Quadrupedal Bounding with an Actuated Spinal Joint

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Abstract—Most legged vertebrates use flexible spines and supporting muscles to provide auxiliary power and dexterity for dynamic behaviors, resulting in higher speeds and additional maneuverability during locomotion. However, most existing legged robots capable of dynamic locomotion incorporate only a single rigid trunk with actuation limited to legs and associated joints. In this paper, we investigate how quadrupedal bounding can be achieved in the presence of an actuated spinal joint and characterize associated performance improvements compared to bounding with a rigid robot body. In the context of both a new controller structure for bounding with a body joint and existing bounding controllers for the rigid trunk, we use optimization methods to identify the highest performance gait parameters and establish that the spinal joint allows increased forward speeds and hopping heights.

I. INTRODUCTION

Legged morphologies offer the necessary mobility, efficiency and agility for autonomous operation in complex outdoor environments. Evidence to this end is provided by the exclusive adoption of legs on land by the most efficient locomotors [1, 2]. Bounding is commonly used particularly by quadrupedal animals for clearing obstacles and running at moderate speeds [3]. Due to its saggitally symmetric nature and relatively simple structure, this behavior also received considerable attention in the robotics community, starting from Raibert's work on dynamic legged robots [4, 5] and continuing with later instantiations on different platforms including KOLT [6], Scout I and II [7, 8], PAW [8] and BigDog [9]. As such, bounding has been one of the most commonly implemented dynamic behaviors for legged robots.

Most bounding robots built to date are very similar in their morphology and feature a single rigid body with four passively compliant legs, occasionally equipped with tunable compliance, and almost always with individually actuated hips (see Fig. 1). The bounding behavior supported by this model, however, is much more constrained than its natural counterpart, which explicitly relies on spinal flexion both for increased speed [10] and thrust [11]. Nevertheless, there has only been a few attempts with limited success to implement similar mechanisms on robotic platforms [12], with substantial research effort devoted instead to the design of different leg structures and associated actuation mechanisms [13–15].

In this paper, we propose a new model and an associated bounding controller with a jointed spine. We show that our controller not only achieves stable bounding, but also higher locomotion speeds due to larger possible stride lengths resulting from spinal flexion. In the design of our bounding controller, we adopt an open-loop state machine structure similar to most existing controllers for bounding [7, 16] and other similar dynamic locomotory behaviors [17-20]. In contrast to these manually tuned behaviors, however, we use optimization methods to identify the "best" possible bounding gait, both for a jointed-spine and a rigid quadruped to support meaningful comparisons. Such optimization methods have previously been used to identify advantageous sets of morphological parameters for bounding behaviors [21, 22] as well as to provide alternatives to manual tuning for achieving gait efficiency [23]. Similarly, there has been substantial recent work on gait adaptation under both model uncertainties and environmental conditions [24, 25]. In contrast to more formal, model-based inquiries on the analysis of the bounding behavior to provide insight to the behavior itself [26, 27], such adaptation mechanisms can be used to obtain very effective locomotion controllers that would otherwise be impossible to design through manual tuning alone.

The primary contributions of this paper include the design of a novel bounding controller for a quadruped with an actuated spinal joint, and the demonstration of how the spinal joint increases the bounding speed that can be reliably achieved compared to a quadruped with a rigid body. In Section II, we describe the planar dynamic models for both the jointed body and rigid body quadruped morphologies. We then describe bounding controllers for both in Section III, followed by our comparative simulation results in Section IV.

II. PLANAR BOUNDING MODELS



Fig. 1. Planar quadruped model with a stiff body.

Due to the limited attention given to spinal actuation in the robotics literature, there is no single, widely accepted mathematical model to serve as a basis for studying locomotion within such a morphology. In contrast, the rigid quadruped model that we will use for our comparative study, shown in Fig. 1, has previously been used in the literature. This rigid model consists of a single rigid body with mass m_b and inertia I_b , located at (x_C, y_C) at an angle of θ_b with the

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horizontal. Two legs with compliance k and damping b are attached to the body at distances d_i away from its center of mass, individually actuated with a torque τ_i relative to the body. Each leg can either be in stance or flight, with the leg angle relative to the body vertical denoted with φ_i and its length with l_i .



Fig. 2. Planar quadruped model with an actuated spinal joint.

In contrast, the new bounding model introduced in this paper is illustrated in Fig. 2 and incorporates an actuated spinal joint connecting two rigid bodies with masses m_{b_i} and inertias I_{b_i} at a horizontal distance d_{s_i} from their respective centers of mass. This spinal joint is assumed to be actuated with a controllable torque τ_s . We also denote the angle between the two segments of the robot body with β_s .

In the rest of the paper, we will use the subscripts f and b to replace i in the definitions above and refer to front and back legs and individual body segments, respectively. We have implemented both models using the *Working Model* 2D environment, which supports numerical simulation of two dimensional rigid-body systems. Since this simulation environment implicitly computes and solves the equations of motion for these models, we do not present detailed derivations of the associated dynamics in this paper. Further details and specific parameters for the simulation environment we use for our results are given in Section IV-A.

III. CONTROL OF QUADRUPEDAL BOUNDING

In this section, we describe bounding controllers for both the stiff-backed and the new jointed-spine quadruped models, consisting of high-level state machines that modulate commands to the actuated degrees of freedom, which are then tracked by local PID controllers. We assume leg touchdown and liftoff events can be detected to trigger state transitions. The only other sensors in both systems are assumed to be encoders on hip joints and the body joint to measure their relative angles and support local PID control laws.

A. Bounding with a Stiff Spine

 TABLE I

 HIGH-LEVEL STATE MACHINE FOR STIFF-BACKED BOUNDING.

State	Target Angles	Trigger Event
Double Flight	$(\varphi_{b_{td}}, \varphi_{f_{td}})$	Back leg lift-off
Front Leg Stance	$(\varphi_{b_{td}}, \varphi_{f_{lo}})$	Front leg touchdown
Double Stance	$(\varphi_{b_{lo}}, \varphi_{f_{lo}})$	Back leg touchdown
Back Leg Stance	$(\varphi_{b_{lo}}, \varphi_{f_{td}})$	Front leg lift-off

The bounding gait controller we use for the stiff backed robot model has the same structure with the controllers used in previous research focused on stiff quadrupedal bounding [7, 8, 21], consisting of a reactive state machine that responds to leg contact events. The desired bounding behavior proceeds through four different states as shown in Table I. The last column in this table indicates the event that leads into the state, whereas the middle column indicates target angles for both the front and back legs that are activated when the corresponding state is entered. Combined with the gains for the local PID controllers, the resulting behavior can hence be represented by the vector

$$\mathbf{p}_{sb} := \left[\varphi_{b_{td}}, \varphi_{b_{lo}}, \dot{\varphi}_{b}, \varphi_{f_{td}}, \varphi_{f_{lo}}, \dot{\varphi}_{f}, K_{p}, K_{i}, K_{d}\right]^{T} \quad (1)$$

where K_p , K_i and K_d denote local PID gains to track linear leg and body joint reference trajectories.

B. Bounding with a Jointed Spine

Bounding with an actuated spine requires the control of the spinal joint. Consequently, we augment the high-level state machine for stiff-backed bounding described in Section III-A with additional semantics, resulting in the state machine illustrated in Fig. 3. Our extensions are loosely inspired from biomechanics literature on quadrupedal bounding observed in legged vertebrates [1, 3, 28], cheetahs in particular.



Fig. 3. High-level state machine for bounding with an actuated spinal joint. Each state illustrates how the robot body and legs are expected to be configured. Transitions indicate associated leg contact events.

The key observation in this context is that the spinal joint has two dominant configurations: convex and concave. The former increases stride length by increasing touchdown angles for both legs while the latter prepares the kinematics for additional thrust from the spinal actuator. The following paragraphs present descriptions of each of these states.

1) Double Flight: This state starts with the back leg lifting off during the back leg stance phase. Both legs are then commanded to go to pre-specified touchdown angles $\varphi_{b_{td}}$ and $\varphi_{f_{td}}$. The robot body assumes a convex shape with an angle β_{cx} , increasing the stride length for the following step.

2) Front Leg Stance: This state starts with the front leg touching the ground during the double flight phase. The spine actuator then assumes a concave shape with an angle β_{cv} ,

helping increase the stride length of the back leg. In this state, the front leg maintains backwards motion with a sweep limit, while the back leg maintains its touchdown angle in flight.

3) Double Leg Stance: This state starts with the back leg touching the ground during the front leg stance phase. Following this transition, the front leg spring quickly starts compressing and the back leg starts to shorten in length. Both legs are commanded to go to their liftoff angles $\varphi_{b_{lo}}$ and $\varphi_{f_{lo}}$ at predefined speeds $\dot{\varphi}_b$ and $\dot{\varphi}_f$. The spine actuator is used to maintain a concave body shape to ensure and prepare for supplying additional power during front leg retraction.

4) Back Leg Stance: This state starts with the front leg lifting off the ground during the double leg stance phase. The spine actuator then forms a convex shape, while the back leg maintains its liftoff angle and experiences maximum compression. On the other hand, the front leg actuator is commanded to go towards its touchdown angle position.

Given these four phases, the bounding controller with the flexible body joint is represented with the parameter vector

$$\mathbf{p}_{fb} := [\varphi_{b_{td}}, \varphi_{b_{lo}}, \dot{\varphi}_{b}, \varphi_{f_{td}}, \varphi_{f_{lo}}, \dot{\varphi}_{f}, \beta_{cx}, \beta_{cv}, \beta_{s}, K_{p}, K_{i}, K_{d}, K_{ps}, K_{is}, K_{ds}]^{T}$$
(2)

where K_p , K_i and K_d denote local PID controller gains whereas angle parameters and their velocities parametrize linear trajectories for the body and leg angles starting from their current position towards their target angles.

C. Local PID Controllers

In all high level controller states for bounding with both the stiff and jointed back morphologies, local PID controllers are used for each actuator to determine associated torque commands. Torque commands for legs in both models and the body joint are computed as

$$\tau_j = K_p e_j(t) + K_j \int_0^t e_j(t) dt + K_d \frac{de_j(t)}{dt}, \qquad (3)$$

with j indexes either the leg number or the body joint. Leg and body tracking errors are respectively defined as $e_i(t) := \varphi_i^*(t) - \varphi_i(t)$, and $e_b(t) := \beta_s^*(t) - \beta_s(t)$.

The desired angles for both the legs and the spine are determined by the current phase of the high-level controller and the controller parameters in \mathbf{p}_{sb} and \mathbf{p}_{fb} for the stiff-backed and flexible-spine models, respectively. For both models, the computation of the desired angles for the legs depend on whether they are individually in stance or flight. During stance, we have

$$\varphi_i^*(t) = \begin{cases} \varphi_i(t_{td}) + \dot{\varphi}_i(t - t_{td}) & \text{if } t - t_{td} < \frac{\varphi_{i_{lo}} - \varphi_{i_{td}}}{\dot{\varphi}_i} \\ \varphi_{i_{lo}} & \text{otherwise }. \end{cases}$$

Similarly, during flight, we have

$$\varphi_i^*(t) = \begin{cases} \varphi_i(t_{lo}) - \dot{\varphi_i}(t - t_{lo}) & \text{if } t - t_{lo} < \frac{\varphi_{i_{td}} - \varphi_{i_{lo}}}{\dot{\varphi_i}} \\ \varphi_{i_{td}} & \text{otherwise }. \end{cases}$$

The spine actuator is controlled in a similar fashion with its concave and convex poses using the target angles determined by the parameter vector in (2).

IV. SIMULATION RESULTS

In this section, we present a comparison of the "best" gait performances for the stiff and jointed back bounding controllers, obtained through an optimization framework to automatically tune gait parameters, resulting in a fair characterization of performance in terms of speed, stability and efficiency for both models.

A. Simulation Environment

We have implemented both models described in Section II in the Working Model 2D environment which supports numerical simulation of articulated rigid-body systems with collisions and models of surface interaction. In order to implement the stiff-backed model, we rigidly locked the body joint of the flexible model, keeping all other system parameters identical. All simulations were run with a small, constant time step of $10^{-3}s$ in order to ensure numerical accuracy of the results. Table II details kinematic and dynamic parameters we have chosen for our models, mirroring the morphology of a cheetah as much as possible except perhaps leg compliance and damping constants [28]. Note that we use the same total body mass for both models in order to ensure a fair comparison of performance even though the spinal motor is likely to increase the total mass of the jointed model. In practice, this choice may be justified by choosing a smaller payload capacity for the jointed robot.

TABLE II

SYSTEM PARAMETERS F	OR BOTH	BOUNDING	MODELS
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Param.	Value	Param.	Value
m_{bi}	$10 \ kg$	m_b	$20 \ kg$
I_{bi}	$1.3 \ kg \cdot m^2$	Ib	$3.85 \ kg \cdot m^2$
d_i	0.365 m	d_{si}	0.25 m
k	3500 N/m	b	55 Nm/s
l	$0.8 \ m$	$ au_{max}$	200 Nm
μ_s	0.9	μ_k	0.8
$i \in \{f, b\}, f$: front, b: back			
μ_s, μ_k : Static and kinetic friction			

Given a set of controller parameters, simulations were started from an initial condition with a forward speed of 1 m/s and a height of 0.75 m and executed for 32s. System trajectories were then transferred to Matlab for stability analysis and visualization. A simulation run was considered stable if the system trajectories converged to a limit-cycle in the steady-state. This was verified using the difference in system states from one touchdown of the front leg to the next with a tolerance of 10^{-1} on the norm of the error. For each successful run, we compute a performance measure that is inspired from the commonly used *specific resistance* [29], modified to slightly favor speed over efficiency with $\epsilon := P/mgv^3$, where P denotes the average of either instantaneous power or its absolute value and v denotes the average velocity across the last five steady-state strides.

B. Nelder-Mead Optimization

Nelder-Mead is an efficient, simplex-based numerical optimization method [30] that approximates gradient descent but requires much fewer evaluations of the objective function. Consequently, it is suitable for robotic applications where the evaluation of the objective function requires running experiments with a physical robot [23], or simulations that may take a long time to complete. Since our automated tuning of the bounding controllers rely on running a rather slow simulation for 32s in simulation time, we will use the Nelder-Mead framework to find controller parameters that minimize ϵ .



Fig. 4. Progression of the Nelder-Mead optimization for stiff backed (left) and actuated spine (right) models. Red squares plot the stopping criteria function C, whereas blue stars represent the best vertex cost values for each simplex. Each "turn" corresponds to five Nelder-Mead iterations.

For the stiff-backed bounding model, we use the parameter set defined in (1) that corresponds to a D = 9 dimensional space of independent variables for optimization. In contrast, for the model with a jointed spine, we use the parameter set defined in (2) that corresponds to a D = 15 dimensional space. In both cases, we start the optimization from a manually tuned initial condition that realizes the desired high-level state transition sequence and a gait with reasonable stability. We then run the Nelder-Mead algorithm until convergence, defined through a stopping function estimating the size of the current simplex as $C = \sqrt{\sum_i (\epsilon_i - \bar{\epsilon})^2 / (D+1)}$, where ϵ_i denotes the objective function value for different vertices. When this function falls below a certain threshold (10^{-3}) in our case), the optimization is terminated and is assumed to be converged. Fig. 4 illustrates the progressions of both the objective function and the stopping criteria function for both models. Parameter sets obtained at the end of these optimizations are given in Tables III and IV.

TABLE III

OPTIMAL GAIT PARAMETER FOR STIFF BACKED BOUNDING

Parameter	Value
K_p, K_i, K_d	(338.2, 0.08, 6.7)
$\varphi_{td_f}, \varphi_{lo_f}, \dot{\varphi_f}$	$(0.4 \ rad, -0.03 \ rad, 4.25 \ rad/s)$
$\varphi_{td_b}, \varphi_{lo_b}, \dot{\varphi_b}$	$(0.27 \ rad, -0.11 \ rad, 4.5 \ rad/s)$

TABLE IV Optimal gait parameters for Actuated Spine Bounding

Parameter	Value
K_p, K_i, K_d	(524.7, 0.11, 6.9)
K_{ps}, K_{is}, K_{ds}	(1737, 0.03, 330.3)
$\varphi_{td_f}, \varphi_{lo_f}, \dot{\varphi_f}$	$(0.3 \ rad, \ 0.17 \ rad, \ 3.43 \ rad/s)$
$\varphi_{td_b}, \varphi_{lo_b}, \dot{\varphi_b}$	$(0.3 \ rad, -0.12 \ rad, 4.93 \ rad/s)$
$\beta_{cx}, \beta_{cv}, \dot{\beta_s}$	(0.1 rad, -0.22 rad, 23.05 rad/s)

C. Simulation Results for Bounding with a Stiff Spine

Optimal control parameters for the stiff-backed bounding model result in the behavior illustrated in the snapshots of



Fig. 5. Snapshots of optimal bounding with the stiff-backed model.



Fig. 6. Body height (top), horizontal velocity (middle) and foot clearance (bottom) trajectories for bounding with the stiff-backed model. Green dashed lines in the top two plots indicate the average horizontal speed and heights. Shaded regions in the bottom indicate different controller phases.

Fig. 5. Body height and forward speed trajectories associated with this gait are shown in Fig. 6. With these parameters, stiff-backed bounding reached an average running speed of 1.75m/s and a maximum body height of approximately 0.76m. The robot has a small but distinct flight phase with a maximum foot clearance of up to 0.035m as shown in Fig. 6.



Fig. 7. Power consumption (top) and torque output (bottom) of motors. Blue dotted line represents front hip motor and black line back hip motor.

TABLE V Specific resistance values for bounding behaviors.

Model	ϵ with Avg. Power	ϵ with Avg. Abs. Power
stiff	0.127	0.309
jointed total	0.227	0.755
jointed legs	0.024	0.371

In order to achieve the bounding gait, both hip motors individually consume less than 400W each instantaneously with averages being much lower. Fig. 7 also shows that both motors output less than 90Nm torque on the legs. Based on these figures, the specific resistance of the robot during steady-state bounding is given in Table V, where values

obtained from the average of both instantaneous power and its absolute value are given. The optimization was done with the average absolute power, but the direct average of the power is included to give a sense of how passive compliance would benefit performance.

D. Simulation Results for Bounding with an Actuated Spine



Fig. 8. Snapshots of optimal bounding with the actuated spine model.

Similar to our simulations with the stiff-backed bounding model, Fig. 8 illustrates snapshots from steady-state bounding with the actuated spine model. The video extension accompanying this paper also shows both real-time and slow motion versions of optimal bounding with this model using the parameters of Table IV. These experiments demonstrate the validity of our controller since both the bending directions of the body and the progression of leg contact states are consistent with the proposed high-level state machine.



Fig. 9. Body height (top), horizontal velocity (middle) and foot clearance (bottom) trajectories for bounding with the actuated spine model. Green dashed lines in the top two plots indicate the average horizontal speed and heights. Shaded regions in the bottom indicate different controller phases.



Fig. 10. Power consumption (top) and torque outputs (bottom) of motors. Red dashed line represents spine motor, blue dotted line front hip motor and black line back hip motor.

As shown in Fig. 9, the actuated spine model achieves a stable bounding gait with an average horizontal speed of 2m/s. As shown in Fig. 9, the hopping height of the robot fluctuates between 0.7m and 0.8m, with the corresponding leg clearances illustrated in Fig. 9. Not surprisingly, the body joint experiences higher loads than the leg actuators, with a maximum instantaneous power of 1200W but a much lower average. Fig. 10 also shows that spine motor torque saturates during the front leg stance. Bounding with the actuated spine model yields specific resistance values shown in Table V.

E. Discussion

The most important outcome of our results is the observation that the actuated spinal joint allows bounding at higher speeds. Consistent with studies of how increases in speed are obtained by quadrupedal animals in the biomechanics literature [10], this increase in speed is primarily a consequence of the increase in stride length as illustrated by Fig. 11. The average stride lengths for the stiff and flexible spine models are 0.58m and 0.86m, respectively. Interestingly, if Figures 6 and 9 are studied carefully, one can notice that the stride frequency for the optimal bounding gait with the flexible spine model is in fact *smaller*, corresponding to a slowing down of the stepping frequency. This means that the increase in stride length through the use of the spinal joint is not only sufficient to compensate for the slowing down in the stride frequency, but also allows an overall increase in the average speed from 1.75m/s to 2.02m/s.



Fig. 11. Stride lengths for the back (left plot) and front (right plot) legs for the last six bounding steps. Red stars illustrate stride lengths for the actuated spine model whereas the blue squares correspond to the stiff backed model.

Some additional insight into the bounding behavior may be gained by inspection of the optimal parameters obtained for the actuated spine model. In particular, the back leg is commanded at a higher velocity (4.93rad/s) following touchdown even though its range of travel is diminished (0.42rad) compared to the front leg (0.47rad). This translates into the back leg exerting larger torques during stance, also confirmed by the torque profiles of Fig. 10. This would motivate stronger actuators for the back legs of a possible physical realization of this gait, which is also intuitively consistent with what we observe in most quadrupedal runners.

When the specific resistance values are compared, a number of interesting features emerge. First of all, there is a substantial difference between specific resistance values computed with the averages of instantaneous power and its absolute value. This suggests that both the leg motors and particularly the body joint do significant negative work and would substantially benefit from some form of passive compliance. This is both intuitive and consistent with existing conclusions on how springs can improve efficiency of locomotion. Naturally, when such passively compliant elements are incorporated into the system, control strategies need to be adapted accordingly. Finally, we can also see that when only the legs are considered for both models using the absolute value of the power for averaging, both bounding models have similar specific resistance values with 0.309 for the stiff and 0.371 for the jointed back models. This shows that the incorporation of the jointed back does not substantially change the power budget of leg actuators.

V. CONCLUSION AND FUTURE WORK

In this paper we have introduced a new quadruped robot model with an actuated spinal joint. Inspired from the morphology of fast running land mammals, we have designed a new bounding gait controller that works with our new model. We compared our model with a commonly used rigid bounding model through corresponding optimized bounding gaits and showed the effects of an actuated spine mechanism on the running speed and hopping height.

Our simulation experiments show that the actuated spine mechanism helps increase stride length, resulting in increased horizontal speed. The actuated spine is also capable of achieving higher hopping heights of the robot body, which helps the bounding behavior overcome bigger obstacles, consistent with its use by quadrupedal animals. The efficiency of the new bounding behavior was found to be not unreasonably different from stiff-backed bounding even with the additional actuator on the body joint.

In the near future, we will investigate using passively compliant elements both for the body joint and the legs to eliminate inefficiencies arising from negative work. This would also address practical problems such as controller bandwidth and required motor size towards a realistic implementation of the bounding behavior with jointed spine. Another important direction is to investigate how the front and back legs should be differentiated to motivate more efficient distribution of the power budget on a physical robot platform. In any case, we have been able to demonstrate in this paper that the incorporation of a flexible spine may be advantageous in improving quadrupedal gait performance.

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