fenchel conjugates [46, p. 95].

Therefore, combining (58) and (59), we obtain
\[
\begin{align*}
    \sum_{j=1}^{m} \inf_{y_j \in \mathbb{R}^{r_j}} \left\{ -x_{b,b,y}^j + \exp(tA_j)x_{j,0} + \nu_j \zeta_j(t), y_j \right\} + \mu_j \int_0^t |\langle y_j, \xi_j(s) \rangle| \, ds = 0.
\end{align*}
\]

For \(j \in [m]\), since each objective in (60) involves an integral of the absolute value of a polynomial in \(s\) of degree \(r_j - 1\) that polynomial can have at most \(r_j - 1\) roots in the interval \([0,t]\), i.e., can have at most \(r_j - 1\) sign changes in that interval. If all \(r_j - 1\) roots of the aforesaid polynomial are in \([0,t]\), we denote these roots as \(s_1 \leq s_2 \leq \cdots \leq s_{r_j-1}\), and write
\[
\begin{align*}
    \int_0^t |\langle y_j, \xi_j(s) \rangle| \, ds &= \pm \int_0^{s_1} |\langle y_j, \xi_j(s) \rangle| \, ds \mp \int_{s_1}^{s_2} |\langle y_j, \xi_j(s) \rangle| \, ds \\
    &\mp \cdots \mp (-1)^{r_j-1} \int_{s_{r_j-2}}^{s_{r_j-1}} |\langle y_j, \xi_j(s) \rangle| \, ds \\
    &= (y_j, +\zeta_j(s_1, s_2) \mp \zeta_j(s_1, s_2) \pm \cdots \mp (-1)^{r_j-1}\zeta_j(s_{r_j-1}, t), y_j).
\end{align*}
\]

Notice that even if the number of roots in \([0,t]\) is strictly less than \(r_j - 1\), the expression (61) is generic in the sense the corresponding summand integrals become zero. Thus, combining (60) and (61), we arrive at
\[
\begin{align*}
    \sum_{j=1}^{m} \inf_{y_j \in \mathbb{R}^{r_j}} \left\{ -x_{b,b,y}^j + \exp(tA_j)x_{j,0} + \nu_j \zeta_j(t), y_j \right\} + \mu_j \zeta_j(s_1, s_2) \pm \cdots \pm \mu_j (-1)^{r_j-1}\zeta_j(s_{r_j-1}, t), y_j) = 0.
\end{align*}
\]

The left-hand side of (62) being the sum of the infimum values of linear functions, can achieve zero if and only if each of those infimum equals to zero, i.e., if and only if
\[
\begin{align*}
    x_{b,b,y}^j &= \exp(tA_j)x_{j,0} + \nu_j \zeta_j(t, y_j) + \mu_j \zeta_j(s_1, s_2) \pm \cdots \pm \mu_j (-1)^{r_j-1}\zeta_j(s_{r_j-1}, t), y_j.
\end{align*}
\]

Using (6), (13), and (14), we simplify (63) to (26), thereby completing the proof. □

E. Proof of Corollary 3

From (26), we get two different parametric representations of \(x_{b,b,y}^j\) in terms of \((s_1, s_2, \ldots, s_{r_j-1})\). One parametric representation results from the choice of positive sign for the \(\pm\) appearing in (26), and another for the choice of negative sign for the same. Denoting the implicit representation corresponding to the parametric representation (26) with \(+\) (resp. \(-\)) sign as
\[
\begin{align*}
    p_{b,b,y}^{\text{up}}(x) &= 0 \quad (\text{resp. } p_{b,b,y}^{\text{low}}(x) = 0).
\end{align*}
\]

F. Proof of Theorem 4

We notice that (26) gives polynomial parameterizations of the components of \(x_{b,b,y}^j\) for all \(j \in [m]\). In particular, for each \(k \in [r_j]\), the right-hand side of (26) is a homogeneous polynomial in \(r_j - 1\) parameters \((s_1, s_2, \ldots, s_{r_j-1})\) of degree \(r_j - k + 1\). By polynomial implicitization [25, p. 134], the corresponding implicit equations \(p_{b,b,y}^{\text{up}}(x_{b,b,y}^j) = 0\) [with fixing plus sign for \pm in (26)] and \(p_{b,b,y}^{\text{low}}(x_{b,b,y}^j) = 0\) [with fixing minus sign for \pm in (26)], must define affine varieties \(V_{R}^{\text{up}}(x) \subseteq V_{R}^{\text{low}}(x) \subseteq R^{x_1, \ldots, x_d}\).
Specifically, denote the right-hand sides of (26) as $g_j^+, \ldots, g_j^n$ for all $j \in [m]$, where the superscripts indicate that either all $g$s are chosen with plus signs, or all with minus signs. Then, write (26) as $x_j^{bly}(1) = g_j^+(s_1, s_2, \ldots, s_{r_j-1}) \ldots, x_j^{bly}(r_j) = g_j^n(s_1, s_2, \ldots, s_{r_j-1})$. Now for each $j \in [m]$, consider the ideal

$$I_j := \left\{ \left< x_j^{bly}(1) - g_j^+, x_j^{bly}(2) - g_j^2, \ldots, x_j^{bly}(r_j) - g_j^n \right> \right\} \subseteq \mathbb{R}[s_1, s_2, \ldots, s_{r_j-1}, x_1, x_2, \ldots, x_n]$$

and let $I_{j,r_j-1} := I_j \cap \mathbb{R}[x_1, \ldots, x_{r_j}]$ be the $(r_j-1)$th elimination ideal of $I_j$. Then, for each $j \in [m]$, the variety

$$V(I_{j,r_j-1}) = V_{x_1, \ldots, x_{r_j}}(p_{j,upper})$$

Likewise, the variety $V(I_{j-1}) = V_{x_1, \ldots, x_{r_j}}(p_{j,lower})$.

Thus, the algebraic boundary (i.e., the Zariski closure of the Euclidean boundary) of $\mathcal{R}_j$ is

$$\partial \mathcal{R}_j = V_{x_1, \ldots, x_{r_j}}(p_{j,upper}) \cup V_{x_1, \ldots, x_{r_j}}(p_{j,lower}).$$

Therefore, $\mathcal{R}_j := \{ x \in \mathbb{R}^{r_j} | p_{j,upper}(x) \leq 0, p_{j,lower}(x) \leq 0 \}$ is semialgebraic for all $j \in [m]$. Since the Cartesian product of semialgebraic sets is semialgebraic, the statement follows from (18). \[\blacksquare\]

**G. Proof of Theorem 5**

We organize the proof in three steps.

**Step 1:** From (18), we have

$$\text{vol} \left( \mathcal{R} \left( \{ x_0 \}, t \right) \right) = \text{vol} \left( \mathcal{R}_1 \times \mathcal{R}_2 \times \cdots \times \mathcal{R}_m \right) = \prod_{j=1}^m \text{vol} \left( \mathcal{R}_j \left( \{ x_0 \}, t \right) \right).$$

**Step 2:** Motivated by (64), we focus on deriving the $r_j$-dimensional volume of $\mathcal{R}_j(\{x_0\}, t)$. For this purpose, we proceed as in [4] by uniformly discretizing the interval $[0, t]$ into $n$ subintervals $[ (i-1)t/n, it/n ]$, $i = 1, \ldots, n$, with $(n+1)$ breakpoints $t_i := it/n$ for $i = 0, 1, \ldots, n$.

From (24), we then have

$$\text{vol} \left( \mathcal{R}_j \left( \{ x_0 \}, t \right) \right) = \lim_{n \to \infty} \frac{1}{n^{r_j}} \text{vol} \left( \sum_{i=0}^n \mu_j \xi_j(t_i) [-1, 1] \right)$$

where $\xi_j$ was defined in (13). We recognize that the set $\sum_{i=0}^n \mu_j \xi_j(t_i) [-1, 1]$ in (65) is a Minkowski sum of $n + 1$ intervals, i.e., a zonotope imbedded in $\mathbb{R}^{r_j}$, wherein each interval is rotated and scaled in $\mathbb{R}^{r_j}$ via different linear transformations $\exp(t_i A_j)$, $i = 0, 1, \ldots, n$. Using the formula for the volume of zonotopes [15, eq. (57)], [69, Exercise 7.19], we can write (65) as

$$\text{vol} \left( \mathcal{R}_j \left( \{ x_0 \}, t \right) \right) = (2t_j)^{r_j} \lim_{n \to \infty} \frac{1}{n^{r_j}} \sum_{i=0}^n \det \left( \xi_j(t_i) \right) .$$

**Step 3:** Our next task is to simplify (69). Observe that the sum

$$\sum_{0 \leq i_1 < i_2 < \cdots < i_{r_j} \leq n} \prod_{k=1}^{r_j} (i_k - i_{k-1})$$

returns a polynomial in $n$ of degree $r_j(r_j + 1)/2$, and hence, the limit in (69) is always well-defined. Specifically, the limit extracts the leading coefficient of this polynomial.

Let us denote the leading coefficient of the sum (70) as $c(r_j)$. By the Euler–Maclaurin formula [71], [72, Ch. II.10]

$$c(r_j) = \int_{0 \leq y_1 < y_2 < \cdots < y_{r_j} \leq 1} \prod_{a=1}^{r_j} (y_a - y_b) \cdot \prod_{a=1}^{r_j} dy_a .$$

One way to unpack (71) is to write it as a sum over the symmetric permutation group $\mathcal{S}_{r_j}$ of the finite set $[r_j]$, i.e.,

$$c(r_j) = \sum_{\sigma \in \mathcal{S}_{r_j}} \text{sgn}(\sigma) \prod_{k=1}^{r_j} (\sigma_1 + \sigma_2 + \cdots + \sigma_k)$$

where $\text{sgn}(\sigma) := (-1)^{\nu}$, $\nu := \#(i, j) | i < j, \sigma(i) > \sigma(j)$, and # stands for “the number of.” We will now prove that

$$c(r_j) = \sum_{k=1}^{r_j-1} \frac{(k)!^2}{(2k+1)!} .$$
To this end, we write \( r_j! \cdot c(r_j) \) as an integral over \([0, 1]^r\)
\[
r_j! \cdot c(r_j) = \int_{[0, 1]^r} \prod_{1 \leq a < b \leq r_j} |y_a - y_b| \, dy_1 \ldots dy_{r_j}. \tag{73}
\]

In 1955, de Bruijn [73, see toward the end of Sec. 9] used certain Pfaffians to evaluate

\[
= \frac{r_j! \cdot \{1! \times 2! \times \cdots \times (r_j - 1)!\}^2}{1! \times 3! \times \cdots \times (2r_j - 1)!}, \quad r_j = 2, 3, \ldots
\]

which upon substitution in (73), indeed yields (72). Combining (69) and (72), we arrive at

\[
\text{vol} (\mathcal{R}_j (\{x_0\}, t)) = \left(\frac{2\mu_j}{r_j} \right)^{r_j} \frac{1}{r_j - 1} \prod_{k=1}^{r_j - 1} k! \prod_{k=1}^{r_j - 1} \frac{k!}{(2k + 1)!}.
\tag{74}
\]

Finally, substituting (74) in (64), and recalling that \( r_1 + r_2 + \ldots + r_m = d \), the expression (40) follows. \hfill \blacksquare

H. Proof of Theorem 6

From (13), the subvector \( \xi_j(s) \), where \( j \in [m] \), is component-wise nonnegative for all \( s \in [0, t] \).

Therefore, by triangle inequality, we have

\[
\int_0^t |\langle \eta, \xi_j(s) \rangle| \, ds \leq \int_0^t \sum_{j=1}^m |\langle \eta_j, \mu_j \xi_j(s) \rangle| = \langle \eta |, \zeta(t) \rangle \tag{75}
\]

where \( |\eta_j| \) denotes the \( j \)th subvector with component-wise absolute values. Let us call \( |\eta| \) as the “absolute unit vector.”

The upper bound in (75) is convex in \( \eta \), and is maximized by an absolute unit vector collinear with \( \zeta(t) \) given by

\[
\eta = \pm \frac{\zeta(t)}{\| \zeta(t) \|_2} \tag{76}
\]

i.e., the unit vectors associated with \( \zeta(t) \) up to plus-sign minus sign permutations among its components.

Out of the \( 2^d \) unit vectors given by (76), the “all plus” and “all minus” unit vectors achieve equality in (75) and, hence, must be the maximizers of (42). Inequality (75) remains strict for the remaining \( 2^d - 2 \) unit vectors in (76), thus, are suboptimal for (42). Therefore, the maximizers in (43) are

\[
\eta^\text{max} = \frac{\zeta(t)}{\| \zeta(t) \|_2}, \quad -\frac{\zeta(t)}{\| \zeta(t) \|_2}
\]

which upon substitution in (42), results in (44). \hfill \blacksquare

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