# 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

## **P**RELIMINARIES

## 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

$$\begin{aligned} \max \ \, \left(\alpha_1 + \alpha_2 + \alpha_3 + \alpha_4\right) &= \textbf{37.25} < \textbf{38} \\ s.t. \ \, \alpha_1 \leq \textbf{11.5}, \ \, \cdots, \ \, \alpha_4 \leq \textbf{14.5}, \\ \alpha_1 + \alpha_2 \leq \textbf{16.5}, \ \, \cdots, \ \, \alpha_3 + \alpha_4 \leq \textbf{22.5}, \\ \cdots, \\ \alpha_1 + \alpha_2 + \alpha_3 + \alpha_4 \leq \textbf{38}. \end{aligned}$$

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

## Models & Analyses

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- 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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A cooperative TU game (V,c) is called an IVPU game if it satisfies the following formulations:

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

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

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

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

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

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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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- When the number of using machines is 1 for the grand coalition, the range of slopes of the line segments in the interval is  $\left(-1, -\frac{1}{n-1}\right]$ , and the number of breakpoints is  $O(v^2)$ ;
- 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  $\alpha_{LP}^{GF}$  with  $\alpha_{LP}^{GF}(N) = \pi_{LP}(N)$ .

## CP Algorithm

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

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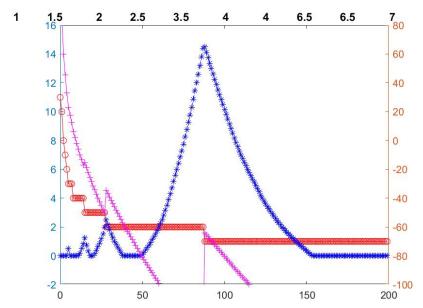
## DP Algorithm

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

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- **Step 3.** Obtain an LPB cost allocation  $\alpha_{LP}^{GF}$  with  $\alpha_{LP}^{GF}(N) = \pi_{LP}(N)$ .

## Computational Results

## **I**mage



## EXTENSION & GENERALIZATION

## Machine Scheduling Game with Weighted Jobs

## Pricing in General IM Games

## Conclusions

- \* Cooperative Game Theory:
  - New Instrument for Stabilization via Setup cost Pricing.
- \* Scheduling 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!