Discrete Optimization

# Scheduling to minimize the maximum total completion time per machine ${ }^{\pi}$ 

Long Wan ${ }^{\text {a }}$, Zhihao Ding ${ }^{\text {b }}$, Yunpeng Li ${ }^{\text {b }}$, Qianqian Chen ${ }^{\text {b }}$, Zhiyi Tan ${ }^{\text {c,* }}$<br>${ }^{\text {a }}$ School of Information Technology, Jiangxi University of Finance and Economics, Nanchang 330013, PR China<br>${ }^{\mathrm{b}}$ Department of Mathematics, Zhejiang University, Hangzhou 310027, PR China<br>${ }^{\text {c D D Department of Mathematics, State Key Lab of CAD \& CG, Zhejiang University, Hangzhou 310027, PR China }}$

## A R T I C L E I N F O

## Article history:

Received 17 July 2013
Accepted 28 September 2014
Available online 7 October 2014

## Keywords:

Scheduling
Parallel machine
Worst-case ratio


#### Abstract

In this paper, we study the problem of minimizing the maximum total completion time per machine on $m$ parallel and identical machines. We prove that the problem is strongly $N P$-hard if $m$ is a part of the input. When $m$ is a given number, a pseudo-polynomial time dynamic programming is proposed. We also show that the worst-case ratio of SPT is at most 2.608 and at least 2.5366 when $m$ is sufficiently large. We further present another algorithm which has a worst-case ratio of 2.


© 2014 Elsevier B.V. All rights reserved.

## 1. Introduction

Minimizing the total (weighted) completion time is one of the most commonly studied criteria in scheduling theory (Smith, 1956; Kawaguchi, \& Kyan, 1986). It is well known that the problem of minimizing the total completion time on parallel and identical machines can be solved optimally by the Shortest Processing Time first (SPT for short) algorithm (Conway, Maxwell, \& Miller, 1967). Here SPT first sorts all jobs in the order of nondecreasing processing times, then assigns the first unprocessed job in the sequence to the machine which can complete it as early as possible.

However, in the schedule produced by SPT, the total completion time of the jobs assigned to each machine may vary quite widely. To keep the total completion time per machine as equal as possible, it is convenient to consider the objective of minimizing the maximum total completion time per machine. Suppose a manager of a company wants to assign some late orders to several parallel subsidiaries. Since the orders are delayed, a penalty should be paid for each order, which is proportional to its completion time. The manager wants to be fair such that the total penalty that each subsidiary needs to be paid is as equal as possible. The situation can be viewed as a scheduling problem, in which orders are modeled as jobs and subsidiaries are modeled as machines. Taking the above objective into considera-

[^0]tion, the manager may be provided with constructive assistance and receive much less complaints from subsidiaries.

The problem can be stated formally as follows. We are given a sequence of $n$ independent jobs $J_{1}, J_{2}, \ldots, J_{n}$, which need to be non-preemptively processed on $m$ parallel and identical machines $M_{1}, M_{2}, \ldots, M_{m}$. The size (processing time) of $J_{j}$ is $p_{j}, j=1, \ldots, n$. Without loss of generality, we assume $p_{1} \leq p_{2} \leq \cdots \leq p_{n}$. Given a schedule $\sigma^{A}$, let $C_{j}^{A}$ be the completion time of $J_{j}$ in $\sigma^{A}, j=1, \ldots, n$. The total completion time of the jobs scheduled on $M_{i}$ is denoted $T C_{i}^{A}$, $i=1, \ldots, m$. Then $T C^{A}=\max _{i=1, \ldots, m} T C_{i}^{A}$ is the objective value of $\sigma^{A}$. By extending the three-field notation, the problem can be denoted as $\operatorname{Pm} \|\left(\sum C_{j}\right)_{\max }$, while the corresponding problem with the objective of minimizing the total completion time is denoted by $\operatorname{Pm} \| \sum C_{j}$ as usual. The problem of finding such a schedule among optimal schedules of $P m \| \sum C_{j}$ that the maximum total completion time per machine is minimized is denoted $\operatorname{Pm} \| \mid \operatorname{Lex}\left(\sum C_{j},\left(\sum C_{j}\right)_{\max }\right)$. We will use $\sigma^{*}$ and $\sigma^{* *}$ to denote the optimal schedules of $\operatorname{Pm\| } \|\left(\sum C_{j}\right)_{\max }$ and $\operatorname{Pm}\left|\mid \operatorname{Lex}\left(\sum C_{j},\left(\sum C_{j}\right)_{\max }\right)\right.$, respectively.

The problem $P m \|\left(\sum C_{j}\right)_{\max }$ was first proposed by Angel, Bampis, and Pascual (2008). They proved that to find schedule $\sigma^{*}$ or $\sigma^{* *}$ is NP-hard. They also proved that the worst-case ratio of SPT for $\operatorname{Pm} \|\left(\sum C_{j}\right)_{\max }$ lies in the interval $\left[2-\frac{2}{m^{2}+m}, 3-\frac{3}{m}+\frac{1}{m^{2}}\right]$. In this paper, we show that the problem is strongly $N P$-hard if $m$ is a part of the input. When $m$ is a given number, a pseudo-polynomial time dynamic programming algorithm is proposed, which will directly lead to a Fully Polynomial Time Approximation Scheme (FPTAS). We also show that the worst-case ratio of SPT is at most 2.608 and at least 2.5366 when $m$ is sufficiently large. The lower and upper bounds of the worst-case ratio of SPT when $m$ is a fixed number are also improved. We further present another algorithm Reverse SPT (RSPT for
short) which is also optimal for $P m \| \sum C_{j}$ and has a worst-case ratio of 2 for $P m \|\left(\sum C_{j}\right)_{\text {max }}$.

Comparing with the FPTAS, the advantage of SPT and RSPT is that both are optimal algorithms for $P m \| \sum C_{j}$. Thus corresponding results have some similarity with hierarchical optimization, an approach widely used for multicriteria scheduling (Hoogeveen, 2005; T'kindt, \& Billaut, 2006; Perez-Gonzalez, \& Framinan, 2014). Inspired by our work, Wan, Ma, and Yuan (2014) considered the problem of $\operatorname{Pm} \| \operatorname{Lex}\left(\sum C_{j},\left(\sum C_{j}\right)_{\max }\right)$, and proved that the worst-case ratios of SPT and RSPT are at most $\frac{11}{6}$ and $\frac{3}{2}$, respectively. The difference on the worst-case ratios of the same algorithm indicates that $\sigma^{*}$ and $\sigma^{* *}$ usually are not identical.

There are fruitful results in the field of multicriteria scheduling problems on parallel machines. Most researches are concerned with two objectives of minimizing the makespan and minimizing the total completion time (Leung, \& Young, 1989; Lin, Fowler, \& Pfund, 2013; Lee, Leung, Jia, Li, \& Pinedo, 2014), which are regarded to be egalitarian and utilitarian objectives, respectively (Myerson, 1981). In fact, the objective of minimizing the maximum total completion time per machine shares the above two characteristic features, which makes our results new and interesting. Another line of research about multicriteria scheduling is the so-called multi-agent scheduling (Baker \& Smith, 2003; Agnetis, Mirchandani, Pacciarelli, \& Pacifici, 2004; Leung, Pinedo, \& Wan, 2010; Mor, \& Mosheiov, 2011; Nong, Cheng, \& Ng, 2011). It arises in situations where multiple agents (customers), each having a different objective, compete for a common processing resource. However, for our problem, jobs belong to the same agent and contribute to the single objective.

The rest of the paper is organized as follows. In Section 2, we present complexity result and dynamic programming. The analyses of SPT and RSPT are given in Sections 3 and 4, respectively. Finally some conclusions are made in Section 5.

## 2. Complexity, dynamic programming and FPTAS

In Angel et al. (2008), it is proved that finding a schedule among all schedules or among optimal schedules of $\operatorname{Pm} \| \sum C_{j}$ such that the maximum total completion time per machine is minimized is $N P$ hard. However, if $m$ is a part of the input, to find a schedule minimizing the maximum total completion time per machine is strongly NP-hard. The following helpful lemma can be proved by a job interchange argument.

Lemma 2.1. There always exists an optimal schedule of $\operatorname{Pm} \|\left(\sum C_{j}\right)_{\max }$ in which all machines process the jobs in non-decreasing order of the size without idle time.

## Theorem 2.1. $P \|\left(\sum C_{j}\right)_{\text {max }}$ is strongly NP-hard.

Proof. The theorem will be proved by reduction from the numerical three-dimensional matching problem, which is known to be strongly NP-complete (Garey, and Johnson (1978)).

Numerical three-dimensional matching problem: Given three multisets of integers $U=\left\{u_{1}, \ldots, u_{m}\right\}, \quad V=\left\{v_{1}, \ldots, v_{m}\right\}, \quad W=$ $\left\{w_{1}, \ldots, w_{m}\right\}$ and an integer $B$ such that $\sum_{i=1}^{m}\left(u_{i}+v_{i}+w_{i}\right)=m B$, do there exist two permutations $\phi$ and $\varphi$ of $\{1, \ldots, m\}$ such that $u_{i}+v_{\phi(i)}+w_{\varphi(i)}=B$ for $i=1, \ldots, m$ ?

Let $I_{M}$ be an instance of numerical three-dimensional matching. Construct an instance with $m$ machines and $3 m$ jobs. Each job belongs to one of three types, $U$-job, $V$-job, or $W$-job. Corresponding to each element $u_{i}$ of $U$, there is a $U$-job $J_{U_{i}}$ with size $\frac{u_{i}}{3}$. Corresponding to each element $v_{i}$ of $V$, there is a $V$-job $J_{v_{i}}$ with size $\frac{v_{i}}{2}+B$. Corresponding to each element $w_{i}$ of $W$, there is a $W$-job $J w_{i}$ with size $w_{i}+2 B$. We will show that $I_{M}$ has "yes" answer if and only if there is a feasible schedule such that the maximum total completion time per machine is no more than $5 B$.

If $I_{M}$ has "yes" answer, then there exist two permutations $\phi$ and $\varphi$ of $\{1, \ldots, m\}$ such that $u_{i}+v_{\phi(i)}+w_{\varphi(i)}=B$ for $i=1, \ldots, m$. Let $J_{U_{i}}, J_{V_{\phi(i)}}, J_{W_{\varphi(i)}}$ be the three jobs successively scheduled on $M_{i}, i=$ $1, \ldots, m$. The total completion time of $M_{i}, i=1, \ldots, m$, is
$3 \frac{u_{i}}{3}+2\left(\frac{v_{\phi(i)}}{2}+B\right)+\left(w_{\varphi(i)}+2 B\right)=5 B$.
On the other hand, suppose there exist feasible schedules such that the maximum total completion time per machine is no more than $5 B$. By Lemma 2.1, we can focus on the one, denoted by $\sigma^{Y}$, that jobs are sequenced in non-decreasing order of the sizes on each machine. Clearly, $T C_{i}^{Y} \leq 5 B, i=1, \ldots, m$. We show the structure of $\sigma^{Y}$ step by step. Firstly, there is exactly one $W$-job on each machine. Otherwise, there exists machine $M_{i}$ which processes at least two $W$-jobs. Thus $T C_{i}^{Y} \geq 2 \cdot 2 B+2 B=6 B>5 B$, a contradiction. Let the $W$-job scheduled on $M_{i}$ be $J w_{\varphi(i)}, i=1, \ldots, m$, thus $\varphi$ is a permutation of $\{1, \ldots, m\}$. Secondly, there is at most one $V$-job on each machine. Otherwise, there exists machine $M_{i}$ which processes at least two $V$-jobs. Thus $T C_{i}^{Y} \geq 3 \cdot B+2 \cdot B+2 B=7 B>5 B$, also a contradiction. Let the $V$-job scheduled on $M_{i}$ be $J_{V_{\phi(i)}}, i=1, \ldots, m$, thus $\phi$ is also a permutation of $\{1, \ldots, m\}$. Finally, let $\mathcal{U}^{i}$ be the set of $U$-jobs which are scheduled on $M_{i}, i=1, \ldots, m$. We claim that $\left|\mathcal{U}^{i}\right|=1$ for any $i$. In fact, if $\mathcal{U}^{i}=\left\{J_{U_{i}}\right\}$, then

$$
\begin{aligned}
T C_{i}^{Y} & =3 \frac{u_{i}}{3}+2\left(\frac{v_{\phi(i)}}{2}+B\right)+\left(w_{\varphi(i)}+2 B\right) \\
& =u_{i}+v_{\phi(i)}+w_{\varphi(i)}+4 B
\end{aligned}
$$

If $\left|\mathcal{U}^{i}\right| \geq 2$, then

$$
\begin{aligned}
T C_{i}^{Y} & >3 \sum_{J_{U_{i}} \in \mathcal{U}^{i}} \frac{u_{i}}{3}+2\left(\frac{v_{\phi(i)}}{2}+B\right)+\left(w_{\varphi(i)}+2 B\right) \\
& =\sum_{J_{U_{i}} \in \mathcal{U}^{i}} u_{i}+v_{\phi(i)}+w_{\varphi(i)}+4 B
\end{aligned}
$$

Hence, if there exists $i$, such that $\left|\mathcal{U}^{i}\right| \geq 2$, then

$$
\begin{aligned}
\sum_{i=1}^{m} T C_{i}^{Y} & >\sum_{i=1}^{m}\left(\sum_{J_{u_{i}} \in \mathcal{U}^{i}} u_{i}+v_{\phi(i)}+w_{\varphi(i)}+4 B\right) \\
& =4 m B+\sum_{i=1}^{m}\left(u_{i}+v_{i}+w_{i}\right)=5 m B \geq m T C^{Y}
\end{aligned}
$$

which is a contradiction. Therefore, $\left|\mathcal{U}^{i}\right|=1$ for any $i$ and we assume that $\mathcal{U}^{i}=\left\{J_{U_{i}}\right\}$. Moreover, since $\sum_{i=1}^{m} T C_{i}^{Y}=5 m B \geq m T C^{Y}=$ $m \max _{i=1, \ldots, m} T C_{i}^{Y}, T C_{i}^{Y}=5 B$ for any $i$. Thus, we find two permutations $\phi$ and $\varphi$ of $\{1, \ldots, m\}$ such that $u_{i}+v_{\phi(i)}+w_{\varphi(i)}=B$ for any $i=1, \ldots, m$, which implies that $I_{M}$ has "yes" answer. $\square$

The result of Theorem 2.1 excludes the possibility that the $P \|\left(\sum C_{j}\right)_{\max }$ admits a FPTAS. However, if $m$ is a fixed number, we can develop a pseudo-polynomial time dynamic programming algorithm $D P$. In other words, the problem is $N P$-hard in the ordinary sense.

Denote by $\left[c_{1}, \ldots, c_{m}, t_{1}, \ldots, t_{m}\right]$ the state vector of a partial schedule without idle time, where $c_{i}$ and $t_{i}$ are the current completion time and current total completion time of $M_{i}$, respectively, $i=1, \ldots, m$. Let the initial state space be $F_{0}=\{[0, \ldots, 0,0, \ldots, 0]\}$. Let $F_{j}$ be the set of state vectors for schedules of the first $j$ jobs, $j=1,2, \ldots, n$. Then the state space $F_{j}$ can be recursively generated from $F_{j-1}$ as follows.

For any schedule of the first $j-1$ jobs, which corresponds to a state vector $\left[c_{1}, \ldots, c_{m}, t_{1}, \ldots, t_{m}\right]$ of $F_{j-1}, J_{j}$ can be assigned to any one of the $m$ machines. By Lemma 2.1, we can assume that $J_{j}$ is the current last job on the machine where it is assigned. Specifically, if

# https://daneshyari.com/en/article/476578 

Download Persian Version:

## https://daneshyari.com/article/476578

## Daneshyari.com


[^0]:    * This work was supported by the National Natural Science Foundation of China $(10971191,11271324)$ and Zhejiang Provincial Natural Science Foundation of China (LR12A01001).
    * Corresponding author. Tel.: +86 057187953867.

    E-mail address: tanzy@zju.edu.cn (Z. Tan).

