

Case Studies Nonlinear Optimization

# Open Cast Mining

Final Presentation

July 09, 2016

InYoung Choi, Olivia Kaufmann, Martin Sperr, Florian Wuttke

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- 1 Project Overview
- 2 Physical Model
- 3 Parameter Identification: Physical Model
- 4 Parameter Identification: Black Box Model
- 5 Summary

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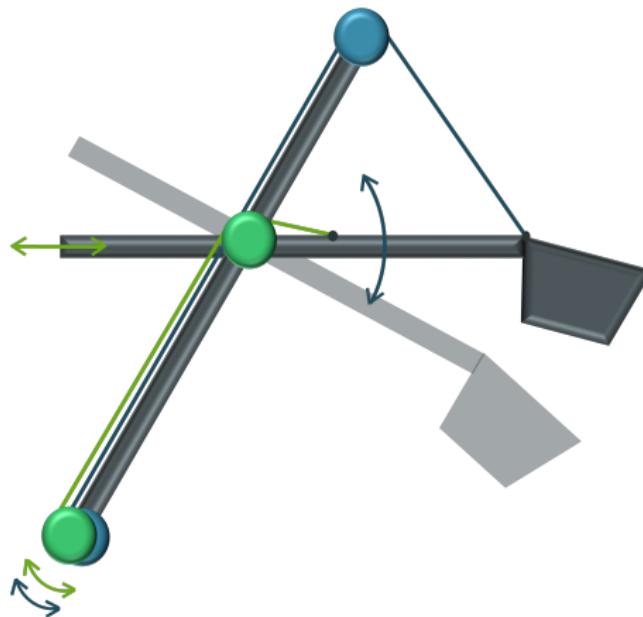
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originally posted to Flickr by FAndrey at <http://flickr.com/photos/43301444@N06/4141786255>

- Goal: **Optimization of model parameters**
- Models of technical system = Physical properties + Control properties

## Problem Setting: Schematic Illustration



# Problem Setting

# Procedure

## Physical Model

- Rope properties
- Lagrange Formalism



## Parameter Identification

## Discretization + Optimization

# Procedure

## Physical Model

- Rope properties
- Lagrange Formalism

## Blackbox Model

- Realistic model
- Confidential information

and

## Parameter Identification

Discretization + Optimization



## Parameter Identification

Derivative-free optimization

# Physical Modeling

## Why?

Building an  
accurate model



Good description of  
the effects of control  
and motion

## Physical Modeling Cont'd

### How?

To consider:

- Friction in cable reels
- Potential/Kinetic energy
- etc

# Physical Modeling Cont'd

## How?

To consider:

- Friction in cable reels
- Potential/Kinetic energy
- etc



Lagrange Formalism  
(ODE)

## Which color do we want to use?

TUMblue

TUMblue1

TUMblue2

TUMblue3

TUMblue4

TUMblue5

# Parameter Identification

## What are parameters?

- Friction coefficients
- Mass
- Inertia

# Parameter Identification

## What are parameters?

- Friction coefficients
- Mass
- Inertia

## Why?

Accurate and  
realistic parameters



Better prediction  
and planning of  
motion

# Parameter Identification Two Ways

**Two independent models acquired,  
two different computational approaches.**

Physical Model



Parameter Identification

and

Blackbox Model



Parameter Identification

# Parameter Identification: Physical Model

Own Physical Model

Lagrange Formalism



Parameter Identification

Discretization

$$\min_p \quad \frac{1}{2} \|\bar{x} - x(p)\|^2$$

# Parameter Identification: Physical Model

Own Physical Model

Lagrange Formalism



Parameter Identification

Discretization

$$\min_p \quad \frac{1}{2} \|\bar{x} - x(p)\|^2$$

- Information for calibration available
- Runge-Kutta method
- Parameter computation for a given control and motion

# Parameter Identification: Blackbox Model

Siemens Blackbox Model

- **Input:** Control
- **Output:** Motion



Parameter Identification

Particle Swarm, Pattern Search,  
Genetic Algorithm, ...

# Parameter Identification: Blackbox Model

- Almost no information available
- Derivative-free optimization
- Only a few parameters studied

Siemens Blackbox Model

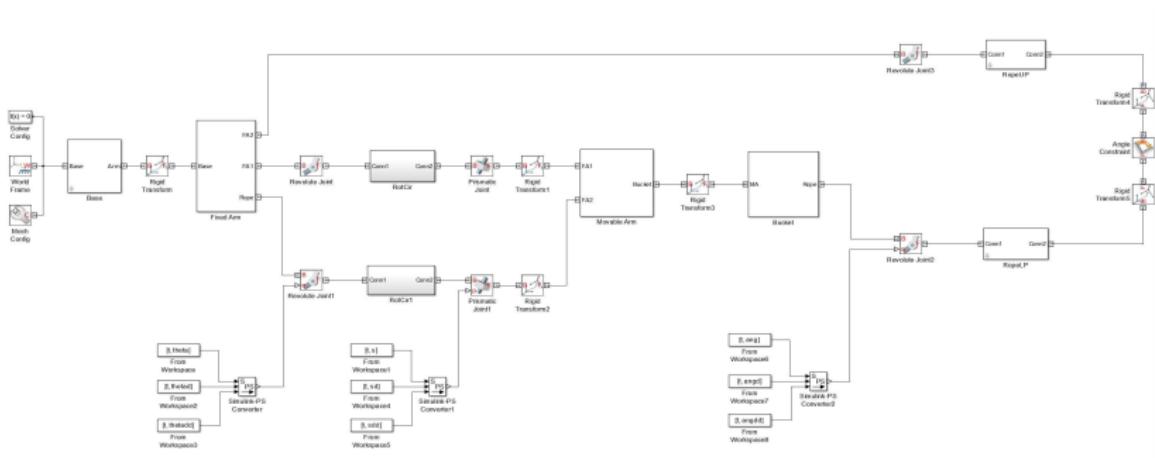
- **Input:** Control
- **Output:** Motion



Parameter Identification

Particle Swarm, Pattern Search,  
Genetic Algorithm, ...

# Visualization: Simulink



## Visualization Example

Same parameters, different load weights

**vs**

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# Lagrange Formalism

Method to describe dynamics of an accelerated system

T    kinetic energy  
V    potentials

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- T    kinetic energy
- V    potenitals
- F    non-conservative external forces
- r    points of actions of forces F
- q    free variables
- Q    generalized forces

# Lagrange Formalism

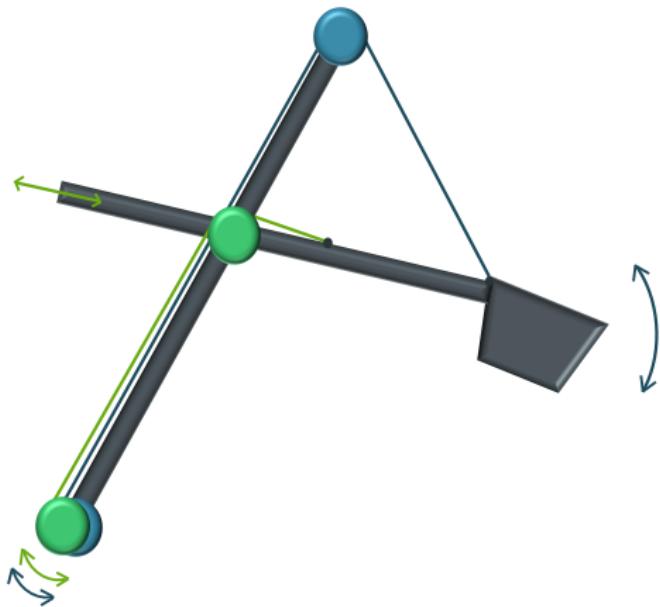
Method to describe dynamics of an accelerated system

$$\frac{d}{dt} \left( \frac{\partial T}{\partial \dot{q}_i} \right) - \frac{\partial T}{\partial q_i} + \frac{\partial V}{\partial q_i} = Q_i$$

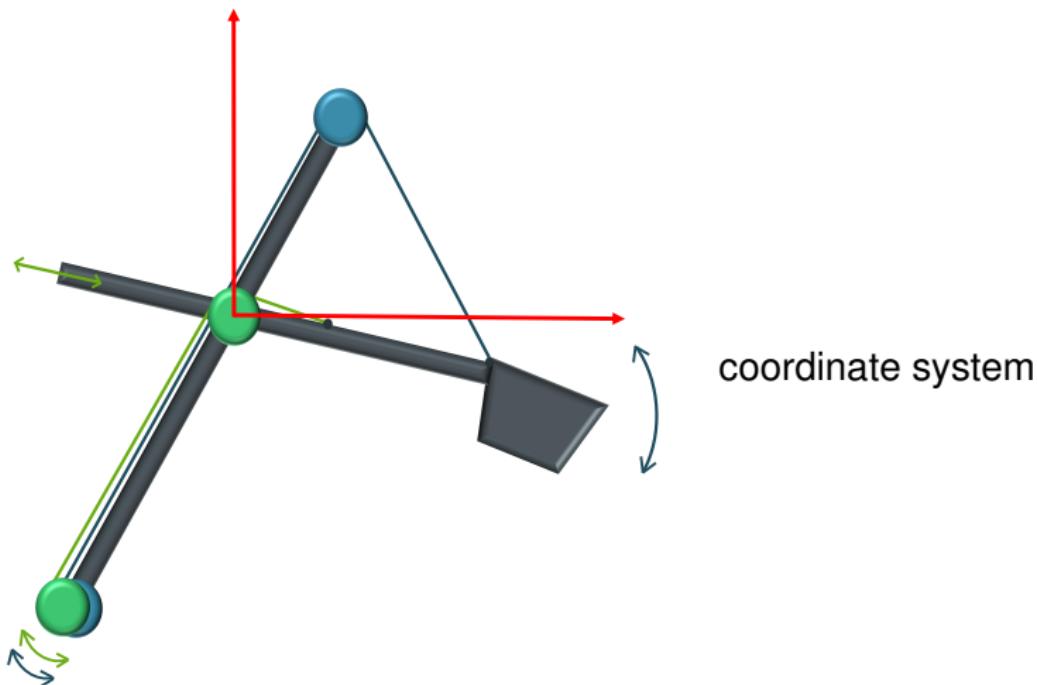
$$Q = \left( \frac{\partial r}{\partial q} \right)^T F$$

- T kinetic energy
- V potentials
- F non-conservative external forces
- r points of actions of forces F
- q free variables
- Q generalized forces

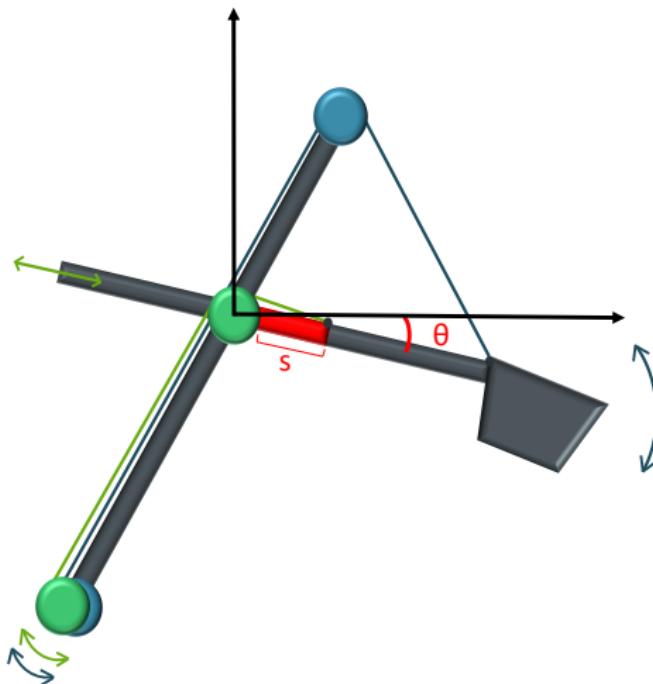
# Physical Model of Excavator



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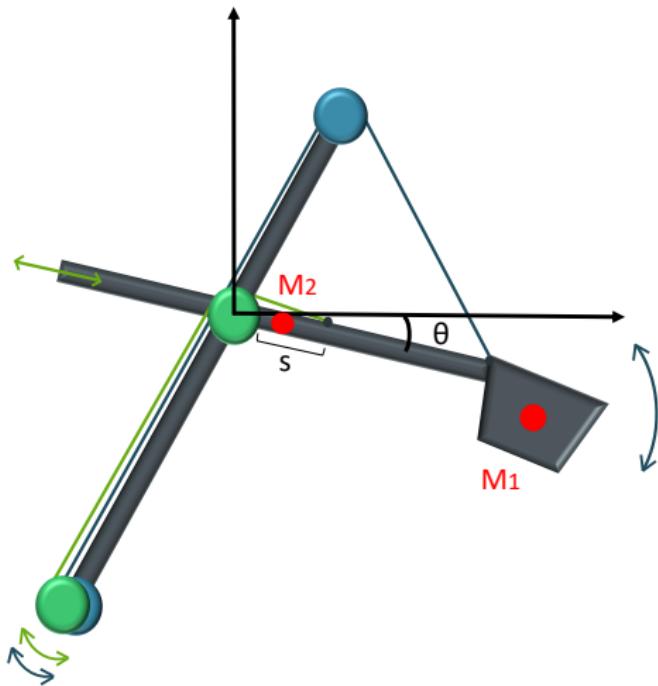
# Physical Model of Excavator



degrees of freedom

- length  $s$
- tilt angle  $\theta$

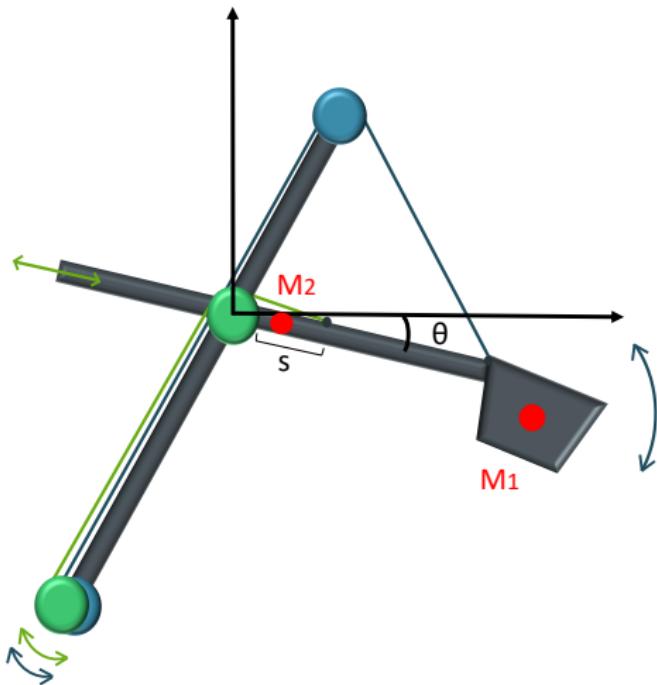
# Physical Model of Excavator



movable centers of gravity of

- shovel  $M_1$
- arm  $M_2$

# Physical Model of Excavator



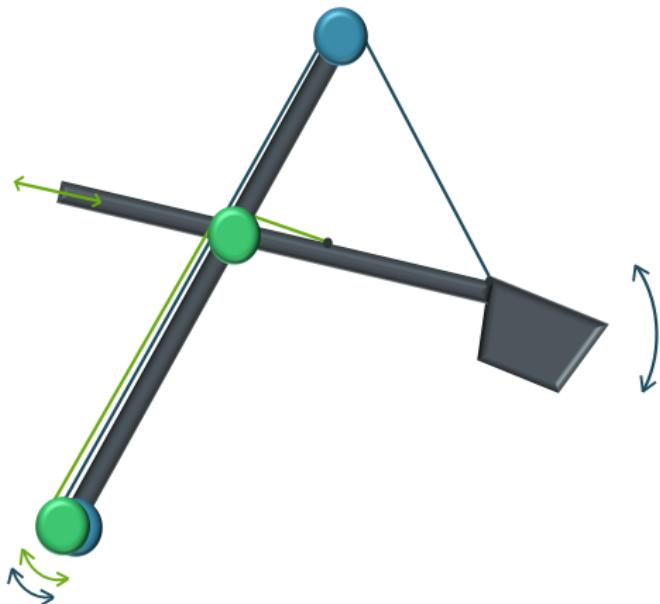
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- shovel  $M_1$
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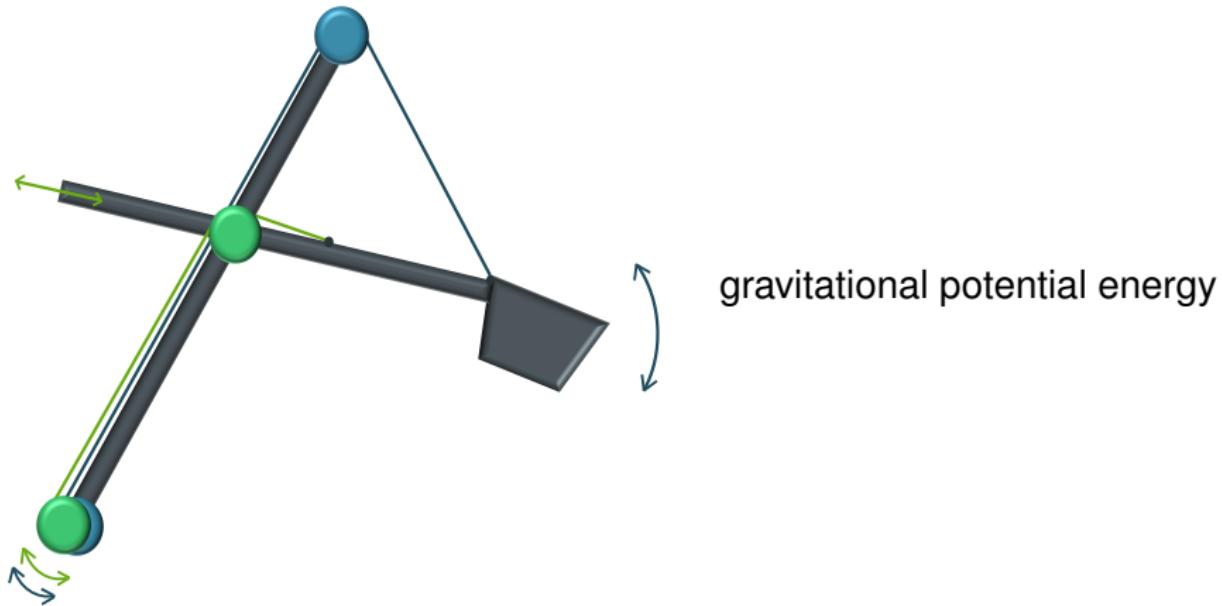
$$Q = \left( \frac{\partial r}{\partial q} \right)^T F$$

# Kinetic Energy

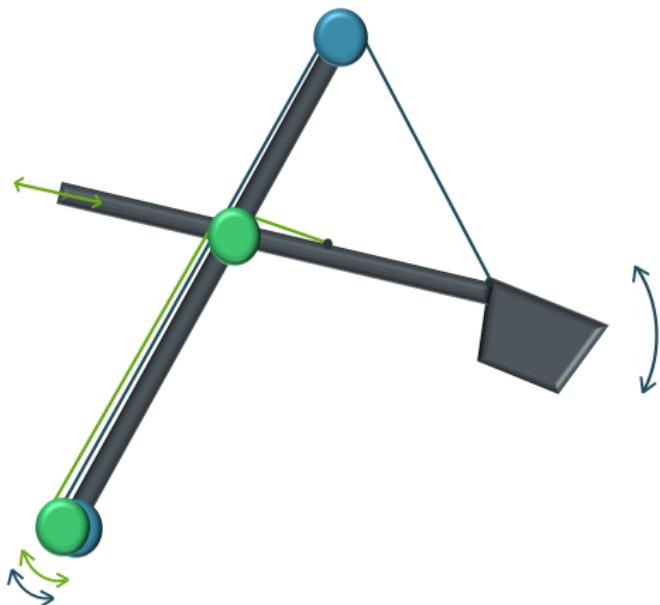


- movement of mass
- rotation of cable reel

# Potential Energy



# Generalized Forces



- torque on cable reels
- friction of cable reels

# Physical Model of Excavator

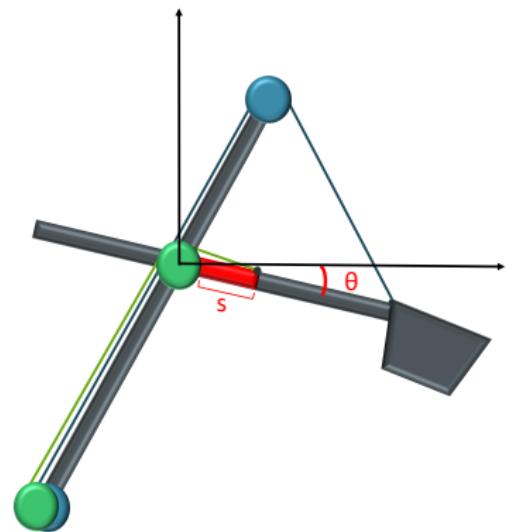
Assumptions to the model:

- no mass for the ropes
- shovel as point mass
- no slack / friction between ropes and cable reels

# Lagrange Formalism

$$\frac{d}{dt} \left( \frac{\partial T}{\partial \dot{s}} \right) - \frac{\partial T}{\partial s} + \frac{\partial V}{\partial s} = Q_s$$

$$\frac{d}{dt} \left( \frac{\partial T}{\partial \dot{\theta}} \right) - \frac{\partial T}{\partial \theta} + \frac{\partial V}{\partial \theta} = Q_\theta$$



# Resulting ODE

$$A(x, p) \begin{pmatrix} \ddot{s} \\ \ddot{\theta} \end{pmatrix} = b(x, u, p)$$

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$$A(x, p) \begin{pmatrix} \ddot{s} \\ \ddot{\theta} \end{pmatrix} = b(x, u, p)$$

→ Transformation into 1st order ODE

$$\frac{d}{dt} \begin{pmatrix} s \\ \theta \\ \dot{s} \\ \dot{\theta} \end{pmatrix} = \begin{pmatrix} \dot{s} \\ \dot{\theta} \\ A^{-1}(x, p)b(x, u, p) \end{pmatrix} = f(x, u, p)$$

state	$x = (s, \theta, \dot{s}, \dot{\theta})^T$
control	$u = (\tau_1, \tau_2)^T$
parameters	$p = (p_1, \dots, p_k)^T$

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## Discretization of the ODE

Discretize time interval:

$$[0, T] \rightarrow \{0 = t_0, t_1, \dots, t_{m-1}, t_m = T\}$$

Discretize state and control:

$$x_n = x(t_n)$$

$$u_n = u(t_n)$$

## Discretization of the ODE

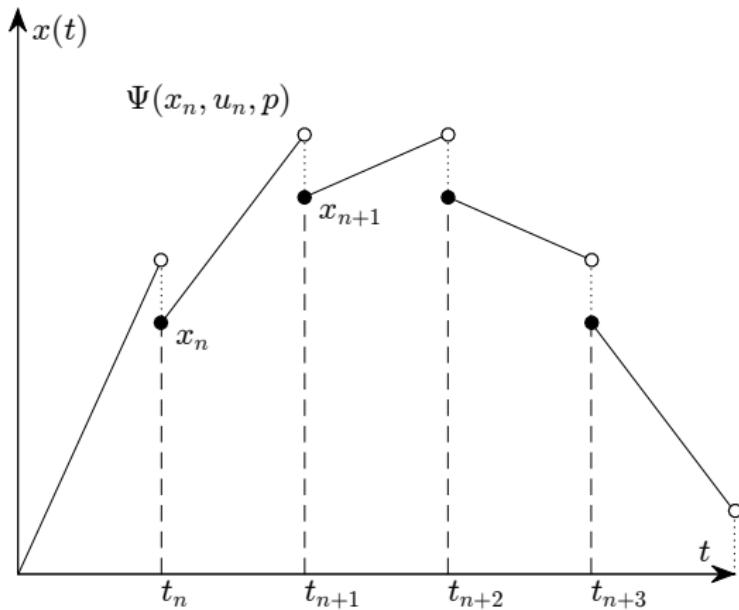
Explicit Euler for every time step  $h_n = t_{n+1} - t_n$ :

$$\Psi(x_n, u_n, p) = x_n + h_n f(x_n, u_n, p)$$

Discrete constraint:

$$0 = x_{n+1} - \Psi(x_n, u_n, p) =: \Phi_n(x, u, p) \quad \forall n = 0, \dots, m-1$$

# Discretization of the ODE



# Problem Setting

Given:

- control  $\bar{u}$
- motion  $\bar{x}$  related to  $\bar{u}$  and  $\bar{p}$

Unknown:

- parameters  $\bar{p}$  of the excavator

Output:

- parameters  $p$

## Problem Formulation

Natural approach: Optimal Control Problem

$$\begin{aligned} \min_{x,p} \quad & \frac{1}{2} \|\bar{x} - x\|^2 \\ \text{s. t.} \quad & \Phi(x, \bar{u}, p) = 0 \\ & p \geq 0 \end{aligned}$$

state	$x = (s, \theta, \dot{s}, \dot{\theta})^T$
parameters	$p = (p_1, \dots, p_k)^T$
control	$\bar{u} = (\tau_1, \tau_2)^T$
desired motion	$\bar{x}$

# Problem Formulation

## Original Problem

$$\begin{aligned} \min_{x,p} \quad & \frac{1}{2} \|\bar{x} - x\|^2 \\ \text{s. t.} \quad & \Phi(x, \bar{u}, p) = 0 \\ & p \geq 0 \end{aligned}$$

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Set  $x \leftarrow \bar{x}$

# Problem Formulation

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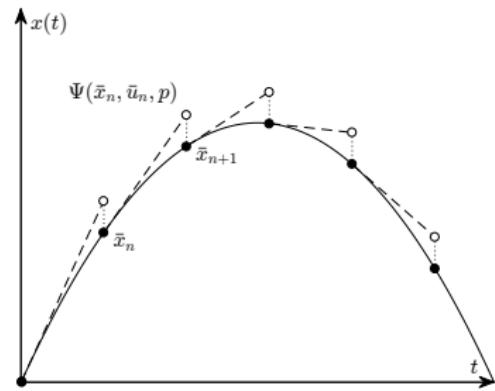
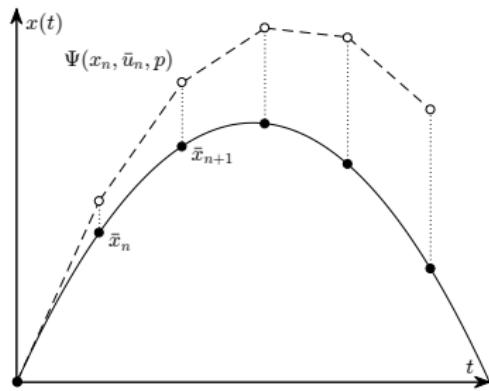
Reinterpreted Problem

$$\begin{aligned} \min_p \quad & \frac{1}{2} \|\Phi(\bar{x}, \bar{u}, p)\|^2 \\ \text{s. t.} \quad & p \geq 0 \end{aligned}$$

$$\begin{aligned} \min_p \quad & \frac{1}{2} \|\Phi(\bar{x}, \bar{u}, p)\|^2 \\ \text{s. t.} \quad & p \geq 0 \end{aligned}$$

- $\bar{x}$  solves ODE for  $\bar{p}$
- $\Phi(\bar{x}, \bar{u}, \bar{p}) \rightarrow 0$  for discretization  $m \rightarrow \infty$
- number of parameters fix

# Comparison of the Approaches

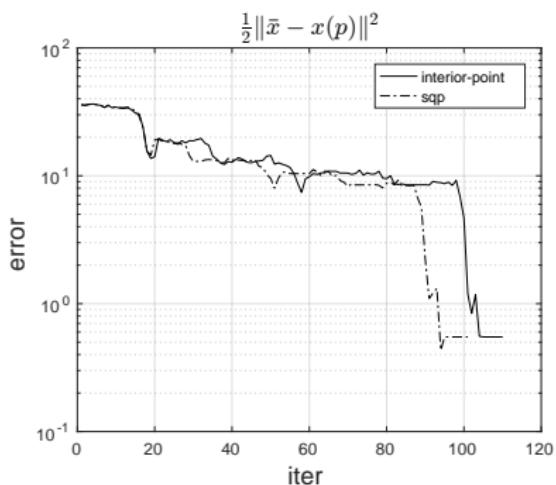
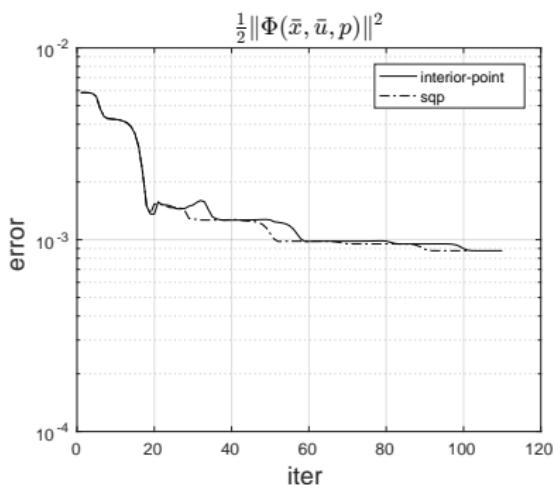


continuous vs. stepwise Approximation

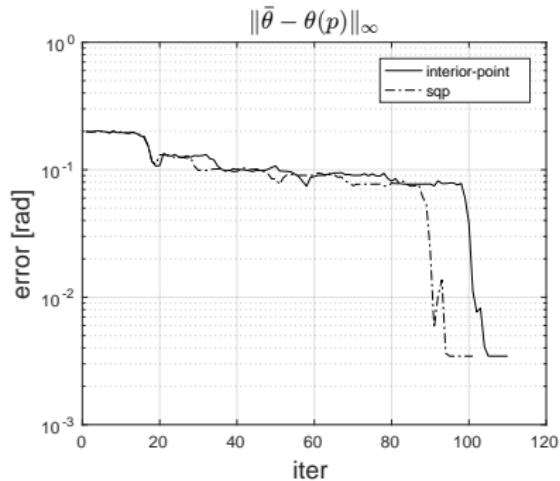
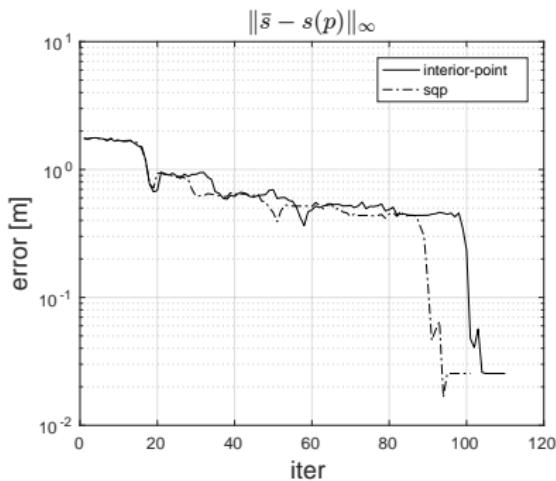
## Example Instance

- $[0, T] = [0, 14s]$
- 1500 time steps
- $p_0 \in [0.8\bar{p}, 1.2\bar{p}]$
- $x(p)$  solution of ODE for given  $p$
- internally 5 trajectories in parallel

# Results

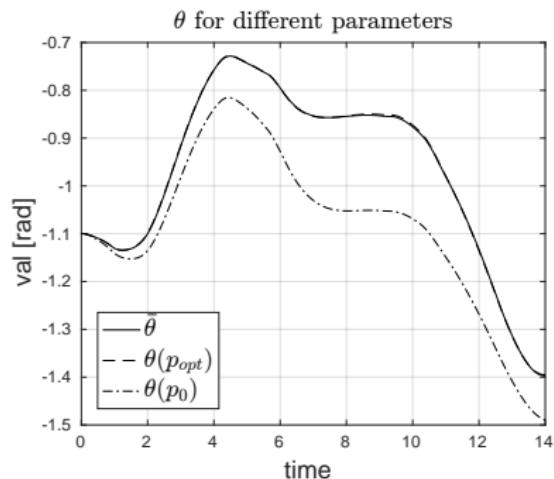
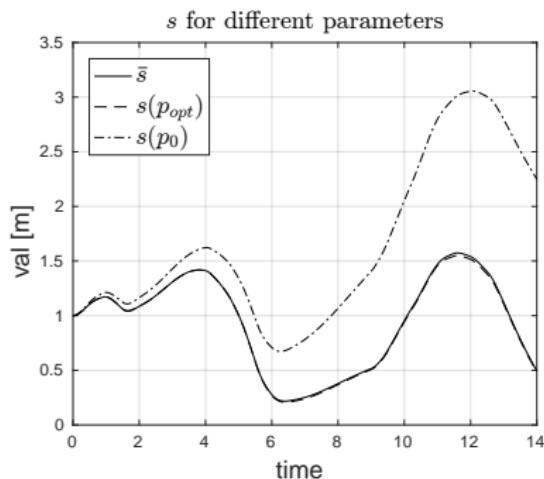


# Results



Exact up to 3cm

# Results



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## Examples

- Friction coefficients
- Masses
- Inertia

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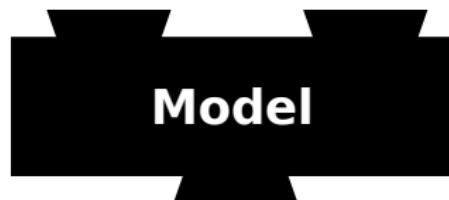
## Black box model

- Realistic model from Siemens
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**Control**



**Model**

## Black box model

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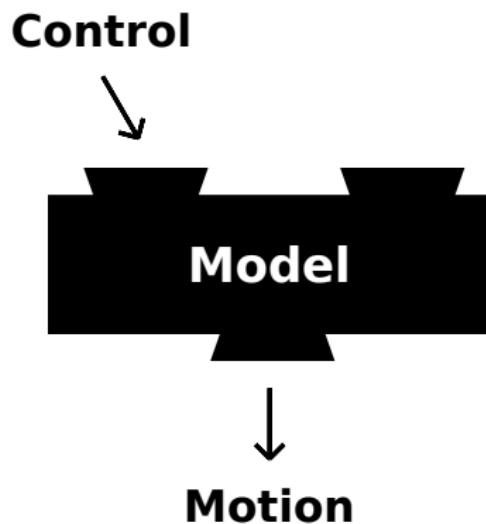
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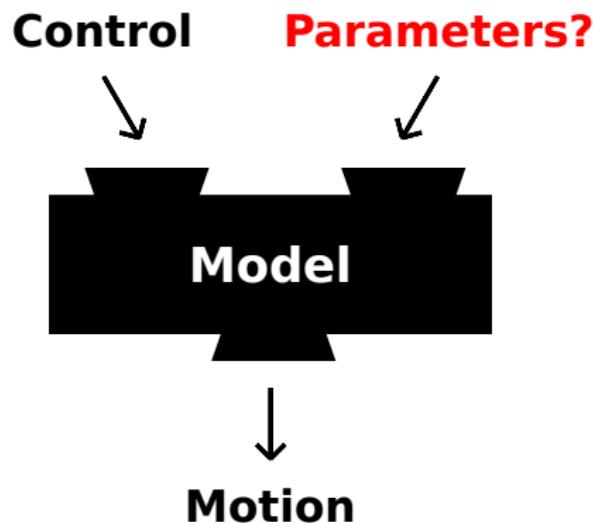
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# Trajectories

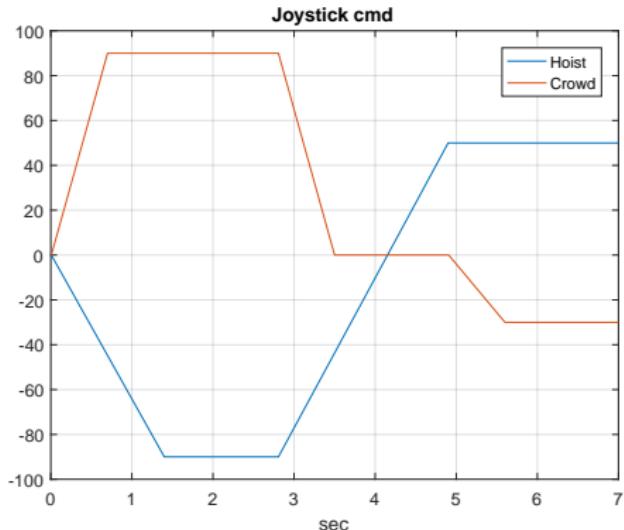
**Input:** Joystick commands for Up/Down and Forth/Back

**Output:** Position of the shovel

# Trajectories

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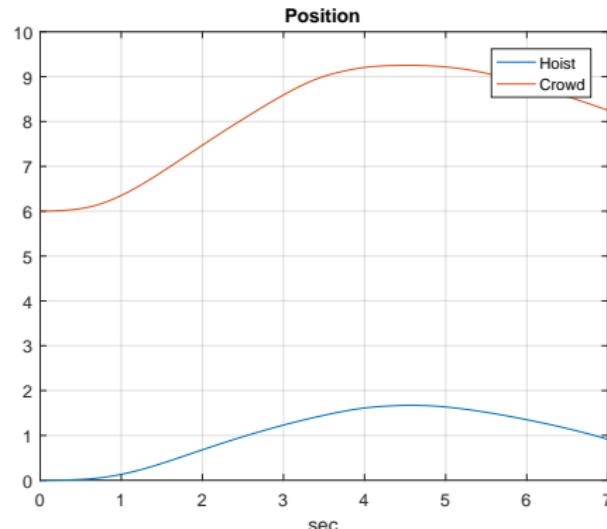
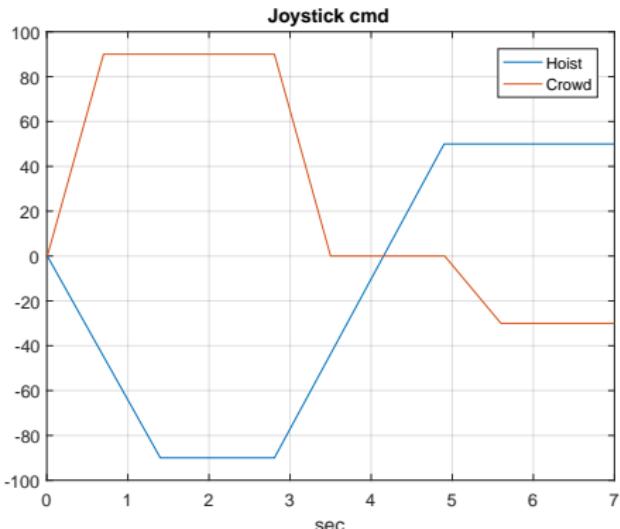
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# Trajectories

**Input:** Joystick commands for Up/Down and Forth/Back

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# Objective Function

## Optimized Parameters:

- Inertia (Engine)
- Inertia (Arm)
- Friction
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## Penalty Term:

$$\|\bar{X}_i - X_i(p)\|^2$$

# Objective Function

## Optimized Parameters:

- Inertia (Engine)
- Inertia (Arm)
- Friction
- Mass

## Penalty Term:

$$\frac{\|\bar{X}_i - X_i(p)\|^2}{\|\bar{X}_i\|^2}$$

# Objective Function

## Optimized Parameters:

- Inertia (Engine)
- Inertia (Arm)
- Friction
- Mass

## Penalty Term:

$$\frac{\|\bar{X}_i - X_i(p)\|^2}{\|\bar{X}_i\|^2} + \frac{\|\bar{Y}_i - Y_i(p)\|^2}{\|\bar{Y}_i\|^2}$$

$\bar{X}_i, \bar{Y}_i$  reference trajectories

# Objective Function

## Optimized Parameters:

- Inertia (Engine)
- Inertia (Arm)
- Friction
- Mass

## Penalty Term:

$$\frac{1}{n} \cdot \sum_{i=1}^n \left( \frac{\|\bar{X}_i - X_i(p)\|^2}{\|\bar{X}_i\|^2} + \frac{\|\bar{Y}_i - Y_i(p)\|^2}{\|\bar{Y}_i\|^2} \right)$$

$\bar{X}_i, \bar{Y}_i$  reference trajectories

# Objective Function

## Optimized Parameters:

- Inertia (Engine)
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- Friction
- Mass

## Penalty Term:

$$\min_{p \in \mathbb{R}^4} f(p) = \frac{1}{n} \cdot \sum_{i=1}^n \left( \frac{\|\bar{X}_i - X_i(p)\|^2}{\|\bar{X}_i\|^2} + \frac{\|\bar{Y}_i - Y_i(p)\|^2}{\|\bar{Y}_i\|^2} \right)$$

s. t.  $p_j \geq 0$

$\bar{X}_i, \bar{Y}_i$  reference trajectories

# Influence of the Parameters

10% parameter deviation:

- Inertia (Engine):  $1 \cdot 10^{-3}$
- Inertia (Arm):  $3 \cdot 10^{-3}$
- Friction:  $8 \cdot 10^{-11}$
- Mass:  $5 \cdot 10^{-2}$

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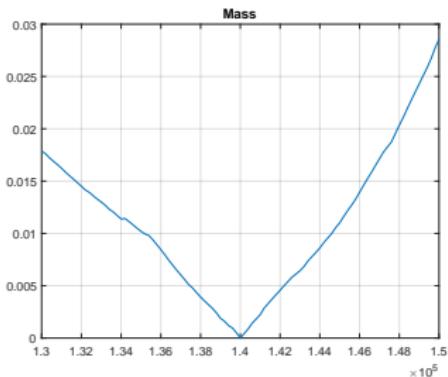
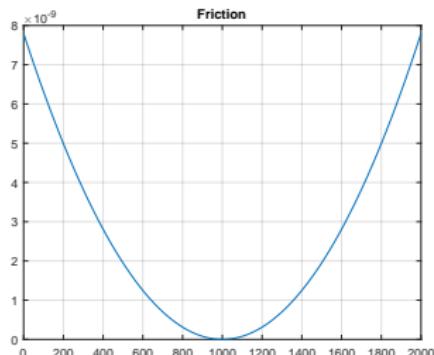
Big parameter changes  $\Rightarrow$  Small effects

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# Optimizers

- Derivative free optimization
- Deterministic or stochastic
- Decrease function value by evaluating systematically

# Optimizers

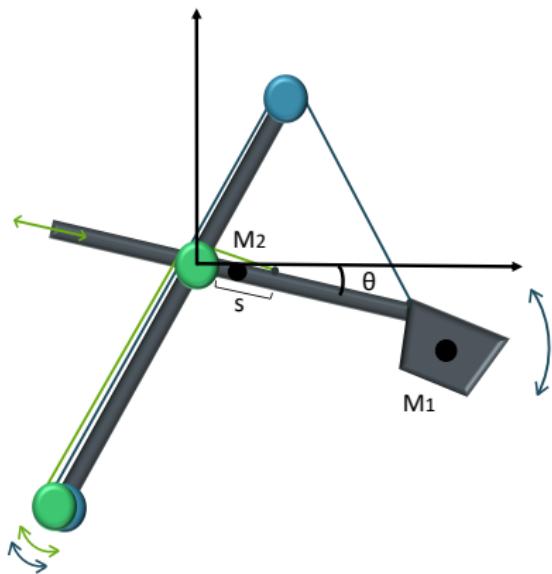
- Derivative free optimization
- Deterministic or stochastic
- Decrease function value by evaluating systematically

	value	evaluations	time	$\text{dev}_{\max}$	$\text{dev}_{\text{mean}}$
Particle Swarm	$10^{-12}$	2200	6 min	$10^{-1.0}$	$10^{-3.8}$
Pattern Search	$10^{-11}$	6500	17 min	$10^{-1.3}$	$10^{-3.7}$
Genetic Algorithm	$10^{-4}$	5500	14 min	$10^{-0.5}$	$10^{-1.0}$
Simulated Annealing	$10^{-3}$	4200	11 min	$10^{-0.3}$	$10^{-0.6}$

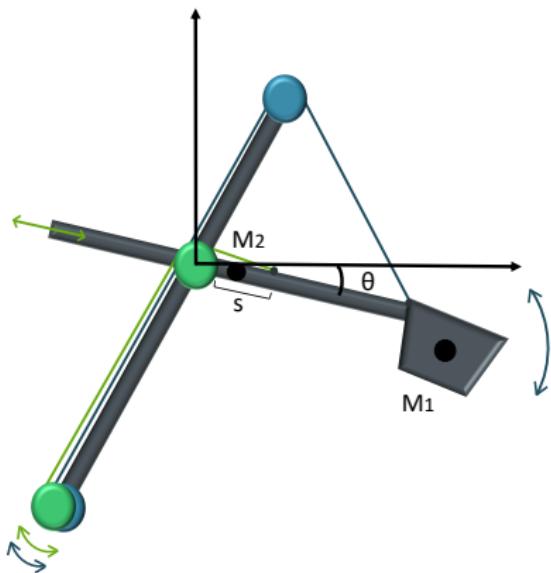
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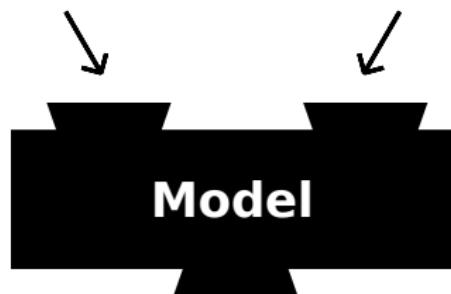


# Summary



**Control**

**Parameters?**



**Model**

**Motion**