

Planar Bipedal Locomotion with Nonlinear Model Predictive Control: Online Gait Generation using Whole- Body Dynamics

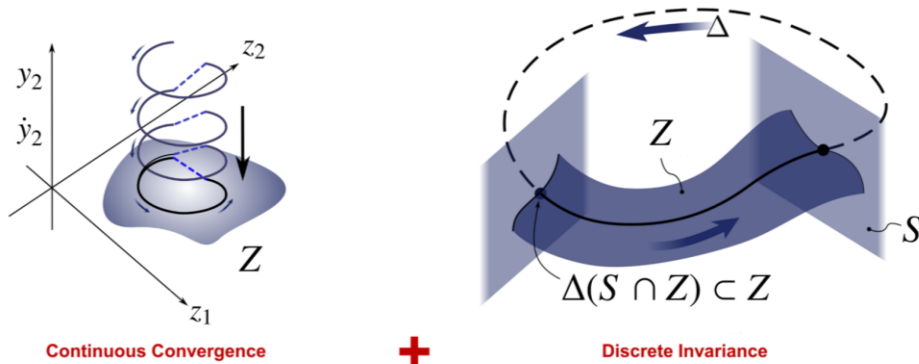
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A J Taylor, F Farshidian, M Hutter, A D Ames
Humanoids 2022



Related Approaches – Dynamic Locomotion

Offline Gait Synthesis using Whole-Body Dynamics

Hybrid Zero Dynamics (HZD)

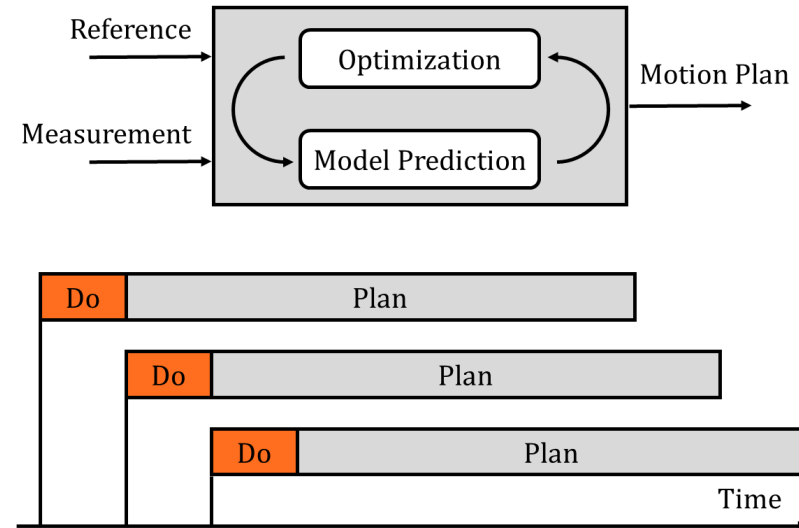


Feedback Control of Dynamic Bipedal Robot Locomotion, Eric R. Westervelt, 2007

- Find periodic trajectory of actuated outputs $y_d(q, \alpha)$ s.t. unactuated DoF exhibit stable periodic behavior

Precomputed Stable Periodic Trajectories

Online Gait Synthesis using MPC



- S-LIP Model, Centroidal Dynamics

A Unified MPC Framework for Whole-Body Dynamic Locomotion and Manipulation, Jean-Pierre Sleiman, 2021

Simplified/Reduced Model Dynamics

Contribution

Whole-Body Nonlinear MPC

Reduced computational cost via HZD Reference & Terminal

Experimental validation on planar biped AMBER-3M

MPC Formulation – Reparametrized Whole-Body Dynamics

Exclude inverse dynamics from MPC prediction

- $x = (q_b, q_j, \dot{q}_b, \dot{q}_j)^T$
- $u = (\lambda_c, \ddot{q}_j)$
- $\dot{x} = (\dot{q}_b, \dot{q}_j, \ddot{q}_b, \ddot{q}_j)^T$

- Recover torques via inverse dynamics

$$\tau = J_c^T F_c - D \ddot{q} - C \dot{q} - G$$

- Dynamics Formulation via Euler Lagrange

$$D(q) \ddot{q} + C(q, \dot{q}) \dot{q} = B \tau + J_c^T(q) \lambda_c$$

$$\begin{pmatrix} D_{bb} & D_{bj} \\ D_{jb} & D_{jj} \end{pmatrix} \begin{pmatrix} \ddot{q}_b \\ \ddot{q}_j \end{pmatrix} + \begin{pmatrix} C_b \\ C_j \end{pmatrix} \dot{q} = \begin{pmatrix} 0 \\ I \end{pmatrix} \tau + \begin{pmatrix} J_{cb} \\ J_{cj} \end{pmatrix} \lambda_c$$

$$\ddot{q}_b = D_{bb}^{-1} (-D_{bj} \ddot{q}_j - C_b \dot{q} + J_{cb} \lambda_c)$$

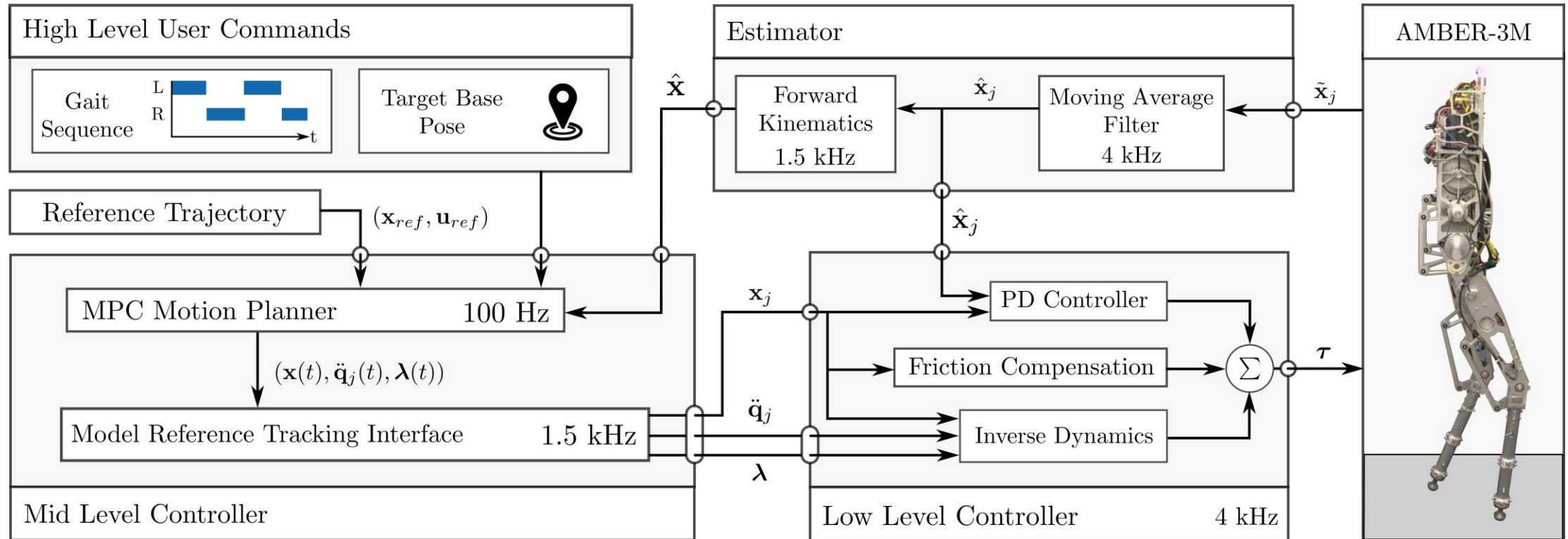


Focus computation on unactuated DoF

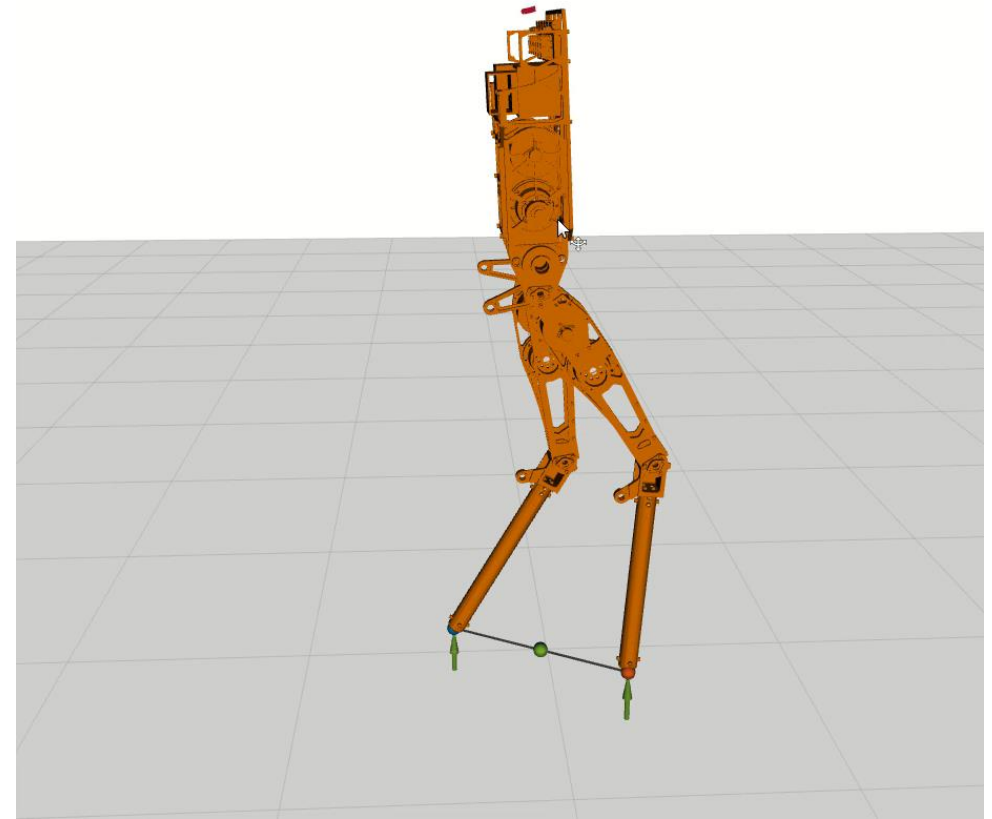
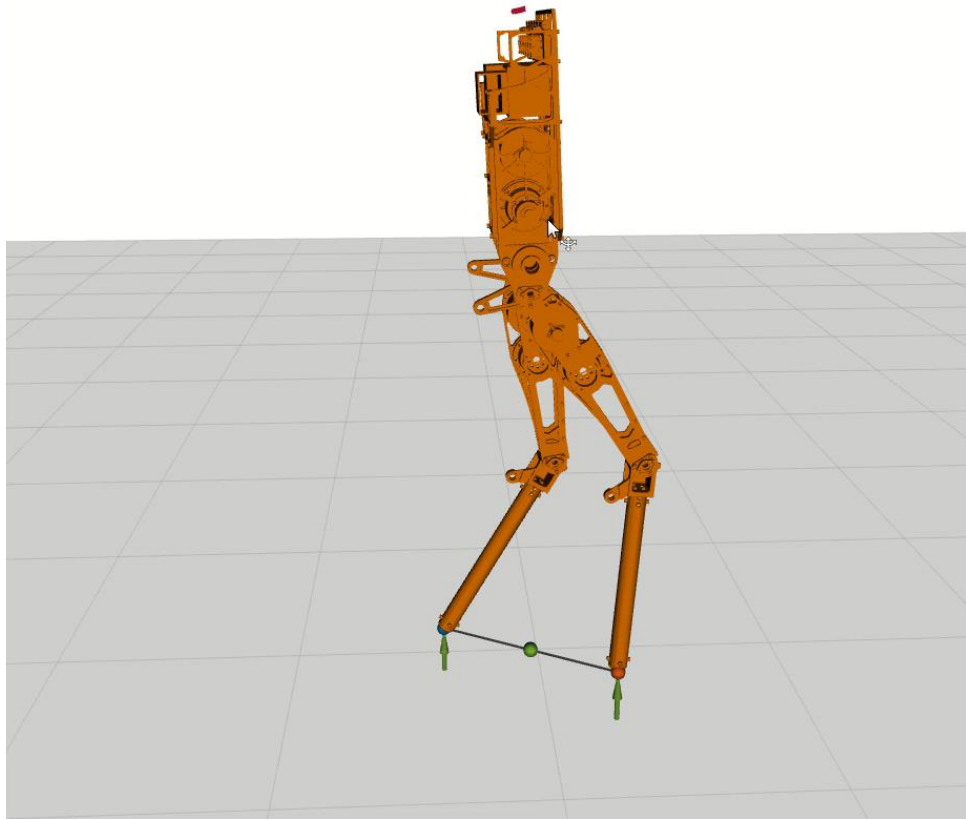


Whole-Body nonlinear dynamics/constraints

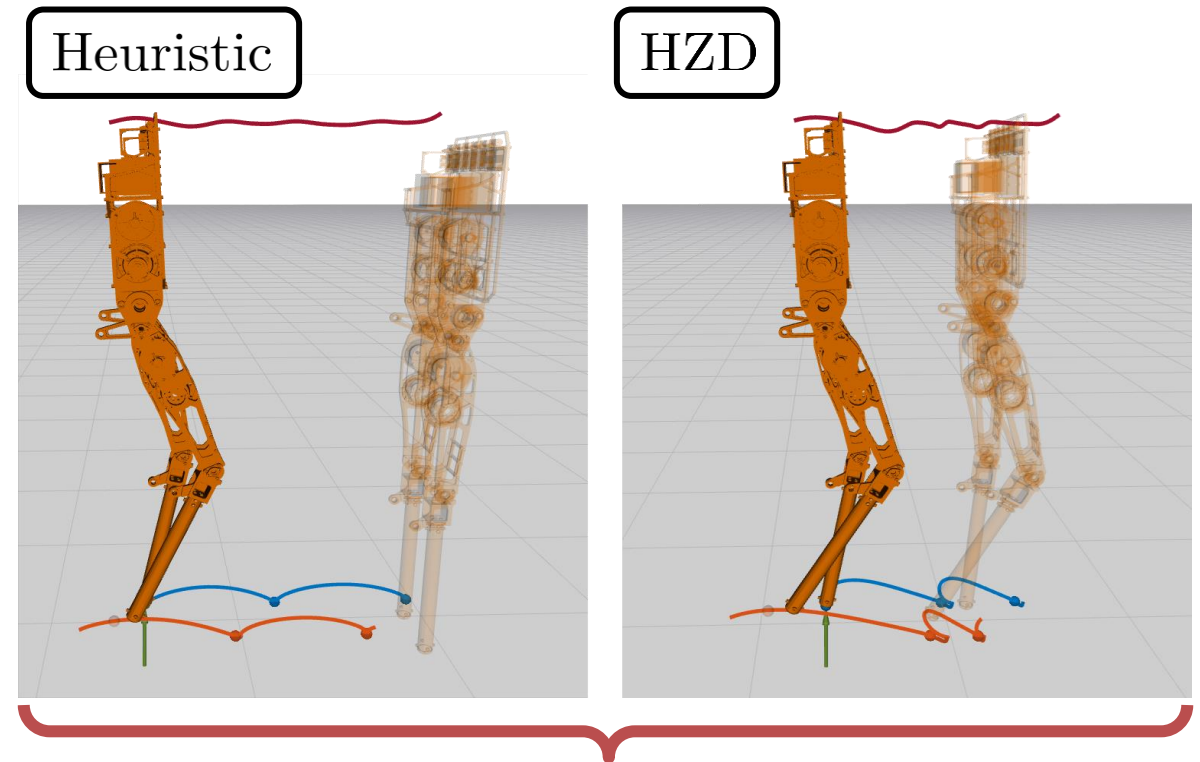
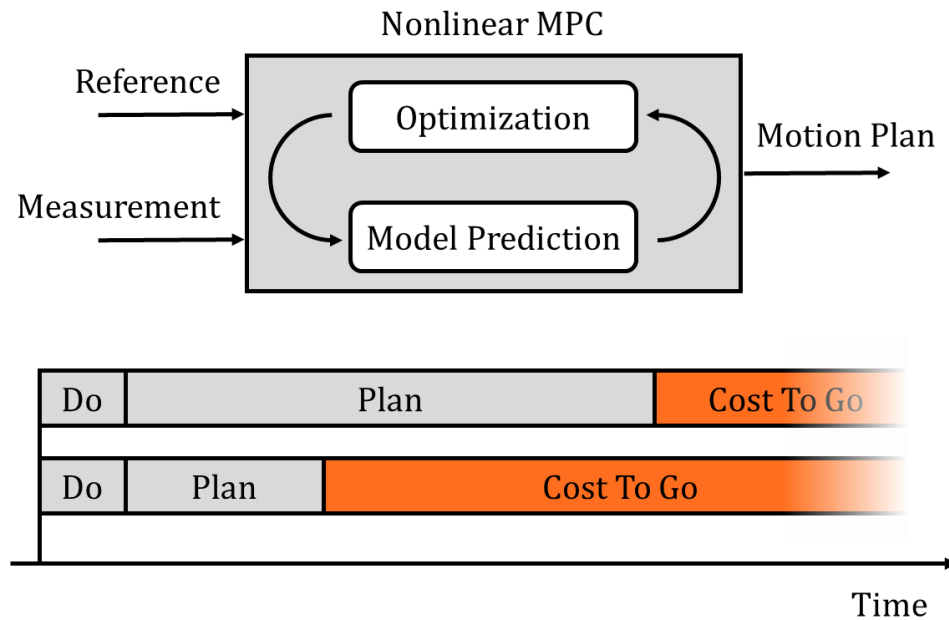
AMBER Implementation – Control Overview



Results - Simulation



Reduce Computational Cost via Horizon Shortening



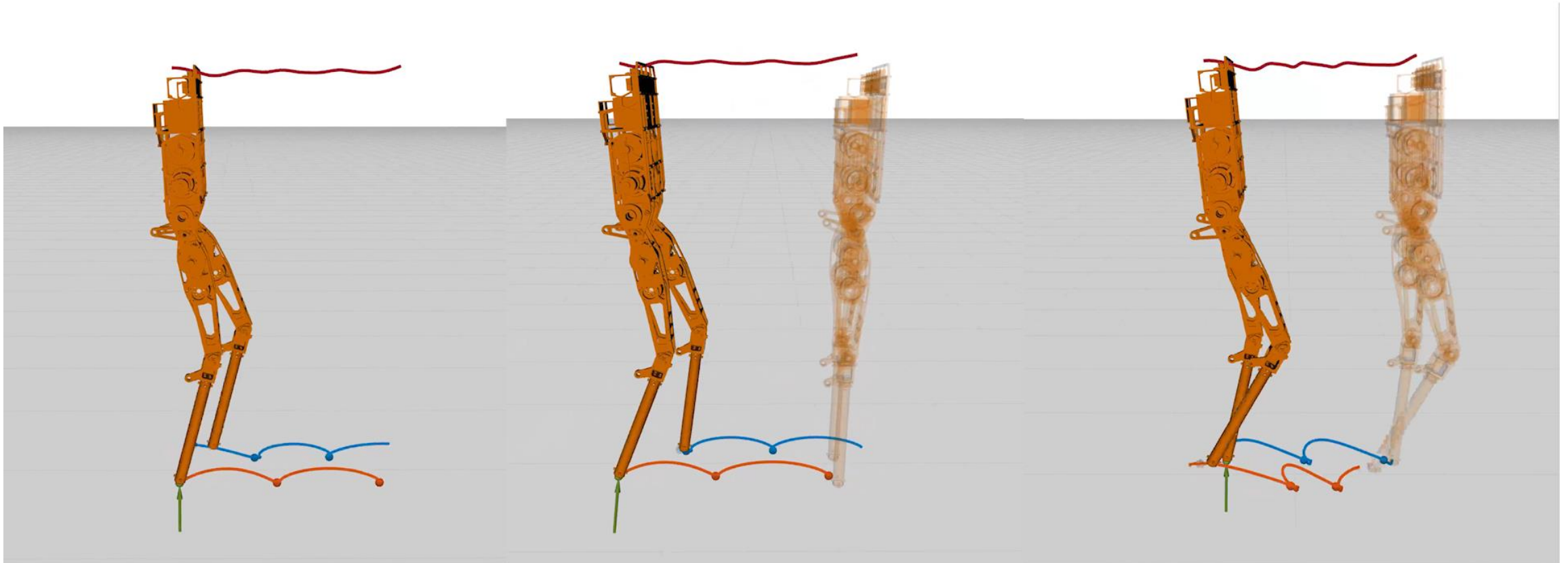
$$\text{minimize}_{\mathbf{u}(\cdot)} \quad \phi(\mathbf{x}(t_H)) + \int_0^{t_H} l(\mathbf{x}(t), \mathbf{u}(t), t) dt$$

MPC Terminal Cost Visualization 2s Horizon

No Terminal

Heuristic Terminal

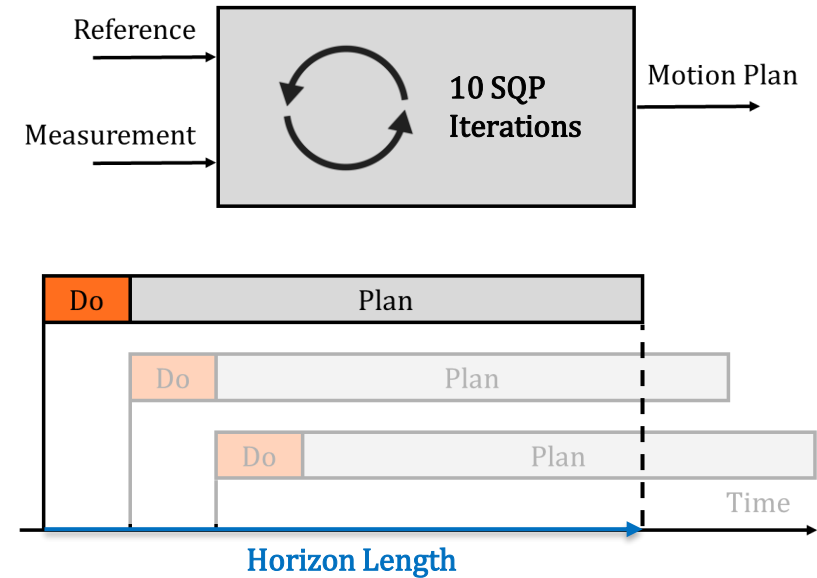
HZD Terminal



Results - Metrics

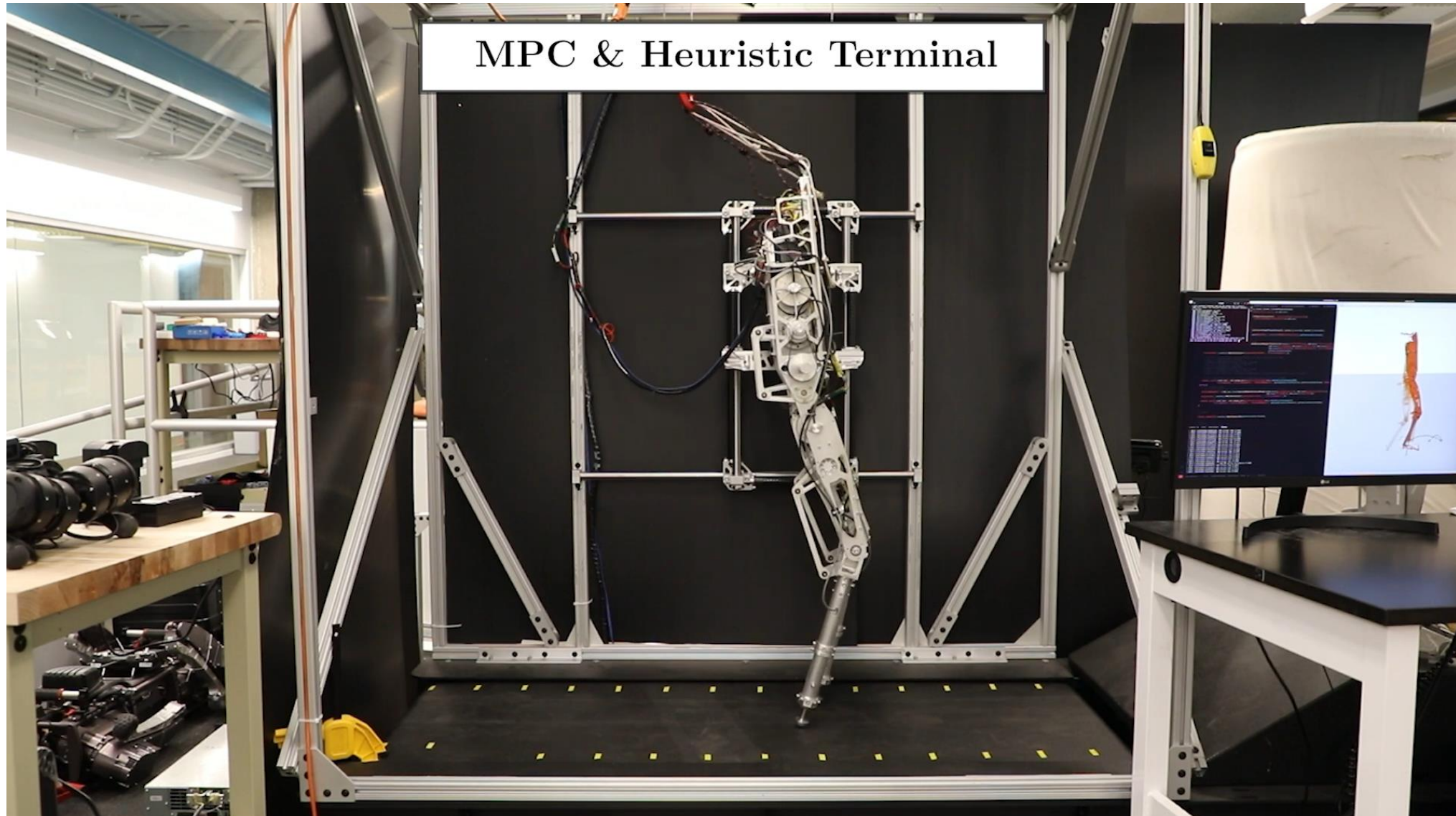
Ryzen 9 5950x at 10 SQP Iterations

Horizon Length [s]	2.0	1.0	0.5	0.2
MPC Frequency [Hz]	270	480	670	850

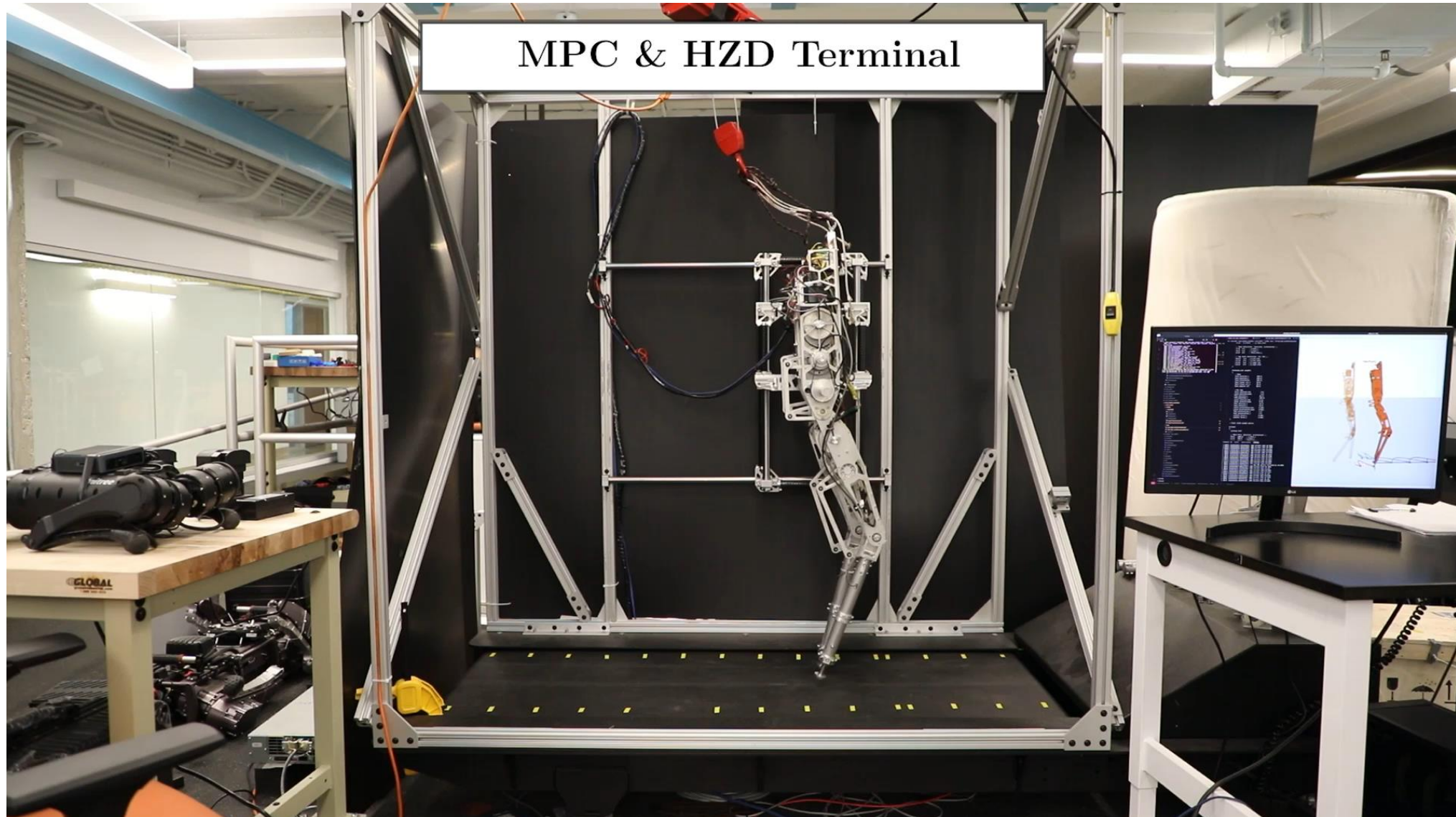


➔ Reduce computational complexity through terminal

Results



Results – MPC & HZD Terminal



Conclusion

Reparametrized Whole-Body NMPC Formulation

Significant Horizon shortening through HZD Terminal

Hardware Demonstration of Whole-Body Online Planning

Thank you for your Attention!



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