

# HEURISTIC ALGORITHMS FOR SCHEDULING IN A MULTIPROCESSOR TWO-STAGE FLOWSHOP WITH 0-1 RESOURCE REQUIREMENTS

Ewa Figielska

Warsaw School of Computer Science, Lewartowskiego 17, 00-169 Warsaw, Poland

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Abstract: This paper deals with the problem of preemptive scheduling in a two-stage flowshop with parallel unrelated machines at the first stage and a single machine at the second stage. At the first stage, jobs use some additional resources which are available in limited quantities at any time. The resource requirements are of 0-1 type. The objective is the minimization of makespan. The problem is NP-hard. Heuristic algorithms are proposed which, while solving to optimality the resource constrained scheduling problem at the first stage of the flowshop, select for simultaneous processing jobs according to rules promising a good (short) schedule in the flowshop. Several rules of job selection are considered. The performance of the proposed heuristic algorithms is analyzed by comparing their solutions with the lower bound on the optimal makespan. The results of computational experiments show that these heuristics are able to produce near-optimal solutions in short computation time.

## 1 INTRODUCTION

During the last years, the flowshops with multiple processors (FSMP) also called hybrid flowshops, have received considerable attention from researchers (e.g. Gupta 1988; Chen 1995; Haouari and M'Hallah, 1997; Brah and Loo (1999), Linn and Zhang, 1999; Ruiz and Maroto, 2006).

In this paper, we extend multiprocessor flowshop scheduling research by including resource constraints. We consider the problem of scheduling in a two-stage flowshop where jobs use additional renewable resources, which are available in limited quantities at any time. This problem can be described as follows. There are  $n$  preemptive jobs to be processed at two stages in the same technological order, first at stage 1 then at stage 2. At stage 1 there are  $m$  parallel unrelated machines, stage 2 has one machine. A job upon finishing its processing at stage 1 is ready to be processed at stage 2; it may be processed at stage 2 when the machine is available there, or it may reside in a buffer space of unlimited capacity following stage 1 until the machine at stage 2 becomes available. At stage 1, a job can be processed on any of the parallel machines, and its processing times may be different

on different machines. The processing times of job  $j$  ( $j=1, \dots, n$ ) are equal to  $p_{ij}$  (if it is executed on machine  $i$  ( $i=1, \dots, m$ )) and  $s_j$  time units, respectively, at stage 1 and at stage 2. The processing of a job on a machine of stage 1 may be interrupted at any moment and resumed later on the same or another machine. A job during its processing at stage 1 does not need a resource or uses one unit of this resource (0-1 resource requirements). There are  $l$  types of resources. A resource of type  $r$  ( $r=1, \dots, l$ ) is available in an amount limited to  $W_r$  units at a time. The total usage of resource  $r$  at any moment by jobs simultaneously executed on parallel machines cannot exceed the availability of this resource. The objective is to find a feasible schedule which minimizes makespan,  $C_{\max}$ , which is equal to the maximum job completion time at stage 2.

The considered problem is NP-hard in the strong sense since the problem of preemptive scheduling in the two-stage flowshop with two identical parallel machines at one stage and one machine at another is NP-hard in the strong sense (Hoogeveen et al., 1996).

The heuristic algorithms proposed for the considered problem, while solving to optimality the resource constrained scheduling problem at the first stage of the flowshop, select for simultaneous processing jobs according to rules promising a good (short) schedule in the flowshop. Several rules of job selection are considered.

The problem under consideration arises in real-life systems that are encountered in a variety of industries, e.g. in chemical, food, cosmetics and textile industries. These systems are often subjected to some additional resource constraints for example on the availability of the additional resources such as skilled labour and tools. Preemption of jobs usually results in shortening the schedule. The problem with parallel unrelated machines at the first stage and a single machine at the second stage may arise in a manufacturing environment in which products are initially processed on any of parallel machines and then each product must go through a final testing operation, which is to be carried out on a common testing machine.

## 2 FRAMEWORK OF THE HEURISTIC ALGORITHMS

The proposed heuristic algorithms proceed in the following steps:

1. A linear programming (LP) problem is solved to minimize time  $T$  needed for finishing all jobs at stage 1 of the flowshop under relaxed resource constraints over time  $T$ . As a result, the minimal value of  $T$  and the values of the time  $t_{ij}$  ( $i = 1, \dots, m, j = 1, \dots, n$ ) during which job  $j$  is processed on machine  $i$  are obtained.
2. Using the values of  $t_{ij}$  obtained in Step 1 as well as the values of  $s_j$  ( $s_j$  is the processing time of job  $j$  at stage 2), weights,  $w_j$ , for all jobs are determined on the basis of 6 different expressions presented in Table 1.
3. The schedule at stage 1 of the flowshop is constructed in the form of a sequence of partial schedules using the values of  $T$ ,  $t_{ij}$ , and  $w_j$ . In a partial schedule at most  $m$  ( $m$  is the number of machines) jobs are assigned to machines for simultaneous processing during some period of time so that resource constraints are satisfied at every moment. The consecutive

partial schedules are created in subsequent iterations of an iterative procedure. Assignment of jobs to machines in a partial schedule is found maximizing the weighted assignment  $\sum_{i=1}^m \sum_{j=1}^n w_j v_{ij}$  ( $v_{ij} = 1$  if job  $j$  is processed on machine  $i$  in a current partial schedule, and 0 otherwise) under resource constraints. In each created partial schedule, conditions on optimality formulated in (Slowinski, 1980, 1981) are satisfied.

4. Completion times of jobs at stage 1 are calculated.
5. A schedule on the machine of stage 2 is constructed using the values of  $s_j$ , and ready times of jobs at stage 2, which are equal to corresponding completion times at stage 1.

Table 1: Weights used in the heuristic algorithms.

Algorithm	Weight $w_j$ of job $j$
A1	1
A2	random number from $U[0,1]$
A3	$\frac{\min_{k=1, \dots, n} \{Z_k\}}{Z_j}$
A4	$\frac{s_j}{\max_{k=1, \dots, n} \{s_k\}}$
A5	$\frac{s_j}{Z_j} \frac{\min_{k=1, \dots, n} \{Z_k\}}{\max_{k=1, \dots, n} \{s_k\}}$
A6	$\frac{\min_{k=1, \dots, n} \{Z_k\}}{Z_j} + 1$ if $Z_j \leq s_j$ , and $\frac{s_j}{\max_{k=1, \dots, n} \{s_k\}}$ if $Z_j > s_j$

$Z_j$  is the processing time of job  $j$  at stage 1,  $Z_j = \sum_{i=1}^m t_{ij}$ . In A3, A4 and A5 the maximal value of  $w_j$  is equal to 1, in A6 the maximal  $w_j$  is equal to 2 if  $Z_j \leq s_j$ , and 1 if  $Z_j > s_j$ .

## 3 ILLUSTRATIVE EXAMPLE

To illustrate the problem and the solution method we present the following example. Consider the case of the two-stage flowshop with 2 machines at stage 1 and a single machine at stage 2. The number of jobs  $n=10$ , the resource availability at any moment,  $W_1=1$ . Job processing times and resource requirements are shown in Figure 1.

job	machine	
	1	2
1	6	10
2	14	9
3	12	8
4	13	10
5	6	12
6	23	16
7	6	8
8	13	9
9	19	22
10	8	23

job	machine 1
1	1
2	8
3	10
4	6
5	6
6	7
7	2
8	9
9	1
10	1

job	1
1	1
2	1
3	0
4	1
5	0
6	0
7	0
8	0
9	1
10	1

resource availability = 1

Figure 1: Data for an illustrative example.

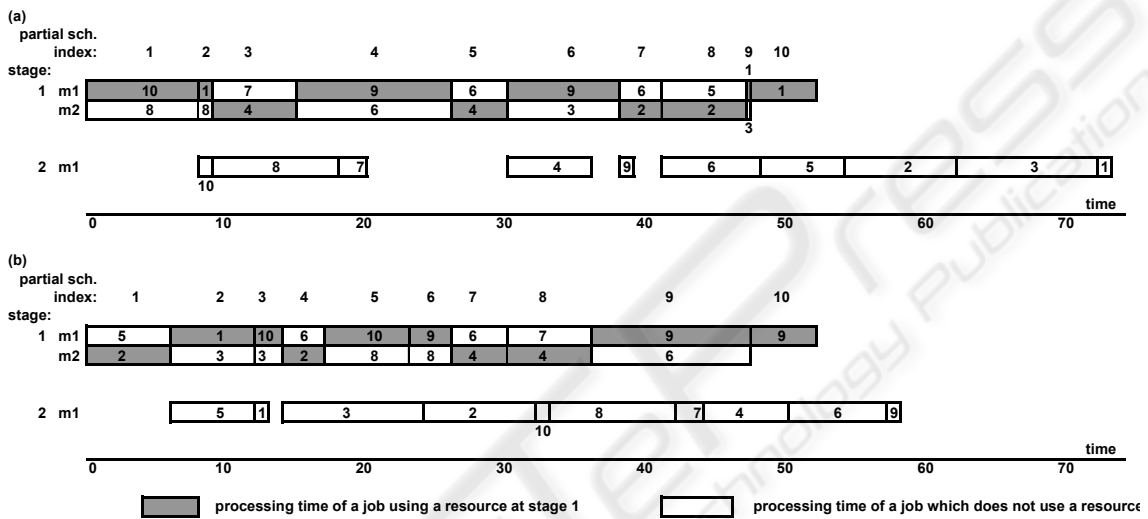


Figure 2: An illustrative example. The resulting schedules: (a) a schedule obtained by heuristic algorithm A1, (b) a schedule obtained by heuristic algorithm A5.

Figure 2 presents two flowshop schedules for this instance, one created by algorithm A1 (Figure 2a) and another created by A5 (Figure 2b). The first stage schedules in Figures 2a and 2b are composed of 10 partial schedules. In each of these partial schedules at most 2 jobs are processed simultaneously and the total resource usage does not exceed the resource availability,  $W_1 = 1$ , e.g. in the partial schedule of index 1 in Figure 2a, jobs 10 and 8 are processed simultaneously and use at every moment 1 and 0 units of the resource, respectively. The first stage schedules in Figures 2a and 2b have the same length, but completion times of jobs in these schedules are different. For example, in the first stage schedule in Figure 2a, job 10 finishes its processing at stage 1 at 8 time units and in Figure 2b - at 23 time units. This results in different lengths of the flowshop schedules provided by A1 and A5.

## 4 COMPUTATIONAL STUDY

In this section, the results of a computational experiment conducted to evaluate the performance of the proposed heuristic algorithm are presented. 720 randomly generated instances were created and examined. Instances were generated for  $n = 50, 100, 150,$  and  $200$ ,  $m = 2, 4,$  and  $6$ , and for one resource type. The resource availability  $W_1$  was set at  $m/2$ , and for 55% of jobs resource requirements were set at 1. Processing times at stage 2,  $s_j$ , were generated from  $U[1, 100]$  ( $U[a, b]$  denotes the discrete uniform distribution in the range of  $[a, b]$ ) for all instances. Processing times at stage 1,  $p_{ij}$ , were generated from 9 ranges:  $U[30, 120]$ ,  $U[45, 180]$ , and  $U[60, 240]$  for instances with  $m = 2$ ,  $U[90, 360]$ ,  $U[120, 480]$ , and  $U[150, 600]$  for  $m = 4$ ,  $U[150, 600]$ ,  $U[200, 800]$ , and  $U[250, 1000]$  for  $m = 6$ . The ranges

for processing times were selected so as to include the cases when the length of the optimal schedule at stage 1,  $C_1^*$ , is close to the sum of job processing times at stage 2,  $\sum_{j=1}^n s_j$ .

As an effectiveness measure we use the relative percentage deviation of a heuristic solution from the lower bound on the optimal makespan defined as

$$\delta = \frac{C_{\max} - LB}{LB} \times 100\%$$

where  $LB = \max\{LB_1, LB_2\}$ , where  $LB_1 = C_1^* + \min_{j=1, \dots, n} (s_j)$  and  $LB_2 = \min_{i=1, \dots, m, j=1, \dots, n} (p_{ij}) + \sum_{j=1}^n s_j$ ,  $C_1^*$  is the minimal makespan at stage 1.

Table 2: Computational results.

n	m	$\delta$ (%)						
		A1	A2	A3	A4	A5	A6	
50	2	1.38	1.40	0.06	0.75	0.06	0.04	
		2.78	3.35	1.30	0.54	0.12	0.06	
		4.15	4.43	2.74	0.27	0.20	0.14	
	4	3.26	1.94	0.31	0.46	0.15	0.40	
		7.21	4.60	3.77	0.60	0.52	0.60	
		7.68	6.25	5.81	0.82	0.89	0.74	
	6	2.65	1.39	0.25	0.52	0.16	0.52	
		8.24	5.16	4.64	0.87	0.77	0.84	
		8.90	7.80	6.87	1.55	1.68	1.51	
	100	2	0.74	0.51	0.03	0.39	0.02	0.01
			1.80	2.49	0.60	0.21	0.04	0.02
			1.52	1.33	1.00	0.03	0.05	0.02
4		1.26	1.08	0.08	0.26	0.08	0.25	
		5.06	3.18	1.87	0.18	0.22	0.18	
		3.45	2.70	2.50	0.17	0.21	0.16	
6		2.03	1.01	0.06	0.23	0.07	0.25	
		5.64	4.03	2.46	0.40	0.31	0.42	
		4.02	4.03	3.76	0.47	0.47	0.43	
150		2	0.21	0.42	0.02	0.19	0.02	0.00
			1.62	1.16	0.28	0.19	0.02	0.01
			0.95	0.90	0.91	0.02	0.02	0.02
	4	0.71	0.61	0.00	0.22	0.01	0.16	
		4.24	2.05	1.44	0.15	0.10	0.19	
		1.94	1.60	1.37	0.06	0.07	0.06	
	6	0.76	0.49	0.04	0.20	0.01	0.20	
		4.14	2.36	1.86	0.24	0.17	0.24	
		2.66	2.46	2.17	0.18	0.20	0.19	
	200	2	0.46	0.42	0.01	0.21	0.02	0.00
			1.10	0.99	0.17	0.30	0.01	0.01
			0.69	0.48	0.38	0.01	0.01	0.01
4		0.60	0.44	0.04	0.19	0.03	0.10	
		2.15	1.70	0.86	0.12	0.05	0.12	
		1.47	1.22	1.16	0.05	0.05	0.05	
6		0.83	0.25	0.02	0.10	0.03	0.10	
		2.46	1.57	0.90	0.12	0.07	0.12	
		1.75	1.67	1.64	0.10	0.10	0.10	

The results of a computational experiment are presented in Table 2. All entries in this table are average values over 20 instances.

From Table 2, we can observe that deviations,  $\delta$ , significantly decrease, as the number of jobs grows, and they increase with the number of machines. We can see that algorithms A3, A4, A5, and A6 always outperform A1 and A2, and A4, A5, and A6 produce near-optimal solutions. On the average over the entire collection of instances, relative deviations of the heuristic makespan from its lower bound are equal to 2.79%, 2.15%, 1.43%, 0.32%, 0.19%, and 0.23% for A1, A2, A3, A4, A5, and A6, respectively.

The CPU times are small for all the heuristic algorithms and equal to about 0.3, 1.5, 3, 4.5, and 6.5 seconds for  $n=50, 100, 150,$  and  $200,$  respectively.

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