

Solving 2-stage Hybrid Flow Shop using Climbing Depth-bounded Discrepancy Search

Abir BEN HMIDA SAKLY^(1,2), Mohamed HAOUARI⁽²⁾
Marie-José HUGUET⁽¹⁾, Pierre LOPEZ⁽¹⁾

⁽¹⁾ LAAS-CNRS, Université de Toulouse, France

⁽²⁾ ROI, Ecole Polytechnique de Tunisie, Tunisia



OUTLINE

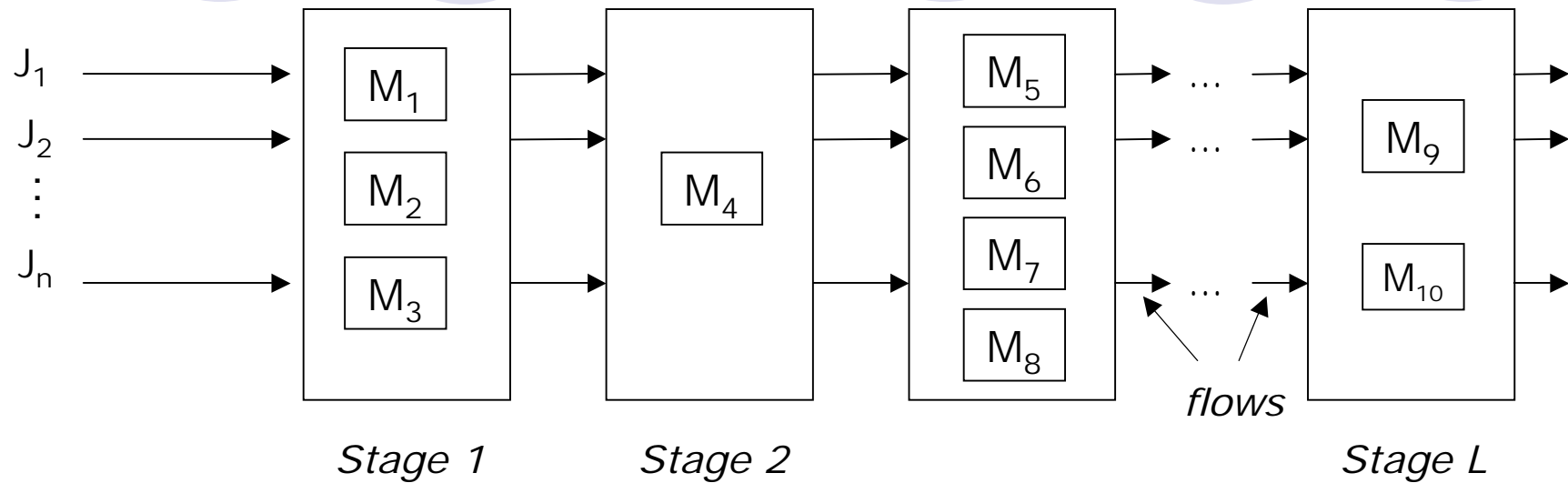
- Objectives
- 2-stage Hybrid Flow Shop Problem
- Discrepancy-based search methods
- Proposed method: CDDS
- Experimental evaluation and results
- Conclusions

Objectives



- To solve 2-stage Hybrid Flow Shop (“2-HFS”)
- Adapting discrepancy-based search methods for the problem under study
- Developing associated algorithms
- Experimental evaluation of the propositions

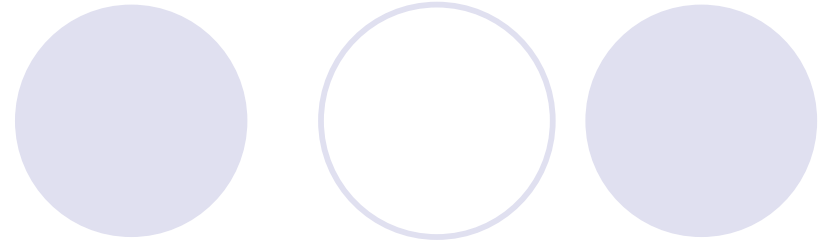
Hybrid Flow Shop



- $J = \{J_1, \dots, J_i, \dots, J_n\}$ jobs
- $E = \{1, \dots, s, \dots, L\}$ stages
- m_s : identical parallel machines / $\max(m_s) > 1$
- Application: semiconductors
(Printed Circuit Boards)



Hybrid Flow Shop



- As in a Flow Shop:
 - Resources = machines
 - Job = sequence of operations
- But:
 - Alternative machines can process an operation
- Double problem:
 - Select a machine for each operation
 - Determine a start time for each operation
- 2 stages \Rightarrow 2-HFS
- 2-HFS | $\max(m_1, m_2) > 1$ | C_{\max} is strongly NP-hard

2-stage Hybrid Flow Shop

- Background

- 2-HFS with $\max(m_s)=2$ for $s \in \{1,2\}$ is NP-hard: [Gupta, 1988]
- Exact methods: [Brah & Hunsucker, 1991]; [Portmann *et al.*, 1992]; [Moursli & Pochet, 2000]; [Carlier & Néron, 2000]; [Lin & Liao, 2003]; [Haouari *et al.*, 2006]
- Lower bounds: [Santos *et al.*, 1995]; [Moursli & Pochet, 2000]; [Carlier & Néron, 2000]; [Haouari and M'Hallah, 1997]
- Heuristics: [Brah & Loo, 1999]; [Engin & Döyen, 2004]; [Haouari and M'Hallah, 1997]

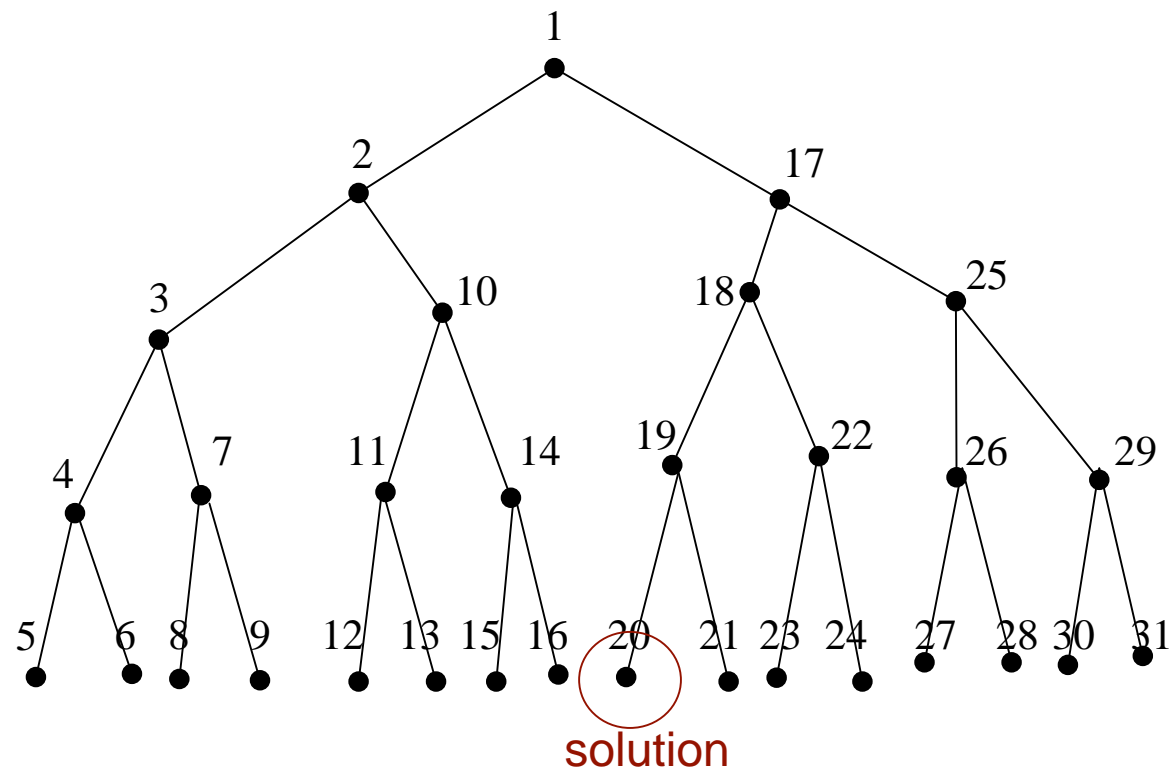
Discrepancy-based search methods (1)

- Limited Discrepancy Search - LDS [Harvey & Ginsberg, 1995]
 - Is a problem satisfiable? → *Satisfaction*
 - Iterative tree search method
 - Instantiation heuristic to guide the search
(the initial global instantiation is not necessarily a solution)
 - When the heuristic does not find a good solution, it is probably because it made a few poor choices → *discrepancy* then makes a choice different than heuristically top-ranked
 - Hope to find a solution before *Depth-First Search*

Depth First Search (DFS)

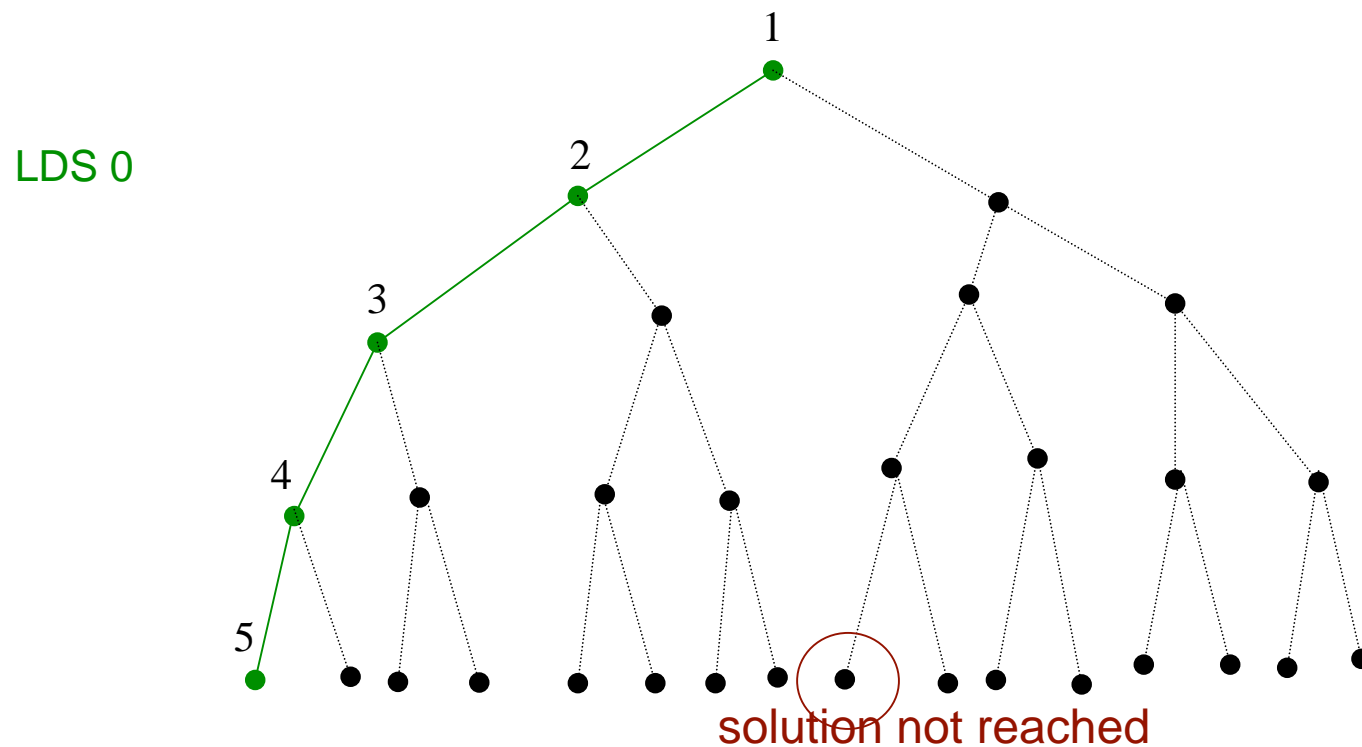
- Search principle
- Example: **binary tree**

DFS



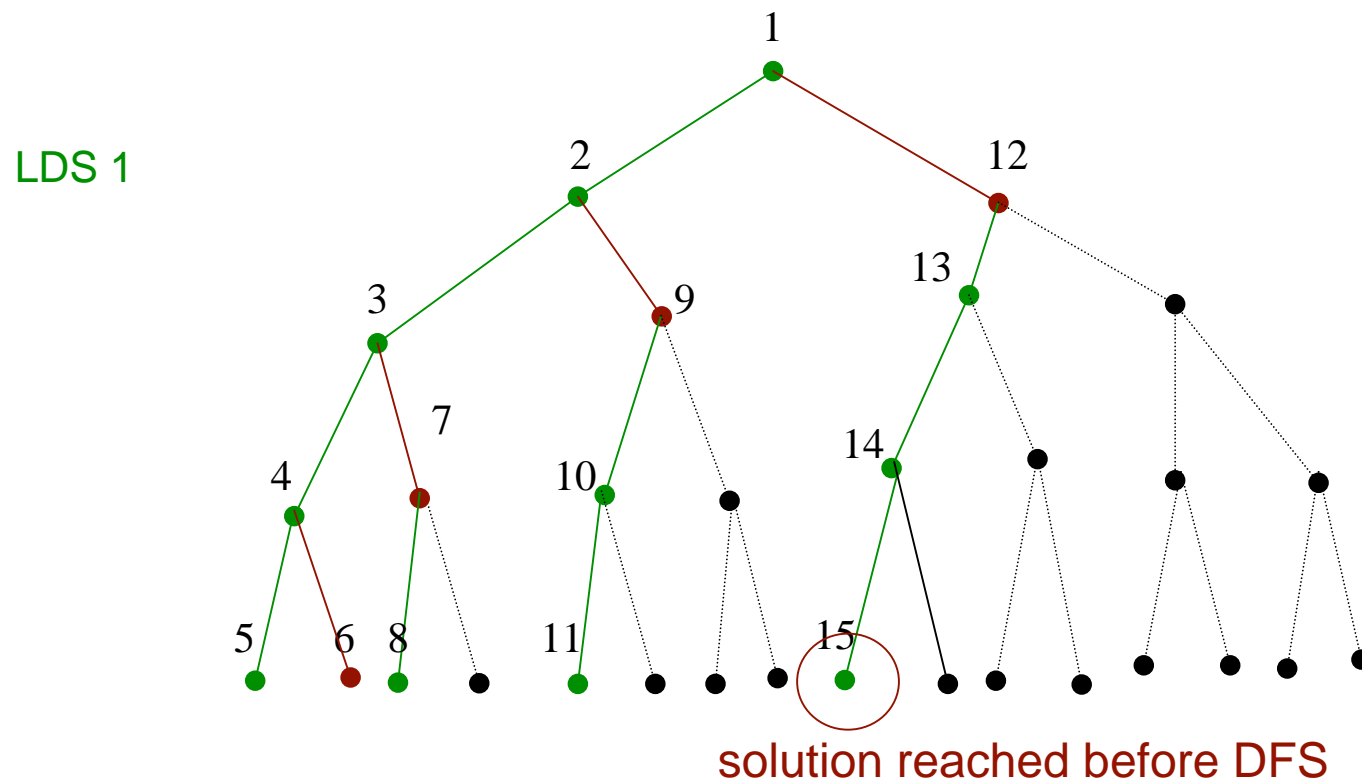
Limited Discrepancy Search (LDS)

- Search principle
 - Example: **binary tree**
 - LDS 0: The choices of the heuristic are satisfied



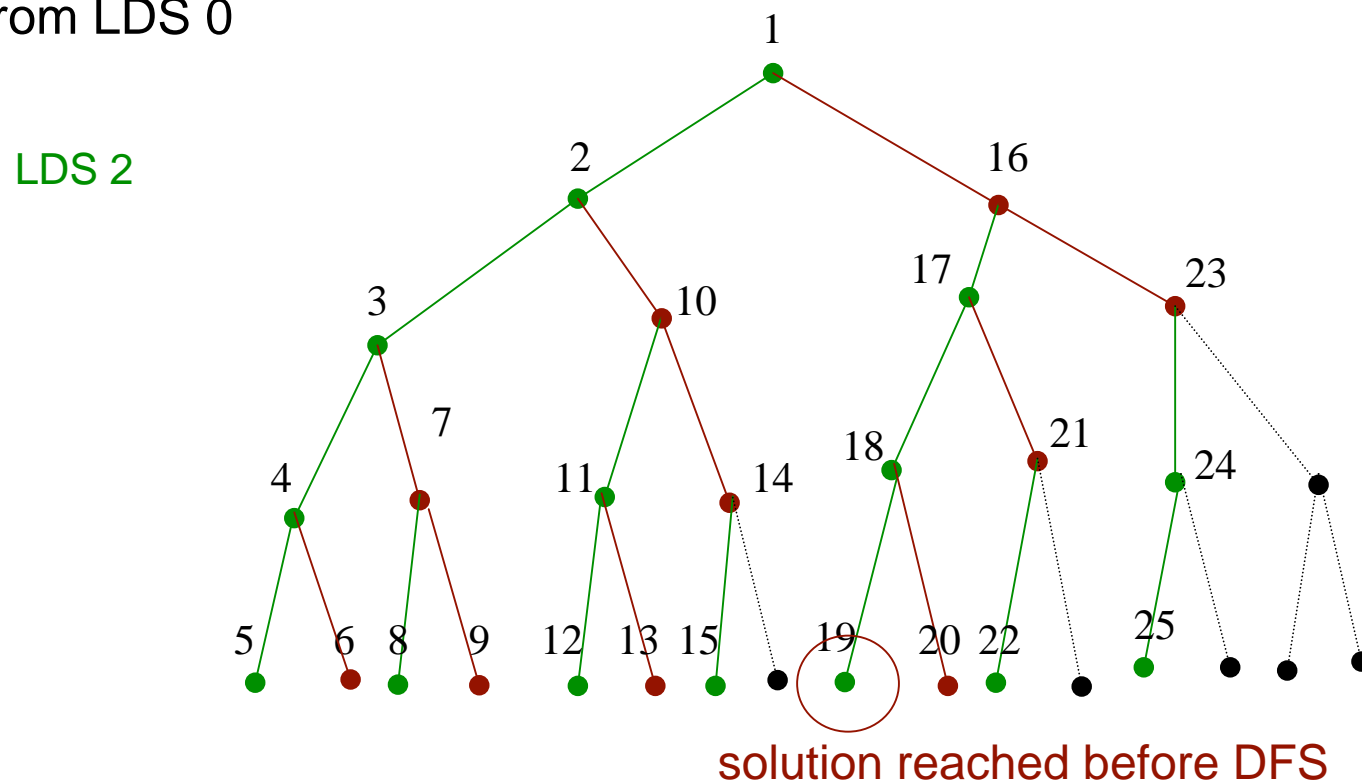
Limited Discrepancy Search (LDS)

- Search principle
 - Example: **binary tree**
 - LDS 1: All the paths differ of ONE decision from LDS 0



Limited Discrepancy Search (LDS)

- Search principle
 - Example: **binary tree**
 - LDS 2: All the paths differ of TWO decisions (for TWO variables) from LDS 0

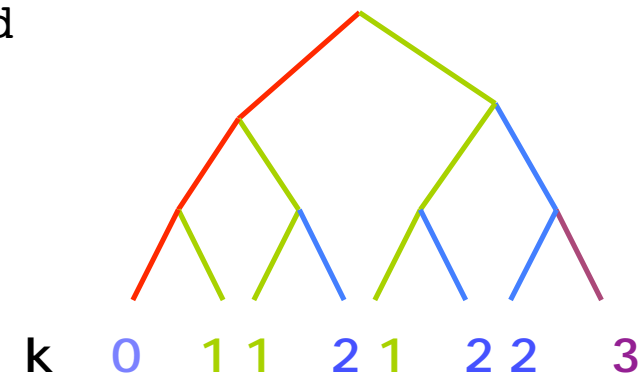


(Improved) Limited Discrepancy Search (I-LDS)

[\[Harvey & Ginsberg, 1995\]](#) and [\[Korf, 1996\]](#)

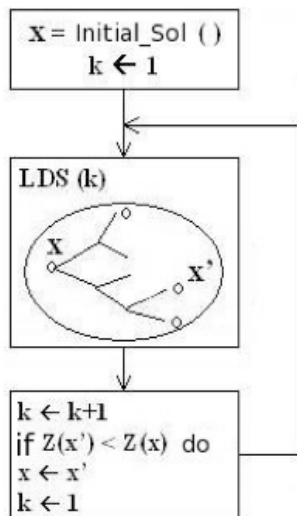
Algorithm

```
k ← 0
kmax ← N
I ← Initial_instantiation()
While no_solution() and (k ≤ kmax) do
  k ← k+1
  -- Generate leaves at discrepancy k from I
  -- Stop when a solution is found
  I ← compute_Leaves (I, k)
End while
```



Discrepancy-based search methods (2)

- Depth-bounded Discrepancy Search - **DDS** [Walsh, 1997]
 - → *Satisfaction*
 - To correct “early mistakes” (the most important)
 - Principle: discrepancies on top of the search tree
 - Stop: a solution is found
- Climbing Discrepancy Search - **CDS** [Milano & Roli, 2005]



- To improve current solution → *Optimization*
- Principle:
 - Initial solution (Reference)
 - Apply LDS principle to explore the neighborhood from this reference
 - Reference ← Improved_Solution
 - Restart with #discrepancy := 0
- Stop: no more improvement, limit on time or #iterations reached
- CDS is close to VNS [Hansen & Mladenovic, 2001]

Proposed method

- To combine 2 discrepancy-based methods
 - Climbing DS (neighborhood search)
 - Depth-bounded DS (neighborhood restricted at the top of the tree)

→ *Climbing Depth-bounded Discrepancy Search (CDDS)*
- Optimization method: approximate solutions
 - Criterion = *makespan* minimization
 - A solution = UB
- LBs to fathom nodes in the search tree

$$\min C_{\max}$$

Proposed method

- Exploration strategy
 - Instantiation heuristics
 - First Stage:
 1. Job Selection : Extension of Johnson's rule for the 2-machine FS having $\{ p_{1,j}/ m_1 , p_{2,j}/ m_2 \}$ as processing times
 2. Machine selection (allocation): First Available machine (FAM)
 - Second Stage:
 1. Job Selection : Earliest release date (and longest $p_{2,j}$)
 2. Machine selection (allocation): First Available machine (FAM)
 - Simple propagation
 - Forward-Checking over:
 - Earliest start time of subsequent operations
 - Release date of selected machine
 - Discrepancies
 - On job selection variables

Proposed method

- Lower bounds: SPT-rule [Haouari and M'Hallah, 1997]

1. On second stage:

$$LB_{SPT}^2(S) = \left\lceil \frac{I_2(S) + \sum_{j \in S} p_{2,j}}{m_2} \right\rceil$$

2. On first stage :

$$LB_{SPT}^1(S) = \left\lceil \frac{I_1(S) + \sum_{j \in S} p_{1,j}}{m_1} \right\rceil$$

3. So,

$$LB = \max \left(LB_{SPT}^1, LB_{SPT}^2 \right)$$

Experiments

- Instances:

- Three sets of instances generated in a similar way as in [\[Lee and Vairaktarakis, 1994\]](#):
- Set A:
 - $n = \{10, 20, 30, 40, 50, 100, 150\}$
 - $M: (m_1, m_2)$ are (2, 2), (2, 4), (4, 2), and (4, 4)
 - processing times: uniform distribution in [1, 20] for the first stage and in [1, 40] for the second stage
 - 560 instances
- Set B:
 - processing times: uniform distribution in [1, 40] for the first stage and in [1, 20] for the second stage
- Set C:
 - processing times: uniform distribution in [1, 40] for both stages
- 1680 instances

Experiments

- Comparison of $CDDS^2$ (2-HFS) with:
 - $CDDS^L$ of [Ben Hmida *et al.*, 2007] (HFS)
 - LBs of [Haouari *et al.*, 2006]
 - TS of [Haouari and M'Hallah, 1997]

$$\text{Dev} = \frac{C \max_best - \text{LowerBound}}{\text{LowerBound}} \times 100$$

- Stop:
 - limit on CPU time=15 sec.
- Instance and its reverse are considered

Results

Table 1. Performance on Set A

n	(m_1, m_2)	$CDDS^2$			$CDDS^L$		
		US	Dev	Time	US	Dev	Time
10	(2, 2)	3	0.21	3.62	11	1.20	8.79
	(2, 4)	10	2.47	5.28	17	6.24	12.83
	(4, 2)	0	0.00	0.07	3	0.06	2.31
	(4, 4)	0	0.00	0.13	4	0.12	3.10
20	(2, 2)	2	0.05	2.03	3	0.16	3.11
	(2, 4)	9	0.91	8.42	20	5.79	15.00
	(4, 2)	0	0.00	0.94	4	0.12	3.75
	(4, 4)	4	0.21	4.60	11	1.34	8.77
30	(2, 2)	1	0.02	0.92	7	1.61	5.85
	(2, 4)	6	0.86	6.89	17	5.61	12.92
	(4, 2)	0	0.00	0.45	3	0.12	2.63
	(4, 4)	3	0.07	3.78	8	0.32	6.76
40	(2, 2)	0	0.00	0.24	2	0.04	1.72
	(2, 4)	4	0.21	5.18	8	0.97	6.78
	(4, 2)	0	0.00	0.96	4	0.56	3.77
	(4, 4)	2	0.05	2.89	4	0.12	4.16
50	(2, 2)	0	0.00	0.40	3	0.07	2.59
	(2, 4)	2	0.15	2.37	6	0.70	5.33
	(4, 2)	0	0.00	0.60	4	0.11	3.48
	(4, 4)	2	0.06	3.91	4	0.08	4.56
100	(2, 2)	1	0.05	1.88	4	0.23	3.75
	(2, 4)	2	0.06	3.68	6	0.30	5.79
	(4, 2)	0	0.00	0.91	5	0.09	4.43
	(4, 4)	2	0.02	3.71	6	0.15	5.80
150	(2, 2)	0	0.00	4.40	3	0.30	5.99
	(2, 4)	0	0.00	2.41	2	0.16	3.67
	(4, 2)	0	0.00	5.27	3	0.18	6.73
	(4, 4)	1	0.01	8.10	2	0.05	8.79
Average		54	0.19	3	174	0.96	5.83

Results

Table 2. Performance on **Set B**

<i>n</i>	(m_1, m_2)	<i>CDDS</i> ²			<i>CDDS</i> ^L		
		US	Dev	Time	US	Dev	Time
10	(2, 2)	2	0.09	1.51	5	0.12	4.32
	(2, 4)	0	0.00	0.44	5	0.38	4.08
	(4, 2)	9	1.60	6.03	19	5.92	14.28
	(4, 4)	2	0.29	1.52	7	1.18	5.74
20	(2, 2)	1	0.03	0.79	5	0.08	4.34
	(2, 4)	0	0.00	0.03	3	0.09	2.28
	(4, 2)	5	0.64	4.91	13	5.70	10.09
	(4, 4)	3	0.14	2.91	4	0.15	3.78
30	(2, 2)	0	0.00	0.03	4	0.16	3.02
	(2, 4)	0	0.00	0.10	3	0.12	2.34
	(4, 2)	7	0.93	6.05	11	5.69	8.64
	(4, 4)	2	0.11	2.49	3	0.19	3.31
40	(2, 2)	0	0.00	0.19	2	0.06	1.67
	(2, 4)	0	0.00	0.09	3	0.53	2.33
	(4, 2)	3	0.28	3.06	9	1.01	7.31
	(4, 4)	2	0.05	13.09	4	0.55	8.24
50	(2, 2)	0	0.00	0.65	3	0.56	2.80
	(2, 4)	0	0.00	0.10	2	0.43	1.59
	(4, 2)	8	0.37	7.23	10	0.69	7.95
	(4, 4)	1	0.02	2.94	3	0.06	4.75
100	(2, 2)	0	0.00	0.52	3	0.09	2.69
	(2, 4)	0	0.00	0.55	2	0.09	2.00
	(4, 2)	2	0.03	4.30	7	0.35	6.65
	(4, 4)	2	0.02	3.30	8	0.17	6.99
150	(2, 2)	0	0.00	1.39	3	0.08	3.43
	(2, 4)	0	0.00	0.90	4	0.13	3.72
	(4, 2)	1	0.03	3.03	5	0.22	6.02
	(4, 4)	1	0.01	5.06	3	0.03	6.55
Average		51	0.17	2.62	153	0.89	5.03

Results

Table 3. Performance on Set C

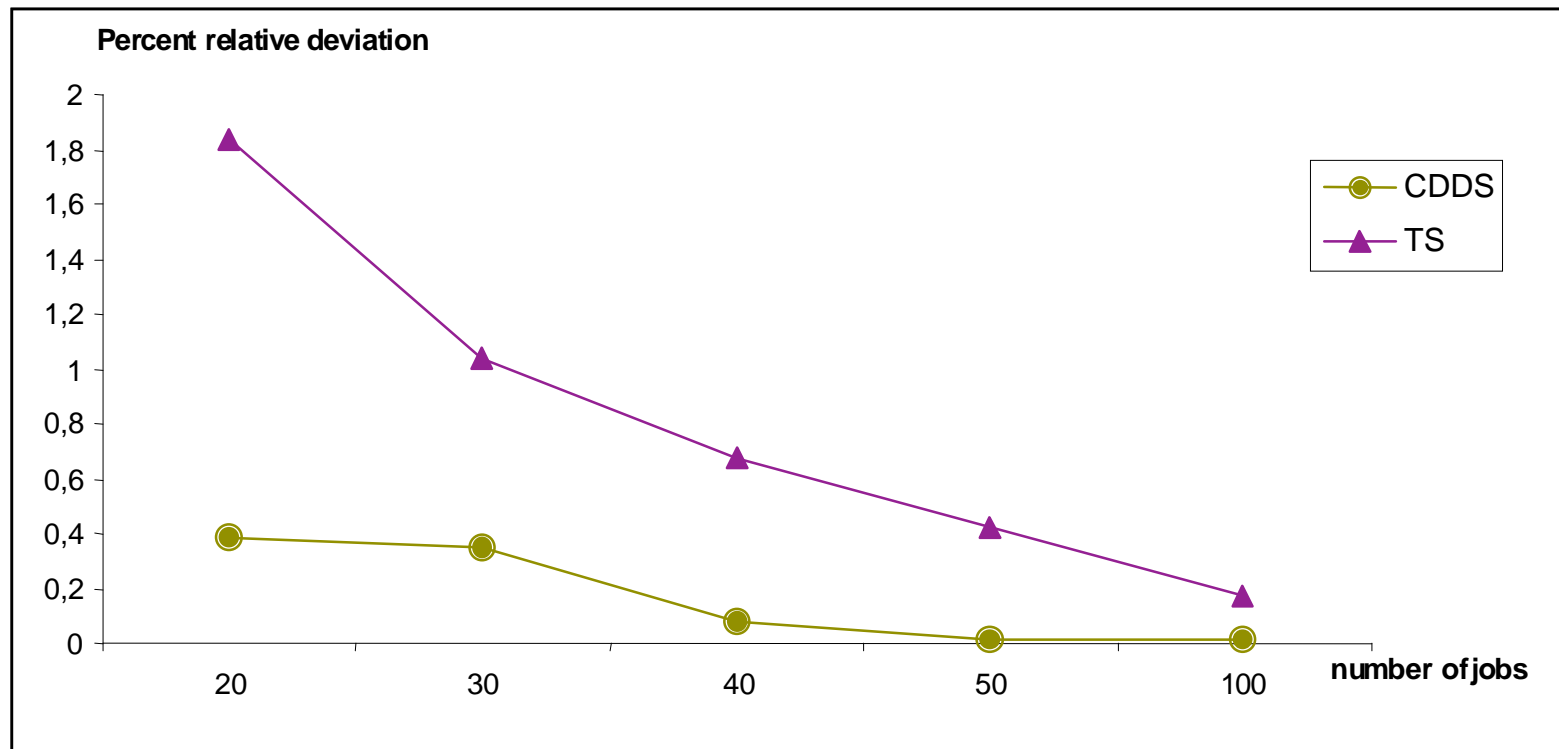
n	(m_1, m_2)	$CDDS^2$			$CDDS^L$		
		US	Dev	Time	US	Dev	Time
10	(2, 2)	3	0.20	2.35	9	0.26	7.18
	(2, 4)	1	0.34	0.81	6	0.45	5.07
	(4, 2)	0	0.00	0.06	5	0.05	3.80
	(4, 4)	9	1.89	6.75	18	1.98	13.58
20	(2, 2)	5	0.39	3.85	11	0.48	8.60
	(2, 4)	0	0.00	0.07	7	0.07	5.30
	(4, 2)	0	0.00	0.91	3	0.23	3.02
	(4, 4)	13	1.19	9.87	20	1.47	15.00
30	(2, 2)	4	0.10	3.07	12	0.63	9.31
	(2, 4)	0	0.00	0.44	8	0.08	6.26
	(4, 2)	1	0.02	1.18	5	0.05	4.64
	(4, 4)	9	1.42	9.10	14	1.95	10.80
40	(2, 2)	3	0.08	2.38	9	0.26	7.19
	(2, 4)	1	0.01	0.81	7	0.14	5.78
	(4, 2)	1	0.01	2.09	6	0.17	5.96
	(4, 4)	5	0.46	5.39	11	0.61	8.74
50	(2, 2)	0	0.00	0.19	6	0.06	4.63
	(2, 4)	0	0.00	0.26	8	0.28	6.16
	(4, 2)	2	0.02	1.88	5	0.05	4.46
	(4, 4)	5	0.84	4.89	12	1.20	9.39
100	(2, 2)	0	0.00	0.73	9	0.16	7.15
	(2, 4)	0	0.00	0.56	11	0.15	8.50
	(4, 2)	2	0.01	3.44	9	0.09	7.70
	(4, 4)	6	0.22	6.73	7	0.32	5.98
150	(2, 2)	1	0.01	4.28	8	0.08	8.57
	(2, 4)	0	0.00	1.62	9	0.09	7.64
	(4, 2)	0	0.00	5.04	7	0.07	8.53
	(4, 4)	7	0.11	9.69	10	0.14	8.19
Average		78	0.26	3.16	252	0.41	7.40

Results

Table 4. Performance comparison between the proposed CDDS and TS [Haouari and M'Hallah, 1997]

<i>n</i>		<i>Performance on Set A</i>			<i>Performance on Set B</i>			<i>Performance on Set C</i>			<i>Average</i>
		(2, 4)	(4, 4)	(4, 2)	(2, 4)	(4, 4)	(4, 2)	(2, 4)	(4, 4)	(4, 2)	
20	<i>CDDS</i>	0.95	0.26	0.00	0.03	0.14	0.73	0.00	1.48	0.05	0.40
	<i>TS</i>	2.90	1.20	0.35	0.92	5.72	0.13	0.56	3.43	1.22	1.83
30	<i>CDDS</i>	0.92	0.10	0.00	0.00	0.11	0.96	0.07	1.45	0.02	0.40
	<i>TS</i>	1.43	0.85	0.06	0.57	3.10	0.05	0.27	1.45	1.46	1.03
40	<i>CDDS</i>	0.21	0.05	0.00	0.00	0.05	0.28	0.02	0.46	0.01	0.12
	<i>TS</i>	0.96	0.43	0.12	0.5	1.57	0.12	0.34	1.08	0.89	0.67
50	<i>CDDS</i>	0.15	0.06	0.00	0.00	0.02	0.37	0.00	0.88	0.02	0.16
	<i>TS</i>	0.54	0.30	0.02	0.26	1.09	0.04	0.20	0.95	0.42	0.42
100	<i>CDDS</i>	0.06	0.02	0.00	0.00	0.02	0.03	0.00	0.22	0.01	0.04
	<i>TS</i>	0.19	0.15	0.02	0.11	0.39	0.01	0.07	0.41	0.18	0.17
Average	<i>CDDS</i>	0.46	0.10	0.00	0.01	0.07	0.48	0.02	0.90	0.02	0.22
	<i>TS</i>	1.20	0.59	0.11	0.47	2.37	0.07	0.29	1.46	0.83	0.82

Results



Conclusions

- Novel method to solve the 2-HFS
- Promising results
 - outperforms the previous algorithm (*CDDS*^L)
 - outperforms the TS procedure of [Haouari and M'Hallah, 1997]
 - produces optimal solutions for more than 89.2% (1498 instances out of 1680)
 - provides a tiny gap (≤ 0.22 if all instances are considered)
- Good results for solving Flexible Job Shop problems ([Ben Hmida *et al.*, 2009] submitted)

instances	#	<i>CDDS</i>	<i>GA</i>	<i>TS</i>	<i>hGA</i>
<i>Brandimarte</i>	10	15.0	17.5	15.1	14.9
<i>Barnes/Chambers</i>	21	22.5	29.6	22.5	22.6
<i>Hurink Edata</i>	43	2.3	6.0	2.2	2.1
<i>Hurink Rdata</i>	43	1.3	4.4	1.2	1.2
<i>Hurink Vdata</i>	43	0.1	2.0	0.1	0.08

Limit on CPU time=15 sec.

