CHAPTER-VI

A NOTE ON FLOW SHOP SCHEDULING PROBLEM WITH INCREASING AND DECREASING LINEAR DETERIORATION ON WEIGHTED DOMINANT MACHINES

6.1 Introduction
6.2 Flow Shop Scheduling
6.3 Formulations of the Problem
6.4 Minimize the Total Weighted Completion Time
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The main results of this chapter have been published as detailed below:

A Note on Flow Shop Scheduling Problem with Increasing and Decreasing Linear Deterioration on Weighted Dominant Machine.  
6.1 INTRODUCTION

In this chapter, we study the flow shop scheduling problem with increasing and decreasing linear deterioration on weighted dominant machines and also deal with some special case of general, no-wait permutation flow shop scheduling problem, respectively. Special cases mean that the machines form an increasing series of dominant machines, and decreasing series of dominant machines. The objectives are to minimize one of the two regular performance criteria, namely, makespan, total completion time and weighted completion time. This objective is considered under following dominant machine constraint: \( idm \) and \( ddm \) are considered. Numerical examples of the solution approaches are provided.

Flow shops are frequently found in industry and are characterized by a set of jobs \( J = \{1, 2, \ldots, n\} \) and a set of machines, \( M = \{1, 2, \ldots, m\} \). The set of jobs is processed sequentially on \( m \) machines. In the traditional flow shop problem use assume deterministic processing times, denoted by \( P_{ij} \) where indices \( i \in M \) and \( j \in J \) represent a machine and a job, respectively. Furthermore all jobs are ready for processing at time zero and no other arrive later; a job may not be preempted by another job; jobs are not allowed to
pass others; no job may be processed by more than one machine; machines may process no more than one job at a time; and there are no down times due to machine breakdown or maintenance. In a flow shop problem, we usually determine the sequence of jobs to satisfy certain performance criteria including the minimization of makespan $c_{\text{max}}$, the sum of the job flow times, the mean tardiness or lateness, the maximum tardiness and the number of tardy or late jobs.

In the classical scheduling theory, a job processing time does not independent on job position in a sequence, however, in many realistic scheduling settings, the production facility improves continuously with time. As a result the processing time of a given job is shorter if it is scheduled later in the production sequence. In literature, this phenomenon is known as learning effect. Biskup [1] was the first one who investigated the effect of learning process as a function dependent on a number of repetitions during a production of similar items, in other words, processing times depend on a job position in the sequence, i.e. $P_{jr} = P_j r^\alpha$, where $P_j$ is the normal processing times of job $J_j$, $r$ is the position of job $J_j$ in the sequence and $\alpha \leq 0$ is the learning index of job $J_j$. He studied the single machine problem of minimizing the total flow time, the weighted sum
of completion time deviations from a common due date and the sum of job completion times. Similar works can be found in Mosheior [2], Mosheior and Sidney [3], Bachman and Janiak [4]. etc.

Wang and Xia [5] consider no-wait of no-idle flow shop scheduling problems with processing times dependent on starting time. In these problems job processing time is a simple linear function of the jobs starting time and some dominating relationships between machines can be satisfied. They showed that for the problems to minimum Makeskpan of minimize weighted sum of completion time polynomial algorithms still exist. When the objectives are to minimize maximum lateness, the solutions of a classical version may not hold. Ng et al. [6] also consider three scheduling problems with a decreasing linear modal of the job processing times, where the objective function is to minimize the total completion time, and two of the problems are solved optimally. Bachman et al. [7] consider the single – machine scheduling problem with start time dependent job processing times. They prove that the problem of minimizing the total weighted completion time is NP- hard. They also consider some special cases. Zhao et al. [8] consider a special type of the actual processing time, which is $P_i(t) = a_i(a + t)$, where $a$ and $b$ are
positive constant. They prove that the single-machine scheduling problems of minimizing makespan, sum of weighted completion times, maximum lateness and maximum cost are polynomials solvable, and the two-machine flow shop scheduling to minimize the makespan can be solved by Johnson’s rule.

We introduced such an interesting scheduling model in which the processing time of a job is a polynomial function of its starting time. This model reflects some real-life situations in which expected processing time of a job increases / decreases linearly or piecewise linearly on its starting time. Examples can be found in financial management, steel production resource allocations and national defiance, where any delay in tackling a task may result in an increasing / decreasing time, cost etc. to accomplish the task.

We consider the general, no-wait flow shop scheduling problem with increasing and decreasing linear deterioration a weighted dominant machines respectively. Deterioration of a job means that its processing time is a function of its execution start time. The “no-wait” constraint means that each job, once started, has to be processed without interruption until all of its operations are completed. In practice, this requirement may arise out of certain job
characteristics or out of the unavailability of intermediate storage machines.

In this chapter, we study flow shop scheduling problems with a learning effect on no-wait dominant machines. That is, the job processing time is a function of its position \( r \) in the sequence and no machine is allowed to have no-wait time between processing any two operations. Since processing time \( p_m \) is zero (as operation \( O_{m5} \) is not to be performed). The objective is to minimize maximum completion time. The previous works on flow shop scheduling in an environment of a series of dominating machines can be found in Nouweland et al. [9], Ho and Gupta [10] and Xiang et al. [11].

The remaining part of the chapter is organized as follows. In the next section 6.2, we give a general introduction to flow shop problem with a learning effect and dominant machines. In section 6.3, we explain the formulation of the problem. In section 6.4, we consider the minimizing the weighted sum of completion time. Final section includes conclusions and remarks about future research. To the author’s knowledge no literature on a note on flow shop scheduling problem with increasing and decreasing linear deterioration on weighted dominant machines has been published.
Some common examples of the problem in which the job processing time is an increasing start time-dependent function can be found in the areas of scheduling maintenance, cleaning assignments or metallurgy, in which any delay often implies additional effort (or time) to accomplish the job. On the other hand, an example considering the so called “learning effect” can be described by a non-increasing start time-dependent function. Assume that a worker has to assemble a large number of similar products. The time required by the worker to assemble one product depends on his knowledge, skills, organization of his working place and others. The worker learns how to produce. After some time, he is better skilled, his working place is better organized and his knowledge has increased. As a result of his learning, the time required to assemble one product decreases. In this case, a radar station has detected some objects approaching it. The time required to recognize the objects decreases as the objects get closer. Thus, the later the objects are detected, the less time needed for their recognition.

6.2 FLOW SHOP SCHEDULING

The flow-shop problem has been concentrated on by many researchers with diverse classical assumptions and different objective
functions and by implementing various optimization techniques. The regular flow shop problem consists of two main elements: (1) a group of \( M \) machines and (2) a set of \( N \) jobs to be processed on this group of machine. Each of the \( N \) jobs has the same ordering of machines for its process sequence. Each job can be processed on one and only one machine at a time (which means no job splitting), and each machine can process only one job at a time. Each job is processed only once on each machine. Operations are not preemtatable and set-up times of operations are independent of the sequences and therefore can be included in the processing time. The scheduling problem is to specify the order and timing of the processing of the jobs on machines, with an objective or objectives respecting above-mentioned assumptions. The flow shop problem with makespan criterion can be shown by \( n/m/F/c_{\text{max}} \) or equivalently \( F/lc_{\text{max}} \), where both show an \(( n \text{-job, } m\text{-machine})\) flow-shop problem with makespan criterion that can be defined as completion time at which all jobs complete processing or equivalently as maximum completion time of jobs.

6.3 FORMULATION OF THE PROBLEM

The sequence flow shop scheduling problem considered in this paper may be state as follows; we are given a set of \( n \) jobs \( J_1 \),
J_2, \ldots, J_n \) that have to be processed on the machines \( M_1, M_2, M_3, \ldots, M_m \) successively. The normal processing time of job \( J_i \) on machine (operation \( Q_{ij} \)) is \( P_{ij} \), the actual processing time of job \( J_i \) on the machine \( I \) is \( P_{ijr} \) if operation \( Q_{ij} \) is the \( r^{th} \) operation on machine \( i \). We are asked to find the order in which these \( n \) jobs should be processed on the \( m \) machines such that a given objective is to find the schedule that minimizes the makespan. In this chapter, we consider the jobs processing times characterized by position-dependent function:

\[
P_{ij} = P_{ijr} \alpha \quad i=1,2,\ldots,m; \ r, \ j = 1,2\ldots n. \quad (6.3.1)
\]

where \( \alpha \leq 0 \) denotes a learning index.

for a given schedule \( \sigma \), let \( C_{ij} = C_{ij}(\sigma) \) represent the completion time of operation \( O_{ij} \), \( C_j = C_{mj} \) represents the completion time of job \( J_i \), \( \sigma = ([1], [2], \ldots, [n]) \) denote a schedule, where \([j]\) denotes the job that occupies the \( j \)th position in \( \sigma \). The problems considered in the paper are denoted according to the three field notation \( \alpha/\beta/\gamma \) introduced by Graham et al. [12]

**Definition 1.** \( M_i \) is dominated by \( M_k \), iff \( \max\{P_{ij}/j = 1,2,\ldots,n\} \leq \min\{P_{kj}/j = 1,2,\ldots,n\} \). In abbreviated notation, it is denoted as \( M_i < M_k \) based on the above concept of dominant machines the five definitions considered in this chapter are as follows.
**Definition 2.** The machines form an increasing series of dominating machines \((idm)\). That is
\[
M_1 < M_2 < ... < M_m.
\]

**Definition 3.** The machines form a decreasing series of dominating machines \((ddm)\), that is
\[
M_1 > M_2 > ... > M_m.
\]

### 6.4 MINIMIZE THE TOTAL WEIGHTED COMPLETION TIME

The following results of lemmas can be easily obtained, and the results can be used in latter:

**Property 1:** *(Mosheior [2])* For the problem \(1/P_{jr} = P_j r^a/\)
\(C_{max}\) and optimal schedule can be obtained by SPT *(shortest processing time first)* rule.

**Property 2:** For the problem \(F_m//P_{ijr} = P_j r^a, \text{no-wait, } idm/\)
\(C_{max}\) and a given schedule \(\sigma = ([1],[2],...,[n])\), the completion time \(C_{(i)}\) of job \(J_{(i)}\) is as follows
\[
C_{(i)} = \sum_{i=1}^{m} P_{i1} + \sum_{k=2}^{i} P_{m[k]}k^a 
\]  \( (6.4.1)\)

Now, we demonstrate the results of property 2 in the following example.
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**Example 1.**  
$N = 4, M = 4, P_{11} = 12, P_{12} = 15, P_{13} = 17, P_{14} = 18, P_{21} = 19,$  
$P_{22} = 20, P_{23} = 23, P_{24} = 24, P_{31} = 26, P_{32} = 28, P_{33} = 29, P_{34} = 31, P_{41} = 31, P_{42} = 33,$  
$P_{43} = 34, P_{44} = 36, & W_1 = 4, W_2 = 6, W_3 = 8, W_4 = 9.$

Learning curve that is $a = -0.2$. Obviously the condition of example 1 conforms to the case of $idm$, that is $M_1 < M_2$.

\[ \begin{array}{c|cccc} M_1 & J_1 & J_2 & J_3 & J_4 \\ \hline M_2 & J_1 & J_2 & J_3 & J_4 \\ M_3 & J_1 & J_2 & J_3 & J_4 \\ M_4 & J_1 & J_2 & J_3 & J_4 \end{array} \]

**Fig. 1.** A sample of $F_{nt}/P_{ijr} = P_{ijr}a$, no - wait, $idm/C_{\text{max}}$, n=4, m=4

<table>
<thead>
<tr>
<th>Machin es</th>
<th>weighted</th>
<th>Processing time of the increasing order</th>
<th>Actual completion time</th>
<th>Weighted completion time</th>
</tr>
</thead>
<tbody>
<tr>
<td>Job</td>
<td></td>
<td>Job</td>
<td>1 2 3 4</td>
<td></td>
</tr>
<tr>
<td>1</td>
<td>4</td>
<td>12 15 17 18</td>
<td>88</td>
<td>352</td>
</tr>
<tr>
<td>2</td>
<td>6</td>
<td>19 20 23 24</td>
<td>116.72</td>
<td>700.32</td>
</tr>
<tr>
<td>3</td>
<td>8</td>
<td>26 28 29 31</td>
<td>144.01</td>
<td>1152.08</td>
</tr>
<tr>
<td>4</td>
<td>9</td>
<td>31 33 34 36</td>
<td>171.29</td>
<td>1541.61</td>
</tr>
</tbody>
</table>

**Property 3:** For the problem $F_{nt}/P_{ijr} = P_{ijr}a$, no - wait, $ddm / C_{\text{max}}$ and a given schedule $\sigma = ([1], [2], .........., [n])$, the completion time $C_{(j)}$ of job $J_{(j)}$ is as follows.

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\[ C_{[j]} = \sum_{k=1}^{n-1} P_{1[k]} k^a + \sum_{k=1}^{m} P_{k[n]} n^a + \sum_{k=j+1}^{n} P_{m[k]} k^a \quad \ldots \ldots (6.4.2) \]

Example 2. \( N = 4, M = 4, P_{11} = 28, P_{12} = 26, P_{13} = 24, P_{14} = 23, P_{21} = 22, \)
\( P_{22} = 21, P_{23} = 19, P_{24} = 18, P_{31} = 18, P_{32} = 17, P_{33} = 15, P_{34} = 12, P_{41} = 10, \)
\( P_{42} = 09, P_{43} = 07, P_{44} = 06, & W_1 = 8, W_2 = 7, W_3 = 7, W_4 = 6. \)

Learning curve that is \( a = -0.2. \) Obviously the condition of example 2 conforms to the case of \( ddm, \) that is \( M_1 > M_2. \)

![Diagram](image1)

**Fig. 2.** A sample of \( F_m | P_{ijr} = P_{ijr}^a, \) no-wait, \( ddm/C_{\text{max}} \)

**Table: 2**

<table>
<thead>
<tr>
<th>Machines</th>
<th>weighted</th>
<th>Processing time of the decreasing order Job</th>
<th>Actual completion time</th>
<th>Weighted completion time</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>8</td>
<td>28 26 24 23</td>
<td>132.61</td>
<td>1060.88</td>
</tr>
<tr>
<td>2</td>
<td>7</td>
<td>22 21 19 18</td>
<td>124.77</td>
<td>873.39</td>
</tr>
<tr>
<td>3</td>
<td>7</td>
<td>18 17 15 12</td>
<td>119.15</td>
<td>834.05</td>
</tr>
<tr>
<td>4</td>
<td>6</td>
<td>10 09 07 06</td>
<td>114.61</td>
<td>687.66</td>
</tr>
</tbody>
</table>

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Theorem-4. For the problem $F_m \mid P_{ijr} = P_{ij} r^a$, no-wait, idm $\mid \sum W_j C_j$ if the first processed job $J_1$ ascertained, then the schedule $\sigma = \{ J_t, \sigma_1 \}$ is an optimal one, $\sigma_1$ is a partial sequence obtained by sequencing the remaining (n-1) jobs is non-decreasing order $\{ P_{mj} \}$.

Proof: Consider the sequence $\sigma = \{ J_1, J_2, \ldots, J_n \}$.

By property 2, we have

$$\sum W_j C_j = \sum_{j=1}^{n} W_j \left[ \sum_{i=1}^{m} P_{i[1]} + \sum_{k=2}^{n} P_{m[k]} K^a \right]$$

where the term $\sum_{j=1}^{n} W_j$ is a constant. If the job processed first as curtailed then combining $\sum_{k=2}^{n} P_{m[k]} K^a$ can be minimized by sequencing the remaining (n-1) jobs in non-decreasing order of their normal processing times on the last machine by Lemma 1. An optimal schedule for the problem $F_m \mid P_{ijr} = P_{ij} r^a$, no-wait, idm $\mid \sum W_j C_j$ is obtained.

Therefore, an optimal schedule of the problem $F_m \mid P_{ijr} = P_{ij} r^a$, no-wait, idm $\mid \sum W_j C_j$ can be constructed as follows.

Select $J_1, J_2, J_3, \ldots, J_n$ as the first processed job in turn, then the remaining (n-1) job are sequenced in non-decreasing order of $\{ P_{mj} \}$ on the last machine, respectively, thus n schedules are generated. The one with the minimum weighted sum of completion time among these ‘n’ schedules is an optimal schedule.
Theorem 5. For the problem $F_m \mid P_{ij} = P_{ij}r^a$, no-wait, ddm $\sum W_j C_j$, if the last processed job $J_s$ ascertained, then the schedule $\sigma = \{\sigma_1, J_s\}$ is an optima one, where $\sigma_1$ is a partial sequence obtained by sequencing the remaining (n-1) job in non-decreasing order of $\{P_{ij}\}$.

Proof. Consider the sequence $\sigma = (J_{[1]}, J_{[2]}, \ldots, J_{[n]})$. By property 3, we have

$$\sum W_j C_j = \sum_{k=1}^{n-1} P_{1[k]} k^a + \sum_{k=1}^{m} P_{k[n]} n^a + \sum_{k=j+1}^{n} P_{m[k]} k^a$$

where the term $\sum_{j=1}^{n} W_j$ is a constant.

If the last processed job ascertained, then combining $\sum_{k=1}^{n-1} P_{1[k]} K^a$ can be minimized by sequencing the remaining (n-1) jobs in non-increasing order of their normal processing times on the first machine by lemma 1. An optimal schedule for the problem $F_m \mid P_{ij} = P_{ij}r^a$, no-wait, ddm $\sum W_j C_j$ is obtained.

Therefore, an optimal schedule of the problem $F_m \mid P_{ij} = P_{ij}r^a$, no-wait, ddm $\sum W_j C_j$ can be constructed as follows.

Select $J_1$, $J_2$, $J_3$, ..., $J_n$ as the first processed job in turn, then the remaining (n-1) job are sequenced in non-decreasing order of $\{P_{ij}\}$ on the last machine, respectively, thus ‘n’ schedules are generated. The
one with the minimum weighted sum of completion time among these ‘n’ schedules is an optimal schedule.

6.5 CONCLUSION

This chapter considers some permutation flow shop scheduling problem with increasing and decreasing linear deterioration on weighted dominant machines. It was show that some special cases of minimizing the discounted total weighted completion time can be solved in polynomial time. The objective is to minimize maximum completion time. For the objective, the following dominant machines constraint: \( idm \) and \( ddm \) are considered. Scheduling problems with such a learning effect in some other machine settings are also interesting and significant for future research.
6.6 REFERENCES


