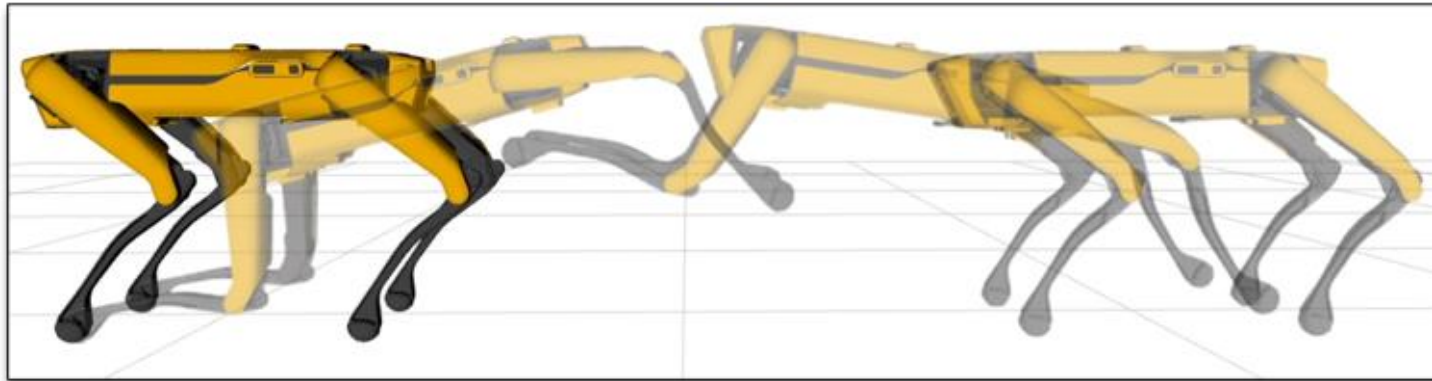


Horizon:

A Trajectory Optimization Framework for Robotic Systems



ISTITUTO ITALIANO
DI TECNOLOGIA
HUMANOIDS AND HUMAN
CENTERED MECHATRONICS

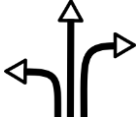
Features



Robotics focus: built-in methods and robotics-oriented utilities



User-friendly interfaces: simplify the formulation process, set up an optimization with minimal effort



Comprehensive pipeline: all the modules to generate a complete robot motion only requiring standard inputs such as XML files



Versatility: generic enough to include all the necessary tools to prototype:

- offline dynamic motion
- receding horizon walking gait
- co-design a robot structure



- **Open-source:** based on open-source packages, and freely available itself

NLP formulation

$$\min_{\mathbf{x}, \mathbf{u}} \int_{t_0}^{t_f} \ell(\mathbf{x}, \mathbf{u}; \mathbf{p}, t) dt + \ell_f(\mathbf{x}_f; \mathbf{p}, t_f) \longrightarrow \begin{array}{l} \text{objective function} \\ \text{to minimize} \end{array}$$

$$\text{subject to:} \quad \dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, \mathbf{u}; \mathbf{p}, t) \longrightarrow \begin{array}{l} \text{dynamics} \\ \text{of the system} \end{array}$$

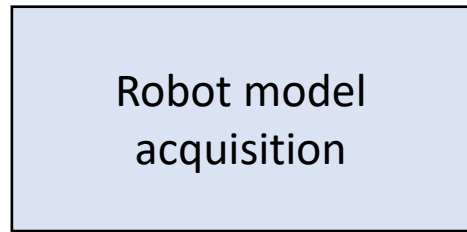
$$\phi^{\min} \leq \phi(\mathbf{x}, \mathbf{u}; \mathbf{p}, t) \leq \phi^{\max} \longrightarrow \begin{array}{l} \text{desired} \\ \text{constraints} \end{array}$$



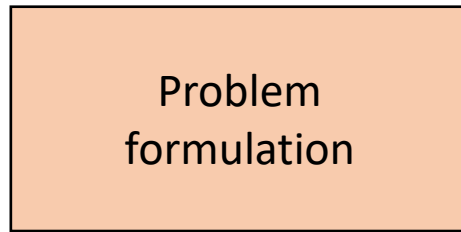
NLP problem set up using the **symbolic framework** provided by CasADi:

- cost and constrain functions are defined using symbolic expressions
- state-of-the-art implementation of algorithmic differentiation (AD)

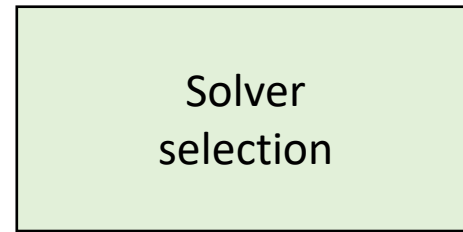
Modules



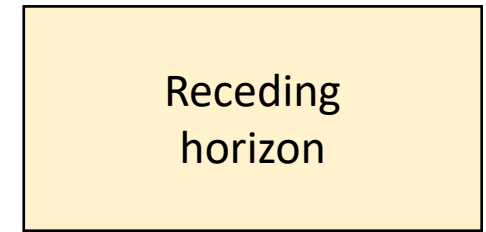
acquires robot model
parsing the URDF



concise syntax to specify
costs and constraints,
and distribute them over
horizon



different solvers
available to meet different
requirements

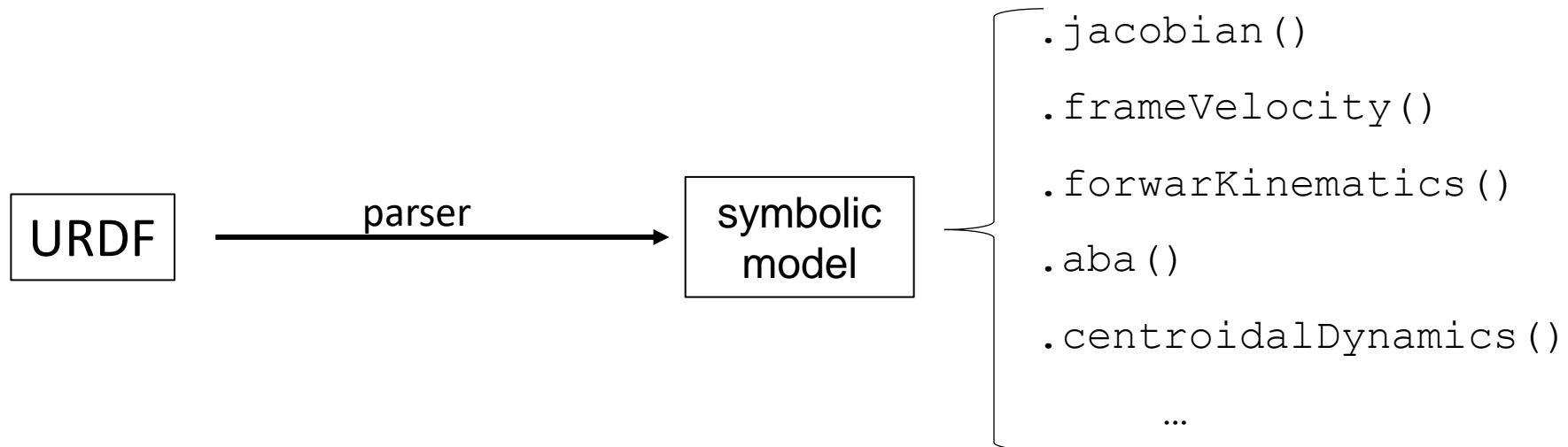


tools for cheaply relocating
costs and constraints
to different nodes

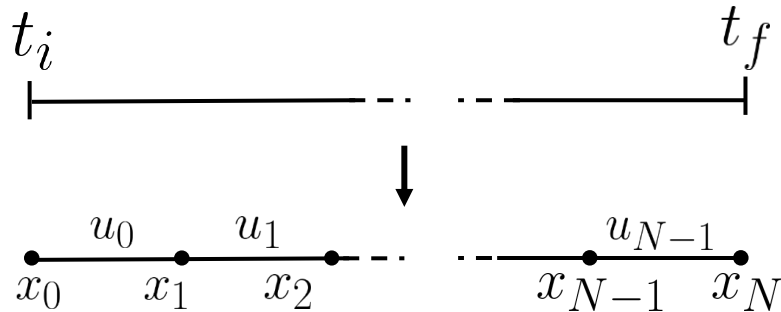
Robot model acquisition



- Parse the URDF file generating a ready-to-use model in Horizon
- From XML file to *symbolic description* compatible with the framework
- Pinocchio library for **robot kinematics and dynamics**

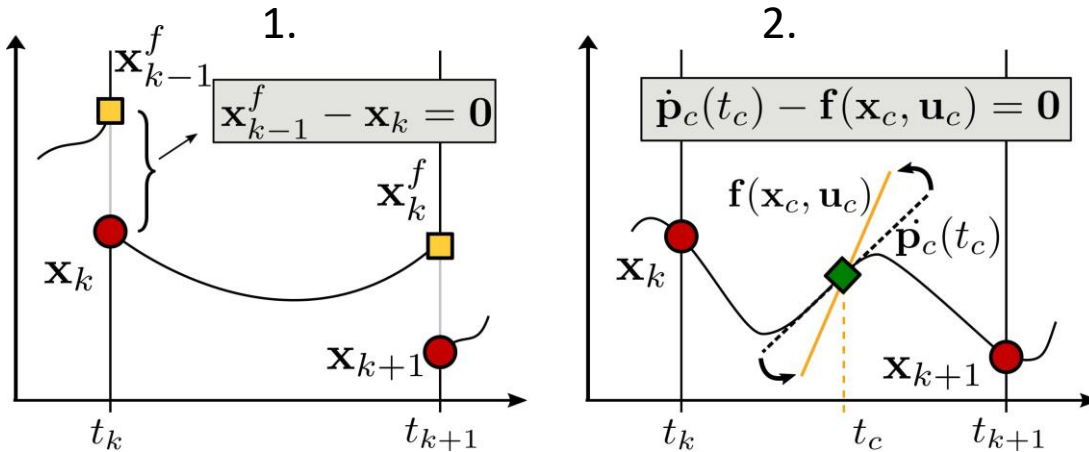


Problem formulation



- **Duration and discretization** of time horizon

```
prb = Problem(nodes=50)
prb.setDt(dt)
```

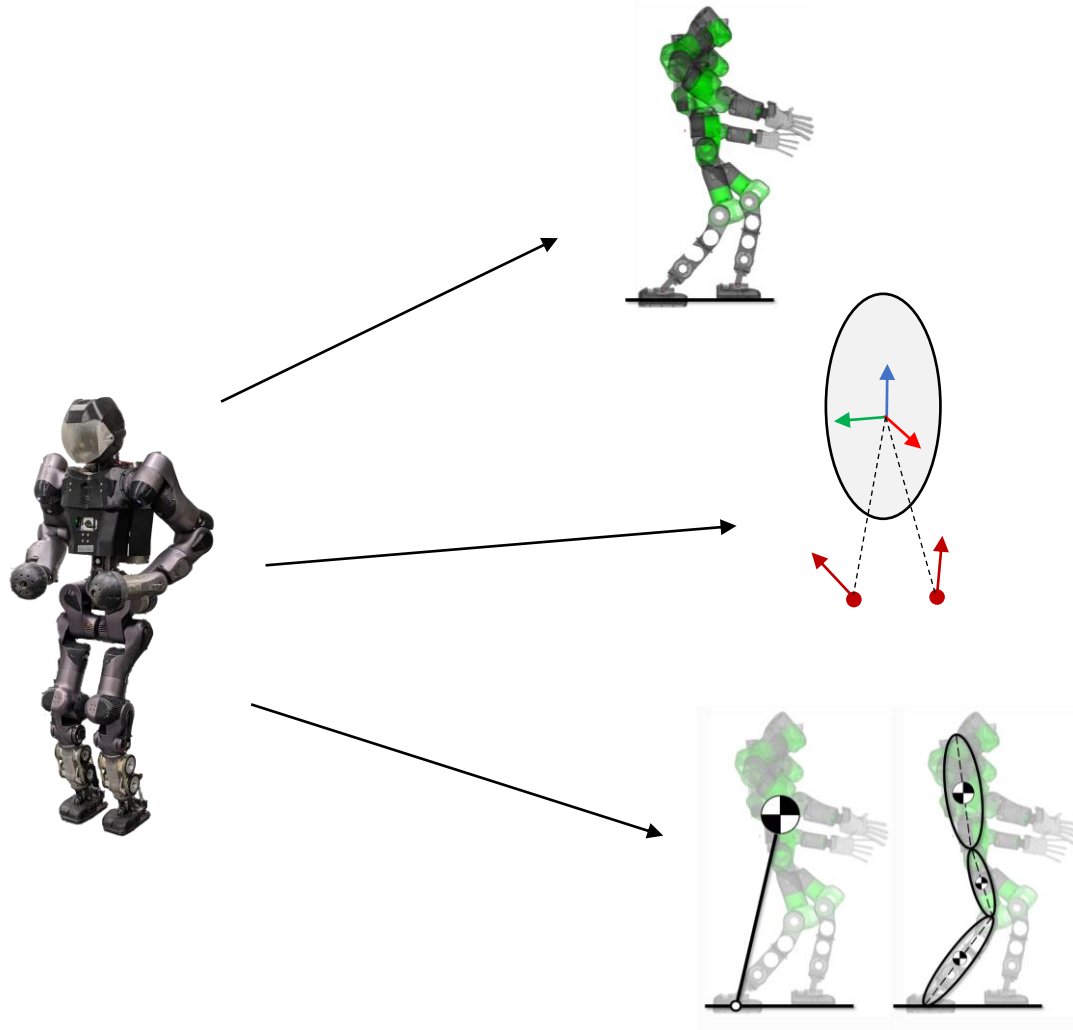


- **Problem transcription:**

1. Direct Multiple Shooting
2. Direct Collocation

```
Transcriptor.make("multiple_shooting", prb)
```

Problem formulation



- **System dynamics:**

1. Built-in models:

- *Full body dynamics*
- *Single Rigid Body Dynamics*
- *Centroidal Dynamics*

2. Custom implementation:

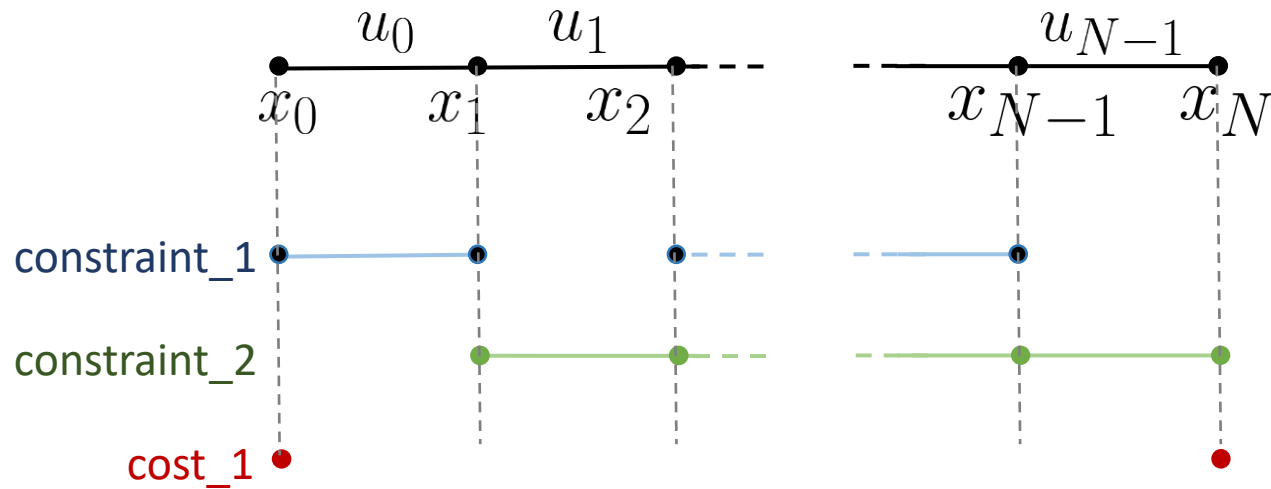
as differential-algebraic system of equation (DAE)

```
x = prb.createStateVariable("x", dim)
```

```
u = prb.createInputVariable("u", dim)
```

```
prb.setDynamics(x_dot)
```

Problem formulation



- **Functions and bounds definition:**

- *constraints* and *costs* defined on desired nodes
- *bounds* for variables and constraints

```
prb.createCost("qddot",
              cs.sumsqr(u),
              nodes=[1, 2, 15])
```

```
prb.createConstraint("frict_cones", f_c)
```

```
x.setBounds(ub, lb, nodes)
```


Solvers

Robot model
acquisition

Problem
formulation

Solver
selection

Receding
horizon

Solver selection depending on the user's requirements.

Custom implementations:

- Gauss-Newton Sequential Quadratic Programming (**GN-SQP**)
- multiple-shooting Iterative Linear-Quadratic Regulator (**ILQR**)

Fast, useful for
receding horizon
formulation

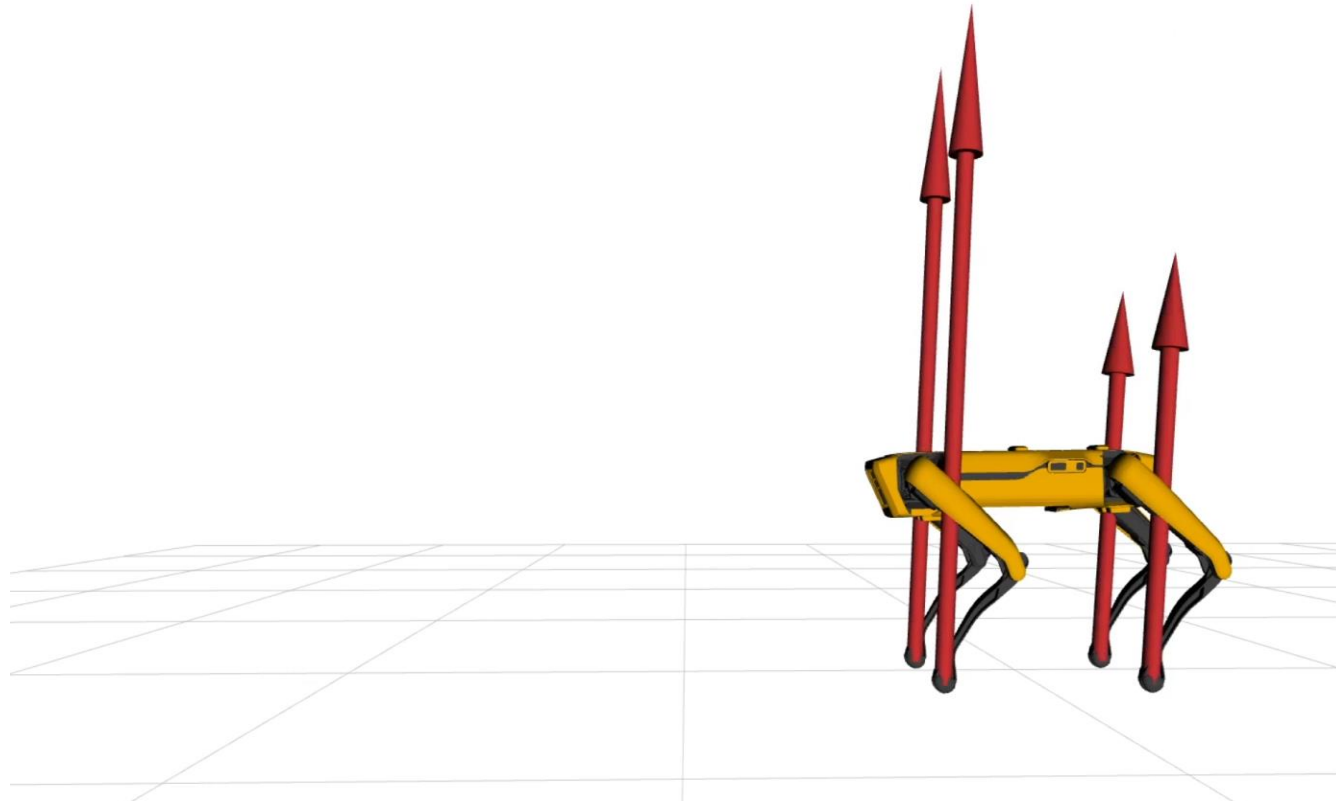
Off-the-shelf solvers available through CasADi interface:

- **IPOPT**/BONMIN,
- BlockSQP
- WORHP
- KNITRO and
- SNOPT

Large-scale nonlinear
optimization, interior-
point methods

Robot motion design

- Optimized trajectory: combined result of bounds, constraints and cost functions
- The desired behavior can be achieved by tuning the formulation of the problem
- High performance index corresponds to the closeness of the trajectory to the desired behavior



Robot model
acquisition

Problem
formulation

Solver
selection

Receding
horizon

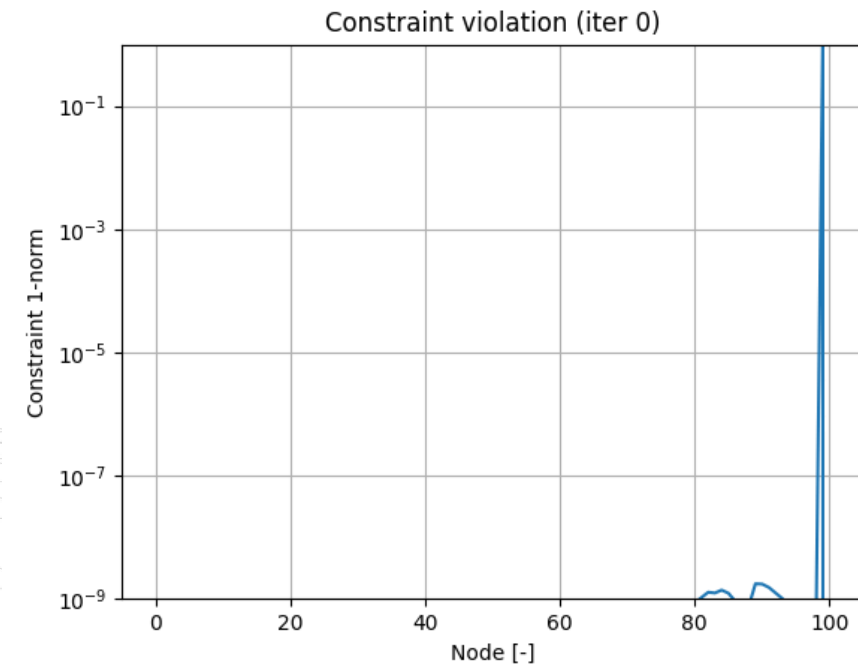
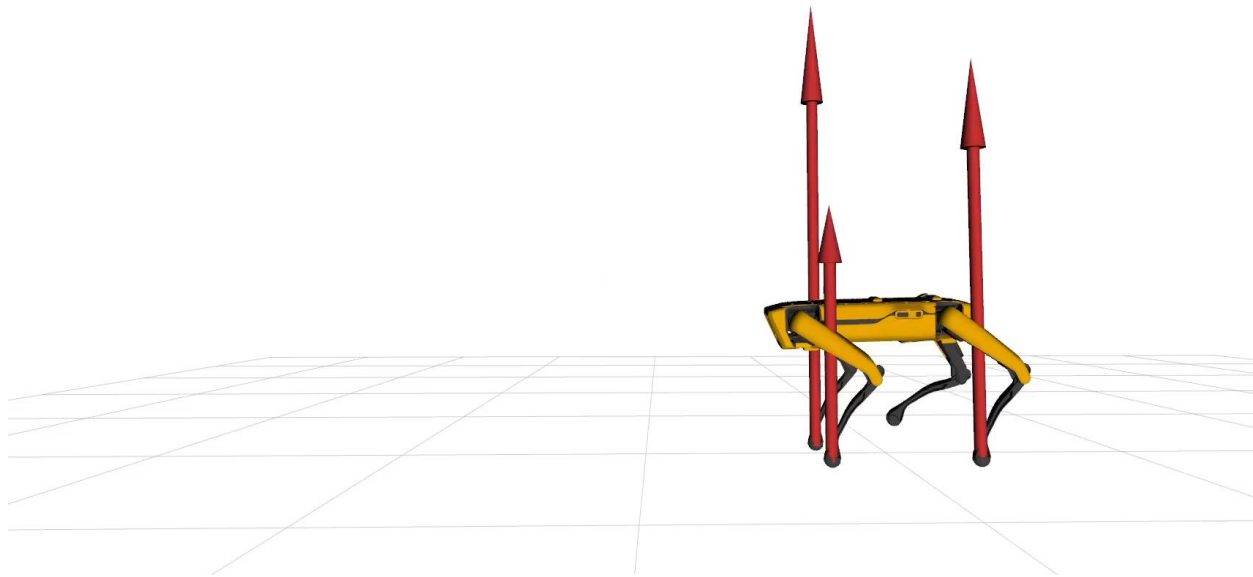
Receding horizon

Online scenario:

- *real system state* is changing
- *user references* are changing

More effective if user
references are changed
“close to the tail”

Idea: update system state and user inputs before each iteration



Spot's receding horizon walking

Problem options:

- Full body dynamics
- $N_{\text{nodes}} = 100$
- $dt = 0.1 \text{ s}$

Constraints:

- [*stance*] zero contact velocity
- [*stance*] unilateral contact force
- [*swing*] zero force
- [*swing*] prescribed z-trajectory
- force-acceleration consistency

Costs:

- reference base link velocity
- postural
- regularization

Parameters:

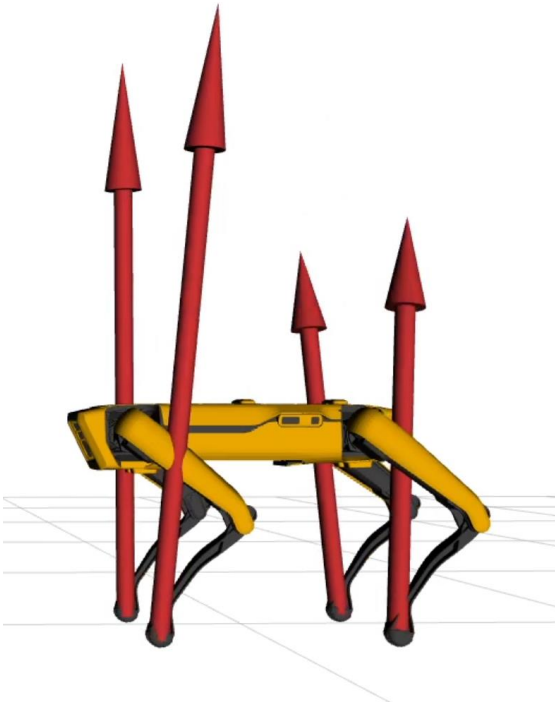
- reference velocity
- contact schedule
- z trajectory

Spot results

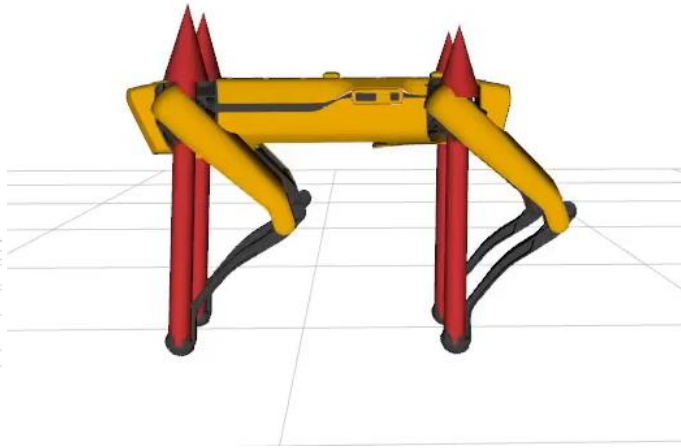
Formulation of the problem is similar between the different motions:

- schedule and duration of the contacts

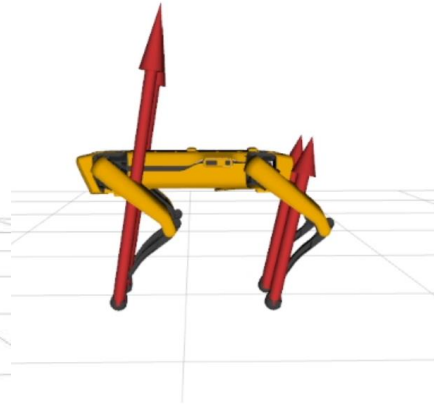
Jump and turn



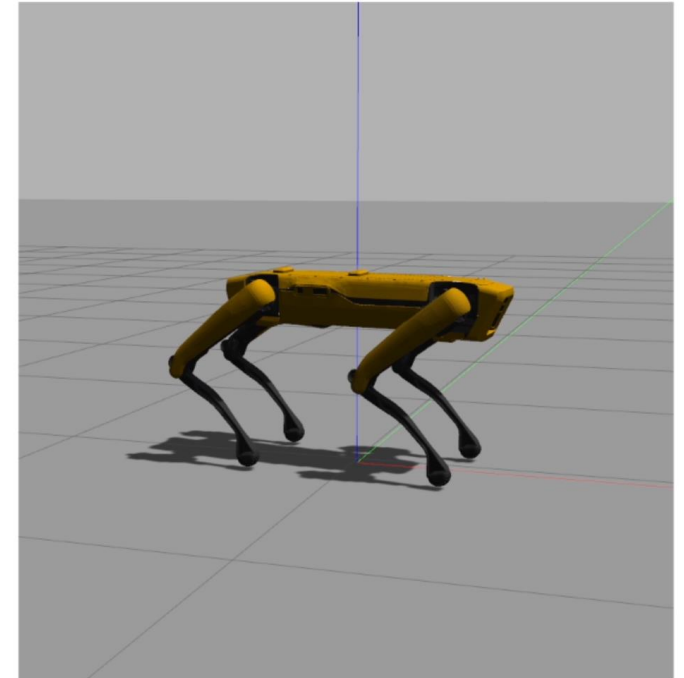
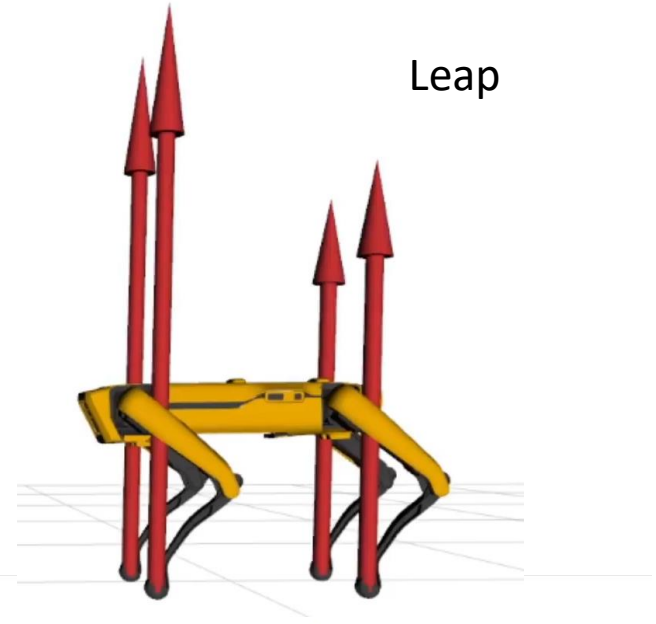
Wheelie



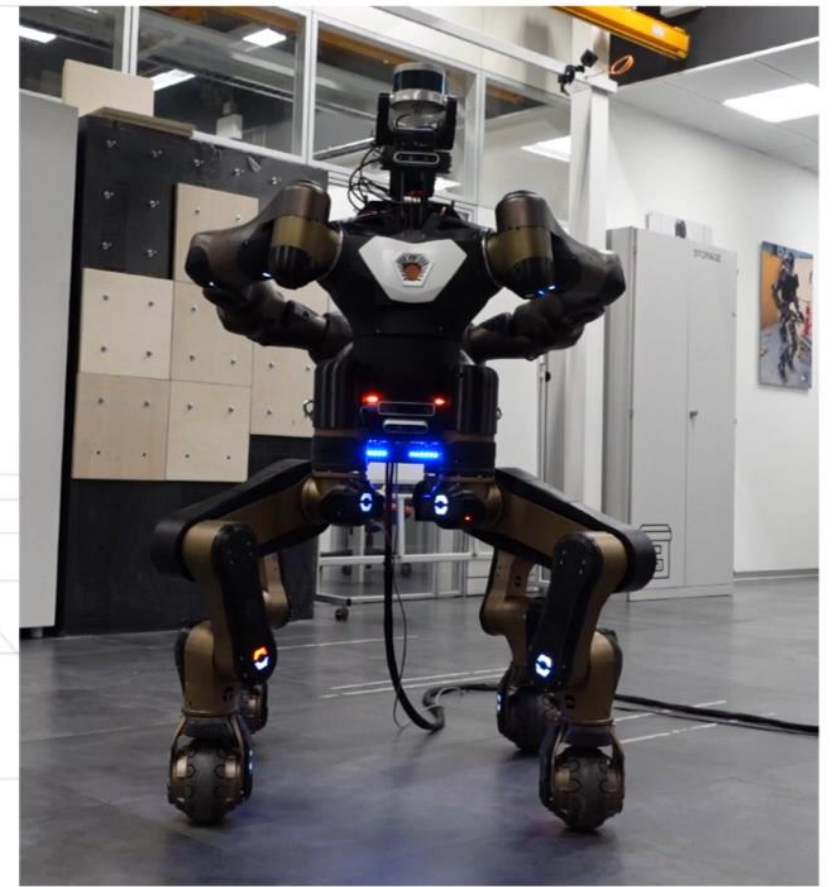
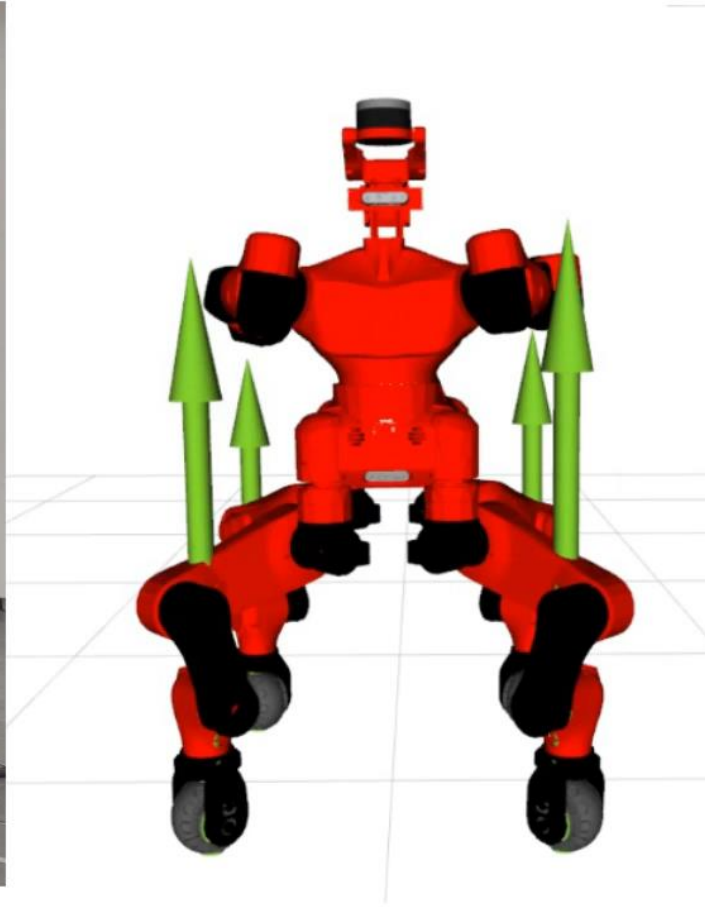
Jump



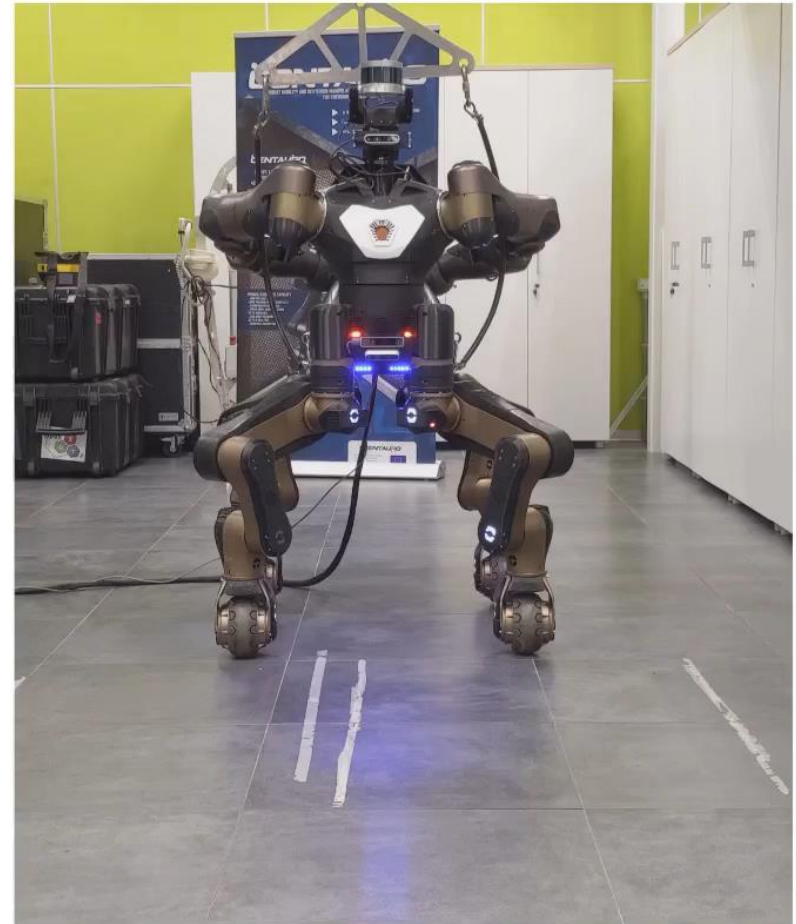
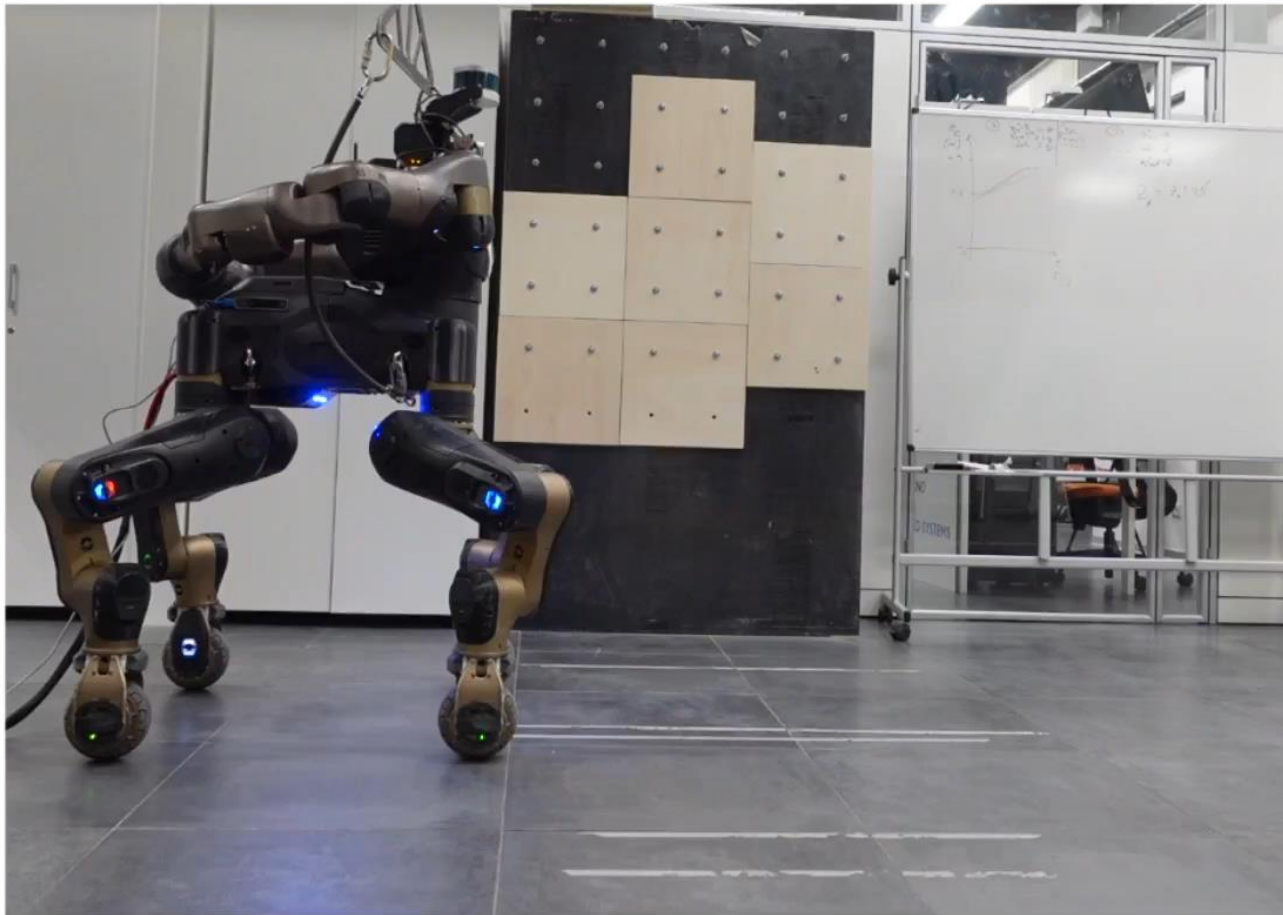
Leap



Centauro robot performing a series of steps to change its heading:



crawling gait on the robot Centauro:
due to the gait scheduling and the final pose, very large strides are required



Dynamic trajectories for a 2-DoF prototype robotic leg for the initial design of a leg intended for agile motions



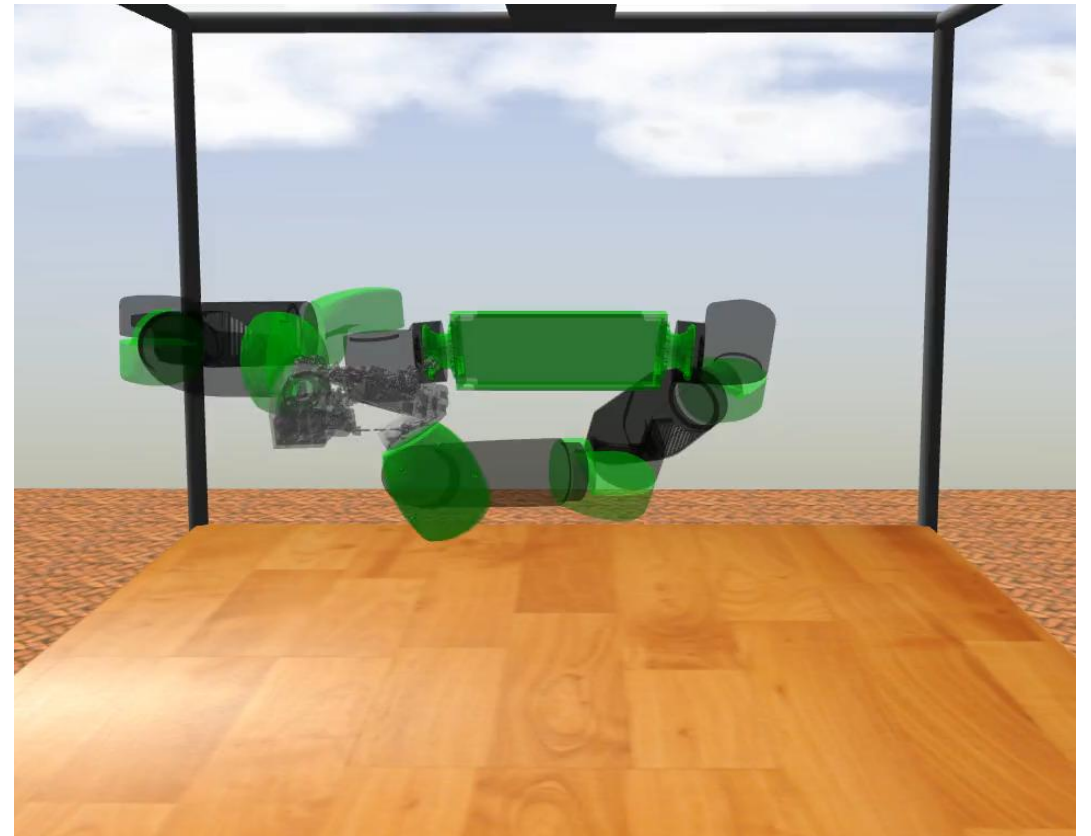
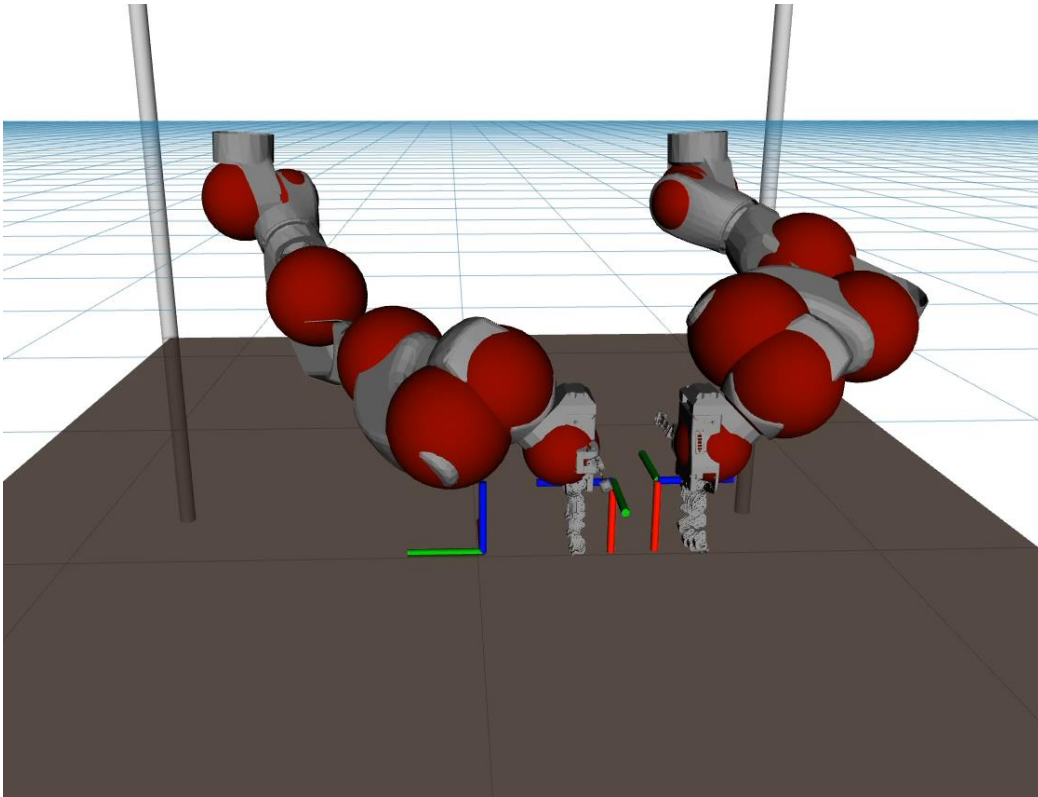
analyze relevant parameters:

- maximum current
- torque
- angular velocity

to carry out the sizing of the motors

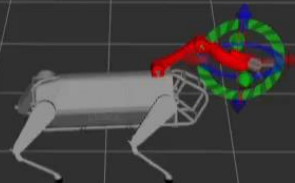
Co-design of bimanual manipulator

- Optimize link length and joint orientation to maximize manipulability



Intuitive robot operation through end-effector Cartesian control

The operator exclusively teleoperate the INAIL-IIT arm end-effector:
the FIELD robot autonomously locomote, reaching the desired goal



Thank you for your attention!

