



Contact-Implicit Trajectory Optimization for Dynamic Object Manipulation

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Introduction

There has been a shift in attention from industrial robotics towards the development of robots that are capable of a more dexterous interaction with their environment.

Locomotion

Dynamic Manipulation



Dynamic Manipulation

Nonprehensile (graspless) manipulation includes phases where the manipulator loses possibility of contact with the object before task completion





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- Ensure dynamic feasibility and physically consistent contact behavior
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- Solve the optimal control problem efficiently for hopes of applying it within an MPC framework
- Ensure dynamic feasibility and physically consistent contact behavior
- Avoid specifying a hand-crafted contact schedule
- Ensure ease of transferability onto real hardware without any post-optimization modifications





Optimal Control Problem

Trajectory optimization techniques aim to solve the following nonlinear optimal control (NLOC) problem:

$$\begin{cases} \min_{\boldsymbol{\tau}(\cdot)} \quad J(\boldsymbol{x}(t), \boldsymbol{\tau}(t), t) = m(\boldsymbol{x}(T)) + \int_{0}^{T} L(\boldsymbol{\tau}(t), \boldsymbol{x}(t)) dt \\ \boldsymbol{s.t.} \quad \dot{\boldsymbol{x}}(t) = \boldsymbol{f}(\boldsymbol{x}(t), \boldsymbol{\tau}(t)) & \text{(nonlinear, smooth ODEs)} \\ \boldsymbol{h}_{min} \leq \boldsymbol{h}(\boldsymbol{x}(t), \boldsymbol{\tau}(t)) \leq \boldsymbol{h}_{max} & \text{(inequality constraints)} \\ \boldsymbol{g}(\boldsymbol{x}(t), \boldsymbol{\tau}(t)) = 0 & \text{(equality constraints)} \\ \forall t \in [0, T] \end{cases}$$

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 How to incorporate the multi-contact behavior into the dynamics?



Non-Smooth Contact Dynamics

(a)

Modeling

A reliable model describing multi-body contacts has to be chosen before simulating the system dynamics



(b)

Contact dynamics can be modeled using either (a) a hard contact-model or (b) a soft contact-model

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Optimal Control Problem

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Non-Smooth Contact Dynamics

Simulation

Two prominent schemes used for simulating hybrid dynamical systems are: **event-driven** and **time-stepping** techniques

Courtesy of Blender Guru and Phymec

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The forward simulation of multi-contact dynamics can be eventually formulated as a so-called **linear complementarity problem (LCP)**



Non-Smooth Contact Dynamics

Trajectory Optimization

Unlike the **multi-phase approach**, **contact-implicit optimization (CIO)** does not require a prespecified contact schedule!





Multiple Shooting

Transcription Process

- Integral expressions replaced with Riemann sums, defines the cost function
- System dynamics integrated with a Runge-Kutta scheme, defines the state transition function

Multi-staged Program

$$\begin{split} f_{\mathbf{z}_1,...,\mathbf{z}_N} & F_N(\mathbf{z}_N) + \sum_{k=1}^{N-1} F(\mathbf{z}_k) \\ s.t & \mathbf{E}_k \mathbf{z}_{k+1} = \mathbf{c}(\mathbf{z}_k) \qquad \forall \ k = 1,...,N-1 \\ & \mathbf{S}_1 \mathbf{z}_1 = \mathbf{z}_{init} \\ & \mathbf{S}_N \ \mathbf{z}_N = \mathbf{z}_{final} \\ & \mathbf{z}_k \leq \mathbf{z}_k \leq \bar{\mathbf{z}}_k \qquad \forall \ k = 1,...,N \\ & \mathbf{\underline{h}}_k \leq \mathbf{h}(\mathbf{z}_k) \leq \bar{\mathbf{h}}_k \qquad \forall \ k = 1,...,N \end{split}$$







The cost function minimizes energy and penalizes the robot's wrist joint-velocities

$$J(oldsymbol{z}_k) = \sum_{k=1}^N \Delta t \cdot \left(rac{oldsymbol{ au}_k^T oldsymbol{R} oldsymbol{ au}_k}{ au_{max}^2} + rac{\dot{oldsymbol{q}}_{r_k}^T oldsymbol{Q} \dot{oldsymbol{q}}_{r_k}}{\dot{oldsymbol{q}}_{max}^2}
ight)$$

- The manipulation task is specified in terms of final boundary conditions in the optimization program
- The signed distance function \u03c6(q) is defined such that a negative distance indicates that the end-effector is penetrating the object





- Normal contact force λ_N , contact impulse Λ_N , and percussion P_N between robot and object
- Frictional force λ_F , frictional impulse Λ_F , and percussion P_F between environment and object
- Variables λ_n and λ_f defined in relation to the percussions: $P_N = \Delta t \cdot \lambda_n \& P_F = \Delta t \cdot \lambda_f$

$$\begin{array}{l} \textbf{Underactuated Dynamics \& Impulse-Momentum Equations} \\ \left\{ \begin{array}{l} M(q) \ddot{q} + h(q, \dot{q}) = S^T \tau + J^T(q) \cdot \begin{bmatrix} \lambda_N \\ \lambda_F \end{bmatrix} \\ M(q) (\dot{q}^+ - \dot{q}^-) = J^T(q) \cdot \begin{bmatrix} \Lambda_N \\ \Lambda_F \end{bmatrix} \\ \downarrow \\ \left\{ \begin{array}{l} \dot{q}_{k+1} = \dot{q}_k + \Delta t \cdot M_k^{-1} \left(S^T \tau_k - h_k + J_k^T \cdot \begin{bmatrix} \lambda_{n_k} \\ \lambda_{f_k} \end{bmatrix} \right) \\ q_{k+1} = q_k + \Delta t \cdot \dot{q}_{k+1} \end{array} \right\}$$

Complementarity Constraints

$$0 \le \phi(q_k) \perp \lambda_{n_k} \ge 0$$
 $(no penetration)$
 0
 $(unilateral cont)$

(no force at a distance)



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- Positive scalar quantity v_d signifying the speed along the desired direction of motion
- ▶ Pre- and post-impact relative separation velocities γ^- / γ^+
- Newton's restitution coefficient ϵ

Newton's Restitution Law of ImpactsCoulomb Friction Model
$$\phi(q) = 0:$$
 $0 \le P_N \perp (\gamma^+ + \epsilon \gamma^-) \ge 0$ $\begin{cases} \lambda_f = \lambda_n & \text{if } v_d = 0 \& \lambda_n < F_s \\ \lambda_f = F_s & \text{if } v_d > 0 \end{cases}$ $P_{N_k} \cdot (\gamma_{k+1} + \epsilon \gamma_k) = 0$ \uparrow \uparrow $\Delta t \cdot \lambda_{n_k} \cdot (\dot{\phi}_{k+1} + \epsilon \dot{\phi}_k) = 0$ \uparrow $\Delta t \cdot \lambda_{n_k} \cdot (J_N(q_{k+1})\dot{q}_{k+1} + \epsilon J_N(q_k)\dot{q}_k) = 0$ $\begin{pmatrix} 0 \le v_{dir} \perp (F_s - F_f) \ge 0 \\ (F_s - F_f)(\lambda - F_f) = 0 \end{pmatrix}$



Without Newton's Restitution Law

With Newton's Restitution Law







- Positive scalar quantity v_d signifying the speed along the desired direction of motion
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Open-loop The optimal input sequence τ_{opt_k} is an open-loop control Contact-Implicit policy. Therefore, trajectory stabilization is still needed. Optimization τ_{opt} q_{opt} $\tau_{i}(t) = \underbrace{\tau_{opt_{i}}(t)}_{(t)} + \underbrace{k_{p_{i}}(q_{opt_{i}}(t) - q_{i}(t)) + k_{d_{i}}(\dot{q}_{opt_{i}}(t) - \dot{q}_{i}(t))}_{(t)}$ \dot{q}_{opt} feedback term feedforward term Feedforward Term & where $\tau_{opt}(t)$ is obtained by a zero-order-hold while $q_{opt}(t)$ Linear-Time-Invariant Feedback Law and $\dot{q}_{opt}(t)$ by linear interpolation



Robot-Door Manipulation Task

Robot-Ball Manipulation Task







Robot-Block Manipulation Task

Compatibility between optimal contact force and measured one



Instantaneous Normal Percussions for Experiment 3



Robot-Block Manipulation Task

- Compatibility between optimal contact force and ► measured one
- Convergence rate two orders of magnitude higher than that of previous works





Robot-Block Manipulation Task

- Compatibility between optimal contact force and measured one
- Convergence rate two orders of magnitude higher than that of previous works
- Satisfaction of manipulation task with inaccuracy proportional to desired displacement







Thank you for your attention!







Any Questions?