

# A Robust Approach for Project Scheduling Problem

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- 1 Introduction
- 2 Simple Case: Modeling without Bad Luck
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# Problem Description

## Eg: Minimal System Realization

Want to design a Low order Linear Time-Invariant(LTI) system

Denote by  $H_n$  the Hankel matrix with parameters  $h_1, h_2, \dots, h_{2n-1} \in \mathbb{R}$ :

$$H_n = \begin{bmatrix} h_1 & h_2 & h_3 & \cdots & h_n \\ h_2 & h_3 & h_4 & \cdots & h_{n+1} \\ h_3 & h_4 & h_5 & \cdots & h_{n+2} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ h_n & h_{n+1} & h_{n+2} & \cdots & h_{2n-1} \end{bmatrix}$$

Fact: Let  $h_1, h_2, \dots, h_n$  be given real numbers. Then there exists a minimal Linear Time-Invariant system of order  $r$  if and only if

$$r = \min_{h_{n+1}, \dots, h_{2n-1} \in \mathbb{R}} \text{rank}(H_n)$$

(Fazel)

# Eg: Minimal System Realization(Continued)

The problem can be expressed as:

$$\begin{array}{ll}
 \underset{H_n, s}{\text{minimize}} & \text{rank } H_n \\
 \text{subject to} & l_i \leq s_i \leq u_i, \quad i = 1, \dots, n \\
 & h_{n+1}, \dots, h_{2n-1} \in \mathbb{R}
 \end{array}$$

where:

- $s_i = \sum_{k=1}^i h_k$  denote the terms in the step response.
- $l_i$  lower bound of step response at the  $i$ th term.
- $u_i$  upper bound of step response at the  $i$ th term.

(Fazel)

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# SDCMPCC formulation

Want to solve:

$$\min_X \{ \text{rank}(X) : X \in \mathcal{C} \text{ and } X \in \mathbb{S}_+^n \}$$

Equivalently:

$$\min_{X, U} \quad n - \langle I, U \rangle$$

$$\text{subject to } X \in \mathcal{C}$$

$$0 \preceq U \preceq I$$

$$0 \preceq X \perp U \preceq 0$$

When  $X$  and  $U$  p.s.d,  $X \perp U$  is equivalent to:

$$\langle X, U \rangle = 0$$



Note that if  $X$  has the eigenvalue decomposition,

$$X = P^T \Sigma P$$

then we can choose

$$U = P_0 P_0^T$$

where  $P_0$  is composed of columns in  $P$  corresponding to 0 eigenvalue of  $X$ .

Thus, it is obvious that  $\text{rank}(X) = n - \langle I, U \rangle$ .

# SDCMPCC formulation

We can apply the SDCMPCC formulation to the general case  $X \in \mathbb{R}^{m \times n}$  by introducing an auxiliary variable  $Z$ :

$$Z = \begin{bmatrix} G & X^T \\ X & B \end{bmatrix} \succeq 0$$

For any  $X$ , can find matrix  $G$  and  $B$  such that  $Z \succeq 0$  and  $\text{rank}(Z) = \text{rank}(X)$

In the objective, we want to minimize the rank of  $Z$ .

# Constraint Qualification of SDCMPCC Formulation

Common Constraint qualifications such as LICQ and Robinson CQ are violated for SDCMPCC.

Here we consider **Local Calmness**.

## Definition

Suppose that  $\bar{x}$  is a local optimal solution to the problem:

$$\underset{x \in \mathcal{X}}{\text{minimize}} \quad f(x) \quad \text{subject to} \quad x \in \mathcal{L} \text{ and } g(x) \in -\mathcal{K} \quad (1)$$

Problem(1) is said to be calm of order  $\alpha > 0$  at  $\bar{x}$  if there exists  $M < \infty$  such that, for any sequence  $\{z^q\}$  with  $0 \neq z^q \rightarrow 0$  and any sequence  $\{x^q\} \subset \mathcal{L}$  satisfying  $x^q \rightarrow \bar{x}$  and  $g(x^q) \in z^q - \mathcal{K}$ , there holds

$$\frac{f(x^q) - f(\bar{x})}{\|z^q\|^\alpha} + M \geq 0 \quad (2)$$

# Constraint Qualification of SDCMPCC Formulation

Huang et.al shows that local calmness or order 1 implies the existence of KKT multipliers:

## Theorem

*Let  $\bar{x}$  be a local optimal solution to Problem(1) and (1) is calm of order 1 at  $\bar{x}$ . Then, there exists  $\mu \in K^*$  such that the system:*

$$0 \in \partial f(\bar{x}) + \mu(\nabla g(\bar{x})) + N_{\mathcal{L}}(\bar{x})$$

$$\mu(g(\bar{x})) = 0$$

*is consistent.  $N_{\mathcal{L}}(\bar{x})$  is the Clarke normal cone of  $\mathcal{L}$  at  $\bar{x}$ .*

# Constraint Qualification of SDCMPCC Formulation

## Proposition

*Calmness of Order 1* holds at each local optimum  $(\bar{X}, \bar{U})$  in the SDCMPCC Formulation.

In the proof, let  $(X^q, U^q)$  be a feasible solution to the perturbed SDCMPCC Formulation with perturbation parameter  $(z^q, r^q, h_1^q, h_2^q)$ .

Want to show the existence of  $M < \infty$  that satisfies:

$$(n - \langle I, U^q \rangle) - (n - \langle I, \bar{U} \rangle) \geq -M \|(z^q, r^q, h_1^q, h_2^q)\|$$

for any  $(X^q, U^q) \rightarrow (\bar{X}, \bar{U})$ .

An upper bound of  $n - \langle I, \bar{U} \rangle$  is  $\text{rank}(\bar{X})$ .

## Sketch of Proof

Want to get a lower bound for  $n - \langle I, U^q \rangle$ .  $(X^q, U^q)$  is feasible to the perturbed problem:

$$\begin{aligned}
 & \underset{X, U \in \mathbb{S}^n}{\text{minimize}} && n - \langle I, U \rangle \\
 & \text{subject to} && X + z^q \in \tilde{\mathcal{C}} \cap \mathcal{S}_+^n \\
 & && -\langle X, U \rangle \leq r^q \\
 & && \langle X, U \rangle \leq r^q \\
 & && I - U \succeq -h_1^q I \\
 & && U \succeq -h_2^q I
 \end{aligned}$$

The lower bound can be acquired by fixing  $X = X^q$  in the perturbed problem.

## Sketch of Proof

By fixing  $X = X^q$  we can get the following problem:

$$\begin{aligned}
 & \underset{U \in \mathbb{S}^n}{\text{minimize}} && n - \langle I, U \rangle \\
 & \text{subject to} && -\langle X^q, U \rangle \leq r^q, && y_1 \\
 & && \langle X^q, U \rangle \leq r^q, && y_2 \\
 & && I - U \succeq -h_1^q I, && \Omega_1 \\
 & && U \succeq -h_2^q I, && \Omega_2
 \end{aligned}$$

where  $y_1, y_2, \Omega_1, \Omega_2$  are the Lagrangian multipliers for the corresponding constraints.

$U^q$  is feasible to the above problem.

Slater condition holds for the above problem. Can find a lower bound the objective by **Strong Duality**

# Sketch of Proof

The dual problem is:

$$\begin{aligned}
 & \underset{y_1, y_2 \in \mathbb{R}, \Omega_1, \Omega_2 \in \mathbb{S}^n}{\text{maximize}} && n + r^q y_1 + r^q y_2 - (1 + h_1^q) \text{trace}(\Omega_1) - h_2^q \text{trace}(\Omega_2) \\
 & \text{subject to} && -y_1 X^q + y_2 X^q - \Omega_1 + \Omega_2 = -I \\
 & && y_1, y_2 \leq 0 \\
 & && \Omega_1, \Omega_2 \succeq 0
 \end{aligned}$$

By diagonalizing  $X^q$  we can get a tightened problem:

$$\begin{aligned}
 & \underset{y_1, y_2 \in \mathbb{R}, f, g \in \mathbb{R}^n}{\text{maximize}} && n + r^q y_1 + r^q y_2 - (1 + h_1^q) \sum_i f_i - h_2^q \sum_i g_i \\
 & \text{subject to} && -y_1 \lambda_i^q + y_2 \lambda_i^q - f_i + g_i = -1, \forall i = 1 \dots n \\
 & && y_1, y_2 \leq 0 \\
 & && f_i, g_i \geq 0, \forall i = 1 \dots n
 \end{aligned}$$



# Sketch of Proof

Since  $X^q \rightarrow \bar{X}$ ,  $\lambda_i^q \rightarrow \lambda_i$ . Can get a lower bound for the objective of the dual problem, which is:

$$\text{rank}(X) - \frac{2r^q}{\tilde{\lambda}} - (n - \text{rank}(X))(h_1^q + (1 + h_1^q)\frac{2}{\tilde{\lambda}}\|z^q\|) - \frac{h_2^q}{\tilde{\lambda}}\|\bar{X}\|^*$$

where  $\tilde{\lambda}$  is the smallest positive eigenvalue of  $\bar{X}$ . We can take

$$M = \frac{2}{\tilde{\lambda}} + \frac{1}{\tilde{\lambda}}\|\bar{X}\|^* + (n - \text{rank}(X))(1 + \frac{4}{\tilde{\lambda}})$$

# KKT Condition of SDCMPCC Formulation

Given  $\mathcal{C} = \{X \mid \langle A_i, X \rangle \geq b_i, \forall i = 1 \dots p\}$

The KKT condition is:

$$\begin{aligned}
 0 &\preceq U \quad \perp \quad -I + \mu X + Y \succeq 0 \\
 0 &\preceq X \quad \perp \quad -\sum \lambda_i A_i + \mu U \succeq 0 \\
 0 &\preceq Y \quad \perp \quad I - U \succeq 0 \\
 0 &\leq \lambda_i \quad \perp \quad b_i - \langle A_i, X \rangle \geq 0, \forall i = 1 \dots p
 \end{aligned} \tag{3}$$

Where  $\lambda, \mu$  and  $Y$  are lagrangian multipliers corresponding to the constraints  $A(X) = b$ ,  $\langle X, U \rangle = 0$  and  $I - U \succeq 0$  respectively.

Any feasible pair  $(X, U)$  with  $U$  given by  $P_0 P_0^T$  with columns of  $P_0$  to be the eigenvectors in the null space of  $X$ , is a KKT stationary point of the SDCMPCC Formulation.

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