Single-machine Scheduling with Periodic Maintenance to Minimize Makespan

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Abstract

We consider a single-machine scheduling problem with periodic maintenance activities. Although the scheduling problem with maintenance has attracted researchers' attention, most of past studies considered only one maintenance period. In this research several maintenance periods are considered where each maintenance activity is scheduled after a periodic time interval. The objective is to find a schedule that minimizes the makespan, subject to periodic maintenance and nonresumable jobs. We first prove that the worst-case ratio of the classical LPT algorithm is 2. Then we show that there is no polynomial time approximation algorithm with a worst-case ratio less than 2 unless $P = NP$, which implies that the LPT algorithm is the best possible.

1 Introduction

Most literature on scheduling theory assumes that machines are continuously available. However, this assumption may not be valid in a real production situation due to preventive maintenance (a deterministic event) or breakdown of machines (a stochastic phenomenon). It is not uncommon to observe in practice that machines are awaiting maintenance while there are jobs waiting to be processed by these machines. This is due to a lack of coordination between maintenance planning and production scheduling. Uncertain machine breakdowns will make the shop behavior hard to predict, thus reducing the efficiency of the production system. Maintenance can reduce the breakdown rate with minor sacrifices in production time. The importance of maintenance has increasingly been recognized by decision makers. Therefore, scheduling the maintenance of manufacturing systems has gradually become a common practice in many companies. The work of periodic maintenance includes periodic inspections, periodic repairs, and preventive maintenance. With proper planning of periodic maintenance, the shop can improve production efficiency and safety, resulting in increased productivity and heightened safety awareness [2].

As maintenance is scheduled periodically in many manufacturing systems, there is a need to develop an approach to handle the scheduling of jobs for processing in systems with periodic maintenance, which usually have more than one maintenance period. However, to the best of our knowledge, only Liao and Chen [12] have considered such a scheduling problem for the objective of minimizing the maximum tardiness. They proposed a branch-and-bound algorithm to derive the optimal schedule and a heuristic algorithm for large-sized problems. They also provided computational results to demonstrate the efficiency of their heuristic. In this paper we consider the scheduling problem with periodic maintenance to minimize the makespan.

Formally, the considered problem can be described as follows: We are given n independent non-resumable jobs $\mathcal{J} = \{J_1, J_2, \cdots, J_n\}$, which are processed on a single machine. Here *nonresumable* means that if a job cannot be finished before a maintenance activity, it has to restart. The processing time of job J_i is p_i. All the jobs are available at time zero. The amount of time to perform each maintenance activity is t. Let the length of the time interval between two consecutive maintenance periods be T. We assume that $T \geq p_i$ for every $i = 1, 2, \cdots, n$, for otherwise there is trivially no feasible schedule. We think of each interval between two consecutive maintenance activities as a batch with a capacity T. Thus, a schedule π can be denoted as $\pi = (B_1, M_1, B_2, M_2, \cdots, M_{L-1}, B_L)$, where M_i is the ith maintenance activity, L is the number of batches, and B_i is the ith batch of jobs. An illustration of the considered problem in the form of a Gantt chart is given in Figure 1. Let C_i be the completion time of job J_i, then the objective is to minimize the makespan, which is defined as $C_{\max} = \max_{i=1,2,\cdots,n} C_i$. Using the three-field notation of Graham et al. [5], we denote this scheduling problem as $1/nr - pm/C_{\max}$. It can easily be shown that this problem is strongly NP-hard [9], but no approximation algorithm has been provided and analyzed in the literature.

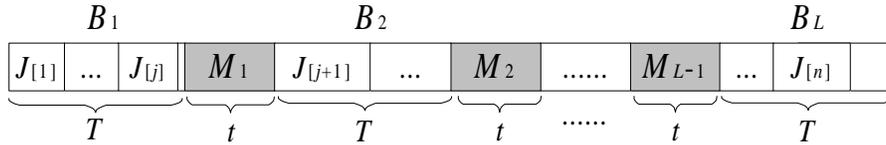We use the worst-case ratio to measure the quality of an approximation algorithm. Specifically,

Figure 1: An illustration of the problem under consideration, where $J_{[j]}$ denotes the job placed in the j-th position of the given schedule.

for the makespan problem, let C_A denote the makespan produced by an approximation algorithm A, and C_{OPT} the makespan produced by an optimal off-line algorithm. Then the *worst-case ratio* of algorithm A is defined as the smallest number c such that for any instance I, $C_A \leq cC_{OPT}$.

The single-machine scheduling problem with single maintenance and nonresumable jobs has been well studied. For the objective of minimizing the makespan, Lee [9] showed that the Longest Processing Time (LPT) rule has a tight worst-case ratio of 4/3, and He et al. [7] presented a fully polynomial time approximation scheme. For the objective of minimizing the total completion time, Lee and Liman [10] proved that the worst-case ratio of the Shortest Processing Time (SPT) rule is 9/7. Sadfi et al. [13] proposed a modified algorithm $MSPT$ with a worst-case ratio of 20/17. He et al. [8] presented a polynomial time approximation scheme. Moreover, Lee [9] presented heuristics for other objectives, such as minimizing the maximum lateness, the number of tardy jobs, and the total weighted completion time, etc. Graves and Lee [6] extended the problem to consider semiresumable jobs. For details on related research, the reader may refer to the survey paper [14].

In this paper we first show that the worst-case ratio of the classical LPT algorithm is 2. Then we prove that there is no polynomial time approximation algorithm with a worst-case ratio of less than 2, which implies that LPT is the best possible algorithm. Finally, we present some concluding remarks.

2 The LPT algorithm and its worst-case ratio

In this section we analyze the LPT algorithm, which is a classical heuristic for solving scheduling problems. It can be formally described as follows.

Algorithm LPT: Re-order all the jobs such that $p_1 \geq p_2 \geq \cdots \geq p_n$; then process the jobs consecutively as early as possible.

Note that if we take each batch as a bin, then the LPT algorithm is equivalent to the First Fit Decreasing (FFD) algorithm, which is a classical heuristic for solving the bin-packing problem. The worst-case ratio for the FFD is 3/2 [11], i.e.,

$$b \leq \frac{3}{2}b^*, \tag{1}$$

where b is the number of bins (i.e., batches) obtained by the FFD (i.e., LPT) algorithm and b^* is

the optimal number of bins (batches) for the bin-packing (scheduling) problem. Before analyzing the LPT algorithm, we first present some properties and lemmas, which are all straightforward.

Property 1 *The optimal schedule must have the minimum number of batches, i.e., it corresponds to an optimal solution for the bin-packing problem.*

Lemma 1 *(see [3], p. 574) In the LPT schedule, if $b > b^*$, then the processing time of each job in batches $B_{b^*+1}, B_{b^*+2}, \cdots, B_b$ is not larger than $T/3$.*

Lemma 2 *(see [3], p. 574-575) In the LPT schedule, if $b > b^*$, then the total number of jobs in batches $B_{b^*+1}, B_{b^*+2}, \cdots, B_b$ is not greater than $b^* - 1$.*

Let the total processing times of the jobs in the last batch of the optimal schedule and the LPT schedule be x and y, respectively. Then from Property 1, the makespan of the optimal schedule is

$$C_{OPT} = (b^* - 1)(T + t) + x, \tag{2}$$

while the makespan of the LPT schedule is

$$C_{LPT} = (b - 1)(T + t) + y. \tag{3}$$

(2) implies that

$$b^* = 1 + \frac{C_{OPT} - x}{T + t}. \tag{4}$$

Substituting (4) into (1), we obtain

$$b \leq \frac{3}{2}\left(1 + \frac{C_{OPT} - x}{T + t}\right). \tag{5}$$

Substituting (5) into (3), we establish

$$\begin{aligned} C_{LPT} &\leq \left[\frac{3}{2}\left(1 + \frac{C_{OPT} - x}{T + t}\right) - 1\right](T + t) + y \\ &= \frac{3}{2}C_{OPT} + \frac{1}{2}(T + t) - \frac{3}{2}x + y \\ &\leq \frac{3}{2}C_{OPT} + \frac{1}{2}(T + t) + y. \end{aligned} \tag{6}$$

On the other hand, it is clear that $y \leq T$. Combining it with (6), we obtain

$$C_{LPT} \leq \frac{3}{2}C_{OPT} + \frac{1}{2}(3T + t). \tag{7}$$

Now we are ready to obtain the worst-case ratio of the LPT algorithm.

Theorem 3 *For the problem $1/nr - pm/C_{max}$, the worst-case ratio of the LPT algorithm is 2.*

Proof. We first claim that $b^* > 1$. Otherwise, we have $C_{LPT} = C_{OPT}$, and we are done. If $b = b^*$, then from (2) and (3), we see that

$$C_{LPT} = C_{OPT} + y - x \le C_{OPT} + T < 2C_{OPT},$$

where the last inequality holds because $b^* > 1$, i.e., $T < C_{OPT}$. So, we assume in the following that $b > b^*$.

Case 1 $b^* \ge 4$. Thus, from (2), we have $C_{OPT} \ge 3(T + t) > 3T + t$. Combining it with (7), we obtain $C_{LPT} \le 2C_{OPT}$.

Case 2 $b^* = 3$. Then, from Lemma 2, we know that in the LPT schedule, the total number of jobs in batches $B_{b^*+1}, B_{b^*+2}, \cdots, B_b$ is not greater than 2. Combining it with Lemma 1, we conclude that $y \le \frac{2}{3}T$. As $b \le \frac{3}{2}b^* = \frac{9}{2}$ and $b > b^*$, we know that $b = 4$. By (3) and $y \le \frac{2}{3}T$, we have $C_{LPT} = 3(T + t) + y < 4(T + t)$. (2) states that $C_{OPT} = 2(T + t) + x > 2(T + t)$. Hence, we have $C_{LPT} < 2C_{OPT}$.

Case 3 $b^* = 2$. By the same reasoning as Case 2, we conclude that $b = 3$, the number of jobs in the 3th batch of the LPT schedule is 1, and $y \le \frac{T}{3}$. Thus, we have $C_{OPT} = (T + t) + x$ and $C_{LPT} = 2(T + t) + y$.

Denote $P = \sum_{i=1}^{n} p_i$. Let A be the total processing time of the jobs in the first batch of the optimal schedule, and B, C be the total processing times of the jobs in the first and the second batches of the LPT schedule, respectively. Then, we have $B + C + y = P = A + x$. Combining it with $A \le T$, we have

$$B + C + y \le T + x. \tag{8}$$

On the other hand, by the LPT rule, we have $B + y > T$ and $C + y > T$, and hence $B + C + y > 2T - y$. Combining it with (8), we obtain $x > T - y$, and hence $x > \frac{2}{3}T$ (since $y \le \frac{T}{3}$). Therefore, we obtain $C_{OPT} = (T + t) + x > \frac{5}{3}T + t$, and $C_{LPT} = 2(T + t) + y \le \frac{7}{3}T + 2t$, implying $C_{LPT} < 2C_{OPT}$.

Hence, we have completed the proof that the worst-case ratio of the LPT algorithm is not greater than 2. To show that the ratio cannot be smaller than 2, consider the following instance: Let $T = 12$, $p_1 = 6$, $p_2 = p_3 = p_4 = 4$, $p_5 = p_6 = 3$, and t be an arbitrary integer. Applying LPT, we obtain $B_1 = \{J_1, J_2\}$, $B_2 = \{J_3, J_4, J_5\}$, $B_3 = \{J_6\}$, and the makespan is $C_{LPT} = 2t + 27$. However, an optimal solution has two batches, where the first batch contains J_1, J_5, J_6 while the second batch contains J_2, J_3, J_4. Hence, $C_{OPT} = t + 24$. It follows that $\frac{C_{LPT}}{C_{OPT}} = \frac{2t+27}{t+24} \to 2$ as $t \to \infty$. The LPT schedule and an optimal schedule are illustrated in Figure 2. ∎
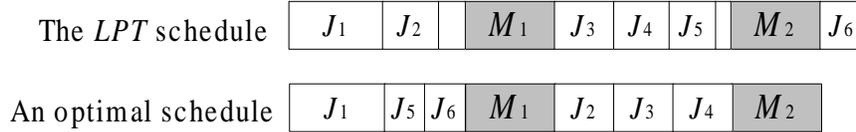
Figure 2: An instance showing the tightness of the worst-case ratio.

3   Non-approximability

It is well known that it is impossible to have a polynomial time approximation algorithm for the bin-packing problem with a worst-case ratio of less than $3/2$ unless $P = NP$ [4]. However, for our problem, the lower bound can be larger. In fact, we show that if there is an approximation algorithm with a worst-case ratio of $2 - \varepsilon$ for any $0 < \varepsilon < 1$, then it can be used to establish a polynomial time algorithm for solving the PARTITION problem, which is NP-hard [4], leading to a contradiction (if $P \neq NP$). Hence, such an algorithm for the considered problem cannot exist unless $P = NP$.

PARTITION: Given n positive integers $h_1, h_2, \cdots, h_n$ with $\sum_{i=1}^{n} h_i = 2H$, does there exist a set $U \subseteq \{1, 2, \cdots, n\}$ with $\sum_{i \in U} h_i = H$?

For any fixed positive number $\varepsilon < 1$ and an instance I of the PARTITION problem, we construct an instance II of our scheduling problem as follows: There are n jobs: $J_1, \cdots, J_n$, and the processing time of job J_i is $p_i = h_i$. Let the time of the maintenance period be $t = \lceil 2H(1 - \varepsilon)/\varepsilon \rceil$, and the time of the interval between two consecutive maintenance periods be $T = H$. It is clear that this construction can be performed in polynomial time. We first give some lemmas.

Lemma 4 *If there exists a solution to the instance I, then the optimal makespan for the instance II is $C_{OPT} = 2H + t$.*

Proof. Suppose that there exists such a subset U for the instance I such that $\sum_{i \in U} h_i = H$. We process jobs $\{J_i \mid i \in U\}$ in the first batch and all the remaining jobs in the second batch. Hence, the corresponding makespan equals $2H + t$, which achieves the trivial lower bound for the optimal makespan and is thus optimal. ∎

Lemma 5 *If there is no solution to the instance I, then the optimal solution value for the instance II satisfies the inequality $C_{OPT} \geq 2H + 2t + 1$.*

Proof. If there is no solution to the instance I, then it can easily be verified that the optimal schedule for instance II has to use at least three batches. Therefore, the optimal value $C_{OPT} \geq 2(T + t) + 1 = 2(H + t) + 1$. ∎

Lemma 6 *If there exists a polynomial time approximation algorithm A_ε with a worst-case ratio of $2 - \varepsilon$ for some positive number $\varepsilon < 1$, then there exists a polynomial time algorithm for the PARTITION problem.*

Proof. Given any instance I of the PARTITION problem, we construct the corresponding instance II of our scheduling problem in polynomial time as above. Define an upper threshold $Z = 2(H+t)$. Then the instance I of the PARTITION problem can be answered by merely comparing the values of C_{A_ε} and Z.

To see this is the case, let us apply A_ε to the instance II. If $C_{A_\varepsilon} \leq Z = 2(H+t)$, then $C_{OPT} \leq C_{A_\varepsilon} \leq 2(H+t)$. By Lemma 5, we deduce that there is a solution to the instance I of the PARTITION problem. On the other hand, if $C_{A_\varepsilon} > Z$, since $C_{OPT} \geq \frac{C_{A_\varepsilon}}{2-\epsilon}$ by the assumption that A_ε has a worst-case ratio of $2 - \varepsilon$, and since $t = \lceil 2H(1-\varepsilon)/\varepsilon \rceil$ implies $\epsilon \geq \frac{2H}{2H+t}$, we have

$$C_{OPT} > \frac{2(H+t)}{2 - \epsilon} \geq \frac{2(H+t)}{2 - \frac{2H}{2H+t}} = 2H + t.$$

Combining it with Lemma 4, we deduce that there is no solution to the instance I.

Hence, we have shown that the schedule for the instance II produced by A_ε gives us a right answer about whether there exists a solution to the instance I. Since the times for constructing instance II and A_ε are all polynomial, A_ε can solve I, an arbitrary instance of PARTITION problem in polynomial time. This completes the proof. ∎

By Lemma 6, and the fact that any NP-hard problem cannot be solved by a polynomial time algorithm unless $P = NP$, we establish the following main result.

Theorem 7 *Unless $P = NP$, the LPT algorithm is the best possible polynomial time approximation algorithm for the problem $1/nr - pm/C_{\max}$.*

4 Conclusions

We showed that the worst-case ratio of the classical LPT algorithm is 2 for the problem $1/nr - pm/C_{\max}$. We also showed that 2 is the best possible for all the polynomial time approximation algorithms unless $P = NP$. So in future research, it is worth considering the design of approximation algorithms with a lower time complexity than the LPT algorithm, while maintaining the worst-case ratio of 2. It is also worth considering the problem with respect to other objectives and in parallel-machine systems.

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