



Extending the Single Machine-Based Relaxation Scheme for the Job Shop Scheduling Problem

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Abstract

The contribution of this paper to the job shop related literature is twofold. First, we provide an efficient way for solving the job shop scheduling problem with release dates, delivery times and delayed precedence constraints. It is shown that the latter problem is equivalent to a classical job shop with precedence constraints. Second, an effective extension of the standard single machine-based relaxation scheme is derived. Preliminary computational results conducted on a set of benchmark instances show that effective multiple machine-based lower bounds can be computed within reasonable CPU time.

Keywords: Lower Bounds, Release Dates, Delivery Times, Delayed Precedence Constraints.

1 Introduction

We address the job shop scheduling problem, denoted by $J||C_{max}$, where n jobs (J_1, \dots, J_n) have to be scheduled without preemption on m machines M_1, \dots, M_m , with the objective of minimizing the makespan. We denote by O_{ij} the operation of job J_j that has to be processed on machine M_i , and by p_{ij} its processing time. Each job is characterized by its proper routing on the machines.

Numerous heuristics have been proposed for this \mathcal{NP} -hard problem. However, only few exact branch-and-bound algorithms have been published. It is well-known that the performance of branch-and-bound algorithms strongly relies on the effectiveness/efficiency of the implemented lower bounds. Unfortunately, the lack of effective relaxation schemes seems to be the Achilles' heel in job shop-related research. Indeed, the overwhelming majority of the published lower bounds are based on the single machine relaxation scheme. Although this latter scheme has been enhanced by various adjustment and propagation techniques, the best results obtained so far for open benchmark instances still fail to close the gap.

This paper constitutes a first attempt to extend the single machine-based relaxation scheme to the multiple machines variant. We propose an efficient procedure to solve the multiple machines subproblem with release dates, delivery times, and delayed precedence constraints. The instance is nontrivially transformed into a (classical) $J|prec|C_{max}$ one with an expanded number of jobs and machines, but with a maximum of three operations per job. Interestingly, the obtained instance proves to be efficiently solvable by the branch-and-bound algorithm of Brucker et al. [1]. Indeed, our preliminary computational results show that such instances with more than 180 jobs and 120 machines are optimally solved within an average CPU time of 0.12 seconds.

2 The single machine subproblem

A classical and widely used job shop relaxation scheme consists in relaxing the constraint that each machine can process at most one job at a time, for all machines but one (say M_k). The obtained problem is a single machine

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problem (denoted by $1|r_j, q_j|C_{max}$) with release dates and delivery times. The release date r_j of a job j is equal to the length of the longest path from the source node to node O_{kj} in the disjunctive graph. Similarly, the delivery time q_j is such that $p_{kj} + q_j$ is equal to the length of the longest path from the O_{kj} node to the sink node in the disjunctive graph. Although the latter problem is strongly \mathcal{NP} -hard, its optimal makespan, denoted hereafter by $C_0^*(M_k)$, can be efficiently computed by using effective existing branch-and-bound algorithms [2], [4]. This yields the standard single machine-based lower bound:

$$LB_0 = \max_{k=1, \dots, m} C_0^*(M_k)$$

Interestingly, several procedures exist which aim at identifying precedence relationships between operations that have to be scheduled on the same machine in a $J||C_{max}$ instance (see for instance [1]). An immediate output of using these techniques is that delayed precedence constraints may occur between operations on the same machine. Indeed, let L_{ij} denote the length of the longest path between operations O_{ki} and O_{kj} in the disjunctive graph. Then, O_{kj} cannot start processing before $a_{ij} = L_{ij} - p_{ki}$ units of time after the completion of O_{ki} . Dauzère-Pérès [3] devised an efficient branch-and-bound algorithm for solving the single machine problem with heads, tails and delayed precedence constraints (denoted by $1|r_j, q_j, del - prec|C_{max}$). Let $C_1^*(M_k)$ denote the optimal makespan of the $1|r_j, q_j, del - prec|C_{max}$ defined on M_k . A valid lower bound for the $J||C_{max}$ that dominates LB_0 is

$$LB_1 = \max_{k=1, \dots, m} C_1^*(M_k)$$

3 The multiple machines subproblem

Consider the relaxation scheme which consists in relaxing the capacity constraint of all the machines except for a subset of K machines (say for simplicity M_1, M_2, \dots, M_K). The obtained problem is a $J_K|r_j, q_j, del - prec|C_{max}$ where the release dates r_j , the delivery times q_j , and the time lags a_{ij} are computed in a similar way than it is described in the previous section. Let Δ denote the set of delayed precedence constraints. For the sake of clarity, we will denote w.l.o.g. by O_1, O_2, \dots, O_T the operations that have to be processed on the non relaxed machines M_1, M_2, \dots, M_K . Interestingly, we state the following result:

Theorem 1: The $J|r_j, q_j, del - prec|C_{max}$ is equivalent to the $J|prec|C_{max}$.

Proof: Let (P_1) denote a $J|r_j, q_j, \text{del} - \text{prec}|C_{\max}$ instance. In the following, we show how to construct a $J|\text{prec}|C_{\max}$ instance, denoted hereafter by (P_2) , which is equivalent to (P_1) . The set of machines of (P_2) includes machines M_1, M_2, \dots, M_K . Moreover, to each operation O_i ($i = 1, \dots, T$) in (P_1) are associated two machines M_i^p and M_i^s in (P_2) . Machine M_i^p has to process all the operations that precede O_i , and machine M_i^s has to process all the operations that succeed to O_i . In addition, to each delayed precedence constraint between operations O_i and O_j in (P_1) is associated a machine D_{ij} . The set of jobs of (P_2) is partitioned into the four following subsets:

- Subset 1 (T jobs): each job is composed of three operations O_i^p, O_i , and O_i^s ($i = 1, \dots, T$) which have to be processed (in that order) on machines $M_i^p, M_{(O_i)}$, and M_i^s , respectively, where $M_{(O_i)} \in \{M_1, M_2, \dots, M_K\}$ denotes the machine on which operation O_i has to be processed in (P_1) . Operations O_i^p and O_i^s have zero processing times while O_i requires p_i units of processing time.
- Subset 2 ($|\Delta|$ jobs): each job consists of three operations $\delta_{ij}^p, \delta_{ij}$, and δ_{ij}^s ($(O_i, O_j) \in \Delta$) which have to be processed on machines M_i^s, D_{ij} and M_j^p , respectively. Machine D_{ij} has to process only operation δ_{ij} for a_{ij} units of time. Operations δ_{ij}^p and δ_{ij}^s have zero processing times. Moreover, δ_{ij}^p has to succeed to O_i^s , and δ_{ij}^s has to precede O_j^p .
- Subset 3 (T jobs): each job consists of a single operation α_i ($i = 1, \dots, T$) which has to be processed before all operations on machine M_i^p for r_i units of time.
- Subset 4 (T jobs): each job consists of a single operation β_i ($i = 1, \dots, T$) which has to be processed after all operations on machine M_i^s for q_i units of time.

It suffices to prove that in an optimal (active) schedule of (P_2) , the start time of operation O_i is equal to $\max\{r_i, \max_{(O_j, O_i) \in \Delta} (C(O_j) + a_{ji})\}$ (where $C(\cdot)$ stems for completion time), and operation β_i starts processing at $C(O_i)$ (which expresses the delivery time constraint). The remainder of the proof is omitted for lack of available space.

An immediate consequence of Theorem 1 is that the $J|r_j, q_j, \text{del} - \text{prec}|C_{\max}$ can be optimally solved by applying a branch-and-bound algorithm on the equivalent $J|\text{prec}|C_{\max}$. In our implementation, the latter problem is solved by using the branch-and-bound algorithm of Brucker et al. [1]. Although the latter algorithm is initially designed for the $J||C_{\max}$, it can be immediately adapted to the $J|\text{prec}|C_{\max}$. Indeed, the disjunctive graph of the $J|\text{prec}|C_{\max}$

is no more than that of the $J||C_{\max}$ where some of the disjunctive arcs are fixed.

The reader may have remarked the huge size of the obtained $J|prec|C_{\max}$ instance ($3T + |\Delta|$ jobs and $2T + K + |\Delta|$ machines). On the other hand, it should be noticed that each job has a maximum number of 3 operations to be performed, which substantially reduces the computational burden. Interestingly, the problem size can be fairly reduced by considering the non-delayed precedence constraints. That is, for a given $(O_i, O_j) \in \Delta$ such that $a_{ij} = 0$, there is no need for operation δ_{ij} neither for machine D_{ij} (i.e. the corresponding subset 2 job has only operations δ_{ij}^p and δ_{ij}^s). If, in addition, O_i and O_j are processed on the same machine, then operations δ_{ij}^p and δ_{ij}^s can be removed and replaced by a conjunctive arc from O_i to O_j . Furthermore, the following results (stated with omitted proofs) consistently improve the performance of the optimization procedure. The first two lemmas show that the only disjunctive arcs that need to be fixed by the branch-and-bound algorithm are those which are related to machines M_1, M_2, \dots, M_K ; while the third one proves that the preemptive single machine-based lower bound $C_0^{(p)}(.)$ (which is implemented in the branch-and-bound algorithm of [1]) only needs to be computed over machines M_1, M_2, \dots, M_K .

Lemma 2: Any sequence of operations δ_{ji}^s ($(O_j, O_i) \in \Delta$) on M_i^p is optimal ($i = 1, \dots, T$).

Lemma 3: Any sequence of operations δ_{ij}^p ($(O_i, O_j) \in \Delta$) on M_i^s is optimal ($i = 1, \dots, T$).

Lemma 4: $\max\{C_0^{(p)}(M_i^p), C_0^{(p)}(M_i^s), \max_{(O_i, O_j) \in \Delta} (C_0^{(p)}(D_{ij}))\} \leq C_0^{(p)}(M_{(O_i)})$ for $i = 1, \dots, T$.

Example 1: Figure 1 displays the disjunctive graph of the $J|prec|C_{\max}$ that is derived from a 5 operation-2 machine $J|r_j, q_j, del - prec|C_{\max}$ instance, where $\Delta = \{(O_1, O_2), (O_1, O_3), (O_3, O_4), (O_5, O_4)\}$. For the sake of clarity, the source node and some of the sink-related conjunctive arcs are not displayed.

For a given value of K , let π_h ($h = 1, \dots, \binom{K}{m}$) denote the h^{th} $J_K|r_j, q_j, del - prec|C_{\max}$ instance that is derived from the $J_m||C_{\max}$ one, and let $C^*(\pi_h)$ denote its optimal makespan. Therefore, a valid K -machine-based lower bound for the $J_m||C_{\max}$ is:

$$LB_K = \max_{h=1, \dots, \binom{K}{m}} C^*(\pi_h)$$

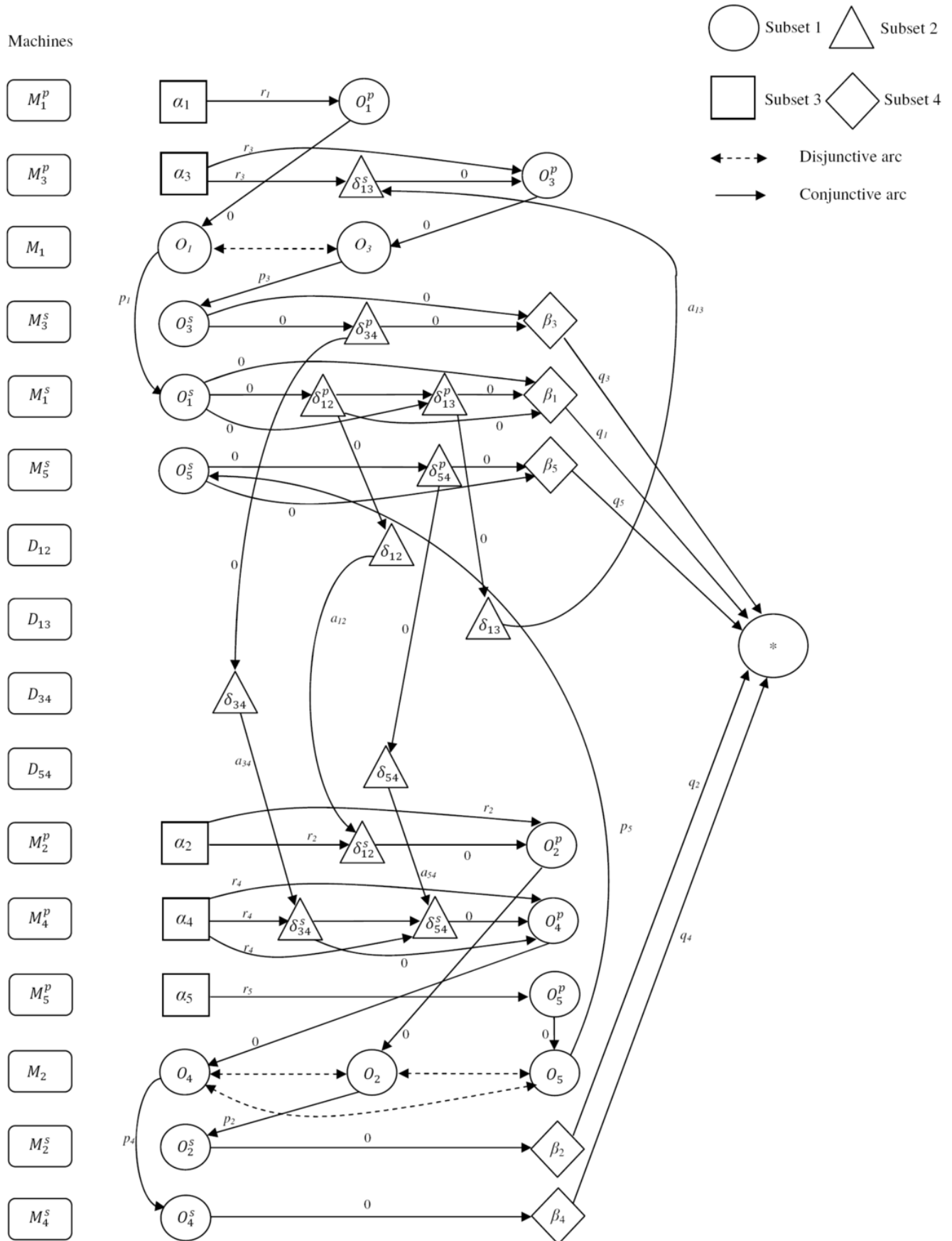


Figure 1. The $J|precl|C_{max}$ disjunctive graph of Example 1.

In our implementation, LB_K ($K \geq 2$) is computed as follows: Let C denote a trial value on the optimal makespan of the $J||C_{\max}$. Initially, the value of C is set to LB_{K-1} . For each value of C , a deadline $d_{ij} = C - q_{ij}$ is set for each operation O_{ij} and the adjustment procedure described in [1] is applied in order to identify precedence constraints. Then, the branch-and-bound algorithm of Brucker et al. [1] is run for each π_h (starting from $h = 1$) after setting at the root node an artificial upper bound $UB = C + 1$. The branch-and-bound algorithm is stopped as soon as a feasible schedule with makespan less than UB has been found. In that case, the computation of $C^*(\pi_h)$ will not be useful in finding a lower bound better than C , and π_h need not to be solved for larger values of C . Otherwise, we will have $C^*(\pi_h) = UB$ i.e. there is no feasible schedule of π_h with makespan less than or equal to C . Therefore, C is incremented by one unit and the procedure is restarted from π_h . The procedure is stopped, yielding $LB_K = C$, when a feasible schedule with makespan less than or equal to C has been found for $\pi_{\binom{K}{m}}$.

4 Preliminary computational results

The effectiveness/efficiency of the implemented multiple machine-based lower bounds (namely LB_2 , LB_3 and LB_4) with respect to the best single machine-based one (namely LB_1) is evaluated on a set of selected benchmark instances [5] for which LB_1 is not equal to the optimal makespan. The results are depicted in Table 1 where we provide, for each lower bound $LB_K \in \{LB_2, LB_3, LB_4\}$:

- Gap_{red} : the percentage by which the gap between LB_1 and the optimal/best found makespan (C_{best}) has been reduced i.e.

$$Gap_{red} = 100(LB_K - LB_1)/(C_{best} - LB_1)$$

The bold values denote the fact that $LB_K > LB_{K-1}$.

- $Time$: the total CPU time on a Quad 2.8 GHz Personal Computer with 4 GB RAM (including the CPU time of LB_{K-1}). The value between parentheses denotes the average CPU time that is required for solving a single $J_K|r_j, q_j, del - prec|C_{\max}$ subproblem.

Table 1 provides evidence that the multiple machine-based lower bounds consistently outperform the single machine-based one, while requiring a (surprisingly) moderate CPU time. For the sake of illustration, we observe that, for all the largest instances (20 jobs and 15 machines), the four machine-based lower bound is computed in about one minute, although it requires

	LB_2		LB_3		LB_4	
<i>Instance (nxm)</i>	<i>Gap_{Red}</i>	<i>Time</i>	<i>Gap_{Red}</i>	<i>Time</i>	<i>Gap_{Red}</i>	<i>Time</i>
<i>ABZ5</i> (10x10)	48.06	0.18 (-)	51.46	0.48 (-)	70.87	1.69 (-)
<i>ABZ6</i> (10x10)	50.93	0.14 (-)	54.63	0.44 (-)	74.07	1.33 (-)
<i>ABZ7</i> (20x15)	0.00	0.36 (-)	0.00	2.96 (-)	0.00	22.31 (0.01)
<i>ABZ8</i> (20x15)	25.00	0.74 (-)	27.94	4.78 (-)	33.82	69.06 (0.04)
<i>ABZ9</i> (20x15)	22.22	0.42 (-)	22.22	4.02 (-)	30.16	51.26 (0.03)
<i>FT6</i> (6x6)	100.00	0.02 (-)	100.00	0.05 (-)	100.00	0.07 (-)
<i>FT10</i> (10x10)	49.18	0.14 (-)	54.10	0.47 (-)	72.95	2.26 (-)
<i>La4</i> (10x5)	69.57	0.03 (-)	69.57	0.06 (-)	91.30	0.14 (-)
<i>La16</i> (10x10)	48.57	0.11 (-)	65.71	0.50 (-)	80.00	1.25 (-)
<i>La17</i> (10x10)	81.08	0.11 (-)	81.08	0.37 (-)	81.08	1.02 (-)
<i>La18</i> (10x10)	69.23	0.17 (-)	69.23	0.37 (-)	69.23	0.91 (-)
<i>La19</i> (10x10)	36.84	0.13 (-)	44.36	0.48 (-)	57.14	1.49 (-)
<i>La20</i> (10x10)	46.32	0.12 (-)	49.47	0.43 (-)	54.74	1.32 (-)
<i>La21</i> (15x10)	74.51	0.15 (-)	74.51	1.10 (-)	74.51	4.15 (0.01)
<i>La22</i> (15x10)	0.00	0.09 (-)	0.00	0.45 (-)	57.14	5.26 (0.02)
<i>La24</i> (15x10)	14.81	0.13 (-)	16.67	0.78 (-)	25.93	3.75 (0.01)
<i>La25</i> (15x10)	43.37	0.28 (-)	49.40	1.02 (-)	61.45	11.30 (0.04)
<i>La29</i> (20x10)	13.16	0.13 (-)	13.16	1.25 (-)	13.16	95.49 (0.44)
<i>La36</i> (15x15)	20.45	0.27 (-)	52.27	2.25 (-)	59.09	11.24 (-)
<i>La38</i> (15x15)	25.21	0.35 (-)	35.29	2.63 (-)	42.02	16.87 (0.01)
<i>La39</i> (15x15)	0.00	0.28 (-)	0.00	2.43 (-)	0.00	17.21 (0.01)
<i>La40</i> (15x15)	42.31	0.38 (-)	42.31	2.86 (-)	53.85	21.52 (0.01)
<i>Average</i>	40.04	0.21 (-)	44.24	1.37 (-)	54.66	15.50 (0.03)

(-) means that the CPU time is less than 0.01 sec.

Table 1. Performance of the multiple machine-based lower bounds

solving more than $1365 J_4|r_j, q_j, del - prec|C_{\max}$ subproblems. Finally, it is worth noting that the $J_K|r_j, q_j, del - prec|C_{\max}$ subproblems are solved within an extremely short average CPU time, frequently less than 0.01 seconds and scarcely exceeding 0.03 seconds.

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