Stabilizing Grand Cooperation of Machine Scheduling Game via Setup Cost Pricing

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Outline

- Preliminaries
- Motivation and Illustrative Example
- Models and Analyses
- Algorithms and Computations
- Extension and Generalization
- Conclusion

PRELIMINARIES

Cooperative Game

A **cooperative game** is defined by a pair (V, C):

- A set $V = \{1, 2, ..., v\}$ of players, grand colaition;
- A characteristic function C(S) = the minimum total cost achieved by the cooperation of members in coalition $S \in \mathbb{S} = 2^V \setminus \{\emptyset\}$.

The game requires:

• A cost allocation $\alpha = [\alpha_1, \alpha_2, \dots, \alpha_v] \in \mathbb{R}^v$, where $\alpha_k =$ the cost allocated to each player $k \in V$.

Core

Define
$$\alpha(S) = \sum_{k \in S} \alpha_k$$
.

A cost allocation $\alpha \in \mathbb{R}^{\nu}$ is in the **core** if it satisfies:

- Budget Balance Constraint: $\alpha(V) = C(V)$;
- Coalition Stability Constraints: $\alpha(S) \leq C(S)$ for each $S \in \mathbb{S}$.

$$\begin{aligned} \operatorname{Core}(V,C) &= & \left\{ \alpha : \ \alpha(V) = C(V), \right. \\ & \left. \alpha(S) \leq C(S), \ \forall S \in \mathbb{S} \setminus \{V\}, \ \alpha \in \mathbb{R}^v \right\}. \end{aligned}$$

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However, Core(V, c) can be empty.

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- S. Caprara and Letchford (2010, MP), Liu et al. (2016, IJOC)
- P. Faigle et al. (2001, IJGT), Schulz and Uhan (2010, OR)
- P&S Liu et al. (2018, OR)
- Inv. Opt. Liu et al. (2020, under review)

ILLUSTRATIVE EXAMPLE

Example: Machine Scheduling Game (MSG)

Game of Parallel Machine Scheduling with Setup Cost:

- Grand coalition: $V = \{1, 2, 3, 4\}$;
- Processing times: $t_1 = 2$, $t_2 = 3$, $t_3 = 4$, $t_4 = 5$;
- Machine setup cost: $t_0 = 9.5$;
- c(S) for S ∈ S: minimizes the total completion time of jobs in S plus the machine setup cost;
- $\pi(N) = \pi(\{1,3\}) + \pi(\{2,4\}) = 38$ (SPT Rule).





Example: Empty Core

Coalitions	Cost
{1}	11.5
{2}	12.5
{3}	13.5
{4}	14.5
$\{1, 2\}$	16.5
$\{1, 3\}$	17.5
$\{1, 4\}$	18.5
$\{2, 3\}$	19.5
$\{2,4\}$	20.5
{3,4}	22.5
$\{1, 2, 3\}$	25.5
$\{1, 2, 4\}$	26.5
$\{1, 3, 4\}$	28.5
{2,3,4}	31.5
$\{1,2,3,4\}$	38

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Optimal Cost Allocation Problem

$$\max (\alpha_{1} + \alpha_{2} + \alpha_{3} + \alpha_{4}) = 37.25 < 38$$

$$s.t. \quad \alpha_{1} \le 11.5, \quad \cdots, \quad \alpha_{4} \le 14.5,$$

$$\alpha_{1} + \alpha_{2} \le 16.5, \quad \cdots, \quad \alpha_{3} + \alpha_{4} \le 22.5,$$

$$\cdots,$$

$$\alpha_{1} + \alpha_{2} + \alpha_{3} + \alpha_{4} \le 38.$$

$$\alpha^* = [6; 8.75; 10.75; 11.75]$$

Models & Analyses

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- Grand coalition: $V = \{1, 2, \dots, v\}$;
- Identical machine: $M = \{1, 2, \dots, m\}$;
- Each machine Price: P and Each job processing time: t_k ;
- Characteristic function: $c(S) = \min(\sum_{k \in S} C_k + Pm_S)$,

where C_k is the completion time of job $k \in S$ and m_S is the number of using machine for the sub-coalition S.

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$$c(S, P) = \min \sum_{k \in V} \sum_{j \in O} c_{kj} x_{kj} + P \sum_{k \in S} x_{k1}$$

$$s.t. \quad \sum_{j \in O} x_{kj} - y_k^s = 0, \forall k \in V,$$

$$\sum_{k \in V} x_{kj} \le m, \forall j \in O,$$

$$x_{kj} \in \{0, 1\}, \forall k \in V, \forall j \in O,$$

$$y_k^s = 1, k \in s : y_k^s = 0, k \notin s.$$

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- Denote the right end of every subinterval as $P_i, 1 \leq i \leq v$, where $P_1 = P^*$.

$$\omega(P) = \min_{\alpha} \{ c(V, m(V, P)) - \alpha(V) : \\ \alpha(s) \le c(s, m(s, P)), \forall s \in S, \alpha \in \mathbb{R}^{v} \};$$

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- P_i , $2 \le i \le v$ can be obtained by SPT rules.
- $P_1 = P_2 + \cdots + P_n = \sum_{i=2}^n P_i$.

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- Define that

$$\omega_1(P) = \min_{\alpha} \{ c(V, m(V, P)) - \alpha(V) : \\ \alpha(s) \le c(s) + P, \forall s \in S, \alpha \in \mathbb{R}^{\nu} \}$$

Then the original problem $\omega(P)$ is equivalent to $\omega_1(P)$ which means that all sub-coalitions only use one machine.

ALGORITHMS & COMPUTATIONS

IPC Algorithm

The Intersection Points Computation(IPC) Algorithm to Construct $\omega(P)$ Function.

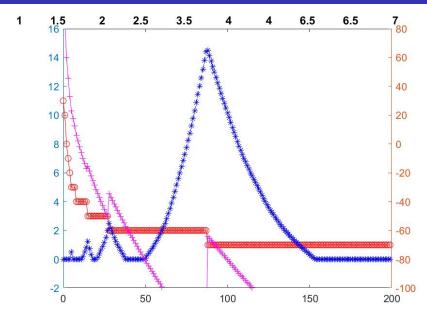
- **Step 1.** Initially, set $I^* = \{P_L, P_H\}$ and $\mathbb{I} = \{[P_L, P_H]\}$.
- **Step 2.** When \mathbb{I} is not empty: Sort values in I^* by $P_0 < P_1 < \cdots < P_q$, where $P_0 = P_L, P_q = P_H$ and $q = |I^*| 1$.
- **Step 3.** Obtain an LPB cost allocation α_{LP}^{GF} with $\alpha_{LP}^{GF}(N) = \pi_{LP}(N)$.

Computational Results

Instrument 1: LRB Cost Allocation

- **Step 1.** Find an LP $\min_{x} \{ cx : Gx \ge F\gamma^{s} \}$ giving a lower bound $\pi_{LP}(S)$ to $\pi(S)$;
- **Step 2.** Compute $(\alpha_{LP}^{GF})_i = (\mu^*)^T F_{.j}$, where μ^* : dual solution; $F_{.j}$: j-th column.
- **Step 3.** Obtain an LPB cost allocation α_{LP}^{GF} with $\alpha_{LP}^{GF}(N) = \pi_{LP}(N)$.

Image



EXTENSION & GENERALIZATION

Machine Scheduling Game with Weighted Jobs

Pricing in General IM Games

Conclusions

- * Cooperative Game Theory:
 - New Instrument for Stabilization via Cost Adjustment.
- * Inverse Problem:
 - Constrained Inverse Optimization Problem.
- * Models, Solution Methods and Applications:
 - Several equivalent LP formulations;
 - Feasibility analyses & How to handle infeasibility;
 - Implementations on WMG and UFL games.

The End

Thank you!