

Limited Discrepancy-Based Neighborhood Search for Parallel Machine Scheduling

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Outline

- 1 Introduction
- 2 Problem statement
- 3 Discrepancy-based search and local search methods
- 4 Branching strategy and node evaluation
- 5 Computational experiments and conclusions

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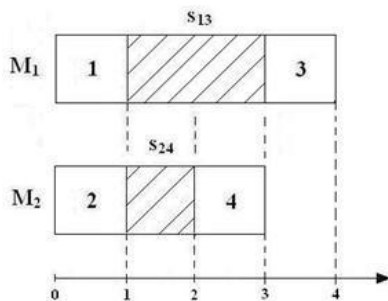
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Introduction

- Parallel machine scheduling problem [Pearn et al., 2007]
- Solved by tree and local search methods [Néron et al., 2006]
- Problem with setup times and precedence constraints between jobs
- Difficulty for list scheduling algorithm [Hurink & Knust, 2001]

Example

- The problems can't be efficiently solved by a list scheduling algorithm



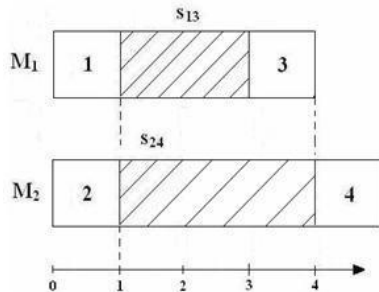
$$\begin{pmatrix} 0 & 10 & 2 & 10 \\ 10 & 0 & 0 & 1 \\ 10 & 10 & 0 & 10 \\ 10 & 10 & 10 & 0 \end{pmatrix}$$

$$\pi = (1, 2, 4, 3)$$

$$\pi' = (2, 1, 4, 3)$$

Example

- Precedence constraint $3 \prec 4$



$$\begin{pmatrix} 0 & 10 & 2 & 10 \\ 10 & 0 & 0 & 1 \\ 10 & 10 & 0 & 10 \\ 10 & 10 & 10 & 0 \end{pmatrix}$$

$$\pi = (1, 2, 4, 3)$$

$$\pi' = (2, 1, 4, 3)$$

- π and π' don't respect the precedence constraint
- Branching Structure: complete enumeration (best list of jobs and best resource allocation)

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Problem statement

Problem data

Decision variables

Optimization criteria

Problem statement

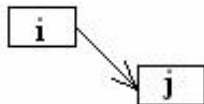
Problem data

- Jobs: $J = \{1, \dots, n\}$
- Resources/Machines: $M = \{M_1, \dots, M_m\}$
- For every job i : p_i, r_i, d_i
- Precedence constraints
- Setup times s_{ij}

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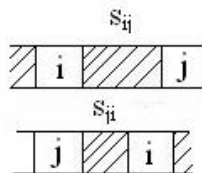


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Problem statement

Decision variables

- S_i ($C_i = S_i + p_i$)

Optimization criteria

- $\min \sum C_i$ [Chu et al., 2005]
- $\min L_{\max}$ [Uzsoy & Velásquez, 2006]

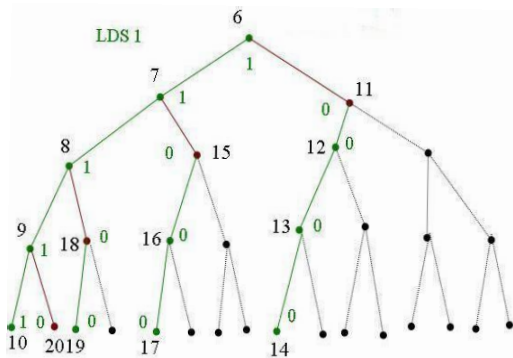
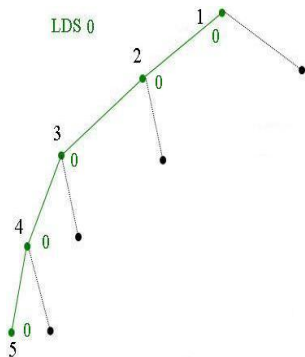
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Discrepancy-based search methods

Limited Discrepancy Search (LDS) [Harvey & Ginsberg, 1995]

- Iterative tree search method
- Instantiation heuristic to guide the search (discrepancies)



Discrepancy-based search methods

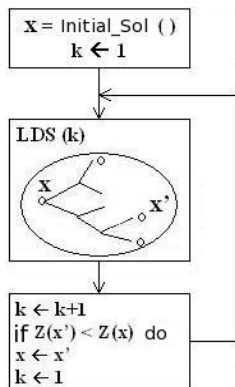
Methods based on LDS in order to increase efficiency:

- Improved LDS (ILDS) [Korf, 1996]
- Depth-bounded Discrepancy Search (DDS) [Walsh, 1997]
- Discrepancy-bounded Depth-First Search (DDFS) [Beck & Perron, 2000]
- Yet Improved LDS (YIELDS) [Karoui et al., 2007]

Large neighborhood local search based on LDS

Climbing Discrepancy Search (CDS) [Milano & Roli, 2002]

- Neighborhood exploration by LDS
- Discrepancies around the best current solution x



CDDS: the neighborhood is also limited by the depth in the tree [Ben Hmida et al., 2007]

Large neighborhood local search based on LDS

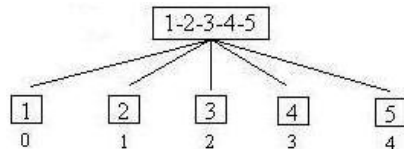
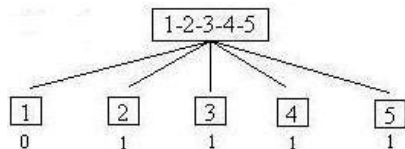
We propose two methods:

- Hybrid Discrepancy CDDS

- 1 CDS search until k discrepancies
- 2 $k \leftarrow k + 1$
- 3 CDDS search (not only on top of the tree), to restrict the neighborhood

- Mix Counting CDS

- Different counting discrepancies ways \rightarrow Depth tree level



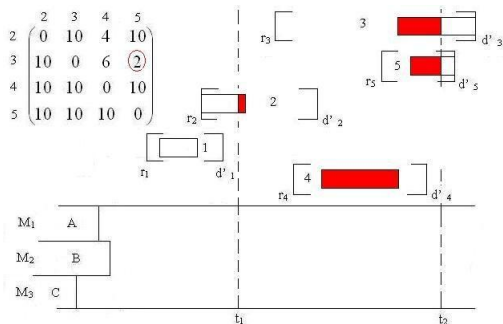
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- Initial heuristic
 - *Earliest Start Time (EST)* → minimize setup times
- Branching strategy
 - *DDFS (Discrepancy-bounded Depth First Search)*
 - *LDS-top*
 - *LDS-low*

Node evaluation

- $\min \sum C_i$: Lower bound [Chu et al., 2005]
- $\min L_{\max}$: Energetic reasoning [Lopez et al., 1992]
 - In a time interval $\Delta = [t_1, t_2] \rightarrow E_{consumed} + E_{setup} \leq E_{produced}$
 - $[r_i, d'_i]$, time windows for the set of not yet allocated jobs



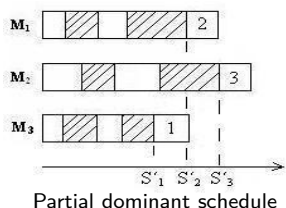
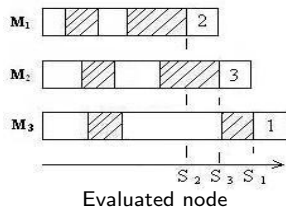
$$E_{produced} = m \times (t_2 - t_1)$$

$$E_{consumed} = \sum_i E_{min}, i \in F$$

$$E_{setup} \rightarrow k - m \text{ shortest } s_{ij}, i, j \in F$$

where F is the set of jobs that $E_{consumed} > 0$

Dominance rules for local search



- Goal: To find a partial dominant schedule ($S'_i < S_i$ and $S'_j \leq S_j$ for the rest of the last jobs executed on every resource)
- It has to be included in the explored neighborhood (authorised discrepancies)

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Computational experiments

Comparison on $P|r_i, q_i|C_{max}$ problems [Néron et al.]($n = 100$,
 $m = 10$, $p_i = [1 - 10]$)

$CPU_{limit} = 30 s$	Best Solution	Best Sol. Strict	CPU(s)
$LDS_{z=1}^{TW}$	1	0	29.64
$LDS_{z=2}^{CHR}$	7	0	28.40
$BS_{\omega=3}^{TW}$	25	3	20.37
$BS_{\omega=4}^{CHR}$	22	0	28.40
CDS	35	6	30 (8.03)
HD-CDDS	38	9	30 (7.02)

Computational experiments

- *ECT* rule evaluation
 - We find the optimal solution for 85 % of instances
 - Consider the job order discrepancies separately of machine allocation discrepancies
- Comparison with local search methods ($n = 100$, $m = \{2, 4\}$, $s_{ij} = \{[1 - 10], [1 - 20]\}$, $p_i = \{[1 - 10], [1 - 5]\}$, $OS = \{0.75, 0.5, 0.25\}$)

$\sum C_i$	<i>Best Solution</i>	<i>Mean Dev.</i>
<i>CDS</i>	12 (20.0 %)	1.4 %
<i>CDDS</i>	2 (3.3 %)	4.8 %
<i>HD-CDDS</i>	27 (45.0 %)	0.9 %
<i>MC-CDS</i>	40 (66.7 %)	1.0 %

L_{\max}	<i>Best Solution</i>	<i>Mean Dev.</i>
<i>CDS</i>	18 (30.0 %)	2.5 %
<i>CDDS</i>	0 (0.0 %)	11.6 %
<i>HD-CDDS</i>	36 (60.0 %)	1.8 %
<i>MC-CDS</i>	34 (56.7 %)	1.7 %

Conclusions and further work

- *LDS-top* branching strategy and *Binary counting* mode are the best strategies for *LDS* implementation
- Hybrid local-tree discrepancy based methods are efficient for parallel machine scheduling problems
- The computation of upper bounds improve the results (*EST*)
- We have efficiently introduced the setup times in the energetic reasoning and we have adapted the dominance rules for local search
- VRP problem: strategies and resolution methods