INTEGRER RELAXATION ON BINARY QUADRATIC PROGRAMMING
FOR BATCHING AND SEQUENCING IN SINGLE FORMULATION
TO MINIMIZE TOTAL ACTUAL FLOW TIMES CRITERIA

Zahedi

Mathematics & Statistics Department, School of Computer Science, Binus University
Jl. K.H. Syahdan No. 9, Palmerah, Jakarta Barat 11480
zahedizahedi@binus.ac.id

ABSTRACT

This paper examines batch scheduling problem that has batching and sequencing in single formulation
for multiple-item case. The first step is to develop a model for single item single resource discussed in Zahedi
(2008). The model determined batch sizes and their schedule simultaneously in single item case. This paper
develops the model and algorithm for multiple-item case. The model functions to minimize total actual flow
times. An algorithm for the model is developed using a relaxation of the binary constraints. The binary values
for the decision variables are obtained from steps provided in algorithm. A numerical experience showing
characteristics of this problem is presented.

Keywords: actual flow time, batching, sequencing, simultaneous formulation, optimal solution

ABSTRAK

Makalah ini membahas masalah penjadwalan batch dengan batching dan sequencing dalam formulasi
tunggal untuk beberapa item kasus. Langkah pertama adalah mengembangkan model sumber daya tunggal satu
item yang dibahas dalam Zahedi (2008). Model ini ditentukan ukuran batch dan jadwal mereka secara
bersamaan dalam kasus item tunggal. Makalah ini mengembangkan model dan algoritma untuk multiple-item
kasus. Model ini berfungsi meminimalkan total waktu aliran yang sebenarnya. Sebuah algoritma untuk model ini
dikembangkan menggunakan relaksasi kendala biner. Nilai biner untuk variabel keputusan diperoleh dari
langkah yang diberikan dalam algoritma. Sebuah pengalaman numerik disajikan yang menunjukkan
karacteristik dari masalah ini.

Kata kunci: waktu aliran sebenarnya, batching, sequencing, formulasi simultan, solusi optimal
INTRODUCTION

This article develops backward scheduling problem and minimizes total actual flow times criteria from Halim et al (1998). He proposed a shop time criterion called actual flow time defined as the time that the part spends in the shop from its due date. The parts to be processed arrive at the production line at the right time in the right quantities, and that completed parts are delivered at their due dates. This means that the actual flow time can be implemented for production systems under Just-in-time (JIT) environment. In solving the problem Halim et al (1998), Lagrangian relaxation is used, in the case determining number of parts in batches and scheduling of batches in two ways. Zahedi (2004, 2007, and 2008) did research for backward scheduling on job and block level in the same criterion, i.e., minimize total actual flow times. Zahedi (2008) developed model and algorithm to solve binary quadratic programming model from batch scheduling that determines number of parts in batches and scheduling in single formulation, in single item single resource case. This article examines batch scheduling problem that has batching and sequencing in single formulation for multiple-item case.

Problem Formulation

Let $K$ different items be processed in a single stage operation. Each item requires one machine or resource to complete the operation. The items, with $n_k$ standing for the numbers of parts of item $k$ ($k = 1, 2, ..., K$) respectively, are demanded at common due date $d$. The processing time and setup time for an item may be different from one to another item. For each item, the processing time, the setup time, and the quantity demanded are assumed to be known and deterministic. The parts to process are not required to be available for processing at time zero but may arrive at the production line at the right times in the right quantities. The completed parts are delivered at their due date.

There are two issues considered here, i.e., creating the model formulation and creating an algorithm to solve the formulation.

Let’s define the following constants and variables:

- $d$ = the common due date
- $i, j$ = the indices for batch numbers. $i, j = 1, 2, ..., N_k$
- $k, m$ = the indices for items. $k, m = 1, 2, ..., K$
- $N_k$ = the number of batches for item $k$
- $n$ = the total number of parts
- $n_k$ = the number of parts of item $k$ demanded
- $t_k$ = the processing time for item $k$
- $s_k$ = the setup time for item $k$
- $B_{ki}$ = the starting time of processing batch $i$ of item $k$
- $Q_{ki}$ = the number of parts of item $k$ in batch $i$
- $X_{ki}$ = 1, if batch $i$ of item $k$ processed on resource is not vacuous, 0, otherwise
- $Y_{ki}$ = 1, if batch $i$ of item $k$ precedes batch $j$ of item $m$, 0, otherwise
- $F^a$ = total actual flow times all of parts
- $F_i^a$ = total actual flow times on $i^{th}$ step in algorithm.

The formulation of multiple items single resource (MISR) to minimize the total actual flow times can be expressed as follows.

Problem MISR

Minimize $F^a = \sum_{k=1}^{K} \sum_{i=1}^{N_k} \left( \sum_{m=1}^{K} \sum_{j=1}^{Q_{ki}} (s_m X_{mj} + t_m Q_{mj}) - s_k \right) Q_{ki}$
subject to
\[ \sum_{j=1}^{k} \sum_{i=1}^{m} Q_{j}(i) = n \]  
(2)
\[ \sum_{j=1}^{k} \sum_{i=1}^{m} Q_{j}(i) = n_k, \forall k \]  
(3)
\[ Q_{j}(i) \leq X_{j} n_k, \forall i,k \]  
(4)
\[ B_{j}(i) + \sum_{m=1}^{k} \sum_{j=1}^{n} (s_{m} x_{j} + t_{m} Q_{j}(i)) = d, \forall k,i \]  
(5)
\[ B_{j}(i) + t_{k} Q_{j}(i) \leq B_{m}(i) + m (Y_{k}^{m}/), \forall i,j,k,m, \text{ except for } i=j=k=m \]  
(6)
\[ Y_{i}^{m}/ + Y_{j}^{k}/ = 1, \forall i,j,k,m, \text{ except for } i=j=k=m \]  
(7)
\[ N, N_k \geq 0 \text{ and integer }, \forall k \]  
(8)
\[ Q_{j}(i) \geq 0, \forall k,i \]  
(9)

Function (1) is the objective of the model i.e. minimize total actual flow times all of the parts. Constraints (2) and (3) accomplish material balance in the shop and the items/respectively. Constraint (4) means that if the size of batch \( i \) of item \( k \) is positive, the setup time for item \( k \) in constraint (5) will be incurred. Constraint (5) ensures that all batches on schedule will be finished at the due date. Constraint (6) shows that the processing of batch \( K \) must be started at or after time zero. Constraint (7) ensures that if batch \( i \) of item \( k \) is to precede batch \( j \) of item \( m \) then starting time of batch \( j \) of item \( m \) must come later after the completion time of batch \( i \) of item \( k \), where constant \( M \) must sufficiently be large enough. Constraint (8) shows that given two batches, one must precede the other. \( X_{k} \) and \( Y_{k}^{m}/ \) are binary variables as shown by constraint (9). Constraint (10) shows values of total number of parts and number of all items must be integer. Number of parts in every batches must be non negative.

**METHOD**

Problem MISR can be viewed as a mix quadratic programming model. There are a lot of softwares such as LINGO that can be used to solve the problem. However, in order to solve Problem MISR using LINGO, it is required to create a relaxation for binary constraint, i.e., Constraint (9). It can developed some steps referring to using LINGO for the modified Problem MISR in a proposed algorithm to solve the original problem MISR. The proposed algorithm is then provided with steps to obtain integer values for the decision variables that must be integer. The remaining steps in the algorithm are based on the previous algorithm that have developed in Zahedi (2008) to solve the single item single resource (SISR) batch scheduling problem. The following algorithm is proposed to solve Problem MISR.

**Algorithm**

Step 1: Set all parts of each item as one batch. This leads to having \( K \) batches.

Step 2: Sequence the \( K \) batches in order of non-decreasing the ratio of \( (t_{k} n_k + s_k)/ n_k, \forall k \), or \( (t_{1} n_1 + s_1)/ Q_1 \leq (t_{2} n_2 + s_2)/ n_2 \leq . . . \leq (t_{K} n_K + s_K)/ n_K \).

Step 3: Generate a backward schedule of the batches on resource, based on the resulting sequence, considering the feasibility of the schedule. Determine \( T_{min} \) constituting the time interval from the starting time, \( B_k \), until the common due date.
Step 4: Problem MISR is considered feasible if and only if $T_{\text{min}} \leq d$, and go to step 5. Otherwise, the problem is not feasible, and stop.

Step 5: Compute $N_{k(maks)}$ for each item using the following equation

$$N_{k(maks)} = \lfloor 1/2 + \sqrt{1/4 + 2n_k t_k / s_k} \rfloor$$

where $\lfloor N \rfloor$ denotes the maximum integer less than or equal to $N$.

Step 6: Substitute all values of $K$ and $N_k$ where $N_k = N_{k(maks)}$, $n_k$, $t_k$, $s_k$ and $d$ to modify Problem MISR.

Step 7: Set $X_{11} = X_{21} = \ldots = X_{k1} = 1$, and set $X_{kj} = 0$ otherwise. Also, set $Y_{k}^{m} = 1$ if $k < m$ or $k=m$ and $i < j$, $\forall$ $i,j,k,m$, and set $Y_{k}^{m} = 0$, otherwise. Compute $F_{i}^{a}$, constituting the initial total actual flow time, for this setting.

Step 8: Set $k = K$.

Step 9: Set $i = 2$, then set $X_{ij} = 1$, for $j = 1, \ldots, i$.

Step 10: Solve the problem.

Step 11: Observe whether $B_{(ki)} \geq 0$ ?,

If $B_{(ki)} \geq 0$, write $F_{i}^{a}$.

Observe whether $F_{i}^{a} < F_{i-1}^{a}$. If $F_{i}^{a} < F_{i-1}^{a}$, set $i = i + 1$ and go to step 9. Otherwise, set $i-1 = N_{k(\text{opt})}$ and go to step 12.

Step 12: Set $k = K-1$. If $k > 0$, go to step 8. If $k = 0$, set $F_{i}^{a}$ as the optimal solution and recompute all decision variables.

RESULTS AND DISCUSSION

To clarify how the proposed algorithm works, the following example is given. Consider a problem with parts of 3 items. The quantities demanded of the respective items at the common due date, $d = 200$, are 40, 100, and 80. The processing times are 0.6, 0.8 and 0.5 respectively. The setup times for respective items are 2.4, 2.0, and 4.0. The computational steps to solve the problem are the followings.

Step 1 ~ Step 4: These steps give the resulting sequences: Item 3, Item 1, and Item 2. The time interval from the starting time, $B_3$, of batch processed first until the common due date, $T_{\text{min}} = 150.4 \leq d=200$.

Step 5: The numbers of batches for items 1, 2, and 3 respectively are 5, 5, and 9.

Step 6 ~ Step 7 substitute all values of the parameters to the formulation of modified Problem MISR and set the condition as shown in step 7 of the proposed algorithm result $F_{i}^{a} = 18592.36$. Step 8 ~ Step 12: These steps give the optimal solution shown in Table 1 with $F_{i}^{a} = 17966.44$.

Table 1.
The Optimal Solution of Example

<table>
<thead>
<tr>
<th>Position</th>
<th>Q_{ki}</th>
<th>B_{ki}</th>
<th>Item</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>44</td>
<td>178</td>
<td>3</td>
</tr>
<tr>
<td>2</td>
<td>36</td>
<td>156</td>
<td>3</td>
</tr>
<tr>
<td>3</td>
<td>40</td>
<td>128</td>
<td>1</td>
</tr>
</tbody>
</table>
On other side, to observe model behavior and the solution, some groups of cases are created as available in Table 2.

Table 2
Optimal Solution Some Groups of Cases

<table>
<thead>
<tr>
<th>Group</th>
<th>No</th>
<th>Data</th>
<th>Item-1</th>
<th>Item-2</th>
<th>Item-3</th>
<th>$F^*$</th>
<th>$N_{(0)}$</th>
<th>$B_{N_{opt}}$</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td>n t s</td>
<td>n t s</td>
<td>n t s</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>I</td>
<td>1</td>
<td>40</td>
<td>0.6 2.4</td>
<td>100</td>
<td>0.8 2.0</td>
<td>80</td>
<td>0.5 4.0</td>
<td>17966.44</td>
</tr>
<tr>
<td></td>
<td>2</td>
<td>40</td>
<td>0.6 2.4</td>
<td>80</td>
<td>0.8 2.0</td>
<td>80</td>
<td>0.5 4.0</td>
<td>14716.60</td>
</tr>
<tr>
<td></td>
<td>3</td>
<td>40</td>
<td>0.6 2.4</td>
<td>60</td>
<td>0.8 2.0</td>
<td>80</td>
<td>0.5 4.0</td>
<td>11781.31</td>
</tr>
<tr>
<td></td>
<td>4</td>
<td>40</td>
<td>0.6 2.4</td>
<td>40</td>
<td>0.8 2.0</td>
<td>80</td>
<td>0.5 4.0</td>
<td>9212.50</td>
</tr>
<tr>
<td></td>
<td>5</td>
<td>40</td>
<td>0.6 2.4</td>
<td>20</td>
<td>0.8 2.0</td>
<td>80</td>
<td>0.5 4.0</td>
<td>7021.10</td>
</tr>
<tr>
<td>II</td>
<td>6</td>
<td>40</td>
<td>0.6 2.4</td>
<td>100</td>
<td>0.8 2.0</td>
<td>80</td>
<td>0.5 2.0</td>
<td>17332.44</td>
</tr>
<tr>
<td></td>
<td>7</td>
<td>40</td>
<td>0.6 2.4</td>
<td>100</td>
<td>0.8 2.0</td>
<td>80</td>
<td>0.5 3.0</td>
<td>17649.94</td>
</tr>
<tr>
<td></td>
<td>1</td>
<td>40</td>
<td>0.6 2.4</td>
<td>100</td>
<td>0.8 2.0</td>
<td>80</td>
<td>0.5 4.0</td>
<td>17966.44</td>
</tr>
<tr>
<td>III</td>
<td>8</td>
<td>40</td>
<td>0.6 2.4</td>
<td>100</td>
<td>0.4 2.0</td>
<td>80</td>
<td>0.5 4.0</td>
<td>13568.67</td>
</tr>
<tr>
<td></td>
<td>9</td>
<td>40</td>
<td>0.6 2.4</td>
<td>100</td>
<td>0.6 2.0</td>
<td>80</td>
<td>0.5 4.0</td>
<td>17150.33</td>
</tr>
<tr>
<td></td>
<td>1</td>
<td>40</td>
<td>0.6 2.4</td>
<td>100</td>
<td>0.8 2.0</td>
<td>80</td>
<td>0.5 4.0</td>
<td>17966.44</td>
</tr>
</tbody>
</table>

CONCLUSION

The foregoing has presented a model and algorithm to solve binary quadratic programming model from batch scheduling that determines number of parts in batches and scheduling in single formulation for multiple-item cases. Problem MISR appears to be an NP-hard problem, and can be viewed as a mixed quadratic programming for which the proposed algorithm takes advantages of the so-called LINGO software. However, a relaxation of binary constraints is needed to make the formulation of Problem MISR solvable using the LINGO software. In view of the Results and Discussion, the conclusions are the proposed algorithm can solve problem MISR effectively. The last item scheduled (first in process) has more batches since batching of the last item can increase total actual flow time. The resulting solution are optimal, since LINGO gives the optimal solution and the steps provided in proposed algorithm for obtaining the best binary value search from all alternative binary values. The research underway is to extend problem MISR to cover the issue in actual problems such as multiple due dates and multiple process stages.

REFERENCES

