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Abstract
In this paper, we consider the problem of allocating a large number of independent, equal-sized tasks to a heterogeneous processor farm. The master processor \( P_0 \) can process a task within \( w_0 \) time-units; it communicates a task in \( d_i \) time-units to the \( i \)-th slave \( P_i \), \( 1 \leq i \leq p \), which requires \( w_i \) time-units to process it. We assume communication-computation overlap capabilities for each slave (and for the master), but the communication medium is exclusive: the master can only communicate with a single slave at each time-step.

We give a polynomial-time algorithm to solve the following scheduling problem: given a time-bound \( T \), what is the maximal number of tasks that can be processed by the master and the \( p \) slaves within \( T \) time-units?

Keywords: heterogeneous processors, scheduling, mapping, master-slave, complexity

Résumé
Nous nous intéressons à l’ordonnancement de tâches indépendantes sur une plateforme maître-esclave hétérogène, modélisée par un graphe fork. Le maître et les \( p \) esclaves ont des vitesses de calcul différentes. De même, les liens de communication entre le maître et chaque esclave ont des capacités différentes. Le modèle autorise le recouvrement calcul-communication sur chaque processeur, mais on suppose que toutes les communications s’effectuent en mode 1-port: à tout instant, le maître ne peut envoyer qu’un seul message (à un esclave donné). Nous présentons un algorithme polynomial pour résoudre le problème d’ordonnancement correspondant: étant donnée une borne temporelle \( T \), maximiser le nombre total de tâches qui peuvent être effectuées par le maître et les \( p \) esclaves en temps \( T \).

Mots-clés: processeurs hétérogènes, heuristiques d’ordonnancement, distribution, maître-esclave, complexité
1 Introduction

In this paper, we deal with the problem of allocating a large number of independent, equal-sized tasks to a heterogeneous processor farm. The master processor $P_0$ can process a task within $w_0$ time-units: it communicates a task in $d_i$ time-units to its $i$-th slave $P_i$, $1 \leq i \leq p$, which requires $w_i$ time-units to process it. We assume communication-computation overlap capabilities for each slave (and for the master), but the communication medium is exclusive. In other words, the master can only communicate with a single slave at each time-step. We state the communication model more precisely:

- a given communication from $P_0$ to $P_i$ lasts $d_i$ units of time
- $P_0$ cannot be involved in any other communication during these units of time
- $P_i$ cannot start the execution of the task before the communication is completed.

We give a polynomial-time algorithm to solve the following scheduling problem (to be stated more formally in Section 2: given a time-bound $T$, what is the maximal number of tasks that can be processed by the master and the $p$ slaves within $T$ time-units? The polynomial complexity of this scheduling problem is established in Sections 3 and 4. This result is rather surprising: indeed, two variants of the problem are shown NP-complete in Section 5. Finally, we survey several related problems from the literature in Section 6.

2 Problem statement

We consider a fork-graph (see Figure 1) with a master-processor $P_0$ and $p$ slaves $P_i$, $1 \leq i \leq p$. The master processor $P_0$ owns an arbitrarily large number of independent, equal-size, non-preemptive tasks; it can process a given task within $w_0$ time-units; in parallel, it can send tasks to the $p$ slaves, but only to a single slave at a given time-step.

![fork-graph](image)

Figure 1: Heterogeneous fork-graph

It takes $d_i$ time-units for slave $P_i$ to receive a task, $1 \leq i \leq p$. More precisely, if $P_0$ sends a task to $P_i$ at time-step $t$, $P_i$ cannot start executing this task before time-step $t + d_i$, while $P_0$ cannot initiate another communication before time-step $t + d_i$. $P_i$ can process a task within $w_i$ time-units; while computing a given task, $P_i$ can receive another task from $P_0$. See Figure 2 for an example.

We deal with the following scheduling problem: given a time bound $T$, determine the scheduling and allocation of tasks to processor which maximize the total number of tasks whose execution is completed no later than $T$. Informally, we have to determine the best ordering of the messages sent by the master, together with the identity of the processors which execute the corresponding tasks, so as to maximize the total number of executed tasks. Note that this scheduling problem is called the NOW-Exploitation problem in [17].
Remark 1. Clearly, because the master processor can compute and communicate in parallel, it should be kept busy; within $T$ time-units, $P_0$ can execute $\lfloor \frac{T}{w_0} \rfloor$ tasks. To deal with the case where the master processor $P_0$ has no processing capability, simply let $w_0 = +\infty$. Similarly, to deal with the case where the master cannot overlap its own computations with the communications to the slaves, add a new child $P_{-1}$ such that $d_{-1} = w_0$ and $w_{-1} = w_0$, and then let $w_0 = +\infty$.

To state our complexity result, we have to be more precise about the problem specification. The expected output of the scheduling algorithm includes the list of tasks to be processed by each slave. The size of this list will be linear in the time-bound $T$, therefore exponential in the problem size if we only let $T$ in the input (because it can be encoded with $O(\log T)$ bits). Rather, we let the input include the original list of all tasks owned by the master. Such an hypothesis makes full sense in practice: even though we assume equal-size tasks, each task will correspond to a different file, and to a different computation, and we must keep track of which task is executed on which computing resource.

We are ready to formally state our scheduling problem (in view of the previous remarks, we omit the computations performed by the master processor):

Definition 1 (MAX-TASKS($n, \mathcal{F}, p, W, D, T$)). Given

- a set $\mathcal{F} = \{F_1, F_2, \ldots, F_n\}$ of $n$ equal-size independent tasks,
- a set $W = \{w_1, w_2, \ldots, w_p\}$ of $p$ execution times
- a set $D = \{d_1, d_2, \ldots, d_p\}$ of $p$ communication delays,
- a time-bound $T$,

compute a maximal-cardinal subset of $\mathcal{F}$ of tasks that can be executed within $T$ time-units on a heterogeneous master-slave platform made up with $p$ slaves $P_i$ of computation time $w_i$ and communication time $d_i$, $1 \leq i \leq p$.

Theorem 1. MAX-TASKS($n, \mathcal{F}, p, W, D, T$) can be solved in polynomial time.

The proof is constructive. In the following, a (somewhat sophisticated) greedy algorithm is shown to be optimal.
3 Reduction to a problem with at most one task per slave

**Proposition 1.** Any instance MAX-TASKS\((n, F, p, W, D, T)\) can be polynomially reduced to an instance MAX-TASKS-SINGLE\((n, F, p', W', D', T)\) where each slave is constrained to execute at most one task.

*Proof.* Each slave \(P_i\) can be dealt with separately, so let \(i, 1 \leq i \leq p\), be fixed. We replace slave \(P_i\) by a collection of \(k\) slaves \(P_{i,j}\), \(1 \leq j \leq k = \min(n, \lceil \frac{T - w_i}{M_i} \rceil + 1)\), where \(M_i = \max(d_i, w_i)\). See Figure 3: all the new slaves have the same communication capability \(d_i\) as \(P_i\), but \(P_{i,j}\) processes a task in \(w_{i,j} = w_i + (j - 1)M_i\) time-units. Obviously, the \(j + 1\)-th task executed by \(P_i\) in the original instance will be executed by \(P_{i,j}\) in the new instance.

After the transformation, the total number of slaves is bounded by \(p \times n\), which is indeed polynomial in the size of the input (because \(F\) is of size at least \(O(n)\) and \(W\), or \(D\), of size at least \(O(p)\)).

![Figure 3: Replacing a slave by a collection of uni-task slaves](image)

4 With at most one task per slave

In this section, we solve MAX-TASKS-SINGLE\((n, F, p, W, D, T)\), the problem where each slave is limited to executing at most one-task. The problem is reduced to deciding which slave to communicate with, and in which order. For simplicity, we still use the notations \(p, W, D\) while in fact there are \(p' = O(pm)\) slaves after the transformation of Section 3.

We start with a technical lemma:

**Lemma 1.** Consider a set of \(k\) slaves capable of executing \(k\) tasks (one each) within \(T\) units with some scheduling. Then, if we sort these slaves by computation times (largest first), and if we schedule the communications from the master \(P_0\) in that order, we can also execute \(k\) tasks within \(T\) time-units.

*Proof.* To simplify notations, let \(P_1, P_2, \ldots, P_k\) be the sorted slaves: \(w_1 \geq w_2 \geq \ldots \geq w_k\). Let \((P_{i_1}, \ldots, P_{i_k})\) be the ordering of the communications from \(P_0\) in the original schedule. The following set of equations is satisfied:

\[
\begin{align*}
(L_1) & \quad d_{i_1} + w_{i_1} \quad \leq T \quad \text{first task} \\
(L_2) & \quad d_{i_1} + d_{i_2} + w_{i_2} \quad \leq T \quad \text{second task} \\
& \vdots \quad \quad \quad \quad \quad \quad \quad \quad \vdots \quad \vdots \\
(L_k) & \quad d_{i_1} + d_{i_2} + \cdots + d_{i_k} + w_{i_k} \quad \leq T \quad \text{last task}
\end{align*}
\]
Let $j_1, \ldots, j_k$ be the indices of the tasks of the sorted slaves in the original schedule: $i_{j_1} = 1$, $i_{j_2} = 2, \ldots$ and $i_{j_k} = k$. $i_{j_1} = 1$ means that processor $P_1$ is assigned the $j_1$-th task in the original schedule. We prove by induction that tasks can be assigned in the ordering $1, 2, \ldots, k$ of sorted computation times:

i) For the first task, we have $d_1 + w_1 \leq T$. Indeed, see equation $L_{j_1}$: $d_{i_1} + \cdots + d_{i_{j_1-1}} + d_1 + w_1 \leq T$.

ii) For the second task, we have $d_1 + d_2 + w_2 \leq T$. Indeed, there are two possibilities:

- Either $j_2 > j_1$; then the inequality $L_{j_2}$ contains the quantity $d_{i_{j_1}} + d_{i_{j_2}} + w_{i_{j_2}}$, hence $d_1 + d_2 + w_2 \leq T$.
- Or $j_2 < j_1$; then the inequality $L_{j_1}$ contains the quantity $d_{i_{j_1}} + d_{i_{j_2}} + w_{i_{j_1}}$; because $w_2 \leq w_1$, we derive that $d_1 + d_2 + w_2 \leq T$.

iii) For the general case, let $i < k$. Assume that for all $l \leq i$ we have $d_1 + \cdots + d_l + w_l \leq T$. We aim at proving that $d_1 + \cdots + d_{i+1} + w_{i+1} \leq T$. Let $m_i = \max(j_1, \ldots, j_i)$.

- Either $j_{i+1} > m_i$; then we consider the inequality $L_{j_{i+1}}$, which contains the terms $d_1, d_2, \ldots, d_i$ (because $j_{i+1} > m_i$) as well as the term $d_{i+1} + w_{i+1}$. Hence $d_1 + \cdots + d_{i+1} + w_{i+1} \leq T$.
- Or $j_{i+1} < m_i$; then we consider the inequality $L_{m_i}$, which contains the terms $d_1, d_2, \ldots, d_{i+1}$ as well as the term $d_{i+1} + w_i$. Hence $d_1 + \cdots + d_{i+1} + w_{i+1} \leq T$. Because $w_i \geq w_{i+1}$, we finally derive that $d_1 + \cdots + d_{i+1} + w_{i+1} \leq T$.

We introduce the following greedy algorithm:

```
MASTER-SLAVE((P_1, \ldots, P_p), T)
1: Sort slaves such that $d_1 \leq d_2 \ldots \leq d_p$
2: $L \leftarrow \emptyset$
3: For $i = 1$ To $p$ Do
4: If $L \cup \{P_i\}$ can be scheduled within $T$ time-units
5: Then $L \leftarrow L \cup \{P_i\}$
6: Return ($L$)
```

The condition at line 4 can be checked using Lemma 1: sort the current list $L$ according to the computation times and verify that all inequalities are satisfied.

**Proposition 2.** The Master-Slave algorithm returns a maximal set of tasks that can be processed within $T$ time-units.

**Proof.** To simplify notations, we say that a set of indices $I = \{i_1, \ldots, i_k\}$ is schedulable if there exists a valid scheduling executing $k$ tasks within $T$ time-units, with processors $\{P_{i_1}, \ldots, P_{i_k}\}$.

Assume that the set $I = \{i_1, \ldots, i_k\}$ is schedulable, and further (without loss of generality) that $d_{i_1} \leq d_{i_2} \leq d_{i_k}$. We show that the MASTER-SLAVE algorithm returns at least $k$ values, i.e. a schedulable set $\mathcal{G} = \{g_1, \ldots, g_k\}$, which will conclude the proof. We proceed by induction. We proceed by induction.

i) Let $I_0 = \{i_1, \ldots, i_k\}$. The set $I_0$ is schedulable. Each subset of $I_0$ is schedulable. In particular, each singleton $\{i_j\}$ is schedulable. By construction of the MASTER-SLAVE algorithm, $g_1$ exists and $g_1 \leq \min(i_1, \ldots, i_k)$. This is because the MASTER-SLAVE algorithm selects the
smallest index of a schedulable task. Now we show that there exists a schedulable set \( \mathcal{I}_1 \) of \( k \) elements which contains \( g_1 \). If \( g_1 \in \mathcal{I}_0 \) we are done; \( \mathcal{I}_1 = \mathcal{I}_0 \). Otherwise consider the ordering \( i_{j_1}, i_{j_2}, \ldots, i_{j_k} \) in which \( \mathcal{I}_0 \) is scheduled: we have the set of inequalities:

\[
\begin{align*}
(L_1) \quad d_{i_{j_1}} + w_{i_{j_1}} & \leq T \\
(L_2) \quad d_{i_{j_1}} + d_{i_{j_2}} + w_{i_{j_2}} & \leq T \\
& \quad \vdots \\
(L_k) \quad d_{i_{j_1}} + d_{i_{j_2}} + \ldots + d_{i_{j_k}} + w_{i_{j_k}} & \leq T
\end{align*}
\]

We replace the first scheduled index \( i_{j_1} \) by \( g_1 \) to build \( I_1 \): we let \( I_1 = I_0 \setminus \{ i_{j_1} \} \cup \{ g_1 \} \). The first inequality \( c_{g_1} + w_{g_1} \leq T \) holds because \( g_1 \) is schedulable. The other inequalities \((L_2)\) to \((L_k)\) still hold because \( d_{g_1} \leq d_{i_{j_1}} \). Hence \( \mathcal{I}_1 = \{ g_1, i_{j_2}, \ldots, i_{j_k} \} \) is schedulable.

ii) Therefore there exists a schedulable set \( \mathcal{I}_1 = \{ g_1, i'_1, \ldots, i'_{k-1} \} \). For all \( j \), \( 1 \leq j \leq k \), the pair \( \{ g_1, i'_j \} \) is schedulable, which establishes that the MASTER-SLAVE algorithm does select an index \( g_2 \) after selecting \( g_1 \). Furthermore, \( g_2 \leq \min(\{ i'_1, \ldots, i'_{k-1} \} \). Now we show that there exists a schedulable set \( I_2 \) of \( k \) elements which contains \( g_1 \) and \( g_2 \). If \( g_2 \in \{ i'_1, \ldots, i'_{k-1} \} \) let \( I_2 = I_1 \). Otherwise, consider the ordering in which \( \mathcal{I}_2 \) is scheduled. There is no reason that \( g_1 \) should be scheduled first in \( I_1 \), hence we have two cases:

- \( g_1 \) is indeed the first task in \( \mathcal{I}_1 \) to be communicated by \( P_0 \). Let \( f \) be the index of the second task in \( \mathcal{I}_1 \) to be communicated by \( P_0 \). Because the pair \( \{ g_1, g_2 \} \) is schedulable, we can replace \( f \) by \( g_2 \): the first two inequalities \((L_1)\) and \((L_2)\) are satisfied (maybe we have to swap \( g_1 \) and \( g_2 \), but anyway this pair can be scheduled), and the following \( k - 2 \) inequalities still hold true because \( d_{g_2} \leq d_{f} \).

- The index \( f \) of the first task in \( \mathcal{I}_1 \) to be communicated by \( P_0 \) is different from \( g_1 \). Then we replace \( f \) by \( g_2 \): the first inequality \((L_1)\) is satisfied because \( \{ g_2 \} \) is schedulable, and the following \( k - 1 \) inequalities still hold true because \( d_{g_2} \leq d_{f} \).

We conclude that \( \mathcal{I}_2 \) is schedulable.

iii) For the general case, assume we have a schedulable set \( I_{j-1} = \{ g_1, \ldots, g_{j-1}, i'_1, \ldots, i'_{k-j+1} \} \). For every \( l, 1 \leq k - j + 1 \), the subset \( \{ g_1, \ldots, g_{j-1}, i'_l \} \) is schedulable, hence the the MASTER-SLAVE algorithm does select an index \( g_j \) after selecting \( g_1, \ldots, g_{j-1} \). Furthermore, \( g_j \leq \min(\{ i'_1, \ldots, i'_{k-j+1} \} \). Now we show that there exists a schedulable set \( I_j \) of \( k \) elements which contains \( \{ g_1, g_2, \ldots, g_j \} \). If \( g_j \in \{ i'_1, \ldots, i'_{k-j+1} \} \) let \( I_j = I_{j-1} \). Otherwise, consider the ordering in which \( I_{j-1} \) is scheduled. We pick up the first task \( f \) to be scheduled that is different from the MASTER-SLAVE indices \( g_1, g_2, \ldots, g_{j-1} \). We replace this index \( f \) by \( g_j \) to derive \( I'_j \). We claim that \( I_j \) is schedulable. To see this, consider the scheduling of \( I_{j-1} \), and let \( l \) be such that the first \( l \) inequalities \((L_1)\) to \((L_l)\) deal with MASTER-SLAVE indices \( g_i, g_{i+1}, \ldots, g_{j-1} \), while \((L_{l+1})\) deals with index \( f \):

(a) \( \{ g_1, \ldots, g_{j}, g_{j} \} \) is schedulable as a subset of \( \{ g_1, \ldots, g_{j} \} \): a valid ordering of these \( l + 1 \) tasks provide the first \( l + 1 \) inequalities.

(b) inequalities \((L_{l+2})\) to \((L_k)\) still hold because \( d_{g_j} \leq d_{f} \).

This establishes Proposition 2.
Proof of Theorem 1. We go back to the proof of Theorem 1. First we note that the complexity of the Master-Slave algorithm is quadratic in the number of slaves: each task is processed once, and the test for its insertion in the returned schedule is easily implemented in time linear in the number of already selected tasks. To see this, simply maintain the list of already selected tasks together with their current termination time. When a new task $F_i$ is inserted, we increment by $d_i$ the termination times of all tasks $F_j$ whose $w_j$ is smaller than $w_i$, and we check on the fly that these termination times still do not exceed $T$.

Going back to the original problem, the complexity is thus bounded by $O(n^2 p^2)$: after the transformation of Section 3, the number of uni-task slaves is bounded by $O(n, p)$, and the complexity of the greedy algorithm is at most quadratic in that quantity. This concludes the proof of Theorem 1.

5 Extensions

In this section we discuss two variants which are natural extensions of the MAX-TASKS($n, F, p, W, D, T$) problem. The first variant deals with a two-port communication model, where the master can communicate to at most two slaves simultaneously. The second variant deals with two masters sharing slaves. Both variants turn out to be NP-complete.

5.1 With a two-port communication model

We state the decision problem associated with the extension to a two-port communication model as follows:

Definition 2 (MAX-TASKS-2PORTS($n, F, p, W, D, T, K)$). Given a set $F = \{F_1, F_2, \ldots, F_n\}$ of $n$ equal-size independent tasks, a set $W = \{w_1, w_2, \ldots, w_p\}$ of $p$ execution times, a set $D = \{d_1, d_2, \ldots, d_p\}$ of $p$ communication delays, a time-bound $T$, and a task-bound $K$, is it possible to execute at least $K$ tasks within $T$ time-units on a heterogeneous master-slave platform made up with $p$ slaves $P_i$ of computation time $w_i$ and communication time $d_i$, $1 \leq i \leq p$, and a master capable of communicating to at most two slaves simultaneously?

Proposition 3. MAX-TASKS-2PORTS($n, F, p, W, D, T, K$) is NP-complete.

Proof. MAX-TASKS-2PORTS($n, F, p, W, D, T, K$) clearly belongs to the class NP. We use a reduction from 2-PARTITION [8] to show its completeness. We are given an instance $I_1$ of 2-PARTITION, i.e. a set of $n$ integers $A = \{a_1, \ldots, a_n\}$, and we ask whether there exists a partition of $\{1, \ldots, n\}$ into two subsets $A_1$ and $A_2$ such that

$$\sum_{i \in A_1} a_i = \sum_{i \in A_2} a_i$$

Let $S = \frac{1}{2} \sum_{i=1}^{n} a_i$.

Given $I_1$, we build an instance $I_2$ of MAX-TASKS-2PORTS as follows. We use a set $F$ of $n$ tasks and $p = n$ slaves with $w_i = 2S$ and $d_i = a_i$ for $1 \leq i \leq n$. We let $T = 3S$ and $K = n$. The size of $I_2$ is polynomial (even linear) in the size of $I_1$.

Assume that $I_1$ has a solution $A_1$ and $A_2$. Let the master communicate along its first communication channel with the slaves whose index is in $A_1$, and along its second communication channel with the slaves whose index is in $A_2$. The ordering of the communications is not important. At time $S$, the last two processors have completed the reception of their message from the master. At time $3S$, each processor has executed one task, hence a solution to $I_2$. 
Assume now that $I_2$ has a solution. Let $A_1$ denote the set of indices of those processors which communicate with the master along its first communication channel, and $A_2$ denote the set of indices of those processors which communicate with the master along its second communication channel. Each processor can execute at most one task, because $2w_i = 4S > T$ for all $i$. Hence, because $K = n$, each processor executes exactly one task. Therefore the sets $A_1$ and $A_2$ represent a partition of $\{1, \ldots, n\}$.

The last processor to receive a message along the first communication channel does complete the execution of its task within $T$ time units, which translates into

$$\sum_{i \in A_1} d_i + 2S \leq T$$

Similarly, $\sum_{i \in A_2} d_i + 2S \leq T$. Since $\sum_{i=1}^n d_i = 2S$, we get

$$\sum_{i \in A_1} d_i = \sum_{i \in A_1} d_2 = S,$$

hence a solution to $I_1$.

5.2 With two masters

We state the decision problem associated with the extension for two masters (see Figure 4) as follows:

![Figure 4: Variant with two masters.]

**Definition 3** (MAX-TASKS-2MASTERS($n, F_1, F_2, p, W, D_1, D_2, T, K$)). Given a master-slave heterogeneous computing platform made up with

- a first master holding a set $F_1 = \{F_1, F_2, \ldots, F_n\}$ of $n$ equal-size independent tasks,
- a second master possessing a set $F_2 = \{F_1', F_2', \ldots, F_n'\}$ of $n$ equal-size independent tasks,
- $p$ slaves,
- a set $W = \{w_i, 1 \leq i \leq p\}$ of execution times of the $p$ slaves,
- a set $D_1 = \{d_{i_1}, 1 \leq i \leq p\}$ of communication delays from the first master to the $p$ slaves,
- a set $D_2 = \{d_{i_2}, 1 \leq i \leq p\}$ of communication delays from the second master to the $p$ slaves,

given a time-bound $T$, and a task-bound $K$, is it possible to execute at least $K$ tasks within $T$ time-units on the platform, assuming that each processor is capable of sending to or receiving from at most another processor at any time-step?
Proposition 4. **MAX-TASKS-2MASTERS**(*n, F_1, F_2, p, W, D_1, D_2, T, K*) is **NP-complete**.

**Proof.** MAX-TASKS-2MASTERS clearly belongs to NP. Again, we use a reduction from 2-PARTITION to show its completeness. We are given an instance *I_1* of 2-PARTITION, i.e. a set of *n* integers \( A = \{a_1, \ldots, a_n\} \), and we ask whether there exists a partition of \( \{1, \ldots, n\} \) into two subsets \( A_1 \) and \( A_2 \) such that
\[
\sum_{i \in A_1} a_i = \sum_{i \in A_2} a_i ?
\]

Let \( S = \frac{1}{2} \sum_{i=1}^{n} a_i \).

Given \( I_1 \), we build an instance \( I_2 \) of MAX-TASKS-2MASTERS as follows. We use \( p = n \) slaves with \( w_i = 2S \) for all \( 1 \leq i \leq n \). We let \( d_{i_1} = d_{i_2} = a_i \) for \( 1 \leq i \leq n \). We let \( T = 3S \) and \( K = n \). The size of \( I_2 \) is polynomial (even linear) in the size of \( I_1 \).

The proof that \( I_1 \) has a solution if and only if \( I_2 \) has a solution is quite similar to that of Proposition 3.

6 Related problems

This paper is a follow-on of two previous papers that study the same scheduling problem. In [3], communication links are assumed to be homogeneous. In other words, \( d_i = d \) for \( 1 \leq i \leq p \). We use a matching algorithm to compute the optimal solution in this context. In [2] we deal with master-slave tasking on more general tree-structured networks: in other words, the underlying architecture is a tree, it is no longer restricted to a fork graph. We do not know the complexity of the master-slave scheduling problem on a general tree graph, which seems to be a challenging open problem. Rather than searching a solution to this problem, we characterize in [2] the best steady-state for various operation models (i.e. we neglect the initialization and termination phases).

We classify several related papers along the following three main lines:

**Scheduling fork graphs** The complexity of scheduling fork graphs has been widely studied in the literature. Given a task graph, the standard macro data-flow model [5, 16] assumes that: (i) processors are homogeneous; (ii) communication delays are paid each time a task and one of its successors are not assigned to the same processor. In this model, scheduling a fork graph with an infinite number of processors is a polynomial problem [9], while scheduling a fork-join graph is NP-complete [6].

Extensions of the standard macro data-flow model includes the one-port model [13, 14] (just as in this paper, each processor can communicate with at most another processor at a given time-step) and the LogP model [7]. Scheduling fork graphs in the one-port model [1] or in the LogP model [20] remains NP-complete, even for an infinity of resources. Extensions of the result under the LogP model to allow message forwarding are dealt in [21].

**Collective communications on heterogeneous platforms** Several papers deal with the complexity of collective communications on heterogeneous platforms: broadcast and multicast operations are addressed in [4, 15], gather operations are studied in [11].

**Master-slave on the computational grid** Master-slave scheduling on the grid can be based on a network-flow approach [19, 18] or on an adaptive strategy [12]. Enabling frameworks to facilitate the implementation of master-slave tasking are described in [10, 22].
7 Conclusion

In this paper, we have shown that master-slave tasking using heterogeneous fork-graphs can be solved in polynomial time, using a non-trivial greedy algorithm. The complexity of this scheduling problem for arbitrary tree graphs is a challenging open problem. A first step to solving this problem would be to study the complexity for a linear chain of processors.

References


