#### MPC control of quadruped robot

Master's Degree in Artificial Intelligence and Robotics

Paradiso Emiliano (1940454) Piccione Brian (1889051) Pisapia Vittorio (1918590) Tedeschi Jacopo (1882789)





# Table of Contents Introduction

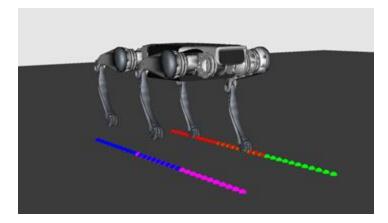
#### ► Introduction

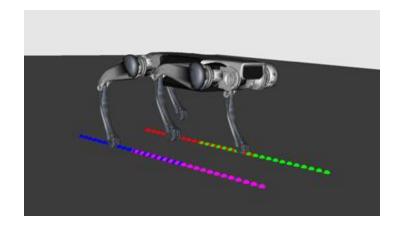
- Footstep planner
- ► MPC
- Controllers
- Results
- Conclusion



## Project overview Introduction

- the project focuses on simulating quadruped locomotion using Model Predictive Control
- the control problem is formulated as a convex quadratic program to compute the optimal contact forces
- feasible contact sequences for stable locomotion is generated by a custom footstep planner
- ground and swing leg controllers apply forces computed by the MPC and execute gait phases
- multiple gait scenarios are used to assess the framework's effectiveness







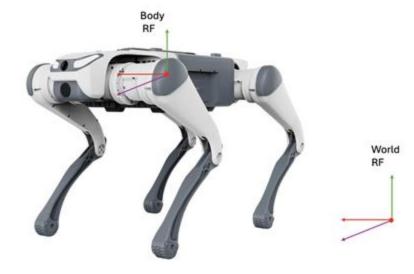
#### **Problem definition**

#### Introduction

- the analysis is carried out using the **Lite 3** quadruped robot from DeepRobotics
- the state of the robot is defined as  $x \in SE(3) \times \mathbb{R}^6$ :

$$\mathbf{x} = (\mathbf{\Theta} \ \mathbf{p} \ \boldsymbol{\omega} \ \dot{\mathbf{p}})^T$$

- **0**: orientation expressed in *ZYX* Euler angles
- *p*: CoM position
- $\omega$ : rigid body angular velocity
- $\dot{p}$ : CoM linear velocity

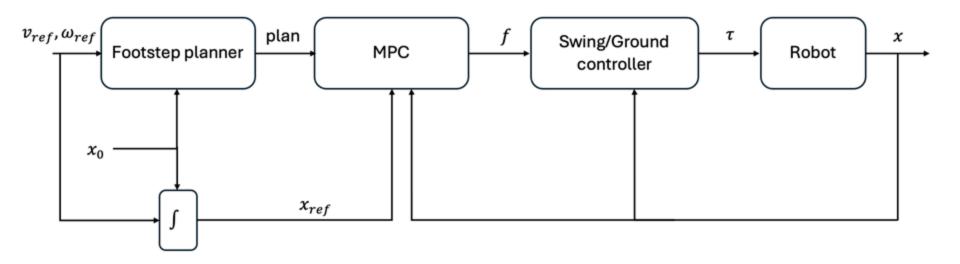


• to enable effective walking, a reference trajectory  $x_{des}$  is properly designed by integrating predefined linear and angular velocities ( $v_{ref}$ ,  $\omega_{ref}$ ), resulting in motion with **constant velocity** 



#### **Block scheme**

Introduction





#### Table of Contents

Footstep planner

- Introduction
- ► Footstep planner
- ► MPC
- Controllers
- Results
- Conclusion



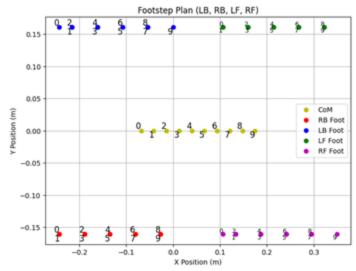
#### **Overview**

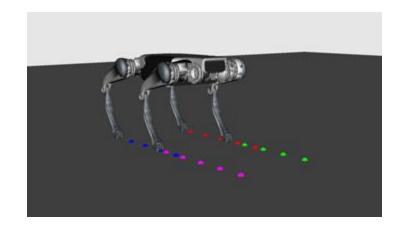
#### Footstep planner

- **input:** initial configuration, linear velocity  $v_{ref}$ , angular velocity  $\omega_{ref}$ , number of steps, double and single support duration
- output: a plan composed by a sequence of pose and the swing/support state for each leg
- **virtual unicycle modelling**: the robot is modelled as a virtual unicycle  $(x, y, \theta)$  placed on the projection on the ground of the robot's center of mass
- **gait generation:** euler integration with T = SS + DS



$$\theta(t + \Delta t) = \theta(t) + \omega_{ref} \cdot \Delta t$$







#### Types of gait

Footstep planner

**Trotting** 

FL Foot					Standing
FR Foot					
HL Foot					Swinging
HR Foot					
	DS	SS			
	S	tep			

**Pronking** 

FL Foot			Standing
FR Foot			
HL Foot			Swinging
HR Foot			
	DS   SS		
	Step		

**Pseudo-gallopping** 

FL Foot				Standing
FR Foot				
HL Foot				Swinging
HR Foot				
	DS   SS	1		
	Step	-1		

**Ambling** 

FL Foot				Standing
FR Foot				
HL Foot				Swinging
HR Foot				
	DS   SS			
	Step			

multi-phase gait such as **galloping** can't be generated

FL Foot						Standing
FR Foot						Swinging
HL Foot						Unfeasible
HR Foot						



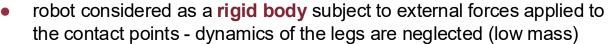
# **Table of Contents**MPC

- Introduction
- Footstep planner
- ► MPC
- Controllers
- Results
- Conclusion



## Robot's dynamics: simplification and model

The **linear system**  $x_{i+1} = A_i x_i + B_i u_i$  has been derived from the following assumptions:





- Considering **small roll and pitch angles** and avoid near-vertical posture
- the state of the robot is then extended to include the gravitational term
- the problem is formulated as a discrete-time linear system.
- $A_i$  and  $B_i$  should be computed using the future predicted position of the feet obtained through the dynamics at instant t + i; desired position of the feet are used instead

$$egin{bmatrix} \dot{oldsymbol{\phi}} \ \dot{oldsymbol{\phi}} \ \dot{oldsymbol{\psi}} \end{bmatrix} pprox oldsymbol{R}_z(\psi) oldsymbol{\omega}$$

World

$$\frac{d}{dt} \begin{bmatrix} \mathbf{0} \\ \mathbf{p} \\ \boldsymbol{\omega} \\ \dot{\mathbf{p}} \\ g \end{bmatrix} = \begin{bmatrix} \mathbf{0}_3 & \mathbf{0}_3 & \mathbf{R}_z(\psi) & \mathbf{0}_3 & \mathbf{0} \\ \mathbf{0}_3 & \mathbf{0}_3 & \mathbf{0}_3 & \mathbf{1}_3 & \mathbf{0} \\ \mathbf{0}_3 & \mathbf{0}_3 & \mathbf{0}_3 & \mathbf{0}_3 & \mathbf{0} \\ \mathbf{0}_3 & \mathbf{0}_3 & \mathbf{0}_3 & \mathbf{0}_3 & \mathbf{0} \\ \mathbf{0}_{1\times3} & \mathbf{0}_{1\times3} & \mathbf{0}_{1\times3} & \mathbf{0}_{1\times3} & \mathbf{1} \end{bmatrix} \begin{bmatrix} \mathbf{0} \\ \mathbf{p} \\ \boldsymbol{\omega} \\ \dot{\mathbf{p}} \\ g \end{bmatrix} + \begin{bmatrix} \mathbf{0}_3 & \dots & \mathbf{0}_3 \\ \mathbf{0}_3 & \dots & \mathbf{0}_3 \\ \mathbf{0}_3 & \dots & \mathbf{0}_{1} \\ \mathbf{1}_3/m & \dots & \mathbf{1}_3/m \\ \mathbf{0}_{1\times3} & \dots & \mathbf{0}_{1\times3} \end{bmatrix} \begin{bmatrix} f_1 \\ \vdots \\ f_n \end{bmatrix}$$



## **Problem definition, cost function and constraints**

 simplifications and linearization of the robot dynamic's leads to the following problems, formulated as multiple-shooting, convex optimization problem:

$$\min_{\mathbf{x}, \mathbf{u}} \sum_{i=0}^{N-1} \|\mathbf{x}_{i+1} - \mathbf{x}_{i+1, ref}\|_{\mathbf{Q}_{i}}^{2} + \|\mathbf{u}_{i}\|_{\mathbf{R}_{i}}^{2}$$
subject to
$$\mathbf{x}_{i+1} = \mathbf{A}_{i} \mathbf{x}_{i} + \mathbf{B}_{i} \mathbf{u}_{i}, \qquad i = 0, ..., N-1$$

$$\underline{\mathbf{c}_{i}} \leq \mathbf{C}_{i} \mathbf{u}_{i} \leq \overline{\mathbf{c}_{i}}, \qquad i = 0, ..., N-1$$

$$\mathbf{D}_{i} \mathbf{u}_{i} = 0 \qquad i = 0, ..., N-1$$

- $x_i$  represents the system's state at the i-th step of the horizon
- $u_i$  is the control input at step i
- $A_i$  and  $B_i$  represent the discrete time system dynamics
- $c_i$  and  $c_i$  represent inequality constraints (bound on the fz component, contact forces within friction cone)
- D<sub>i</sub> selects forces corresponding to swinging foot (whose ground forces should be zero)
- $Q_i$  and  $R_i$  are diagonal positive semidefinite matrices of weights



# Table of Contents Controllers

- Introduction
- Footstep planner
- ► MPC
- ► Controllers
- Results
- Conclusion



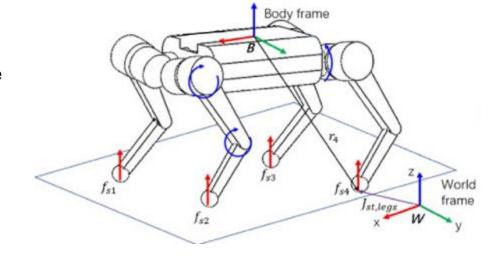
#### **Ground controller**

Controllers

• given a leg **planned to be in contact** at a specific *timestep* (*i*), is possible to map the computed **reaction force** (by the MPC) to the torque the joints of the leg should apply by simply computing:

$$\tau_i = J_i^T R_i^T f_i$$

- $J_i^T$ : i-th leg Jacobian
- $R_i^T$ : rotation from body frame to world frame
- f<sub>i</sub>: output of the MPC relative to the i-th leg





#### **Swing trajectory definition**

Controllers

- each foot of the robot should track a predefined trajectory for the movement during the swing phase
- using a **cubic polynomial** defined in the plane x-y allows continuous velocity and acceleration profile and zero derivatives at the boundary:

$$p(t) = p_i + (p_f - p_i) \left(-2\left(\frac{t}{T}\right)^3 + 3\left(\frac{t}{T}\right)^2\right), \quad t \in [0, T]$$

where *T* is the duration of the single support time

 using a quartic polynomial with a bell-shaped profile for the z component, allowing the liftoff and landing on 0 vertical velocity and acceleration:

$$z(t) = 16\frac{h}{T^4}t^4 - 32\frac{h}{T^3}t^3 + 16\frac{h}{T^2}t^2, \quad t \in [0, T]$$

where the value of the *h* is an *hyperparameter* describing the maximum height of the step



## Swing leg controller Controllers

- simple control strategy using feedback + feedforward to track the reference trajectory
- commanding the torque for the swinging leg as:

$$\tau_i = J_i^T [K_p(p_{i,ref} - p_i) + K_d(v_{i,ref} - v_i)] + \tau_{i,ff}$$

where the  $\tau_{i\_ff}$  is the feedforward term computed taking account of the Dynamical term of the leg:

$$\tau_{i\_ff} = J_i^T M_i (a_{i,ref} - \dot{J}_i \dot{q}_i) + C_i \dot{q}_i + G_i$$

- $J_i^T$ : Jacobian of the i-th leg
- $\dot{q}_i$ : joint velocity vector
- $M_i$ : operational space inertia matrix (apparent mass along the i-th direction)
- C<sub>i</sub>: Coriolis term for the i-th leg
- G<sub>i</sub>: gravitational term for the i-th leg



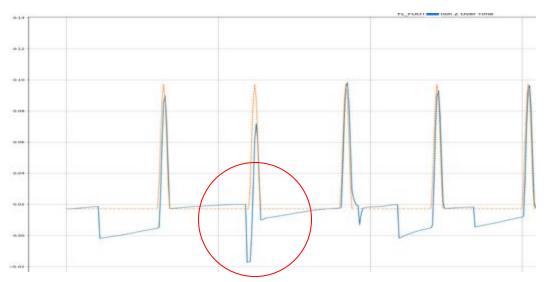
#### **Table of Contents**

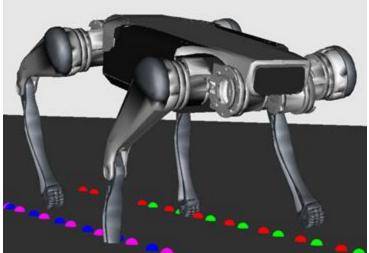
- Introduction
- Footstep planner
- ► MPC
- Controllers
- ► Results
- Conclusion



## Simulation environment and MPC frequencies Results

- The proposed results are obtained using the python based DARTPy simulation environment
- MPC solver achieved solve frequencies in the range of 40-60 Hz with horizon length of 60
- By decreasing the horizon length to 30, tracking performance were slightly worse but solve frequencies exceed 100 Hz, making real-time, on-board implementation feasible (simulation time step of 0.01 seconds)
- One major problem was compenetration between feet and ground; it was solved by enlarging collision spheres of the feet and by adding small mass and inertia to them

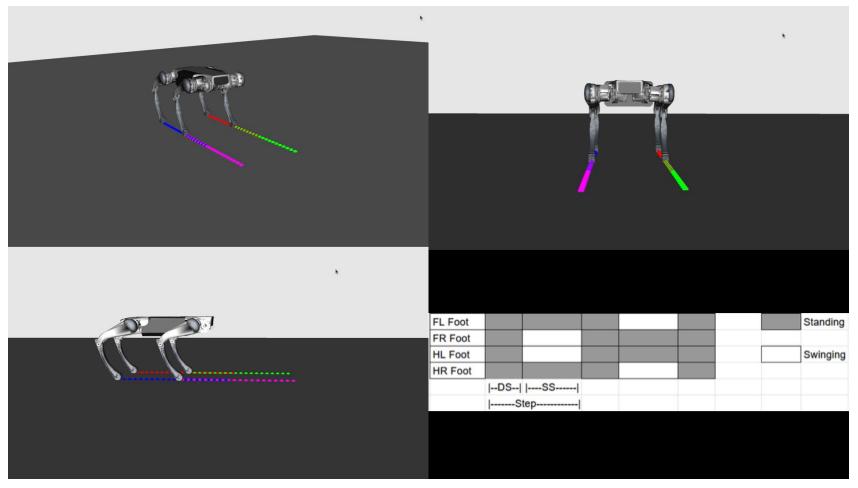






#### **Trotting gait - Video**

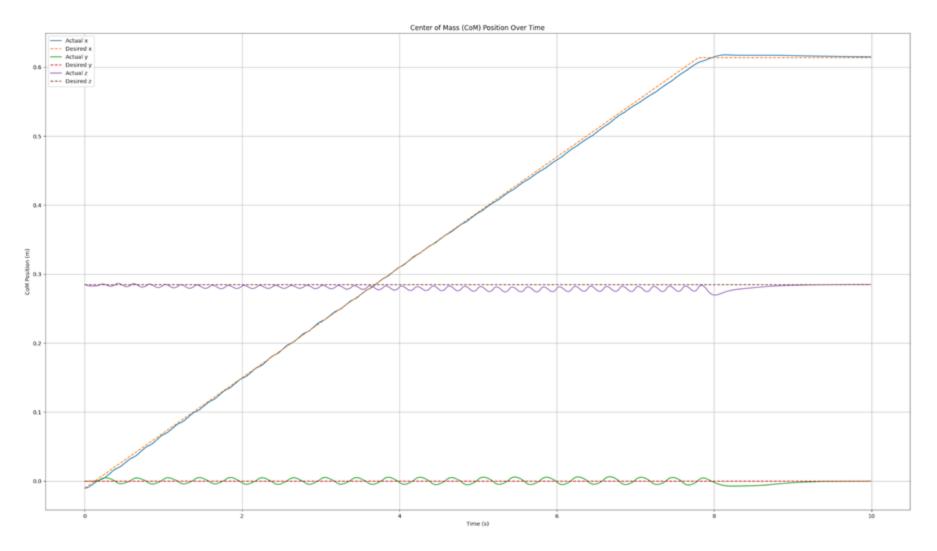
Results



 $ss\_duration = 0.1s$ ,  $ds\_duration = 0.1s$ ,  $v\_com\_ref\_x = 0.08m/s$ 

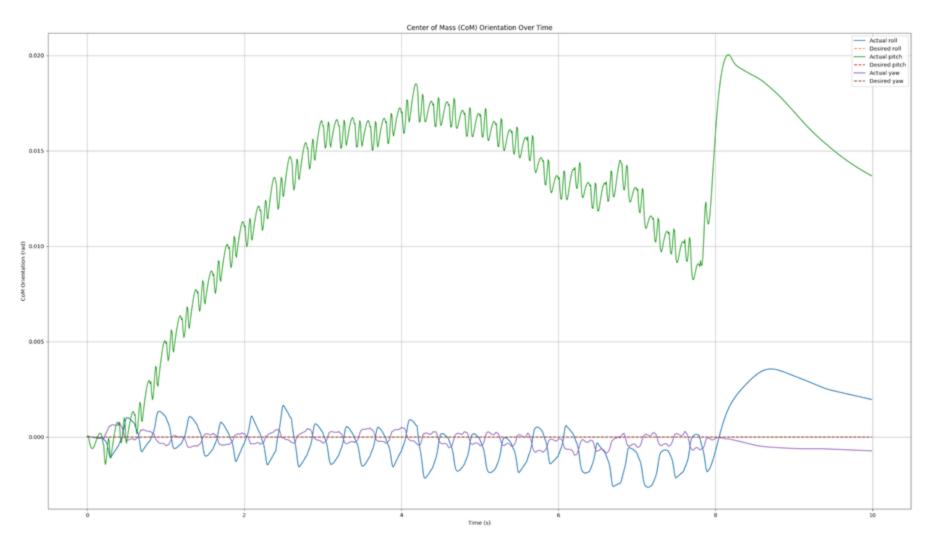


### **Trotting gait - Position of CoM**



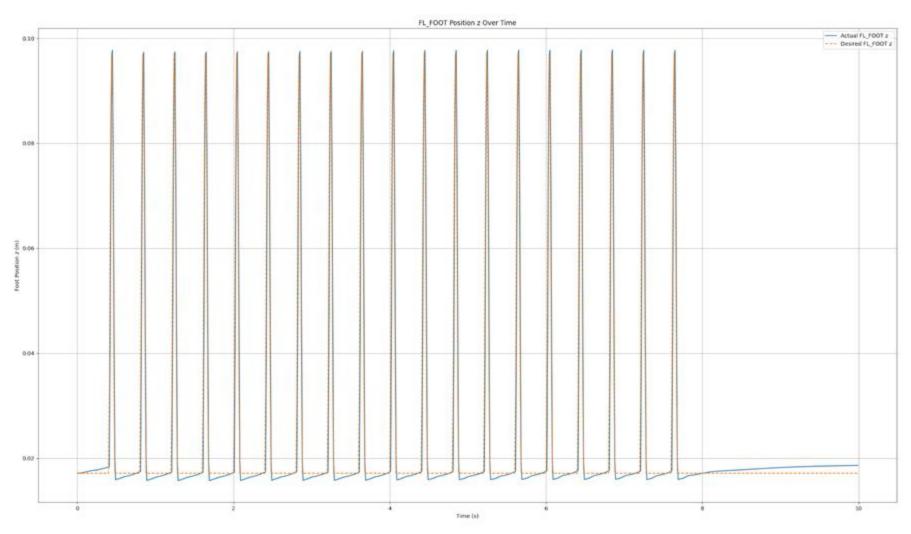


### **Trotting gait - Orientation of CoM**



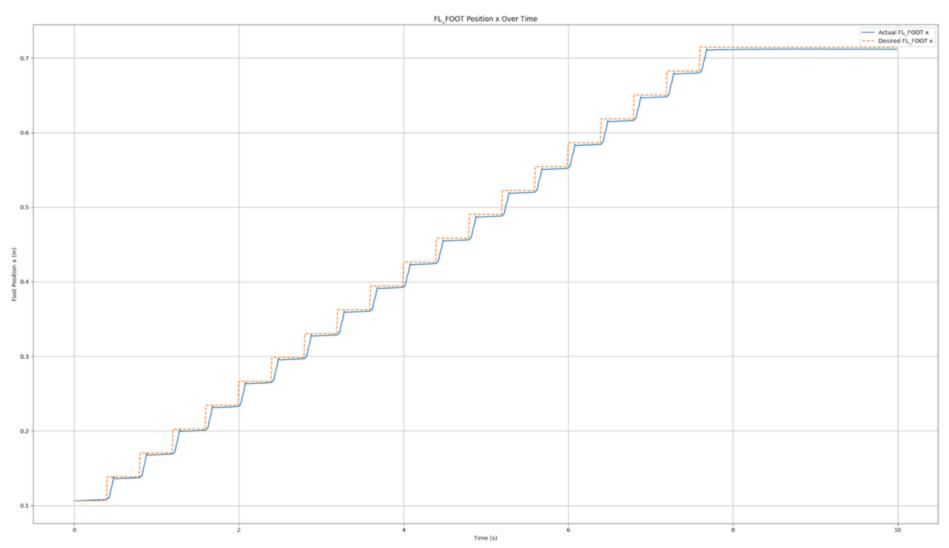


#### **Trotting gait - FL foot z component**



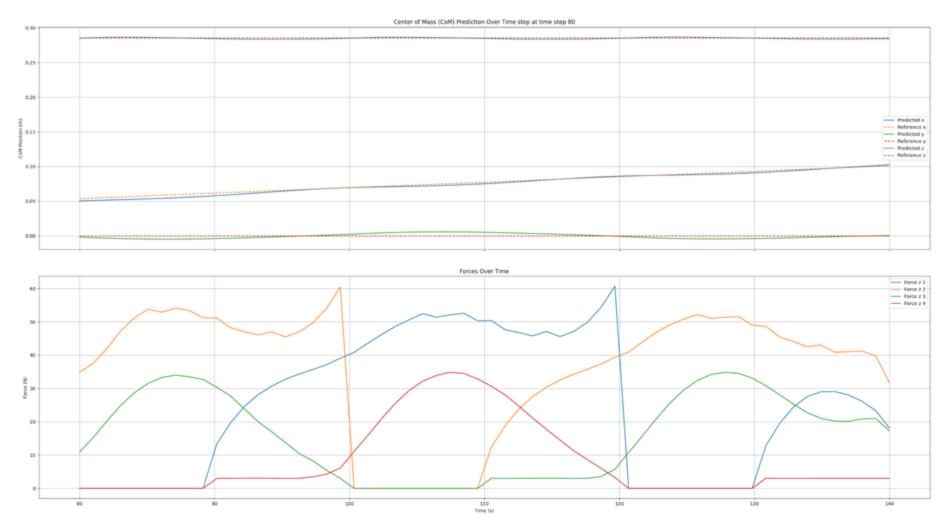


### **Trotting gait - FL foot x component**



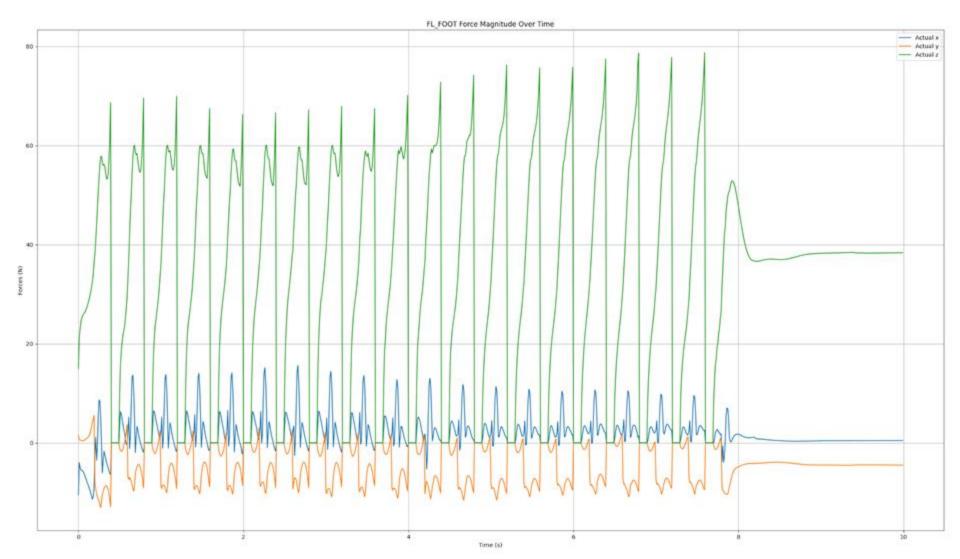


# **Trotting gait - MPC prediction at timestep 80**Results



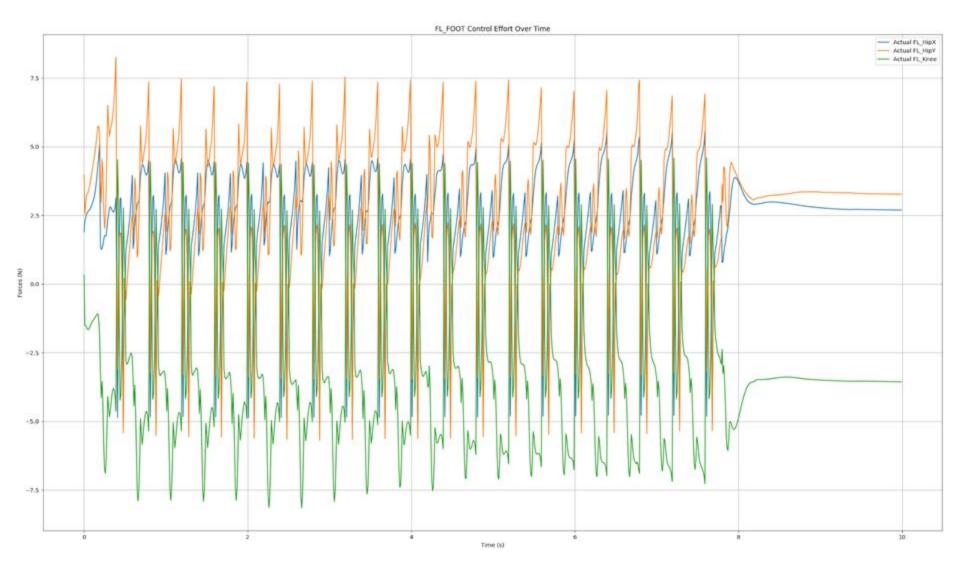


### **Trotting gait - FL foot forces**





### **Trotting gait - FL leg control effort**





#### Pesudo-gallopping gait - Video

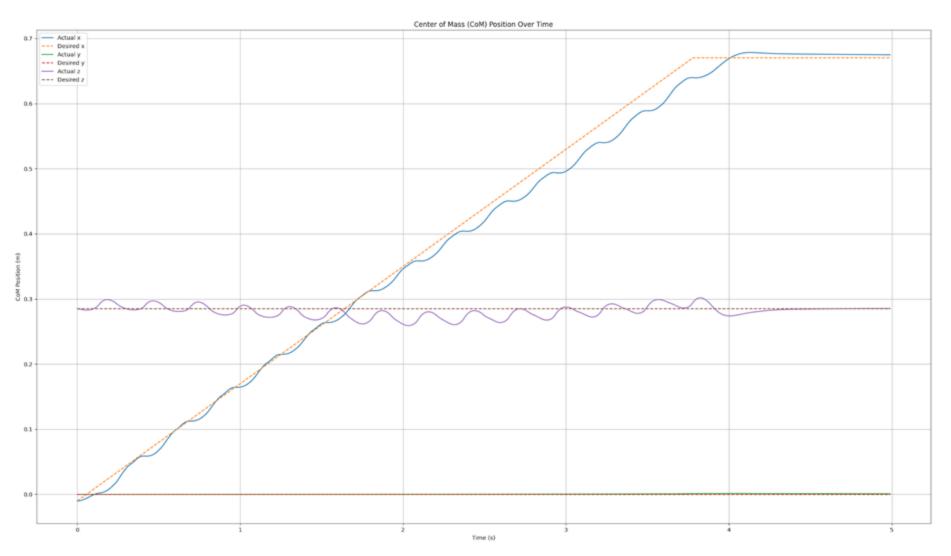
Results



 $ss\_duration = 0.1s$ ,  $ds\_duration = 0.04s$ ,  $v\_com\_ref\_x = 0.18m/s$ 

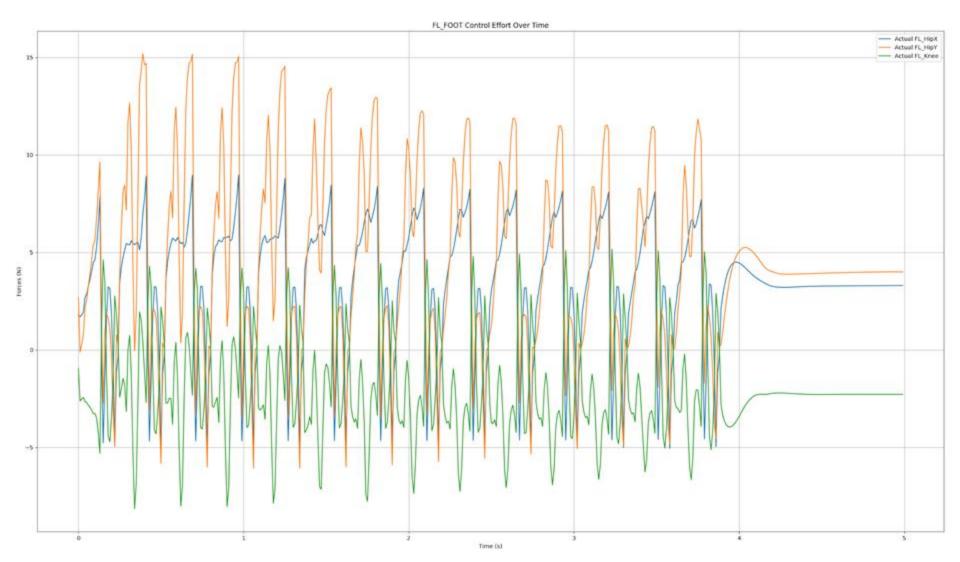


### **Pesudo-gallopping gait - Position of CoM**





#### Pesudo-gallopping gait - FL leg control effort





#### **Ambling gait - Video**

Results



 $ss\_duration = 0.1s$ ,  $ds\_duration = 0.1s$ ,  $v\_com\_ref\_x = 0.08m/s$ 



#### **Pronking gait - Video**

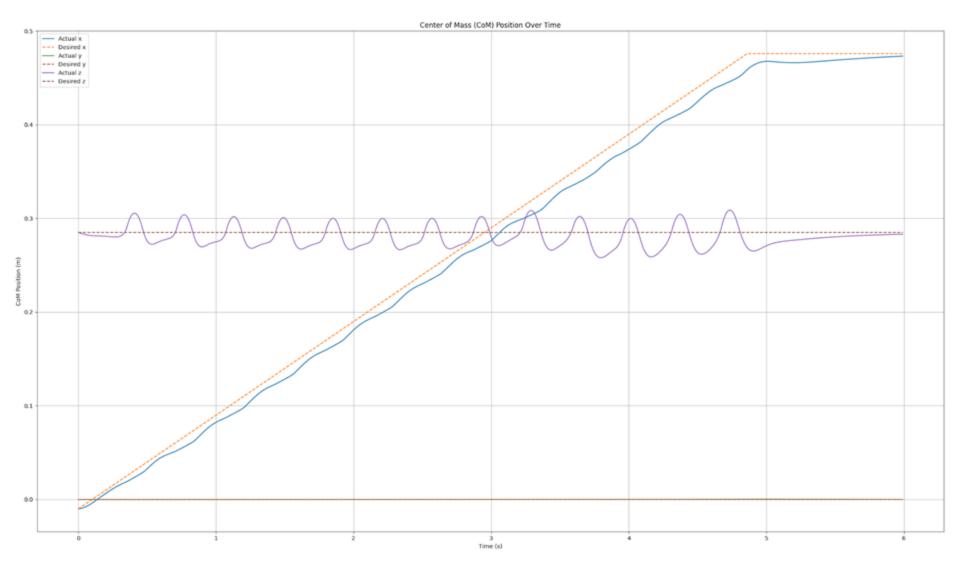
Results



 $ss\_duration = 0.1s$ ,  $ds\_duration = 0.08s$ ,  $v\_com\_ref\_x = 0.1m/s$ 

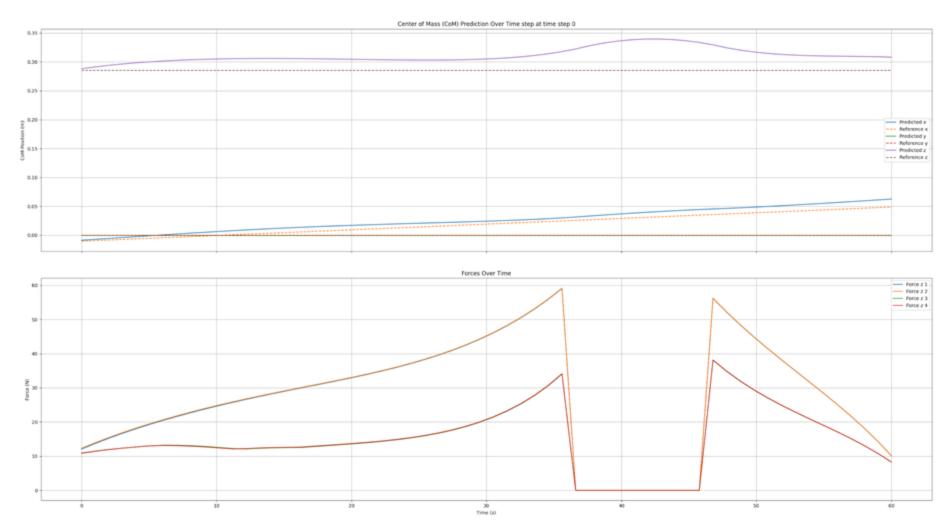


#### **Pronking gait - Position of CoM**



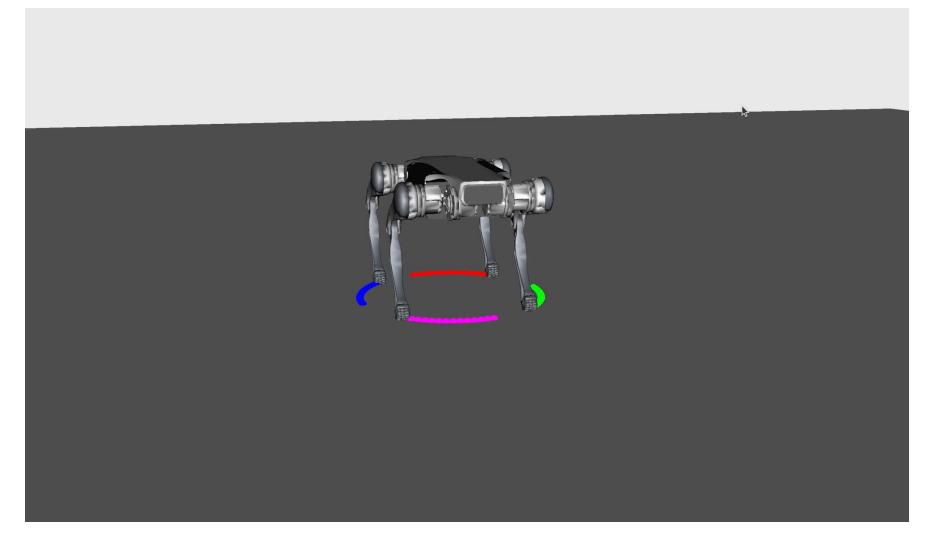


### Pronking gait - MPC prediction at timestep 0





# Spinning in place Results





# Side-trotting Results



## **Table of Contents**Conclusion

- Introduction
- Footstep planner
- ► MPC
- Controllers
- Results
- ► Conclusion



- complete locomotion control pipeline: implemented and validated Model Predictive Control for the Lite 3 quadruped robot
- **simulation handling of the problem**: effectively addressed the issue of ground penetration
- integrated key components: footstep planner, trajectory generator, ground and swing leg controllers
- MPC convergence: demonstrated successful convergence in simulation, with valid ground reaction forces enabling trajectory tracking.
- smooth and stable gaits: achieved reasonably smooth and stable locomotion with different gaits



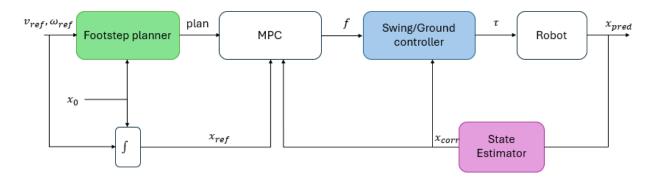
#### **Future works**

Conclusion



**Solid foundation for future work**: framework sets the stage for real-world deployment and extensions to complex gaits or terrains. Some examples are:

- multi-phase gait planner for able the robot to handle different and more complex gaits
- on-line planning for real-time control
- contact detection algorithm for early or late contacts
- state estimator for real-time application





### Thanks for your attention!



#### References

- [1] J. Di Carlo, P. M. Wensing, B. Katz, G. Bledt, and S. Kim, "Dynamic Locomotion in the MIT Cheetah 3 Through Convex Model-Predictive Control", 2018 IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS) Madrid, Spain, October 1-5, 2018
- [2] F. Stark, J. Middelberg, D. Mronga, S. Vyas, F. Kirchner, "Benchmarking Different QP Formulations and Solvers for Dynamic Quadrupedal Walking", arXiv:2502.01329v1 [cs.RO] 3 Feb 2025
- [3] Yijie Zhu, "Lite 3 urdf", <a href="https://github.com/TopHillRobotics/quadruped-robot/blob/mpc-wbc/quadruped/config/lite3/lite3\_robot.yaml">https://github.com/TopHillRobotics/quadruped-robot/blob/mpc-wbc/quadruped/config/lite3/lite3\_robot.yaml</a>