

A geometric characterization of observability in inertial parameter identification

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Abstract

This paper presents an algorithm to geometrically characterize inertial parameter identifiability for an articulated robot. The geometric approach tests identifiability across the infinite space of configurations using only a finite set of conditions and without approximation. It can be applied to general open-chain kinematic trees ranging from industrial manipulators to legged robots, and it is the first solution for this broad set of systems that is provably correct. The high-level operation of the algorithm is based on a key observation: Undetectable changes in inertial parameters can be represented as sequences of inertial transfers across the joints. Drawing on the exponential parameterization of rigid-body kinematics, undetectable inertial transfers are analyzed in terms of observability from linear systems theory. This analysis can be applied recursively, and lends an overall complexity of $O(N)$ to characterize parameter identifiability for a system of N bodies. MATLAB source code for the new algorithm is provided.

Keywords

Dynamics theoretical foundations, calibration and identification, theoretical foundations, legged robots, humanoids and animaloids

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1. Introduction

A classic problem in robotics is the identification of inertial parameters (mass, center of mass, and inertia) for each link of a mechanism. This problem has received attention through multiple decades, with original work on the identification of manipulators (Atkeson et al., 1986; Khalil and Bennis, 1995; Swevers et al., 1997) seeing extensions to the identification of mobile robots and humans in more recent applications (Ayusawa et al., 2008, 2010, 2014; Jovic et al., 2016; Lee et al., 2020; Traversaro et al., 2013). Across domains, an enabling property is that the inverse dynamics of a rigid-body system are linear in its inertial parameters, motivating least-squares solutions to identify parameters from the measurement of joint kinematics, joint torques, and external forces.

It is well known, however, that not all inertial parameters are identifiable from these measurements (Atkeson et al., 1986). This observation has motivated studying which parameters (or combinations thereof) can theoretically be identified from an infinite amount of data (Ayusawa et al., 2014; Chen et al., 2002; Gautier and Khalil, 1990; Kawasaki et al., 1991; Khalil and Bennis, 1995; Mayeda et al., 1990; Ros et al., 2012; Sheu and Walker, 1991). This problem is one of characterizing the *structural identifiability* (Bellman and Aström, 1970) of a model form. It is emphasized that this problem is distinct from the problem of fitting a model

to experimental data. Naturally, the quality of any identified model depends both on the accuracy of the model form assumed and the quality of the data collected. Thus, the related problem of *practical identifiability* addresses how uncertainty in measurements relates to uncertainty in the identified parameters (e.g., Calafiore and Indri, 2000; Ramdani and Poinet, 2005). In short, structural identifiability considers what model properties can be identified from data, whereas practical identifiability considers how non-ideal aspects of that data impact the inferred model.

This paper presents new tools for characterizing the structural identifiability of a rigid-body system via a geometric approach to the problem. We provide an algorithm that takes a kinematic model as an input, and returns the

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linear combinations of the system's inertial parameters that are identifiable from measurements of the joint kinematics, joint torques, and external forces. The approach taken offers an alternative to previous numerical (e.g., [Gautier, 1991](#)) or symbolic approaches (e.g., [Gautier and Khalil, 1990](#); [Khosla, 1989](#)) to the problem. The main benefit of our approach is that the algorithm requires only a finite set of conditions to test the infinite space of configurations. It also comes with a theoretical proof of correctness that holds for arbitrary open-chain fixed- or floating-base systems with generic joint types. This proof of correctness is achieved by working with spatial (6D) inertias in a parameterization-free sense and only adopting inertial parameters for the implementation of the theory.

We show that undetectable changes in the inertial parameters can be represented by exchanges of mass and inertia between neighboring bodies. Exchanges are considered undetectable if they leave the system dynamics unchanged across the entire state space. Addressing this infinity of possibilities is the main hurdle that has previously prevented a rigorous treatment of structural identifiability for general robot systems.

1.1. Related previous work

We begin by motivating the consideration of identifiability through its relevance to applications in system identification and adaptive control. We then discuss, in detail, past approaches for identifiability analysis to contextualize our contribution.

1.1.1. Motivation. Early work on system identification emphasized determining minimal sets of parameters (base parameters) to be identified ([Gautier, 1991](#)). Base parameters group together linear combinations of link parameters that appear together in the equations of motion. Isolating these groupings allows using a reduced parameter set for faster dynamics computations ([Khalil and Kleinfinger, 1987](#)), and enables identification to have a unique solution.

Since this work back in the 1980s and 90s, increases in computing power and new identification methodologies suggest revisiting these original motivations. Regarding computation power, the recursive Newton–Euler algorithm (RNEA) can now be carried out with the full parameter set in microseconds for complex systems ([Carpentier et al., 2019](#)). In terms of methodology, recent advances in enforcing physical consistency of the parameters ([Sousa and Cortesão, 2014](#); [Traversaro et al., 2013](#); [Wensing et al., 2017a](#)) and geometric regularization ([Lee et al., 2020](#); [Lee and Park, 2018](#)) jointly suggest the benefits of considering the full parameter set when carrying out identification.

Despite these recent advances, structural identifiability considerations remain fundamental for robot system identification. Recall that structural identifiability analysis characterizes which parameters (or combinations) can be deduced from an *infinite* amount of training data. Methods that design exciting trajectories ([Calafiore et al., 2001](#);

[Gautier and Khalil, 1992](#); [Jovic et al., 2015](#); [Lee et al., 2021](#); [Swevers et al., 1997](#)) seek to maximally identify model information with a *finite* amount of training data. It is only possible to certify that a given dataset is maximally exciting, however, via comparison to a structural identifiability analysis. Another motivation for characterizing base parameters comes from instrumental variable identification techniques ([Janot et al., 2014a, 2014b](#)) that address model bias from noisy data, but currently require the use of a non-redundant model parameterization. Methods that characterize uncertainty in the parameter estimates likewise only do so through considering uncertainty in the base parameters ([Calafiore and Indri, 2000](#); [Ramdani and Poignet, 2005](#)). Yet other threads of work have exploited how the full parameters project onto a base parameter set for payload identification ([Gaz and De Luca, 2017](#)), or have sought to invert this relationship for inferring the full parameters from the base set ([Gaz et al., 2016](#)).

As a practical matter, mobile legged systems are often identified in restricted setups (e.g., with the torso fixed ([Bonnet et al., 2018](#); [Tournois et al., 2017](#); [Wensing et al., 2017a](#))), and so it is important to understand how these setups affect identifiability. When the accuracy of certain joint torques is affected by these restricted setups, it also has implications for which joints are best used for proprioceptive detection of external contacts ([Bledt et al., 2018b](#); [De Luca et al., 2006](#); [Haddadin et al., 2017](#)).

Considerations of structural identifiability likewise find applications in the area of adaptive control (e.g., [Pucci et al., 2015](#); [Slotine and Li, 1987](#); [Wang and Xie, 2012](#); [Zhu et al., 2023](#)), where so-called persistency of excitation (PE) conditions (i.e., ensuring maximal excitation over recurring finite intervals over time) can only be satisfied when adopting base parameters. These considerations hold regardless of whether one adopts a direct adaptive control strategy (e.g., [Chung and Slotine, 2009](#); [Garofalo et al., 2021](#); [Pucci et al., 2015](#); [Slotine and Li, 1987](#)), an indirect adaptive control strategy ([Li and Slotine, 1989](#)), or a composite of the two ([O'Connell et al., 2022](#); [Slotine and Li, 1989](#); [Wang, 2013](#)). More recently, less strict alternatives to PE conditions, known as interval excitation (IE) conditions, have been proposed ([Pan and Yu, 2018](#)) (see, e.g., [Wang et al. \(2023\)](#) for broader context beyond robot control). Again, however, IE conditions can only be satisfied when adopting a minimal parameterization.

1.1.2. Approaches. Previous methods for characterizing identifiability include symbolic and numerical approaches, with the methods herein primarily advancing the state-of-the-art for symbolic approaches.

Some symbolic approaches consider finding common groupings of parameters within the equations of motion ([Khalil et al., 1986](#); [Khosla, 1986, 1989](#)), which can become untenable at a system-level scale due to the complexity of the equations of motion. Thus, other work focused on carrying out regroupings of parameters by hand for special cases such as revolute manipulators with parallel or

perpendicular joints (Gautier and Khalil, 1990; Kawasaki et al., 1991; Mayeda et al., 1990). Many special cases still have to be considered separately, for example, whether joints are revolute or prismatic, parallel or orthogonal to each other, parallel or orthogonal to gravity, and combinations of the above (see Section 9.4 of the text by Khalil and Dombre (2002) for a summary, or an implementation in the SYMORO package (Khalil et al., 2014)). When regroupings are missed in the application of these rules, it leads one to believe that there are more base parameters than there truly are. Indeed, Khalil and Dombre (2002) state that the rules give “*in most cases* the exact number of base inertial parameters.” Further, the available formulas restrict to attaching frames via the Denavit–Hartenberg convention, which is limiting for practical application with model descriptions such as URDF that allow for any frame attachments. By comparison, the approach herein borrows inspiration from these methods but does not require any special cases or restricted frame attachments due to the generality of its geometric treatment of the problem.

On the flip side, numerical methods for assessing identifiability (e.g., Gautier, 1991) provide a lower bound on the number of base parameters of a model. Such methods often generate a finite set of random data, which is assumed to be maximally exciting. If the data is indeed maximally exciting, and there are no numerical issues, then correct conclusions can be drawn regarding the identifiability of the model. In cases when the data is not maximally exciting, the number of base parameters is underestimated. Since numeric methods provide a lower bound on the number of base parameters, while symbolic methods provide an upper bound, they can be combined together (Gautier, 1991) to mutually certify their outputs. Overall, neither the symbolic nor numeric approaches individually are provably correct for general mechanisms.

A complementary approach to finding common parameter groupings is to consider inconsequential/undetectable transfers of inertia between pairs of bodies across each joint. This strategy was developed originally by Niemeyer (1990) for use with floating-base systems with revolute and prismatic joints. It was later independently discovered by Chen and colleagues for 2D (Chen et al., 2002) and 3D (Chen and Beale, 2002) mechanisms, and developed further by Ros et al. (2012, 2015) and Iriarte et al.

(2013). These latter papers represent the state of the art in base parameter determination. However, much like original symbolic approaches (e.g., Gautier and Khalil, 1990) they require a skilled individual to exercise discretion in applying a set of special rules for determining identifiable parameters of links with motion restrictions close to the ground. By comparison, our algorithm is fully automatic. The user provides a model description (e.g., specifying the kinematic data found in a URDF file), and our algorithm provides a provably correct description of which parameter combinations are identifiable, and which are not.

The closest work in spirit is from Niemeyer and Slotine (1991) and Ayusawa et al. (2014), where they carried out provably correct identifiability analysis for floating-base robots. Ayusawa et al. (2014) treated general joint models, and also contributed a surprising result regarding the ability to identify floating-base models without measurement of joint torques. Beyond our extensions to fixed-base robots, we revisit the main theorem of Ayusawa et al. (2014) to provide a shorter proof from a new angle.

1.2. Contribution

The main contribution of this paper is the first provably correct algorithm to characterize the identifiable inertial parameter combinations for general fixed- and floating-base open-chain systems (Table 1). The algorithm is named the recursive parameter nullspace algorithm (RPNA), and it provides a modern update to past symbolic identifiability work through its applicability with generic frame assignments and general joint models beyond prismatic and revolute. Our geometric treatment of the problem, while more advanced in its derivation, allows the algorithm to achieve this generality in a compact final form that avoids complex special cases needed in past symbolic approaches.

The algorithm has a structure reminiscent of the outward kinematics pass of the RNEA. Rather than computing the velocities of each link on the outward pass, we geometrically characterize all possible velocities of each link. This information enables the algorithm to systematically detect motion restrictions for each body and to assess how they influence parameter identifiability without any special rules.

Table 1. Feature comparison.

	Type	General Joints	General Frames	Fixed Base	Floating Base	Provably Correct	Symbolic Output
Niemeyer (1990); Niemeyer and Slotine (1991)	Geometric		✓		✓	✓	✓
Ayusawa et al. (2014)	Symbolic	✓	✓		✓	✓	✓
Gautier and Khalil (1990); Khalil and Dombre (2002)	Symbolic			✓			✓
Gautier (1991)	Numeric	✓	✓	✓	✓		
Ros et al. (2015)	Symbolic		✓	✓			✓
This paper	Geometric	✓	✓	✓	✓	✓	✓

1.3. Organization

The paper is organized as follows. Section 2 introduces the main concepts of an inertial transfer and how it relates to identifiability. We then develop the identifiability theory by focusing on a pair of bodies connected via a joint, in cases where the combined bodies can move freely in 6D space (Section 3) versus when one body has additional motion restrictions (Section 4). We develop this theory in a parameterization-free sense, but rephrase the results in terms of inertial parameters in Section 5. These developments then enable a recursive treatment of identifiability in chains of bodies in Section 6, where we present the RPNA. Extensions of the basic algorithm to kinematics trees, multi-DoF joints, and floating-base systems are provided in Section 7. Section 8 provides a further extension to joint motors, an example of simple closed chains. Results in Section 9 consider system-level identifiability for classical manipulators, the PUMA and SCARA, as well as a mobile robot, the MIT Cheetah 3. As a key practical takeaway, we illustrate the pitfalls of constrained identification experiments often used to identify individual limbs of mobile legged systems. Concluding comments are provided in Section 10.

2. Main concepts

2.1. Dynamics and unidentifiable parameters

We consider identifying a rigid-body robot with N bodies whose joint-space dynamics take the standard form

$$\mathbf{H}(\mathbf{q})\ddot{\mathbf{q}} + \mathbf{c}(\mathbf{q}, \dot{\mathbf{q}}) + \mathbf{g}(\mathbf{q}) = \boldsymbol{\tau}_{\text{total}} \quad (1)$$

with \mathbf{q} the configuration variable, \mathbf{H} the mass matrix (also known as the joint-space inertia matrix), \mathbf{c} and \mathbf{g} the Coriolis and gravity forces, and $\boldsymbol{\tau}_{\text{total}}$ the total generalized

force summing contributions from joint torques and any external contact forces. It is well known that \mathbf{H} , \mathbf{c} , and \mathbf{g} can be expressed linearly in the inertial parameters $\boldsymbol{\pi} \in \mathbb{R}^{10N}$ of the bodies, which include masses, first moments, and rotational inertias. Thus, the dynamics (1) can be written as

$$\mathbf{Y}(\mathbf{q}, \dot{\mathbf{q}}, \ddot{\mathbf{q}}) \boldsymbol{\pi} = \boldsymbol{\tau}_{\text{total}} \quad (2)$$

where \mathbf{Y} is the classical regressor (Atkeson et al., 1986).

Unidentifiable parameters of the system are then given by those that don't affect the generalized force in any case

$$\mathcal{N} = \{ \delta\boldsymbol{\pi} \in \mathbb{R}^{10N} \mid \mathbf{Y}(\mathbf{q}, \dot{\mathbf{q}}, \ddot{\mathbf{q}}) \delta\boldsymbol{\pi} = \mathbf{0}, \forall \mathbf{q}, \dot{\mathbf{q}}, \ddot{\mathbf{q}} \}$$

For any parameter variations $\delta\boldsymbol{\pi} \in \mathcal{N}$, we say that the change $\delta\boldsymbol{\pi}$ does not affect the dynamics, or equivalently that it is not identifiable through measurement of the total generalized force $\boldsymbol{\tau}_{\text{total}}$. Likewise, $\delta\boldsymbol{\pi}$ is undetectable from the measurement of the joint torques and any external contact forces in this case.

Since the first two terms in (1) are determined by the kinetic energy, we will depart from these vector equations and instead focus on scalar energy $T = 1/2 \dot{\mathbf{q}}^\top \mathbf{H}(\mathbf{q}) \dot{\mathbf{q}}$. We temporarily dismiss potential energy before introducing gravity in Section 6.5 as an upward acceleration. Rewriting $T = \mathbf{Y}_T(\mathbf{q}, \dot{\mathbf{q}}) \boldsymbol{\pi}$, we can then equivalently express

$$\mathcal{N} = \{ \delta\boldsymbol{\pi} \in \mathbb{R}^{10N} \mid \mathbf{Y}_T(\mathbf{q}, \dot{\mathbf{q}}) \delta\boldsymbol{\pi} = 0, \forall \mathbf{q}, \dot{\mathbf{q}} \}$$

2.2. Identifiable parameters and combinations

Generally, any variation $\delta\boldsymbol{\pi}$ may affect a combination of parameters. For illustration, consider the $\boldsymbol{\pi} \in \mathbb{R}^3$ parameter space shown in Figure 1, decomposed into the unidentifiable subspace \mathcal{N} as well as the identifiable orthogonal subspace \mathcal{N}^\perp . Axis π_3

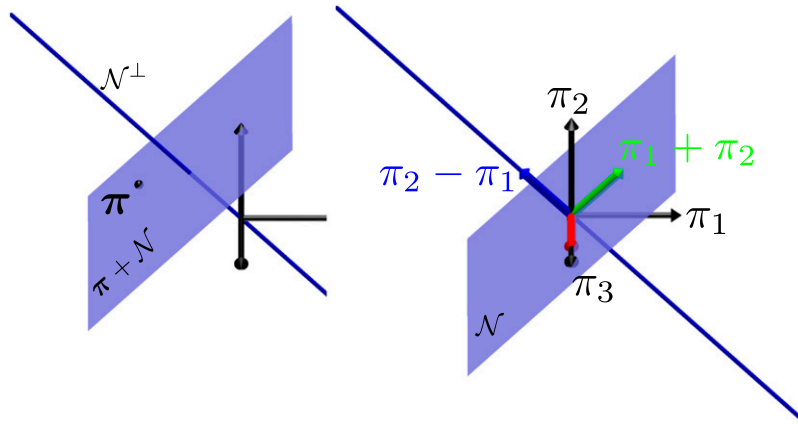


Figure 1. (left) Example 3D parameter space showing the nominal parameter values $\boldsymbol{\pi}$. Any parameter vectors in the affine (i.e., offset) subspace $\boldsymbol{\pi} + \mathcal{N}$ give the same model. (right) Variations $\delta\boldsymbol{\pi}$ split across unidentifiable \mathcal{N} (2D) and identifiable \mathcal{N}^\perp (1D) subspaces.

$$\begin{aligned} \pi_3 : \delta\boldsymbol{\pi} &= \begin{bmatrix} 0 & 0 & 1 & 0 & \dots \end{bmatrix} \in \mathcal{N} \\ (\pi_1 + \pi_2) : \delta\boldsymbol{\pi} &= \begin{bmatrix} 1 & 1 & 0 & 0 & \dots \end{bmatrix} \in \mathcal{N} \\ (\pi_2 - \pi_1) : \delta\boldsymbol{\pi} &= \begin{bmatrix} -1 & 1 & 0 & 0 & \dots \end{bmatrix} \in \mathcal{N}^\perp \end{aligned} .$$

lies within \mathcal{N} , making the parameter π_3 unidentifiable. But the effects of parameters π_1 and π_2 are combined: $(\pi_1 + \pi_2)$ is unidentifiable while $(\pi_2 - \pi_1)$ is identifiable. We can also equally imagine a π_4 axis within \mathcal{N}^\perp so that π_4 is identifiable by itself. So, any individual parameter may be identifiable in isolation, identifiable in combination, or unidentifiable.

Mathematically, if we can characterize $\mathcal{N} = \text{Null}(\mathbf{N})$ as the nullspace of some matrix \mathbf{N} , then the subspace \mathcal{N}^\perp orthogonal to \mathcal{N} is given by $\mathcal{N}^\perp = \text{Range}(\mathbf{N}^\top)$ such that the rows of \mathbf{N} give the identifiable parameter combinations.

2.3. Interpretation and parameter transfers

The set \mathcal{N} is a linear subspace of \mathbb{R}^{10N} , and any basis for it will generally consist of combinations of parameters from multiple bodies. In deriving a basis for \mathcal{N} , we will show it is enough to consider combinations of parameters from only two neighboring bodies at a time, and that these combinations represent a parameter transfer between the bodies.

Indeed, the approach derives two criteria on $\delta\pi$, which we interpret as two identification mechanisms: (1) how a body's momentum is projected on the preceding joint (and hence identifiable via joint torques) or (2) how a body's inertia combines with preceding (parent) bodies and whether it remains identifiable in the conglomeration of bodies. For the latter, we notice that the child body's parameters may be identified if they add to a parent differently, depending on joint configurations. But if a child parameter adds to the parent in a fixed manner, independent of configuration, it remains undetectable via (2).

Figure 2 illustrates this idea at a high level, showing three potential point masses to be attached to body 2. Each point mass affects the body's total mass, center of mass, and inertia and presents a physical interpretation of a particular $\delta\pi$. Adding either m_b or m_c ($\delta\pi_b$ or $\delta\pi_c$) affects the angular momentum seen on the joint (mechanism (1)), but, in the absence of gravity, the rotational joint alone can only determine their sum and not distinguish between them. However, considering the conglomeration of both bodies, we see m_b versus m_c at different locations with respect to body 1 based on the joint position, and hence can identify them distinctly (mechanism (2)). So, both $\delta\pi_b \in \mathcal{N}^\perp$ and $\delta\pi_c \in \mathcal{N}^\perp$. Meanwhile, m_a ($\delta\pi_a$) does not affect the angular momentum about the joint axis and adds to the previous body in a fixed manner/location, making it unidentifiable by itself. Indeed, we could equivalently rigidly attach m_a to the parent body without any effect on the dynamics. We consider this reassignment a transfer of $\delta\pi_a$ between the neighboring bodies. The combination of all such undetectable exchanges will be shown to span \mathcal{N} .

2.4. Spatial notation

To analyze the parameter combinations that do not affect the system dynamics, we will equivalently consider parameter changes that do not modify the kinetic energy. A simple

extension in Section 6 will address gravity. Effects of rotational and linear kinetic energy will be captured together using 6D spatial notation of Featherstone (2008) (see Appendix A for a short review). For example, the kinetic energy of a single body takes the form $T = 1/2 \mathbf{v}^\top \mathbf{I} \mathbf{v}$, with $\mathbf{v} \in \mathbb{R}^6$ its spatial velocity and $\mathbf{I} \in \mathbb{R}^{6 \times 6}$ its spatial inertia, given by

$$\mathbf{v} = \begin{bmatrix} \omega_x \\ \omega_y \\ \omega_z \\ v_x \\ v_y \\ v_z \end{bmatrix} \mathbf{I} = \begin{bmatrix} I_{xx} & I_{xy} & I_{xz} & 0 & -mc_z & mc_y \\ I_{xy} & I_{yy} & I_{yz} & mc_z & 0 & -mc_x \\ I_{xz} & I_{yz} & I_{zz} & -mc_y & mc_x & 0 \\ 0 & mc_z & -mc_y & m & 0 & 0 \\ -mc_z & 0 & mc_x & 0 & m & 0 \\ mc_y & -mc_x & 0 & 0 & 0 & m \end{bmatrix}$$

where $[\omega_x, \omega_y, \omega_z]^\top$ is the angular velocity of a body-fixed coordinate system, $[v_x, v_y, v_z]^\top$ the linear velocity of the origin of that system, m the body mass, $[c_x, c_y, c_z]^\top$ the CoM location in body coordinates, and I_{xx}, I_{yy} , etc. the mass moments and products of inertia about the body coordinate origin. For later use, the first moments are abbreviated as $h_x = mc_x$ for the x -axis and similarly for the others, while \mathbb{I} denotes the subspace of 6×6 matrices taking the above form. Rather than working with $T = 1/2 \dot{\mathbf{q}}^\top \mathbf{H}(\mathbf{q}) \dot{\mathbf{q}}$ in our analysis, we'll instead consider the system kinetic energy through $T = \sum_i 1/2 \mathbf{v}_i^\top \mathbf{I}_i \mathbf{v}_i$ where i sums over bodies.

Our preliminary development will consider the special case of rigid-body chains with bodies connected by single-degree-of-freedom joints. Bodies are numbered 1 to N from the base to the end of the chain with a frame attached to each body (Figure 3). Joint i connects body $i-1$ to body i , with its configuration noted by $q_i \in \mathbb{R}$. Under these definitions, the spatial velocities of neighboring bodies are related as

$$\mathbf{v}_i = {}^i\mathbf{X}_{i-1}(q_i) \mathbf{v}_{i-1} + \Phi_i \dot{q}_i \quad (3)$$

where ${}^i\mathbf{X}_{i-1} \in \mathbb{R}^{6 \times 6}$ is a spatial transform between frames and the vector $\Phi_i \in \mathbb{R}^{6 \times 1}$ describes the free motion for the joint. For instance, for a revolute joint about \hat{z}_i , $\Phi_i = [0 \ 0 \ 1 \ 0 \ 0 \ 0]^\top$.

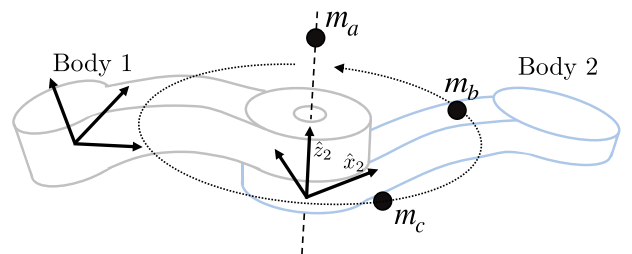


Figure 2. Additional masses m_a , m_b , and m_c are rigidly attached to body 2. The masses m_b and m_c are detectable by the joint torque and appear in differing locations w.r.t. body 1 based on the joint position. The mass m_a neither affects the angular momentum about the joint or joint torque, nor shifts location, and remains undetectable by itself.

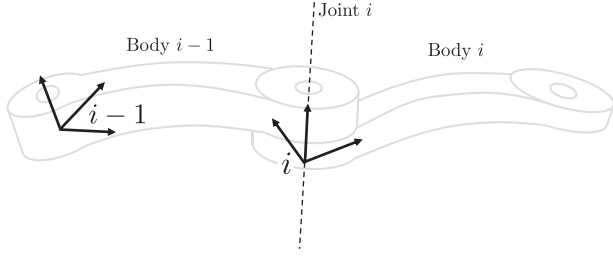


Figure 3. Example two-body system. Body $i-1$ can move freely, while body i is attached via a single-degree-of-freedom joint (revolute as shown). Frames $i-1$ and i are rigidly attached to bodies $i-1$ and i , respectively.

3. Two body case: General spatial motion

We begin by considering a single pair of bodies within the chain, and we use this simple case to mathematically develop the main ideas of the previous section. Consider bodies $i-1$ and i where body $i-1$ is able to move with any general spatial velocity. This case could occur, for example, if body $i-1$ were a floating base or if it were a body in a fixed base system that was far enough away from the base of the mechanism to experience all six spatial degrees of freedom in its movement.

3.1. Undetectable changes in inertia

We proceed to examine conditions under which the inertial parameters of these bodies can be changed without affecting the system's dynamics. To do so, we consider a collection of changes in inertias $\delta \mathbf{I}_{i-1}$ and $\delta \mathbf{I}_i$. For these changes to have no effect on the dynamics, they must not change the kinetic energy

$$\delta T = \frac{1}{2} \mathbf{v}_{i-1}^\top \delta \mathbf{I}_{i-1} \mathbf{v}_{i-1} + \frac{1}{2} \mathbf{v}_i^\top \delta \mathbf{I}_i \mathbf{v}_i = 0$$

Using (3), the variation δT can be factored as

$$\frac{1}{2} \begin{bmatrix} \mathbf{v}_{i-1} \\ \dot{q}_i \end{bmatrix}^\top \begin{bmatrix} \delta \mathbf{I}_{i-1} + {}^i \mathbf{X}_{i-1}^\top \delta \mathbf{I}_i {}^i \mathbf{X}_{i-1} & {}^i \mathbf{X}_{i-1}^\top \delta \mathbf{I}_i \Phi_i \\ \Phi_i^\top \delta \mathbf{I}_i {}^i \mathbf{X}_{i-1} & \Phi_i^\top \delta \mathbf{I}_i \Phi_i \end{bmatrix} \begin{bmatrix} \mathbf{v}_{i-1} \\ \dot{q}_i \end{bmatrix} \quad (4)$$

For the kinetic energy variation δT to always be zero, each entry of the above matrix must be zero. Considering the off-diagonal blocks, this condition requires

$$\mathbf{0} = \Phi_i^\top \delta \mathbf{I}_i \quad (5)$$

since ${}^i \mathbf{X}_{i-1}$ is full rank. The condition $\Phi_i^\top \delta \mathbf{I}_i \Phi_i = 0$ for the lower-right block is redundant with (5). For the upper-left block to be zero, it must be that

$$\mathbf{0} = \delta \mathbf{I}_{i-1} + {}^i \mathbf{X}_{i-1}^\top (q_i) \delta \mathbf{I}_i {}^i \mathbf{X}_{i-1}(q_i) \quad \forall q_i \quad (6)$$

Toward more general results later the manuscript, we note that since \mathbf{v}_{i-1} and \mathbf{v}_i can individually take any values, these two conditions are equivalent to

$$\mathbf{0} = \Phi_i^\top \delta \mathbf{I}_i \mathbf{v}_i \quad \forall \mathbf{v}_i \quad (7)$$

$$\mathbf{0} = \mathbf{v}_{i-1}^\top [\delta \mathbf{I}_{i-1} + {}^i \mathbf{X}_{i-1}^\top (q_i) \delta \mathbf{I}_i {}^i \mathbf{X}_{i-1}(q_i)] \mathbf{v}_{i-1} \quad \forall \mathbf{v}_{i-1}, q_i \quad (8)$$

The first condition (7) encodes that the change in inertia $\delta \mathbf{I}_i$ must not change the projection of body i 's momentum ($\mathbf{I}_i \mathbf{v}_i$) along the joint free mode Φ_i . For example, for a revolute joint about \hat{z}_i , the angular momentum about \hat{z}_i must not change due to $\delta \mathbf{I}_i$.

Returning to condition (6), we see the first appearance of a condition regarding an inertia transfer. The sum $\delta \mathbf{I}_{i-1} + {}^i \mathbf{X}_{i-1}^\top (q_i) \delta \mathbf{I}_i {}^i \mathbf{X}_{i-1}(q_i)$ represents the change in the total inertia of the two bodies combined, where ${}^i \mathbf{X}_{i-1}^\top (q_i) \delta \mathbf{I}_i {}^i \mathbf{X}_{i-1}(q_i)$ gives how $\delta \mathbf{I}_i$ maps back to its parent. As a result, (6) requires that the combination of $\delta \mathbf{I}_{i-1}$ and $\delta \mathbf{I}_i$ must represent an even exchange of inertia between the bodies for all joint angles. In the case when $q_i = 0$, (6) requires

$$\delta \mathbf{I}_{i-1} = -{}^i \mathbf{X}_{i-1}^\top (0) \delta \mathbf{I}_i {}^i \mathbf{X}_{i-1}(0) \quad (\text{Transfer Assign.}) \quad (9)$$

which encodes an assignment for a transfer of inertia between the bodies. Considering changes in configuration via a time derivative of (6) then requires

$$\mathbf{0} = \frac{d}{dt} ({}^i \mathbf{X}_{i-1}^\top (q_i) \delta \mathbf{I}_i {}^i \mathbf{X}_{i-1}(q_i)) \quad (10)$$

which enforces that the mapping of $\delta \mathbf{I}_i$ to body $i-1$ must not change over time. Consider the property from equation (2.45) of Featherstone (2008)

$$\frac{d}{dt} {}^i \mathbf{X}_{i-1}(q_i) = -(\Phi_i \dot{q}_i) \times {}^i \mathbf{X}_{i-1}(q_i) \quad (11)$$

where $(\mathbf{v}) \times \in \mathbb{R}^{6 \times 6}$ gives the spatial cross-product matrix with its form noted further in Appendix A. This property enables simplifying condition (10) to

$$\mathbf{0} = {}^i \mathbf{X}_{i-1}^\top [(\Phi_i \dot{q}_i \times)^\top \delta \mathbf{I}_i + \delta \mathbf{I}_i (\Phi_i \dot{q}_i \times)] {}^i \mathbf{X}_{i-1} \quad (12)$$

since ${}^i \mathbf{X}_{i-1}$ is full rank and \dot{q}_i can be chosen arbitrarily, (12) is equivalent to

$$(\Phi_i \times)^\top \delta \mathbf{I}_i + \delta \mathbf{I}_i (\Phi_i \times) = \mathbf{0} \quad (13)$$

3.2. Summary

In the case of a body that experiences general spatial motion, an exchange of inertia (9) with its child will not affect the dynamics if and only if it (1) does not modify the projection of the child body's momentum along the joint and (2) has a variation to the child inertia that maps to its parent in a constant way across configuration

$$\mathbf{0} = \Phi_i^\top \delta \mathbf{I}_i \quad (\text{Momentum}) \quad (14)$$

$$\mathbf{0} = (\Phi_i \times)^\top \delta \mathbf{I}_i + \delta \mathbf{I}_i (\Phi_i \times) \quad (\text{Mapping}) \quad (15)$$

We name these conditions the momentum and mapping conditions. Conditions (14) and (15) are equivalent to equation (18) of Niemeyer and Slotine (1991) (equiv., equations (4.13) and (4.14) of Niemeyer (1990)) and to equation (41) of Ayusawa et al. (2014).

Remark 1. Ayusawa et al. (2014) include an additional condition in their equation (42), which comes from considering Coriolis/centripetal effects. However, since these effects are determined by the kinetic energy, the additional conditions are redundant with those herein, enabling a more compact treatment.

Remark 2. Building on this simple case, we revisit the main result of Ayusawa et al. (2014) on parameter identifiability for floating-base systems with only external force/torque measurements in Section 7.4 following the extension of our theory to multi-DoF joints.

Returning to the high-level description of how parameters are identified, the momentum condition (14) enforces that the changes $\delta \mathbf{I}_i$ must not affect the local joint torque, where the mapping condition (15) enforces that the changes $\delta \mathbf{I}_i$ must not be detectable via how they are mapped to the parent.

3.3. Example: Revolute joint

This subsection gives an example of the momentum and mapping conditions with a revolute joint. We only consider parameter changes to body i , since the corresponding changes to body $i-1$ are determined with the transfer assignment (9). Consider a revolute joint about the local \hat{z}_i axis such that $\Phi_i = [0 \ 0 \ 1 \ 0 \ 0 \ 0]^\top$. In this case, the momentum condition (14) imposes

$$\delta h_x = \delta h_y = \delta I_{xz} = \delta I_{yz} = \delta I_{zz} = 0 \quad (16)$$

for the second link. The last three restrictions ($\delta I_{xz} = \delta I_{yz} = \delta I_{zz} = 0$) ensure that the angular momentum about the \hat{z}_i axis will remain unchanged for pure angular velocities of the body frame, while the first two $\delta h_x = \delta h_y = 0$ likewise ensure the same for linear velocities. The implication is that, for the second body, h_x , h_y , I_{xz} , I_{yz} , and I_{zz} are identifiable via the joint torque.

Likewise the mapping condition (15) imposes

$$\delta h_x = \delta h_y = \delta I_{xy} = \delta I_{xz} = \delta I_{yz} = \delta I_{xx} - \delta I_{yy} = 0 \quad (17)$$

to ensure that $\delta \mathbf{I}_i$ maps in a fixed way to the parent (i.e., that ${}^i\mathbf{X}_{i-1}^\top(q_i) \delta \mathbf{I}_i {}^i\mathbf{X}_{i-1}(q_i)$ is configuration invariant). Physically, these conditions are satisfied for any mass distribution that is symmetric about the \hat{z}_i axis. Comparing with (16), (17) implies that I_{xy} and $I_{xx} - I_{yy}$ are also identifiable, but via configuration dependence in the way these parameters are combined with the parent inertia.

To satisfy both the momentum and mapping conditions, $\delta \mathbf{I}_i$ is left with three degrees of freedom

$$\delta m, \delta h_z, \delta I_{xx} = \delta I_{yy}$$

Notation for this third freedom signifies that changes in I_{xx} must match those to I_{yy} . In this example, the inertia exchange has a physical interpretation. The transfer freedoms represent an exchange of any infinitely thin rod along \hat{z}_i . The point mass m_a in Figure 2 is a special case of such a rod.

In general, a rigid body attached to another by a revolute joint will add a maximum of seven parameters, in the absence of further motion restrictions.

4. Two bodies: Restricted spatial motion

Building on the previous section, we explore how additional spatial motion restrictions affect identifiability. Similar to the previous case, parameters of a body are identified via two mechanisms: (1) directly via how the parameters are sensed on the preceding joint or (2) indirectly via variations in how the parameters map to their parent body. However, if the parent has motion restrictions, some of its parameters may not appear in the equations of motion and would be unidentifiable via the generalized force. Thus, for a child parameter to use the second identification mechanism, variations in its mapping to the parent must appear on identifiable parameters of the parent. We develop these conditions mathematically and then work through examples to illustrate them.

4.1. Coupled conditions for $\delta T = 0$

In the previous development, body $i-1$ was assumed able to experience any velocity \mathbf{v}_{i-1} . This simplified analysis when reducing conditions (7) and (8) for $\delta T = 0$

$$0 = \Phi_i^\top \delta \mathbf{I}_i \mathbf{v}_i \quad \forall \mathbf{v}_i \quad (18)$$

$$0 = \mathbf{v}_{i-1}^\top [\delta \mathbf{I}_{i-1} + {}^i\mathbf{X}_{i-1}^\top(q_i) \delta \mathbf{I}_i {}^i\mathbf{X}_{i-1}(q_i)] \mathbf{v}_{i-1} \quad \forall \mathbf{v}_{i-1}, q_i \quad (19)$$

Now, instead, we consider these conditions in the context where \mathbf{v}_{i-1} does not freely take values in \mathbb{R}^6 .

4.2. Decoupling effects of inertia changes in the mapping condition

The condition (19) is impractical to verify as written due to its dependence on q_i and \mathbf{v}_{i-1} . It is also complicated by the fact that both $\delta \mathbf{I}_{i-1}$ and $\delta \mathbf{I}_i$ appear, which we address first.

For a given fixed \mathbf{v}_{i-1} , (19) must hold for all q_i . Equivalently, the expression on the right side of (19) must be zero at some point, and have zero derivative everywhere. We again consider the transfer assignment (9):

$$\delta \mathbf{I}_{i-1} = -{}^i \mathbf{X}_{i-1}^\top(0) \delta \mathbf{I}_i {}^i \mathbf{X}_{i-1}(0)$$

so that (19) holds at $q_i = 0$. Then, enforcing a zero derivative condition for (19) gives

$$\frac{d}{dq_i} \mathbf{v}_{i-1}^\top [{}^i \mathbf{X}_{i-1}^\top(q_i) \delta \mathbf{I}_i {}^i \mathbf{X}_{i-1}(q_i)] \mathbf{v}_{i-1} = 0 \quad (20)$$

which no longer has a dependence on $\delta \mathbf{I}_{i-1}$.

4.3. Simplification into a finite form

4.3.1. Addressing dependence on q_i . We can enforce a zero derivative everywhere by ensuring that all (first and higher-order) derivatives (w.r.t., q_i) are zero at $q_i = 0$. For the first derivative,

$$\begin{aligned} 0 &= \frac{d}{dq_i} \mathbf{v}_{i-1}^\top [\delta \mathbf{I}_{i-1} + {}^i \mathbf{X}_{i-1}^\top(q_i) \delta \mathbf{I}_i {}^i \mathbf{X}_{i-1}(q_i)] \mathbf{v}_{i-1} \Big|_{q_i=0} \\ &= \mathbf{v}_{i-1}^\top {}^i \mathbf{X}_{i-1}^\top(0) [(\Phi_i \times)^\top \delta \mathbf{I}_i + \delta \mathbf{I}_i (\Phi_i \times)] {}^i \mathbf{X}_{i-1}(0) \mathbf{v}_{i-1} \end{aligned}$$

while the analogous condition for the k th derivative w.r.t. q_i is equivalent to

$$\mathbf{v}_{i-1}^\top {}^i \mathbf{X}_{i-1}^\top(0) \delta \mathbf{I}_i^{(k)} {}^i \mathbf{X}_{i-1}(0) \mathbf{v}_{i-1} = 0 \quad (21)$$

where $\delta \mathbf{I}_i^{(0)} = \delta \mathbf{I}_i$ and

$$\delta \mathbf{I}_i^{(k+1)} = (\Phi_i \times)^\top \delta \mathbf{I}_i^{(k)} + \delta \mathbf{I}_i^{(k)} (\Phi_i \times)$$

The matrix $\delta \mathbf{I}_i^{(k)}$ captures the k th derivative in how $\delta \mathbf{I}_i$ maps to the parent. The mapping condition for the case of unrestricted motion is equivalent to $\delta \mathbf{I}_i^{(1)} = \mathbf{0}$. Each successive derivative is linear in the previous, and thus, for all derivatives after $k = 10$, the condition (21) can be guaranteed to be redundant via the Cayley–Hamilton theorem (Rugh, 1996).

4.3.2. Addressing velocity dependence. Repeating the momentum condition, the previous development is rewritten as

$$\mathbf{0} = \Phi_i^\top \delta \mathbf{I}_i \mathbf{v}_i \quad \forall \mathbf{v}_i \quad (22)$$

$$\mathbf{0} = \mathbf{v}_{i-1}^\top [{}^i \mathbf{X}_{i-1}^\top(0) \delta \mathbf{I}_i^{(k)} {}^i \mathbf{X}_{i-1}(0)] \mathbf{v}_{i-1} \quad \forall \mathbf{v}_{i-1} \quad (23)$$

$$\forall k = 1, \dots, 10$$

Both conditions remain impractical to verify as written since they give an infinite number of constraints. To simplify them, we introduce spans and create an equivalent finite set of conditions. This step provides the key to our contribution. We address each condition separately.

4.3.3. Linear velocity span for momentum condition (22). First, let the set of possible velocities for \mathbf{v}_{i-1} be denoted by \mathcal{V}_{i-1}^* . The set of possible velocities for body i is

$$\mathcal{V}_i^* = \{ {}^i \mathbf{X}_{i-1}(q_i) \mathbf{v}_{i-1} + \Phi_i \dot{q}_i \mid \mathbf{v}_{i-1} \in \mathcal{V}_{i-1}^*, q_i, \dot{q}_i \in \mathbb{R} \} \quad (24)$$

Then, consider the following linear velocity span \mathcal{V}_i , which is subspace of \mathbb{R}^6 and characterizes the motion restrictions:

$$\mathcal{V}_i = \text{span}\{ \mathbf{v} \mid \mathbf{v} \in \mathcal{V}_i^* \} \quad (25)$$

Remark 3. Note that $\mathcal{V}_i^* \subseteq \mathcal{V}_i$, since not all elements of \mathcal{V}_i need be attainable velocities. Rather, all elements of \mathcal{V}_i can be expressed as the linear combination of some attainable velocities.

Since the set \mathcal{V}_i is a finite-dimensional subspace, we consider a set of basis vectors

$$\mathcal{V}_i = \text{span}\{ \mathbf{v}_{i,1}, \dots, \mathbf{v}_{i,n_i^v} \} \quad (26)$$

where we collect the basis elements for \mathcal{V}_i into a matrix $\mathbf{V}_i = [\mathbf{v}_{i,1}, \dots, \mathbf{v}_{i,n_i^v}]$ so that $\text{Range}(\mathbf{V}_i) = \mathcal{V}_i$. We will later provide methods to compute \mathbf{V}_i . With this, the infinite set of conditions (22) are equivalent to the finite set of conditions

$$\mathbf{0} = \Phi_i^\top \delta \mathbf{I}_i \mathbf{v}_{i,\ell}, \forall \ell = 1, \dots, n_i^v \quad (27)$$

which can easily be collected into the vector equation

$$\mathbf{0} = \Phi_i^\top \delta \mathbf{I}_i \mathbf{V}_i \quad (28)$$

4.3.4. “Quadratic” velocity span for mapping condition (23). Unfortunately, the mapping condition depends on quadratic velocity terms. We employ the Trace () operator toward creating a “quadratic” velocity span. Fortunately, leveraging the linear inertial parameterization in the next section will enable simplifying this step.

Using Trace(), we rewrite the mapping condition (23)

$$\begin{aligned} 0 &= \text{Trace} \left(\mathbf{v}_{i-1} \mathbf{v}_{i-1}^\top [{}^i \mathbf{X}_{i-1}^\top(0) \delta \mathbf{I}_i^{(k)} {}^i \mathbf{X}_{i-1}(0)] \right) \\ &\quad \forall \mathbf{v}_{i-1} \in \mathcal{V}_{i-1}^*, \forall k = 1, \dots, 10 \end{aligned} \quad (29)$$

and create the “quadratic” velocity span

$$\mathcal{W}_{i-1} = \text{span}\{ \mathbf{v} \mathbf{v}^\top \mid \mathbf{v} \in \mathcal{V}_{i-1}^* \} \quad (30)$$

where \mathcal{W}_{i-1} is a finite-dimensional subspace of $\mathbb{R}^{6 \times 6}$. We also consider a set of basis elements

$$\mathcal{W}_{i-1} = \text{span}\{ \mathbf{W}_{i-1,1}, \dots, \mathbf{W}_{i-1,n_{i-1}^w} \} \quad (31)$$

for converting the mapping condition into the finite set of conditions

$$0 = \text{Trace} \left(\mathbf{W}_{i-1,j} \left[{}^i\mathbf{X}_{i-1}^\top(0) \delta \mathbf{I}_i^{(k)} {}^i\mathbf{X}_{i-1}(0) \right] \right) \quad (32)$$

$$\forall k = 1, \dots, 10, \quad j = 1, \dots, n_{i-1}^w$$

Unlike (27), the iteration over all basis elements cannot be collected into a vector equation. Section 5 will avoid this issue.

Again, physically, (28) encodes that $\delta \mathbf{I}_i$ must not change the components of body i 's momentum associated with the joint Φ_i , while (32) encodes that $\delta \mathbf{I}_i$ maps in a fixed way onto the identifiable parameters of the parent.

4.4. Main theorem

Although we've developed this theory for a pair of bodies, an inductive argument (in the [appendix](#)) shows that we can consider the momentum and mapping conditions on inertia transfers joint-by-joint in chains of bodies to build out all unidentifiable changes to the inertia.

Theorem 1. (Main Result) Consider a serial-chain rigid-body system in the absence of gravity, with the following inertia transfer subspaces for each joint ($i \in \{1, \dots, N\}$):

$$\begin{aligned} \overline{\mathcal{T}}_i &= \{ \delta \mathbf{I}_1, \dots, \delta \mathbf{I}_N \in \mathbb{I} \mid \exists \delta \mathbf{I}_0 \in \mathbb{I}, \delta \mathbf{I}_j = \mathbf{0} \text{ if } j \notin \{i, i-1\}, \\ \delta \mathbf{I}_{i-1} &= -{}^i\mathbf{X}_{i-1}^\top(0) \delta \mathbf{I}_i {}^i\mathbf{X}_{i-1}(0), \\ \mathbf{0} &= \Phi_i^\top \delta \mathbf{I}_i \mathbf{V}_i, \\ \mathbf{0} &= \text{Trace} \left(\mathbf{W}_{i-1,j} {}^i\mathbf{X}_{i-1}^\top(0) \delta \mathbf{I}_i^{(k)} {}^i\mathbf{X}_{i-1}(0) \right) \\ &\quad \forall k = 1, \dots, 10, \quad j = 1, \dots, n_{i-1}^w \} \end{aligned}$$

Then, the set of all structurally unobservable inertia changes is given by $\overline{\mathcal{T}}_1 \oplus \dots \oplus \overline{\mathcal{T}}_N$, where \oplus denotes a direct sum of vector subspaces.

Proof: See [Appendix C](#).

We proceed first with a set of examples before focusing on making this result amenable to practical computation with inertial parameters in the subsequent sections.

4.5. Example

We work through these conditions within the context of the two 2R manipulators shown in [Figure 4](#).

4.5.1. Simple system with parallel joint axes. The system in [Figure 4\(a\)](#) is planar, with the spatial velocities of its bodies and a basis matrix \mathbf{V}_2 given by

$$\mathbf{v}_1 = \begin{bmatrix} 0 \\ 0 \\ \dot{q}_1 \\ 0 \\ 0 \\ 0 \end{bmatrix} \quad \mathbf{v}_2 = \begin{bmatrix} 0 \\ 0 \\ \dot{q}_1 + \dot{q}_2 \\ \ell s_2 \dot{q}_1 \\ \ell c_2 \dot{q}_1 \\ 0 \end{bmatrix} \quad \mathbf{V}_2 = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix}$$

where $c_2 = \cos(q_2)$ and $s_2 = \sin(q_2)$. Note that the velocity span for body 2 includes linear velocities in both the x and y

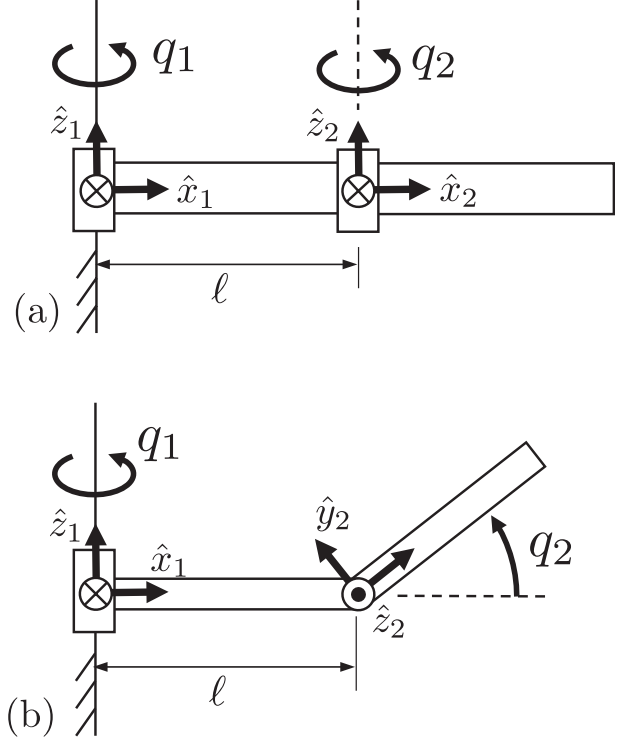


Figure 4. Two simple RR manipulators.

directions, since effects from \dot{q}_1 will be in the \hat{x}_2 or \hat{y}_2 direction depending on the value of q_2 . Considering the span for $\mathbf{v}_1 \mathbf{v}_1^\top$, we likewise have a basis for \mathcal{W}_1 consisting of the single element:

$$\mathbf{W}_{1,1} = \Phi_1 \Phi_1^\top = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

which has a 1 in the slot of the matrix corresponding to I_{zz} (the only identifiable parameter for body 1).

The body 1 momentum condition $\mathbf{0} = \Phi_1^\top \delta \mathbf{I}_1 \mathbf{V}_1$ imposes

$$\delta I_{zz_1} = 0$$

The body 1 mapping condition is empty as \mathcal{W}_0 is empty—any parameters transferred to ground are unidentifiable. The body 2 momentum condition $\mathbf{0} = \Phi_2^\top \delta \mathbf{I}_2 \mathbf{V}_2$ imposes

$$\delta h_{x_2} = \delta h_{y_2} = \delta I_{zz_2} = 0$$

and the body 2 mapping condition

$$0 = \text{Trace}(\mathbf{W}_{1,1} {}^2\mathbf{X}_1^\top(0) \delta \mathbf{I}_2^{(k)} {}^2\mathbf{X}_1(0))$$

or, equivalently, using $\mathbf{W}_{1,1} = \Phi_1 \Phi_1^\top$, written as

$$0 = \Phi_1^\top {}^2\mathbf{X}_1^\top(0) \delta \mathbf{I}_2^{(k)} {}^2\mathbf{X}_1(0) \Phi_1$$

imposes

$$\begin{aligned} \delta h_{y_2} &= 0 & \text{when } k = 1 \\ \delta h_{x_2} &= 0 & \text{when } k = 2 \end{aligned}$$

The constraints are redundant for all k higher. Thus, we find four base parameters: I_{zz_1} for body 1 and h_{x_2} , h_{y_2} , and I_{zz_2} for body 2.

For reference, consider the mass matrix $\mathbf{H}(\mathbf{q})$

$$\begin{bmatrix} I_{zz_1}^\star + I_{zz_2} + 2\ell h_{x_2}c_2 - 2\ell h_{y_2}s_2 & I_{zz_2} + \ell h_{x_2}c_2 - \ell h_{y_2}s_2 \\ I_{zz_2} + \ell h_{x_2}c_2 - \ell h_{y_2}s_2 & I_{zz_2} \end{bmatrix}$$

where $I_{zz_1}^\star = I_{zz_1} + m_2\ell^2$. We see the expected four parameter groupings: $I_{zz_1} + m_2\ell^2$, I_{zz_2} , h_{x_2} , h_{y_2} . Note the effect of m_2 is transferred from body 2 and combined with I_{zz_1} .

In more detail, body 1 experiences only a pure rotation, so the body 1 momentum condition shows we can only identify the rotational inertia about the joint axis. Body 2 rotates in the same plane but may also translate. So the body 2 momentum condition also considers the torque needed to linearly move the center of mass (h_{x_2} , h_{y_2} depending on q_2). The mapping condition also certifies the identifiability of h_{x_2} , h_{y_2} , since they map inconsistently to body 1 (compare to m_b , m_c in Figure 2). But m_2 itself maps to body 1 in a fixed manner (compare to m_a in Figure 2). Overall, seven parameters of body 2 are not identifiable individually, and can be transferred to body 1.

4.5.2. Simple system with perpendicular joint axes. For the system in Figure 4(b), the second body can move out of the \hat{x}_1, \hat{y}_1 plane, resulting in additional motion freedom. A full derivation for this example is in Appendix B, with main results summarized here. The body 2 momentum condition provides

$$\delta I_{xz_2} = \delta I_{yz_2} = \delta I_{zz_2} = 0 \quad (33)$$

which implies that I_{xz_2} , I_{yz_2} , and I_{zz_2} can be identified through the second joint torque. The body 2 mapping condition requires

$$\delta h_{x_2} = \delta h_{y_2} = \delta I_{xx_2} - \delta I_{yy_2} = \delta I_{xy_2} = 0 \quad (34)$$

which implies that h_{x_2} , h_{y_2} , $I_{xx_2} - I_{yy_2}$, and I_{xy_2} can be identified via the first joint torque. Note again, due to motion restrictions, conditions (33) and (34) represent a subset of those in the free-floating case (16) and (17), respectively. However, unlike the previous manipulator with parallel joints, the seven conditions from (33) and (34) are independent. As a result, the unobservable transfers across joint 2 have three degrees of freedom, which must coincide with those in the free-floating case. This gives eight base

parameters for the mechanism (one from body 1 and seven from body 2).

5. Momentum/mapping conditions in terms of inertial parameters

Up until this point, we have phrased the momentum and mapping conditions in terms of changes to the inertia matrices. In this section, we instead translate these conditions to ones on the inertial parameters, relying heavily on the linearity of the inertia matrices in the inertial parameters of each body:

$$\boldsymbol{\pi}_i = [m, h_x, h_y, h_z, I_{xx}, I_{xy}, I_{xz}, I_{yy}, I_{yz}, I_{zz}]^\top \quad (35)$$

We switch from the matrix form of the spatial inertia to the parameter vector via the notation $\boldsymbol{\pi}_i = [\mathbf{I}_i]^\vee$ where the vee \vee demotes an inertia to a parameter vector.

5.1. Momentum condition

To express the momentum condition $\Phi_i^\top \delta \mathbf{I}_i \mathbf{V}_i = \mathbf{0}$ from (28) using inertial parameters, we temporarily consider $\boldsymbol{\pi} \in \mathbb{R}^{10}$ for a single body with and define the matrix:

$$\mathbf{M}(\mathbf{V}, \boldsymbol{\Phi}) = \left[\frac{\partial}{\partial \boldsymbol{\pi}} \mathbf{V}^\top \mathbf{I}(\boldsymbol{\pi}) \boldsymbol{\Phi} \right]^\top \quad (36)$$

Then, inertia variations $\delta \mathbf{I}_i$ satisfy the momentum condition if and only if the corresponding parameter changes $\delta \boldsymbol{\pi}_i$ satisfy $\mathbf{M}(\mathbf{V}_i, \boldsymbol{\Phi}_i)^\top \delta \boldsymbol{\pi}_i = \mathbf{0}$.

5.2. Mapping condition

The earlier version of the mapping condition from (23) is recalled below:

$$0 = \mathbf{v}_{i-1}^\top \left[{}^i\mathbf{X}_{i-1}^\top(0) \delta \mathbf{I}_i^{(k)} {}^i\mathbf{X}_{i-1}(0) \right] \mathbf{v}_{i-1} \quad \forall \mathbf{v}_{i-1} \quad (37)$$

$$\forall k = 1, \dots, 10$$

which we favor over (32) to simplify our development in this section.

Since this condition is a bit more complicated, we will work on it from the outside moving in, making use of three separate helper functions:

$$\mathbf{k}(\mathbf{v}) := \nabla_{\boldsymbol{\pi}} \frac{1}{2} \mathbf{v}^\top \mathbf{I}(\boldsymbol{\pi}) \mathbf{v} \quad (38)$$

$$\mathbf{B}(\mathbf{X}) := \frac{\partial}{\partial \boldsymbol{\pi}} [\mathbf{X}^\top \mathbf{I}(\boldsymbol{\pi}) \mathbf{X}]^\vee \quad (39)$$

$$\mathbf{A}(\boldsymbol{\Phi}) := \frac{\partial}{\partial \boldsymbol{\pi}} [(\boldsymbol{\Phi} \times)^\top \mathbf{I}(\boldsymbol{\pi}) + \mathbf{I}(\boldsymbol{\pi}) (\boldsymbol{\Phi} \times)]^\vee \quad (40)$$

which we now use in three separate simplifying stages.

For the first stage, we note that $\mathbf{k}(\mathbf{v}_i)$ gives the gradient of the kinetic energy for the i th body w.r.t. its

inertia parameters. In this regard, if $\mathbf{k}(\mathbf{v})^\top \delta \boldsymbol{\pi}_i = 0$ for all attainable velocities $\mathbf{v} \in \mathcal{V}_i^*$, the linear combination of parameters given via $\delta \boldsymbol{\pi}_i$ does not appear in the kinetic energy of the mechanism (i.e., that combination is unidentifiable).

With this motivation, consider the span of the vectors $\mathbf{k}(\mathbf{v}_i)$ over all attainable velocities for body i :

$$\mathcal{K}_i = \text{span}\{\mathbf{k}(\mathbf{v}) \mid \mathbf{v} \in \mathcal{V}_i^*\}$$

Analogous to the previous span \mathcal{V}_i , \mathcal{K}_i is a *vector* subspace of \mathbb{R}^{10} and thus has a finite basis representation that we set via selecting any matrix \mathbf{K}_i with $\text{Range}(\mathbf{K}_i) = \mathcal{K}_i$. A change $\delta \boldsymbol{\pi}_i$ to body i is unidentifiable if and only if $\mathbf{K}_i^\top \delta \boldsymbol{\pi}_i = \mathbf{0}$. Turning this around, the columns of \mathbf{K}_i form a basis for the coefficients of the inertial parameters of body i that are identifiable themselves or in combination with other bodies.

Using this matrix, we provide equivalent conditions for the form of the mapping condition in (37) as

$$\mathbf{0} = \mathbf{K}_{i-1}^\top \left[{}^i\mathbf{X}_{i-1}^\top(0) \delta \mathbf{I}_i^{(k)} {}^i\mathbf{X}_{i-1}(0) \right]^\vee \quad \forall k = 1, \dots, 10 \quad (41)$$

which enforces that $\delta \mathbf{I}_i$ maps in a fixed way onto the identifiable parameters of the parent.

Remark 4. To connect this development to the previous section, we could alternatively compute \mathcal{K}_i via the “quadratic” velocity span \mathcal{W}_i as

$$\mathcal{K}_i = \text{span}\{\nabla_{\boldsymbol{\pi}} \text{Trace}(\mathbf{W} \mathbf{I}(\boldsymbol{\pi})) \mid \mathbf{W} \in \mathcal{W}_i\}$$

Proceeding to our second stage of simplifications, consider the 10×10 parameter transformation matrix

$${}^{i-1}\mathbf{B}_i = \mathbf{B}({}^i\mathbf{X}_{i-1}(0)) \quad (42)$$

that maps parameters to the predecessor. With this definition, we re-express the transfer assignment (9) as

$$\delta \boldsymbol{\pi}_{i-1} = -{}^{i-1}\mathbf{B}_i \delta \boldsymbol{\pi}_i \quad (43)$$

Via this matrix, (41) is equivalent to

$$\mathbf{K}_{i-1}^\top {}^{i-1}\mathbf{B}_i [\delta \mathbf{I}_i^{(k)}]^\vee = \mathbf{0} \quad \forall k = 1, \dots, 10 \quad (44)$$

As our last stage of simplification, we use the final helper function $\mathbf{A}(\boldsymbol{\Phi})$ from (40), which gives the rate of change in how parameters map to the parent with changes in the joint angle. It follows from the definition of $\mathbf{A}(\boldsymbol{\Phi})$ that $[\delta \mathbf{I}^{(k)}]^\vee = \mathbf{A}(\boldsymbol{\Phi}_i)^k \boldsymbol{\pi}_i$. This property finally enables re-factoring of (44) as

$$\mathbf{K}_{i-1}^\top {}^{i-1}\mathbf{B}_i \mathbf{A}(\boldsymbol{\Phi}_i)^k \delta \boldsymbol{\pi}_i = \mathbf{0} \quad \forall k = 1, \dots, 10 \quad (45)$$

5.2.1. Summary. Inertia transfers (43) across any joint i are unobservable to the kinetic energy if they satisfy

$$\mathbf{0} = \mathbf{M}(\mathbf{V}_i, \boldsymbol{\Phi}_i)^\top \delta \boldsymbol{\pi}_i \quad (\text{Momentum}) \quad (46)$$

$$\mathbf{0} = \mathbf{K}_{i-1}^\top {}^{i-1}\mathbf{B}_i \mathbf{A}(\boldsymbol{\Phi}_i)^k \delta \boldsymbol{\pi}_i \quad (\text{Mapping}) \quad (47)$$

$$\forall k = 1, \dots, 10$$

6. Kinematic chains of bodies: Recursion, theorem, and algorithm

With the results of the previous sections in place, a remaining hurdle is how to compute bases for the attainable velocity spans \mathcal{V}_i and identifiable parameter spans \mathcal{K}_i . In this section, recursive application of controllability and observability from linear systems theory is shown to play the key role needed. We develop these steps, and then present our main algorithm. While the focus remains on serial chains, extensions to tree-structure systems and multi-DoF joints are presented in Section 7.

6.1. Velocity spans

Bases \mathbf{V}_i for the velocity spans can be computed starting from $\mathbf{V}_0 = \mathbf{0}_{6 \times 1}$ and proceeding outward, as in Figure 5.

Lemma 1. Suppose a matrix \mathbf{V}_{i-1} such that $\mathcal{V}_{i-1} = \text{Range}(\mathbf{V}_{i-1})$. Consider the quantities

$$\begin{aligned} \mathbf{V}_{i-1+} &= {}^i\mathbf{X}_{i-1}(0) \mathbf{V}_{i-1} \\ \mathbf{V}_i &= \text{Ctrb}((\boldsymbol{\Phi}_i \times), \mathbf{V}_{i-1+}) \\ &= [\mathbf{V}_{i-1+}, (\boldsymbol{\Phi}_i \times) \mathbf{V}_{i-1+}, \dots, (\boldsymbol{\Phi}_i \times)^5 \mathbf{V}_{i-1+}] \end{aligned}$$

where $\text{Ctrb}((\boldsymbol{\Phi}_i \times), \mathbf{V}_{i-1+})$ is the controllability matrix (Rugh, 1996) associated with the pair $((\boldsymbol{\Phi}_i \times), \mathbf{V}_{i-1+})$. Then, the matrix

$$\mathbf{V}_i = [\mathbf{V}_i \quad \boldsymbol{\Phi}_i] \quad (48)$$

satisfies $\mathcal{V}_i = \text{Range}(\mathbf{V}_i)$,

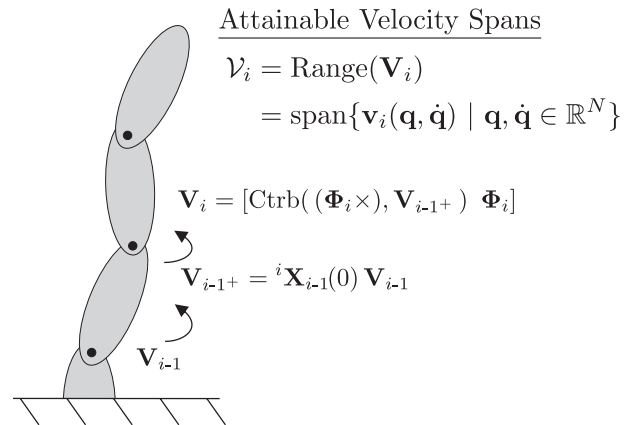


Figure 5. Controllability analysis is applied recursively to obtain a basis \mathbf{V}_i for the span of the attainable velocities for each body, starting with $\mathbf{V}_0 = \mathbf{0}_{6 \times 1}$.

Proof: See [Appendix D](#).

To provide intuition into this result, \mathbf{V}_{i-1+} first transforms \mathbf{V}_{i-1} across the link. Then $\mathbf{V}_{i-} = \text{Ctrb}((\Phi_i \times), \mathbf{V}_{i-1+})$ captures all possible velocity effects from the predecessor following transformation across the joint. Finally, the second term Φ_i in (48) adds in relative velocities from the joint itself.

6.2. Identifiable parameter coefficient spans

Using the previous helper function definitions, \mathbf{K}_i can be computed recursively, starting from $\mathbf{K}_0 = \mathbf{0}_{10 \times 1}$ and propagating outward via the following lemma.

Lemma 2. Suppose a matrix \mathbf{K}_{i-1} such that $\mathcal{K}_{i-1} = \text{Range}(\mathbf{K}_{i-1})$. Consider the quantities

$$\begin{aligned} \mathbf{K}_{i-1+} &= {}^{i-1}\mathbf{B}_i^\top \mathbf{K}_{i-1} \\ \mathbf{K}_{i-} &= \text{Ctrb}(\mathbf{A}(\Phi_i)^\top, \mathbf{K}_{i-1+}) \\ &= [\mathbf{K}_{i-1+}, \mathbf{A}(\Phi_i)^\top \mathbf{K}_{i-1+}, \dots, \mathbf{A}(\Phi_i)^9 \mathbf{K}_{i-1+}] \end{aligned} \quad (49)$$

Then, the matrix

$$\mathbf{K}_i = [\mathbf{K}_{i-} \quad \mathbf{M}(\mathbf{V}_i, \Phi_i)]$$

satisfies $\mathcal{K}_i = \text{Range}(\mathbf{K}_i)$.

Proof: See [Appendix E](#), which proves the equivalent characterization in Remark 5 below.

Similar to before, $\mathbf{K}_{i-1+} = {}^{i-1}\mathbf{B}_i^\top \mathbf{K}_{i-1}$ transforms the identifiable parameters of body $i-1$ across the link, while $\mathbf{K}_{i-} = \text{Ctrb}(\mathbf{A}(\Phi_i)^\top, \mathbf{K}_{i-1+})$ then captures all possible transformations across the joint. Finally, the extra columns $\mathbf{M}(\mathbf{V}_i, \Phi_i)$ add on additional parameters for body i that can be identified from measuring the torque at joint i .

As an additional outcome, the mapping condition (45) can be written as

$$\mathbf{K}_{i-}^\top \mathbf{A}(\Phi_i) \delta \pi_i = \mathbf{0}$$

Remark 5. Due to the duality between controllability and observability ([Rugh, 1996](#)), we can equivalently view Lemma 2 as characterizing the identifiable parameter combinations via an observability matrix:

$$\mathbf{K}_i^\top = \begin{bmatrix} \text{Obs}(\mathbf{K}_{i-1+}^\top, \mathbf{A}(\Phi_i)) \\ \mathbf{M}(\mathbf{V}_i, \Phi_i)^\top \end{bmatrix}$$

where $\text{Obs}(\mathbf{K}_{i-1+}^\top, \mathbf{A}(\Phi_i))$ gives the observability matrix for the pair $(\mathbf{K}_{i-1+}^\top, \mathbf{A}(\Phi_i))$.

6.3. Summary

An inertia transfer $\delta \pi_{i-1} = -{}^{i-1}\mathbf{B}_i \delta \pi_i$ across joint i is unobservable to the kinetic energy if it satisfies

$$\mathbf{M}(\mathbf{V}_i, \Phi_i)^\top \delta \pi_i = \mathbf{0} \quad (\text{Momentum}) \quad (50)$$

$$\mathbf{K}_{i-}^\top \mathbf{A}(\Phi_i) \delta \pi_i = \mathbf{0} \quad (\text{Mapping}) \quad (51)$$

A summary of the steps leading to these final conditions is provided in [Figure 6](#).

Both conditions can further be written together using a transfer nullspace descriptor \mathbf{N}_i given by

$$\mathbf{N}_i = \begin{bmatrix} \mathbf{M}(\mathbf{V}_i, \Phi_i)^\top \\ \mathbf{K}_{i-}^\top \mathbf{A}(\Phi_i) \end{bmatrix}$$

In summary, Lemmas 1 and 2 provide the main recursive steps for computing the spans \mathcal{V}_i and \mathcal{K}_i , with the transfer nullspace descriptors \mathbf{N}_i then collecting the momentum and mapping conditions together in the single equation $\mathbf{N}_i \delta \pi_i = \mathbf{0}$.

6.4. Theorem

Theorem 2. (Main Result, Restated) Consider a serial-chain rigid-body system in the absence of gravity, with the

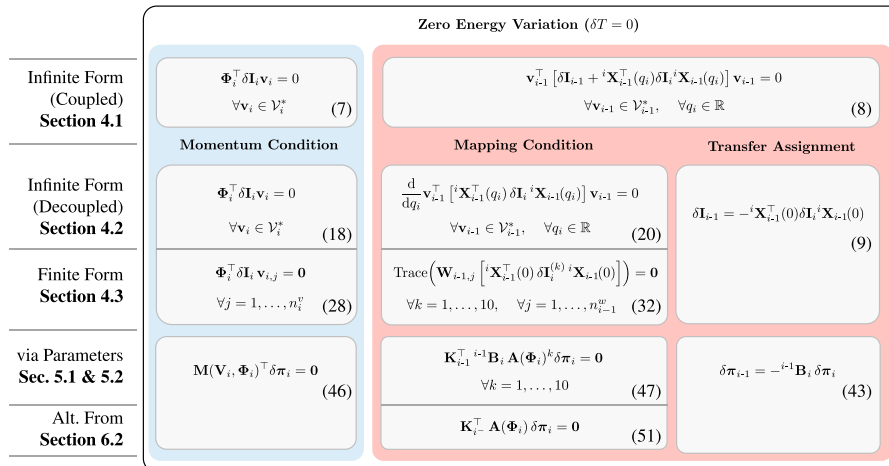


Figure 6. Summary of conditions for a parameter transfer at joint i to be unobservable to the kinetic energy.

following inertia transfer subspaces for each joint ($i \in \{1, \dots, N\}$):

$$\mathcal{T}_i = \{ \delta \boldsymbol{\pi} \in \mathbb{R}^{10N} \mid \begin{aligned} \delta \boldsymbol{\pi}_0 &\in \mathbb{R}^{10}, \delta \boldsymbol{\pi}_{i-1} = -{}^{i-1}\mathbf{B}_i \delta \boldsymbol{\pi}_i, \\ \mathbf{0} &= \mathbf{N}_i \delta \boldsymbol{\pi}_i \\ \mathbf{0} &= \delta \boldsymbol{\pi}_j \text{ if } j \notin \{i, i-1\} \end{aligned} \}$$

Then, the structurally unobservable parameter subspace \mathcal{N} satisfies $\mathcal{N} = \mathcal{T}_1 \oplus \dots \oplus \mathcal{T}_N$.

Proof: The proof follows from the same logic as Theorem 1.

Remark 6. The above theorem only applies to the gravity-free case. In fixed-base robots, gravitational forces provide opportunity to identify additional mechanism parameters, decreasing the dimensionality of \mathcal{N} . The following section provides a simple method to address these effects within a recursive algorithm.

6.5. Addressing gravity

Within rigid-body dynamics algorithms, effects of gravity are often addressed by fictitiously accelerating the base opposite gravity. This trick is applied in the recursive-Newton–Euler algorithm (Luh et al., 1980) for inverse dynamics and the articulated-body algorithm for forward dynamics (Featherstone, 2008). The same approach also works to address gravitational effects for identifiability. By seeding $\mathbf{V}_0 = {}^0\mathbf{a}_g$, with ${}^0\mathbf{a}_g$ the gravity acceleration in the world coordinate, the recursive computations of this section result in modified transfer subspaces \mathcal{T}_i that include gravitational considerations. Intuitively, this modification corresponds to adding a fictitious prismatic joint aligned with gravity at the base whose force is not measured. Appendix F rigorously analyzes the role of gravity on identifiability and further justifies this simple modification.

6.6. Algorithm summary

Algorithm 1 provides a compact method to recursively compute the parameter nullspace descriptors \mathbf{N}_i . We name this method the recursive parameter nullspace algorithm (RPNA). As a practical matter, linearly dependent columns of \mathbf{V}_i or \mathbf{K}_i can be removed at any step in the algorithm. A MATLAB implementation of the RPNA is provided open source (Wensing, 2024a). Unlike pure numeric methods, the algorithm can be run with either numeric or symbolic values for the frame transformations ${}^i\mathbf{X}_{i-1}(0)$. When symbolic values are used, the RPNA returns a symbolic output as a function of mechanism kinematic parameters. Compared to previous symbolic approaches, while the derivation strategy herein employed more advanced mathematical tools, the final algorithm has a less complex structure, avoiding a need to treat different joint types or special cases separately, all while applying to a more general set of mechanisms with general frame assignments. In terms of ease of use, we

provide sample code to show how the algorithm can be applied with a UDRF for specifying the frame assignments directly (Wensing, 2024b).

Remark 7. The RPNA has described all unobservable parameter combinations through a direct sum of local transfers. The nullspace descriptors \mathbf{N}_i can also be used to compute bases for the system-wide parameter nullspace \mathcal{N} and its orthogonal complement \mathcal{N}^\perp . Details are provided in Appendix G. Any basis for \mathcal{N} gives the unidentifiable parameter combinations for the mechanism, while any basis for \mathcal{N}^\perp gives identifiable parameter combinations. When the RPNA is run with the joint transformations ${}^i\mathbf{X}_{i-1}(0)$ in symbolic form, the provided code uses the basis of \mathcal{N}^\perp to generate parameter regroupings for base parameters in symbolic form (i.e., as a function of the kinematic parameters of the mechanism).

Remark 8. One might alternatively be interested in considering the parameter nullspace under static experiments.

$$\mathcal{N}_{\text{static}} = \{ \delta \boldsymbol{\pi} \mid \mathbf{Y}(\mathbf{q}, \mathbf{0}, \mathbf{0}) \delta \boldsymbol{\pi} = \mathbf{0}, \quad \forall \mathbf{q} \in \mathbb{R}^N \}$$

Characterizing identifiability in this way would mean only considering the term \mathbf{g} in (1), which could be of interest for developing a gravity-compensation model. An algorithm for this set is obtained by modifying line 7 of the RPNA to $\mathbf{V}_i = \mathbf{V}_i^-$ which intuitively removes local joint velocities from consideration. See Appendix F for justification.

7. Extensions for open-chain systems

This section considers extending the RPNA algorithm to tree-structure systems with multi-DoF joints. These extensions build toward the analysis of floating-base systems (e.g., mobile legged systems). We also reflect on our energy analysis so far and revisit how it relates to dynamics, which also provides a new perspective on an important past result of Ayusawa et al. (2014).

7.1. Tree-structure systems

The inertia transfer concept readily generalizes to branched open-chain rigid-body systems. In branched systems, each body has a predecessor, denoted $p(i)$, toward the base. The transfer assignment is re-written as

$$\delta \mathbf{I}_{p(i)} = -{}^i\mathbf{X}_{p(i)}^\top(0) \delta \mathbf{I}_i {}^i\mathbf{X}_{p(i)}(0)$$

All recursive steps of the RPNA generalize to branched systems by likewise replacing $i-1$ with $p(i)$.

7.2. Multi-DoF joints

Suppose each joint i has n_{d_i} DoFs with free modes:

$$\boldsymbol{\Phi}_i = [\phi_{i,1} \quad \dots \quad \phi_{i,n_{d_i}}]$$

Algorithm 1: Recursive Parameter Nullspace Algorithm for Serial-Chains with Single-DoF Joints

```

1  $\mathbf{V}_0 = {}^0\mathbf{a}_g$ ,  $\mathbf{K}_0 = \mathbf{0}_{10 \times 1}$ 
2 for  $i = 1, \dots, N$  do
   Transfer across the link
3  $\mathbf{V}_{i-1+} = {}^i\mathbf{X}_{i-1}(0) \mathbf{V}_{i-1}$ 
4  $\mathbf{K}_{i-1+} = {}^{i-1}\mathbf{B}_i^\top \mathbf{K}_{i-1}$ 

   Transfer across the joint
5  $\mathbf{V}_{i-} = \text{Ctrb}((\Phi_i \times), \mathbf{V}_{i-1+})$ 
6  $\mathbf{K}_{i-} = \text{Ctrb}(\mathbf{A}(\Phi_i)^\top, \mathbf{K}_{i-1+})$ 

   Add the current joint's effect
7  $\mathbf{V}_i = [\mathbf{V}_{i-} \quad \Phi_i]$ 
8  $\mathbf{K}_i = [\mathbf{K}_{i-} \quad \mathbf{M}(\mathbf{V}_i, \Phi_i)]$ 

   Collect the momentum and mapping conditions
9  $\mathbf{N}_i = \begin{bmatrix} \mathbf{M}(\mathbf{V}_i, \Phi_i)^\top \\ \mathbf{K}_{i-}^\top \mathbf{A}(\Phi_i) \end{bmatrix}$ 
10 end
return  $\mathbf{N}_i, \quad i = 1, \dots, N$ 

```

where each $\phi_{i,j} \in \mathbb{R}^6$ is fixed. A fixed free-mode matrix of this form could accommodate, for example, spherical or floating-base joints.

In any case, additional modifications are required to accommodate multi-DoF joints in the RPNA. First, the propagation of the attainable velocity spans must be generalized. For a single-DoF joint, joint kinematics follow a linear system (11). In contrast, for a multi-DoF joint:

$$\frac{d}{dt} {}^i\mathbf{X}_{p(i)}(\mathbf{q}_i) \in \text{span}\left((\phi_{i,1} \times) {}^i\mathbf{X}_{p(i)}, \dots, (\phi_{i,n_{d_i}} \times) {}^i\mathbf{X}_{p(i)}\right)$$

As a result, the span

$$\text{span}\{\mathbf{v} \mid \exists \mathbf{q}_i, \mathbf{v} \in \text{Range}({}^i\mathbf{X}_{p(i)}(\mathbf{q}_i)\mathbf{V})\}$$

can be seen as the smallest set containing $\text{Range}({}^i\mathbf{X}_{p(i)}(0)\mathbf{V})$ that is invariant under each $(\phi_{i,k} \times)$. This set is equivalent to the controllable subspace of a switched linear system (Sun et al., 2002) with pairs $\{((\phi_{i,k} \times), {}^i\mathbf{X}_{p(i)}(0)\mathbf{V})\}_{k=1}^{n_{d_i}}$. If the controllability matrix in Lemma 1 is replaced by a matrix whose range equals the switched controllable subspace of these pairs, then Lemma 1 holds more generally. The conditions in Lemma 2 generalize to multi-DoF joints in a similar manner. With these developments, Algorithm 2 provides a modification to the RPNA for open-chain systems with multi-DoF joints. The algorithm relies on an updated definition of the matrix $\mathbf{M}(\mathbf{V}, \Phi)$ as

$$\mathbf{M}(\mathbf{V}, \Phi) = \left[\frac{\partial}{\partial \boldsymbol{\pi}} \text{Vec}(\mathbf{V}^\top \mathbf{I}(\boldsymbol{\pi}) \Phi) \right]^\top \quad (52)$$

Algorithm 2: Recursive Parameter Nullspace Algorithm for Open-Chain Systems with Multi-DoF Joints

```

1  $\mathbf{V}_0 = {}^0\mathbf{a}_g$ ,  $\mathbf{K}_0 = \mathbf{0}_{10 \times 1}$ 
2 for  $i = 1, \dots, N$  do
   Transfer across the link
3  $\mathbf{V}_{i-1+} = {}^i\mathbf{X}_{p(i)}(0) \mathbf{V}_{p(i)}$ 
4  $\mathbf{K}_{i-1+} = {}^{p(i)}\mathbf{B}_i^\top \mathbf{K}_{p(i)}$ 

   Transfer across the joint
5  $\mathbf{V}_{i-} = \text{Ctrb}\left(\{(\phi_{i,k} \times), \mathbf{V}_{i-1+}\}_{k=1}^{n_{d_i}}\right)$ 
6  $\mathbf{K}_{i-} = \text{Ctrb}\left(\{\mathbf{A}(\phi_{i,k})^\top, \mathbf{K}_{i-1+}\}_{k=1}^{n_{d_i}}\right)$ 

   Add the current joint's effect
7  $\mathbf{V}_i = [\mathbf{V}_{i-} \quad \Phi_i]$ 
8  $\mathbf{K}_i = [\mathbf{K}_{i-} \quad \mathbf{M}(\mathbf{V}_i, \Phi_i)]$ 

   Collect the momentum and mapping conditions
9  $\mathbf{N}_i = \begin{bmatrix} \mathbf{M}(\mathbf{V}_i, \Phi_i)^\top \\ \mathbf{K}_{i-}^\top \mathbf{A}(\phi_{i,1}) \\ \vdots \\ \mathbf{K}_{i-}^\top \mathbf{A}(\phi_{i,n_{d_i}}) \end{bmatrix}$ 
10 end
return  $\mathbf{N}_i, \quad i = 1, \dots, N$ 

```

where the $\text{Vec}(\cdot)$ operation stacks the columns of a matrix into a vector.

7.3. Floating-base systems

With the generalization to multi-DoF joints, Algorithm 2 can be applied directly to floating-base systems. Note, however, that the six degrees of freedom of the floating base can enable one to simplify the algorithm. For example, since the possible velocities of any link satisfy $\mathcal{V}_i^* = \mathbb{R}^6$, basis matrices \mathbf{V}_i and \mathbf{K}_i can be chosen as identity matrices $\mathbf{V}_i = \mathbf{1}_{6 \times 6}$ and $\mathbf{K}_i = \mathbf{1}_{10 \times 10}$, with \mathbf{K}_{i-} correspondingly set to $\mathbf{K}_{i-} = \mathbf{1}_{10 \times 10}$ without effect on the algorithm. As a result, the nullspace descriptors can be constructed directly as

$$\mathbf{N}_i = \begin{bmatrix} \mathbf{M}(\mathbf{1}_{6 \times 6}, \Phi_i)^\top \\ \mathbf{A}(\phi_{i,1}) \\ \vdots \\ \mathbf{A}(\phi_{i,n_{d_i}}) \end{bmatrix}$$

such that $\mathbf{N}_i \delta \boldsymbol{\pi}_i = \mathbf{0}$ simply combines the original momentum (14) and mapping (15) conditions from Section 3. For the floating-base link (Link 1), the momentum condition is $\delta \mathbf{I}_1 = \mathbf{0}$, and so the first transfer subspace \mathcal{T}_1 is the zero set $\mathcal{T}_1 = \{\mathbf{0}\}$. Algorithm 3 shows the revised RPNA for floating-base systems.

Algorithm 3: Recursive Parameter Nullspace Algorithm for Open-Chain Floating-Base Systems

```

1  $\mathbf{N}_1 = \mathbf{1}_{10 \times 10}$ 
2 for  $i = 2, \dots, N$  do
3
4    $\mathbf{N}_i = \begin{bmatrix} \mathbf{M}(\mathbf{1}_{6 \times 6}, \Phi_i)^\top \\ \mathbf{A}(\phi_{i,1}) \\ \vdots \\ \mathbf{A}(\phi_{i,n_{d_i}}) \end{bmatrix}$ 
5 end
return  $\mathbf{N}_i, \quad i = 1, \dots, N$ 

```

Remark 9. Gravity effects provide no additional identifiable parameters in floating-base systems. This result is due to the fact that the floating base can be accelerated opposite gravity to excite the same dynamic effects as does gravity itself.

Remark 10. With the generalizations in this section, our main Theorems 1 and 2 apply to floating-base systems without motion restrictions, and certify the maximum number of parameter combinations that can be identified from a maximally exciting trajectory. However, for purely floating systems (without actuation on the base), the underactuation and constraints from physics (e.g., conservation of linear and angular momentum) can prevent resulting trajectories from being maximally exciting. For example, if the system undergoes ballistic motion without external forces other than gravity, it will lose at least one identifiable parameter (the total mass (Ayusawa et al., 2014)), and potentially more depending on the simplicity of the mechanism and its achievable motions. See Ayusawa et al. (2014) for additional discussion.

7.4. From energy to dynamics

Throughout, we have primarily considered energy analysis to characterize unidentifiable parameters rather than examining the dynamics directly. These two views are equivalent: In the absence of gravity, ensuring a zero variation to the kinetic energy $\delta T(\mathbf{q}, \dot{\mathbf{q}}) \equiv 0$ is the same as ensuring a zero variation to the mass matrix $\delta \mathbf{H}(\mathbf{q}) \equiv \mathbf{0}$, which, in turn, is the same as ensuring zero variation to the total generalized force $\delta \boldsymbol{\tau}_{\text{total}} = \delta[\mathbf{H}(\mathbf{q})\ddot{\mathbf{q}} + \mathbf{c}(\mathbf{q}, \dot{\mathbf{q}})] \equiv \mathbf{0}$.

Comparing these views, however, can help determine parameter identifiability via specific or individual forces/torques on/in the mechanism. Let us consider a two-body floating-base system similar to in Section 3. The system has kinetic energy:

$$T = \frac{1}{2} \begin{bmatrix} \mathbf{v}_1 \\ \dot{q}_2 \end{bmatrix}^\top \mathbf{H}(q_2) \begin{bmatrix} \mathbf{v}_1 \\ \dot{q}_2 \end{bmatrix}$$

where the mass matrix takes the form:

$$\begin{aligned} \mathbf{H}(q_2) &= \begin{bmatrix} \mathbf{I}_1 + {}^2\mathbf{X}_1^\top \mathbf{I}_2 {}^2\mathbf{X}_1 & {}^2\mathbf{X}_1^\top \mathbf{I}_2 \Phi_2 \\ \Phi_2^\top \mathbf{I}_2 {}^2\mathbf{X}_1 & \Phi_2^\top \mathbf{I}_2 \Phi_2 \end{bmatrix} \\ &= \begin{bmatrix} \mathbf{H}_{11} & \mathbf{H}_{12} \\ \mathbf{H}_{21} & H_{22} \end{bmatrix} \end{aligned}$$

The generalized force in this case would be comprised of the net external force on the base \mathbf{f}_1 and the torque τ_2 at joint 2, so that the equations of motion take the form:

$$\begin{bmatrix} \mathbf{f}_1 \\ \tau_2 \end{bmatrix} = \begin{bmatrix} \mathbf{H}_{11} & \mathbf{H}_{12} \\ \mathbf{H}_{21} & H_{22} \end{bmatrix} \begin{bmatrix} \dot{\mathbf{v}}_1 \\ \ddot{q}_2 \end{bmatrix} + \begin{bmatrix} \mathbf{c}_1 + \mathbf{g}_1 \\ c_2 + g_2 \end{bmatrix}$$

In this case, the momentum and mapping conditions for a transfer at joint 2 would collectively enforce a zero variation $\delta \mathbf{f}_1$ to the base force, and zero variation $\delta \tau_2$ to the joint torque. More generally, *without any additional motion restrictions*, these two conditions for body i ensure that a transfer with the parent would not disturb the connecting joint torque τ_i nor the total 6D predecessor force $\mathbf{f}_{p(i)}$ (acting from body $p(i)$ onto body i).

For this two-body system, a zero variation to the mass matrix \mathbf{H} overall is equivalent to a zero variation in its first row (\mathbf{H}_{11} and \mathbf{H}_{12}) alone. This result follows due to symmetry ($\mathbf{H}_{12} = \mathbf{H}_{21}^\top$) and the fact that the lower right block gives a projection of the upper right one ($H_{22} = \Phi_2^\top \mathbf{I}_2 {}^2\mathbf{X}_1^\top \mathbf{H}_{12}$). Physically, this occurs since any reaction torque τ_2 is balanced by and shows up on the base force \mathbf{f}_1 . For $\delta \mathbf{f}_1 \equiv \mathbf{0}$, then we necessarily have $\delta \tau_2 \equiv 0$ as well, such that the joint torque provides no additional information nor can identify any additional parameters not already identifiable via the base force.

We generalize this argument to open-chain floating-base systems in Appendix H: Under proper excitation and full knowledge of net external forces, measurements of joint torques do not enable the identification of any additional parameters. This is inspired by and equivalent to a key result of Ayusawa et al. (2014).

8. Special case extension to closed kinematic loops: Joint motors

Joint motors represent a simple and common closed kinematic chain, driving the joint between two connected bodies. The spinning rotor is coupled to the joint, but often spins a fixed ratio faster and may spin along a distinct axis. Hence, we must treat the rotor as a separate body. Toward understanding this case, we consider a three-body system with a floating base (body 1), a child link (body 2), and a motor's rotor (body m), building on Section 3.

We will also assume the rotor to be rotationally symmetric about the motor axis, giving it only four inertial parameters versus the general 10 parameters.

8.1. Dynamics and constraints

The total kinetic energy of the three bodies is

$$T = \frac{1}{2} \mathbf{v}_1^\top \mathbf{I}_1 \mathbf{v}_1 + \frac{1}{2} \mathbf{v}_2^\top \mathbf{I}_2 \mathbf{v}_2 + \frac{1}{2} \mathbf{v}_m^\top \mathbf{I}_m \mathbf{v}_m$$

where, similar to (3),

$$\begin{aligned} \mathbf{v}_2 &= {}^2\mathbf{X}_1(q_2) \mathbf{v}_1 + \Phi_2 \dot{q}_2 \\ \mathbf{v}_m &= {}^m\mathbf{X}_1(q_m) \mathbf{v}_1 + \Phi_m \dot{q}_m \end{aligned}$$

are the child and rotor body velocities, respectively. The motor position $q_m = n_R q_2$ and speed $\dot{q}_m = n_R \dot{q}_2$ are modified by the fixed gear ratio n_R .

Using inertia variations $\delta \mathbf{I}$ and the gear ratio n_R , we can factor the energy variation δT as

$$\delta T = \frac{1}{2} \begin{bmatrix} \mathbf{v}_1 \\ \dot{q}_2 \end{bmatrix}^\top \begin{bmatrix} \delta \mathbf{I}_1^C & {}^2\mathbf{X}_1^\top \delta \mathbf{I}_2 \Phi_2 + {}^m\mathbf{X}_1^\top \delta \mathbf{I}_m \Phi_m n_R \\ \Phi_2^\top \delta \mathbf{I}_2 \Phi_2 + \Phi_m^\top \delta \mathbf{I}_m \Phi_m n_R^2 \end{bmatrix} \begin{bmatrix} \mathbf{v}_1 \\ \dot{q}_2 \end{bmatrix}$$

where the composite inertia variation $\delta \mathbf{I}_1^C$ is

$$\delta \mathbf{I}_1^C = \delta \mathbf{I}_1 + {}^2\mathbf{X}_1^\top(q_2) \delta \mathbf{I}_2 {}^2\mathbf{X}_1(q_2) + {}^m\mathbf{X}_1^\top(q_m) \delta \mathbf{I}_m {}^m\mathbf{X}_1(q_m)$$

Mirroring the developments of Section 3, enforcing $\delta T = 0$, we use the top-left composite inertia $\delta \mathbf{I}_1^C$ to define a fixed transfer assignment analogous to (9)

$$\delta \mathbf{I}_1 = -{}^2\mathbf{X}_1^\top(0) \delta \mathbf{I}_2 {}^2\mathbf{X}_1(0) - {}^m\mathbf{X}_1^\top(0) \delta \mathbf{I}_m {}^m\mathbf{X}_1(0)$$

and uniformly ensure $\delta \mathbf{I}_1^C = \mathbf{0}$ over all configurations by zeroing its derivative.

$$\mathbf{0} = \frac{d}{dt} ({}^2\mathbf{X}_1^\top(q_2) \delta \mathbf{I}_2 {}^2\mathbf{X}_1(q_2) + {}^m\mathbf{X}_1^\top(q_m) \delta \mathbf{I}_m {}^m\mathbf{X}_1(q_m))$$

Using (11), we find the mapping condition

$$\begin{aligned} \mathbf{0} &= {}^2\mathbf{X}_1^\top(q_2) [(\Phi_2 \times)^\top \delta \mathbf{I}_2 + \delta \mathbf{I}_2 (\Phi_2 \times)] {}^2\mathbf{X}_1(q_2) \dot{q}_2 \\ &\quad + {}^m\mathbf{X}_1^\top(q_m) [(\Phi_m \times)^\top \delta \mathbf{I}_m + \delta \mathbf{I}_m (\Phi_m \times)] {}^m\mathbf{X}_1(q_m) \dot{q}_m \end{aligned} \quad (53)$$

From the off-diagonal elements for δT , we obtain the momentum condition

$$\mathbf{0} = \Phi_2^\top \delta \mathbf{I}_2 {}^2\mathbf{X}_1(q_2) + \Phi_m^\top \delta \mathbf{I}_m {}^m\mathbf{X}_1(q_m) n_R \quad (54)$$

Finally, from the bottom-right element determining δT , we obtain a third condition, which we refer to as the torque condition.

$$0 = \Phi_2^\top \delta \mathbf{I}_2 \Phi_2 + \Phi_m^\top \delta \mathbf{I}_m \Phi_m n_R^2 \quad (55)$$

Note, in this three-body case the momentum condition does not automatically guarantee the torque condition.¹

8.2. Symmetric rotor

In the common case that the motor's rotor is rotationally symmetric, these conditions simplify a great deal. Consider

a rotor spinning about its z axis such that $\Phi_m = [0 \ 0 \ 1 \ 0 \ 0 \ 0]^\top = [\hat{\mathbf{z}}^\top \mathbf{0}^\top]^\top$. Rotational symmetry implies $I_{xy} = I_{xz} = I_{yz} = h_x = h_y = 0$ and $I_{xx} = I_{yy}$. Intuitively, variations respecting this symmetry satisfy

$$\mathbf{0} = (\Phi_m \times)^\top \delta \mathbf{I}_m + \delta \mathbf{I}_m (\Phi_m \times) \quad (56)$$

This means a rotor only has four parameters: I_{zz} , $I_{xx} = I_{yy}$, h_z , m . Further, from the example in Section 3.2, we know that the revolute attachment via the z axis allows the transfer of the last three to the parent body 1. So, practically, we need only consider changes to the rotor inertia δI_{zz} about its axis. This further gives

$$\Phi_m^\top \delta \mathbf{I}_m = \Phi_m^\top \delta I_{zzm} \quad (57)$$

$$\Phi_m^\top \delta \mathbf{I}_m \Phi_m = \delta I_{zzm} \quad (58)$$

8.3. Simplified conditions

We simplify the mapping condition (53) using symmetry via (56), acknowledging that ${}^2\mathbf{X}_1(q_2)$ is full rank, and allowing \dot{q}_2 to take any value. For the momentum condition (54), we first post-multiply with ${}^1\mathbf{X}_2(q_2)$. We then recognize that $\Phi_m^\top {}^m\mathbf{X}_1(q_m) = \Phi_m^\top {}^m\mathbf{X}_1(0)$ since the motor rotation does not change the motor axis, and note ${}^m\mathbf{X}_2(q_2) = {}^m\mathbf{X}_1(0) {}^1\mathbf{X}_2(q_2)$ is a transform from joint 2 to the motor. Finally, for the torque condition, we use (58). In summary, the mapping, momentum, and torque conditions are

$$\mathbf{0} = (\Phi_2 \times)^\top \delta \mathbf{I}_2 + \delta \mathbf{I}_2 (\Phi_2 \times) \quad (59)$$

$$\mathbf{0} = \Phi_2^\top \delta \mathbf{I}_2 + \delta I_{zzm} n_R \Phi_m^\top {}^m\mathbf{X}_2(q_2) \quad (60)$$

$$0 = \Phi_2^\top \delta \mathbf{I}_2 \Phi_2 + \delta I_{zzm} n_R^2 \quad (61)$$

where the mapping condition reverts to the two-body case from Section 3, while the momentum and torque conditions retain the rotor inertia terms.

Since δI_{zzm} is the only parameter appearing for the motor, it can at most add this one additional identifiable parameter. However, if both conditions (60) and (61) are satisfied, δI_{zzm} can be transferred to $\delta \mathbf{I}_1$ and it becomes unidentifiable.

Breaking down Φ_2 into its rotational and linear components as $\Phi_2 = [\hat{\mathbf{e}}_{\omega,2}^\top, \hat{\mathbf{e}}_{v,2}^\top]^\top$ and then post-multiplying (60) with Φ_2 then gives a projected version of the momentum condition as

$$\begin{aligned} 0 &= \Phi_2^\top \delta \mathbf{I}_2 \Phi_2 + n_R \delta I_{zzm} \Phi_m^\top {}^m\mathbf{X}_2(q_2) \Phi_2 \\ &= \Phi_2^\top \delta \mathbf{I}_2 \Phi_2 + n_R \delta I_{zzm} \hat{\mathbf{z}}^\top {}^m\mathbf{R}_2(q_2) \hat{\mathbf{e}}_{\omega,2} \end{aligned} \quad (62)$$

where the structure of the spatial transform ${}^m\mathbf{X}_2(q_2)$ (given in Appendix A) is used in the simplification to (62).

We see that (61) and (62) can simultaneously be satisfied if and only if

$$n_R^2 \delta I_{zzm} = n_R \delta I_{zzm} \hat{\mathbf{z}}^\top {}^m\mathbf{R}_2(q_2) \hat{\mathbf{e}}_{\omega,2}$$

or equivalently that

$$n_R = \hat{\mathbf{z}}^\top {}^m \mathbf{R}_2(q_2) \hat{\mathbf{e}}_{\omega,2} \quad (63)$$

If the joint is purely translational, (63) cannot hold since $\hat{\mathbf{e}}_{\omega,2} = \mathbf{0}$, and so δI_{zz_m} is identifiable in this case

If the joint has a rotational component (e.g., for a revolute or helical joint), we assume $\|\hat{\mathbf{e}}_{\omega,2}\| = 1$. In this case, (63) holds when

1. the gear ratio n_R is unity and
2. the rotational component of joint 2 ($\hat{\mathbf{e}}_{\omega,2}$) is parallel to the motor axis (making $\Phi_m^\top {}^m \mathbf{X}_2(q_2) \Phi_2$ unity).

Note that this allows only a motor without no reduction ($n_R = 1$), but that may be offset from the joint. In all other cases, the rotor adds one identifiable parameter I_{zz_m} .

8.4. Generalization to fixed bases

More generally, the fixed-base case requires considerations of motion restrictions to determine whether the single additional motor parameter is identifiable at each joint.

The main conceptual difference in the fixed-base case is that the motor inertia is identifiable **only** if it can be felt earlier in the chain. This follows since, when \mathbf{v}_1 has motion restrictions, the momentum condition takes the form:

$$\mathbf{0} = \Phi_2^\top \delta \mathbf{I}_2 {}^m \mathbf{X}_1(q_2) \mathbf{v}_1 + \Phi_m^\top \delta \mathbf{I}_m {}^m \mathbf{X}_1(q_m) n_R \mathbf{v}_1$$

where $\Phi_m^\top \delta \mathbf{I}_m {}^m \mathbf{X}_1(q_m) \mathbf{v}_1 = \mathbf{0}$ when the rotor inertia is not felt along any of the directions \mathbf{v}_1 can take. For instance, consider the simple manipulators in Figure 4 and suppose each motor rotates along the corresponding joint axis $\hat{\mathbf{z}}_i$. The inertia of the first motor is not felt earlier in the chain for either mechanism (since there are no previous joints), and thus it does not add an identifiable parameter. For the system with parallel joints in Figure 4(a), the rotational inertia of the second motor about its axis does contribute rotational inertia about the first joint axis, and it is found to add an identifiable parameter. For the system with perpendicular joint axes, the rotational inertia of the second motor rotor does not lead to any rotational inertia about the first joint axis. Thus neither motor inertia contributes an identifiable parameter in this case.

The full details of generalizing the RPNA to consider rotors are described in Appendix I. They are also implemented in the companion MATLAB code where \mathbf{V}_i and \mathbf{K}_i are generalized to include effects from both the rotor and link associated with joint i . Returning to the original motivation for the RPNA, these generalizations avoid the algorithm having to consider special cases, such as when joints are/aren't parallel, as was necessary in the above example.

Remark 11. *The reason that the RPNA can be extended to this class of closed kinematic chains is that we can characterize the velocities that are simultaneously experienced by the link and its corresponding rotor. In the more general*

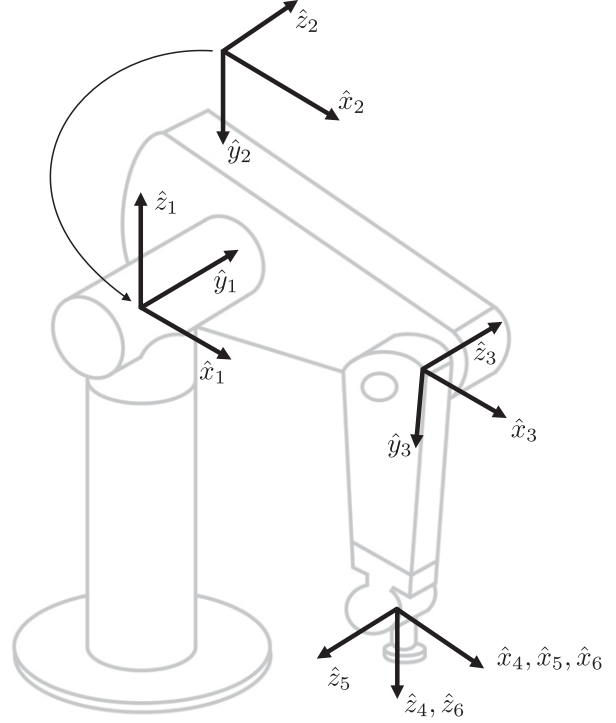


Figure 7. PUMA 560 robot and coordinate assignment.

case for parallel robots, however, it is difficult to characterize the velocities that can be simultaneously achieved by neighboring bodies in a closed kinematic loop. Doing so, in general, has a complicated (and nonlinear) dependency on all the joint variables within the kinematic loop, which prevents using the linear systems tools that we've employed for the RPNA one joint at a time.

9. Verification and system-level examples

This section provides verification of the recursive parameter nullspace algorithm (RPNA) for fixed- and floating-base systems. The RPNA is unique in that it requires only the kinematic parameters of a mechanism as its input, it is provably correct, and it can be run without any symbolic manipulations or assumed exciting input data. We use numerical identification approaches (Atkeson et al., 1986) to empirically verify the RPNA output, with examples given as part of the released code (Wensing, 2024c).

9.1. PUMA 560

We first consider the classical industrial manipulator PUMA 560 shown in Figure 7. The mechanism has three joints to position the wrist, followed by three wrist joints with intersecting orthogonal axes.

Table 2 details the parameter identifiability for this mechanism, with the symbols explained in Table 3. We recall that there are three possibilities for each parameter: identifiable by itself, unidentifiable, and identifiable in

Table 2. PUMA 560 parameters—see Table 3 for legend. We find 36/40 base parameters ex-/including motor inertias. Dimensions in parentheses consider gravitational effects.

	1	2	3	4	5	6
m	×	×	○	○	○	○
mc_x	×	•	•	★	★	★
mc_y	×	★	•	•	•	★
mc_z	×	×	○	○	○	○
I_{xx}	×	•	•	•	•	•
I_{yy}	×	○	○	○	○	○
I_{zz}	•	•	•	•	•	★
I_{yz}	×	★	★	★	★	★
I_{xz}	×	•	★	★	★	★
I_{xy}	×	★	•	★	★	★
I_m	○	○	★	★	★	★
$\dim(\mathcal{V}_i)$	1 (2)	3 (5)	6	6	6	6
$\dim(\mathcal{K}_i)$	1 (1)	6 (8)	10	10	10	10
$\dim(\mathcal{T}_i)$	9	3	3	3	3	3

Table 3. Legend for results tables. A minimal set of parameters is denoted by the filled stars and circles. Filled circles are only identifiable in combination with open circles.

Symbol	Explanation
★	Identifiable individually
•	Identifiable in combination with others (Selected as a base parameter)
○	Identifiable in combination with others
×	Unidentifiable parameter

linear combinations only (Atkeson et al., 1986). A minimal set of parameters representing identifiable combinations is indicated with symbols (•) in the table. There is some freedom in this assignment since, for example, there is an arbitrary choice as to whether $I_{xx,6}$ or $I_{yy,6}$ is chosen as the base parameter for the identifiable combination $I_{xx,6} - I_{yy,6}$. To resolve this ambiguity, we always give preference to the parameter that appears first in the table as we move down the columns and right across the rows. With this designation, the total number of solid entries (★ or •) in any given column represents the number of base parameters contributed by that body. The total number of × and ○ symbols for each body indicates the number of degrees of freedom in the inertia transfer with its parent. Using the RPNA, the PUMA is found to have 36 identifiable parameter combinations for its bodies. This is consistent with previous symbolic approaches (Gautier and Khalil, 1990; Mayeda et al., 1990) that relied on many special cases. Additional details on the identifiable linear combinations (i.e., full specification of the parameter combinations for the given choice of base parameters) can be obtained from the supplementary MATLAB code.

Motion restrictions play an important role on the structure of the identifiable parameters for the first two

bodies of the PUMA. The true attainable velocity spans have sub-maximal dimensions 1 and 3, and have dimensions 2 and 5 within the algorithm when considering gravity as a fictitious prismatic joint at the base. All remaining bodies have full dimension for \mathcal{V}_i and \mathcal{K}_i . Despite the motion restrictions on body 2, its undetectable transfers coincide with that of an unconstrained body. The mass m_2 is unidentifiable since it is not sensed by joint 2, and it is mapped onto the parent parameter m_1 , which is itself unidentifiable. Likewise, mc_{z_2} is not sensed by torques on joint 2 and maps to parent parameters mc_{x_1} and mc_{y_1} depending on the value of q_2 . Both of these parameters of the parent are unidentifiable, and thus so is mc_{z_2} . With respect to the motor inertias, the first two joints of the PUMA are perpendicular and thus its first two motor inertias are not identifiable (see the end of Section 8).

To show the benefits of the method compared to past symbolic methods, we compared the output to the popular SYMORO package (Khalil et al., 2014) available at <https://github.com/symoro/symoro>. The default RX90 manipulator in SYMORO has the same kinematic structure as the Puma 560, and so will have the same number of identifiable parameters. In the case with gravity along \hat{z}_1 , both the RPNA and SYMORO reported 36 identifiable parameter combinations. However, when gravity was removed (e.g., as may be of interest for a space robotics application), the RPNA reported only 34 identifiable parameters, while SYMORO continued to report 36. Using the same random sampling of data that was certified as maximally exciting from the previous steps, the result of 34 parameter combinations was confirmed as correct via numerical approaches. SYMORO's incorrect output in this instance is due to it not accounting for a special case with no gravity. Past symbolic regrouping rules (e.g., such as those in Khalil and Dombre (2002)) could undoubtedly be considered for SYMORO to address this special case. However the need for such additional steps underlies the benefit of the RPNA in that it provides a systematic method for determining the identifiable parameter regroupings without requiring special cases. Further, compared to a numerical approach, the RPNA does not require a second method to certify its output (due to its proof of correctness), and it can provide all identifiable parameter combinations symbolically.

9.2. SCARA

The second example considered is a SCARA robot depicted in Figure 8. The SCARA is a 4-DoF RRPR manipulator traditionally used in pick and place operations. All rotations and translations take place about the local \hat{z}_i axes. Motion restrictions play a key role in parameter identifiability for this robot, as described in Table 4.

Each of the joints in the SCARA admits more transfer freedoms than in the floating case. The first two links of the SCARA resemble the parallel joint example from Section 4.5.1. As a result, the second revolute joint admits seven

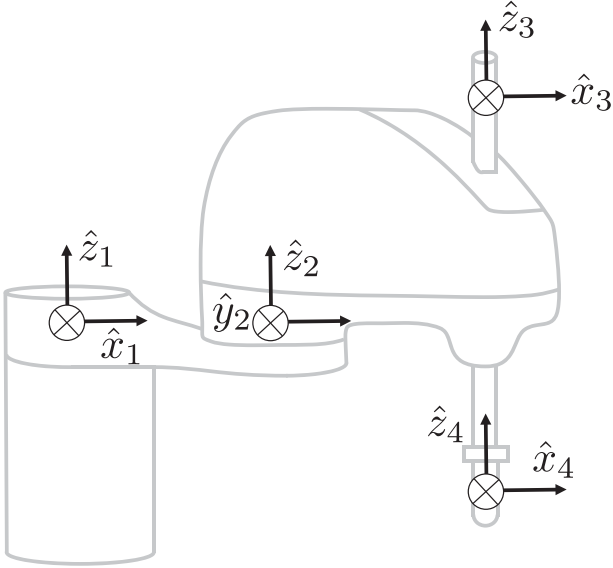


Figure 8. SCARA robot and coordinate assignment.

Table 4. SCARA parameters—see Table 3 for legend. We find 8/11 base parameters ex-/including motor inertias. Dimensions in parentheses consider gravitational effects.

	1(R)	2(R)	3(P)	4(R)
m	×	○	●	○
mc_x	×	●	○	★
mc_y	×	●	○	★
mc_z	×	×	×	×
I_{xx}	×	×	×	×
I_{yy}	×	×	×	×
I_{zz}	●	●	○	★
I_{yz}	×	×	×	×
I_{xz}	×	×	×	×
I_{xy}	×	×	×	×
I_m	○	★	★	★
$\dim(\mathcal{V}_i)$	1 (2)	3 (4)	4 (4)	4 (4)
$\dim(\mathcal{K}_i)$	1 (1)	4 (4)	4 (4)	4 (4)
$\dim(\mathcal{T}_i)$	9	7	9	7

transfer degrees of freedom and contributes three identifiable parameter combinations. Motion restrictions likewise enlarge the undetectable transfers across the prismatic joint. While a free-floating prismatic joint admits six transfer degrees of freedom, the SCARA prismatic joint admits nine transfer degrees of freedom.

These extra transfer freedoms for the SCARA prismatic joint can be understood physically from the momentum and mapping conditions. The momentum condition $\Phi_3^\top \delta \mathbf{I}_3 \mathbf{V}_3 = \mathbf{0}$ requires that $\delta \mathbf{I}_3$ must not modify the linear momentum of body 3 along \hat{z}_3 . Motions of joint 3 will create pure linear momentum along \hat{z}_3 with magnitude $m_3 \dot{q}_3$, while motions of joints 1 and 2 do not create any linear momentum in this direction. Thus, the momentum condition requires $\delta m_3 = 0$.

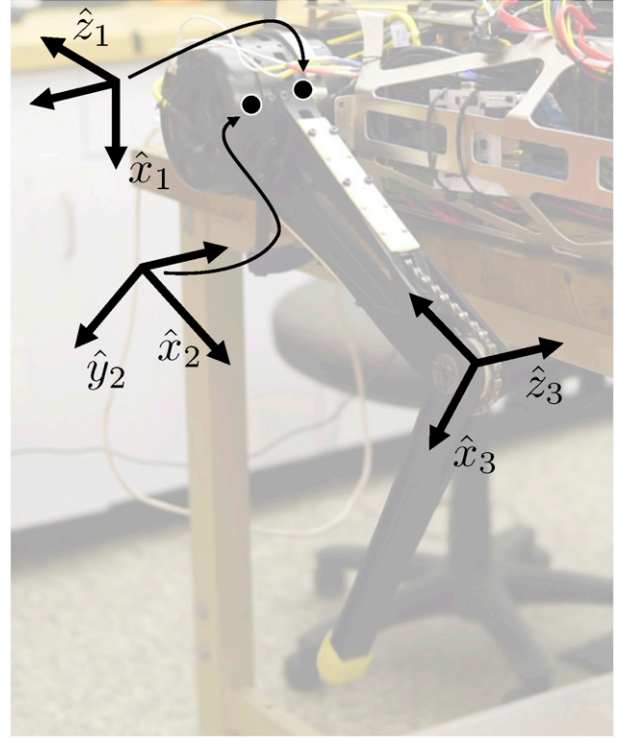


Figure 9. Cheetah 3 leg coordinates. Table-top experiments like the one shown are often used to identify leg parameters.

It turns out that the mapping condition (51) for joint 3 holds without restriction on $\delta \mathbf{I}_3$. Recall that the mapping condition considers changes in the way $\delta \mathbf{I}_3$ maps to parameters of its parent. It holds when any changes in this mapping with q_3 appear only on unidentifiable parameters for the parent. Changes in q_3 affect the vertical distribution of mass for body 3 relative to its parent. Yet, any of the parameters affected by the vertical distribution of mass (i.e., mc_z , I_{xx} , I_{yy} , I_{xz} , I_{yz}) are unidentifiable for body 2. Thus, the mapping condition holds without any restriction on $\delta \mathbf{I}_3$. As a result, transfers between body 2 and body 3 need only satisfy the momentum condition $\delta m_3 = 0$, providing nine transfer freedoms across this joint. Discounting motors, the mechanism has 32 unobservable parameter combinations and therefore only eight identifiable parameter combinations. This was confirmed empirically through an SVD applied to random samples of the regressor \mathbf{Y} .

9.3. Cheetah 3 leg

The last example considers the MIT Cheetah 3 robot (Bledt et al., 2018a) consisting of a torso and four independent legs. Each leg contains three rigid bodies, driven by three proprioceptive actuators (Wensing et al., 2017b). It is common to identify the legs separate from the torso (Toussaint et al., 2017; Wensing et al., 2017a): Fixing the torso as a base, leg swing experiments identify the leg parameters, as depicted in the Figure 9. We explore whether this common setup is appropriate to fully identify the leg,

Table 5. Cheetah 3 leg parameters—see Table 3 for legend. We find 17/18 versus 21/24 base parameters in fixed versus floating conditions, ex-/including motor inertias. Dimensions in parentheses consider gravitational effects.

	Fixed			Free		
	1	2	3	1	2	3
m	×	○	○	○	○	○
mc_x	★	●	★	★	●	★
mc_y	●	★	★	●	★	★
mc_z	×	○	○	○	○	○
I_{xx}	×	●	●	●	●	●
I_{yy}	×	○	○	○	○	○
I_{zz}	●	●	★	●	●	★
I_{yz}	×	★	★	★	★	★
I_{xz}	×	●	★	★	★	★
I_{xy}	×	★	★	★	★	★
I_m	○	○	★	★	★	★
$\dim(\mathcal{V}_i)$	1 (3)	3 (6)	6	6	6	6
$\dim(\mathcal{K}_i)$	1 (3)	6 (9)	10	10	10	10
$\dim(\mathcal{T}_i)$	7	3	3	3	3	3

testing two cases: floating versus fixed torso. We find that motion restrictions in the fixed experiments prevent exciting all the parameters affecting the floating torso case.

Table 5 compares parameter identifiability in the fixed-versus floating-base case. Similar to the PUMA and SCARA examples, the Cheetah 3 leg model possesses unidentifiable parameters in the fixed-base case. As expected, the parameter set is a subset of the floating-base case.

To analyze the effects of motion restrictions in a concrete situation, consider the scenario of Cheetah 3 executing a transverse gallop. For consistency across cases, motion data is collected in simulation, shown in Figure 10, and includes the configuration \mathbf{q} , generalized velocity $\dot{\mathbf{q}} \in \mathbb{R}^{18}$, and generalized acceleration $\ddot{\mathbf{q}} \in \mathbb{R}^{18}$. Motor rotors are modeled as rigid bodies themselves connected to the preceding link via a revolute joint.

In the floating case, we retain the motion as simulated and use inverse dynamics to determine the matching generalized force $\boldsymbol{\tau}_{\text{total}} \in \mathbb{R}^{18}$ with effects from active joint torques and ground forces. We estimate all 178 inertial parameters $\hat{\boldsymbol{\pi}} \in \mathbb{R}^{178}$, stemming from the 13 bodies (10 parameters each) and 12 rotationally symmetric rotors (4 parameters each, as described in Section 8). Following (2), the parameters of the system are found by solving a least-squares problem:

$$\min_{\boldsymbol{\pi}} \sum_{j=1}^{N_s} \left\| \boldsymbol{\tau}_{\text{total}}^{[j]} - \mathbf{Y}(\mathbf{q}^{[j]}, \dot{\mathbf{q}}^{[j]}, \ddot{\mathbf{q}}^{[j]}) \hat{\boldsymbol{\pi}} \right\|^2 \quad (64)$$

where N_s is the number of samples used, the superscript $(\cdot)^{[j]}$ indicates the j th sample of the quantity. Note that the experiment captures both torques for the legs at the joints, as well as associated dynamic coupling forces on the body.

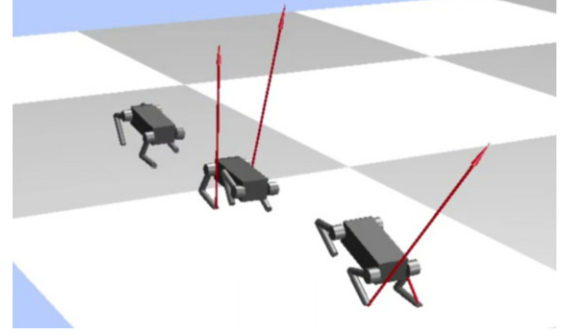


Figure 10. Full dynamic simulation of galloping was used to obtain the identification dataset for the MIT Cheetah 3 model.

In the fixed case, we lock the torso in place, high above the ground, mimicking the table-top setup in Figure 9. For simplicity, only the front-left (FL) leg is considered. To provide a fair comparison with the floating case, the same swing-leg trajectories are employed, providing configuration $\mathbf{q}_{FL} \in \mathbb{R}^3$, velocity $\dot{\mathbf{q}}_{FL}$, and acceleration $\ddot{\mathbf{q}}_{FL}$. Again, inverse dynamics determine the required joint torques $\boldsymbol{\tau}_{FL} \in \mathbb{R}^3$, this time inherently without ground reaction forces. Equivalent to (64), we estimate the 42 leg parameters $\hat{\boldsymbol{\pi}}_{FL} \in \mathbb{R}^{42}$ (three bodies and rotors).

An SVD on the regressor shows that the fixed case includes 18 identifiable combinations (17 from links, and 1 from the knee rotor), while the floating case includes 24 identifiable combinations for each leg (21 from links, and 3 from rotors). The provably-correct output of the RPNA, summarized in Table 5, thus certifies that this motion is maximally exciting for both the fixed- and floating-base cases. Details on the identifiable combinations for both cases are available by running the supplementary MATLAB code.

We computed solutions $\hat{\boldsymbol{\pi}}$ and $\hat{\boldsymbol{\pi}}_{FL}$ to the optimization problems without leveraging prior knowledge by using a pseudo-inverse. In practice, regularization is often used to include a prior parameter estimate (often from CAD) (e.g., Lee et al., 2020), which also occurs implicitly (Boffi and Slotine, 2021) in adaptive settings (e.g., Lee et al., 2018).

Table 6 shows the validation error across the two cases. For identification with the floating torso, as expected, validation errors are zero in both the floating-base and fixed-base validation cases. In contrast, the fixed torso identification only displays favorable generalization when applied to another fixed-base data set. When the identified leg model is used within a full floating-base model, validation errors appear on both the leg torques and the coupling forces/torques.

The parameter identifiability analysis in Table 5 explains the leg torque errors in the floating-base validation. These validation errors occur when motions of the body excite new dynamic effects for the leg that were not captured in the mock table-top experiments. It is observed that these motion restrictions are only on the first two links (ab/ad and hip), while the shank (body 3) can fully excite all its parameters, as signified by $\dim(\mathcal{V}_3) = 6$ and $\dim(\mathcal{K}_3) = 10$. As a result

Table 6. RMS validation errors on the galloping dataset using fixed-base versus floating-base identification. Leg torque vector report ab/ad, hip, and knee residuals. Force and torque residuals are reported in body coordinates with +x forward, +y left, and +z up.

Identification	Fixed-base validation	Floating-base validation		
	Leg joint torques	Leg joint torques	Body torques	Body forces
Floating torso	[0,0,0] Nm	[0,0,0] Nm	[0,0,0] Nm	[0,0,0] N
Fixed torso	[0,0,0] Nm	[1.35, 2.39, 0.00] Nm	[7.40, 20.24, 4.71] Nm	[16.31, 12.40, 61.73] N

of this full excitation of the shank parameters in the fixed-base case, the knee joint experiences zero validation errors when generalizing to the free-base case.

10. Conclusions

This paper has introduced the recursive parameter null-space algorithm (RPNA) to geometrically characterize the identifiability of inertial parameters in a rigid-body system. We have shown that unidentifiable parameter combinations have an interpretation as representing a sequence of undetectable inertial transfers across the joints. In arriving at this result, we have transformed the nonlinear parameter identifiability problem for the system as a whole into a sequence of classical linear systems observability problems, proceeding recursively across each joint of the mechanism. As a result of these new theoretical advances, the final algorithm is compact (it can be expressed in 10 lines), while generalizing the results of multiple previous authors. Further, due to the generality in this theoretical development, the final algorithm admits general frame assignments, and applies to generic joint models, both of which provide a modern update to past symbolic results for broader practical use. While our method departs from past symbolic methods in its derivation strategy, the final algorithm can be run with symbolic inputs for the mechanism kinematics parameters to obtain symbolic base parameters. The final algorithm is easy to use and provided open-source, requiring only a URDF file or equivalent to specify the kinematics of the bodies. Extensions have been discussed to handle general multi-DoF joint models, branched kinematic trees, and simple closed loops arising from geared motors. The results verify the correctness of the algorithm, highlight cases where past symbolic methods may fail, and illustrate the importance of considering motion restrictions when designing identification strategies for mobile systems. Overall, the theoretical advances underpinning the RPNA, combined with its practical usability and generality, should make the algorithm a valuable tool for practitioners in system identification and adaptive control applications.

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Notes

1. Connecting back to Section 7.4, a related implication for floating-base systems is that motor rotors are generally not identifiable from ground forces alone.
2. Note that the reasoning for $C \Rightarrow D$ no longer holds for systems that have closed kinematics loops, as is observed in Section 8.

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Appendix

A. Rigid-body dynamics details

A.1. Spatial velocities. The spatial velocity is a 6D velocity that collects traditional 3D rotational and linear velocities. When expressed in its body-fixed coordinate frame the spatial velocity of body i takes the form

$$\mathbf{v}_i = \begin{bmatrix} \boldsymbol{\omega}_i \\ \mathbf{v}_i \end{bmatrix} \quad (65)$$

where $\omega_i \in \mathbb{R}^3$ is the angular velocity in body coordinates, and $v_i \in \mathbb{R}^3$ is the linear velocity of the coordinate origin (given in body coordinates).

A.2. Spatial transform. The 6×6 matrix ${}^i\mathbf{X}_{i-1}$ is a spatial transformation matrix that converts spatial velocities expressed in frame $i-1$ to frame i :

$${}^i\mathbf{X}_{i-1} = \begin{bmatrix} {}^i\mathbf{R}_{i-1} & \mathbf{0} \\ -{}^i\mathbf{R}_{i-1} \mathbf{S}({}^{i-1}\mathbf{p}_i) & {}^i\mathbf{R}_{i-1} \end{bmatrix} \quad (66)$$

where ${}^i\mathbf{R}_{i-1} \in \mathbb{R}^3$ the rotation matrix from frame $i-1$ to frame i , ${}^{i-1}\mathbf{p}_i$ the vector from the origin of frame $i-1$ to the origin of frame i , and $\mathbf{S}(\mathbf{x}) \in \mathbb{R}^{3 \times 3}$ is the skew-symmetric 3D cross-product matrix satisfying $\mathbf{S}(\mathbf{x})\mathbf{y} = \mathbf{x} \times \mathbf{y}$ for all $\mathbf{x}, \mathbf{y} \in \mathbb{R}^3$. Note that if ${}^i\mathbf{T}_{i-1}$ is the homogenous transform between frames i and $i-1$, then ${}^i\mathbf{X}_{i-1}$ is equivalent to the (big) Adjoint matrix $\text{Ad}_{{}^i\mathbf{T}_{i-1}}$. We note that these transformations satisfy

$${}^{i-1}\mathbf{X}_i = {}^i\mathbf{X}_{i-1}^{-1}$$

However, unlike with rotation matrices, ${}^{i-1}\mathbf{X}_i^\top \neq {}^i\mathbf{X}_{i-1}^{-1}$.

A.3. Spatial cross product. The 6×6 spatial “cross product” matrix is given by

$$\begin{bmatrix} \omega \\ v \end{bmatrix} \times = \begin{bmatrix} \mathbf{S}(\omega) & \mathbf{0} \\ \mathbf{S}(v) & \mathbf{S}(\omega) \end{bmatrix}$$

Similar to the standard 3D cross product, the spatial cross product can be used to provide the rate of change in a 6D quantity due to its expression in moving coordinates (Featherstone, 2008). Note that the 6×6 cross product matrix $(v \times)$ is equivalent to the (little) adjoint matrix ad_v .

B. Second example for fixed-based systems

For the system in Figure 4(b), the velocity of body 2 and its span can be given by

$$\mathbf{v}_2 = \begin{bmatrix} \ell s_2 \dot{q}_1 \\ \ell c_2 \dot{q}_1 \\ \dot{q}_2 \\ 0 \\ 0 \\ -\ell \dot{q}_1 \end{bmatrix} \quad \mathbf{V}_2 = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

In comparison to the example system in Figure 4(a), the span \mathbf{V}_2 has an extra column, representing additional motion freedoms for the second body of this non-planar system. This example also highlights that not all velocities in the span \mathbf{V}_2 are themselves attainable. It is observed that while the last column of \mathbf{V}_2 is in the span of attainable velocities, a pure linear velocity in the \hat{z}_2 direction is not possible.

The first body again only has a single identifiable parameter I_{zz_1} . Considering a transfer between body 1 and body 2, the momentum condition $\Phi_2^\top \delta \mathbf{I}_2 \mathbf{V}_2 = \mathbf{0}$ enforces

$$\delta I_{xz_2} = \delta I_{yz_2} = \delta I_{zz_2} = 0 \quad (67)$$

Similar to the previous case, these conditions impose that inertial changes to body 2 must not create angular momentum about \hat{z}_2 . However, the change in joint geometries between the examples provides a different set of parameters that are identifiable via the second joint torque.

The mapping condition:

$$\Phi_1^\top {}^2\mathbf{X}_1^\top(0) \delta \mathbf{I}_2^{(k)} {}^2\mathbf{X}_1(0) \Phi_1 = 0 \quad \forall k = 1, \dots, 10$$

collectively enforces

$$\delta h_{x_2} = \delta h_{y_2} = \delta I_{xx_2} - \delta I_{yy_2} = \delta I_{xy_2} = 0 \quad (68)$$

for $k = 1, \dots, 4$. The conditions are redundant for all larger k . Again, these conditions ensure that any variations how in $\delta \mathbf{I}_2$ maps to δI_{zz_1} must be zero. Note that the rotational inertia of Body 2 about \hat{z}_1 is

$$\begin{aligned} \Phi_1^\top {}^2\mathbf{X}_1^\top \mathbf{I}_2 {}^2\mathbf{X}_1 \Phi_1 &= m_2 \ell^2 + c_2^2 I_{yy_2} + s_2^2 I_{xx_2} + 2c_2 s_2 I_{xy_2} \\ &\quad + 2\ell c_2 h_{x_2} - 2\ell s_2 h_{y_2} \end{aligned}$$

This term staying constant with changes in q_2 is equivalent to (68), and can be deduced from conditions on its first four derivatives with respect to q_2 .

C. Proof of theorem 1

We restate the theorem from the main text:

Theorem 1. (Main Result) Consider a serial-chain rigid-body system in the absence of gravity, with the following inertia transfer subspaces for each joint ($i \in \{1, \dots, N\}$):

$$\begin{aligned} \overline{\mathcal{T}}_i &= \{ \delta \mathbf{I}_1, \dots, \delta \mathbf{I}_N \in \mathbb{I} \mid \exists \delta \mathbf{I}_0 \in \mathbb{I}, \delta \mathbf{I}_j = \mathbf{0} \text{ if } j \notin \{i, i-1\}, \\ \delta \mathbf{I}_{i-1} &= -{}^i\mathbf{X}_{i-1}^\top(0) \delta \mathbf{I}_i {}^i\mathbf{X}_{i-1}(0), \\ \mathbf{0} &= \Phi_i^\top \delta \mathbf{I}_i \mathbf{V}_i, \\ \mathbf{0} &= \text{Trace} \left(\mathbf{W}_{i-1,j} {}^i\mathbf{X}_{i-1}^\top(0) \delta \mathbf{I}_i^{(k)} {}^i\mathbf{X}_{i-1}(0) \right) \\ &\quad \forall k = 1, \dots, 10, j = 1, \dots, n_{i-1}^w \} \end{aligned}$$

Then, the set of all structurally unobservable inertia changes is given by $\overline{\mathcal{T}}_1 \oplus \dots \oplus \overline{\mathcal{T}}_N$, where \oplus denotes a direct sum of vector subspaces.

Proof: Consider the set of inertia variations up to body i that don't change the kinetic energy:

$$\overline{\mathcal{N}}_i = \{ \delta \mathbf{I}_1, \dots, \delta \mathbf{I}_N \in \mathbb{I} \mid \delta T(\mathbf{q}, \dot{\mathbf{q}}) = 0 \forall \mathbf{q}, \dot{\mathbf{q}}, \delta \mathbf{I}_j = \mathbf{0} \text{ if } j > i \}$$

we show via induction that for $i = 0, \dots, N$

$$\overline{\mathcal{N}}_i = \bigoplus_{j=1}^i \overline{\mathcal{T}}_j \quad (69)$$

Base case ($i = 0$): (69) holds since $\overline{\mathcal{N}}_0$ is the set containing only the zero element.

Inductive step: We assume (69) holds at some $i-1$, and show this implies it holds at i . To do so, we consider perturbations $\delta\mathbf{I}_1, \dots, \delta\mathbf{I}_N$ such that $\delta\mathbf{I}_j = \mathbf{0}$ for all $j > i$. We will show that these perturbations don't change the kinetic energy if and only if they are in the set $\bigoplus_{j=1}^i \overline{\mathcal{T}}_j$. Now, not changing the kinetic energy requires $0 = \delta T = 1/2 \sum_{j=1}^i \mathbf{v}_j^\top \delta\mathbf{I}_j \mathbf{v}_j$. Then, using (3), the kinetic energy variation can be expressed as

$$\begin{aligned} \delta T = & \frac{1}{2} \sum_{j=1}^{i-2} (\mathbf{v}_j^\top \delta\mathbf{I}_j \mathbf{v}_j) \\ & + \mathbf{v}_{i-1}^\top \mathbf{X}_{i-1}^\top \delta\mathbf{I}_i \Phi_i \dot{q}_i + \frac{1}{2} \Phi_i^\top \delta\mathbf{I}_i \Phi_i \dot{q}_i^2 \\ & + \frac{1}{2} \mathbf{v}_{i-1}^\top [\delta\mathbf{I}_{i-1} + {}^i\mathbf{X}_{i-1}^\top \delta\mathbf{I}_i {}^i\mathbf{X}_{i-1}] \mathbf{v}_{i-1} \end{aligned} \quad (70)$$

Consider a linear change of variables for $\delta\mathbf{I}_{i-1}$:

$$\delta\mathbf{I}_{i-1} = \delta\mathbf{I}'_{i-1} - {}^i\mathbf{X}_{i-1}^\top(0) \delta\mathbf{I}_i {}^i\mathbf{X}_{i-1}(0) \quad (71)$$

which forms $\delta\mathbf{I}_{i-1}$ via an inertia transfer from the child plus an additional change $\delta\mathbf{I}'_{i-1}$. Under this change of variables:

$$\delta T = \frac{1}{2} \sum_{j=1}^{i-2} (\mathbf{v}_j^\top \delta\mathbf{I}_j \mathbf{v}_j) + \frac{1}{2} \mathbf{v}_{i-1}^\top \delta\mathbf{I}'_{i-1} \mathbf{v}_{i-1} \quad (72)$$

$$+ \mathbf{v}_{i-1}^\top {}^i\mathbf{X}_{i-1}^\top \delta\mathbf{I}_i \Phi_i \dot{q}_i + \frac{1}{2} \Phi_i^\top \delta\mathbf{I}_i \Phi_i \dot{q}_i^2 \quad (73)$$

$$+ \frac{1}{2} \mathbf{v}_{i-1}^\top \Delta_i(q_i) \mathbf{v}_{i-1} \quad (74)$$

where $\Delta_i(q_i) = {}^i\mathbf{X}_{i-1}^\top(q_i) \delta\mathbf{I}_i {}^i\mathbf{X}_{i-1}(q_i) - {}^i\mathbf{X}_{i-1}^\top(0) \delta\mathbf{I}_i {}^i\mathbf{X}_{i-1}(0)$.

Considering the cases when $q_i = 0$ and $\dot{q}_i = 0$ shows that δT will be identically zero if and only if (72)–(74) are all individually identically zero.

The terms in (72) being zero is equivalent to $\delta\mathbf{I}_1, \dots, \delta\mathbf{I}_{i-2}, \delta\mathbf{I}'_{i-1}$ giving $\delta T = 0$, which is the same as the perturbations being in $\overline{\mathcal{N}}_{i-1}$.

The terms from (73) being zero for all $\mathbf{q}, \dot{\mathbf{q}}$ is equivalent to $\mathbf{v}_i^\top \delta\mathbf{I}_i \Phi_i = 0$ holding for all $\mathbf{q}, \dot{\mathbf{q}}$, which, via the argument in the main body, is the same as $\Phi_i^\top \delta\mathbf{I}_i \mathbf{V}_i = 0$.

Finally, the terms from (74) being zero is equivalent to

$$0 = \frac{\partial}{\partial q_i} \mathbf{v}_{i-1}^\top [{}^i\mathbf{X}_{i-1}^\top(q_i) \delta\mathbf{I}_i {}^i\mathbf{X}_{i-1}(q_i)] \mathbf{v}_{i-1} \quad \forall \mathbf{q}, \dot{\mathbf{q}} \quad (75)$$

which, via the argument in the main body, is the same as $0 = \text{Trace}(\mathbf{W}_{i-1,j} {}^i\mathbf{X}_{i-1}^\top(0) \delta\mathbf{I}_i^{(k)} {}^i\mathbf{X}_{i-1}(0))$ for $k = 1, \dots, 10$, $j = 1, \dots, n_{i-1}^w$.

Thus, we've shown that a variation to the inertias $(\delta\mathbf{I}_1, \dots, \delta\mathbf{I}_N) \in \overline{\mathcal{N}}_i$ iff it can be written as a parameter variation in $\overline{\mathcal{N}}_{i-1}$ added with one in $\overline{\mathcal{T}}_i$. This proves the inductive step.

Considering (69) when $i = N$ proves the theorem.

D. Proof of Lemma 1

To prove the lemma in the main text, we begin with a proposition to compute the span of velocities that can be reached after a joint given a span of velocities before it.

Proposition 1. Consider a spatial transform as a function of a single angle q , denoted $\mathbf{X}(q)$. Suppose $\mathbf{X}(0) = \mathbf{I}$ and that

$$\frac{d}{dq} \mathbf{X}(q) = -(\Phi \times) \mathbf{X}(q) \quad (76)$$

for some $\Phi \in \mathbb{R}^{6 \times 1}$. Then, for any $\mathbf{V} \in \mathbb{R}^{6 \times k}$

$$\begin{aligned} \text{span}\{\mathbf{v} \mid \exists q \in \mathbb{R}, \mathbf{v} \in \text{Range}(\mathbf{X}(q) \mathbf{V})\} \\ = \text{Range}(\text{Ctrb}((\Phi \times), \mathbf{V})) \end{aligned} \quad (77)$$

where $\text{Ctrb}((\Phi \times), \mathbf{V})$ is the controllability matrix associated with the pair $((\Phi \times), \mathbf{V})$ (Rugh, 1996).

Proof of Proposition 1: We define

$$\mathcal{S}(\Phi, \mathbf{V}) = \text{span}\{\mathbf{X}(q)\mathbf{v} \mid q \in \mathbb{R}, \mathbf{v} \in \text{Range}(\mathbf{V})\}$$

and recall, from the Lemma statement, that

$$\frac{d}{dq} \mathbf{X}(q) = -(\Phi \times) \mathbf{X}(q)$$

From the definition of the matrix exponential for a linear system (Rugh, 1996):

$$\mathbf{X}(q)\mathbf{V} = e^{-q(\Phi \times)} \mathbf{V}$$

The Cayley–Hamilton theorem then ensures that

$$\mathcal{S}(\Phi, \mathbf{V}) \subseteq \text{Range}([\mathbf{V}, (\Phi \times)\mathbf{V}, \dots, (\Phi \times)^5 \mathbf{V}])$$

and thus

$$\mathcal{S}(\Phi, \mathbf{V}) \subseteq \text{Range}(\text{Ctrb}((\Phi \times), \mathbf{V}))$$

Note, the range of the controllability matrix provides the smallest $(\Phi \times)$ -invariant subspace containing $\text{Range}(\mathbf{V})$. Yet, $\mathcal{S}(\Phi, \mathbf{V})$ is invariant under $(\Phi \times)$ and contains $\text{Range}(\mathbf{V})$. This proves the reverse containment.

Proof of Lemma 1: The propagation of the attainable velocity span:

$$\mathbf{V}_i = [\text{Ctrb}((\Phi_i \times), {}^i\mathbf{X}_{i-1}(0) \mathbf{V}_{i-1}) \Phi_i]$$

follows from Proposition 1 and equations (3) and (11).

E. Proof of Lemma 2

Before proving (2), to streamline the equations, we introduce a hat operator $[\cdot]^\wedge$ that maps inertial parameters to a 6D inertia matrix:

$$[\pi]^\wedge = \mathbf{I}(\pi)$$

The following proposition is key to proving Lemma 2.

Proposition 2. Consider a spatial transform as a function of a single angle q , denoted $\mathbf{X}(q)$. Suppose $\mathbf{X}(0) = \mathbf{1}$ and that there exists $\Phi \in \mathbb{R}^{6 \times 1}$ such that

$$\frac{d}{dq} \mathbf{X}(q) = -(\Phi \times) \mathbf{X}(q)$$

Then, for any $\mathbf{C} \in \mathbb{R}^{k \times 10}$, the following holds

$$\begin{aligned} \{ \pi \in \mathbb{R}^{10} \mid \mathbf{C}[\mathbf{X}^\top(q)[\pi]^\wedge \mathbf{X}(q)]^\vee = \mathbf{0}, \forall q \in \mathbb{R} \} \\ = \text{Null}(\text{Obs}(\mathbf{C}, \mathbf{A}(\Phi))) \end{aligned}$$

where $\text{Obs}(\mathbf{C}, \mathbf{A}(\Phi))$ is the observability matrix (Rugh, 1996) associated with the pair $(\mathbf{C}, \mathbf{A}(\Phi))$.

Proof of Proposition 2: Let $\pi_0 \in \mathbb{R}^{10}$ and denote

$$\pi(q) = [\mathbf{X}^\top(q)[\pi_0]^\wedge \mathbf{X}(q)]^\vee$$

Using (11), (40), and the fact that $\mathbf{X}(q)$ and $(\Phi \times)$ commute

$$\begin{aligned} \frac{d}{dq} \pi(q) &= -[(\Phi \times)^\top [\pi(q)]^\wedge + [\pi(q)]^\wedge (\Phi \times)]^\vee \\ &= -\mathbf{A}(\Phi) \pi(q) \end{aligned}$$

Linear systems observability results (Rugh, 1996) then guarantee that the following are equivalent

$$\pi_0 \in \text{Null}(\text{Obs}(\mathbf{C}, \mathbf{A}(\Phi))) \Leftrightarrow \pi(q) \in \text{Null}(\mathbf{C}) \quad \forall q$$

Proof of Lemma 2: We aim to construct \mathbf{K}_i so that $\text{Range}(\mathbf{K}_i) = \mathcal{K}_i = \text{span}(\{\nabla_{\mathbf{v}} \mathbf{v}^\top \mathbf{I}(\pi) \mathbf{v} \mid \mathbf{v} \in \mathcal{V}_i^*\})$. Equivalently, we construct \mathbf{K}_i^\top so that $\text{Null}(\mathbf{K}_i^\top) = \mathcal{K}_i^\perp$.

Suppose $\delta \pi_i \in \mathcal{K}_i^\perp$. That is, we consider a corresponding $\delta \mathbf{I}_i$ such that $\mathbf{v}_i^\top \delta \mathbf{I}_i \mathbf{v}_i = 0 \quad \forall \mathbf{q}, \dot{\mathbf{q}}$. This is equivalent to

$$[{}^i \mathbf{X}_{i-1} \mathbf{v}_{i-1} + \Phi_i \dot{q}_i]^\top \delta \mathbf{I}_i [{}^i \mathbf{X}_{i-1} \mathbf{v}_{i-1} + \Phi_i \dot{q}_i] = 0$$

for all $\mathbf{q}, \dot{\mathbf{q}}$. Expanding terms, this implies

$$0 = \mathbf{v}_{i-1}^\top {}^i \mathbf{X}_{i-1}^\top(q_i) \delta \mathbf{I}_i {}^i \mathbf{X}_{i-1}(q_i) \mathbf{v}_{i-1} \quad \forall \mathbf{q}, \dot{\mathbf{q}} \quad (78)$$

$$0 = \Phi_i^\top \delta \mathbf{I}_i \mathbf{v}_i \quad \forall \mathbf{q}, \dot{\mathbf{q}} \quad (79)$$

We take the transformation ${}^i \mathbf{X}_{i-1}(q_i)$ from frame $i-1$ to i , and instead decompose it using a fixed transformation to a frame J_i right before the joint so that ${}^J_i \mathbf{X}_{i-1} = {}^i \mathbf{X}_{i-1}(0)$. With this definition, we express ${}^i \mathbf{X}_{i-1}(q_i) = {}^i \mathbf{X}_{J_i}(q_i) {}^J_i \mathbf{X}_{i-1}$. The first condition (78) above is then equivalent to

$$\mathbf{0} = \mathbf{K}_{i-1}^\top {}^{i-1} \mathbf{B}_i [{}^i \mathbf{X}_{J_i}^\top(q_i) \delta \mathbf{I}_i {}^i \mathbf{X}_{J_i}(q_i)]^\vee \quad \forall q_i$$

and the second (79) equivalent to

$$\mathbf{M}(\mathbf{V}_i, \Phi_i)^\top \delta \pi_i = \mathbf{0}$$

where $\mathbf{M}(\mathbf{V}_i, \Phi_i)$ is given by (36). Using Proposition 2, it follows that \mathbf{K}_i^\top can be selected as

$$\mathbf{K}_i^\top = \begin{bmatrix} \text{Obs}(\mathbf{K}_{i-1}^\top {}^{i-1} \mathbf{B}_i, \mathbf{A}(\Phi_i)) \\ \mathbf{M}(\mathbf{V}_i, \Phi_i)^\top \end{bmatrix}$$

F. Identifiability from gravity

Similar to the nullspace for the kinetic energy, parameter variations ensuring $\delta \mathbf{g} \equiv \mathbf{0}$ can be formed via sequences of inertia transfers. The variation $\delta \mathbf{g}$ to the generalized gravitational force is equal to zero if and only if it does not affect the rate of change in potential energy \dot{V} :

$$0 = \delta \dot{V} = \dot{\mathbf{q}}^\top (\delta \mathbf{g}) = - \sum_{j=1}^N \mathbf{v}_j^\top \delta \mathbf{I}_j {}^j \mathbf{X}_0 {}^0 \mathbf{a}_g$$

where ${}^0 \mathbf{a}_g$ is the spatial acceleration due to gravity. Each entry of the sum characterizes a change in the instantaneous power of the gravitational force on Body j . We again assume that body i is the largest body with $\delta \mathbf{I}_i \neq \mathbf{0}$, and follow a similar approach to the kinetic energy nullspace analysis. Following an equivalent derivation to Appendix C it can be shown that $\delta \mathbf{g} = \mathbf{0}$ iff

$$\mathbf{0} = \Phi_i^\top \delta \mathbf{I}_i {}^i \mathbf{X}_0 {}^0 \mathbf{a}_g \quad (80)$$

$$0 = \mathbf{v}_{i-1}^\top \Lambda_i(q_i) {}^{i-1} \mathbf{X}_0 {}^0 \mathbf{a}_g \quad (81)$$

for all $\mathbf{q}, \dot{\mathbf{q}}$, and subsequent changes $\delta \mathbf{I}_1, \dots, \delta \mathbf{I}_{i-2}, \delta \mathbf{I}_{i-1}'$ independently satisfy $\delta \mathbf{g} = \mathbf{0}$. Similar to before, the substitution introducing $\delta \mathbf{I}_{i-1}'$ via (71) decouples considerations of transfers across joint i from transfers earlier in the chain.

Condition (80) motivates the attainable gravity vector span

$$\mathcal{G}_i = \text{span}\{ {}^i \mathbf{X}_0(\mathbf{q}) {}^0 \mathbf{a}_g \mid \mathbf{q} \in \mathbb{R}^N \}$$

Analogous to Lemma 1, we seed $\mathbf{G}_0 = {}^0 \mathbf{a}_g$, let $\mathbf{G}_{i-1+} = {}^i \mathbf{X}_{i-1}(0) \mathbf{G}_{i-1}$, and recursively apply

$$\mathbf{G}_i = \text{Ctrb}((\Phi_i \times), \mathbf{G}_{i-1+})$$

which ensures each $\text{Range}(\mathbf{G}_i) = \mathcal{G}_i$. Intuitively, changes satisfying $\Phi_i^\top \delta \mathbf{I}_i \mathbf{G}_i = \mathbf{0}$ cannot be detected via the preceding joint torque in static cases.

The second condition (81) can be addressed by generalizing the propagation of \mathbf{K}_i from Lemma 2 to include gravitational effects. This extension can be accomplished by including the new parameters that are identified via torques on each joint:

$$\mathbf{K}_i^\top = \begin{bmatrix} \text{Obs}(\mathbf{K}_{i-1}^\top {}^{i-1} \mathbf{B}_i, \mathbf{A}(\Phi_i)) \\ \mathbf{M}([{}^i \mathbf{V}_i \mathbf{G}_i], \Phi_i)^\top \end{bmatrix}$$

Comparing the propagation of \mathbf{G}_i and \mathbf{V}_i

$$\mathbf{V}_0 = \mathbf{0} \quad \mathbf{V}_i = [\text{Ctrb}((\Phi_i \times), \mathbf{V}_{i-1+}) \quad \Phi_i] \quad (82)$$

$$\mathbf{G}_0 = {}^0 \mathbf{a}_g \quad \mathbf{G}_i = \text{Ctrb}((\Phi_i \times), \mathbf{G}_{i-1+}) \quad (83)$$

Figure 11. An alternate proof of the main result from Ayusawa et al. (2014) uses mechanics principles. The implication from the blue arrow is replaced by those in green.

where \mathbf{I}_i^C is the composite rigid-body inertia (Featherstone and Orin, 2008) of the subchain of bodies rooted at body i . Indeed, one may be able to see that if a variation in inertias does not change the first block row of \mathbf{H} , then it will not change \mathbf{H} overall. We proceed to justify this point alternatively via physical arguments.

Let us consider the case when only joint j is moving. In that case, we have from (86) that $\mathbf{H}_1 \ddot{\mathbf{q}} = {}^j\mathbf{X}_1^\top \mathbf{I}_j^C \Phi_j \ddot{q}_j$, which represents the total wrench required to move all bodies in the system. By comparison, consider the joint torque at any joint i earlier in the chain ($i < j$), given by

$$\tau_i = \mathbf{H}_{ij} \ddot{q}_j = \Phi_i^\top {}^j\mathbf{X}_i^\top \mathbf{I}_j^C \Phi_j \ddot{q}_j = ({}^1\mathbf{X}_i \Phi_i)^\top \mathbf{H}_1 \ddot{\mathbf{q}}$$

In this case, τ_i is simply the projection of the net wrench onto the i th joint. As such, the i th joint torque carries strictly less information than $\mathbf{H}_1 \ddot{\mathbf{q}}$. By this logic, if

$$\delta[\mathbf{H}(\mathbf{q})\ddot{\mathbf{q}}] = \mathbf{0} \quad \forall \mathbf{q}, \ddot{\mathbf{q}} \quad (87)$$

then the upper triangle of $\delta\mathbf{H}(\mathbf{q})$ will be zero for all \mathbf{q} . Via symmetry of \mathbf{H} , this implies that $\delta\mathbf{H}(\mathbf{q}) = \mathbf{0}$ for all \mathbf{q} . This reasoning shows that C implies D.²

We finally argue the equivalence of A and D in the figure. Per Remark 9, we can ignore gravity, so $\delta\pi$ leaves the dynamics unchanged if and only if

$$\delta[\mathbf{H}(\mathbf{q})\ddot{\mathbf{q}} + \mathbf{c}(\mathbf{q}, \dot{\mathbf{q}})] = \mathbf{0} \quad \forall \mathbf{q}, \dot{\mathbf{q}}, \ddot{\mathbf{q}} \quad (88)$$

since the Coriolis terms are determined uniquely from the form of the mass matrix (e.g., via Christoffel symbols (Echeandia and Wensing, 2021; Siciliano et al., 2008)), (88) is equivalent to requiring that

$$\delta[\mathbf{H}(\mathbf{q})\ddot{\mathbf{q}}] = \mathbf{0} \quad \forall \mathbf{q}, \ddot{\mathbf{q}} \quad (89)$$

This proves the equivalence of A and D in the figure, which completes the proof.

We hope this conceptual but physically grounded proof will help enhance the appeal of the original main result in Ayusawa et al. (2014) for a broader range of readers.

I. Generalization of RPNA to address rotors

The momentum and torque conditions (54) and (55) are generalized here in the case of a serial chain:

$$\mathbf{0} = [\Phi_i^\top \delta \mathbf{I}_i {}^i\mathbf{X}_{i-1}(q_i) + n_R \Phi_{m_i}^\top \delta \mathbf{I}_{m_i} {}^{m_i}\mathbf{X}_{i-1}(q_i)] \mathbf{v}_{i-1} \quad (90)$$

$$\quad \forall \mathbf{v}_{i-1} \in \mathcal{V}_{i-1}^*, q_i \in \mathbb{R}$$

$$\mathbf{0} = \Phi_i^\top \delta \mathbf{I}_i \Phi_i + \Phi_{m_i}^\top \delta \mathbf{I}_{m_i} \Phi_{m_i} n_R^2 \quad (91)$$

we can collect these two conditions together into a larger equation:

$$\mathbf{0} = \begin{bmatrix} \Phi_i \\ n_R \Phi_{m_i} \end{bmatrix}^\top \begin{bmatrix} \delta \mathbf{I}_i & \mathbf{0} \\ \mathbf{0} & \delta \mathbf{I}_{m_i} \end{bmatrix} \left(\begin{bmatrix} {}^i\mathbf{X}_{i-1} \\ {}^{m_i}\mathbf{X}_{i-1} \end{bmatrix} \mathbf{v}_{i-1} + \begin{bmatrix} \Phi_i \\ n_R \Phi_{m_i} \end{bmatrix} \dot{q}_i \right) \quad (92)$$

for all \mathbf{v}_{i-1} and \dot{q}_i . To simplify these conditions, we define

$$\underline{\Phi}_i = \begin{bmatrix} \Phi_i \\ n_R \Phi_{m_i} \end{bmatrix} \quad (93)$$

$$\delta \underline{\mathbf{I}}_i = \begin{bmatrix} \delta \mathbf{I}_i & \mathbf{0} \\ \mathbf{0} & \delta \mathbf{I}_{m_i} \end{bmatrix} \quad (94)$$

$${}^i\mathbf{X}_{i-1}(q_i) = \begin{bmatrix} {}^i\mathbf{X}_{i-1}(q_i) \\ {}^{m_i}\mathbf{X}_{i-1}(q_i) \end{bmatrix} \quad (95)$$

and assume a matrix $\underline{\mathbf{V}}_i$ so that

$$\text{Range}(\underline{\mathbf{V}}_i) = \text{span}(\{ {}^i\mathbf{X}_{i-1}(q_i) \mathbf{v}_{i-1} + \underline{\Phi}_i \dot{q}_i \mid \quad (96)$$

$$\mathbf{v}_{i-1} \in \mathcal{V}_{i-1}^*, q_i, \dot{q}_i \in \mathbb{R} \}) \quad (97)$$

with these definitions, we re-express the momentum and torque conditions in a combined form as

$$\underline{\Phi}_i^\top \delta \underline{\mathbf{I}}_i \underline{\mathbf{V}}_i = \mathbf{0}$$

Additionally, we can recover a choice of \mathbf{V}_i as

$$\mathbf{V}_i = [\mathbf{1}_{6 \times 6} \quad \mathbf{0}_{6 \times 6}] \underline{\mathbf{V}}_i \quad (98)$$

A remaining question regards how to construct $\underline{\mathbf{V}}_i$, however, this can be done using controllability analysis similar to how \mathbf{V}_i was constructed in the main text.

$$\underline{\mathbf{V}}_{i-1+} = {}^i\mathbf{X}_{i-1}(0) \mathbf{V}_{i-1}$$

$$\underline{\mathbf{V}}_{i-} = \text{Ctrb} \left(\begin{bmatrix} (\Phi_i \times) & \mathbf{0} \\ \mathbf{0} & n_R(\Phi_{m_i} \times) \end{bmatrix}, \underline{\mathbf{V}}_{i-1+} \right)$$

$$\underline{\mathbf{V}}_i = [\underline{\mathbf{V}}_{i-} \quad \underline{\Phi}_i]$$

which motivates the definition:

$$\underline{\Phi}_i \times = \begin{bmatrix} (\Phi_i \times) & \mathbf{0} \\ \mathbf{0} & n_R(\Phi_{m_i} \times) \end{bmatrix}$$

While we used the rotational symmetry of the rotor in the main text to simplify the mapping condition, we can discharge that assumption to improve the generality of the RPNA with rotors. Specifically, we consider $\underline{\pi} \in \mathbb{R}^{20}$ to represent both rotor and link inertia parameters. We then extend the definitions in the main text using $\underline{\mathbf{V}} \in \mathbb{R}^{12 \times m}$, $\underline{\mathbf{I}}(\underline{\pi}) \in \mathbb{R}^{12 \times 12}$, $\underline{\mathbf{v}} \in \mathbb{R}^{12}$, $\underline{\mathbf{X}} \in \mathbb{R}^{12 \times 6}$ as

$$\underline{\mathbf{M}}(\underline{\mathbf{V}}, \underline{\Phi}) := \left[\frac{\partial}{\partial \underline{\pi}} \underline{\mathbf{V}}^\top \underline{\mathbf{I}}(\underline{\pi}) \underline{\Phi} \right]^\top$$

$$\underline{\mathbf{k}}(\underline{\mathbf{v}}) := \nabla_{\frac{1}{\pi_2} \underline{\mathbf{v}}}^\top \underline{\mathbf{I}}(\underline{\pi}) \underline{\mathbf{v}}$$

$$\underline{\mathbf{B}}(\underline{\mathbf{X}}) := \frac{\partial}{\partial \underline{\pi}} [\underline{\mathbf{X}}^\top \underline{\mathbf{I}}(\underline{\pi}) \underline{\mathbf{X}}]^\vee$$

$$\underline{\mathbf{A}}(\underline{\Phi}) := \frac{\partial}{\partial \underline{\pi}} [(\underline{\Phi} \times)^\top \underline{\mathbf{I}}(\underline{\pi}) + \underline{\mathbf{I}}(\underline{\pi})(\underline{\Phi} \times)]^\vee$$

we consider a matrix $\underline{\mathbf{K}}_i$ so that

$$\text{Range}(\underline{\mathbf{K}}_i) = \text{span}(\{\underline{\mathbf{k}}(\underline{\mathbf{v}}_i) \mid \underline{\mathbf{v}}_i = {}^i\mathbf{X}_{i-1}(q_i) \underline{\mathbf{v}}_{i-1} + \underline{\Phi}_i \dot{q}_i \mid \underline{\mathbf{v}}_{i-1} \in \mathcal{V}_{i-1}^*, q_i, \dot{q}_i \in \mathbb{R}\})$$

with these definitions, the momentum and torque conditions take the form:

$$\underline{\mathbf{M}}(\underline{\mathbf{V}}_i, \underline{\Phi}_i)^\top \delta \underline{\boldsymbol{\pi}}_i = 0$$

while the mapping condition takes the form:

$$\underline{\mathbf{K}}_i^\top \mathbf{B}({}^i\mathbf{X}_{i-1}(0)) \underline{\mathbf{A}}(\underline{\Phi}_i)^k \delta \underline{\boldsymbol{\pi}}_i = \mathbf{0}, k = 1, \dots, 10$$

Mirroring the development in the main text, we can consider

$$\underline{\mathbf{K}}_{i-1+} = \underline{\mathbf{B}}({}^i\mathbf{X}_{i-1}(0))^\top \underline{\mathbf{K}}_{i-1}$$

$$\underline{\mathbf{K}}_{i-} = \text{Ctrb}(\underline{\mathbf{A}}(\underline{\Phi}_i)^\top, \underline{\mathbf{K}}_{i-1+})$$

$$\underline{\mathbf{K}}_i = [\underline{\mathbf{K}}_{i-} \underline{\mathbf{M}}(\underline{\mathbf{V}}_i, \underline{\Phi}_i)]$$

$$\underline{\mathbf{N}}_i = \begin{bmatrix} \underline{\mathbf{M}}(\underline{\mathbf{V}}_i, \underline{\Phi}_i)^\top \\ \underline{\mathbf{K}}_{i-}^\top \underline{\mathbf{A}}(\underline{\Phi}_i) \end{bmatrix}$$

where we then compute

$$\mathbf{K}_i = [\mathbf{1}_{10 \times 10} \quad \mathbf{0}_{10 \times 10}] \underline{\mathbf{K}}_i \quad (99)$$

for the link alone. Aside from the extra steps (98) and (99), the rest of these computations take the same structure as the RPNA in the main text, allowing us to quickly generalize its operation to accommodate single-DoF joints with rotors.

Remark 12. *This approach can be directly extended to the case when m bodies ($m \geq 2$) move with q_i as long as their velocities $\underline{\mathbf{v}}_i \in \mathbb{R}^{6m}$ can be represented in the form*

$$\underline{\mathbf{v}}_i = {}^i\mathbf{X}_{i-1}(q_i) \underline{\mathbf{v}}_{i-1} + \underline{\Phi}_i \dot{q}_i$$

for some fixed $\underline{\Phi}_i$.