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European Journal of Operational Research 149 (2003) 377-389

EUROPEAN JOURNAL OF OPERATIONAL RESEARCH

www.elsevier.com/locate/dsw

# Scheduling multiprocessor tasks on parallel processors with limited availability

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#### Abstract

In this work we consider the problem of scheduling multiprocessor tasks on parallel processors available only in restricted intervals of time called time windows. The multiprocessor task model applies to modern production systems and parallel applications in which several processors can be utilized in parallel. Preemptable tasks are considered. Polynomial time algorithms are given in three cases: the case of maximum lateness criterion and a fixed number of processors, the case of schedule length criterion when tasks have various ready times and require either one or all processors, and in case of schedule length criterion when the sizes of the tasks are powers of 2.

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Keywords: Scheduling; Parallel computing; Multiprocessor tasks; Parallel tasks; Gang scheduling; Co-scheduling; Bandwidth allocation; Time windows

#### 1. Introduction

In a multiprocessor task system it is allowed for some tasks to use several processors in parallel. This assumption is justified by various technological and efficiency reasons: In fault-tolerant systems tasks may be duplicated to obtain results securely [13]. Mutual testing of processors needs at least two processors working in parallel [16]. In

parallel computer systems with time-sharing, co-scheduled parallel applications, i.e. running on several processors in the same time quantum, utilize processors with higher efficiency than applications scheduled according to other policies [13,18,26]. In bandwidth allocation [1] applications have to reserve sufficient channel capacity for a given amount of time in advance. The required bandwidth plays the role of processors. In the scheduling literature multiprocessor tasks are also referred to as parallel, concurrent, or malleable tasks. The idea of multiprocessor task scheduling is also conveyed by gang scheduling. For a more detailed survey of scheduling multiprocessor tasks, we refer the reader to [3,11,25].

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In this work we assume that the availability of processors is restricted, and some machines are available only in certain intervals called time windows. Time windows may appear in case of computer breakdowns or maintenance periods. In any multitasking computer system, and in hard real time systems in particular, urgent tasks have high priority and are pre-scheduled in certain time intervals, thus creating time windows of processor availability. Scheduling in time windows was considered in e.g. [19,22–24].

Now we shall introduce the notation used in the paper. We consider a set  $\mathcal{P} = \{P_1, \dots, P_m\}$  of m > 11 parallel identical processors. Processors are accessible in time windows. There are l time windows. Each time window i = 1, ..., l is characterized by:  $s_i$ —its start time,  $m_i$ —the number of processors in window i. The end of one time window is also the beginning of the succeeding window. Without the loss of generality we assume that  $s_1 = 0$ ,  $s_i < s_{i+1}$ ,  $m_i \neq m_{i+1}$ , and after time  $s_{l+1}$  no processor is available. Our definition of time windows is equivalent to the so-called staircase patterns of processor availability [23]. In the staircase pattern processors can be renumbered so that the intervals available on processor  $P_{i+1}$  are also available on processor  $P_i$  (cf. Fig. 1b). In this paper we consider the preemptive schedules, and any processor availability pattern can be converted to a staircase pattern by renumbering the processors within the time windows (see Fig. 1a and b).

The multiprocessor task set is  $\mathcal{T} = \{T_1, \dots, T_n\}$ . Tasks are preemptable, which means that they may be suspended and restarted later and possibly on different processors without incurring any additional costs. Nonpreemptable tasks must be continuously executed on the same processor from the beginning until the very end, i.e. they cannot be suspended or migrate. Each task  $T_i \in \mathcal{T}$  is defined by: size,—the number of processors required in parallel,  $p_i$ —the processing time,  $r_i$ —the ready time,  $d_i$ —the due-date. The number of processors simultaneously required by a task will be called the size of the task. The set  $\mathcal{T}$  of tasks can be divided into subsets according to the sizes of the tasks. Thus,  $\mathcal{T} = \mathcal{T}^1 \cup \mathcal{T}^2 \cup \cdots \cup \mathcal{T}^m$ , where  $\mathcal{T}^i$  is a set of the  $n_i$  tasks which are executed by i processors.

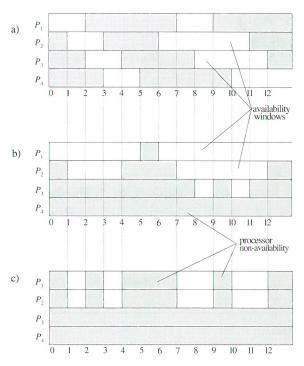


Fig. 1. Converting restricted intervals of availability. (a) The original pattern, (b) availability windows converted into a staircase pattern, (c) availability windows converted into a staircase pattern when allocation across processor partitions is not allowed.

Note that  $n = n_1 + n_2 + \cdots + n_m$ , and that the number of tasks with the size next to size, is  $n_{\text{size}_{i+1}}$ (not  $n_{i+1}$ ). For the conciseness of the text we will name the tasks of size i as i-tasks. We assume that tasks are numbered according to their size, i.e.  $size_1 \geqslant size_2 \geqslant \cdots \geqslant size_n$ . In case of parallel computers with hypercube interconnection or in buddy processor allocation systems [12,17] tasks may not have arbitrary sizes. To avoid the excessive fragmentation of the processor set, tasks have sizes that are multiples of each other. For example, sizes can be powers of 2. This situation will be called a cube case. In the cube case only compact allocations are allowed, i.e. a k-task can be assigned only to processors  $P_{(i-1)k+1}, \ldots, P_{ik}$ , where  $i=1,\ldots,m/k$ . Compact allocations result in perceiving time windows differently by tasks of different sizes (e.g. see Fig. 1c for 2-tasks). This assumption is discussed in Section 4.

The optimality criteria we consider in this paper are: schedule length  $C_{\max} = \max_{T_j \in \mathscr{T}} \{c_j\}$ , and maximum lateness  $L_{\max} = \max_{T_j \in \mathscr{T}} \{c_j - d_j\}$ , where  $c_j$  is the completion time of task  $T_j \in \mathscr{T}$ .

We shall denote the analyzed scheduling problems according to the three-field notation introduced in [15,25] (cf. also [3,11]). Additionally, symbol *win* in the processor field denotes the limited availability of the processors.

Below we shall review the related results on the preemptable scheduling of multiprocessor tasks. NP-hard problems will be denoted as NPh, and strongly NP-hard problems as sNPh. When the number of processors is a part of the instance input, scheduling multiprocessor tasks is computationally hard (strictly problem  $P|\text{size}_i, pmtn|C_{\text{max}}$ is NPh) [10]. Adding precedence constraints makes problem  $P3|\text{size}_i, pmtn, \text{chains}|C_{\text{max}}|$  sNPh, which follows directly from the complexity of problem  $P3|\text{size}_i, p_i = 1, \text{chain}|C_{\text{max}}$  [6] for unit execution time nonpreemptable tasks. Mixing precedence constraints and task due-dates makes scheduling multiprocessor tasks sNPh even for two processors. Hence,  $P2|\text{size}_i$ , pmtn, chain  $|L_{\text{max}}|$  is computationally hard, which follows from the proof given for problem  $P2|\text{size}_i, p_i = 1$ , chain  $|L_{\text{max}}|$ [7]. Consequently, P2, win|size<sub>j</sub>, pmtn, chain| $L_{max}$  is sNPh.

When the number of processors is fixed, problem  $Pm|\text{size}_i, pmtn|C_{\text{max}}$  can be solved in polynomial time [2]. Restricting possible sizes of the tasks to multiples of each other reduces the problem to case *cube* which is computationally easier because matching task sizes to fit in the processor system is easy [8]. Therefore, the problem of scheduling preemptable tasks on identical hypercubes  $P|\text{cube}_i, pmtn|C_{\text{max}}$  can be solved in  $O(n^2 \log^2 n)$ time [27], and scheduling on hypercubes with different speeds (problem  $Q|\text{cube}_i, pmtn|C_{\text{max}}$ ) is solvable in  $O(n \log n + nm)$  time [9]. When the precedence constraints and ready times are not present simultaneously, scheduling the multiprocessor tasks on two processors is computationally tractable. Thus, problem  $P2|\text{size}_i, pmtn, r_i|L_{\text{max}}$ is solvable in polynomial time using a linear programming (LP) formulation, while problem  $P2|\text{size}_j, pmtn, \text{prec}|C_{\text{max}}$  can be solved in  $O(n^2)$ time [4].

In the following sections we present the polynomial time algorithms for three cases of scheduling multiprocessor tasks in time windows. We start with the most general problem Pm, win|size<sub>j</sub>, pmtn,  $r_j|L_{\max}$  to be considered in Section 2. Then, we present low-order polynomial time algorithms for more specialized problems: P, win|size<sub>j</sub>  $\in$   $\{1, m\}$ , pmtn,  $r_j|C_{\max}$  in Section 3, and problem P, win|cube<sub>j</sub>,  $pmtn|C_{\max}$ , in Section 4.

## 2. Pm, win|size<sub>i</sub>, pmtn, $r_i|L_{max}$

The method we propose here is based on the use of LP and the notion of processor feasible sets of tasks. The processor feasible set of tasks is such a set of tasks that can be feasibly executed in parallel on the given number of processors. The number of different processor feasible sets is at most  $\sum_{i=1}^{m} \binom{n}{i}$ which is  $O(n^m)$ , and is polynomially bounded with respect to the input size, provided m is fixed. Let us consider the tasks which can be executed in parallel for some given value L of maximum lateness criterion  $L_{\text{max}}$ . According to the definition of the maximum lateness,  $T_i$  cannot be executed after time  $d_i + L$ . We will call this value the deadline of  $T_i$ . Only tasks with ready times smaller than  $d_i + L$  can be executed in parallel with  $T_i$ . Also the number of available processors determines the set of tasks which can be executed in parallel. If, due to a change of L, deadline of  $T_i$  shifts from one window of processor availability to another, also the admissible processor feasible sets change. Thus, the collection of processor feasible sets changes with changing L in two types of cases: when  $r_i = d_i + L$ for two tasks  $T_i$ ,  $T_j$ , and when  $s_k = d_j + L$  for some task  $T_i$  and window k. The first case may take place  $\binom{n}{2}$  times, the second ln times. This defines  $O(n^2 + nl)$  intervals  $[L_a, L_{a+1}]$  of criterion  $L_{max}$  values for which the collection of processor feasible sets is constant. In each of the  $[L_a, L_{a+1}]$  intervals the sequence of events (ready times, deadlines, starting a window) is constant and known. The events are ordered according to the increasing time of their appearance. Let us assume that we know that the optimal value of the maximum lateness satisfies  $L \in [L_a, L_{a+1}]$ . We will denote by  $e_i^a(i =$  $1, \ldots, 2n + l + 1$ ) the time instant at which event i

takes place. However, if event i is representing a deadline of some task  $T_i$  then  $e_i^a = d_i$  (value of the maximum lateness is not added). Let us associate with each event i value  $g_i = 1$  when event i represents a deadline, and  $g_i = 0$  otherwise. For each task  $T_i$  it is possible to find index  $u_i$  of the event associated with task ready time  $r_i$ , and index  $v_i$  of the event related to  $T_j$ 's deadline. Hence,  $T_j$  can be executed only in the interval  $[e_{u_i}^a + g_{u_j}L, e_{v_i}^a + g_{v_j}L]$ . Let  $M_i$  denote the family of the processor feasible sets of tasks existing between a pair (i, i + 1) of events, and  $M_{ij}$  the collection of the indices of processor feasible sets in this interval which include task  $T_i$ . With a processor feasible set with number k existing between events (i, i + 1) we associate variable  $x_{ik}$  denoting the duration of executing the processor feasible set. Assuming we know that value  $L \in [L_a, L_{a+1}]$ , our problem can be formulated as a linear program (LP):

Minimize L

subject to 
$$\sum_{i=u_{j}}^{v_{j}-1} \sum_{k \in M_{ij}} x_{ik} = p_{j} \quad j = 1, \dots, n$$

$$\sum_{i=1}^{|M_{i}|} x_{ik} \leqslant e_{i+1}^{a} - e_{i}^{a} + L(g_{i+1} - g_{i})$$
(1)

$$i = 1, \dots, 2n + l \tag{2}$$

$$L_a \leqslant L \leqslant L_{a+1} \tag{3}$$

$$x_{ik} \geqslant 0$$
  $i = 1, \dots, 2n + l,$   
 $k = 1, \dots, |M_i|$  (4)

In the above LP equalities (1) guarantee that tasks receive the required processing time. Inequalities (2) ensure that processor feasible sets defined for the interval between events (i, i + 1) are not executed beyond this interval. The above LP also defines a feasible schedule because in the interval between events (i, i + 1) only a processor feasible set with  $x_{ik} > 0$  is executed. The LP has  $O(n^m(n^2 +$ nl) variables and O(n+l) constraints (not including constraints (4)). Thus, LP can be formulated and solved in polynomial time provided that m is fixed. To identify the interval  $[L_a, L_{a+1}]$  of  $L_{\text{max}}$ values which contains the optimal value of maximum lateness (cf. constraint (3)) one needs to perform  $O(\log n + \log l)$  binary search steps solving a linear program of the type described above. Since the processor system has restricted availability, a feasible solution may not exist. In this case LP is not feasible even in the last acceptable interval of  $L_{\text{max}}$  values.

# 3. P, win|size<sub>i</sub> $\in$ {1, m}, pmtn, $r_i$ | $C_{max}$

In this section we examine the problem of preemptively scheduling multiprocessor tasks with two possible sizes: 1 or m. Tasks have different ready times. The optimality criterion is schedule length.

Before we go into any further details we give an informal description of the algorithm. When new tasks appear at some ready time, the algorithm must decide which of the ready tasks to execute. In any feasible schedule, if an m-task can be executed in some earlier time interval occupied by 1-tasks, then swapping the m-task with the 1-tasks is feasible and does not change the schedule length. Therefore, *m*-tasks can be executed as soon as they are ready and m processors are available. 1-tasks, on the other hand, must be scheduled more carefully, because the longest 1-task, and the sum of processing requirements of 1-tasks in the last occupied window determine the schedule length. The algorithm given by Muntz and Coffman [21] is capable of minimizing these two values before reaching the last occupied interval. The algorithm assigns to the tasks processing capacities  $\bar{\beta} =$  $[\beta_1, \ldots, \beta_n]$  which can be intuitively understood as fractions of the available processing capability. The capacities are calculated on the basis of the task height. The height h(j) of task  $T_i$  is the processing time required to complete it. Initially  $h(j) = p_j$ , for  $T_j \in \mathcal{T}^1$ . The ready times, and window beginnings are events in the system. Let us sort all  $q \le l + n + 1$  events according to their time value, i.e.  $e_1 = 0 \le e_2 \le e_3 \cdots \le e_q$ . In the following algorithm we denote by A the set of 1-tasks ready by time t, by B the set of m-tasks ready by time t, and by  $\mu_k$  the number of processors available in interval  $(e_k, e_{k+1})$ .

An Algorithm for  $P, \text{win} | \text{size}_j \in \{1, m\}, \\ pmtn, r_j | C_{\text{max}} \\ 1: t := 0; B := \emptyset; A := \emptyset; \\ 2: \text{ for } k := 1 \text{ to } q - 1 \text{ do begin}$ 

```
B := B \cup \{T_i | T_i \notin B, \text{size}_i = m, r_i \leqslant t\};
3:
       \tau := \min\{\sum_{T_i \in B} p_j, e_{k+1} - e_k\}; \ \varepsilon := \tau;
       if \tau \neq 0 and \mu_k = m then
       begin (*m-tasks are ready and interval k has
       enough processors*)
6:
         while B \neq \emptyset and \varepsilon > 0 do
          if \varepsilon \geqslant p_i then begin execute p_i units of T_i
          in [e_k, e_k + \tau]; \varepsilon := \varepsilon - p_i; B := B - \{T_i\}
          end
          else begin execute \varepsilon units of T_i in
          [e_k, e_k + \tau]; p_i := p_i - \varepsilon; \varepsilon := 0; end;
7:
          t := t + \tau;
          end;
8:
          A := A \cup \{T_i | T_i \not\in A, \text{size}_i = 1, r_i \leqslant t\};
9:
          order tasks in A according to nonincreas-
          ing h(j);
10:
          while (e_{k+1} > t) and (\exists_{T_i \in A} h(j) > 0) do
         CAPABILITIES (A, k, \overline{\beta});
11:
12:
         calculate times:
         \tau' := \min \left\{ \infty, \min_{j} \left\{ \frac{h(j) - h(j+1)}{\beta_j - \beta_{j+1}} : \right. \right.
            T_j, T_{j+1} \in A, \beta_j \neq \beta_{j+1}, h(j) > h(j+1)
         (*\tau' is the shortest time required for two
         1-tasks T_i, T_{i+1} with
         different heights to become equal with re-
         spect to their height*) if \beta_{|A|} > 0 then
         \tau'' := h(|A|)/\beta_{|A|} else \tau'' := \infty;
         (*\tau" is the time to the earliest completion
         of any 1-task*)
         \tau := \min\{\tau', \tau'', e_{k+1} - t\};
13:
14:
         for T_i \in A do schedule a piece of length
         \tau \beta_i of task T_i in interval [t, t + \tau] accord-
         ing to McNaughton rule [20];
15:
         for T_i \in A do
             begin h(j) := h(j) - \tau \beta_i; if h(j) = 0
             then A := A - \{T_i\} end;
16:
        t := t + \tau;
      end;
procedure CAPABILITIES(in:X, k; out:\overline{\beta});
  begin
```

 $\beta := 0$ ; avail :=  $\mu_k$ ; (\* $\mu_k$  is the number of

free processors in interval  $k^*$ )

```
18:
      while avail > 0 and |X| > 0 do
19:
         Y: = the set of the highest tasks in X;
20:
        if |Y| > avail then
21:
           begin for T_i \in Y do \beta_i := avail/|Y|;
          avail := 0; end;
        else (*tasks in Y cannot use all avail
        processors*)
22:
         begin for T_j \in Y do \beta_i := 1; avail :=
         avail - |Y|; end;
23:
        X := X - Y;
    end; (*of while loop*)
   end; (*of procedure CAPABILITIES*)
```

**Theorem 1.** The above algorithm constructs an optimal schedule for problem P, win|size<sub>j</sub>  $\in \{1, m\}$ ,  $pmtn, r_i | C_{max}$  and has time complexity  $O(n^2 + nl)$ .

**Proof.** We start with the proof of schedule feasibility. Only the tasks in set A and B are executed. Task  $T_i$  is added to one of these sets only if the current time is bigger than  $r_i$  (lines 3 and 8). Thus, ready times are observed. m-tasks are executed only in windows with a sufficient number of available processors (line 5). Partial schedules for 1-tasks (built in line 14) are feasible because  $\tau \beta_i \leqslant \tau$ for  $T_j \in A$ , and  $\sum_{T_i \in A} \beta_j \tau \leq \mu_k \tau$  by the formulation of the procedure CAPABILITIES. Hence, the conditions set in the McNaughton rule [20] are satisfied. A task is removed from the respective set A or B, and stops being processed only when the sum of the pieces of work on the task is equal to the task processing requirement (lines 6, 15). Thus, the schedule is feasible.

Now, we prove the optimality of the schedules. The schedule length is determined by the last completion time of a task finished in the last used window. *m*-tasks are processed as soon as they appear in the system and a free window is available. Thus, *m*-tasks are completed in the earliest possible time. 1-tasks executed in the last occupied window are scheduled according to the McNaughton rule [20]. Therefore, the completion time for the tasks in the last window depends on: the longest remaining piece of a single 1-task, and the sum of the processing requirements of all remaining 1-tasks. By lines 19 to 22 the highest (i.e. the longest) 1-tasks always receive the biggest

processing capabilities. By calculation of  $\tau'$  in line 12 an initially higher task may not reduce its height below some initially lower task. Thus, the remaining processing requirement of the longest 1-task is reduced in the biggest feasible way before reaching the last occupied window. By line 18 no processor remains idle when a task ready to use the processor exists. Thus, also the sum of the remaining processing requirements is minimized in the maximum possible way before reaching the last used window. We conclude that the length of the schedule in the last occupied window is the shortest possible and the schedule is optimal.

The computational complexity of the algorithm can be calculated as follows: Loop 2 is executed O(n+l) times. Lines 3–8 can be executed in time O(n). Line 9 can be executed in  $O(n_1^2)$  total time over all algorithm run. This is a result of sorting 1-tasks once and merging the arriving tasks with tasks in set A. The procedure CAPABILITIES is executed in  $O(n_1)$  time (assuming  $\mu_k \leq n_1$ ). Lines 12–15 are done in  $O(n_1)$  time. Thus, loop 10 is executed in time  $O(n_1)$ . The total running time of the algorithm is  $O(n^2 + nl)$ .

When a feasible solution does not exist due to insufficient capacity of the processor system, the algorithm stops with  $A \neq \emptyset$  or  $B \neq \emptyset$ . The above algorithm can be executed on-line because we only need information about the ready tasks, and the time of the next event.

#### 4. P, win|cube<sub>i</sub>, pmtn| $C_{\text{max}}$

In this section we consider scheduling tasks the sizes of which are multiples of each other. This implies that m is a multiple of the greatest task size. Only compact task assignments are allowed. The schedule length is the optimality criterion. Ready times and due-dates are not taken into account. For the sake of the presentation simplicity we assume that the sizes of the tasks and m are powers of 2. The method we use bears some similarity to the one used in [5,9] to solve problems

 $Q|\text{size}_j \in \{1, k\}, pmtn|C_{\text{max}}, Q|\text{cube}_j, pmtn|C_{\text{max}}, \text{ and in [14] to solve } Q|pmtn|C_{\text{max}}.$ 

As mentioned earlier, only *compact* assignments are allowed, i.e. k-tasks can be assigned only to processors  $P_{(a-1)k+1}, \ldots, P_{ak}$ , where  $a = 1, \ldots, m/k$ . This assumption has both theoretical and practical justifications. Any interval of a feasible schedule for problem P, win|cube<sub>i</sub>,  $pmtn|C_{max}$  can be converted to a compact assignment by renumbering the processors within the interval. In real computer systems [12,18] processor partitions are used to eliminate the fragmentation of the resources, and assigning tasks across the partition border is impossible. This may lead to different windows available for tasks of different sizes. For example, when the pattern presented in Fig. 1a is converted to a staircase pattern in Fig. 1b, window [1,4] is available for 2-tasks. This window comprises three subintervals in which different real processors are available. If the processor partitions are  $\{P_1, P_2\}$ ,  $\{P_3, P_4\}$ , then interval [2,3] is not accessible for 2-tasks because it is an assignment across the partition border. Consequently, when the task assignment across the processor partitions is forbidden, the availability windows depend on the size of the tasks and the bigger the size of the tasks is the more restricted a window can be. Yet, the algorithm presented below can also be applied in this situation. The schedules are optimal because Theorem 3 (presented in the sequel) holds and the amount of processing capacity left after assigning a task of a certain size is the biggest possible.

Let us outline the main stages of the algorithm before presenting the details. First the algorithm calculates a lower bound on the schedule length. Then tasks are scheduled in the order of descending size. The tasks with the same size are scheduled in the order of nonincreasing processing time. When a feasible schedule exists, then its length is equal to the lower bound and thus it is optimal. When a feasible solution does not exist, then there is some task  $T_j$  which needs additional processing capacity. In such a case the schedule length is extended so that the processing requirement of  $T_j$  can be met. In the following we describe each stage of the algorithm.

## 4.1. Lower bound on the schedule length

As already stated, we consider windows of the processor availability which are staircase patterns. Therefore,  $P_1$  is the processor capable of accommodating the biggest processing requirement. Processor  $P_m$  has the least processing ability. Let us call by *processing capacity* PC(i,t) the amount of work that can be performed on processor  $P_i$  until time t. To calculate the processing capacity of a processor we use function g(i,k,t) which is free processing time that can be used on  $P_i$  in window k until time t, defined as follows:

$$g(i,k,t) = \begin{cases} 0 & \text{if } i > m_k \text{ or } t < s_k \\ t - s_k & \text{if } i \le m_k \text{ and } s_k \le t \le s_{k+1} \\ s_{k+1} - s_k & \text{if } i \le m_k \text{ and } s_{k+1} < t \end{cases}$$

The processing capacity is  $PC(i, t) = \sum_{k=1}^{t} g(i, k, t)$ . Due to the staircase pattern we have  $PC(1, t) \ge PC(2, t) \ge \cdots \ge PC(m, t)$  for all t.

For the simplicity of the presentation let us assume that  $\operatorname{size}_1 \geqslant \operatorname{size}_2 \cdots \geqslant \operatorname{size}_n$ , and  $p_{b+1} \geqslant \cdots \geqslant p_{b+n_k}$ , where  $b = \sum_{i=(\log_2 k)+1}^{\log_2 m} n_{2^i}$  is the number of tasks with the size bigger than k. A k-task can be assigned only to processors  $P_{(a-1)k+1}, \ldots, P_{ak}$  existing in the staircase pattern, where a = 1, ..., m/k. Hence, the processing capacity available for k-tasks until time t is defined by PC(ka, t), where a = $1, \ldots, m/k$ . The value of the lower bound on the schedule length must guarantee that the longest task fits on the processor with the greatest processing capacity, the two longest tasks can be accommodated on two processors with the greatest processing capacity, etc. (cf. inequalities (5)). The total processing requirement must fit in the total available processing capacity (cf. (6)). For the completeness of the following argument, and without the loss of generality we assume that  $n_k \ge m/k$ . Let  $C_k'$  be the value of the lower bound calculated solely for k-tasks. The above observations can be formulated as the following requirement:

$$\sum_{i=1}^{a} PC(ki, C'_k) \ge \sum_{j=b+1}^{b+a} p_j,$$

$$a = 1, \dots, (m/k) - 1$$
(5)

$$\sum_{i=1}^{m/k} PC(ki, C'_k) \ge \sum_{i=h+1}^{b+n_k} p_j$$
 (6)

Inequalities (5) and (6) determine the capacity available for k-tasks. However, there are tasks of various sizes in our problem. To deal with sizes greater than k we propose to substitute each task  $T_j$  of size j > k, with size j/k artificial k-tasks with the processing requirement  $p_j$ . Let  $p_{1k} \ge p_{2k} \ge \cdots \ge p_{n'_k k}$  be the processing times of the real k-tasks and the artificial tasks obtained by the reduction to size k.  $n'_k = \sum_{i=(\log_2 k)}^{\log_2 m} n_{2i} 2^i/k$  is the total number of such tasks. Then, the lower bound on the schedule length for the original k-tasks and the artificial ones can be found as the minimum  $C_k$  satisfying the following inequalities:

$$\sum_{i=1}^{a} PC(ki, C_k) \geqslant \sum_{j=1}^{a} p_{jk},$$

$$a = 1, \dots, (m/k) - 1$$
(7)

$$\sum_{i=1}^{m/k} PC(ki, C_k) \geqslant \sum_{i=1}^{n'_k} p_{jk}$$
 (8)

The lower bound on the schedule length can be found as  $C = \max_{i=0}^{\log_2 m} \{C_{2^i}\}$ . Let us illustrate the above concepts by means of an example.

**Example 1.** Task data: n = 8

Task	$T_1$	$T_2$	$T_3$	$T_4$	$T_5$	$T_6$	$T_7$	$T_8$
Size	4	4	4	2	2	2	1	1
Processing	4	2	1	3	2	2	4	2
time								

Processor window data: l = 4

11000	SOI WILL	iow data	$\iota$ . $\iota$ — $\iota$		
Windo numb		2	3	4	5
$S_j$	0	2	4	6	$\infty$
$m_j$	6	4	8	6	0

The following table presents the processing capacities PC(i,t) available on the consecutive processors  $P_i$  by time t.

Interval of <i>t</i>	[0, 2]	[2, 4]	[4, 6]	$[6,\infty]$
PC(1,t)	t	t	t	t
PC(2, t)	t	t	t	t
PC(3, t)	t	t	t	t
PC(4, t)	t	t	t	t
PC(5,t)	t	2	t-2	t-2
PC(6, t)	t	2	t-2	t-2
PC(7,t)	0	0	t-4	2
PC(8,t)	0	0	t-4	2

Now, we calculate the lower bound on the schedule length. We have  $C_4 = 5.5$ , because PC(4, 5.5) +  $PC(8, 5.5) = p_1 + p_2 + p_3 = 7$ . For the calculation of  $C_2$  we have the following processing times of the original and artificial tasks of size 2: 4, 4, 3, 2, 2, 2, 2, 1, 1.  $C_2$  must be in interval  $[6, \infty]$  to accommodate this amount of work. For this interval we have  $PC(2, C_2) + PC(4, C_2) + PC(6, C_2) +$  $PC(8, C_2) = 3C_2 \ge 21$ , i.e. processing capacity is at least equal to the total processing requirement of real and artificial 2-tasks. Hence, we have  $C_2 \ge 7$ . The processing times of real and artificial 1-tasks are: 4, 4, 4, 4, 4, 3, 3, 2, 2, 2, 2, 2, 2, 2, 2, 2, 1, 1, 1, 1. To accommodate this amount of work t must be in interval  $[6, \infty]$ . In case of  $C_1$  and inequality (8) processing capacity  $\sum_{i=1}^{8} PC(i, t) = 6C_1$ , must be at least equal to 48. Hence  $C_1 = 8 = C$ .

Let us analyze the time complexity of calculating bound C. The construction of artificial tasks requires creating at most nm tasks of size 1, nm/2tasks of size 2, etc. Thus, there will be O(nm) artificial tasks altogether. From these tasks we have to select a the longest ones for inequalities (7). This can be done in O(nm) time for all sizes because a is O(m). Thus, the construction of data for the righthand side of inequalities (7), (8) needs O(nm) time. Functions g(i, k, t) can be calculated in O(ml) time. The processing capacity PC(i,t), as a piecewiselinear function of time t can be constructed in O(l)time. The calculation of  $C_k$  in each of inequalities (7) and (8) requires O(l) time. All inequalities can be verified in O(ml) time. Hence, the running time of calculating the lower bound is O(nm + ml).

## 4.2. Task scheduling rule

Tasks are assigned to processors in descending order of sizes and processing times. The processing capacity that remains after assigning tasks of a certain size is utilized by the tasks of smaller sizes. Therefore, the tasks of bigger sizes should use as little processing capacity as possible. Let us suppose tasks of sizes  $m, \ldots, 2k$  are already scheduled, and tasks of size k are about to be assigned to the processors. Let  $PC(1) \ge \cdots \ge PC(m)$  denote the remaining processing capacities of the processor system for the schedule length C. Without the loss of generality we assume that  $n_k \ge m/k$ . For the existence of a feasible assignment of k-tasks only (tasks with smaller sizes are not taken into account) it is necessary and sufficient that the following inequalities hold:

$$\sum_{i=1}^{a} PC(ki) \geqslant \sum_{j=b}^{b+a} p_{j} \quad a = 1, \dots, (m/k) - 1$$
 (9)

$$\sum_{i=1}^{m/k} PC(ki) \geqslant \sum_{j=b}^{b+n_k} p_j \tag{10}$$

where  $b = \sum_{i=(\log_2 k)+1}^{\log_2 m} n_{2^i}$ . The necessity of satisfying the above inequalities is obvious. The sufficiency is demonstrated in the following discussion.

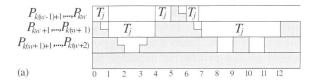
Scheduling rule R

In order to schedule k-task  $T_j$  find pair  $(P_{kw}, P_{k(w+1)})$  of processors satisfying:  $PC(kw) \ge p_j > PC(k(w+1))$ .

Fill processors  $P_{kw+1}, \ldots, P_{k(w+1)}$  with PC(k(w+1)) units of task  $T_j$  in the windows available on  $P_{k(w+1)}$ .

Assign the remaining processing requirement of  $T_j$  equal to  $p_j - PC(k(w+1))$  to processors  $P_{k(w-1)+1}, \ldots, P_{kw}$  from left to right in the windows which are not occupied by  $T_j$  on processors  $P_{kw+1}, \ldots, P_{k(w+1)}$ , and are available on  $P_{kw}$  (cf. Fig. 2a).

Update the remaining processing capacities:  $PC(kw - j) := PC(kw - j) - (p_j - PC(k(w + 1)),$  for j = 0, ..., k - 1. Construct new composite processors from the remaining intervals of



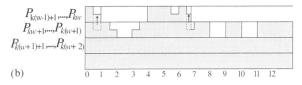


Fig. 2. Scheduling rule. (a) Assignment of a *k*-task to the processors, (b) restoring of the staircase pattern.

availability on pairs of processors  $P_{k(w+1)-j}$  and  $PC_{k(w+2)-j}$ , for  $j=0,\ldots,k-1$  (cf. Fig. 2b). PC(k(w+2)-j):=PC(k(w+2)-j)+PC(k(w+1)-j)-PC(k(w+1)), for  $j=0,\ldots,k-1$ . Remove processors  $P_{kw+1},\ldots,P_{k(w+1)}$  from data structures holding information about the available processing capacity. Renumber processors  $P_{k(w+1)+1},\ldots,P_m$  to  $P_{kw+1},\ldots,P_{m-k}$ , respectively. Rebuild the windows to respect staircase patterns, if necessary.

Let us comment on the above rule. The assignment of task  $T_i$  is feasible because  $T_i$  receives the required processing and is executed exactly on k processors. The rule has a special form when  $PC(m) \ge p_i$ . If this is the case we assume that PC(m+k) = 0, and apply the rule as described above. Since multiprocessor tasks require several processors simultaneously, some intervals in the windows on  $P_{k(w-1)+1}, \ldots, P_{k(w+1)-1}$  may remain idle, because processors  $P_{kw}$  and  $P_{k(w+1)}$  determined the processing capacity available for a k-task. After the assignment of  $T_i$  some intervals on  $P_{kw+1}, \dots, P_{k(w+1)-1}$  may still be free but unavailable on  $P_{k(w-1)+1}, \ldots, P_{kw}$  (cf. Fig. 2b). Such free intervals must be moved to  $P_{k(w-1)+1}, \ldots, P_{kw}$  in order to restore the staircase pattern. The scheduling rule violates neither inequalities (9) nor (10) as shown in the following theorem.

**Theorem 2.** Inequalities (9) and (10) are invariant when assigning k-tasks, according to scheduling rule R.

**Proof.** Let us suppose that initially inequalities (9) and (10) were satisfied, and  $T_j$  was assigned to  $P_{k(w-1)+1}, \ldots, P_{kw}$  and  $P_{kw+1}, \ldots, P_{k(w+1)}$  by rule R. We begin with the proof of the invariance of inequalities (9).

Due to applying rule R the number of inequalities in (9) is reduced to (m/k) - 2. The initial processor  $P_{k(w+1)}$  is removed from data structures holding information about the available processing capacity. Values of  $PC(k), \ldots, PC(k(w-1))$  are the same as before applying rule R. The new PC(kw) is equal to the initial  $PC(kw) + PC(k(w+1)) - p_j$ . The new values of PC(k(w+i)) are the old values PC(k(w+i+1)) for  $i = 1, \ldots, (m/k) - w - 2$ .

As  $T_j$  was the longest k-task, its processing time  $p_j$  is subtracted on the right-hand side of all inequalities (9). It remains to be shown that the left-hand side is not reduced by more than  $p_j$  in any of inequalities (9). Values of  $PC(k), \ldots, PC(k(w-1))$  did not change. Hence, inequalities (9) hold for  $a=1,\ldots,w-1$ , because the left-hand side did not change. In the new value of the sum on the left-hand side of (9), for  $a=w,\ldots,(m/k)-2$ , only the new PC(kw) is different than before applying rule R, and is equal to the initial  $PC(kw)+PC(k(w+1))-p_j$ . Hence, the left-hand side did not decrease more than  $p_j$ , and inequalities (9) hold for  $a=i,\ldots,(m/k)-2$ .

Inequality (10) holds because  $p_j$  has been subtracted on both sides of the inequality.  $\square$ 

The longest k-task can be feasibly scheduled because (9), (10) hold. The above reasoning can be applied inductively to the following k-tasks. Thus, tasks of size k can be feasibly scheduled provided inequalities (9), (10) hold.

After the assignment of k-tasks the processing capacities of the remaining processor system will be used by the tasks of the smaller sizes. The existence of a feasible schedule for these tasks depends on satisfying inequalities (9) and (10) for the new, smaller size tasks. Therefore, the values of the remaining processing capacities on the left-hand side of inequalities (9) and (10) should be as large as possible. In the next theorem we demonstrate that there is no rule which leaves more processing capacity than rule R. For the purposes of the

theorem let  $PC_R(i)$  denote the processing capacity remaining on  $P_i$  after assigning k-tasks according to our rule R, and  $PC_O(i)$  according to some alternative method O, while PC(i) is the processing capacity available for both rules before assigning any k-task.

**Theorem 3.** There is no rule O for assigning k-tasks such that the remaining processing capacities satisfy

$$\sum_{i=1}^{a} PC_{O}(ki) > \sum_{i=1}^{a} PC_{R}(ki)$$
 (11)

for at least one  $a \in \{1, ..., (m/k) - 1\}$ .

**Proof.** Let us suppose, on the contrary, that such a rule exists and it assigns task  $T_j$  to processors  $P_{qk+1}, \ldots, P_{qk+k}$ . Our rule assigns  $T_j$  to processors  $P_{k(w-1)+1}, \ldots, P_{kw}$  and  $P_{kw+1}, \ldots, P_{k(w+1)}$ .

 $P_{k(w-1)+1}, \ldots, P_{kw}$  and  $P_{kw+1}, \ldots, P_{k(w+1)}$ . If q < w then  $\sum_{i=1}^{a} \operatorname{PC}_{O}(ki) \leqslant \sum_{i=1}^{a} \operatorname{PC}_{R}(ki)$  for  $a = q, \ldots, w+1$ , and rule O is not better. If  $q \geqslant w+1$  then the schedule built by an alternative rule O is infeasible because  $p_j > \operatorname{PC}(k(w+1)) \geqslant \operatorname{PC}(kq)$ .

Let us suppose O assigns  $T_i$  to more than kprocessors,  $P_{k(q-1)+1}, \ldots, P_{kq}$  are the last processors used by  $T_i$  according to O, and  $q \ge w + 1$ . O may feasibly execute at most PC(k(w+1)) units work on  $P_{kw+1}, \ldots, P_{kq}$ . Thus,  $x \ge p_j - PC(k(w+1))$ units of work must be assigned to processors  $P_1, \ldots, P_{kw}$ . If rule O uses any of processors  $P_1, \ldots, P_{kw}$  $P_{k(w-1)}$  then rule R is better for  $a=1,\ldots,w-1$ because these processors are not used by our rule. For a = w rule O may not be better than R because amount x units of work assigned to  $P_1, \ldots, P_{kw}$  by *O* is at least equal to the amount  $p_j - PC(k(w+1))$ assigned by R. For  $a = w + 1, \dots, q - 1$  rule O is not better because any smaller (than by rule R) consumption of processing capacity on  $P_{k(q-1)+1}, \ldots, P_{kq}$  according to O must have been compensated for by the use of processing capacity on  $P_1, ..., P_{k(w-1)}$ .

Consequently, rule O is either infeasible, or inequality (11) is not satisfied. The induction over the tasks of size k completes the proof.  $\square$ 

The above theorem shows that rule R leaves the biggest possible processing capacity for the tasks of other sizes. Tasks of sizes smaller than k can

consume all the processing capacity remaining after scheduling  $\mathcal{T}^k$ , because k is a multiple of all the smaller sizes. Tasks with a bigger size may not access all the processing capacity remaining after the assignment of the k-tasks because only compact assignments are allowed. Therefore, our rule guarantees the existence of a feasible schedule for smaller size tasks, provided that a feasible schedule exists for the current schedule length. As a result, the tasks should be scheduled in order of descending sizes.

Let us calculate the time complexity of the above scheduling method. Finding a collection of processors with sufficient processing capacities requires  $O(\log m)$  time. The schedule for some task may have at most l+1 preemptions. Restoring the staircase pattern may require the change of the position of at most O(l) windows. Thus, all tasks can be scheduled in  $O(n(l + \log m))$  time.

#### 4.3. Dealing with infeasible schedules

Rule *R* builds a feasible schedule provided that such a schedule exists for the given schedule length. However, the method of calculating lower bound *C* is not sufficient to grasp the interactions between the tasks of different sizes. Consequently, schedule with length *C* may be infeasible. Consider an example.

**Example 2.** Let l = 1,  $s_1 = 0$ ,  $s_2 = \infty$ ,  $m_1 = 4$ . There are n = 3 tasks defined as follows:  $\text{size}_1 = \text{size}_2 = 2$ ,  $\text{size}_3 = 1$ ,  $p_1 = 2$ ,  $p_2 = 1$ ,  $p_3 = 2$ . From inequalities (7) and (8), we obtain  $C_2 = C_1 = C = 2$ . The schedule for 2-tasks is shown in Fig. 3a.

Processing capacities that remain in the system after scheduling the 2-tasks are PC(1) = PC(2) = 1. Inequalities (9) are not satisfied and  $T_3$  cannot be scheduled. There is enough processing capacity to accommodate  $T_3$ , yet the capacity is located on two processors in parallel and cannot be exploited by a 1-task. The deficiency of the processing capacity for  $T_3$  is  $p_3 - PC(1) = 1$ .

This deficiency must be compensated for by extending schedule length by  $\delta$ . Suppose the increase of the processing capacities of all processors can be accumulated to compensate for the defi-

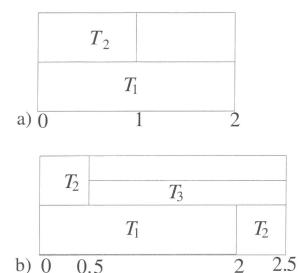


Fig. 3. Example 2. (a) Infeasible schedule, (b) optimal schedule.

ciency. Thus,  $m\delta=1$ , and C=2.25. However, the schedule with C=2.25 is infeasible, and we have to extend it again by 1/8. On the other hand, assume that only one processor produces the capacity which will compensate for the deficiency. Then  $\delta=1$ , and C=3, but a schedule with this length is not optimal. The optimal schedule is presented in Fig. 3b.

From Example 2 we conclude that there are cases where no feasible schedule with length C exists. The total processing capacity is at least equal to the total processing requirement of the tasks. However, not all this capacity can be used to schedule tasks of a certain size. To calculate the minimum extension making the schedule feasible we have to know the number of processors m' which will produce additional processing capacity usable for compensating the deficiency of processing capacity.

As soon as the tasks of sizes  $m, \ldots, 2k$  have been scheduled, the infeasibility of the schedule for k-tasks can be identified by verifying inequalities (9). By Theorem 2 a feasible schedule for tasks of size k exists if inequalities (9) hold. Hence, in the case of infeasibility, we have to increase schedule length by some amount  $\delta$  such that the above inequalities become valid. We calculate the deficiencies of

processing capacity in inequalities (9):  $DP(a,k) = \sum_{j=b}^{b+a} p_j - \sum_{i=1}^{a} PC(ki)$  for a = 1, ..., (m/k) - 1. Suppose inequality for the index a is violated in (9). Then, the extension is  $\delta = DP(a,k)/m'$ . If there are more violated inequalities, then we extend the schedule iteratively to satisfy the inequalities one by one. Note that inequality (10) is satisfied by the way of calculating C.

To calculate  $\delta$  we have to determine m'. For ktasks m' is in the set  $\{1, \ldots, m/k\}$ . If m' is too small, then the new deficiency of processing capacity is DP'(a,k) < 0, and the schedule is not optimal. If m' is too big then DP'(a, k) > 0 and, if perfect precision of the calculations is possible, the search for a feasible schedule never stops. When the number is right, then the violated inequality in (9) will be satisfied with equality. This means that the exactly required amount of processing capacity is created. Thus, a wrong number m' can be easily recognized. The value of m' results from the number of processors contributing to PC(ki). This number is a result of creating composite processors by rule R. Since tracing interactions between tasks and windows in rule R is hard we propose to use a binary search over integers in set  $\{1, \dots, m/k\}$ , to find m'. A faster, but more involved algorithm of determining m' has been proposed in [9].

To verify the correctness of this method observe the following facts: Any increase of processing capacity is created on the processors and the value of the increase accumulated on the left-hand side of inequalities (9) is a multiple of  $\delta$ . As schedule length increases, also the left-hand sizes of (9) increase because the capacity on the real processors increases, while tasks of size  $m, \ldots, 2k$  require the same amount of work as before the schedule extension.

Let us analyze the worst-case running time of the approach described above. Each extension made to accommodate a k-task requires rescheduling tasks of sizes  $m, \ldots, 2k$ , which can be done in time  $O(nl + n \log m)$  (see Section 4.2). For each violated inequality in the formulation (9),  $O(\log m)$  trials must be made to find m'. The number of violated inequalities can be calculated as a sum of geometric sequence. Hence, no more than 2m inequalities can be violated over all the task sizes. The extensions may require  $O(n(l + \log m)m \log m)$ 

time in total which is also the worst-case running time of the whole algorithm.

#### 4.4. The algorithm

We summarize the algorithm described above by presenting its pseudocode. In the following formulation we denote by A the set of the violated inequalities (9).

**Algorithm for** P, win | cube<sub>i</sub>,  $pmtn | C_{max}$ 

- 1:  $C := \max\{C_m, \dots, C_1\}$ ; (\*C is the lower bound on the makespan\*)
- 2: **for**  $f := \log_2 m$  **downto** 0 **do** (\*assignment of  $2^f$ -tasks\*)

begin

- 3: **if** inequalities (9) do not hold for  $a \in A$  **then** begin
- 4: **for each**  $a \in A$  **do begin** (\*loop over violated inequalities\*)

begin

- 5: calculate  $DP(a, 2^f)$ ;  $m_u := m/2^f$ ;  $m' := m_u/2$ ;  $m_l := 1$ ; flag := **true**;
- 6: **while**  $DP(a, 2^f) > 0$  **and** flag **do begin** (\*binary search for the number of processors  $m'^*$ )
- 7:  $\delta := DP(a, 2^f)/m'; C' := C + \delta;$
- 8: schedule tasks of sizes  $m, \ldots, 2^{f+1}$  in windows contained in the interval [0, C'] using rule R;
- 9: calculate new deficiency of processing capacity  $DP'(a, 2^f)$ ;
- 10: **if**  $DP'(a, 2^f) > 0$  **then**  $m_u := m'$  **else**
- 11: **if**  $DP'(a, 2^f) < 0$  **then**  $m_l := m'$  **else**
- 12: **begin** flag := **false**; C := C'; **end**;
- 13:  $m' := \lfloor \frac{m_u + m_l}{2} \rfloor;$

end; (\*of binary search for the number
of processors\*)

end; (\*of the search for the extension accommodating  $2^f$ -tasks\*)

end:

14: assign size  $2^f$  tasks using the scheduling rule R;

end;

#### 5. Conclusions

In this work we have considered scheduling multiprocessor tasks in the windows of processor availability. A polynomial time algorithm based on linear programming has been proposed for the general case of the problem and  $L_{\rm max}$  criterion when the number of processors is fixed. Low-order polynomial time algorithms were proposed for problems  $P, {\rm win}|{\rm size}_j \in \{1, m\}, pmtn, r_j|C_{\rm max}$  and  $P, {\rm win}|{\rm cube}_j, pmtn|C_{\rm max}$ .

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