## Switched Systems Optimal Control Problems

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#### Overview

- (1) Introduction
- (2) Problem formulation
- (3) SSOCP with fixed sequence of phases
- (4) SSOCP with unknown sequence of phases





#### Introduction

#### Switched System

A dynamics system that operates by switching between different subsystems or phases.

#### Switched System Optimal Control Problem

Problem of designing an optimal sequence of phases and optimal control signals for each phase, such that certain cost function is minimized.

Applications:

- Aircraft modelling
- Air traffic control
- Robotics and industrial processes
- Logistics





# Introduction

## **Example Problem**

#### Interaction between quadcopter and robot.







# Introduction

## Example

Quadcopter flying over several industrial robots







#### Problem Formulation

Given *M* phase intervals  $[T_0, T_1], [T_1, T_2], \ldots, [T_{M-1}, T_M]$  with length  $p_1, p_2, \ldots, p_M$ , we consider the following Switched System Optimal Control Problem

#### Switched System Optimal Control Problem

For each phase  $[T_{k-1}, T_k]$ 

minimize 
$$\varphi^k(x(T_{k-1}), x(T_k))$$
 (1)

with respect to  $x \in W^{1,\infty}([T_{k-1}, T_k]; \mathbb{R}^{n_x}), u \in L^{\infty}([T_{k-1}, T_k]; \mathbb{R}^{n_u^k})$  and  $p_k \in \mathbb{R}$ , subject to

$$\dot{x}(t) - f^{k}(t, x(t), u(t)) = 0_{\mathbb{R}^{n_{x}}}$$
 a.a.  $t \in (T_{k-1}, T_{k})$  (2)

$$g^{k}(x(t), u(t)) \leq \underset{\mathbb{R}}{0} g^{n_{g}^{k}} \quad \forall t \in (T_{k-1}, T_{k})$$
(3)

$$\phi^{k}(x(T_{k-1}), x(T_{k})) = 0_{\mathbb{R}^{n_{\phi}^{k}}}$$

$$\tag{4}$$



## **Problem Formulation**

Two cases may occur:

- 1) The sequence of phases is fixed:
  - SSOCP can be transformed to standard optimal control problem
  - Solvable by gradient type methods (SQP, Interior point, Quasi-Newton)
  - Easier to implement and solve
- 2) The sequence of phases is unknown:
  - Optimal Control Problem for each phase
  - · Feasible and infeasible sequences of phases may arise
  - Two phases may not occur simultaneously
  - Need of binary (decision) variables
  - Mixed-Integer solvers have to be used





Since the length of the time intervals  $[T_{k-1}, T_k]$  is unknown, for each  $k \in \{1, ..., M\}$  we consider the time transformation

$$t^{(k)}:[0,1]\to[T_{k-1},T_k]$$

defined as

$$t^{(k)}(\tau) = T_{k-1} + \tau \cdot (T_k - T_{k-1}) = T_{k-1} + \tau \cdot p_k.$$
(5)

Note that the transformation  $t^{(k)}(\cdot)$  is continuous and it holds

$$\frac{dt^{(k)}(\tau)}{d\tau} = T_k - T_{k-1} = p_k \quad \forall \ \tau \in (0, 1)$$
(6)









Let us define the new dimensions

$$N_x = M \cdot n_x$$
 and  $N_u = \sum_{k=1}^M n_u^k$  (7)

and the new states and controls

$$x \in W^{1,\infty}([0,1]; \mathbb{R}^{N_x})$$
  $x = (x^{(1)}, \dots, x^{(M)})$   $x^{(k)} \in W^{1,\infty}([0,1]; \mathbb{R}^{n_x})$  (8)

$$u \in L^{\infty}([0,1]; \mathbb{R}^{N_u})$$
  $u = (u^{(1)}, \dots, u^{(M)})$   $u^{(k)} \in L^{\infty}([0,1]; \mathbb{R}^{n_u^k})$  (9)



In the same way, we redefine the functions involved in the Switched System Optimal Control Problem as follows.

# Cost Function Reformulation

Define

$$\varphi: \mathbb{R}^{N_x} \times \mathbb{R}^{N_x} \to \mathbb{R}$$

such that

$$\varphi(x, y) = \sum_{k=1}^{M} \varphi^{k}(x^{(k)}, y^{(k)})$$
(10)

for every  $x = (x^{(1)}, \ldots, x^{(M)})$  and  $y = (y^{(1)}, \ldots, y^{(M)})$  in  $\mathbb{R}^{N_x}$ .





#### **Dynamics Reformulation**

Define

$$f:(0,1)\times\mathbb{R}^{N_x}\times\mathbb{R}^{N_u}\to\mathbb{R}^{N_x},\quad f=(f^{(1)},\ldots,f^{(M)})$$
(11)

where for every  $k = 1, \ldots, M$ , the  $k^{th}$  component

$$f^{(k)}: (0,1) \times \mathbb{R}^{N_x} \times \mathbb{R}^{N_u} \to \mathbb{R}^{n_x}$$
(12)

is defined as

$$f^{(k)}(\tau, x, u) = f^{k}(t^{(k)}(\tau), x^{(k)}, u^{(k)})$$
(13)

for every  $\tau \in (0, 1), x = (x^{(1)}, \dots, x^{(M)}) \in \mathbb{R}^{N_x}$  and  $u = (u^{(1)}, \dots, u^{(M)}) \in \mathbb{R}^{N_u}$ .



#### Constraints Reformulation

Define  $N_g = \sum_{k=1}^{M} n_g^k$  and consider the function

$$g: \mathbb{R}^{n_x} \times \mathbb{R}^{n_u} \to \mathbb{R}^{N_g}, \quad g = (g^{(1)}, \dots, g^{(M)})$$
 (14)

where for every  $k = 1, \ldots, M$ , the  $k^{th}$  component

$$g^{(k)}: \mathbb{R}^{n_x} \times \mathbb{R}^{n_u} \to \mathbb{R}^{n_x^g}$$

is defined as

$$g^{(k)}(x,u) = g^k(x^{(k)}, u^{(k)})$$
(15)

for every  $x = (x^{(1)}, \dots, x^{(M)}) \in \mathbb{R}^{n_x}$  and  $u = (u^{(1)}, \dots, u^{(M)}) \in \mathbb{R}^{n_u}$ .





#### **Boundary Conditions Reformulation**

Define  $N_{\phi} = \sum_{k=1}^{M} n_{\phi}^{k}$  and consider the function  $\phi : \mathbb{R}^{N_{x}} \times \mathbb{R}^{N_{x}} \to \mathbb{R}^{N_{\phi}}, \quad \phi = (\phi^{(1)}, \dots, \phi^{(M)})$ 

where for every  $k = 1, \ldots, M$ , the  $k^{th}$  component

 $\phi^{(k)}: \mathbb{R}^{N_x} \times \mathbb{R}^{N_x} \to \mathbb{R}^{n_x^{\phi}}$ 

is defined as

$$\phi^{(k)}(x,y) = \phi^{k}(x^{(k)}, y^{(k)})$$
(17)

for every  $x = (x^{(1)}, \dots, x^{(M)})$  and  $y = (y^{(1)}, \dots, y^{(M)})$  in  $\mathbb{R}^{N_x}$ .





(16)

#### Remark

Note that additional boundary conditions have to be imposed, in case continuity of the states between the different phases is not ensured by  $\phi^k$ . Indeed, in that case we consider the function

$$\Phi: \mathbb{R}^{N_X} \times \mathbb{R}^{N_X} \to \mathbb{R}^{(M-1) \cdot n_X}$$
(18)

defined as

$$\Phi(x, y) = \begin{bmatrix} x^{(2)} - y^{(1)} \\ \vdots \\ x^{(M)} - y^{(M-1)} \end{bmatrix}$$
(19)

for every  $x = (x^{(1)}, \ldots, x^{(M)})$  and  $y = (y^{(1)}, \ldots, y^{(M)})$  in  $\mathbb{R}^{N_x}$ .





Finally, let us define the  $N_x \times N_x$  diagonal matrix









With the previous reformulation, the Switched System Optimal Control Problem becomes

**Reformulated Optimal Control Problem** 

 $\begin{array}{rcl} & \textit{Minimize} \quad \varphi(x(0), x(1)) \\ \text{with respect to } x \in W^{1,\infty}\left([0,1]; \mathbb{R}^{N_x}\right), u \in L^{\infty}\left([0,1]; \mathbb{R}^{N_u}\right) \text{ and} \\ \rho = (p_1, \ldots, p_M) \in \mathbb{R}^M, \text{ subject to} \\ & \dot{x}(\tau) - P \cdot f(\tau, x(\tau), u(\tau)) &= 0_{\mathbb{R}^{N_x}} & \text{a.a. } \tau \in (0,1) \\ & g(x(\tau), u(\tau)) &\leq 0_{\mathbb{R}^{N_g}} & \forall \tau \in (0,1) \\ & \phi(x(0), x(M)) &= 0_{\mathbb{R}^{(M-1) \cdot n_x}} \end{array}$ 









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## **Example Problem**

Interaction between quadcopter and youBot platform.







We consider a model of the omni-directional mobile platform youBot provided by KUKA, consisting of:

- 1) youBot omni-directional platform
  - 4 KUKA omniWheels
  - 3 degrees of freedom
- 2) youBot arm with gripper
  - 5 axes robot arm
  - 5 degrees of freedom
  - 2 finger gripper







### **Robot Base**

#### Base Model

- 3 states: x, y and φ representing the position of the robot base in the xy-plane and its orientation
- 3 states:  $v_x$ ,  $v_y$  and  $v_{\varphi}$  representing the velocities of the base (translational and angular)
- 3 controls: u<sub>x</sub>, u<sub>y</sub> and u<sub>φ</sub> representing the accelerations (translational and angular)

#### Equations of Motion

$\dot{x}(t)$	=	$V_X(t)$	
ÿ(t)	=	$v_{y}(t)$	
$\dot{\varphi}(t)$	=	$v_{oldsymbol{arphi}}(t)$	(21)
$\dot{v}_x(t)$	=	$u_x(t)\cos(\varphi(t)) + u_y(t)\sin(\varphi(t))$	
$\dot{v}_y(t)$	=	$u_x(t)\sin(\varphi(t)) - u_y(t)\cos(\varphi(t))$	
$\dot{v}_{\varphi}(t)$	=	$u_{\varphi}(t)$	



# **Robot Base**







#### Robot Arm

#### Arm Model

- 5 states:  $q_1, \ldots, q_5$  representing the angles of the joints
- 5 states: v<sub>1</sub>,..., v<sub>5</sub> representing the angular velocities of the joints
- 5 controls:  $u_1, \ldots, u_5$  representing the angular acceleration of the joints

#### Equations of Motion

$$\dot{q}_{1}(t) = v_{1}(t), \quad \dot{v}_{1}(t) = u_{1}(t) 
\dot{q}_{2}(t) = v_{2}(t), \quad \dot{v}_{2}(t) = u_{2}(t) 
\dot{q}_{3}(t) = v_{3}(t), \quad \dot{v}_{3}(t) = u_{3}(t) 
\dot{q}_{4}(t) = v_{4}(t), \quad \dot{v}_{4}(t) = u_{4}(t) 
\dot{q}_{5}(t) = v_{5}(t), \quad \dot{v}_{5}(t) = u_{5}(t)$$

$$(22)$$





## Robot Arm







#### Robot Arm

Let *r* be the offset vector of the first joint with respect to the base and let  $l_1, \ldots, l_4$  be the lengths of the four arms. We define the rotation matrices

$$S_{0}(\alpha) = \begin{pmatrix} \cos(\alpha) & -\sin(\alpha) & 0\\ \sin(\alpha) & \cos(\alpha) & 0\\ 0 & 0 & 1 \end{pmatrix} \quad S_{1}(\beta) = \begin{pmatrix} \cos(\beta) & 0 & \sin(\beta)\\ 0 & 1 & 0\\ -\sin(\beta) & 0 & \cos(\beta) \end{pmatrix}. \quad (23)$$
$$S_{01}(\alpha, \beta) = S_{0}(\alpha)S_{1}(\beta)$$
$$S_{012}(\alpha, \beta, \gamma) = S_{0}(\alpha)S_{1}(\beta)S_{1}(\gamma) \quad (24)$$
$$S_{0123}(\alpha, \beta, \gamma, \delta) = S_{0}(\alpha)S_{1}(\beta)S_{1}(\gamma)S_{1}(\delta)$$

Then, the mount points  $P_1(q), \ldots, P_4(q)$  and the gripper position  $P_5(q)$  are given by the following equations





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# Robot Arm

# **Gripper Position**

$$P_{1}(q) = S_{0}(q_{1})r$$

$$P_{2}(q) = P_{1}(q) + S_{01}(q_{1}, q_{2}) \begin{pmatrix} 0 \\ 0 \\ l_{1} \end{pmatrix}$$

$$P_{3}(q) = P_{2}(q) + S_{012}(q_{1}, q_{2}, q_{3}) \begin{pmatrix} 0 \\ 0 \\ l_{2} \end{pmatrix}$$

$$P_{4}(q) = P_{3}(q) + S_{0123}(q_{1}, q_{2}, q_{3}, q_{4}) \begin{pmatrix} 0 \\ 0 \\ l_{3} \end{pmatrix}$$

$$P_{5}(q) = P_{4}(q) + S_{0123}(q_{1}, q_{2}, q_{3}, q_{4}) \begin{pmatrix} 0 \\ 0 \\ l_{4} \end{pmatrix}$$
(25)



Ultralight UAV with four rotors:

- 1) Six states:
  - *x*, *y*, *z* position on the quadcopter
  - yaw, roll and pitch angle
- 2) Four controls:
  - RPM of the rotors













#### Equations of Motion

$$m \cdot \left[ \begin{array}{c} \ddot{x} \\ \ddot{y} \\ \ddot{z} \end{array} \right] = F_A + F_m + F_W$$

where

- $F_A$ : lift force generated by the rotors
- Fm: gravitational force

F<sub>W</sub> : Drag





## Lift generated by the rotors

$$F_{A} = A \left( N_{Blades} \cdot C_{A} \cdot \frac{1}{2} \cdot \rho(z) \cdot A_{Blades} \cdot \left[ \begin{array}{c} 0 \\ 0 \\ U_{1}^{2} + U_{2}^{2} + U_{3}^{2} + U_{4}^{2} \end{array} \right] \right)$$

## Drag force

$$F_{W} = C_{W} \cdot \frac{1}{2} \cdot \rho(z) \begin{bmatrix} -sign(v_{x}) \cdot v_{x}^{2} \cdot A_{eff,x} \\ -sign(v_{y}) \cdot v_{y}^{2} \cdot A_{eff,y} \\ -sign(v_{z}) \cdot v_{z}^{2} \cdot A_{eff,z} \end{bmatrix}$$





### Moment generated by the Lift force

$$M_{A} = A \left( \frac{1}{4} \cdot \rho(z) \cdot N_{Blades} \cdot C_{A} \cdot r^{2} \cdot d \cdot \begin{bmatrix} U_{2}^{2} - U_{4}^{2} \\ U_{3}^{2} - U_{1}^{2} \\ 0 \end{bmatrix} \right)$$

# Moment generated by the rotors

$$M_R = A \left( \rho(z) \cdot N_{Blades} \cdot A_{Blades} \cdot C_M \cdot r^3 \cdot \begin{bmatrix} 0 \\ 0 \\ U_1^2 - U_2^2 + U_3^2 - U_4^2 \end{bmatrix} \right)$$





## Applications

#### Problem 1: Two interacting robots

- approach phase
- interaction phase
- return phase
- solution first case
- solution second case

#### Problem 2: Robot intercepted by a quadcopter

- approach phase
- fly-over phase
- return phase
- solution





- Each phase is an Optimal Control Problem
- Sequencing the single phases
- Not every sequence of phases is feasible





## Example

Quadcopter flying over several industrial robots







Example

#### Air traffic scheduling and control





# (1) BILEVEL OPTIMIZATION APPROACH

- Fixed sequence of phases are transformed to standard Optimal Control Problem (OCP)
- Mixed-Integer Optimization Problem (MIOP) is formulated for the unknown sequence of phases
- Bilevel Optimization Problem is solved with MIOP as upper level problem and OCP as lower level one

# (2) EQUILIBRIUM CONSTRAINTS APPROACH

- Fixed sequence of phases are transformed to standard OCP and first order necessary conditions are formulated
- MIOP is formulated for the unknown sequence of phases
- Large scale Mixed-Integer Mathematical Program with Equilibrium Constraints is solved





## **Decision Variables**

# Problem:

- Two phases  $\Phi_1, \Phi_2$  with starting times  $T_1, T_2$  and phase lengths  $p_1, p_2$
- · One of the states only starts when the other one is finished

# Approach:

- Define "big" constant  $N = T_1 + T_2 + p_1 + p_2$
- Define decision variable  $x \in \{0, 1\}$
- Consider the constraints:

$$T_1 + p_1 - T_2 \le (1 - x) \cdot N$$
  
 $T_2 + p_2 - T_1 \le x \cdot N$ 

corresponding to

- $x = 1 \quad \Leftrightarrow \quad \Phi_1 \text{ occures before } \Phi_2$
- $x = 0 \quad \Leftrightarrow \quad \Phi_2 \text{ occures before } \Phi_1$





Denote with OCP(k) the Optimal Control Problem corresponding to the *k*-th phase (k = 1, 2), i.e.

## $O\overline{CP(k)}$

$$Minimize \quad \varphi^k(x(T_k), x(T_k + p_k)) \tag{26}$$

with respect to  $x \in W^{1,\infty}([T_k, T_k + p_k]; \mathbb{R}^{n_x}), u \in L^{\infty}([T_k, T_k + p_k]; \mathbb{R}^{n_u^k})$  and  $p_k \in \mathbb{R}$ , subject to

$$\dot{x}(t) - f^{k}(t, x(t), u(t)) = 0_{\mathbb{R}^{n_{x}}}$$
 a.a.  $t \in (T_{k}, T_{k} + p_{k})$  (27)

$$g^{k}(x(t), u(t)) \leq 0_{\mathbb{R}^{n_{g}^{k}}} \quad \forall t \in (T_{k}, T_{k} + p_{k})$$

$$(28)$$

$$\phi^{k}(x(T_{k}), x(T_{k} + \rho_{k})) = 0_{\mathbb{R}^{n_{\phi}^{k}}}$$
<sup>(29)</sup>



Let  $p_1^*(T_1)$  and  $p_2^*(T_2)$  be the optimal phase length of OCP(1) and OCP(2) respectively, and let  $\frac{\partial p_1^*}{\partial T_1}(T_1)$  and  $\frac{\partial p_2^*}{\partial T_2}(T_2)$  be their sensitivities.

#### MIOP

Minimize 
$$\frac{\partial p_1^*}{\partial T_1}(T_1) \cdot d_1 + \frac{\partial p_2^*}{\partial T_2}(T_2) \cdot d_2$$
 (30)

with respect to  $d_1, d_2 \in \mathbb{R}$  and  $x \in \{0, 1\}$ , subject to

$$T_1 + p_1^*(T_1) + \frac{\partial p_1^*}{\partial T_1}(T_1) \cdot d_1 - T_2 \le (1 - x) \cdot N$$
(31)

$$T_2 + \rho_2^*(T_2) + \frac{\partial \rho_2^*}{\partial T_2}(T_2) \cdot d_2 - T_1 \le x \cdot N$$
(32)





#### Algorithm

- (1) Choose feasible starting times  $T_1$  and  $T_2$  and phase lengths  $p_1$  and  $p_2$
- (2) Solve OCP(k) for each phase k, compute sensitivities
- (3) Evaluate stopping criteria
- (4) Solve *MIOP*, compute search directions  $d_1$  and  $d_2$
- (5) Update  $T_1 = T_1 + d_1$ ,  $T_2 = T_2 + d_2$  and GOTO (2)











#### Discretization

Define the discretization grid

$$\mathbb{G}_k = \left\{ t_0 + i \cdot h_k \mid t_0 = T_k, h_k = \frac{p_k}{N_k}, i = 0, \dots, N_k \right\}$$

where  $N_k \in \mathbb{N}$ . Let  $x_i = x(t_i)$ ,  $u_i = u(t_i)$  for  $i = 0, ..., N_k$  be the discretized states and controls on  $\mathbb{G}_k$ .

#### DOCP(k)

$$Minimize \quad \varphi^k(x_0, x_{N_k}) \tag{33}$$

with respect to  $x_i \in \mathbb{R}^{n_x}$ ,  $u_i \in \mathbb{R}^{n_u^k}$  for  $i = 0, \ldots, N_k$  and  $p_k \in \mathbb{R}$ , subject to

$$x_{i+1} - x_i - f^k(t_i, x_i, u_i) = 0_{\mathbb{R}^{n_k}} \quad \forall i = 0, \dots, N_k - 1$$
 (34)

$$\mathcal{Y}^{k}(x_{i}, u_{i}) \leq 0_{\mathbb{R}^{n_{g}^{k}}} \quad \forall i = 1, \dots, N_{k} - 1$$
 (35)

$$\phi^k(x_0, x_{N_k}) = 0_{\mathbb{R}^{n_{\phi}^k}}$$
(36)



#### Lagrangian Function

We define the Lagrangian function of the DOCP(k)

$$\mathcal{L}_k: \mathbb{R}^{(N_k+1) \cdot n_x} \times \mathbb{R}^{N_k \cdot n_u^k} \times \mathbb{R}^{N_k \cdot n_x} \times \mathbb{R}^{(N_k-1) \cdot n_g} \times \mathbb{R}^{n_{\phi}} \to \mathbb{R}$$

as

$$\mathcal{L}_{k}(x, u, \lambda, \mu, \sigma) = \varphi^{k}(x_{0}, x_{N_{k}}) + \sum_{i=0}^{N_{k}-1} \lambda_{i+1}^{T} \cdot \left[x_{i+1} - x_{i} - f^{k}(t_{i}, x_{i}, u_{i})\right] + \sum_{i=1}^{N_{k}-1} \mu_{i}^{T} \cdot g^{k}(x_{i}, u_{i}) + \sigma^{T} \cdot \phi^{k}(x_{0}, x_{N_{k}})$$
(37)





#### First Order Necessary Conditions for DOCP(k)

Let  $(x^*, u^*, \rho_k^*)$  be an optimal solution of DOCP(k), than there exist multipliers  $\lambda^* \in \mathbb{R}^{N_k \cdot n_x}, \mu^* \in \mathbb{R}^{(N_k - 1) \cdot n_g}$  and  $\sigma^* \in \mathbb{R}^{n_{\phi}}$ , such that

Þ

$$\boldsymbol{\nabla}_{\boldsymbol{X}} \boldsymbol{\mathcal{L}}_{\boldsymbol{k}}(\boldsymbol{x}^*, \boldsymbol{u}^*, \boldsymbol{\lambda}^*, \boldsymbol{\mu}^*, \boldsymbol{\sigma}^*) = \boldsymbol{0}_{\mathbb{R}^{n_{\boldsymbol{X}}}}$$
(38)

$$x_{i+1} - x_i - f^k(t_i, x_i, u_i) = 0_{\mathbb{R}^{n_x}} \quad \forall i = 0, \dots, N_k - 1$$
 (39)

$$g^{k}(x_{i}, u_{i}) \leq 0_{\mathbb{R}^{n_{g}^{k}}} \quad \forall i = 1, \dots, N_{k} - 1$$

$$(40)$$

$$u_i \ge 0_{\mathbb{R}^{n_g^k}} \quad \forall i = 1, \dots, N_k - 1$$
 (41)

$$\mu_i^T \cdot g^k(x_i, u_i) = 0 \quad \forall i = 1, \dots, N_k - 1$$
 (42)

$$\phi^{k}(x_{0}, x_{N_{k}}) = 0_{\mathbb{R}^{n_{\phi}^{k}}}$$

$$\tag{43}$$





## Mixed-Integer Mathematical Program with Equilibrium Constraints

*Minimize* 
$$p_1 + p_2$$
 (44)  
if th respect to  
 $T_1, T_2, p_1, p_2 \in \mathbb{R}$   
 $x \in \{0, 1\}$   
 $(x^{(k)}, u^{(k)}, \lambda^{(k)}, \mu^{(k)}, \sigma^{(k)}) \in \mathbb{R}^{(N_k+1) \cdot n_x} \times \mathbb{R}^{N_k \cdot n_u^k} \times \mathbb{R}^{N_k \cdot n_x} \times \mathbb{R}^{(N_k-1) \cdot n_g} \times \mathbb{R}^{n_{\phi}}$   
ubject to  
 $T_1 + p_1 - T_2 \leq (1 - x) \cdot N$  (45)

$$T_2 + p_2 - T_1 \le x \cdot N \tag{46}$$



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#### Summary

- Introduction and Problem formulation
- SSOCP with fixed sequence of phases
- Models and test problems
- SSOCP with unknown sequence of phases
- Bilevel optimization approach
- Equilibrium constraints approach





# Thank you for your attention!

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