

# Kinematic Planning for Mobile Manipulators Using Optimal Control

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\* Gifftthaler, Markus, et al. "Efficient kinematic planning for mobile manipulators with non-holonomic constraints using optimal control." *2017 IEEE International Conference on Robotics and Automation (ICRA)*. IEEE, 2017.

# Motivation

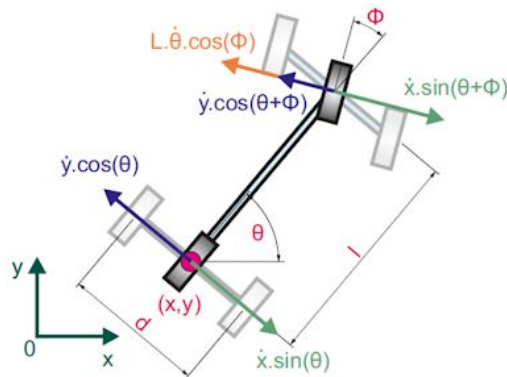
- Efficiently design controllers that satisfies non-holonomic constraints

$$g(\mathbf{q}, \dot{\mathbf{q}}, t) = 0$$

# Motivation

- Efficiently design controllers that satisfies non-holonomic constraints

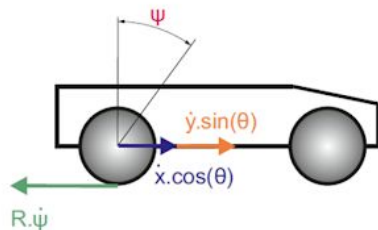
$$g(\mathbf{q}, \dot{\mathbf{q}}, t) = 0$$



**Example:** Car-like robot

Assuming pure rolling  
contact:

$$-\dot{x} \sin \theta + \dot{y} \cos \theta = 0$$



# Motivation

- Operational space tracking for mobile manipulation



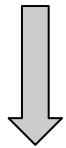
In-Situ Fabrication 1



In-Situ Fabrication 2

# Motivation

- Most commercially available robots are position/velocity controlled

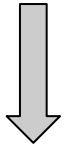


## Kinematic planning

(i.e. dynamics of system is neglected)

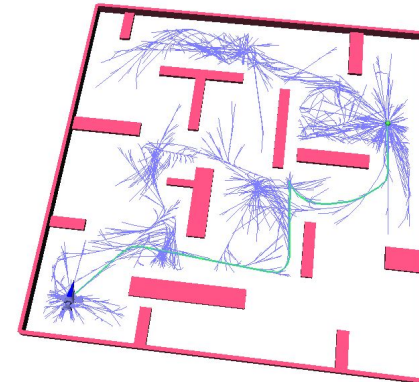
# Motivation

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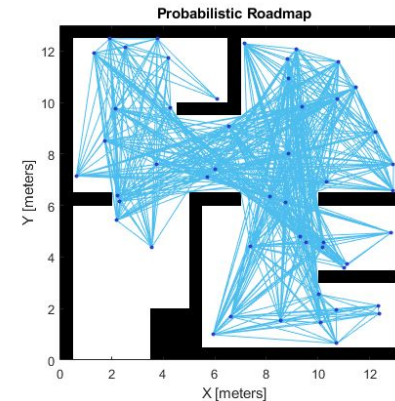


## Kinematic planning

Prior works produce kinematically feasible solutions but without optimality guarantees



RRT



PRM

# Contributions

- Introduces 'constrained SLQ' to kinematic planning
  - Kinematic feedback laws are compliant with constraints
  - Does not suffer from issues in other algorithms that employ inverse kinematics (such as when approaching singularities)

^ Constrained Sequential Linear Quadratic Optimal Control

# Outline

- Primers:
  - Sequential Quadratic Programming
  - Constrained SLQ
- Algorithm Overview
  - Formulation of planning and control problem
  - Receding Horizon Control Setup
- Experiments & Results



# Sequential Quadratic Programming

- Used commonly for constrained non-linear optimization problems

$$\begin{array}{ll} \text{(NLP)} & \text{minimize } f(x) \quad \leftarrow \text{Linear / quadratic function} \\ & \text{over } x \in \mathbb{R}^n \\ & \text{subject to } \left. \begin{array}{l} h(x) = 0 \\ g(x) \leq 0 \end{array} \right\} \text{Affine functions} \end{array}$$

- Iterative procedure which models NLP for given iterate as a Quadratic Programming (QP) subproblem

# Sequential Quadratic Programming

- Formulation of QP subproblem for current iterate  $x^k$

- Local quadratic approximation of objective  $f$  :

$$f(x) \approx f(x^k) + \nabla f(x^k)(x - x^k) + \frac{1}{2} (x - x^k)^T H f(x^k)(x - x^k)$$

- Local affine approximations of constraints

$$g(x) \approx g(x^k) + \nabla g(x^k)(x - x^k)$$

$$h(x) \approx h(x^k) + \nabla h(x^k)(x - x^k)$$

# Sequential Quadratic Programming

- Formulation of QP subproblem for current iterate  $x^k$

- Set:

$$d(x) := x - x^k, \quad B_k := Hf(x^k)$$

- QP subproblem:

$$\text{minimize} \quad \nabla f(x^k)^T d(x) + \frac{1}{2} d(x)^T B_k d(x)$$

$$\text{over} \quad d(x) \in \mathbb{R}^n$$

$$\text{subject to} \quad h(x^k) + \nabla h(x^k)^T d(x) = 0$$

$$g(x^k) + \nabla g(x^k)^T d(x) \leq 0,$$

# Constrained SLQ

- Class of dynamic programming algorithm
  - Scales linearly with optimization horizon problem
  - Designs both feedforward plans and feedback controller
- Solves 
$$\min_{\mathbf{u}(\cdot)} \left\{ \Phi(\mathbf{x}(t_f)) + \int_{t_0}^{t_f} L(\mathbf{x}, \mathbf{u}, t) dt \right\}$$
 control problem subject to
 
$$\begin{aligned} \dot{\mathbf{x}} &= \mathbf{f}(\mathbf{x}, \mathbf{u}) & \mathbf{x}(t_0) &= \mathbf{x}_0 \\ \mathbf{g}_1(\mathbf{x}, \mathbf{u}, t) &= 0 & \mathbf{g}_2(\mathbf{x}, t) &= 0 \end{aligned}$$

# Constrained SLQ

$$\min_{\mathbf{u}(\cdot)} \left\{ \Phi(\mathbf{x}(t_f)) + \int_{t_0}^{t_f} L(\mathbf{x}, \mathbf{u}, t) dt \right\}$$

subject to

$$\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, \mathbf{u})$$

$$\mathbf{x}(t_0) = \mathbf{x}_0$$

$$\mathbf{g}_1(\mathbf{x}, \mathbf{u}, t) = 0$$

$$\mathbf{g}_2(\mathbf{x}, t) = 0$$

Time-varying  
linear feedback  
control law

$$\mathbf{u}(\mathbf{x}, t) = \mathbf{u}_{ff}(t) + \mathbf{K}(t)\mathbf{x}(t)$$

In each iteration:

1. Approximate above with local QP subproblem
2. Solve using Riccati-based approach

# Constrained SLQ

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## Algorithm 1 SLQ Algorithm

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### Given

- Mode switch sequence and switching times
- Cost function (4) and system dynamics (5)

### Initialization

- Initialize the controller with a stable control law,  $\mathbf{u}_0(\cdot)$

### repeat

- Forward integrate system dynamics:  $\{ \bar{\mathbf{x}}(z), \bar{\mathbf{u}}(z) \}_{z=0}^I$ ,
- Compute the LQ approximation of the problem along the nominal trajectory, (7) and (8).
- Solve the final value differential equations, (9-11).
- Line search for the optimal  $\alpha$  with policy (17)
- Update control law:  $\mathbf{u}^*(z, \mathbf{x}) = \bar{\mathbf{u}}(z) + \alpha^* \mathbf{l}(z) + \mathbf{L}(z) \delta \mathbf{x}$

**until**  $\|\mathbf{l}(\cdot)\|_2 < l_{min}$  or maximum number of iterations

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# Kinematic Planning and Control Problem

State  $\mathbf{x} = \mathbf{q}$

Input  $\mathbf{u} = \dot{\mathbf{q}}$

$$\min_{\mathbf{u}(\cdot)} \left\{ \Phi(\mathbf{x}(t_f)) + \int_{t_0}^{t_f} L(\mathbf{x}, \mathbf{u}, t) dt \right\}$$

subject to

$$\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, \mathbf{u}) \quad \mathbf{x}(t_0) = \mathbf{x}_0$$

$$\mathbf{g}_1(\mathbf{x}, \mathbf{u}, t) = 0 \quad \mathbf{g}_2(\mathbf{x}, t) = 0$$



# Kinematic Planning and Control Problem

State  $\mathbf{x} = \mathbf{q}$

Input  $\mathbf{u} = \dot{\mathbf{q}}$

$$\int_{t_0}^{t_f} \mathbf{u}(t)^\top \mathbf{R} \mathbf{u}(t) dt + (\mathbf{x}(t_f) - \mathbf{x}_r)^\top \mathbf{Q}_f (\mathbf{x}(t_f) - \mathbf{x}_r)$$

$$\min_{\mathbf{u}(\cdot)} \left\{ \Phi(\mathbf{x}(t_f)) + \int_{t_0}^{t_f} L(\mathbf{x}, \mathbf{u}, t) dt \right\}$$

subject to

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$$\mathbf{g}_1(\mathbf{x}, \mathbf{u}, t) = 0 \quad \mathbf{g}_2(\mathbf{x}, t) = 0$$

Desired terminal  
state

# Kinematic Planning and Control Problem

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Desired terminal state

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$$\mathbf{x}(t_0) = \mathbf{x}_0$$

$$\mathbf{g}_1(\mathbf{x}, \mathbf{u}, t) = 0$$

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Covers all non-holonomic constraints

$$\mathbf{g}_1(\mathbf{q}, \dot{\mathbf{q}}, t) = 0$$

# Kinematic Planning and Control Problem

State  $\mathbf{x} = \mathbf{q}$

Input  $\mathbf{u} = \dot{\mathbf{q}}$

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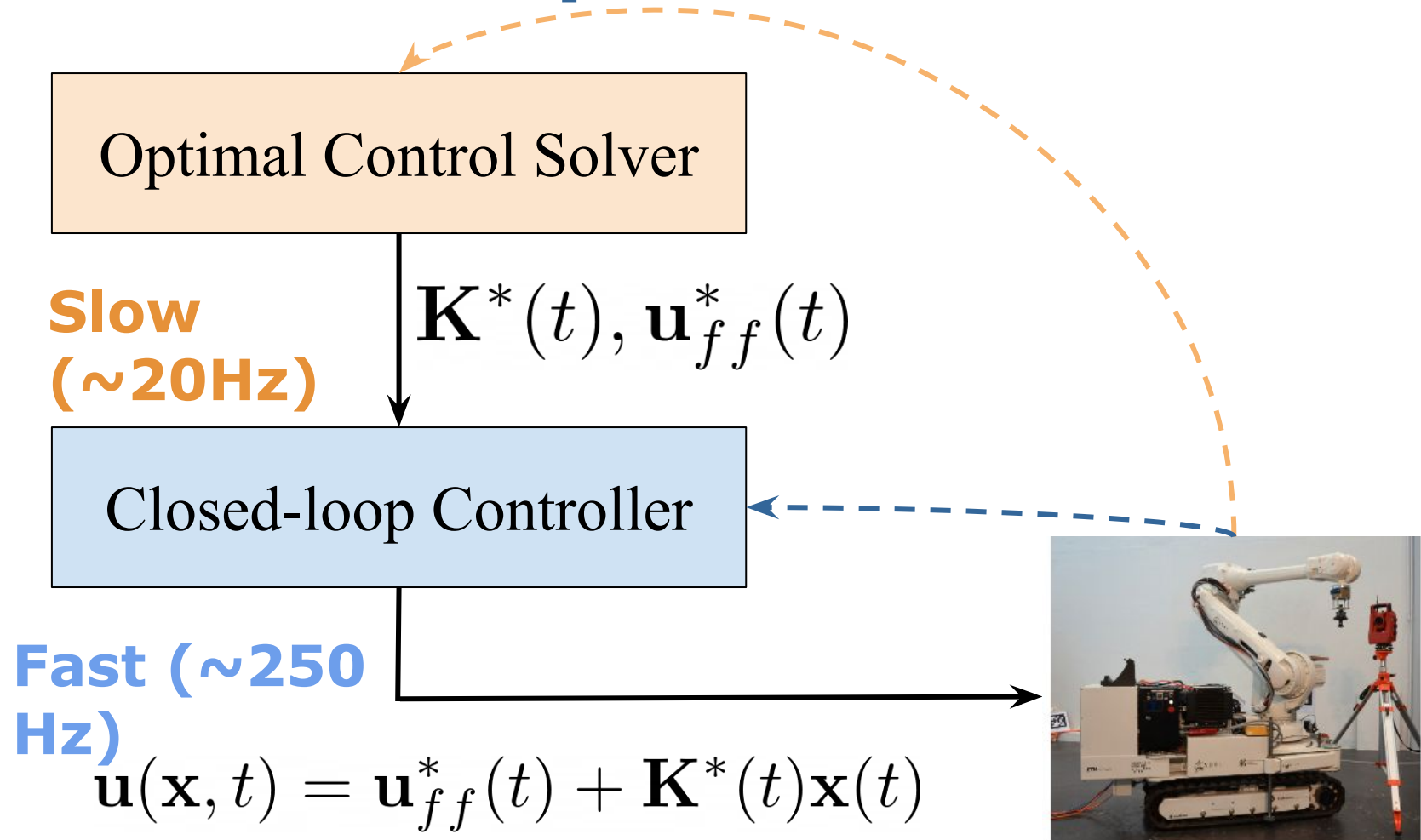
Covers  
holonomic  
operational-space

e.g.  $\mathbf{g}_2(\mathbf{q}, t) : \mathbf{r}_{ee}^{frw} = \mathbf{r}_{ee,ref}^{frw}(t, \mathbf{q})$   
constraints

Covers all non-holonomic  
constraints

$$\mathbf{g}_1(\mathbf{q}, \dot{\mathbf{q}}, t) = 0$$

# Receding Horizon Optimal Control Setup

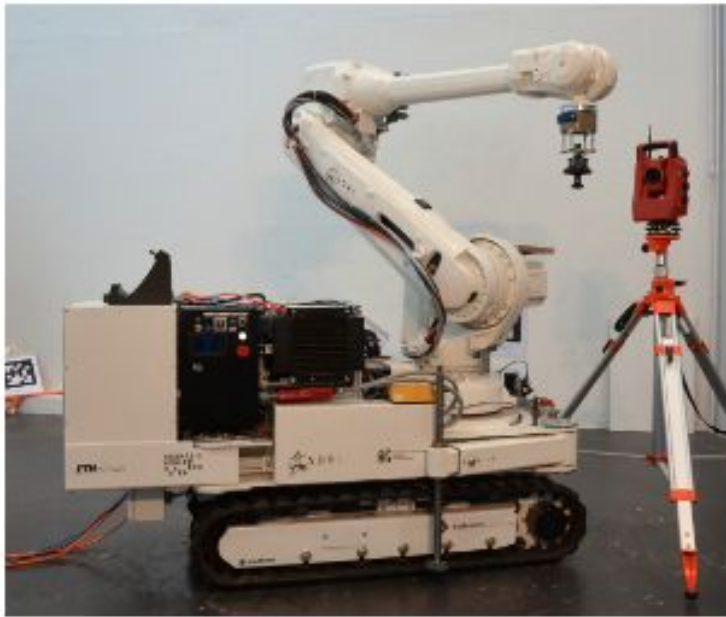


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# Experiments

- IF1: Tracked Mobile Manipulator
- IF2: Legged/wheeled mobile robot with DoF 26



In-Situ Fabrication 1



In-Situ Fabrication 2

# Experiments: IF2

## Wheel Constraints:

- Rolling condition
- Wheel should not lift off of the ground

$$\mathbf{g}_1(\mathbf{q}, \dot{\mathbf{q}}, t) := \mathbf{v}_P^{frW} = 0$$



$$\mathbf{v}_P^{frW} = \mathbf{R}_{frB}^{frW}(\boldsymbol{\theta}) \cdot \left( \mathbf{v}_B^{frB} + \mathbf{v}_P^{frB} + \boldsymbol{\omega}_B^{frB} \times \mathbf{r}_{BP}^{frB} \right)$$

$$\begin{bmatrix} \boldsymbol{\omega}_C^{frB\top} & \mathbf{v}_C^{frB\top} \end{bmatrix}^\top = \mathbf{J}_C^{frB}(\boldsymbol{\varphi}) \cdot \dot{\boldsymbol{\varphi}}$$

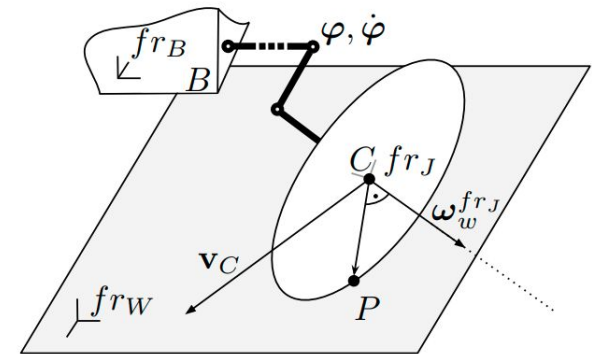
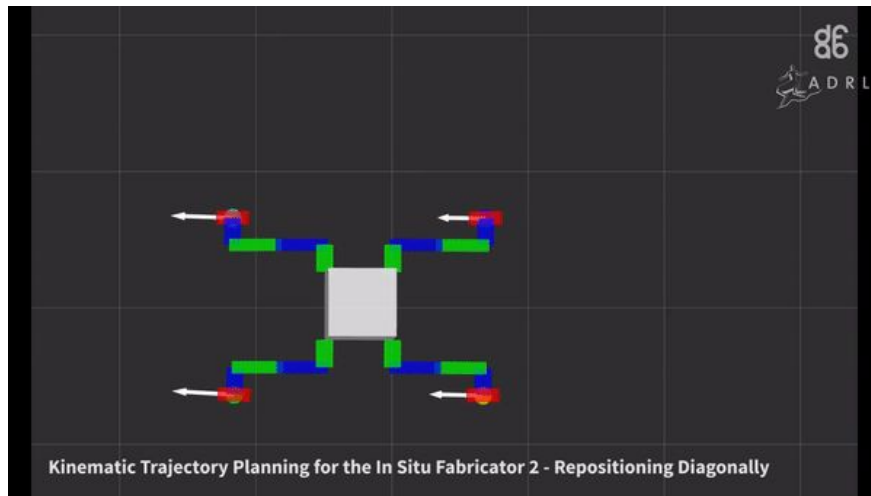
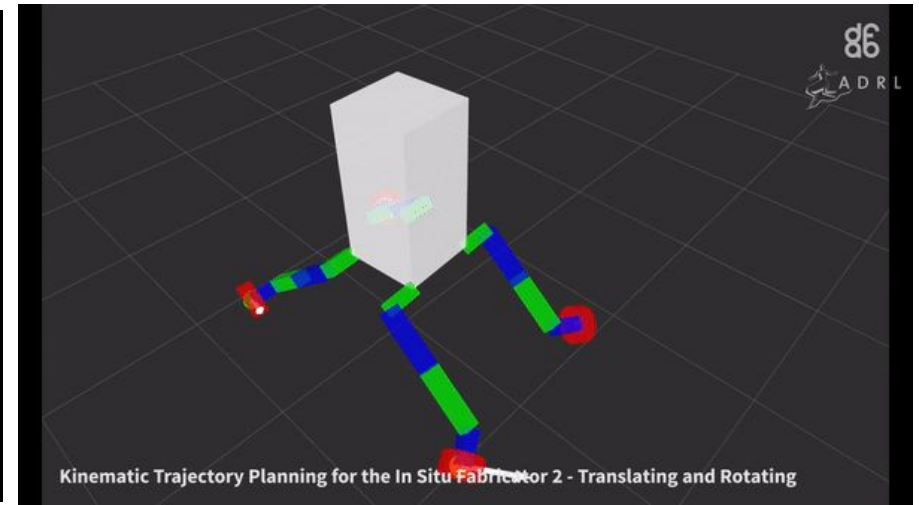


Fig. 2. Sketch of an ideal wheel attached to the robot's trunk through an arbitrary serial chain of joints and links.

# Experiments: IF2



Robot needs to move  
to location (1, 1)



Robot has to  
translate 1m away  
and rotate trunk by  
180 degrees



# Experiments: IF1

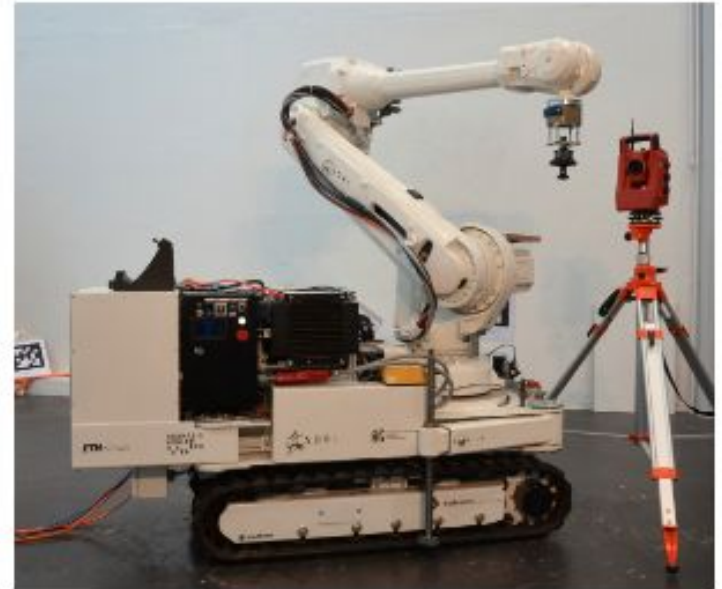
Tracks simplified to two -wheel model

$$\dot{y}_c \cos \theta - \dot{x}_c \sin \theta - \dot{\theta} d = 0$$

Operational-space tracking constraint

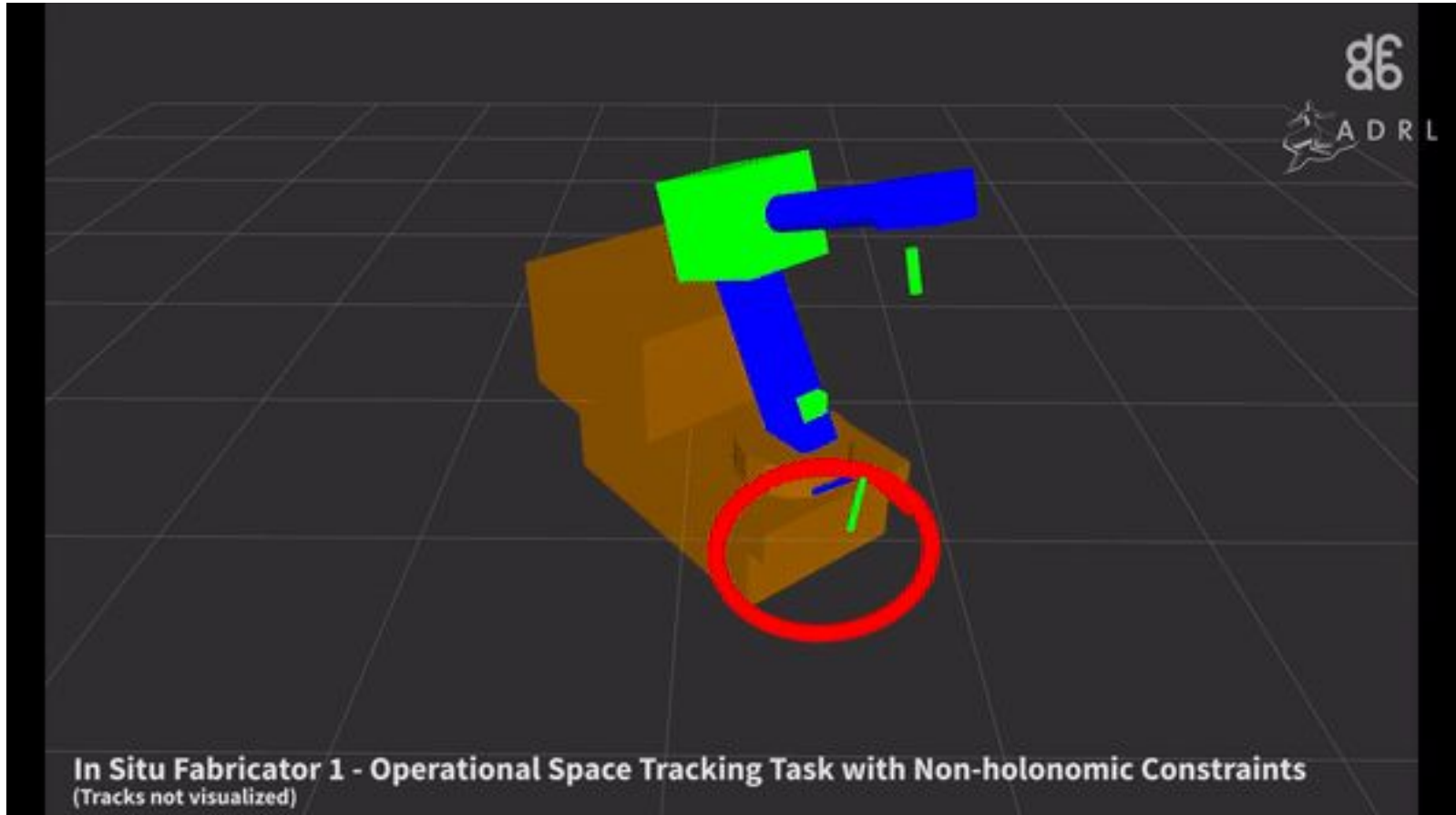
$$\mathbf{g}_1(\mathbf{q}, \dot{\mathbf{q}}, t) : \quad \mathbf{v}_{ee}^{frw} = \mathbf{v}_{ee, \text{ref}}^{frw}(t, \mathbf{q}, \dot{\mathbf{q}}) \quad \text{or}$$

$$\mathbf{g}_2(\mathbf{q}, t) : \quad \mathbf{r}_{ee}^{frw} = \mathbf{r}_{ee, \text{ref}}^{frw}(t, \mathbf{q}) .$$



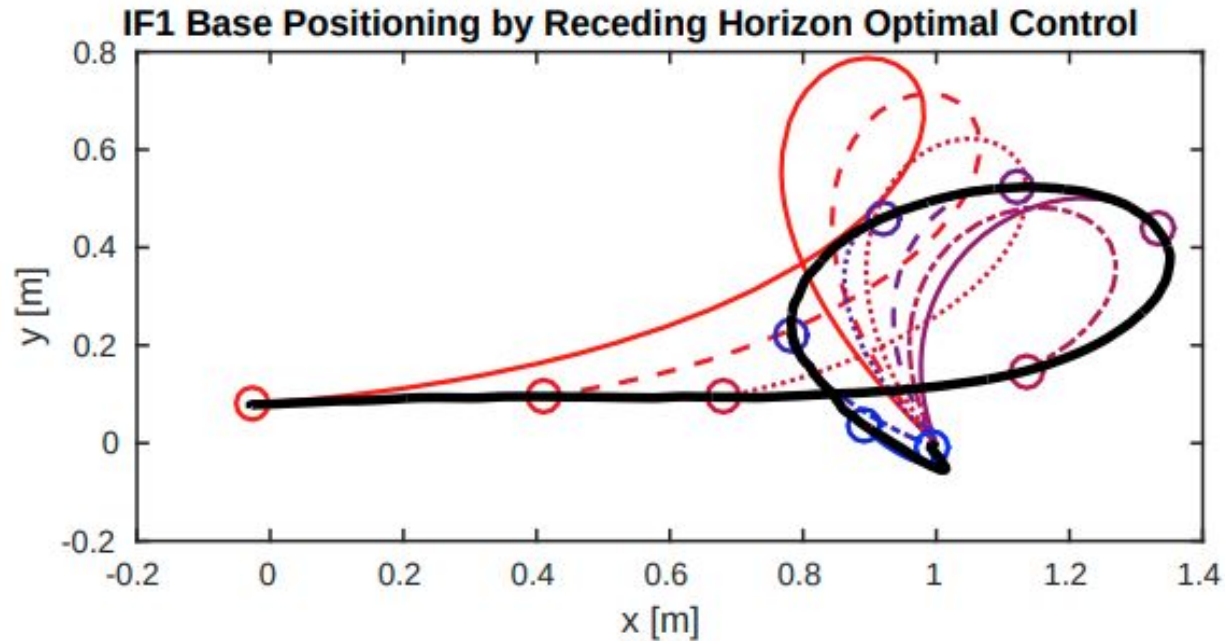
**Note:** Reference positions and velocities trajectories need to be continuous and differentiable.

# Experiments: IF1



End-effector needs to follow given task-space trajectory

# Experiments: IF1



Computed paths for base repositioning/  
reorientation task using RHOC

# Remarks

- Proposed method has linear time complexity in local regime w.r.t. time horizon
  - Note: not intended to compete with planners in cluttered environments
- Uses adaptive step-size ODE solver
- Achieves 50-100 Hz replanning rates (for IF2) on single core CPU

# Remarks

## Ongoing usages of constrained SLQ



Grandia, Ruben, et al. "Feedback MPC for Torque-Controlled Legged Robots." *IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS 2019)*. 2019.



Minniti, Maria Vittoria, et al. "Whole-body mpc for a dynamically stable mobile manipulator." *IEEE Robotics and Automation Letters* 4.4 (2019): 3687-3694.

**Thank you for your attention!**