

Simultaneous Contact, Gait and Motion Planning for Robust Multi-Legged Locomotion via Mixed-Integer Convex Optimization

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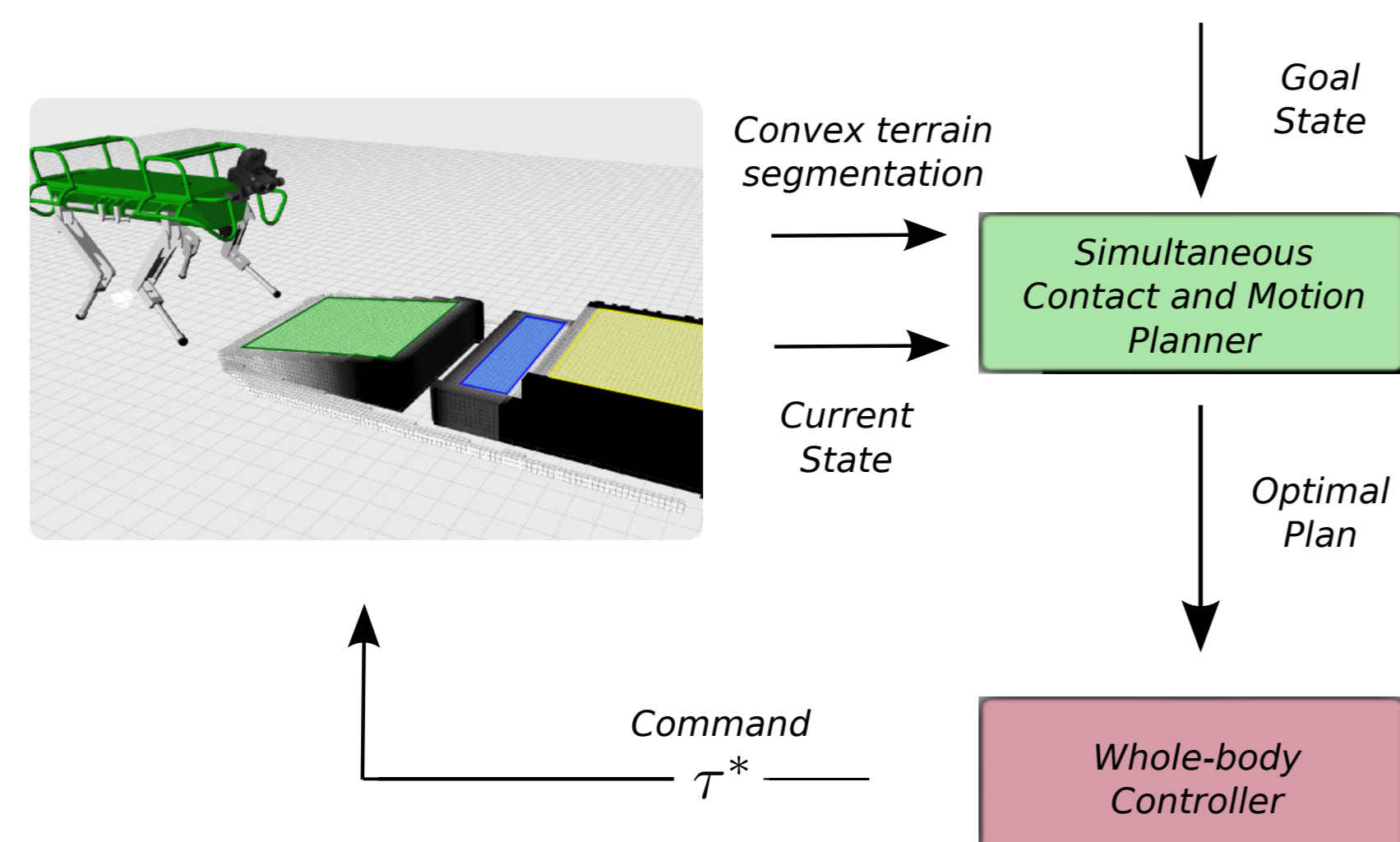
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Introduction

Reasoning about contacts and motions simultaneously is crucial for generating complex whole-body behaviors. We propose a mixed-integer convex formulation to plan simultaneously contact locations, gait transitions and motion, in a computationally efficient fashion.

Approach Overview



Trajectory Optimization

We formulate a convex trajectory optimization:

$$\min_{\mathbf{r}, \mathbf{k}, \mathbf{p}_l, \lambda_l} g_T + \sum_{k=1}^N g(k) \quad (1)$$

The running cost $g(k)$ maximizes the stability of the motion, while seeking for the fastest and smoothest gait:

$$g(k) = \|\ddot{\mathbf{r}}_k\|_{\mathbf{Q}_v} + \|\lambda_{l,k}\|_{\mathbf{Q}_f} + q_u \mathbf{U}_k + q_t t_k - q_\alpha \alpha_k, \quad (2)$$

- 1 Minimize the CoM acceleration $\ddot{\mathbf{r}}$.
- 2 Minimize the contact forces magnitude $\|\lambda\|$.
- 3 Minimize the upper bound of quadratic terms $\mathbf{U} = (\mathbf{u}^-, \mathbf{u}^+)$.
- 4 Maximize the stability margin α .
- 5 Minimize the execution time.

The terminal cost g_T biases the plan towards its goal (\mathbf{r}_G):

$$g_T = \|\mathbf{r}_T - \mathbf{r}_G\|_{\mathbf{Q}_g} \quad (3)$$

In practice, we add a small cost to $\|\dot{\mathbf{k}}\|_{\mathbf{Q}_k}$ in order to generate smoother motions.

Whole-body Control

Reference CoM acceleration ($\ddot{\mathbf{r}}^r \in \mathbb{R}^3$) and body angular acceleration ($\dot{\omega}_b^r \in \mathbb{R}^3$) through a *virtual model*:

$$\begin{aligned} \ddot{\mathbf{r}}^r &= \ddot{\mathbf{r}}^d + \mathbf{K}_r(\mathbf{r}^d - \mathbf{r}) + \mathbf{D}_r(\dot{\mathbf{r}}^d - \dot{\mathbf{r}}), \\ \dot{\omega}_b^r &= \dot{\omega}_b^d + \mathbf{K}_\theta e(\mathbf{R}_b^d \mathbf{R}_b^T) + \mathbf{D}_\theta(\omega_b^d - \omega_b), \end{aligned} \quad (4)$$

where $\mathbf{K}_r, \mathbf{D}_r, \mathbf{K}_\theta, \mathbf{D}_\theta \in \mathbb{R}^{3 \times 3}$ are positive-definite diagonal matrices of proportional and derivative gains, respectively.

We formulate the tracking problem using QP with the generalized accelerations and contact forces as decision variables, $\mathbf{x} = [\ddot{\mathbf{q}}^T, \lambda^T]^T \in \mathbb{R}^{6+n+3n_l}$:

$$\mathbf{x}^* = \arg \min_{\mathbf{x}} g_{err}(\mathbf{x}) + \|\mathbf{x}\|_{\mathbf{W}}; \quad \mathbf{A}\mathbf{x} = \mathbf{b}, \quad \underline{\mathbf{d}} < \mathbf{C}\mathbf{x} < \bar{\mathbf{d}} \quad (5)$$

$$g_{err}(\mathbf{x}) = \left\| \begin{array}{c} \ddot{\mathbf{r}} - \ddot{\mathbf{r}}^r \\ \dot{\omega}_b - \dot{\omega}_b^r \end{array} \right\|_{\mathbf{S}} \quad (6)$$

The equality constraints $\mathbf{A}\mathbf{x} = \mathbf{b}$ encodes dynamic consistency, stance condition and swing task. The inequality constraints $\underline{\mathbf{d}} < \mathbf{C}\mathbf{x} < \bar{\mathbf{d}}$ encode friction, torque, and kinematic limits [1].

We map the optimal solution \mathbf{x}^* into desired feed-forward joint torques $\tau_{ff}^* \in \mathbb{R}^n$:

$$\tau_{ff}^* = [\mathbf{M}_{bj}^T \quad \mathbf{M}_j] \ddot{\mathbf{q}}^* + \mathbf{h}_j - \mathbf{J}_{cj}^T \lambda^* \quad (7)$$

These are summed with the joint PD torques (i.e. feedback torques τ_{fb}) to form the desired torque command τ^d :

$$\tau^d = \tau_{ff}^* + PD(\mathbf{q}_j^d, \dot{\mathbf{q}}_j^d), \quad (8)$$

which is sent to a low-level joint-torque controller.

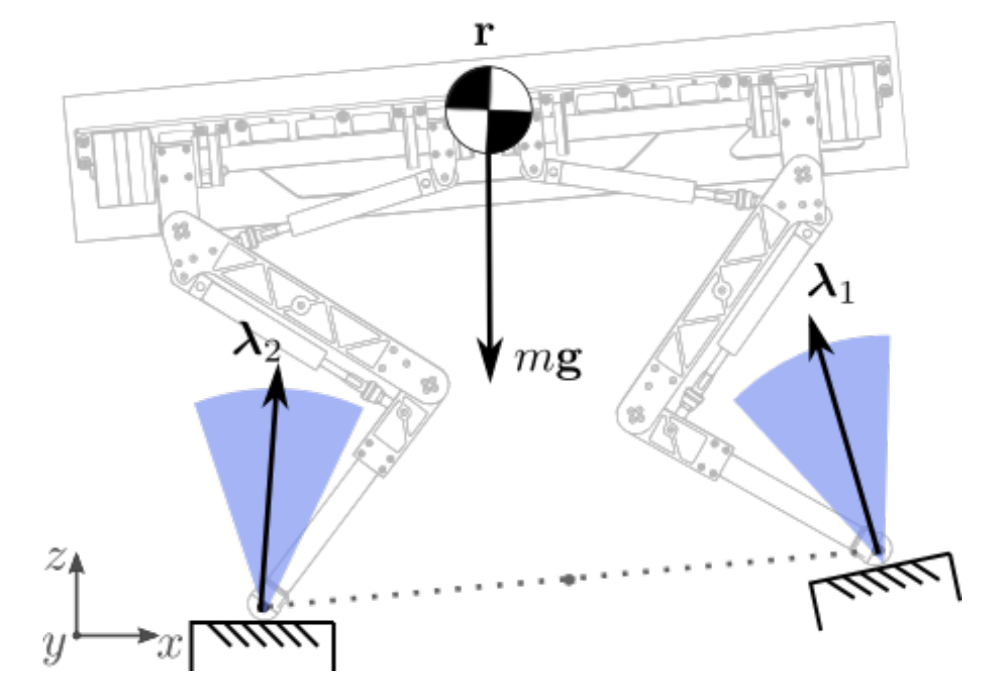
Simultaneous Contact and Motion Planning

A. Centroidal Dynamics

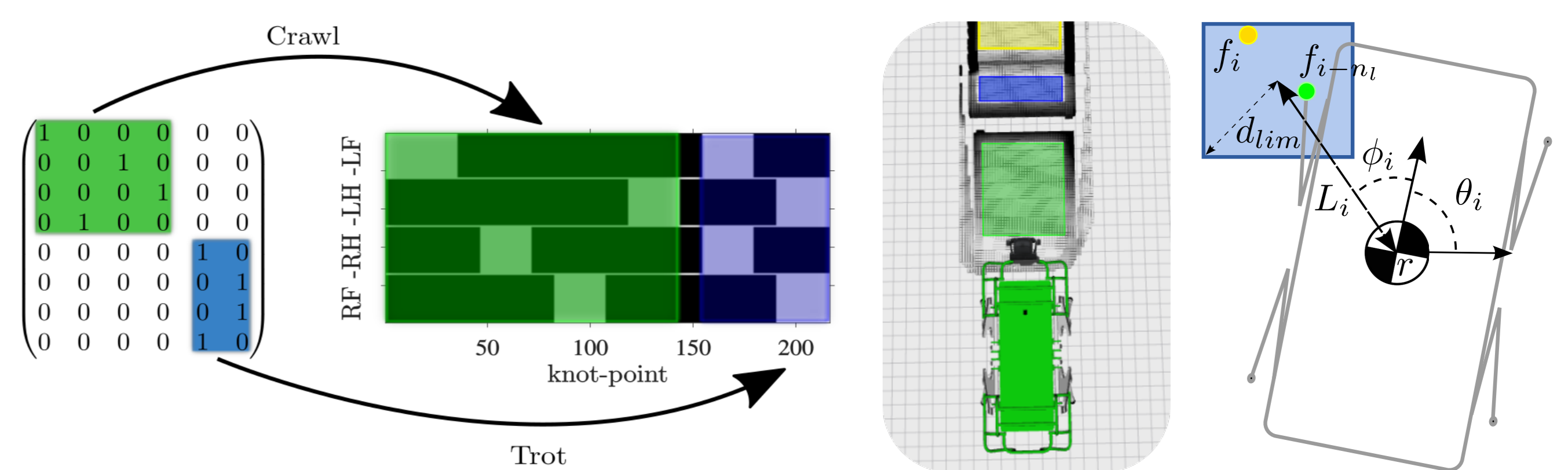
$$\begin{bmatrix} m\ddot{\mathbf{r}} \\ \dot{\mathbf{k}} \end{bmatrix} = \begin{bmatrix} m\mathbf{g} + \sum_{l=1}^{n_l} \lambda_l \\ \sum_{l=1}^{n_l} (\mathbf{p}_l - \mathbf{r}) \times \lambda_l \end{bmatrix}, \quad (9)$$

- CoM position \mathbf{r}
 - Angular Momentum \mathbf{k}
 - Contact force of end-effector λ_l
 - Position of end-effector \mathbf{p}_l
- where $\mathbf{p}_l - \mathbf{r}$ can be described as bilinear function [2] and decomposed as:

$$\mathbf{a}\mathbf{b} = \frac{\mathbf{u}^+ - \mathbf{u}^-}{4} \quad \mathbf{u}^+ \geq (\mathbf{a} + \mathbf{b})^2 \quad \mathbf{u}^- \geq (\mathbf{a} - \mathbf{b})^2. \quad (10)$$



B. Gait Sequence



A gait matrix [3] $\mathbf{T} \in \{0, 1\}^{N_r \times N_t}$ where $\mathbf{T}_{ij} = 1$ means that the robot will move to the i^{th} contact location at the j^{th} time-slot.

- number of contacts N_r
- number of time slots N_t

Each contact location is reached once: $\sum_{j=1}^{N_t} \mathbf{T}_{ij} = 1, \forall i = 1, \dots, N_r$.

C. Contact Location

We constrain the contacts to lie within one of N_r convex safe contact surfaces (each represented as a polygon $\mathcal{R} = \{\mathbf{c} \in \mathbb{R}^3 | \mathbf{A}_r \mathbf{c} \leq \mathbf{b}_r\}$).

After assigning the contact to swing, we optimize the contact locations $\mathbf{f} = (f_x, f_y, f_z, \theta)$ and we assign this contact to one of the N_r using a binary matrix:

$$\sum_{r=1}^{N_r} \mathbf{H}_{ir} = 1, \quad \mathbf{H}_{ir} \Rightarrow \mathbf{A}_r \mathbf{f}_i \leq \mathbf{b}_r \quad (11)$$

We approximate the kinematic limits as a bounding box with respect to the CoM:

$$\left| \mathbf{f}_i - \left[\mathbf{r}_{T(i)} + L_i \begin{pmatrix} \cos(\theta_i + \phi_i) \\ \sin(\theta_i + \phi_i) \end{pmatrix} \right] \right| \leq d_{lim}, \quad (12)$$

- CoM position at i transition $\mathbf{r}_{T(i)}$
- Diagonal of the bounding box d_{lim}
- Distance from the trunk of leg L_i

(the trigonometric functions are decomposed in piecewise linear functions) [4].

D. End-effector Trajectories

We define $\gamma(j, t)$ as swing reference trajectory, connected to adjacent contacts, where:

- j indicates the time-slot
- $t \in [1, \dots, N_k]$ all the knots per time-slot

The leg reaches the contact position \mathbf{f}_i at the end of the j slot:

$$\mathbf{T}_{ij} \Rightarrow \mathbf{p}_{l(i)\gamma(j, N_k)} = \mathbf{f}_i, \quad (13)$$

where $l(i)$ is the leg number for the i^{th} contact.

To constrain that the leg remains stationary when there is no transition:

$$\sum_{i \in C(l)} \mathbf{T}_{ij} = 0 \Rightarrow \mathbf{p}_{l\gamma(j, t)} = \mathbf{p}_{l\gamma(j, 1)} \quad \forall t \in [2, \dots, N_k], \quad (14)$$

where $C(l)$ are the contact indexes assigned to the l^{th} leg.

We ensure kinematic feasibility by constraining the CoM position with respect to the end-effectors (bounding box constraint):

$$\mathbf{d}_- < \mathbf{r}_j - \frac{\sum_{l=1}^{n_l} \mathbf{p}_{lj}}{n_l} < \mathbf{d}_+ \quad (15)$$

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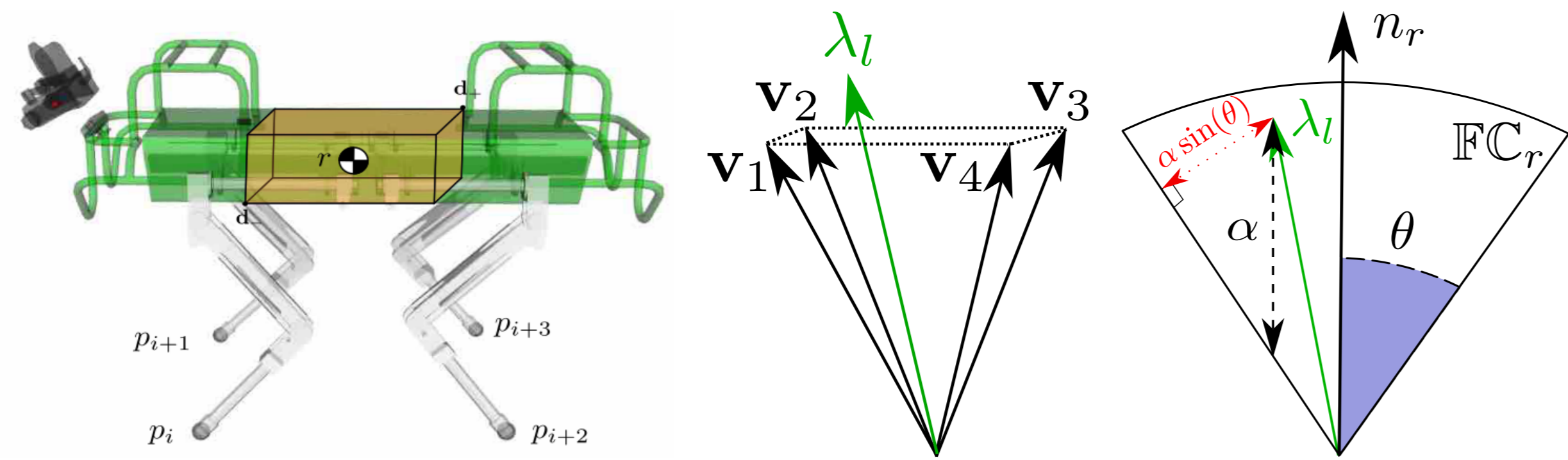
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Simultaneous Contact and Motion Planning



E. Contact Dynamics

If the l^{th} leg is in swing mode, there is no contact force:

$$\sum_{i \in C(l)} \mathbf{T}_{ij} = 1 \Rightarrow \lambda_{l\gamma(j,t)} = 0, \quad \forall t \in NC(j), \quad (16)$$

where $NC(j)$ is the set of knots in the j^{th} slot used for the swing (complementarity constraint [5]).

Stability in non-coplanar contact conditions:

$$\lambda_{lj} \in \mathbb{F}C_r \Rightarrow \lambda_{lj} = \sum_{e=1}^{N_e} \rho_e \mathbf{v}_{r_e}, \quad \rho_1, \dots, \rho_{N_e} > 0,$$

where ρ_e are positive multipliers on each cone edge.

To add robustness to the motion, we maximize the distance between the nonlinear friction cone boundary and the force vector:

$$\alpha = \arg \max_{\alpha} \text{ s.t. } \lambda_{lj} - \alpha \hat{\mathbf{n}}_r \in \mathbb{F}C_r,$$

We introduce the following linear constraint over each safe surface:

$$\mathbf{T}_{ij} \text{ and } \mathbf{H}_{ri} \Rightarrow \lambda_{l(i)\gamma(j)} - \alpha_{l(i)\gamma(j)} \hat{\mathbf{n}}_r \in \mathbb{F}C_r, \quad \alpha \geq 0. \quad (17)$$

Since the contact cone must not change when it is in stance phase, we also add the constraint:

$$\sum_{l \in C(i)} \sum_{t \in NS(j)} \mathbf{T}_{lt} = 0, \quad (18)$$

$$\Rightarrow \lambda_{l(i)\gamma(j)} - \alpha_{l(i)\gamma(j)} \mathbf{n}_r \in \mathbb{F}C_r,$$

where $NS(j)$ is the set of time-slots succeeding j .

G. Approximate Torque Limits

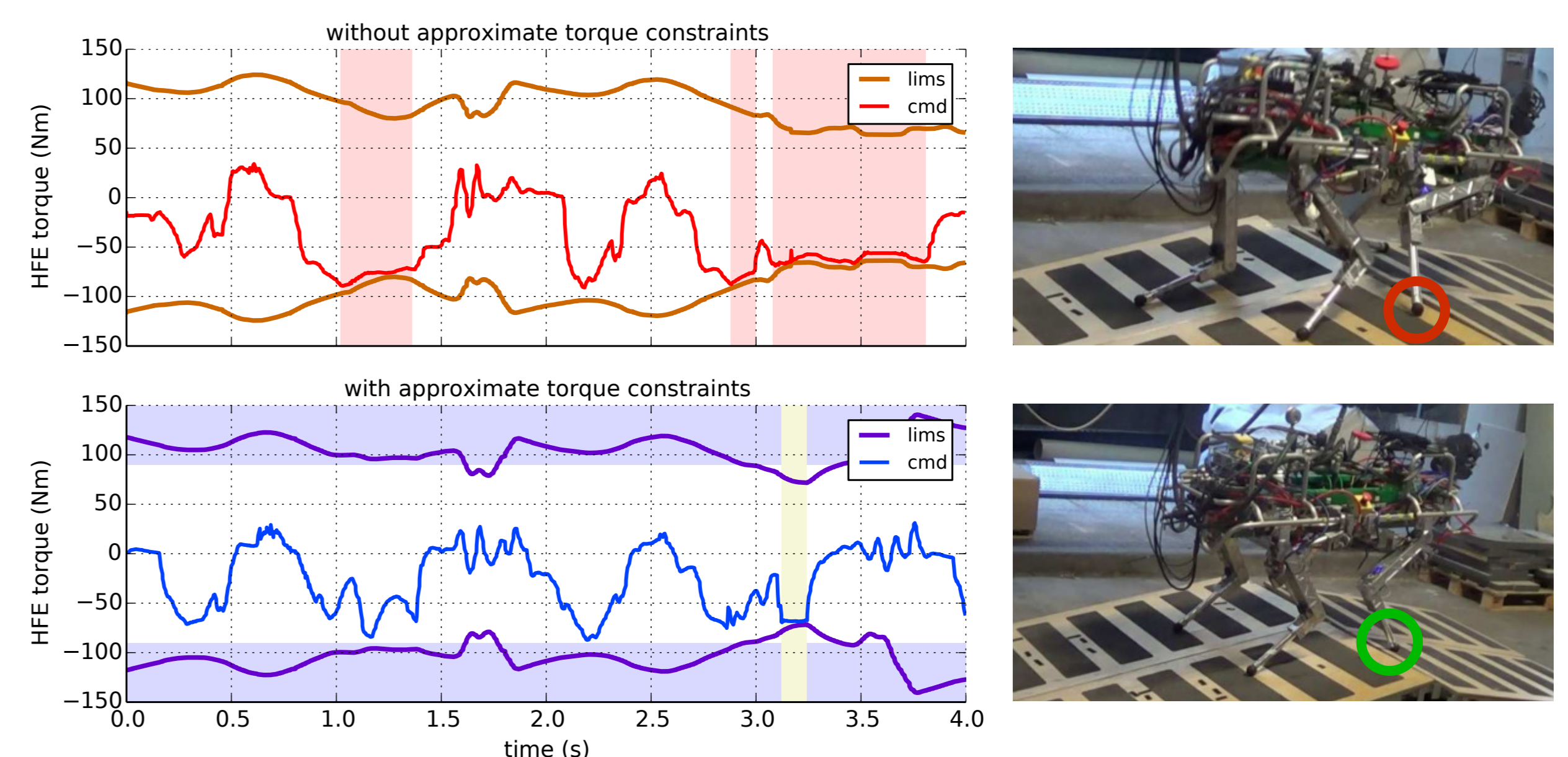
We approximate the torque limits using a quasi-static motion assumption:

$$\mathbf{J}_{l,j}^T \lambda_{l,j} \leq \boldsymbol{\tau}_{max}, \quad \Rightarrow \quad \mathbf{J}_{l,j}^{*T} \lambda_{l,j} \leq \boldsymbol{\tau}_{max},$$

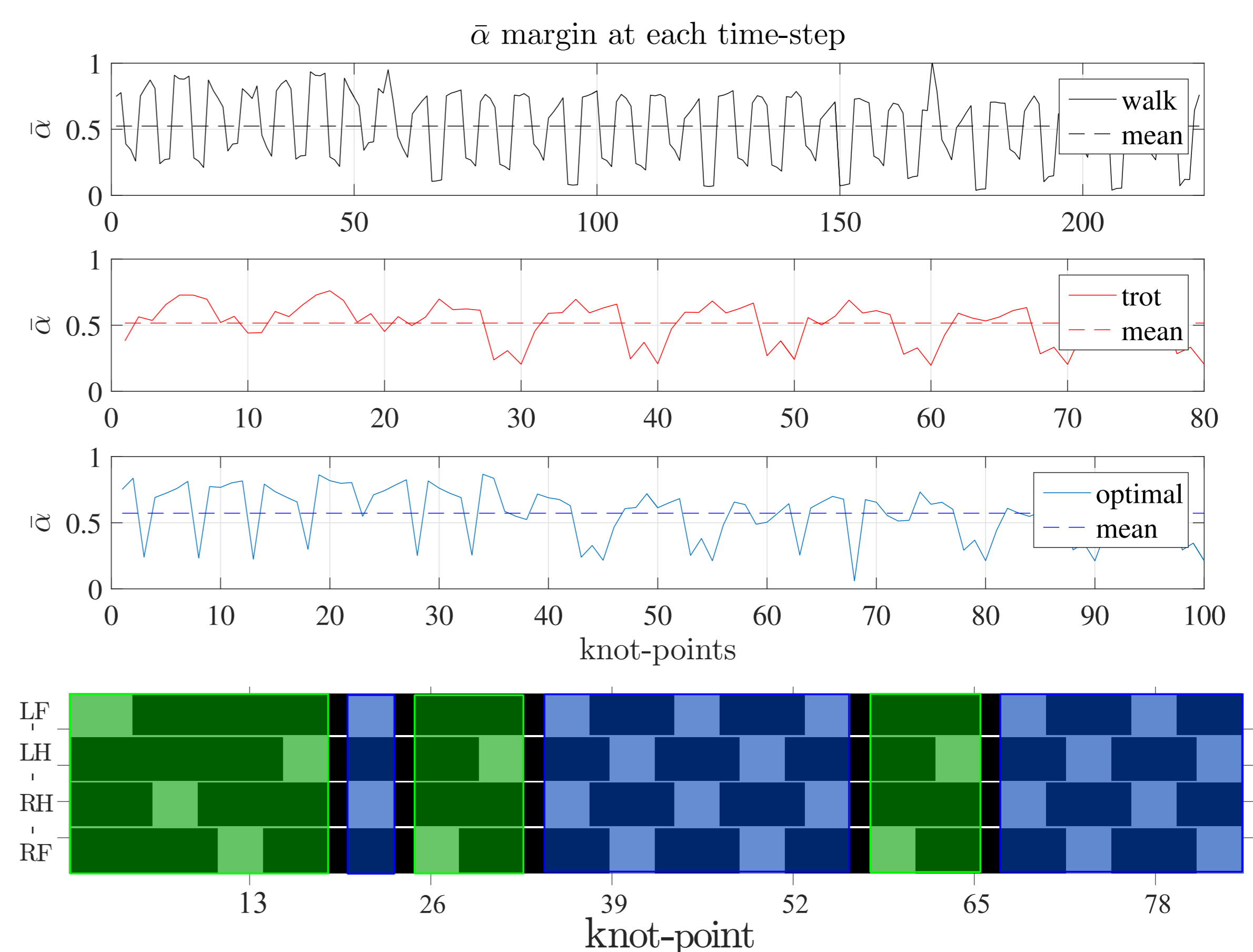
where $\mathbf{J}_{l,j} \in \mathbb{R}^{3 \times 3}$ is the operational space foot Jacobian for the l^{th} leg at the j^{th} knot, and $\boldsymbol{\tau}_{max}$ are the joint torque limits of the leg.

Experimental Validation

Approximate Torque Constraints



Automatic Gait Discovery



Top: Normalized α margin for different gaits. Bottom: Resulting optimal gait sequence for navigating in roof-like terrain.

Computation Time

Experiment	convex surfaces	Gait	mean time (s)
Exp. 1	3	Walk	0.47
Exp. 2	3	Walk	0.64
Exp. 3	4	Walk	0.44
Exp. 4	3	Walk	0.48
Exp. 4	3	Trot	0.51
Exp. 4	3	Free	1.62

Computation time in an un-optimized Matlab code.

Conclusions

We have presented a novel approach for simultaneously planning contacts and motions on multi-legged robots based on MICP. Our approach is able to handle **complex terrain**, while also providing formal **robustness guarantees** on the plan and allows for **automatic gait discovery**. We employ both a state-of-the-art whole-body controller [1] and state estimation [6]. We demonstrate the approach's capabilities by traversing challenging terrains with the HyQ robot.

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