Stabilizing Grand Cooperation of Machine Scheduling Game via Setup Cost Pricing

Lindong Liu

School of Management; International Institute of Finance
University of Science and Technology of China

Co-authored with Zikang Li (PG Student, USTC)

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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 $S \in \mathbb{S}$: 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

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

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

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

$$\cdots,$$

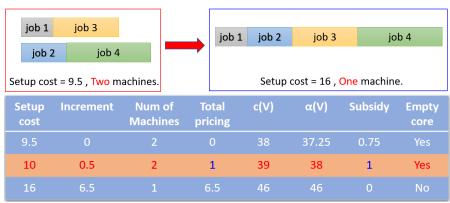
$$\alpha_{1}+\alpha_{2}+\alpha_{3}+\alpha_{4} < {\bf 38}.$$

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

The minimum subsidy: $c(V) - \alpha(V) = 38 - 37.25 = 0.75$

Example: Pricing Instrument

Increase the setup cost from 9.5 to 16. For the grand coalition, it only needs one machine.



The total pricing can cover the subsidy, which means the grand coalition can be stabilized by the players themselves.

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

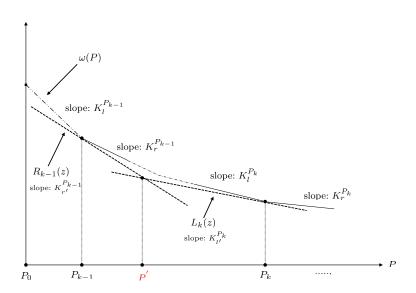
The Intersection Points Computation(IPC) Algorithm to Construct $\omega(P)$ Function.

- **Step 1.** Initially, set $I^* = \{P_L, P_H\}$ and $\mathbb{I} = \{[P_L, P_H]\}$.
- **Step 2.** If \mathbb{I} is not empty, update I^* and \mathbb{I} by the following steps:
- **Step 3.** Sort values in f^* by $P_0 < P_1 < \cdots < P_q$, where $P_0 = P_L, P_q = P_H$ and $q = |f^*| 1$.
- **Step 4.** Select any interval from \mathbb{I} , denoted by $[P_{k-1}, P_k]$ with $1 \le k \le q$.
- **Step 5.** Construct two linear function $R_{k-1}(P)$ and $L_k(P)$ so that $R_{k-1}(P)$ passes $(P_{k-1}, \omega(P_{k-1}))$ with a slope equal to a right derivative $K_r^{P_{k-1}}$ of $\omega(P)$ at P_{k-1} , and that $L_k(z)$ passes $(P_k, \omega(P_k))$ with a slope equal to a left derivative $K_r^{P_k}$ of $\omega(P)$ at P_k .

IPC Algorithm

- **Step 6.** If $R_{k-1}(P)$ passes $(P_k, \omega(P_k))$ or $L_k(P)$ passes $(P_{k-1}, \omega(P_{k-1}))$, then update $\mathbb I$ by removing $[P_{k-1}, P_k]$. Otherwise, $R_{k-1}(P)$ and $L_k(P)$ must have a unique intersection point at P = P' for some $P' \in (P_{k-1}, P_k)$. Update I^* by adding P', and update $\mathbb I$ by removing $[P_{k-1}, P_k]$, adding $[P_l, P']$ and $[P', P_r]$.
- Step 7. Go to step 2.
- **Step 8.** Return a piecewise linear function by connecting points $(P, \omega(P))$ for all $P \in I^*$.

IPC Algorithm



CP Algorithm

The Cutting Plane(CP) Algorithm to compute $\omega(P)$ for a given P.

- **Step 1.** Let $\mathbb{S}' \subseteq \mathbb{S} \setminus \{N\}$ indicates a restricted coalition set, which includes some initial coalitions, e.g., $\{1\}, \{2\}, \dots, \{v\}$.
- **Step 2.** Find an optimal solution $\bar{\alpha}(\cdot, P)$ to LP $\tau(P)$:

$$\max_{\alpha \in \mathbb{R}^n} \big\{ \alpha(\textit{N},\textit{P}) : \alpha(\textit{s},\textit{P}) \leq \textit{c}(\textit{s}) + \textit{P}, \text{ for all } \textit{s} \in \mathbb{S}' \big\}.$$

Step 3. Find an optimal solution s^* to the separation problem:

$$\delta = \min \{ c(s) + P - \bar{\alpha}(s, z) : \forall s \in \mathbb{S} \setminus \{N\} \}.$$

Step 4. If $\delta < 0$, then add s^* to \mathbb{S}' , and go to step 2; otherwise, return $\omega(P) = c(N) - \bar{\alpha}(N, P)$.

DP Algorithm

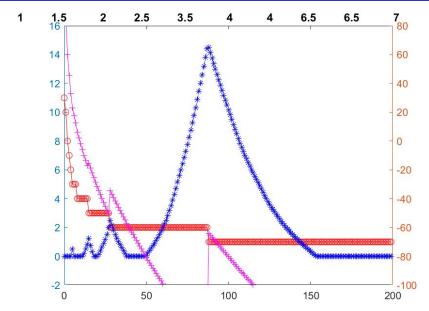
The Dynamic Programming(DP) Algorithm to solve the separation problem.

- **Step 1.** Initially, let D(k, u) indicate the minimum objective value of the restricted problem of separation problem, where $k \in \{1, 2, ..., v\}$ and $u \in \{0, 1, ..., v\}$.
- **Step 2.** Given the initial conditions D(1,0) = P and $D(1,1) = t_1 \beta_1 + P$. The boundary conditions are D(k,u) = if u > k, for all $k \in V$.
- **Step 3.** Given the recursion:

$$D(k,u) = \min \begin{cases} D(k-1,u), \text{ for the case when } s^* \text{ does not contain } k, \\ D(k-1,u-1) + ut_k - \alpha_k, \text{ for the case when } s^* \text{ contains } k. \end{cases}$$

Step 4. Obtain the optimal objective value of separation problem by $\delta_{AIPU} = \min\{D(v,u): u \in \{1,2,\ldots,v-1\}\}$. return δ_{AIPU} .

Computational Results



EXTENSION & GENERALIZATION

Machine Scheduling Game with Weighted Jobs

- Each job $k \in V$ has a processing time and a weight denoted by t_k, ω_k , respectively.
- c(S) can be obtained by assuming that $t_1/\omega_1 \le t_2/\omega_2 \le \ldots \le t_v/\omega_v$.
- $P_m = c_0(V, m) c_0(V, m-1), 2 \le m \le v$.
- $\omega(P)$ is piecewise linear, and convex in price P at each subinterval.
- The separation problem can be solved in polynomial time.

Pricing in General IM Games

- ILP: $c(S, m(S, P)) = \min_{x} \{cx + Pm(x) : Ax \ge By^{s} + D, \tilde{\alpha}x \le m, x \in \mathbb{Z}^{t \times 1}\}$
- Decompose c(S, m(S, P)) into $c_0(S, m(S)) + Pm$.
- $c_0(V, m) c_0(V, m-1) > 0 \Leftrightarrow P_m > 0, m = 2, ..., v.$
- $c_0(V, m) c_0(V, m+1) < c_0(V, m-1) c_0(V, m) \Leftrightarrow P_m < P_{m+1}, m = 2, 3, \dots, v-1.$
- $c_0(S_1, m-1) c_0(S_1, m) \ge c_0(S_1, m) c_0(S_1, m+1)$ $m=2, \ldots, v-1$.
- $c_0(S_1, m-1) c_0(S_1, m) \le c_0(S_2, m-1) c_0(S_2, m)$ $m=2,...,v, where <math>S_1 \subset S_2$.

Conclusions

- * Cooperative Game Theory:
 - New Instrument for Stabilization via Setup cost Pricing.
- * Scheduling Problem:
 - Parallel Machine Scheduling with Setup Cost.
- * Models, Solution Methods and Applications:
 - Several ILP formulations;
 - Cutting Plane to solve the seperation problem;
 - Implementations on the MSGW game.

The End

Thank you!