Complex 3D terrains, such as leaf litter, fallen branches on a forest floor, and landslide debris (figure 1), can pose a major challenge for small animals and robots alike because obstacles are often comparable to, or even larger than, the animal or robot itself [1] and can induce large perturbations [2–8]. Despite this, small animals like insects, reptiles, and small birds agilely traverse complex 3D terrains [2, 9] as diverse as inclined and vertical surfaces [10–16], thin ledges [17], large gaps [4, 7, 18–21] and bumps [3, 5, 22–26], uneven surfaces [6, 27–30], cluttered terrain [8], and even confined spaces [31], with a locomotor performance far exceeding that of even the best robots today [32–36].

To move through varying complex environments, animals not only use multi-modal sensory feedback and feedforward neural commands to control and adjust their body and limbs, but their body and limbs also often have well-tuned morphology to accommodate perturbations in the environment via mechanical feedback [2, 37]. During slow locomotion where sensing is sufficient, animals like stick insects [4, 38], cockroaches [3, 23, 39, 40], lizards [5, 24], and snakes [7, 19] tend to use deliberate, seemingly well-planned, and often precisely-controlled, body and limb movements to negotiate and traverse complex 3D terrains, presumably using antennae or vision in the process. For example, stick insects use their antennae to sense the terrain and use quasi-static, ‘follow-the-leader’ stepping to slowly walk on branches [41], climb over steps [11, 42] and bridge over large gaps. Slow running [39, 43] or walking [3, 23, 40, 44] cockroaches use their antennae to sense obstacles in front and alter their kinematics to either climb over steps [3, 22], tunnel under steps [23, 45], approach and climb up pillars [44, 46], or follow walls [39, 43, 47], depending on the location of the obstacle. Lizards frequently jump onto and over large bumps [48], and snakes either quasi-statically

Dynamic traversal of large gaps by insects and legged robots reveals a template

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Dynamic traversal of large gaps by insects and legged robots reveals a template

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Abstract

It is well known that animals can use neural and sensory feedback via vision, tactile sensing, and echolocation to negotiate obstacles. Similarly, most robots use deliberate or reactive planning to avoid obstacles, which relies on prior knowledge or high-fidelity sensing of the environment. However, during dynamic locomotion in complex, novel, 3D terrains, such as a forest floor and building rubble, sensing and planning suffer bandwidth limitation and large noise and are sometimes even impossible. Here, we study rapid locomotion over a large gap—a simple, ubiquitous obstacle—to begin to discover the general principles of the dynamic traversal of large 3D obstacles. We challenged the discoid cockroach and an open-loop six-legged robot to traverse a large gap of varying length. Both the animal and the robot could dynamically traverse a gap as large as one body length by bridging the gap with its head, but traversal probability decreased with gap length. Based on these observations, we developed a template that accurately captured body dynamics and quantitatively predicted traversal performance. Our template revealed that a high approach speed, initial body pitch, and initial body pitch angular velocity facilitated dynamic traversal, and successfully predicted a new strategy for using body pitch control that increased the robot’s maximal traversal gap length by 50%. Our study established the first template of dynamic locomotion beyond planar surfaces, and is an important step in expanding terradynamics into complex 3D terrains.

1. Introduction

Complex 3D terrains, such as leaf litter, fallen branches on a forest floor, and landslide debris (figure 1), can pose a major challenge for small animals and robots alike because obstacles are often comparable to, or even larger than, the animal or robot itself [1] and can induce large perturbations [2–8]. Despite this, small animals like insects, reptiles, and small birds agilely traverse complex 3D terrains [2, 9] as diverse as inclined and vertical surfaces [10–16], thin ledges [17], large gaps [4, 7, 18–21] and bumps [3, 5, 22–26], uneven surfaces [6, 27–30], cluttered terrain [8], and even confined spaces [31], with a locomotor performance far exceeding that of even the best robots today [32–36].

To move through varying complex environments, animals not only use multi-modal sensory feedback and feedforward neural commands to control and adjust their body and limbs, but their body and limbs
cantilever their body to reach across a smaller gap [19] or dynamically lunge to traverse a larger gap [19], presumably all using vision in the process.

By contrast, during rapid locomotion such as prey chasing and predator escape, particularly in terrains with frequent large disturbances [25], sensory noise [37] and sensory and neuromuscular delays [49] limit an animal’s ability to plan its course of locomotion and control its body and limb movement during each step or gait cycle. To accommodate this, animals primarily move their body and limbs in a more feed-forward manner [37] to simplify control, and use neural and sensory feedback to adjust their body and limbs in response to large environmental perturbations. Because of this, how its body and limbs passively respond to the environment has a crucial impact on the dynamics and locomotor performance of a small animal moving dynamically through complex 3D terrains [6, 8, 28, 50]. For example, cockroaches have streamlined body shapes that facilitate the traversal of cluttered terrain, highly compressible yet robust exoskeletons that help them move through small crevices [31], sprawled leg posture that self-stabilizes lateral perturbations [51], viscoelastic leg muscles and tendons that dampen external perturbations [52, 53], and distributed leg spines that increase the probability of a firm foothold on low contact area surfaces [28].

Traditionally, the field of mobile robotics has mainly dealt with locomotion in complex environments by solving the problem of obstacle avoidance [54–56], which requires either prior knowledge or high-fidelity sensing to plan clear locomotion paths [56–58]. More recently, thanks to cross-discipline collaborations between biologists, applied mathematicians, and engineers, the neuromechanical principles from biological studies have enabled many under-actuated bio-inspired robots [32, 35, 59–61]. By combining high-level sensing and planning with mechanical feedback via mechanically tuned designs and control algorithms [62–64], these robots have achieved unprecedented locomotion performance on simple ground [32, 59] and are beginning to traverse complex 3D terrains [8, 31, 36, 61].

For both legged animals and robots, the primary focus of mechanical feedback studies has been how leg morphology and mechanics interact with simple ground to assist locomotion [37, 63, 65]. In complex 3D terrains where obstacles could be larger than the animal or robot, the body of a small legged animal or robot, which is more robust to collision [31, 50, 66], may also physically contact and interact with the terrain to help locomotion. However, much less attention has been paid to the role of body–terrain interaction in legged locomotion until very recently [31, 50].

Here, we take the next step in understanding how body–terrain interaction affects the dynamic traversal of large 3D obstacles. We chose to focus on two simple large obstacles: (1) a gap obstacle as large as the animal/robot’s body length, reported in this paper; and (2) a bump obstacle as high as four times the animal/robot’s hip height, reported in a companion paper [67]. Such simple, well-defined, parameterizable laboratory models of natural terrains are useful towards understanding complex interactions between moving animals and their natural environment [3, 8, 10, 14, 68]. We chose to study the discoid cockroach, which lives on the floor of tropical rainforests and naturally traverses a wide variety of 3D obstacles, such as dense vegetation and litter [69], because mechanical feedback plays a major role in its locomotion at high speeds [6, 28, 37]. A better ability to traverse large gap and bump obstacles is also important for small legged robots in complex 3D terrains like building rubble and landslides (figure 1(d)) during applications like search and
rescue [70]. We chose to study a cockroach-inspired RHex-like robot [32] because it not only has a similar body plan and running dynamics [71] as the animal and allows a direct comparison with animal observations, but also provides a physical model to precisely control and systematically vary locomotor parameters [70,72].

In this paper, we focus on understanding how a legged animal and robot’s body dynamics and body–terrain interaction affects the dynamic traversal of a large gap obstacle. We challenged the rapidly running animal and the open-loop robot with a large gap of variable size, and tested how the ability to traverse depended on gap length, running speeds towards the gap, and body orientation. Comparison of the animal with the open-loop robot provided insights into the role of active sensory feedback. Inspired by the similarities of animal and robot observations, we developed a reduced-order dynamic model, or template [10, 37, 73], to capture low-order dynamic traversal of a large gap obstacle. Finally, we added an active tail [21, 74] to the robot and tested a control strategy revealed by our template to enhance large gap traversal. In our companion paper [67], we report our experiments and the modeling of the dynamic traversal of a large bump obstacle, and discuss common and potentially general principles and distinct differences between the dynamic traversal of large gaps and bump obstacles.

2. Methods

2.1. Animals
For the animal experiments, we used seven male Blaberus discoidalis cockroaches (Pinellas County Reptiles, St Petersburg, FL, USA), as females were often gravid and under different load bearing conditions. Prior to the experiments, we kept the animals in individual plastic containers at room temperature (22 °C) on a 12 h:12 h light:dark cycle and added food (fish and rabbit pellets) ad libitum. The animals weighed 2.7 ± 0.2 g and measured 4.6 ± 0.2 cm in length, 2.3 ± 0.2 cm in width, and 0.7 ± 0.1 cm in thickness.

2.2. Legged robot
For the robot experiments, we constructed a cockroach-inspired, six-legged robot by modifying the RHex-class design [32] (figure 2(b)). The slightly-flexible robot chassis was cut from a 3.1 mm thick acrylic sheet using a VLS 6.60 laser cutter (Universal Laser Systems Inc., Scottsdale, AZ). We attached each motor (460 RPM micro-gear DC motor, 50:1 gear ratio, ServoCity, Winfield, KS) that drove the legs to 2 cm × 2 cm pieces of 0.15 cm thick polystyrene that were subsequently attached to the chassis. To increase the maximal leg frequency of the motor used to drive the legs, we chose s-shape legs. The s-shape legs were 3D printed with PLA plastic (Ultimaker 2 Extended+, Ultimaker North America, Cambridge, MA). We wrapped each of the legs with friction tape (Duck Brand, Avon, OH) to increase ground traction. To generate stable spring-mass-like [75] running, we tuned the stiffness of the chassis and legs and leg friction. To approximate the anterior shape of the cockroach body for direct comparison of the terrain interaction between the robot and the animal, we thermo-formed the bottom half of a quarter ellipsoid polystyrene shell and attached it to the front of the robot. The robot measured 25 cm long, 16 cm wide, 8 cm tall, and weighed 194 g. We calculated the robot moment of inertia by approximating the body as a simple rigid rod measuring 25 cm long and weighing 194 g, which rotates about a fixed end ($I = 1/3mb^2 = 0.004$ kg m$^2$, where $m$ is body mass and $b$ is body length).

To test the effect of passive mechanics, we did not implement any sensors and drove the robot with an open-loop leg control. We varied the robot’s running speed from 50 cm s$^{-1}$–200 cm s$^{-1}$ by changing the voltage supplied to the DC motors to adjust the leg frequency. At the maximal voltage of 25 V, the robot ran stably at 8.0 ± 0.5 body lengths s$^{-1}$ (200 ± 12 cm s$^{-1}$).

2.3. Gap obstacle track
We constructed a 90 cm long, 30 cm wide track with a gap that spanned the entire width of the track (figure 2(c)) using t-slotted extruded aluminum and acrylic (McMaster-Carr, Elmhurst, IL, USA). For both the animal and robot experiments, we varied the gap length $L$ from 0.2 body length to 1 body length ($b ≈ 5$ cm for the animal, $b = 25$ cm for the robot) by sliding the far side of the track. To measure the maximal traversable gap length, we made an infinite length gap (or cliff) by removing the far side of the gap. We covered the entire test surface with white paper cardstock (Pacon 4-ply railroad poster board, Appleton, WI, USA) for the animal experiments. For the robot experiments, we covered the near side of the test surface with 50 grit sandpaper to increase the leg traction and prevent slipping when the legs started moving. To prevent the robot shell from being scratched during impact, we covered the opposite side of the gap with polystyrene.

2.4. Experiment protocol
We filmed the animal and robot running over the gap at 500 frames s$^{-1}$ using three synchronized high-speed cameras, two from a dorsal view and one from a side view (figure 2(c)), with a shutter time of 500 μs. The dorsal cameras were placed directly above the near edge of the gap. A small lens aperture was used to maximize the focal depth of field.

We illuminated and heated the test area to 35 °C with 500 W work lamps (Coleman Cable, Waukegan, IL, USA), three from the dorsal side, and two laterally. To track the animal and robot, we attached a BEEtag [76], printed on standard office paper, and attached it to the cardstock and to the dorsal side of the body using double-sided tape.
2.4.1. Animal experiments

We filmed the animal experiments using Photron Mini UX100 cameras (Photron USA, San Diego, CA, USA) with a resolution of 1280 × 1024 pixels. To track the animals, we attached a 1.6 cm × 1.6 cm BEEtag [76] to the dorsal surface of the wings directly above the body center of mass (CoM) [77] (figure 2(a)). We attached the tag to the animal using a combination of hot glue, super glue, and baking powder (as an accelerant). The animals were allowed to rest for at least 1 h after the tags were attached.

We placed the cockroaches on the track one at a time for testing. To elicit a rapid escape response, we prodded the posterior and abdomen of the animal with a rod wrapped in paper tape. The animal ran towards the obstacle between two walls that funneled it towards the middle of the track. To encourage the animal to seek shelter [23, 45, 78], we placed a shaded overhang after the obstacle. The animals were allowed to rest for 1–2 min before each trial.

2.4.2. Robot experiments

We filmed all the robot experiments using Fastec IL5 cameras (Fastec Imaging, San Diego, CA, USA) with a resolution of 1920 × 1080 pixels. For robot tracking with no tail, we attached a 5 cm × 5 cm tag above the robot body CoM using 0.3 cm thick polystyrene. The polystyrene was robust in cases where the robot flipped over. For experiments with a tail, we placed an additional 5 cm × 5 cm tag on the front shell of the robot, and the body CoM tag was moved to the side of the tail servo motor so that it was not obstructed by the tail when actuated. Three 3.8 cm × 3.8 cm tags were added to the dorsal side, ventral side, and tip of the tail to track tail motion.

2.4.3. Tailed robot experiment

To study how initial body pitch and initial body pitch angular velocity affected gap traversal, we added an active tail to the posterior end of the robot. We fastened the base of the 12 cm long active tail (3D printed from PLA plastic) to a high torque servo motor (Futaba BLS274SV, Futaba, Champaign, IL) and attached a 33 g mass at its distal end. The active tail rotated within the sagittal plane at a maximal angular speed of 315 ± 115° s⁻¹. We tuned the base position (6 cm from the posterior end) and actuation timing (140 ± 14 ms prior to reaching the obstacle) of the active tail so that the robot’s body pitch increased from −4° ± 8° prior to actuation to 8° ± 6° after actuation. The servo motor actuation was controlled by an Arduino Uno micro-controller and a motor controller (Qunqi L298 Dual H-bridge motor driver module). For this experiment, the robot legs were still under open-loop control. Adding the tail increased the robot’s total body inertia along the pitch axis about its posterior end from 0.004 kg m²–0.007 kg m² and did not change the robot’s maximal running speed (P = 0.4, Student’s t-test).

To test the effect of body pitching, we ran the tailed robot perpendicular to the gap at a constant speed of 190 ± 16 cm s⁻¹ and varied whether the tail was activated or not for each gap length. We empirically timed the tail activation so that the robot’s body pitch increased from −4° ± 8° prior to actuation to 8° ± 6° after actuation. The servo motor actuation was controlled by an Arduino Uno micro-controller and a motor controller (Qunqi L298 Dual H-bridge motor driver module). For this experiment, the robot legs were still under open-loop control. Adding the tail increased the robot’s total body inertia along the pitch axis about its posterior end from 0.004 kg m²–0.007 kg m² and did not change the robot’s maximal running speed (P = 0.4, Student’s t-test).
nite body length, and a total of 50 trials for the experiment with tail actuation for a gap of 0.8, 1.0, 1.2, 1.4, and infinite body length \((N = 1, n = 90)\).

2.5. Data analysis

2.4.4. Kinematic tracking

To calibrate the cameras, we made a calibration grid with Lego bricks (The Lego Group, Bilund, Denmark) and placed it in view of all the cameras prior to each experiment session. We obtained intrinsic (focal length and lens distortion) and extrinsic (relative position and rotation) camera parameters using direct linear transformation (DLT) 3D reconstruction [79]. After the experiments, we imported image sequences into a custom MATLAB script that tracked the tags in each camera view using the BEEtag code [76]. With an additional custom DLT script [79], we obtained the 3D position \((x, y, z)\) and orientation (Euler angles yaw \(\alpha\), pitch \(\beta\), and roll \(\gamma\)) of the tags (figures 2(a) and (b)). To verify the BEEtag accuracy, we 3D printed a calibration object with nine BEEtags equally spaced 7 cm apart in a grid in the horizontal plane, but orientated at pitch and yaw angles from \(0°\)–\(60°\) in an increment of \(30°\). We then calculated the error by measuring the 3D position and orientation of all the tags and compared them with the designed values. We found that the BEEtags accurately measured the body orientation angles (standard deviation (s.d.) of error = \(1.1°\)) and fore- and aft position and body height (s.d. of error = \(1.2\) mm). We measured the position and orientation of the CoM relative to the tag and inferred the CoM for each animal. To examine how the animal and robot’s head interacted with the obstacle, we calculated the head position from the tag position, assuming that the body and head together acted as a rigid body. This was a reasonable approximation: our manual tracking verified that the head moved little (<\(0.13\) cm) relative to the tag throughout each running cycle, and only moved slightly longer (\(0.27\) cm) when the animal flexed its neck joint and abdomen while gripping and climbing onto the far side of the gap. For the robot, we generated a 3D point cloud of the shell using a CAD model (Solidworks, Solidworks Corporation, Waltham, MA, USA) to obtain the head position (anterior end of the shell) relative to the tag position.

2.5. Data analysis

For all the experiments, we observed only a small difference between the velocity heading and body yaw immediately prior to the gap encounter (\(7° \pm 8°\) for the animal, \(2° \pm 8°\) for the robot). Therefore, we assumed that the velocity heading always equaled body yaw. We defined the angle of incidence \(\theta_b\) as the angle between the velocity heading and the forward \(+x\) direction at the time of the obstacle encounter, which equaled body yaw \(\alpha_0\) at the time of the obstacle encounter (see figures 2(a) and (b)). We calculated the angle of incidence \(\theta_b\) initial body pitch \(\beta_0\), initial body pitch angular velocity \(\omega_\beta\), initial body roll \(\gamma_0\), and approach speed \(v_0 = v_0 \cos \theta_b\) (speed perpendicular to the gap) immediately before the animal and robot encountered the near edge of the gap obstacle. Our use of the approach speed accounted for any motion not perpendicular to the gap. All metrics were averaged over four frames to reduce instantaneous measurement error, except for the initial body pitch angular velocity averaged from when the head reached, until the body CoM passed, the near edge of the gap to account for the large noise. We reported both the body yaw \(\alpha\) and angle of incidence \(\theta_b\) in absolute values due to lateral symmetry.

To determine whether the animal and robot approached and traversed the gap obstacle at speeds comparable to that during walking or running, we used the Froude number [80] \(Fr = v_0 (g_hip)^{-1/2}\), where \(g\) is acceleration due to gravity, \(v_0\) is the approach speed, and \(z_\text{hip}\) is hip height (\(0.5\) cm for the animal, \(2.5\) cm for the robot). As a common form of normalized speed of terrestrial locomotion, the Froude number is a good predictor of the speed at which legged animals transition from a walking to a running gait [80]. For the dis- cooid cockroach, this transition occurs at normalized speeds of \(Fr = 1.5–1.7\) [6, 71]. In the remainder of the paper, we refer to the locomotion speed perpendicular to the gap as the approach speed.

2.6. Statistics

For the animal gap experiments, 7 animals ran 10 trials each over 6 different gap lengths of 0.2, 0.4, 0.6, 0.8, 1, and infinite body length, resulting in a total of 420 accepted trials \((N = 7, n = 420)\). We rejected a trial whenever the animal collided with the wall, turned back, or stopped moving forward before encountering the gap, or if it moved out of the camera view during a traversal attempt. We varied the gap length but allowed the animal to run with its own chosen speed and velocity heading during each trial.

For the robot gap experiment, we systematically varied the gap length from 0.2–1 body length in increments of 0.2 body length and tested five different running speeds of \(78 \pm 15\) cm s\(^{-1}\), \(120 \pm 17\) cm s\(^{-1}\), \(170 \pm 18\) cm s\(^{-1}\), \(190 \pm 10\) cm s\(^{-1}\), and \(200 \pm 12\) cm s\(^{-1}\). The robot initial velocity heading was always perpendicular to the gap. We collected five trials for each combination of gap length and speed, resulting in a total of 110 trials \((n = 1, N = 110)\). For the
infinite length gap experiment, we tested five different approach speeds of 56 ± 13 cm s⁻¹, 78 ± 3.0 cm s⁻¹, 130 ± 11 cm s⁻¹, 180 ± 7.6 cm s⁻¹, and 200 ± 6.2 cm s⁻¹ and collected five trials at each speed for a total of 25 trials (n = 1, N = 25). With no tail, the robot had a small, nearly constant initial body pitch of 1° ± 3° (mean ± s.d.) immediately before reaching the near edge of the gap.

To compare metrics (approach speed, traversal probability, etc) across a gap of different lengths, we first calculated the mean for each individual for each gap, and then averaged over individual means to obtain the cross-individual average for that gap. Because we used only one robot, we averaged all trials for each gap to obtain the mean. To test which metrics (approach speed, initial body pitch, etc) affected traversal success or failure, we used multiple logistic regression. To test all the other metrics reported, we used repeated-measures ANOVA for the animal experiments and ANOVA for the robot experiments. Our multiple logistic regression and repeated-measures ANOVA accounted for individual variance by using the individual as one of the factors. We used Tukey’s honestly significant difference method for the post-hoc analysis where needed. All data are reported as means ± 1 s.d. unless otherwise specified.

3. Results and discussion

3.1. Dynamic gap traversal performance

In our experiments, both the discoid cockroach and the robot dynamically ran at high speeds to traverse the gap obstacle. The animal running speeds were 61 ± 16 cm s⁻¹ (12 ± 4 body length s⁻¹), up to nine times that in previous experiments of slow, quasi-static gap traversal [3, 16, 23]. The majority (83%) of the animal trials had a Froude number above that for a walking-to-running transition (Fr = 1.5) [6, 71]. Robot running speeds were 146 ± 48 cm s⁻¹ (6 ± 2 body length s⁻¹).

During such high-speed locomotion, both the animal and the robot were able to dynamically traverse a large gap of up to one body length. The probability of dynamic traversal decreased with gap length (figures 5(c) and (d), solid curve; animal: P < 0.0001, multiple logistic regression; robot: P < 0.0001, multiple logistic regression). When the gap was small enough (up to 0.4 body length), both the animal and the robot almost always traversed (93% for the animal and 100% for the robot, respectively). By contrast, for the largest gap of one body length tested, traversal was unlikely for both the animal and the robot. Only two out of the seven cockroaches were able to traverse it, while the rest always failed to traverse; the robot was never able to traverse it at its highest running speed (8.0 ± 0.5 body lengths s⁻¹). We verified that no animals could traverse a gap larger than 1 body length.

3.2. Gap bridging by head accurately predicts dynamic gap traversal

Careful observation of how the animal and robot’s body interacted with the gap revealed that the dynamic traversal of a gap obstacle was accurately predicted by whether the head was able to bridge across the gap. As the animal and robot encountered a large gap, each pair of legs sequentially lost contact with the near side, and the body began falling, with the head pitching downward (figures 3(a), (b), (e) and (f)), frames 1, 2, and 3; supplementary video 1, 2 (available online at stacks.iop.org/BB/13/026006/mmedia). Gap bridging by the head occurred when the animal or robot’s head reached the far edge of the gap before falling below it (figures 3(a) and (e); figures 3(c), (d), (g) and (h), red solid curve). When gap bridging by the head did not occur, the animal and robot always fell into the gap and failed to traverse (figures 3(b) and (f); figures 3(c), (d), (g) and (h)), blue dashed curves). We found that when gap bridging by the head occurred, traversal was successful 97% of the time for the animal and 81% of the time for the robot (the exceptional cases were due to a failure to grip, as discussed in section 3.6). When the head did not bridge the gap, both the animal and the robot failed to traverse 100% of the time. For a small enough gap (below 0.6 body length), the animal and robot always traversed because the head could not fall below the surface over such a small forward displacement (figures S4 and S5), and gap bridging was guaranteed.

The animal’s gap bridging using the head was likely a passive process, similar to our robot under open-loop control. Although cockroaches actively sweep their antenna to sense the physical environment during slow exploration [40], we observed that the animal’s antennae were held straight forward and slightly upward [40] and rarely came in contact with the ground. In addition, at the high speeds in our experiments, the animal had a very short time to respond if it did detect the gap (given a sensory delay of 6–40 ms [45] and a neuromuscular delay of 47 ms [6]), comparable to the time (60 ± 10 ms) for the animal’s head to fall and bridge across the largest gap. Further, we observed that the animal rarely changed speed (by 3 ± 4 cm s⁻¹), body pitch (by 2° ± 3°), or heading (by 6° ± 6°) before reaching the gap. Together, these observations indicate that the animal was likely unable to respond in time. By contrast, later during dynamic gap traversal, the animal likely used sensory feedback to initiate and control active body and leg adjustments (see section 3.6).

3.3. Template for dynamic gap traversal

The striking similarities in the traversal performance of the animal and the robot and the signature observation of the gap bridging by the head for successful traversal suggested a template for dynamic
gap traversal. A template is a useful modeling concept that allows a general, fundamental understanding of high-dimensional, nonlinear, multi-body dynamic locomotion phenomena by reducing the problem to as few degrees of freedom as possible [37, 75]. A few templates have already captured fundamental dynamics and revealed control strategies for common forms of terrestrial locomotion on 2D surfaces, such as dynamic walking [81, 82] and running on level ground [62, 63, 83] and dynamic climbing on vertical walls [10, 15]. Inspired by these successes, we take the next step in creating a template for the dynamic traversal of a large gap as representative of complex 3D terrains.

We approximated the animal and robot body as a rigid ellipsoid traveling forward at a constant approach speed, and calculated its dynamics during passive falling under gravity as it encountered a large gap (figure 4(a)). For simplicity, we assumed that the rigid body only rotated about the body pitch axis in the sagittal plane and had no body yaw or roll rotations. In addition, we assumed that the body gradually losing support as the legs gradually lost contact when the animal or robot ran past the near edge, we assumed that the force due to gravity increased proportionally to the length of the body beyond the near edge.

Using the Lagrangian method, we obtained the equation of motion during the pre-free falling phase (before the posterior passed the near edge) as:

$$\frac{d}{dt} \left( \frac{\partial L}{\partial \omega} \right) - \frac{\partial L}{\partial \beta} = 0$$  \hspace{1cm} (1)

with the Lagrangian

$$L = \frac{1}{2} I \omega^2 - \frac{mgb}{2} \sin \beta$$  \hspace{1cm} (2)

where $\beta$ is body pitch, $I = 1/3 mb^2$ is the moment of inertia for body pitching about the hinge on its end, $b$ is the length of the body (5 cm for the animal, 25 cm for the robot), $m$ is body mass (2.5 g for the animal, 194 g for the robot), $\omega$ is the pitch angular velocity, and $g = 9.81 \text{ m s}^{-2}$ is the gravitational acceleration.

From the equation of motion, we obtained the angular acceleration about the body pitch axis:

\begin{figure}
\centering
\includegraphics[width=\textwidth]{figure3}
\caption{Dynamic locomotion of the discoid cockroach and the robot over a large gap obstacle. Representative trials of (a) and (e) successful traversal and (b) and (f) failure to bridge the gap. (c) and (g) Head height as a function of the forward position of the head. (d) and (h) Body pitch as a function of the forward position of the head. In (c), (d), (g) and (h), the solid red and blue dashed curves and shaded areas represent means ±1 s.d. for the success and failure cases. Data are shown for the 0.8 body length (4 cm) gap as an example; data for other gap lengths have similar trends (figures S4–S7). For simplicity, only movement perpendicular to the gap (within the $x$-$z$ plane) is shown.}
\end{figure}
\[ \hat{\omega}(t) = \frac{\xi \text{mg} b}{2I} \cos(\beta(t)) \]  

where \( \xi = \left( \frac{\xi}{\beta(t)} \right) \frac{\cos(\beta(t))}{\cos(\theta)} \) is a linear force reduction factor that characterizes the loss of body support due to the legs losing surface contact, \( v \) is the speed along the long axis of the body, and \( \beta(t) \) is the angle of incidence that measures the deviation of body yaw and velocity heading from the forward direction (direction perpendicular to the gap).

Once the posterior end of the body reached the near edge of the gap, the body started to fall like a projectile with zero angular acceleration and \( \omega(t) = \omega_{\text{edges}} \), where \( \omega_{\text{edges}} \) is the angular velocity of the body when the posterior reached the near gap edge. The CoM height during the free falling phase is:

\[ z_{\text{CoM}}(\tilde{t}) = \frac{1}{2} g \tilde{t}^2, \]  

\[ \dot{\omega}(\tilde{t}) = 0 \]  

where \( \tilde{t} \) is the time after the posterior end passed the near edge of the gap. Note that we assumed that the velocity heading and body yaw were always aligned based on experimental observations (see section 2.5).

Finally, we used the Euler method to integrate forward in time to obtain the body pitch angular velocity \( \omega \), body pitch \( \beta \), the CoM position \( x_{\text{CoM}} \) along the forward direction, and the vertical CoM position \( z_{\text{CoM}} \) as a function of time \( t \):

\[ \omega(t + \Delta t) = \omega(t) + \dot{\omega}(t)\Delta t \]  

\[ \beta(t + \Delta t) = \beta(t) + \dot{\beta}(t)\Delta t \]  

\[ x_{\text{CoM}}(t + \Delta t) = x_{\text{CoM}}(t) + \dot{x}_{\text{CoM}}(t)\Delta t \]  

\[ z_{\text{CoM}}(t + \Delta t) = z_{\text{CoM}}(t) + \dot{z}_{\text{CoM}}(t)\Delta t \]  

where time step \( \Delta t = 0.001 \text{ s} \).

Assuming that gap bridging by the head results in traversal, the template allowed us to predict the maximal traversable gap length that the body could traverse for any given initial body pitch \( \beta_0 \), initial body pitch angular velocity \( \omega_0 \), and approach speed \( v_0 \) (see supplementary video 4). We first numerically calculated the time that the body had to attempt to bridge the gap, \( t_{\text{bridges}} \), i.e. the time for the rigid body’s head (free anterior end) to fall to surface level (zero height). The forward distance over which the head travelled, \( d = v_0 t_{\text{fall}} \), was then the maximal traversable gap length (figure 4(b)).

To validate the template, we ran the animal and the robot into an infinite gap (a cliff) to measure the maximal traversable gap length and compared it with the model predictions. Without any model fitting parameters, the template accurately predicted the maximal traversable gap length \( d \) for both the animal and the robot (figures 5(a) and (b)). We noted that the animal and the robot ran at a broad range of approach speed (animal: \( 37 \text{ cm s}^{-1} \leq v_0 \leq 100 \text{ cm s}^{-1} \); robot: \( 39 \text{ cm s}^{-1} \leq v_0 \leq 210 \text{ cm s}^{-1} \) ), initial body pitch (animal: \( 3^\circ \leq \beta \leq 24^\circ \); robot: \( -5^\circ \leq \beta \leq 9^\circ \) ), and initial body pitch angular velocity (animal: \( -401^\circ \text{ s}^{-1} \leq \omega_0 \leq 34^\circ \text{ s}^{-1} \); robot: \( -143^\circ \text{ s}^{-1} \leq \omega_0 \leq 103^\circ \text{ s}^{-1} \) ) in the experiment.

The template slightly under-predicted the maximal traversable gap length for the robot. One reason was due to its bouncier running gait, which resulted in a significant initial upward speed (\( 20 \pm 20 \% \) of its forward speed, versus the animal’s \( 5 \pm 4 \% \)), not accounted for by the horizontally approaching template. Additionally, because the robot’s head was in
front of its legs by a large distance (0.3 body length), it did not start to lose ground reaction force and fall into the gap until it had already entered the gap substantially. By contrast, in the model, the body started to lose ground reaction force and fall immediately upon reaching the gap.

Next, we tested the predictive power of the template for predicting traversal probability measured in the finite length gap experiments. Using the measured approach speed $v_0 = v \cos \beta_0$, initial body pitch $\beta_0$, and initial body pitch angular velocity $\omega_0$ from each experimental trial as initial conditions, we calculated the expected maximal traversable gap length from the model. If it exceeded the gap length being tested, the model predicted successful traversal (figure 4(b), red solid curve); otherwise, the model predicted failure (figure 4(b), blue dashed curve). Without any model fitting parameters, the template accurately predicted the observed monotonic decrease of traversal probability with gap length for both the animal and the robot (figures 5(c) and (d), dashed curves). For a small gap length, the quantitative agreement between the animal data and model predictions was excellent. The over-prediction of the traversal probability for the two largest gaps was because the model did not account for the animal or robot failing to grip the far side of the gap after the head bridged (see section 3.6). The template under-predicted the robot traversal probability for the 0.4 and 0.6 body length gaps due to its bouncier gait and its leg being substantially behind the head, as discussed above.

3.4. Principles of dynamic gap traversal from template

Our experimentally validated template allowed us to gain insights into the general principles of dynamic gap traversal. Using the template, we numerically calculated how the maximal traversable gap length depended on approach speed $v_0$, initial body pitch $\beta_0$, and initial body pitch angular velocity $\omega_0$ over a broad range of parameter space. We discovered that the maximal traversable gap length increased with all these locomotor parameters (figure 6). This is intuitive because faster running allows an animal or robot to travel forward by a larger distance before its head falls below the surface level, and a higher initial body pitch and higher initial pitch angular velocity gave the animal or robot a longer time to travel forward before the head fell below the surface level.

These general principles from the template gave us two predictions: first, for a given gap length, the animal or robot running at a higher approach speed, higher initial body pitch, and/or higher initial body pitch angular velocity should be more likely to successfully traverse. Second, as the gap length increased, the approach speed, initial body pitch, and initial body pitch angular velocity required to traverse should also increase (supplementary video 4). Indeed, in our finite

Figure 5. The template accurately predicted the dynamic gap traversal. (a) and (b) Comparison of maximal traversable gap length measured from the infinite length gap experiment with that predicted from the model. The black diagonal line indicates a perfect match between the data and the model. (c) and (d) Comparison of experimental (filled circles, solid curve) and model-predicted (open circles, dashed curve) traversal probability. Error bars represent 95% confidence interval.
In the infinite length gap experiments, both the animal and robot reached a maximal traversable gap length longer than its body length (figures 5(a) and (b); animal: 1.25 body length; robot: 1.6 body length). However, the traversal of these gap lengths is unlikely. The animal would collide with the far side of the gap at high speeds, making gripping difficult due to the slow reaction time. Additionally, the animal and the robot often could not grip before falling into the gap. Therefore, the model likely predicted a larger maximal traversable gap length than is physically possible due to the grip failure (see section 3.6).

Given its simplicity and predictive power of dynamic gap traversal over a broad range of locomotor and terrain parameter space, our model provided
3.5. Body pitch control to enhance dynamic gap traversal

The principles from our template provided useful control strategies for robots to traverse large gap obstacles. For example, for a given robot already reaching its maximal speed with a head-on approach (zero angle of incidence \( \theta_0 \)), a higher initial body pitch and higher initial body pitch angular velocity would allow it to traverse a larger gap. To demonstrate this, we tested the robot with a bio-inspired active tail \([21, 33]\) that enabled adjustment of body pitching (figure 8(a)).

As the robot approached the gap, the tail was rotated backwards and suddenly stopped, and its angular momentum was quickly transferred to the body, causing the robot to pitch upward (supplementary video 3).

As predicted by our template, we found that the active tail significantly increased the robot’s ability to traverse a large gap by increasing \( \omega_0 \) and the robot’s pitch moment of inertia (figure 8(b)). At an approach speed of \( 190 \text{ cm s}^{-1} \pm 20 \text{ cm s}^{-1} \), the robot’s initial body pitch angular velocity increased from \( -11 \pm 74 \text{ ° s}^{-1} \) without tail actuation to \( 57 \pm 57 \text{ ° s}^{-1} \) with tail actuation (figure 6(b), circles; \( P = 0.011 \), ANOVA).

Although tail actuation did not significantly increase initial body pitch (\( P = 0.8 \), Student’s \( t \)-test), it did increase the maximal body pitch as the robot ran over the near edge of the gap, from \( 5^\circ \pm 6^\circ \) without tail actuation to \( 8^\circ \pm 6^\circ \) with tail actuation (\( P = 0.045 \), ANOVA). Together, these changes in body pitching not only increased the robot’s traversal probability for all but the smallest gap tested, but also allowed it to traverse a gap as large as 1.2 body length—a 50% increase (figure 9(b), open circles).

Although we only demonstrated using an active tail to increase the initial body pitch and initial body pitch angular velocity to aid large gap obstacle traversal, the body pitch can continue to be controlled during the entire falling phase \([21, 74]\). Further, other body pitch control methods may also be employed, such as pushing more forcefully with the fore and mid legs \([34, 36]\) and hyperextending a flexible body \([90]\). Future studies should test the feasibility and performance of these control strategies and add sensory feedback \([2]\) to further improve robot performance traversing large gap obstacles.

3.6. Body flexibility and leg gripping ability enhance large gap traversal

Our template of dynamic large gap traversal accurately captured the body dynamics until the head bridged across the gap using kinetic energy. However, to traverse, the animal or robot must also be able to continue to move forward until its entire body makes it across the gap. This is especially challenging over a large gap, because when the head bridges, the majority of the body or even the entire body is still above the gap without ground reaction forces (figures 3(a, b, e, and f), frame 2). Therefore, generating sufficient upward and forward forces against the far edge of the gap is essential for traversal.

Careful examination revealed that the successful traversal of a large gap obstacle after gap bridging by the head often required additional leg gripping and pulling (66% and 100% of the time for 0.8 and 1 body length gaps, respectively), presumably using sensory feedback (figure 9(a)). The animal’s ability to flex its body and use the various structures on its articulated legs and feet \([91–93]\) to grip when they touched the far side surface played an important role in this process. We observed that, after the animal’s head bridged across a large gap, the body often flexed substantially, while its middle and fore legs pulled on the far edge and its hind legs pushed against the vertical surface (figure 9(a), frames 3, 4, 5). Body flexion not only allowed the fore legs to better reach forward and downward...
to pull, but also allowed the hind legs to reach downward and forward to push [90]. When the animal was unable to do these sufficiently in time, its body pitched backwards, resulting in falling back into the gap, even when the head bridged across (figure 9(b), frames 3, 4, and 5).

By contrast, the robot’s body and legs are relatively rigid and lack gripping mechanisms to pull itself onto the far side of the gap (figure 9(c)). Even when the robot succeeded in traversing a large gap, it did not grip, but simply continued to use the same open-loop gait. As a result, even when the head bridged across the gap, the robot was more likely (19% probability) to fall backwards and fall back into the gap than the animal (3% probability) ($P < 0.0001$, repeated-measures ANOVA). In addition, the probability of grip failure for both the animal and robot increased for the largest gap lengths (animal: $P < 0.0001$, repeated-measures ANOVA; robot: $P < 0.0001$, ANOVA).

Based on these observations, we posit that a two-link body modeling a flexible body with three spring legs modeling fore, mid, and hind feet is a likely candidate for an anchor-level model [37] to better capture the dynamics of the final phase of the dynamic traversal of large gaps and to further explore the role of active body pitch control [21] (figure 9(d)). Future experiments should better understand the biological principles of such active body and leg use and validate the anchor model, and use them to improve the ability of robots to dynamically traverse large gaps.

4. Conclusions

We comparatively studied rapid-running discoid cockroaches and a cockroach-inspired robot as a physical model to discover the performance limits and general principles of the dynamic traversal of a large gap obstacle. We discovered that, similar to bridging small gaps between footholds on low contact area surfaces [28] and uneven terrain [6], both the animal and the robot can use translational and rotational kinetic energy to dynamically traverse a gap obstacle as large as its body length. This is rarely possible during quasi-static gap traversal [4, 38, 94] and is more comparable to snakes using kinetic energy to lunge across large gaps [19]. Traversal was less likely as the gap became larger and was accurately predicted by whether the head bridged across the gap. Inspired by the similarity in the animal and robot observations, we created a template that well described body dynamics during passive falling over the gap and quantitatively predicted traversal performance. Our template revealed that a high approach speed, high initial body pitch, and high initial body pitch angular velocity all facilitated dynamic traversal by allowing the head to travel further to bridge a gap larger than would be possible during quasi-static gap traversal [4, 38, 94]. Despite their similarities, the animal is still far better than the robot at dynamically traversing large gap obstacles thanks to its stronger ability to grip, push, and pull itself onto the far side of the gap.

![Figure 9. Traversal often required gripping and body flexion. (a) Successful leg gripping aided by body flexion by the animal when dynamically traversing a large gap. (b) Grip failure of the animal. (c) Grip failure of the robot. (d) A proposed two-link anchor model for dynamic gap traversal with active body flexion and leg gripping.](image)
Our study is a major step in expanding the emerging field of terrodynamics of biological and robotic locomotion [8]. Our template is the first to quantitatively predict dynamic locomotion beyond planar surfaces [10, 37, 43], and has expanded the usefulness of templates [37] into complex 3D terrains. Future studies to systematically vary locomotor and terrain parameters [8, 72, 95, 96] and create new templates for other types of terrains will advance our understanding of how animals move in nature and improve robotic locomotion in complex natural and artificial environments. An immediate next step is to discover the general principles for the dynamic traversal of another simple yet general obstacle—a large bump—which we explore in our companion study [67].

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Supplementary Movies:

**Movie 1.** A cockroach attempts to dynamically traverse a gap.

**Movie 2.** Gap traversal by a legged robot.

**Movie 3.** Active body pitch control increased gap traversal performance.

**Movie 4.** A template model for gap traversal.