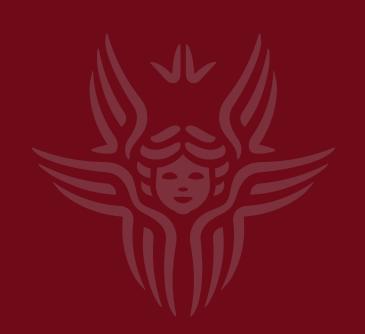
# Grasping On The Move using a mobile manipulator

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**Autonomous Mobile Robotics Course** 2023/2024 Prof: **Giuseppe Oriolo** 

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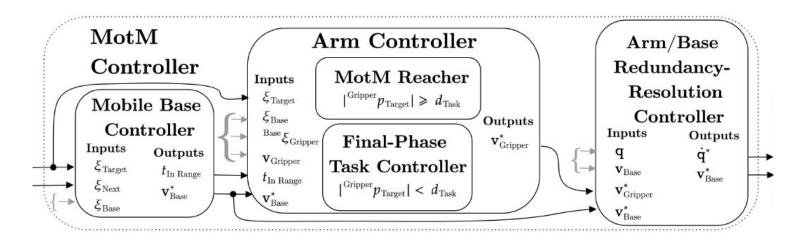
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# **Task - Object Pick and Place**

- Object pick and place.
- Grasping an object while the robot is in motion to decrease the execution time compared to when the base stops to pick up the target.
- Develop a **reactive controller** to **improve robustness** against perception errors, environmental disturbances, and inaccuracies in robot control.
- Reactive and Non-Reactive approaches comparison.

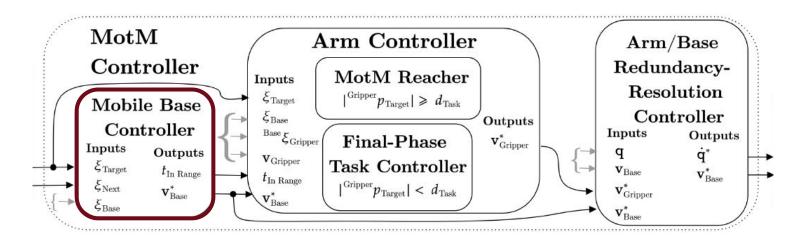


## **Architecture**



- Mobile Base Controller: approaching the robot to the grasping and final target.
- Arm Controller: guides the arm to the grasping target considering base motion.
- Redundancy Resolution Controller: smoothly coordinates arm and base motion.

### **Mobile Base Controller**



- Navigate the robot base to a desired pose  $\xi_C$ 
  - In which the target can be reached by the arm.
- **Key point**: Penalize the steering velocity  $\omega_B$  when the robot is oriented towards the desired pose and penalize the forward velocity  $v_B$  when the robot has reached the desired pose.

#### **Mobile Base Controller**

#### **Driving and steering velocities:**

Grasping phase

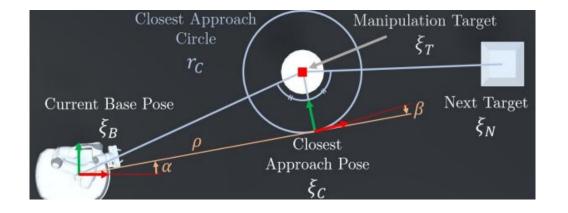
$$v_B = cost.$$

$$\omega_B = (k_\alpha \alpha) \frac{v_B}{\rho}.$$

Placing phase

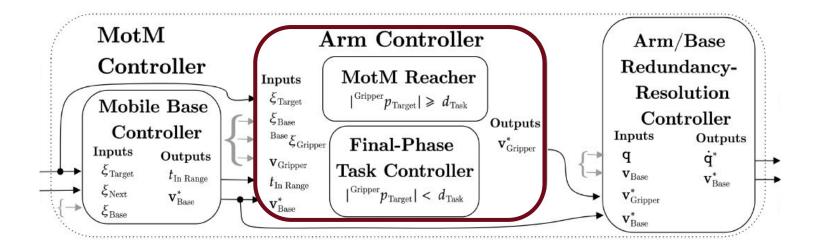
$$v_B = cost.$$

$$\omega_B = (k_\beta \beta) \frac{v_B}{\rho}.$$



- Base-Target distance:  $d = \sqrt{(\xi_{T,x} \xi_{B,x})^2 + (\xi_{T,y} \xi_{B,y})^2}$ .
- Distance error:  $ho = \sqrt{d^2 r_C^2}$
- Orientation error:  $\alpha = \text{atan2}(\xi_{B,y} \xi_{C,y}, \xi_{B,x} \xi_{C,x}) + \pi \theta$

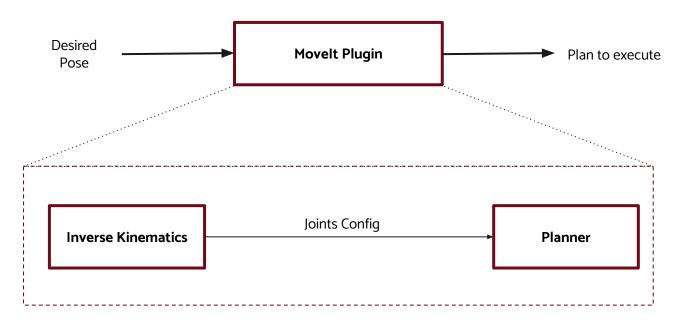
#### **Arm Controller**



• **Arm controller:** guides the arm to desidered end effector pose considering the motion of the base.

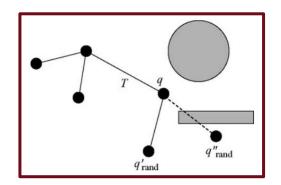
# **Arm Controller - Non Reactive Approach**

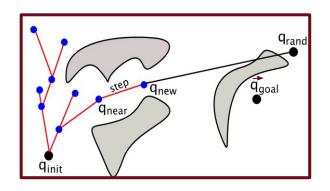
- Movelt Approach: motion planning plugin for ROS.
  - Move TIAGo's gripper grasping frame to a desired pose in Cartesian space
  - Given the desired pose, an arm joints config is computed and the planner generates the plan to reach it.

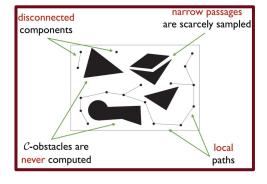


# **Arm Controller - Non Reactive Approach**

- Open Motion Planning Library planners family
  - Single-query Bi-directional Lazy Collision Checking (SBLK)
  - Rapidly-exploring Random Tree (RRT)
  - Probabilistic RoadMap (PRM)







SBLK RRT PRM

# **Arm Controller - Reactive Approach**

- Quintic polynomial timing law  $s(\tau)$  over a total time T
  - End-effector progress along a desired linear cartesian trajectory
  - $\circ$  From an initial position  $\xi_i$  to a final position  $\xi_f$
  - The trajectory is computed online from the current end effector position
- Rest-to-rest trajectory:

$$s(\tau) = 6\tau^{5} - 15\tau^{4} + 10\tau^{3}.$$

$$\dot{s}(\tau) = 30\tau^{4} - 60\tau^{3} + 30\tau^{2}.$$

$$\ddot{s}(\tau) = 120\tau^{3} - 180\tau^{2} + 60\tau.$$

$$\xi_{des}(s(\tau)) = \xi_{i} + s(\tau)(\xi_{f} - \xi_{i}).$$

$$\dot{\xi}_{des}(s(\tau)) = \dot{s}(\tau)(\xi_{f} - \xi_{i}).$$

$$\ddot{\xi}_{des}(s(\tau)) = \ddot{s}(\tau)(\xi_{f} - \xi_{i}).$$

$$\ddot{\xi}_{des}(s(\tau)) = \ddot{s}(\tau)(\xi_{f} - \xi_{i}).$$

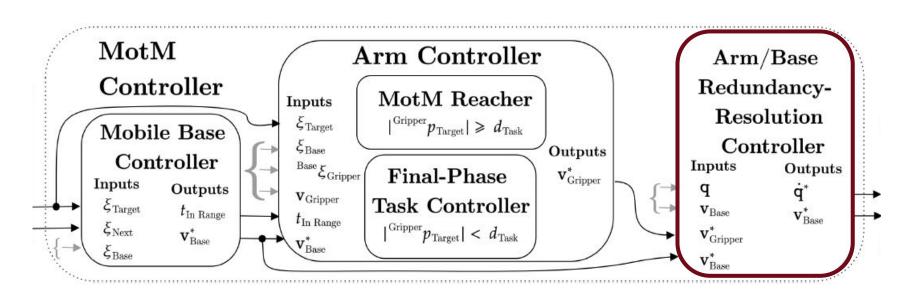
$$\ddot{\xi}_{des}(s(\tau)) = \ddot{s}(\tau)(\xi_{f} - \xi_{i}).$$

# **Arm Controller - Reactive Approach**

- Proportional control:  $\dot{\xi}_r = \dot{\xi}_{des} + k\Delta \xi$ 
  - Cartesian error:  $\Delta \xi = \xi_{des} \xi_i$
  - $\circ$  Gains: k
- **TIAGo arm** controlled in position
  - $\circ$  Joint velocities:  $\dot{q} = J^{\dagger} \dot{\xi}_r$
  - Euler integration:  $q_{k+1} = q_k + \dot{q}\Delta t$ 
    - To get joint positions

# **Redundancy Resolution Controller**

- Formulated as a **Quadratic Programming** (QP) problem
  - Ensures motion coordination between the arm and base.



# **Redundancy Resolution Controller**

argmin 
$$f(x) = \frac{1}{2}x^{T}Hx$$
  
subject to  $\mathcal{J}x = \nu$   
 $\mathcal{A}x \leq \mathcal{B}$   
 $\mathcal{X}^{-} \leq x \leq \mathcal{X}^{+}$ 

- ullet Optimization variable  $x=(\omega_L \quad \omega_R \quad \dot{q}_a \quad \delta)^T$ 
  - $\circ$  Wheels angular velocities (  $\omega_L$  ,  $\omega_R$  )
  - $\circ$  Joint velocities  $\dot{q}_a$
  - Slack variables  $\delta = (\delta_a \quad \delta_{\nu_r})$
- Costs:  $H = diag\{\lambda_q, \lambda_\delta\}$
- Velocities:  $\nu = \left(\frac{\dot{\xi}_r}{\nu_r}\right)$

# **Redundancy Resolution Controller - Costs**

argmin 
$$f(x) = \frac{1}{2}x^{T}Hx$$
subject to 
$$\mathcal{J}x = \nu$$

$$\mathcal{A}x \leq \mathcal{B}$$

$$\mathcal{X}^{-} \leq x \leq \mathcal{X}^{+}$$

- Joints costs consider
  - A factor inversely proportional to base error for angular velocities
  - A constant factor for joint velocities

$$\lambda_q = \left(\frac{1}{\|\rho\|}, \frac{1}{\|\rho\|}, k_a, k_a, k_a, k_a, k_a, k_a, k_a\right)$$

- Slack variables costs consider
  - A factor inversely proportional to the end-effector error

$$\lambda_{\delta} = \left(\frac{1}{\|\Delta\xi\|}, \frac{1}{\|\Delta\xi\|}, \frac{1}{\|\Delta\xi\|}, \frac{1}{\|\Delta\xi\|}\right)$$

# Redundancy Resolution Controller - Equality constraints

argmin 
$$f(x) = \frac{1}{2}x^{T}Hx$$
  
subject to  $\boxed{\mathcal{J}x = \nu}$   
 $\mathcal{A}x \leq \mathcal{B}$   
 $\mathcal{X}^{-} \leq x \leq \mathcal{X}^{+}$ 

- End-effector velocity comes from direct differential kinematics
  - Can be decreased to improve coordination with base movement

$$J\dot{q}_a = \dot{\xi}_r - \delta_a$$

- $\circ$  where  $\dot{\xi}_r$  is the end effector velocity.
- Moving base speed is linked to the wheels angular velocity
  - Can be decreased to improve coordination with the arm movement

$$\frac{\omega_L + \omega_R}{2} = \nu_r - \delta_{\nu}$$

# Redundancy Resolution Controller - Inequality constraints

argmin 
$$f(x) = \frac{1}{2}x^{T}Hx$$
  
subject to  $\mathcal{J}x = \nu$   
 $\mathcal{A}x \leq \mathcal{B}$   
 $\mathcal{X}^{-} \leq x \leq \mathcal{X}^{+}$ 

- Future arm joint positions must fall within the system mechanical limits
  - Considering the used integration method (Euler)

$$q_{k+1} = q_k + \dot{q}\Delta t \in [q_{a,min}, q_{a,max}]$$

• The values of the slack variables must be positive

$$\delta = (\delta_a \quad \delta_{\nu_r}) \ge 0$$

Joint position and velocity must satisfy limits

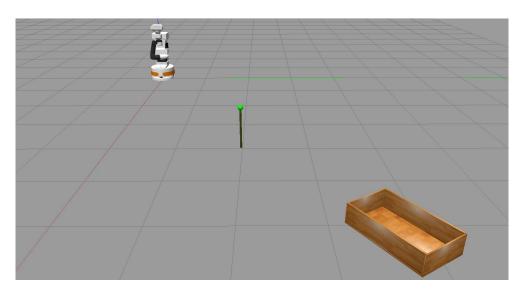
$$\mathcal{X}^- \le x \le \mathcal{X}^+$$

# **Simulation Settings - Environment**

Gazebo simulation environment.

#### Task:

- Grasp the green ball
- Place it inside the brown box



# Simulation Settings - TIAGo Robot



- Developed by PAL Robotics.
- Main features:
  - Differential Drive Robot as mobile base
  - Articulated arm with 7 degrees of freedom
  - Extendable Torso with a Prismatic Joint
  - Gripper or customizable end-effector
- Based on ROS (Robot Operating System).

## **Simulations**

#### On-the-stop simulations:

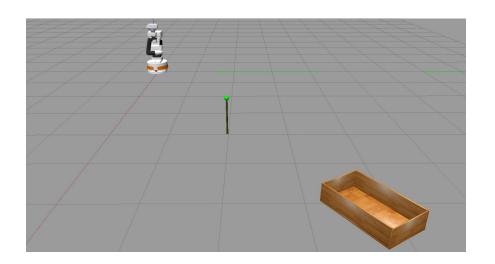
- MoveIT planners
- Quintic polynomial

#### On-the-move simulations:

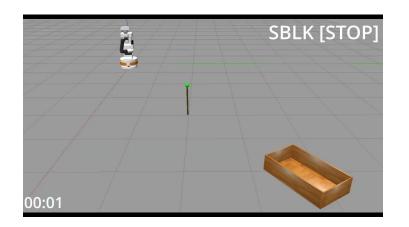
- MoveIT planners
- Quintic polynomial
- Redundancy resolution controller

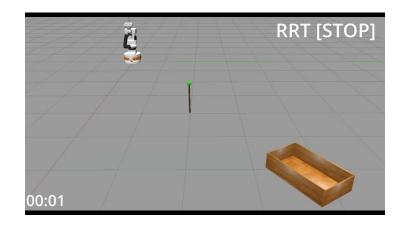
#### Measurements:

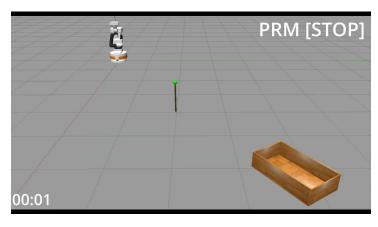
- Task completion
- Grasping trajectory
- Plan finding (only for MovelT planners)
- Success rate



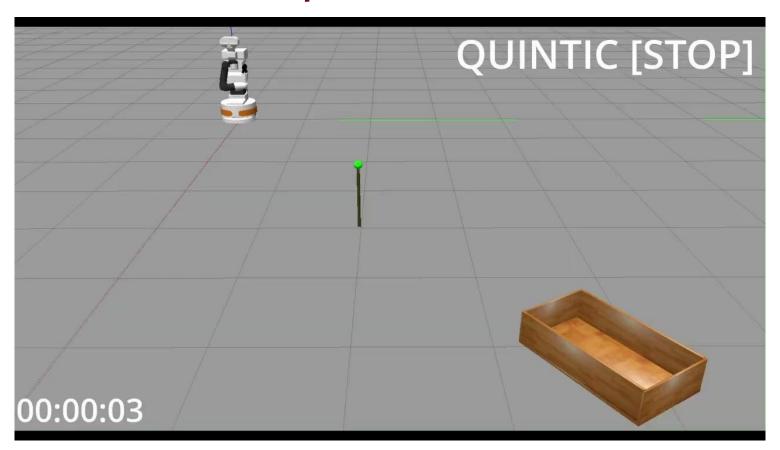
# Simulations - On the stop - Moveit



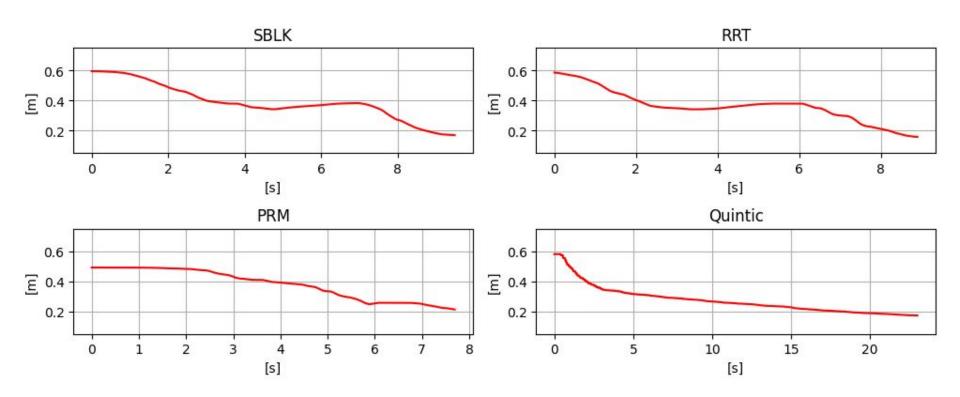




# Simulations - On the stop - Quintic

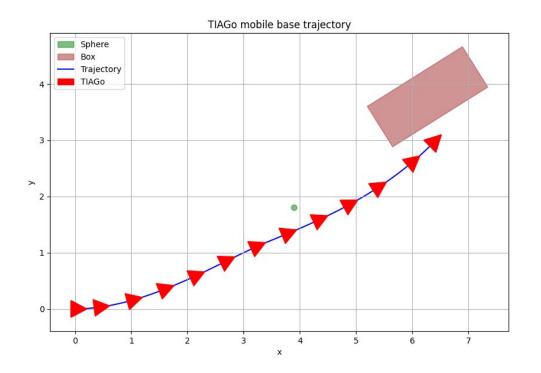


# Simulations - On the stop - Planners & Quintic

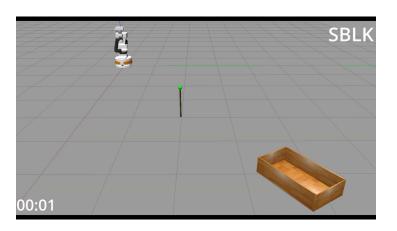


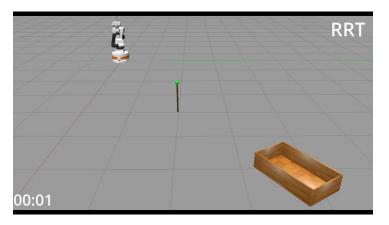
# Simulations - On the move - Base trajectory

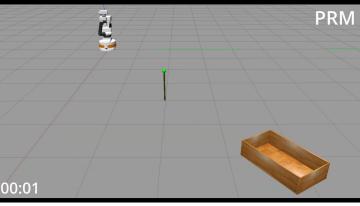
- Simulation base trajectory
- Grasping phase:
  - Ball approaching
  - Ball grasping
- Placing phase:
  - Box approaching
  - Ball placing



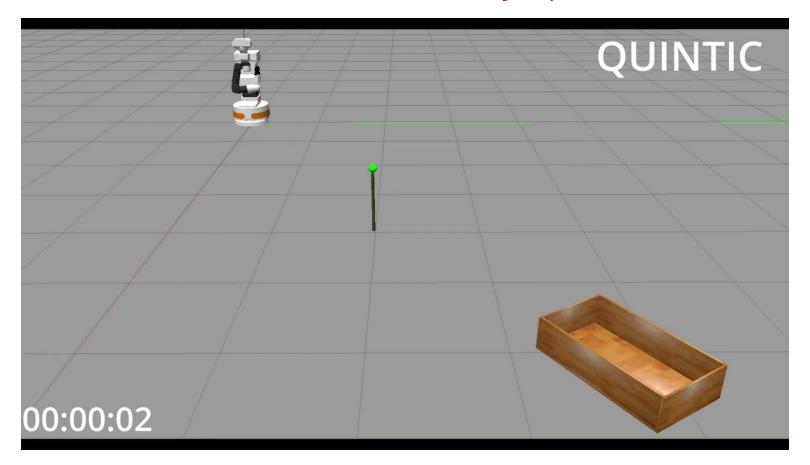
# Simulations - On the move - SBLK



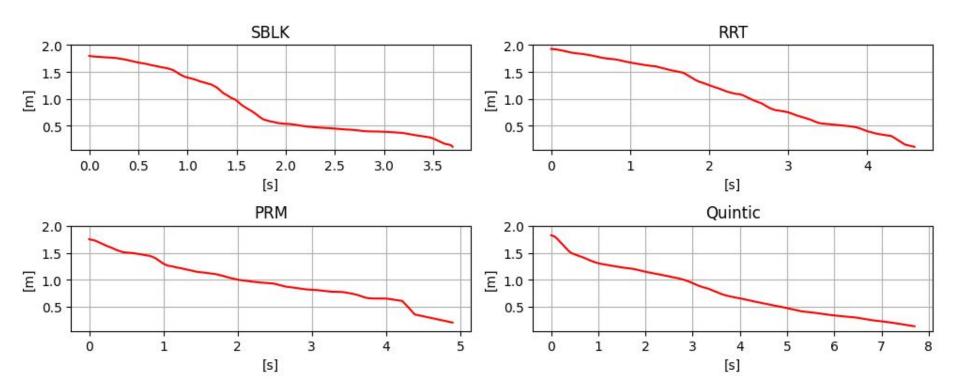




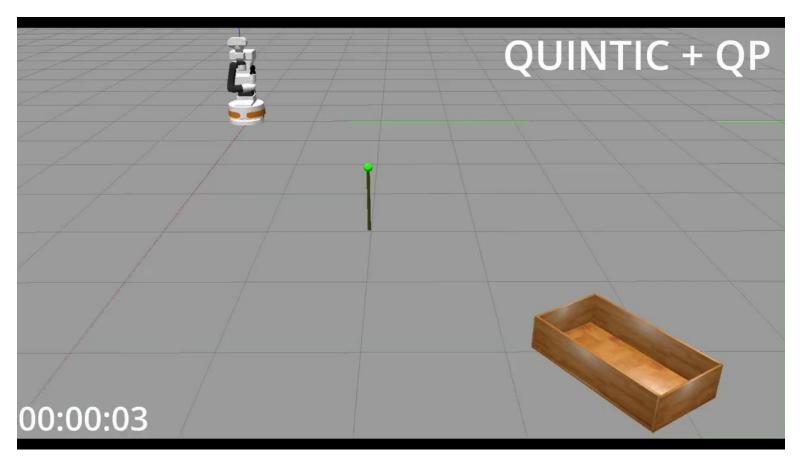
# Simulations - On the move - Quintic polynomial



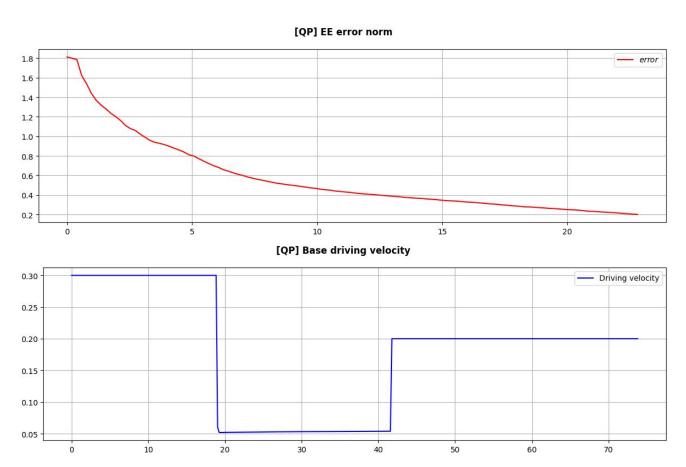
## Simulations - On the move - Planners & Quintic



## Simulations - On the move - Quintic + QP



# Simulations - On the move - Quintic + QP



## **Simulations - Times and Success Rate**

	On-The-Stop			On-The-Move			
Approach	$T_{plan}[s]$	$T_{traj}[s]$	$T_{tot}[s]$	$T_{plan}[s]$	$T_{traj}[s]$	$T_{tot}[s]$	sr
QQP	570	7		107	23.00	74.00	10/10
Quintic		23.02	60.00	1/3	7.67	37.80	9/10
SBLK	6.29	9.17	49.00	3.38	3.58	25.00	3/10
RRT	10.01	8.64	50.00	3.35	4.53	23.00	4/10
PRM	6.39	7.12	42.00	3.73	4.49	22.00	2/10

- Simulations time measurements comparison
  - QQP: Quintic + QP

#### Values:

- $\circ$  Planning time  $T_{plan}$
- $\circ$  Trajectory time  $T_{traj}$
- $\circ$  Total time  $T_{tot}$
- Success rate sr

#### **Conclusions**

- On-The-Move problem solved with a different approaches
  - Base Controller as a Geometric Regulation Task
  - Arm Controller performing planning in Cartesian Space
    - Non reactive with Movelt plugin.
    - **Reactive** with Quintic Polynomial and Kinematic Integration.
  - Redundancy Resolution Controller as a QP
    - Motion coordination between arm and base.
- Reactive approach allows real-time adaptation
- Reduction of task execution time on-the-move.
- Best approach: Quintic Polynomial + Redundancy Resolution Controller
  - In terms of success rate.

#### References

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- [4] Jesse Haviland, Niko Sunderhauf, and Peter Corke. A holistic approach to reactive mobile manipulation. IEEE Robotics and Automation Letters, 7(2):3122–3129, April 2022.
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