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A Tunable, Simplified Model for Biological Latch Mediated Spring Actuated Systems

Andrés Cook,* Kaanthi Pandhigunta,* Mason A. Acevedo,* Adam Walker,* Rosalie L. Didcock,* Jackson T. Castro,* Declan O’Neill,* Raghav Acharya,† M. Saad Bhamla †,† Philip S. L. Anderson †, † and Mark Ilton †, †

*Department of Physics, Harvey Mudd College, Claremont, CA 91711; †School of Chemical and Biomolecular Engineering, Georgia Institute of Technology, Atlanta, Georgia 30318; ‡Department of Evolution, Ecology, and Behavior, University of Illinois at Urbana-Champaign, Urbana, IL 61801

1E-mail: milton@g.hmc.edu

Synopsis We develop a model of latch-mediated spring actuated (LaMSA) systems relevant to comparative biomechanics and bioinspired design. The model contains five components: two motors (muscles), a spring, a latch, and a load mass. One motor loads the spring to store elastic energy and the second motor subsequently removes the latch, which releases the spring and causes movement of the load mass. We develop freely available software to accompany the model, which provides an extensible framework for simulating LaMSA systems. Output from the simulation includes information from the loading and release phases of motion, which can be used to calculate kinematic performance metrics that are important for biomechanical function. In parallel, we simulate a comparable, directly actuated system that uses the same motor and mass combinations as the LaMSA simulations. By rapidly iterating through biologically relevant input parameters to the model, simulated kinematic performance differences between LaMSA and directly actuated systems can be used to explore the evolutionary dynamics of biological LaMSA systems and uncover design principles for bioinspired LaMSA systems. As proof of principle of this concept, we compare a LaMSA simulation to a directly actuated simulation that includes either a Hill-type force-velocity trade-off or muscle activation dynamics, or both. For the biologically-relevant range of parameters explored, we find that the muscle force-velocity trade-off and muscle activation have similar effects on directly actuated performance. Including both of these dynamic muscle properties increases the accelerated mass range where a LaMSA system outperforms a directly actuated one.

Introduction

A diverse array of organisms use stored elastic energy to drive rapid movements. These organisms use motors, springs, and latches to perform a latch mediated spring actuated (LaMSA) motion, and remarkably, they can use this mechanism to outperform current engineering design for repeatable motion at small size-scales (Longo et al. 2019). Models have been developed to understand the extreme biomechanics of LaMSA organisms. Organism-specific models, including both continuum mechanics-based models (Liu et al. 2017; Cooper et al. 2018; Larabee et al. 2018; Tadayon et al. 2018; Bolmin et al. 2019; Berg et al. 2019; Hamlet et al. 2020; Li et al. 2020; Wan and Hao 2020) and physical modeling with biomimetic devices (Cox et al. 2014; Liu et al. 2017; Xu and Bhamla 2019; Li et al. 2020; Singh et al. 2020; Büsse et al. 2021), have been used to test hypotheses about the movement of specific organisms (Table 1 summarizes examples of recent work).

In contrast to organism-specific models, “simple models” with reduced complexity (Anderson et al. 2020) are primarily used for making inter-species comparisons, and for testing scaling relationships and the sensitivity of kinematic performance to different characteristics of the organism. These simple models can also have broad applicability and enable the rapid testing of ideas (Anderson et al. 2020), and typically include muscle motors, springs, masses, and other mechanical linkages. In recent work, these models have been applied to jumping organisms...
Table 1 Recent examples (since 2018) of modeling LaMSA organisms, which includes both mathematical and physical approaches. For a review of earlier work see ref. (Ilton et al. 2018).

| Modeling approach  | Biomechanical system                                   | Reference |
|--------------------|--------------------------------------------------------|-----------|
| Continuum mechanics| click beetle latch                                      | Bolmin et al. (2019) |
|                    | bladderwort trap suction feeding                        | Berg et al. (2019) |
| Fluid dynamics     | *Ruellia ciliatflora* seed aerodynamics                 | Cooper et al. (2018) |
|                    | nematocyst discharge                                    | Hamlet et al. (2020) |
| Finite elements    | locust jump                                             | Wan and Hao (2020) |
|                    | dracula ant mandible strike                             | Larabee et al. (2018) |
|                    | mantis shrimp strike                                    | Tadayon et al. (2018) |
|                    | *Oxalis* sp. seed ejection                              | Li et al. (2020) |
| Physical modeling  | *Oxalis* sp. seed ejection                              | Li et al. (2020) |
|                    | bladderwort trap suction feeding                        | Singh et al. (2020) |
|                    | dragonfly larvae strike                                 | Büss e et al. (2021) |
|                    | *Spirostomum ambiguus* contraction                      | Xu and Bhamla (2019) |

(Davranoglou et al. 2019; Jarur et al. 2019; Niechcial et al. 2019; Olberding et al. 2019; Sutton et al. 2019; Hong et al. 2020; Mo et al. 2020; Zhang et al. 2020) and augmented human movements (Sutrisno and Braun 2019, 2020). General models have also been used to test hypotheses about the scaling and effectiveness of biological spring mechanisms (Galantis and Woledge 2003; Ilton et al. 2018; Abbott et al. 2019; Sutton et al. 2019; Divi et al. 2020). These types of models have similarities to template models—simple biomechanical models that demonstrate a particular mechanical behavior (Full and Koditschek 1999).

Previous work used a simplified mathematical model to illustrate trade-offs between the components of a general LaMSA system (Ilton et al. 2018). The components of a LaMSA system (the latch, spring, loading motor, and load mass) were modeled as a simplified mechanical system and given material, geometric, and dynamic properties; however, the properties of the system components were limited to motors and springs with linear properties, specific latch shapes, frictionless interactions between components, and a fixed unlatching velocity.

Here, we develop a LaMSA Template Model with accompanying software. Our model here includes a more general framework for defining LaMSA components, such that previous LaMSA modeling efforts (Galantis and Woledge 2003; Ilton et al. 2018; Sutton et al. 2019; Divi et al. 2020) are all particular cases of this new model. This broader approach allows the model to be tuned to a specific organism, group of organisms, or a biological scaling relationship to explore questions in comparative biomechanics and LaMSA system design. Our approach also includes non-linear and time-dependent properties of the spring material during unloading. Additionally, we provide a generalized treatment of the latch that includes friction, allows for different latch shapes, and includes an unlatching motor that drives the latch removal of the system, similar to the one recently hypothesized to occur in some biological systems (Büss e et al. 2021).

Finally, as an example of this LaMSA model’s utility, we use the model to explore how dynamic muscle properties affect the power output of both a LaMSA system and a system where the muscle is used to directly actuate movement. Two important dynamic aspects of muscle are a force-velocity trade-off (the muscle exerts less force at higher velocities) and an activation rate (it takes some time for the muscle to reach its maximum force) (Rosario et al. 2016). Previous work has been focused on how muscle force-velocity trade-offs limit power output for a directly actuated system (Galantis and Woledge 2003; Ilton et al. 2018). This force-velocity trade-off is a principal reason LaMSA systems can outperform comparable muscle-driven ones at small load mass; however, it is unclear how significant this force-velocity effect is compared to the activation dynamics of muscle. Here, we directly compare the effect of the muscle force-velocity trade-off to the effect of muscle activation. Using the LaMSA Template Model with
inputs guided by biologically-relevant sizes and masses, we find that the muscle force-velocity trade-off and activation dynamics cause a similar reduction in directly actuated kinematics. Combining the two effects, the mass range where a LaMSA system outperforms a directly actuated one increases by a factor of ≈5 times compared to systems where only one of the two time dependent motor properties is included.

**Methods**

**LaMSA Template Model**

In our model, the motion of a LaMSA system is composed of three distinct phases: loading, unlatching, and spring actuation. In the loading phase (Fig. 1A, first panel), a loading motor (e.g., muscle) deforms a spring starting from the spring's stress-free equilibrium length. We make the simplifying assumption that the loading occurs slowly enough to approximate it as a quasi-static motor contraction—that is, the loading follows the isometric force-length curve in the case of a muscle motor. Loading is complete when the loading motor force pulling down (in the $-y$ direction) matches the spring force pulling up. After the loading phase, the loading motor remains at a fixed displacement and the spring is held in place by a latch (Fig. 1A, second panel). The second phase of motion, the unlatching phase (Fig. 1A, third panel), begins with the activation of an unlatching motor that pulls the latch out of the way. During the unlatching phase the load mass and latch undergo a complex interaction. The interaction between the load mass and latch is modeled as a frictional contact between two rigid bodies, and the unlatching phase ends when there is no longer any contact between the load mass and latch. Once the contact breaks, the load mass is actuated solely by the spring, which undergoes a rapid unloading (Fig. 1A, fourth panel). Spring actuation continues until the spring returns to its equilibrium length where it no longer applies a force to the load mass, which corresponds to the "take-off" of the load mass (Fig. 1A, fifth panel). In the model, we assume that the latch shape is sufficiently smooth that after the latch disengages, it does not re-engage at a later time. This assumption enables the clear delineation of the unlatching and spring actuation phases.

The dynamics of a LaMSA system depends on its components and the interaction between them. In our model, these components are classified into motors, springs, latches, and load masses (Fig. 1B). Each motor is constrained to move along a single coordinate axis in the model (the loading motor moves along the $y$ axis; the latch and unlatching motor move along the $x$ axis). We develop our model with the aim to give general properties to each component. The motors and springs in the LaMSA system are characterized by their force output. The loading motor force ($F_{lm}$), the
unlatching motor force \((F_{um})\), and spring force \((F_{sp})\) are all assumed to be functions of time, displacement, and velocity. Latches are given a shape function \(y_l(x)\) that describes the geometry of the latch. The shape function relates horizontal motion of the latch (in the \(x\) direction) to vertical displacements of the load (in the \(y\) direction). For example, the rounded latch used in this work, which has circular edges of radius \(R\), has a shape function shown in Fig. 1B. The shape function describes the shape of the latch where it contacts the load mass. The derivatives of this shape function with respect to \(x\) determine the latch slope function \(y_l'(x)\) and latch concavity \(y_l''(x) = \frac{d^2 y_l}{dx^2}\). The functions describing shapes and forces are taken as inputs into the model to allow for hypothesis testing of non-linear properties. In addition, the mass of the system can be distributed in the spring mass, latch mass, and load mass. With these definitions, we lay out the mathematical description of the model according to its three phases of motion.

LaMSA Template Model: Loading Phase

In the loading phase, the loading motor slowly applies a force causing a displacement of the spring. The final displacement of the spring at the end of the loading phase, \(y_0\), is the displacement in which the loading motor force and spring force are equal and opposite, namely

\[
F_{um}(t = \infty, y_0, \dot{y} = 0) = -F_{sp}(t = \infty, y_0, \dot{y} = 0),
\]

where \(\dot{y}\) is the velocity in the \(y\) direction. The condition that \(t = \infty\) and \(\dot{y} = 0\) corresponds to a slow, quasi-static loading of the spring. The loaded displacement, \(y_0\), depends on how the force-displacement properties of the loading motor and spring interact.

LaMSA Template Model: Unlatching Phase

The unlatching phase starts with the activation of the unlatching motor at time \(t = 0\). The spring starts with an initial displacement \(y_0\) and velocity \(\dot{y} = 0\), while the latch has an initial horizontal position \(x = 0\) and velocity \(\dot{x} = \dot{v}_0\). By analyzing the spring force acting on the load mass, the unlatching motor force pulling on the latch, and the contact force between the load mass and latch, we derive that the differential equation for the acceleration of the latch, \(\ddot{x}\), during the unlatching phase of motion

\[
\ddot{x} = \frac{(F_{um} + F_{sp}y_l' - m_{eff}y_l'\dot{x}^2) + \mu_k(F_{um}y_l' - F_{sp} + m_{eff}y_l'\dot{x}^2)}{(m_l + m_{eff}y_l') - \mu_k(m_{eff}y_l' - m_l y_l')},
\]

where \(\mu_k\) is the coefficient of friction between the latch and load, and \(m_l\) is the mass of the latch. The term \(m_{eff}\) in Equation (2) is the overall effective mass for the mass-spring system, with \(m_{eff} = m_{load} + m_l/3\) (Ilton et al. 2018), where \(m_l\) is the spring mass and \(m_{load}\) is the effective load mass that depends on load mass and its effective mechanical advantage (EMA). A full derivation of Equation (2) is provided in Appendix B for a system undergoing small angular displacements. For a LaMSA system undergoing large angular displacements during rotational motion, the effective mass and mapping onto Equation (2) is provided in Appendix C. From the dynamics and shape of the latch, the acceleration of the load mass during the unlatching phase is given by the chain rule,

\[
\dot{y} = y_l'\ddot{x}^2 + y_l'\ddot{x}.
\]

To determine the end of the unlatching phase, we solve for the magnitude of the normal component of the contact force between the load mass and latch,

\[
F_N = \frac{-m_l F_{sp} + m_l m_{eff}y_l'\dot{x}^2 + m_{eff}y_l'F_{um}}{m_{eff}y_l'\mu_k - m_{eff}y_l'\dot{x}^2 - m_l \mu_k y_l' - m_l \mu_k \sqrt{1 + (y_l')^2}},
\]

and require that this magnitude be non-negative during the unlatching phase to ensure there is still contact between the load mass and latch. Therefore, we solve for when \(F_N = 0\) to determine the unlatching duration \(t_u\), which marks the end of the unlatching phase and the beginning of the spring actuation phase of motion.

LaMSA Template Model: Spring Actuation Phase

After unlatching, the load mass undergoes a purely spring-driven motion given by

\[
\ddot{y} = \frac{F_{sp}}{m_{eff}},
\]

where the spring force can depend on position, velocity, and time. The initial conditions for this phase are given by the ending condition from the unlatching phase: for the spring actuation phase, the initial position of the load mass is \(y(t = t_u)\), and its initial velocity is \(\dot{y}(t = t_u)\). The spring actuation phase ends when the spring stops pushing on the load mass, that is, when \(F_{sp} = 0\).

Direct Actuation Model

The direct actuation model uses the loading motor of the LaMSA system to directly drive the load mass. To ensure the motor in the directly actuated model is being used in a comparable way to the LaMSA model, the mass is accelerated by the motor using a motor contraction. Therefore, the equation of motion for the load
mass is given directly by the force applied by the motor as it contacts,

\[
\dot{y} = \frac{F_{\text{lm}}}{m_{\text{eff}}}, \tag{6}
\]

where the loading motor force can depend on position, velocity, and time. The initial condition for the directly actuated system is that the motor and load mass are initially at rest, with the motor at its undisplaced initial length. Take-off occurs when the load mass reaches its maximum velocity and \( F_{\text{lm}} = 0 \).

LaMSA and Direct Actuation Software Implementation

The LaMSA and direct actuation models were implemented in MATLAB. This software implementation is freely redistributable and available at https://posmlab.github.io (Didcock et al. 2020). The software allows a user to select a LaMSA system from a library of components (motors, springs, latches, and load masses), set parameters for each component, and run a simulation to determine the dynamics of that system (as both a LaMSA system and a directly actuated system). The software can be used to iterate over the LaMSA system component parameters (e.g., spring stiffness) and rapidly generate the dynamics for variety of LaMSA systems.

Model Input Parameters

The input parameters to the model were chosen based on the accelerated mass, characteristic velocities, and typical accelerations of the larger biological LaMSA systems listed in the supplementary materials of ref. (Ilton et al. 2018). To explore the role of the dynamic properties of muscle, we used a Hill-type muscle motor based on ref. (Rosario et al. 2016), which is one of the default components included in the LaMSA Template Model software. A muscle activation rate of 200 s\(^{-1}\) was chosen as a typical rate based on the force generation delay of small animals reported in ref. (More and Donelan 2018). The full list of parameters used in this work are reported in Table A1.

Results and Discussion

Using the components and parameters in Table A1, the output from a single simulation generated using the software is shown in Fig. 2. The software output includes information about the loading phase, and the dy-
of motors, the simulation generates the position $y(t)$, velocity $\dot{y}(t)$, and forces acting on the load mass. From the position and velocity of the load mass, commonly used metrics for kinematic performance in biomechanics (e.g., maximum acceleration and maximum power (Longo et al. 2019)) are calculated. The maximum load mass acceleration $(\max |\ddot{y}(t)|)$ and maximum power delivered to the load mass ($P_{\max} = \max (m y(t) \dot{y}(t))$) depend on the input parameters to the model, and the freely redistributable software enables a rapid iteration over a range of input parameters.

For a motor directly actuating a load mass, the maximum power output depends on accelerated mass, with an upper bound set by the dynamic properties of the motor (Fig. 3, red curves). Driving the mass with a motor that has only a force-velocity trade-off (setting $r_{\text{act}} = \infty$ and $v_{\max} = 5 \text{ m/s}$ in the model) has a similar effect to a motor that only has activation dynamics (setting $r_{\text{act}} = 200 \text{ s}^{-1}$ and $v_{\max} = \infty$ in the model). Both motors reach an upper bound on their maximum power output when driving small masses (Fig. 3, dashed and dotted red curves). Therefore, even in the absence of a force-velocity trade-off, motors with slow activation rates still have performance limitations when driving small masses. Including both the effects of force-velocity and activation in the motor, as projectile mass is decreased the maximum power output of a directly actuated movement not only saturates to a maximum value, but further decreases for the smallest masses (Fig. 3, solid red curve).

In contrast to the directly actuated systems, the LaMSA system is insensitive to the force-velocity trade-offs and activation dynamics of the loading motor. Varying the loading motor in the LaMSA system using the same three conditions as the directly actuated one (activation dynamics only, F-v trade-off only, F-v trade-off and activation dynamics), the maximum output for those three LaMSA systems is identical (Fig. 3, solid blue curve). The independence of the LaMSA system on the dynamic properties of the loading motor is a result of the slow, quasi-static loading assumption made in the model. This assumption is justified for biological LaMSA systems like mantis shrimp where typical loading rates are orders of magnitude slower than the rate of elastic energy release (Patek 2019), but the loading motor dynamic properties can be important when considering simultaneous loading and release of a series elastic system (Galantis and Woldge 2003).

The power output of comparable LaMSA and directly actuated systems have a mass dependent transition that is affected by the dynamic properties of the motor. Comparing the three different motor conditions in Fig. 3, the crossovers between the power output of the directly actuated and LaMSA systems is shifted to a larger mass (by a factor of $\approx 5$ times) when both the force-velocity and activation dynamics of the motor are included in the simulation. This result suggests that in systems where there is a development and transition of a LaMSA mechanism (e.g., in some species of mantis shrimp (Harrison et al. 2021)), care should be given to both muscle force-velocity and activation dynamics when modeling the transition from LaMSA to directly actuated movement.

Although the results of Fig. 3 were generated using general, biologically-relevant parameter values for LaMSA systems, a more specific biological system could be used to guide further inquiry into the relative importance of force-velocity versus muscle activation dynamics. For example, although here we assumed a fixed value of $\text{EMA} = 1$, the mechanical advantage in both muscle-driven and spring-driven systems can significantly alter dynamics (Richards and Clemente 2012; Olberding et al. 2019). Decreasing the EMA in the current model shifts the drop-off in muscle-driven performance to smaller masses. In addition, for most biologically-relevant systems maximum muscle force typically increases as system size increases. With a specific system in mind, appropriate scaling (Rospars and Meyer-Vernet 2016) and fair

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**Fig. 3** Both a motor’s activation rate and its force-velocity trade-off affect its maximum power output when it directly actuates a projectile. Compared to a using a motor in a LaMSA system (blue solid curve), the maximum power output of a directly actuated system (red curves) is worse for smaller masses. A motor that has both a force-velocity and activation limitation (solid red curve) has a significantly reduced performance at low masses compared to one with only a force-velocity trade-off (dashed red curve) or only an activation rate limitation (dotted red curve). The intersection between the LaMSA and directly actuated curve shifts to a higher mass when both dynamic effects of the motor are included.
comparisons (Ilton et al. 2019) could be made across size-scales for both motor-driven and elastically-driven systems.

Beyond this proof of principle example, the LaMSA Template Model and freely redistributable software provides an extensible platform for exploring biological LaMSA systems. Although this model was formulated generally to encompass a broad range of LaMSA systems, the model can be tuned to specific biological systems because of the flexibility in how system components are defined. The relevant range of input parameters and any interdependence between them can be informed by observed biological data and scaling. For example, depending on the system, the characteristic lengths of the system (i.e., muscle lengths, latch radius, spring length) could be constrained in the model to follow an isometric scaling. The software allows the user to enforce mathematical couplings between the different input parameters to the model, which can be used make inter-species comparisons and to investigate to what extent kinematic performance changes over the course of development for a given species.

Flexible component definitions also enables new components to be created that address specific biological questions. For example, Deban et al. performed a comparative analysis of tongue projection across salamander species which actuate their tongue projection with a LaMSA mechanism or by direct muscle actuation (Deban et al. 2020). The LaMSA projection mechanism can not only lead to higher kinematic performance, but is also robust to temperature variations (Deban et al. 2020). To explore this system with the LaMSA model presented here, new components can be created in the software that introduce a temperature-dependent motor and spring. Adding these components would yield a theoretical prediction of the relative sensitivity of the tongue projection performance to temperature for the two groups of salamanders. Comparing this prediction to the observed kinematics could be used to inform the modeling of how biological motors and springs depend on temperature. As an additional example, Acharya et al. built on the general LaMSA framework here to include nonlinear soft frictional latches to understand the ultrafast motion of human finger snaps (Acharya et al. 2021).

Finally, the model and software presented here can offer insights into how the interrelationships between input parameters and performance may influence the evolution of these biological systems via the concept of mechanical sensitivity. Mechanical sensitivity refers to the idea that variation between parts of a multi-part system are not necessarily equal in relation to their influence on the output of the system (Koehl 1996; Anderson and Patek 2015). Applied to a LaMSA system, we might hypothesize that variation in the spring would result in a larger variation in maximum power than variation in the latch mass. If so, that could mean that the latch mass has more freedom to evolve without altering performance. Such patterns have been identified in both mantis shrimp and fish (Anderson and Patek 2015; Hu et al. 2017) and have been shown to influence rates of morphological evolution (Muñoz et al. 2017, 2018; Muñoz 2019). The model presented here offers an opportunity to quantitatively map how shifts in input parameters affect multiple performance metrics simultaneously, allowing for a comprehensive analysis of mechanical sensitivity.

Conclusion
The LaMSA Template Model and software presented here balances modeling principles of simplicity and extensibility. Simplicity is provided by making explicit assumptions about how the components are connected in the model, and extensibility is achieved though flexibility of defining the individual components. With these principles, the model enables the rapid testing of ideas by simulating kinematic output across the varying model parameters. This model also opens possible new directions for future work by providing a framework for others to build upon. Case studies using the model will inform best practices for tuning the model to explore a specific biological system. Exploring biological and bioinspired LaMSA systems with this model will require input from members of comparative biomechanics community through the use of the software (available at https://poslab.github.io (Didcock et al. 2020)), requesting new features, and actively contributing to software development.

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Author contributions
AC and MI designed the research; all authors contributed to the model development; AC, KP, MAA, AW, RLD, JTC, DO, RA, and MI wrote the software; AC and
MI wrote a first draft of the manuscript; all authors revised and edited the manuscript.

Declaration of competing interest
The authors declare no competing interests.

Supplementary data
Supplementary Data available at IOB online.

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**Appendix A: Table of Parameters Used**

| Parameter | Value |
|-----------|-------|
| $F_{\text{max}}$ | 20 kN |
| $V_{\text{max}}$ | [5 m/s, $\infty$] |
| $L_0$ | 10 mm |
| $a$ | 2.08 |
| $b$ | -2.89 |
| $s$ | -0.75 |
| $r_{\text{act}}$ | [200, $\infty$] s$^{-1}$ |

**Spring**

| Parameter | Value |
|-----------|-------|
| $l_s$ | 10 mm |
| $k_0$ | 2 kN/m |
| $F_{\text{sp, max}}$ | 20 N |
| $m_s$ | 20 mg |

---

**Latch**

| Parameter | Value |
|-----------|-------|
| $R$ | 0.2 mm |
| $m_l$ | 3 g |
| $\mu_k$ | 0 |
| $\nu_0$ | 0 |
Table A1. Continued.

### Unlatching Motor

**Force Function:** (ref. [Ilton et al. 2018])

\[
F_{um}(t, x, \dot{x}) = \begin{cases} 
F_{max}(1 - \frac{\dot{x}}{v_{max}}), & \text{for } 0 \leq x \leq d \\
0, & \text{otherwise} 
\end{cases}
\]

Unlatching Motor parameters used in this work:

- \(F_{max} = 0.25 \text{ N}\)
- \(v_{max} = 1 \text{ m/s}\)
- \(d = 5 \text{ mm}\)

### Load Mass

Load Mass parameters used in this work:

- \(m_{load} = 0.1 - 100 \text{ kg}\)
- \(\text{EMA} = 1\)

---

**Appendix B: Derivation of the Model**

To simplify the model derivation, we will first reduce our general model that can include rotation down to a one-dimensional representation. For an applied force \(F\) at one end of a rotating rod with a fixed pivot:

![Diagram of a rotating rod with applied force](image)

The dynamics of the system is given by relating the applied torque about the pivot to the angular acceleration of the rod,

\[
FL_1 = I \ddot{\theta}, \tag{B.1}
\]

where \(I\) is the moment of inertia of the load mass about the fixed pivot point, which for a uniform rod of mass \(m\) is given by

\[
I = \frac{1}{12}m(L_1 + L_2)^2 + m\left(\frac{L_2 - L_1}{2}\right)^2. \tag{B.2}
\]

If the angular displacement is small (see Appendix for a derivation of the reduced model for large angular displacements), then the linear displacement of the point where force is applied \(y \approx L_1 \theta\), can be substituted into the equation of motion to give,

\[
F = \frac{I}{L_1^2} \ddot{y}, \tag{B.3}
\]

which for a uniform load mass simplifies to

\[
F = m_{load} \ddot{y}, \tag{B.4}
\]

with the effective load mass

\[
m_{load} = m \left(\frac{(1 + \frac{1}{\text{EMA}})^2}{12} + (1 - \frac{1}{\text{EMA}})^2\right)^2. \tag{B.5}
\]

In other words, the rotational system reduces down to one-dimensional dynamics of the point of where force is applied, but with an effective load mass that takes into account the EMA of the system.

With that simplification, we consider a reduced complexity one dimensional LaMSA system:
Our goal is to derive a single ordinary differential equation describing \( x(t) \) of the latch while it is in contact with the load mass.

**Setting Up the Problem**

Let us approximate the load mass and latch as point masses and draw isolation diagrams. In this model, we will consider the latch to have some shape that governs the unlatching process. Therefore, we have some function \( y_L(x) \) that determines the curve of the latch.

**Variables:**
- \( m_{\text{eff}} = m_{\text{load}} + \frac{m_{\text{spring}}}{3} \): the effective mass of the load mass and spring mass combined
- \( F_{sp} \): force exerted by the spring on the load mass
- \( \mu_k \): coefficient of friction between the load mass and the latch
- \( F_N \): normal force
- \( F_f \): force of friction between latch and load mass

\( \theta \): The angle between the normal force vector and the vertical

Among these variables, we’ll consider the following to be given:
- \( F_{sp}, \mu_k \)

With names for our variables, we can write Newton’s second law to get the following:

\[
\sum F_y = m_{\text{eff}} \ddot{y} = F_{sp} - F_N - F_f
\]

\[
m_{\text{eff}} \ddot{y} = F_{sp} - F_N \cos \theta - \mu_k F_N \cos (90 - \theta)
\]

\[
m_{\text{eff}} \ddot{y} = F_{sp} - F_N \cos \theta - \mu_k F_N \sin \theta
\]

**Variables:**
- \( m_L \): mass of the latch
- \( F_{um} \): force of the unlatching motor pulling the latch away
\( F_N \): Normal force from load mass on latch  
\( F_f \): Friction force from load mass on latch  
\( y_L(x) \): function describing the latch geometry

Known Values:  
\( m_L \): mass of the latch  
\( F_{um} \): force of the unlatching motor pulling the latch away  
\( y_L(x) \): function describing the latch geometry

We get the following equations from Newton’s 2nd Law:

\[
\sum F = m_L(\ddot{x}) = F_{um} + F_N - F_f  \\
\dot{m}_L(\ddot{x}) = F_{um} + F_N \sin \theta - \mu_k F_N \sin (90 - \theta)  \\
m_L(\ddot{x}) = F_{um} + F_N \sin \theta - \mu_k F_N \cos \theta
\]

Rewriting Unknowns in Terms of Other Variables

We will use these replacements later in the derivation.  
**Rewriting \( \ddot{y} \)** - Our goal is to get a differential equation for \( \ddot{x} \), but we will end up with \( \ddot{y} \) in our equations. So, we can use the following to rewrite \( \ddot{y} \) in terms of \( \ddot{x} \) and the latch curve:

\[
\ddot{y} = \frac{d}{dt} \left( \frac{dy}{dt} \right)  \\
\dot{y} = \frac{d}{dt} \left( \frac{dy}{dx} \cdot \frac{dx}{dt} \right)  \\
\dot{y} = \frac{d}{dt} (y_L' \cdot \dot{x})  \\
\ddot{y} = y_L'' \cdot \dot{x}^2 + y_L' \cdot \ddot{x}
\]

**Rewriting \( \tan \theta \)** - We will need to replace \( \tan \theta \) later in the derivation.

Because the latch geometry is described by the function \( y_L \), the slope of the latch is described by the derivative \( y_L' \).

\[
y_L' = \frac{\text{rise}}{\text{run}}  \\
\tan \theta = \frac{\text{rise}}{\text{run}}  \\
\tan \theta = y_L'
\]

Solving for \( \ddot{x} \)

**Solving for \( F_N \) in each equation**- Recall that we obtained the following two equations from applying Newton’s 2nd Law to both:

\[
m_{\text{eff}}(\ddot{y}) = F_{sp} - F_N \cos \theta - \mu_k F_N \sin \theta
\]
\[ m_l(\ddot{x}) = F_{um} + F_N \sin \theta - \mu_k F_N \cos \theta \]

Let us replace \( \ddot{y} \) with the expression we obtained in the previous section. Now we have

\[ m_{\text{eff}}(y''_L \cdot \ddot{x}^2 + y'_L \cdot \ddot{x}) = F_{sp} - F_N \cos \theta - \mu_k F_N \sin \theta \]  \hspace{1cm} (B.6)
\[ m_l(\ddot{x}) = F_{um} + F_N \sin \theta - \mu_k F_N \cos \theta \]  \hspace{1cm} (B.7)

The only element that is not known is \( F_N \). Let us eliminate it by solving for \( F_N \) in both equations. Solving for \( F_N \) in Eq. (B.6) gives us:

\[ F_N = \frac{F_{sp} - m_{\text{eff}}(y''_L \cdot \ddot{x}^2 + y'_L \cdot \ddot{x})}{\cos \theta + \mu_k \sin \theta} \]

Solving for \( F_N \) in Eq. (B.7) gives us:

\[ F_N = \frac{-m_l \ddot{x} - F_{um}}{\sin \theta - \mu_k \cos \theta} \]

**Expressing \( F_N \) without using \( x \)** - While it is our ultimate goal to solve for \( \ddot{x} \), a side goal that is useful for determining the end of the unlatching phase is obtaining an expression for \( F_N \) that does not include \( \ddot{x} \).

We can achieve this by taking Eqs. (B.6) and (B.7) from the previous section, isolating \( \ddot{x} \) in each, and setting them equal to each other.

Rearranging Eq. (B.6):

\[ m_{\text{eff}}(y''_L \cdot \ddot{x}^2 + y'_L \cdot \ddot{x}) = F_{sp} - F_N \cos \theta - \mu_k F_N \sin \theta \]
\[ m_{\text{eff}}y'_L \ddot{x} = F_{sp} - F_N \cos \theta - \mu_k F_N \sin \theta - m_{\text{eff}}y''_L \ddot{x}^2 \]
\[ \ddot{x} = \frac{F_{sp} - F_N \cos \theta - \mu_k F_N \sin \theta - m_{\text{eff}}y''_L \ddot{x}^2}{m_{\text{eff}}y'_L} \]

Rearranging Equation (B.7):

\[ m_l(\ddot{x}) = F_{um} + F_N \sin \theta - \mu_k F_N \cos \theta \]
\[ \ddot{x} = \frac{F_{um} + F_N \sin \theta - \mu_k F_N \cos \theta}{m_l} \]

Setting these equal to each other, isolating \( F_N \):

\[ \frac{F_{sp} - F_N \cos \theta - \mu_k F_N \sin \theta - m_{\text{eff}}y''_L \ddot{x}^2}{m_{\text{eff}}y'_L} = \frac{F_{um} + F_N \sin \theta - \mu_k F_N \cos \theta}{m_l} \]
\[ m_l(F_{sp} - F_N \cos \theta - \mu_k F_N \sin \theta - m_{\text{eff}}y''_L \ddot{x}^2) = m_{\text{eff}}y'_L(F_{um} + F_N \sin \theta - \mu_k F_N \cos \theta) \]
\[ -F_N m_l \cos \theta - F_N m_l \mu_k \sin \theta - F_N m_{\text{eff}}y''_L \sin \theta + F_N m_{\text{eff}}y'_L \mu_k \cos \theta = \]
\[ m_l m_{\text{eff}}y''_L \ddot{x}^2 - m_l F_{sp} \]
\[ + m_{\text{eff}}y'_L F_{um} \]
\[ F_N(m_{\text{eff}}y'_L \mu_k \cos \theta - m_{\text{eff}}y''_L \sin \theta - m_l \mu_k \sin \theta - m_l \cos \theta) = \]
\[ m_l m_{\text{eff}}y''_L \ddot{x}^2 - m_l F_{sp} \]
\[ + m_{\text{eff}}y'_L F_{um} \]
\[ F_N = \frac{m_l m_{\text{eff}} y''_L \ddot{x}^2 - m_l F_{sp} + m_{\text{eff}} y'_L F_{um}}{m_{\text{eff}} y'_L \mu_k \cos \theta - m_{\text{eff}} y''_L \sin \theta - m_l \mu_k \sin \theta - m_l \cos \theta} \]

It’s somewhat inconvenient to have \( \theta \) in this expression, so we can make the following substitutions based on the geometry of our problem:
\[
\sin \theta = \frac{y_L'}{\sqrt{1 + (y_L')^2}} \quad \cos \theta = \frac{1}{\sqrt{1 + (y_L')^2}}
\]

Plugging these in:
\[
F_N = \frac{m_1 m_{ct} y_L''^2 - m_1 F_{sp} + m_{ct} y_L' F_{um}}{m_{ct} y_L' L_k - m_1 m_{ct} y_L' - m_1 \sqrt{1 + (y_L')^2}}
\]

Great! Now that we have this, we’ll resume with our other goal, of solving for \( \dot{x} \).

**Solving for \( \dot{x} \)** - With two expressions for \( F_N \) in Eqs. (B.6) and (B.7), we can set them equal to each other to solve for \( \dot{x} \):
\[
\frac{F_{sp} - m_{ct} (y_L'' \cdot \dot{x}^2 + y_L' \cdot \dot{x})}{\cos \theta + \mu_k \sin \theta} = \frac{m_1 \dot{x} - F_{um}}{\sin \theta - \mu_k \cos \theta}
\]

Cross-multiply to get:
\[
(sin \theta - \mu_k \cos \theta)(F_{sp} - m_{ct} (y_L'' \cdot \dot{x}^2 + y_L' \cdot \dot{x})) = (\cos \theta + \mu_k \sin \theta)(m_1 \dot{x} - F_{um})
\]

Expanding:
\[
F_{sp} \sin \theta - m_{ct} \sin \theta (y_L'' \cdot \dot{x}^2 + y_L' \cdot \dot{x}) - F_{sp} \mu_k \cos \theta + m_{ct} \mu_k \cos \theta (y_L'' \cdot \dot{x}^2 + y_L' \cdot \dot{x}) = m_1 \dot{x} \cos \theta - F_{um} \cos \theta + m_1 \mu_k \dot{x} \sin \theta - F_{um} \mu_k \sin \theta
\]

Divide both sides by \( \cos \theta \):
\[
F_{sp} \tan \theta - m_{ct} \tan \theta (y_L'' \cdot \dot{x}^2 + y_L' \cdot \dot{x}) - F_{sp} \mu_k \tan \theta + m_{ct} \mu_k \tan \theta (y_L'' \cdot \dot{x}^2 + y_L' \cdot \dot{x}) = m_1 \dot{x} - F_{um} + m_1 \mu_k \dot{x} \tan \theta - F_{um} \mu_k \tan \theta
\]

We can replace \( \tan \theta \) with \( y_L' \):
\[
F_{sp} y_L' - m_{ct} y_L' (y_L'' \cdot \dot{x}^2 + y_L' \cdot \dot{x}) - F_{sp} \mu_k + m_{ct} \mu_k (y_L'' \cdot \dot{x}^2 + y_L' \cdot \dot{x}) = m_1 \dot{x} - F_{um} + m_1 \mu_k \dot{x} y_L' - F_{um} \mu_k y_L'
\]

If we expand the equation, move all terms that contain \( \dot{x} \) and \( \dot{x}^2 \) to one side, and factor out \( \dot{x} \) and \( \dot{x}^2 \), we get:
\[
\dot{x}(m_{ct} \mu_k y_L' - m_{ct} (y_L')^2 - m_1 - m_1 \mu_k y_L') + \dot{x}^2 (m_{ct} \mu_k y_L' - m_{ct} (y_L')^2) = -F_{um} - F_{sp} \mu_k y_L' - F_{sp} y_L' + F_{sp} \mu_k
\]

Now, let us solve for \( \dot{x} \) and regroup some terms to arrive at the final result:
\[
\dot{x} = \frac{(F_{um} + F_{sp} y_L') + \mu_k (F_{um} y_L' - F_{sp}) - \dot{x}^2 (m_{ct} \mu_k y_L' - m_{ct} (y_L')^2)}{(m_1 + m_{ct} (y_L')^2) - \mu_k (m_{ct} y_L' - m_1 y_L')}
\]

And if there is no friction such that \( \mu_k = 0 \):
\[
\dot{x} = \frac{(F_{um} + F_{sp} y_L') - \dot{x}^2 (m_{ct} y_L' y_L'')}{(m_1 + m_{ct} (y_L')^2)}
\]
Appendix C: Model for Large Angular Displacements

To model large amplitude rotational motion, we can no longer assume \( F_{spring} = m_{proj} \cdot \ddot{y} \) as we do for the case of linear (or small amplitude rotational) motion. Instead, there is a changing mechanical advantage as a function of the angle between the spring and the lever (pictured below) that complicates the dynamics of the system.

![Diagram of spring and lever system](image)

However, if we derive expressions for effective spring force and projectile mass \( F_{eff} \) and \( m_{eff} \) as functions of the displacement of the spring, we can use these expressions in our linear model to accurately describe rotational motion. In the following section, we derive these expressions.

We begin by writing Newton’s second law for rotational motion:

\[
F \cdot \sin(\alpha) \cdot L_1 = I \ddot{\theta}.
\]

Our end goal is to write this as

\[
F_{eff} = m_{eff} \cdot \ddot{y},
\]

where \( y \) is the displacement of the spring.

First, we will write \( \ddot{\theta} \) in terms of \( \ddot{y} \). There is a complex exact relationship between \( y \) and \( \theta \) that an interested reader can calculate using the law of cosines a few times, but it is very well approximated by \( y = L_1 \cdot \sin(\theta) \). Using this, we find

\[
\begin{align*}
y &= L_1 \cdot \sin(\theta) \\
\dot{y} &= L_1 \cdot \cos(\theta) \cdot \dot{\theta} \\
\ddot{y} &= -L_1 \cdot \sin(\theta) \cdot \dot{\theta}^2 + L_1 \cdot \cos(\theta) \cdot \ddot{\theta}.
\end{align*}
\]

Rearranging these equations, we find

\[
\ddot{\theta} = \frac{1}{L_1 \cos(\theta)} \ddot{y} + \frac{\sin(\theta)}{L_1^2 \cos^3(\theta)} \dot{y}^2
\]

(C.2)

Now, if we substitute equation C.2 into equation C.1, we have

\[
F \cdot \sin(\alpha) = \frac{I}{L_1^2} \left( \frac{1}{\cos(\theta)} \ddot{y} + \frac{\sin(\theta)}{L_1 \cos^3(\theta)} \dot{y}^2 \right)
\]

(C.3)

Next, we find \( I \). Using the parallel axis theorem, we find the moment of inertia of the lever (a rod with uniformly distributed mass \( m \)) and projectile (a point mass \( M \)) about the axis of rotation will be

\[
I = M(L_2)^2 + \frac{1}{12} m (L_1 + L_2)^2 + m \left( \frac{L_2 - L_1}{2} \right)^2.
\]
Dividing by $L_1^2$ and substituting the EMA $EMA = \frac{L_1}{L_2}$, we have found an intermediate mass quantity $m_{int}$ such that

$$m_{int} = \frac{I}{L_1^2} = \left( \frac{M}{EMA^2} + \frac{m}{12} \left( \left( 1 + \frac{1}{EMA} \right)^2 + 3 \left( \frac{1}{EMA} - 1 \right)^2 \right) \right)$$

and we can rewrite equation C.3 as

$$F \cdot \sin(\alpha) = m_{int} \cdot \left( \frac{1}{\cos(\theta)} \ddot{y} + \frac{\sin(\theta)}{L_1 \cos^3(\theta)} \dot{y}^2 \right)$$

or

$$F \sin(\alpha) - m_{int} \frac{\sin(\theta)}{L_1 \cos^3(\theta)} \dot{y}^2 = m_{int} \cos(\theta) \cdot \ddot{y}$$

(C.4)

Equation C.4 is in the desired form, so we can now extract

$$F_{eff} = F \sin(\alpha) - m_{int} \frac{\sin(\theta)}{L_1 \cos^3(\theta)} \dot{y}^2$$

and

$$m_{eff} = \frac{m_{int}}{\cos(\theta)}.$$