

Trajectory Optimization for Walking

Optimization-based Robot Control

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Task-Space Inverse Dynamics needs reference trajectories.

How to compute them for a walking robot?

Table of contents

1. Limits of Instantaneous Control
2. Linear Inverted Pendulum Model (LIPM)
3. Center of Mass Trajectory Optimization with LIPM
4. Foot-step Planning
5. Implementation in Python
6. Connection with TSID

Limits of Instantaneous Control

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Example of car moving towards wall.

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CONS: More computationally expensive.

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Solution

Use traj-opt offline to compute reference trajectory.

Use TSID online to track reference trajectory.

Trajectory Optimization through Contacts

Traj-opt for locomotion/manipulation is really hard!

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Option 2: Soft (but stiff) Contacts

Stiff differential equations → Veeeery slow!

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Option 2: Soft (but stiff) Contacts

Stiff differential equations → Veeeery slow!

Solution

Use rigid contacts, but fix contact sequence → Time-varying dynamical system (not hybrid!)

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Common models for locomotion:

- Inverted Pendulum
- **Linear Inverted Pendulum**
- Centroidal Dynamics (i.e. single rigid body dynamics)

Linear Inverted Pendulum Model (LIPM)

Center of Mass and Angular Momentum

Newton equation (center-of-mass dynamics):

$$m(\ddot{c} + g) = \sum_i f_i \quad (1)$$

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where:

- c : center of mass (CoM)
- l : angular momentum (expressed at CoM)
- m : robot mass
- g : gravity acceleration
- f_i : i -th contact force
- p_i : i -th contact point

Flat Ground (1/2)

Assume:

- contacts with flat ground: $p_i^z = 0$
- constant angular momentum: $\dot{l} = 0$
- constant CoM height: $\dot{c}^z = \ddot{c}^z = 0$

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Then we get (Wieber, Tedrake, and Kuindersma 2015):

$$c^{xy} - \frac{c^z}{g^z} \ddot{c}^{xy} = \frac{\sum_i f_i^z p_i^{xy}}{\underbrace{\sum_i f_i^z}_{\text{Center of Pressure}}} \quad (3)$$

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$$f_i^z \geq 0 \iff z^{xy} \triangleq \frac{\sum_i f_i^z p_i^{xy}}{\sum_i f_i^z} \in \text{conv}(p_i^{xy})$$

Rearrange (3) as:

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CoM acc \ddot{c}^{xy} given by force pushing CoM c^{xy} away from CoP z^{xy}

Flat Ground (2/2)

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Interpretation

CoM acc \ddot{c}^{xy} given by force pushing CoM c^{xy} away from CoP $z^{xy} \rightarrow$

UNSTABLE!

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CoM acc \ddot{c}^{xy} given by force pushing CoM c^{xy} away from CoP $z^{xy} \rightarrow$

UNSTABLE!

Same dynamics as linearized Inverted Pendulum.

LIPM as Linear Dynamical System

Rewrite (3) as:

$$\underbrace{\begin{bmatrix} \dot{c}^{xy} \\ \ddot{c}^{xy} \end{bmatrix}}_{\dot{x}} = \begin{bmatrix} 0 & I \\ \omega^2 & 0 \end{bmatrix} \underbrace{\begin{bmatrix} c^{xy} \\ \dot{c}^{xy} \end{bmatrix}}_x + \begin{bmatrix} 0 \\ -\omega^2 \end{bmatrix} \underbrace{z^{xy}}_u \quad (5)$$

where $\omega^2 \triangleq \frac{g^z}{c^z}$.

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where $\omega^2 \triangleq \frac{g^z}{c^z}$.

Discretize with time step δt :

$$x^+ = \underbrace{\begin{bmatrix} \cosh(\omega\delta t) & \omega^{-1} \sinh(\omega\delta t) \\ \omega \sinh(\omega\delta t) & \cosh(\omega\delta t) \end{bmatrix}}_A x + \underbrace{\begin{bmatrix} 1 - \cosh(\omega\delta t) \\ -\omega \sinh(\omega\delta t) \end{bmatrix}}_B u \quad (6)$$

Center of Mass Trajectory Optimization with LIPM

Follow reference trajectory of:

- CoP $P = [p_0 \ \dots \ p_{N-1}]$ (i.e. foot steps),
- CoM position $C^{ref} = [c_0^{ref} \ \dots \ c_N^{ref}]$
- CoM velocity $\dot{C}^{ref} = [\dot{c}_0^{ref} \ \dots \ \dot{c}_N^{ref}]$

Key Idea

Follow reference trajectory of:

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Foot steps and timing P predefined by user.

Keep CoP close to foot center for robustness.

C^{ref} could be straight line.

Formulation

$$\begin{aligned} & \underset{C, \dot{C}, U}{\text{minimize}} && \sum_k \frac{\beta}{2} \|c_k - c_k^{ref}\|^2 + \frac{\gamma}{2} \|\dot{c} - \dot{c}^{ref}\|^2 + \frac{\alpha}{2} \|u_k - p_k\|^2 \\ & \text{subject to} && p_k - \frac{s}{2} \leq u_k \leq p_k + \frac{s}{2} && k = 0 \dots N - 1 \\ & && x_{k+1} = Ax_k + Bu_k && k = 0 \dots N - 1 \\ & && x_0 = x_{initial} \\ & && x_N = x_{final} \end{aligned} \quad (7)$$

where:

- $s \in \mathbb{R}^2$ = foot size in x and y directions
- $C = \begin{bmatrix} c_0 & \dots & c_N \end{bmatrix}$
- $\dot{C} = \begin{bmatrix} \dot{c}_0 & \dots & \dot{c}_N \end{bmatrix}$
- $x_k = (c_k, \dot{c}_k)$
- α, β, γ = user-defined weights

Foot-step Planning

Optimize for foot step positions, but...

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Add P to decision variables \rightarrow Problem remains QP! (Herdt et al. 2010)

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Bound distance between successive foot steps.

CoM Trajectory Optimization with Foot-Step Planning

$$\begin{aligned} & \underset{C, \dot{C}, U, P}{\text{minimize}} && \sum_k \frac{\beta}{2} \|c_k - c_k^{ref}\|^2 + \frac{\gamma}{2} \|\dot{c} - \dot{c}^{ref}\|^2 + \frac{\alpha}{2} \|u_k - p_k\|^2 \\ & \text{subject to} && p_k - \frac{s}{2} \leq u_k \leq p_k + \frac{s}{2} && k = 0 \dots N - 1 \\ & && x_{k+1} = Ax_k + Bu_k && k = 0 \dots N - 1 \quad (8) \\ & && x_0 = x_{initial} \\ & && x_N = x_{final} \\ & && p_{k+1} - p_k \in \mathcal{P}_k && k = 0 \dots N - 1 \end{aligned}$$

Implementation in Python

```
# Inverted pendulum parameters:
# -----
foot_length = conf.lxn + conf.lxp # foot size in x direction
foot_width  = conf.lyn + conf.lyp # foot size in y direction
nb_dt_per_step = int(conf.T_step/conf.dt_mpc)
N = conf.nb_steps * nb_dt_per_step # nb of time steps
```



```
# CoM initial state: [x_0, xdot_0].T
#                       [y_0, ydot_0].T
# -----
x_0 = np.array([conf.foot_step_0[0], 0.0])
y_0 = np.array([conf.foot_step_0[1], 0.0])
```

Code

```
# compute foot steps
foot_steps = manual_foot_placement(conf.foot_step_0,
                                   conf.step_length, conf.nb_steps)
foot_steps[1:,0] -= conf.step_length
```

Code

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# compute foot steps
foot_steps = manual_foot_placement(conf.foot_step_0,
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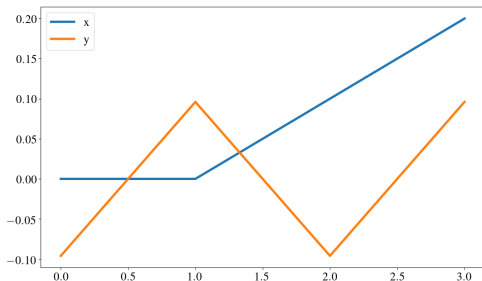


Figure 1: Foot steps.

Code

```
# compute CoP reference trajectory:  
cop_ref = create_CoP_trajectory(conf.nb_steps,  
                                foot_steps, N, nb_dt_per_step)
```

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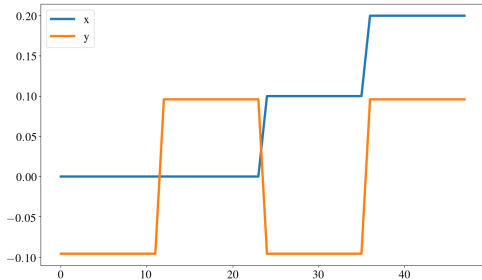


Figure 2: Foot steps and CoP.

```
cd orc/lipm  
python3 lipm_ocp.py
```

Connection with TSID

Two issues:

1. Different time steps
2. Foot trajectories

Interpolation

Input: CoM (pos, vel) and CoP trajectories with traj-opt (large) time step.

Output: CoM (pos, vel, acc) with control (small) time step.

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Compute pos-vel with:

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Compute acc with:

$$\ddot{c} = \frac{g^z}{c^z} (c - z) \quad (10)$$

Common choice: **polynomials**.

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For instance: 3rd order with constraints:

- initial pos
- initial vel (zero)
- final pos
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For instance: 3rd order with constraints:

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- initial vel (zero)
- final pos
- final vel (zero)

Use higher order if you wanna add constraints (e.g., zero initial/final acc).

Run script

```
cd orc/lipm
python lipm_to_tsid.py
cd orc/reactive_control/tsid
python ex_4_walking.py
```

References



Herdt, Andrei et al. (2010). “Online Walking Motion Generation with Automatic Foot Step Placement”. In: *Advanced Robotics* 24.5-6.



Wieber, Pierre-Brice, Russ Tedrake, and Scott Kuindersma (2015). “Modeling and Control of Legged Robots”. In: *Springer Handbook of Robotics*. Ed. by Bruno Siciliano and Khatib Oussama. 2nd. Chap. 48.