# **Lower Bounds for the Two-Machine Flow Shop Problem with Time Delays**

**Mohamed Amine Mkadem, Aziz Moukrim and Mehdi Serairi**

**Abstract** We consider the flow shop problem with two machines and time delays with respect to the makespan, i.e., the maximum completion time. We recall the lower bounds of the literature and we propose new relaxation schemes. Moreover, we investigate a linear programming-based lower bound that includes the implementation of a new dominance rule and a valid inequality. A computational study that was carried out on a set of 480 instances including new hard ones shows that our new relaxation schemes outperform the state of the art lower bounds.

### **1 Introduction**

This paper is devoted to dealing with the flow shop scheduling problem with two machines and time delays, denoted by  $F2|l_j|C_{max}$ . Let  $I = (J, p_1, l, p_2)$  be an instance<br>of  $F2|l_j|C_{max}$ , and  $p_j$  are the vectors of of  $F2|l_j|C_{max}$ , where  $J = \{1, 2, ..., n\}$  is a set of *n* jobs,  $p_1$  and  $p_2$  are the vectors of processing times on the first and the second machines and *l* is the vector of the time processing times on the first and the second machines, and *l* is the vector of the time delays. Each job *j* has two operations. The first operation (resp. the second operation) must be executed without preemption during  $p_{1j}$  (resp.  $p_{2j}$ ) time units on  $M_1$  (resp.  $M_1$ ). For each job  $i \in I_1$  a time delay of at minimum *l* time units must separate the *M*<sub>2</sub>). For each job *j* ∈ *J*, a time delay of at minimum *l<sub>j</sub>* time units must separate the end of the first operation and the start of the second one. The objective is to find a end of the first operation and the start of the second one. The objective is to find a feasible schedule that minimizes the completion time of the last scheduled job on  $M_2$ . A feasible schedule is such that at most one operation is processed at a time on a priven machine. In addition, the operations are executed without preemption, where given machine. In addition, the operations are executed without preemption, where interruption and switching of operations are not allowed.

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Mitten [\[2](#page-6-0)] proves that the permutation flow shop  $F2\pi l_j |C_{max}$ , where a feasible consists in having the same job sequence on both machines, can be solved schedule consists in having the same job sequence on both machines, can be solved in polynomial time. However, solving our problem as a permutation flow shop does not necessarily provide an optimal solution.  $F2|l_j|C_{max}$  is an NP-hard problem in the strong sense even with unit-time operations [3] strong sense even with unit-time operations [\[3](#page-6-1)].

The objective of this paper is to introduce new lower bounds. First, we improve the most promising lower bound of the literature. Second, we investigate a linear programming-based lower bound.

#### **2 Combinatorial Lower Bounds**

We present here the lower bounds of the literature and propose new ones. Hereafter,  $C^*_{max}(I)$  represents the optimal makespan value of instance *I* and  $C_{max}(S)$  stands for the makespan value of schedule *S* the makespan value of schedule *S*.

First, we survey four lower bounds of Yu [\[3\]](#page-6-1). We start by two  $O(n)$  basic lower  $nds$   $B_r = \max_{r} (n_r + l_r + n_s)$  and  $IB_r = \max(\sum_{r=1}^{n} n_r + \min_{r=1} (l_r + n_s))$ bounds  $LB_1 = \max_{1 \le j \le n} (p_{1,j} + l_j + p_{2,j})$  and  $LB_2 = \max(\sum_{j=1}^n p_{1,j} + \min_{1 \le j \le n} (l_j + p_{2,j}))$ <br> $\sum_{j=1}^n p_{1,j} + \min_{1 \le j \le n} (l_j + p_{2,j})$ *j*<sup> $n$ </sup> +  $p_{2j}$  + min<sub>1≤*j*≤*n*</sub>(*l<sub>j</sub>* + *p*<sub>1*j*</sub>)). Moreover, Yu [\[3](#page-6-1)] interested in the problem where  $c$ h iob *i* ∈ *I* is splitted into min(*n*, , , *n*, .) unitary sub-iobs. The lower bound *I.B*, = each job *j* ∈ *J* is splitted into min( $p_{1,j}, p_{2,j}$ ) unitary sub-jobs. The lower bound  $LB_3 =$   $\lceil ( \sum_{i=1}^{n} p_{1i} ) (p_{2i}, p_{3i}) \rceil + 1 + \sum_{i=1}^{n} p_{1i} (p_{2i}, p_{3i})$  was introdu- $[(\sum_{j=1}^{n} \min(p_{1,j}, p_{2,j}), l_j^u)/\sum_{j=1}^{n} \min(p_{1,j}, p_{2,j})] + 1 + \sum_{j=1}^{n} \min(p_{1,j}, p_{2,j})$  was introduced where  $l^u - l + \max(n, n) - 1$  is the time delay observed by each sub-job ced, where  $l_j^u = l_j + \max(p_{1,j}, p_{2,j}) - 1$  is the time delay observed by each sub-job derived from  $i \in I$ derived from  $j \in J$ .

The fourth lower bound is presented as follows. Let *S*<sup>∗</sup> be an optimal schedule and  $p_{k}$ [ $\ell$ ] the processing time of the job scheduled at position  $\ell$  on  $M_k$ ,  $k \in$ {1, 2}. Moreover, let *j*<sup>k</sup> be the position of job *j* on  $M_k$ ,  $k \in \{1, 2\}$ . For each job  $j \in J$ , it holds that  $C_{max}(S^*) \ge \sum_{\ell=1}^{j^1} p_{1,\ell} + l_j + \sum_{\ell=j^2}^n p_{2,\ell}$ . By adding together<br>the above equations for all jobs and by considering that the makespan is integral, *LB*4 <sup>=</sup> ⌈( ∑*<sup>n</sup> <sup>j</sup>*=1 *lj* <sup>+</sup> <sup>∑</sup>*<sup>n</sup> <sup>m</sup>*=1 1*,<sup>m</sup>* <sup>+</sup> <sup>∑</sup>*<sup>n</sup> <sup>m</sup>*=1 2*,<sup>m</sup>*)∕*n*⌉ is a valid lower bound, where *<sup>k</sup>,<sup>m</sup>* is the sum of the *m* smallest values in  $\{p_{k,1}, p_{k,2}, \ldots, p_{k,n}\}$ .<br>The following lower bounds were introduced by Del

The following lower bounds were introduced by Dell'Amico [\[1](#page-6-2)]. In the first one, it is assumed that all jobs are executed at time 0 on  $M_1$ . The problem is then a<br>single machine scheduling problem with release dates denoted by  $1|x|C = I$  at single-machine scheduling problem with release dates denoted by  $1|r_j|C_{max}$ . Let <br>*I* be the instance for  $1|r_j|C_{min}$  with  $r = n_j + l_j$  and  $n = n_j$ ,  $i \in I$  Obviously *I<sub>r</sub>* be the instance for  $1|r_j|C_{max}$  with  $r_j = p_{1,j} + l_j$  and  $p_j = p_{2,j}$ ,  $j \in J$ . Obviously,  $I = C^*$  (*I*) is a valid lower bound on the *F*2*U* [*C*] original instance, which can  $L_1 = C^*_{max}(I_r)$  is a valid lower bound on the  $F2|l_j|C_{max}$  original instance, which can<br>be computed in  $O(n \log n)$ -time by scheduling the jobs in a nondecreasing order of be computed in *<sup>O</sup>*(*<sup>n</sup>* log *<sup>n</sup>*)-time by scheduling the jobs in a nondecreasing order of *r<sub>j</sub>*, *j* ∈ *J*. By interchanging the role of *M*<sub>1</sub> and *M*<sub>2</sub>, we yield a symmetric lower bound called *I*. Finally, we define the lower bound *I B*<sub>2</sub> – max(*I*, *I*, ) called  $L_2$ . Finally, we define the lower bound  $LB_5 = \max(L_1, L_2)$ .<br>Solving our problem as a permutation flow shop does not necessary

Solving our problem as a permutation flow shop does not necessarily provide an optimal solution. However, special cases exist where it is true. Dell'Amico [\[1\]](#page-6-2) proved that permutation schedules are dominant if  $l_j \leq \min_{1 \leq i \leq n} (p_{1,i} + l_i)$ ,  $j \in J$  and then he<br>introduced the following lower hound  $I \propto \overline{I} - (I \cdot \overline{I} \cdot \overline{I})$  has a new instance that is introduced the following lower bound. Let  $\overline{I} = (J, p_1, \overline{l}, p_2)$  be a new instance that is derived from instance  $I = (I, p_1, p_2)$  where  $\overline{I} = \min(I, \min(I, + p_2))$ ,  $i \in I$ derived from instance  $I = (J, p_1, l, p_2)$ , where  $\overline{l}_j = \min(l_j, \min_{1 \le i \le n} (l_i + p_{1,i}))$ ,  $j \in J$ .<br>Since  $\overline{l}$  verifies Dell'Amico's [1] conditions, an ortimal solution for  $\overline{l}$  can be found Since *I* verifies Dell'Amico's [\[1\]](#page-6-2) conditions, an optimal solution for *I* can be found

in polynomial time using Mitten algorithm [\[2](#page-6-0)]. Therefore,  $LB_6 = C^*_{max}(I)$  is a valid lower bound lower bound.

Furthermore, we introduce two new lower bounds which can be considered as a generalization of *LB*<sub>6</sub>. In fact, Yu [\[3\]](#page-6-1) extended Dell'Amico's [\[1](#page-6-2)] result after showing<br>that the permutation schedules are dominant if  $l \leq min$ ,  $(l + max(n, n, \lambda))$ ,  $l \in$ that the permutation schedules are dominant if  $l_j \leq \min_{1 \leq i \leq n} (l_i + \max(p_{1,i}, p_{2,i}))$ ,  $j \in I$ <br>Therefore, from an instance  $I = (I, p_1, I, p_2)$  of  $F2|I \cdot C$  problem, we derive a *J*. Therefore, from an instance  $I = (J, p_1, l, p_2)$  of  $F2|l_j|C_{max}$  problem, we derive a pay instance  $\tilde{I}(I, p, \tilde{l}, p_1)$  where  $\tilde{l} = \min(l, \min(l, l, \max(l, p_1)))$   $i \in I$ . new instance  $I(J, p_1, l, p_2)$ , where  $l_j = \min(l_j, \min_{1 \le i \le n} (l_i + \max(p_{1,i}, p_{2,i}))), j \in J$ . As a consequence of Yu's [\[3\]](#page-6-1) result,  $LB_1^N = C_{max}^*(\tilde{I})$  is a valid lower bound on instance *I*. which is computed in  $O(n \log n)$ -time using Mitten [2]. *I*, which is computed in  $O(n \log n)$ -time using Mitten [\[2\]](#page-6-0).

Moreover, we consider two instances  $I = (J, p_1, l, p_2)$  and  $I' = (J', p_1, l, p_2)$  of  $I \cup C$  where  $I' \subset I$  Then any valid lower bound on *I'* is also a valid lower  $F2|l_j|C_{max}$ , where *J′*  $\subset$  *J*. Then, any valid lower bound on *I'* is also a valid lower bound on *I*  $\wedge$  new lower bound called *I*  $P^N$  can be obtained by invoking *I*  $P^N$  on dif bound on *I*. A new lower bound called  $LB_2^N$  can be obtained by invoking  $LB_1^N$  on different sub-instances of *I*. Interestingly, we consider *n* sub-instances. We start by the ferent sub-instances of *I*. Interestingly, we consider *n* sub-instances. We start by the original instance *I*, the next sub-instance is built from the one in hand by removing the job that has the minimum value of  $l_j + \max(p_{1,j}, p_{2,j}), j \in J$ .

#### **3 Linear Programming-Based Lower Bound**

We consider a mathematical formulation that is based on determining the precedence relationships between jobs on the two machines where it is supposed that the jobs are continuously processed on  $M_1$  and  $M_2$ . Indeed, any valid schedule on an  $F2|l_j|C_{max}$ <br>instance can be transformed to a schedule with the same makesnan value C where instance can be transformed to a schedule with the same makespan value *C* where jobs are continuously processed on the two machines from time 0 and from time  $C - \sum_{j=1}^{n} p_{2,j}$  on  $M_1$  and  $M_2$ , respectively.

<span id="page-2-0"></span>The decision variables are defined for each pair of jobs *i*,  $j \in J$ , where  $X_{i,j}^k$  takes the value 1 if *i* precedes *j* on  $M_k$  and 0 otherwise,  $k \in \{1, 2\}$ . Furthermore,  $C_{k,j}$  represents the completion time of job *j* on  $M_k$  and the total idle time on  $M_2$  is denoted by L. Using these definitions, the model can be formulated as follows:

$$
\mathbf{min} \qquad L \tag{1}
$$

s.t. 
$$
X_{i,j}^k + X_{j,i}^k = 1
$$
,  $\forall i, j \in J \ i \neq j; k \in \{1, 2\}$  (2)

$$
X_{i,j}^k \ge X_{i,v}^k + X_{v,j}^k - 1, \qquad \forall i, j, v \in J; k \in \{1, 2\}
$$
 (3)

$$
C_{1,i} = \sum_{j=1}^{n} p_{1,j} X_{j,i}^{1} + p_{1,i}, \quad \forall i \in J
$$
\n(4)

$$
C_{2,i} \ge C_{1,i} + l_i + p_{2,i}, \qquad \forall i \in J
$$
\n(5)

$$
C_{2,i} = L + \sum_{j=1}^{n} p_{2,j} X_{j,i}^2 + p_{2,i}, \forall i \in J
$$
\n(6)

$$
L \ge 0, C_{k,j} \ge 0, X_{i,j}^k \in \{0, 1\} \ \forall i, j \in J, k \in \{1, 2\}
$$
 (7)

The objective function [\(1\)](#page-2-0) minimizes the total idle time on  $M_2$ . Constraints [\(2\)](#page-2-0) ensure<br>that for each pair of jobs, one of them has to precede the other on each machine that for each pair of jobs, one of them has to precede the other on each machine. Constraints [\(3\)](#page-2-0) guarantee the absence of cyclic precedence relationships between all jobs. Constraints [\(4\)](#page-2-0) and [\(6\)](#page-2-0) take into account the job's precedence and enforce them to be processed continuously without idle on  $M_1$  and  $M_2$ , respectively. In addition,<br>Constraints (5) ensure that a job after being processed on M, has to wait its time delay Constraints [\(5\)](#page-2-0) ensure that a job after being processed on  $M_1$  has to wait its time delay<br>to be executed on  $M_2$ . The nature of decision variables  $I_1$ ,  $C_2$ , and  $X^k$  is displayed to be executed on  $M_2$ . The nature of decision variables *L*,  $C_{k,j}$  and  $X_{i,j}^k$  is displayed by Constraints (7) by Constraints [\(7\)](#page-2-0).

In order to strengthen the LP relaxation of the model, we propose a valid inequality, which is based on the additional waiting time that a job has to fulfill after being available for processing on  $M_2$ . We remark that given a sequence of jobs on  $M_1$ , solutions in which the jobs are scheduled on  $M_1$ , according to their arrival times are solutions in which the jobs are scheduled on  $M_2$  according to their arrival times are<br>dominant. Therefore, if a job *i* is preceded by a job *i* on *M*, then a lower bound dominant. Therefore, if a job *j* is preceded by a job *i* on  $M_1$ , then a lower bound<br>on the minimum additional waiting time observed by job *i* or job *i* is *w* where on the minimum additional waiting time observed by job *j* or job *i* is  $w_{i,j}$ , where (i)  $w_{i,j} = \max(0, l_i + p_{2,i} - p_{1,j} - l_j)$ , if  $l_i \le p_{1,j} + l_j$  (ii)  $w_{i,j} = \max(0, p_{1,j} + l_j + p_{2,j} - l_j)$  if  $l_i > p_{i,j} + l_j$ *l<sub>i</sub>*), if  $l_i > p_{1,j} + l_j$ .<br>A lower bound

A lower bound on the total additional waiting time  $\Delta$  can be obtained by solving the assignment problem where the assignment costs are  $\delta_{i,j}$ ,  $i, j \in \{1, ..., n\}$ . In the following we describe how the assignment costs are computed. Note that the first following, we describe how the assignment costs are computed. Note that the first scheduled job (resp. the last scheduled job) on  $M_1$  is assumed to be preceded (resp.<br>followed) by a dummy job (job 0, resp. job  $n + 1$ ). Obviously, since job 0 cannot followed) by a dummy job (job 0, resp. job  $n + 1$ ). Obviously, since job 0 cannot precede job *n* + 1, and a job cannot precede itself, then we set  $\delta_{0,n+1} = \infty$  and  $\delta_{i,j} =$ ∞*,* ∀*<sup>j</sup>* ∈ {1*,*…*, <sup>n</sup>*}.

*Remark 1* Let us consider an instance *I* of  $F2|l_j|C_{max}$  and *LB* (resp. *UB*) a lower hound (resp. an upper bound) on the value of the makespan. If a schedule of bound (resp. an upper bound) on the value of the makespan. If a schedule of makespan *LB* exists, then the jobs can be continuously processed without any idle time, from time 0 on  $M_1$  and from time  $(LB - \sum_{j=1}^n p_{2,j})$  on  $M_2$ . Then, we obtain the following assignment costs: following assignment costs:

- $\bullet$  ∀*i* ∈ {1*,*…*,n*}*,*  $\delta_{0,i} = \max(0, LB \sum_{j=1}^{n} p_{2,j} l_i p_{1,i}$ <br> $\bullet$  ∀*i* ∈ {1}  $n!$  if  $\sum_{i=1}^{n} p_{i} + l_{i} + p_{i} \geq \frac{IIB}{IIB}$  isometric
- $\forall i \in \{1, ..., n\}$ , if  $\sum_{j=1}^{n} p_{1,j} + l_i + p_{2,i} > UB$ , *i* cannot be processed at the last position on  $M_1$ , then  $\delta_{i,n+1} = \infty$ . Otherwise  $\delta_{i,n+1} = 0$

In order to set  $\delta_{i,j}$ ,  $\forall i, j \in \{1, ..., n\}$ ,  $i \neq j$ , we introduce in the following lemma ew dominance rule a new dominance rule.

**Lemma 1** *Let I* = (*J*, *p*<sub>1</sub>, *l*, *p*<sub>2</sub>) *be an instance of*  $F2|l_j|C_{max}$  *and two jobs i*, *j* ∈ *J such that p*<sub>*n*</sub> + *l*, ≤ *p*<sub>*n*</sub> + *l*, ≤ *p*<sub>*n*</sub> + *l*, *Eor any schedule S* of *I* if *i* and *that*  $p_{1j} + l_j \leq p_{1,i} + l_i \leq p_{2j} + l_j$ . For any schedule S of I, if j and i are adjacent on<br>*M*, then i should precede i on *M*. *M*1 *then j should precede i on M*1*.*

*Proof* Let us suppose that *j* is executed before *i* on  $M_1$ . First, thanks to the rela-<br>tion  $n + l \le n + l$  *i* is ready for processing on *M* while the processing of job tion  $p_{1,i} + l_i \leq p_{2,j} + l_j$ , *i* is ready for processing on  $M_2$  while the processing of job *i* has not vet ended. Then these two jobs are executed continuously without idle on *j* has not yet ended. Then these two jobs are executed continuously without idle on *M*<sub>2</sub>. Second, since  $p_{1,j} + l_j \leq p_{1,i} + l_i$ , the operations  $O_{2,j}$  and  $O_{2,i}$  would have started earlier than if *i* had preceded *i* on *M*. earlier than if *i* had preceded *j* on *M*1.

**Corollary 1** *Let*  $I = (J, p_1, l, p_2)$  *be an instance of*  $F2|l_j|C_{max}$  *and two jobs i,j*  $\in J$ .<br>*If*  $p_k + l \leq p_k + l \leq p_1 + l$  *then*  $\delta_k = \infty$  *Otherwise*  $\delta_k = w$ . If  $p_{1,j} + l_j \leq p_{1,i} + l_i \leq p_{2,j} + l_j$ , then  $\delta_{i,j} = \infty$ . Otherwise  $\delta_{i,j} = w_{i,j}$ .

Similarly, by interchanging the role of  $M_1$  and  $M_2$ , we obtain  $\Delta'$  another lower<br>and on the total additional waiting time. Therefore, the following valid inequality bound on the total additional waiting time. Therefore, the following valid inequality holds:  $\sum_{j=1}^{n} C_{2,j} \ge \sum_{j=1}^{n} C_{1,j} + \sum_{j=1}^{n} (p_{2,j} + l_j) + \max(\Delta, \Delta').$ 

#### **4 Computational Results**

We present in this section the computational results of the new lower bounds and we compare their performance. We test them on a set of six classes A–F that was proposed by Dell'Amico [\[1\]](#page-6-2). Furthermore, preliminary computational results conducted on the literature classes show that previous lower bounds give bad performance when time delays are very large compared to processing times. To that aim, we introduce two new classes of instances where the processing times on  $M_1$  and  $M_2$  and the time<br>delays are randomly generated between  $[1 \alpha]$ ,  $[1 \alpha]$  and  $[1 \alpha]$  respectively delays are randomly generated between  $[1 \dots \alpha]$ ,  $[1 \dots \beta]$  and  $[1 \dots \gamma]$ , respectively, where  $\alpha = \beta = 20$  and  $\gamma = \frac{n}{2}10$  (resp.  $\alpha = \beta = 100$  and  $\gamma = \frac{n}{2}100$ ) for class 1 (resp. class 2) For each class the number of jobs is  $n = 10, 30, 50, 100, 150, 200$ . For each 2 2 class 2). For each class, the number of jobs is *<sup>n</sup>* = 10*,* 30*,* 50*,* 100*,* 150*,* 200. For each combination of class and number of jobs, 10 instances were randomly generated. All algorithms were coded in C++ and compiled under CentOS 6.6. Moreover, we used CPLEX 12.6 to implement the linear programming-based lower bound. The experiments were conducted on an Intel(R) Xeon(R) @ 2.67 GHz processor. For pages limitation, we interest only to the most competitive lower bounds.

In the following, we denote by  $LB_3^N$  the optimal objective value that is obtained<br>er solving the LP relaxation of the mathematical model (1)–(7) including the valid after solving the LP relaxation of the mathematical model  $(1)-(7)$  $(1)-(7)$  $(1)-(7)$  including the valid inequality and by  $LB_4^N$  a version of  $LB_3^N$  without Constraints [\(3\)](#page-2-0). We conducted pre-<br>liminary computational results on  $LB_3^N$  and  $LB_4^N$ . Clearly,  $LB_3^N$  dominates  $LB_4^N$ . How-<br>ever *LB<sup>N</sup>* offers a good trade of ever,  $LB_4^N$  offers a good trade-off between effectiveness and efficiency. Indeed, for all<br>the considered instances where  $n < 100$ ,  $IR^N$  achieves the same lower bound values the considered instances where  $n < 100$ ,  $LR_{\text{4}}^N$  achieves the same lower bound values<br>as  $LR^N$  within a very short time. The average computational time of  $LR^N$  on these as  $LB_3^N$  within a very short time. The average computational time of  $LB_4^N$  on these<br>instances is 0.77 s while  $LB_4^N$  needs 61.54 s. Furthermore,  $LB_4^N$  fails to solve all large instances is 0.77 s while  $LB_3^N$  needs 61.54 s. Furthermore,  $LB_3^N$  fails to solve all large<br>scale instances (i.e.  $n > 100$ ) within 1800 s, while  $LR^N$  solves them in an average scale instances (i.e.  $n \ge 100$ ) within 1800 s, while  $LB_4^N$  solves them in an average time of 1.47 s. time of 1*.*47 s.

In order to present a detailed image of the performance of lower bounds  $LB_3$ ,  $LB_4$ ,  $LB_4$ ,  $LB_5$ ,  $LB_6$ ,  $LB_7$ ,  $LB_6$ ,  $CB_7$ ,  $CB_7$  $LB_5$ ,  $LB_6$ ,  $LB_2^N$  and  $LB_4^N$ , a pairwise comparison between them is given in Table [1.](#page-5-0) In this table we illustrate for each pair of lower bounds *LB* and *LB*, which are disthis table, we illustrate for each pair of lower bounds  $LB_{row}$  and  $LB_{col}$ , which are displayed in some given row and column, respectively, the percentage of times where  $LB_{col} > LB_{row}$ . We observe on classes A–F that  $LB_2^N$  outperforms  $LB_5$  in 10.83% of instances and  $LR_2$  in 26.38% of instances However, on the new classes 1–2, we instances and  $LB_6$  in 26.38% of instances. However, on the new classes 1–2, we<br>notice that  $IR^N$  provides a much better performance than the rest, since it outpernotice that  $LB_4^N$  provides a much better performance than the rest, since it outper-<br>forms  $IB$  and  $IB^N$  in 77.5% and 75% of instances, respectively. forms  $LB_5$  and  $LB_2^N$  in 77.5% and 75% of instances, respectively.<br>To get a better picture of the lower bounds performance we

To get a better picture of the lower bounds performance, we provide in Table [2](#page-5-1) the average percentage deviation (over the instances of each class) with respect to the maximal lower bound value, that is delivered by the considered lower bounds. Note that (-) means that the average CPU time is less than  $10^{-2}$  s. From Table [2,](#page-5-1) we

	Classes A–F							Classes $1-2$						
	LB <sub>3</sub>	LB <sub>4</sub>	LB <sub>5</sub>	LB <sub>6</sub>	$LB^N_{\alpha}$	$LB^N_A$	LB <sub>3</sub>	LB <sub>4</sub>	LB <sub>5</sub>	$LB_{6}$	$LB^N_{\alpha}$	$LB^N_4$		
LB <sub>3</sub>		63.33	99.44	98.05	99.44	100		46.66	60.83	52.5	61.66	100		
LB <sub>4</sub>	36.66		99.72	98.33	99.72	100	50		61.66	54.16	61.66	100		
LB <sub>5</sub>	0.55	0.27	$\overline{\phantom{0}}$	3.33	10.83	2.5	39.16	38.33		7.5	52.5	77.5		
LB <sub>6</sub>	1.94	1.66	23.33	$\overline{\phantom{0}}$	26.38	2.77	47.5	45.83	67.5		72.5	79.16		
$LB_2^N$	0.55	0.27	$\Omega$	$\overline{0}$	-	1.94	38.33	38.33	$\Omega$	$\Omega$	-	75		
$LB_4^N$	$\mathbf{0}$	$\Omega$	97.5	97.22	98.05	$\overline{\phantom{0}}$	$\theta$	$\mathbf{0}$	20.83	19.16	23.33	$\overline{\phantom{0}}$		

<span id="page-5-0"></span>**Table 1** Pairwise comparison between lower bounds

<span id="page-5-1"></span>**Table 2** Relaxation performance by class

Class	LB <sub>3</sub>		$LB_4$		LB <sub>5</sub>		LB <sub>6</sub>		$LB_2^N$		$LB^N_4$	
	Gap	Time	Gap	Time	Gap	Time	Gap	Time	Gap	Time	Gap	Time
$\mathbf{A}$	29.9		30.26	$\overline{\phantom{0}}$	0.04	-	0.3		$\mathbf{0}$	$\overline{\phantom{0}}$	19.81	0.82
B	27.8	$\overline{\phantom{m}}$	28.26	$\overline{\phantom{0}}$	0.06		0.33	-	0.01	$\qquad \qquad$	17.82	0.81
C	23.96		24.16	$\overline{\phantom{a}}$	0.89		2.56		0.69		13.73	0.79
D	32.24	$\overline{\phantom{m}}$	32.29	$\overline{\phantom{0}}$	0.04		0.03		$\mathbf{0}$		$21.46 \mid 0.84$	
E	53.96		46.26	$\overline{\phantom{0}}$	0.003		0.02		$\mathbf 0$	$\overline{\phantom{0}}$	33.09	0.82
F	$53.32$ –		45.85	$\overline{\phantom{0}}$	0.02		0.22		$\mathbf{0}$		32.93	0.85
$\mathbf{1}$	10.35	$\overline{\phantom{a}}$	11	$\overline{\phantom{m}}$	1.78		2.19		1.56		$0.92 \mid 0.58$	
2	10.42	$\equiv$	10.42	$\overline{\phantom{0}}$	13.89		19.26	$\overline{\phantom{0}}$	13.66		0.11	0.69
Avg	30.24		28.56	$\overline{\phantom{0}}$	2.09		3.11		1.99		17.48	0.77

observe that the average gaps significantly depend on the classes. On one hand, *LBN* exhibits an average gap of 1.99% on all classes. However, for the instances of class<br>2. its average gap jumps to 13.66%. On the other hand,  $IR^N$  presents a much better 2, its average gap jumps to 13.66%. On the other hand,  $LB_4^N$  presents a much better performance on the new classes. Indeed, the average gap of this bound is equal to performance on the new classes. Indeed, the average gap of this bound is equal to 0*.*92% and 0*.*11% on class 1 and class 2, respectively.

## **5 Conclusion**

This paper addressed the two-machine flow shop problem with time delays. We recalled the lower bounds of the literature and proposed new ones. In particular, the linear relaxation of a mathematical formulation with the consideration of a valid inequality and a dominance rule provides the best performance on a set of 120 new instances. Future research needs to be focused on investigating new valid inequalities and dominance rules in order to improve the resolution of the considered model.

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