
This document has been downloaded from Chalmers Publication Library (CPL). It is the author’s version of a work that was accepted for publication in:

Citation for the published paper:

Downloaded from: http://publications.lib.chalmers.se/publication/198974

Notice: Changes introduced as a result of publishing processes such as copy-editing and formatting may not be reflected in this document. For a definitive version of this work, please refer to the published source. Please note that access to the published version might require a subscription.

Chalmers Publication Library (CPL) offers the possibility of retrieving research publications produced at Chalmers University of Technology. It covers all types of publications: articles, dissertations, licentiate theses, masters theses, conference papers, reports etc. Since 2006 it is the official tool for Chalmers official publication statistics. To ensure that Chalmers research results are disseminated as widely as possible, an Open Access Policy has been adopted. The CPL service is administrated and maintained by Chalmers Library.

(article starts on next page)

Zhennan Fei, Spyros Reveliotis, Sajed Miremadi and Knut Åkesson

Abstract

This electronic document provides some supportive material to the paper entitled “A BDD-Based Approach for Designing Maximally Permissive Deadlock Avoidance Policies for Complex Resource Allocation Systems” that has been submitted to IEEE Transactions on Automation Science and Engineering (T-ASE).

I. MODELING A GIVEN RAS INSTANCE $\Phi$ BY THE CORRESPONDING EFA $E(\Phi)$

This section provides a formal statement of the procedure for obtaining, for any RAS instance $\Phi$ coming from the RAS class that is considered in the aforementioned paper, the corresponding EFA $E(\Phi)$. This procedure is called DevEFA(RAS $\Phi$), and it is detailed as follows:
**Procedure** DevEFA(RAS $\Phi$)

**Input:** A RAS instance $\Phi = \langle R, C, P, A \rangle$

**Output:** An EFA $E(\Phi)$ that models the (resource allocation) dynamics of $\Phi$

1 /* Define the resource variables based on $R$ and $C$ */

   foreach $R_i$ in $R = \{ R_1, \ldots, R_m \}$ do
   | define $vR_i : \{ 0, \ldots, C_i \}$ where $C_i = C(R_i)$
   | mark the value $C_i$ as the initial and marked values of $vR_i$
   end

2 /* Construct the EFAs modeling the sequential logic of each processing type in $P$ */

   foreach $J_j$ in $P = \{ J_1, \ldots, J_n \}$ do
   | build an EFA $E_j$ with only one location named $\ell_j$
   | mark $\ell_j$ as the initial and marked location of $E_j$
   | foreach $\Xi_{jk}$ where $k \leq l(j)$ in $S_j$ do
   |   if $\Xi_{jk}$ is a non-terminal processing stage then
   |     define $v_{jk} : \{ 0, \ldots, \theta_{jk} \}$ where:
   |     $\theta_{jk} = \min \{ \frac{C_i}{A_{jk}[i]} : A_{jk}[i] > 0 \}$
   |     mark the value 0 as the initial and the marked values of $v_{jk}$
   |   end
   | endforeach
   | foreach stage $\Xi_{jk}$ corresponding to a source node in $G_j$ do
   |   build a self-loop transition $tr$ labeled by the event $\langle J_{\text{loading}}, \Xi_{jk} \rangle$ and add it to $E_j$
   |   attach the action $v_{jk} := v_{jk} + 1$ to $tr$
   | endforeach
   | foreach $e = \langle \Xi_{jk}, \Xi_{jk'} \rangle$ in $G_j$ do
   |   build a self-loop transition $tr$ labeled by the event $\langle \Xi_{jk}, \Xi_{jk'} \rangle$ and add it to $E_j$
   |   attach the guard $v_{jk} \geq 1$ to $tr$
   |   if $\Xi_{jk'}$ is a non-terminal processing stage of $S_j$ then
   |     attach the action $v_{jk} := v_{jk} - 1; v_{jk'} := v_{jk'} + 1$ to $tr$
   |   else attach the action $v_{jk} := v_{jk} - 1$ to $tr$
   | // no instance variable defined for a terminal stage
   | endforeach
   | endforeach
   | foreach transition $tr$ of $E_j$ do
   |   let $\sigma$ denote the labeling event of $tr$
   |   if $\sigma$ is $\langle J_{\text{loading}}, \Xi_{jk} \rangle$ then
   |     allocate the associated resources required at $\Xi_{jk}$
   |     append $vR_i \geq A_{jk}[i]$ to the guard of $tr$ for all $i$ s.t. $A_{jk}[i] > 0$
   |     append $vR_i := vR_i - A_{jk}[i]$ to the action of $tr$ for all $i$ s.t. $A_{jk}[i] > 0$
   |   end
   | else
   |   if $A_{jk}[i] > A_{jk'}[i]$ then
   |     append $vR_i \geq (A_{jk'}[i] - A_{jk}[i])$ to the guard of $tr$
   |   end
   | if $\Xi_{jk'}$ is a terminal processing stage then
   |   just deallocate the resources used at $\Xi_{jk}$
   |   append $vR_i := vR_i + A_{jk}[i]$ to the action of $tr$ for all $i$ s.t. $A_{jk}[i] > 0$
   | end
   | if $\Xi_{jk'}$ is not a terminal processing stage then
   |   append $vR_i := vR_i - (A_{jk'}[i] - A_{jk}[i])$ to the action of $tr$ for all $i$ s.t. $A_{jk'}[i] - A_{jk}[i] \neq 0$
   | end
   | end
end
end
II. A COMPLETE CORRECTNESS ANALYSIS FOR ALGORITHM 2

To prove the effectiveness of Algorithm 2 w.r.t. the penultimate objective of the implementation of the maximally permissive DAP through the one-step-lookahead scheme that was outlined in the earlier parts of this manuscript, we need to show that (i) the algorithm terminates in a finite number of steps, (ii) the returned set $\chi_{FB}$ contains all the feasible boundary unsafe states, and furthermore, (iii) $\chi_{FB}$ does not contain any feasible safe state. The finiteness of Algorithm 2 depends on whether the backward search performed in Lines 6-15 can terminate in a finite number of iterations. We notice that the termination of this search is determined by the set of the new unsafe states, $\chi_{U_{new}}$, computed at each iteration; if $\chi_{U_{new}}$ is empty, the backward search terminates. We also notice that the set $\chi_{U_{new}}$ will finally be empty during the search, since the set of states in $\Delta_A$ is finite. Hence, Algorithm 2 terminates in a finite number of steps. In the following, we focus on establishing the correctness of the algorithm, by establishing items (ii) and (iii) in the above list. In the paper, these two items are addressed by Theorem IV.1. However, that manuscript provides only a sketch of the corresponding proof. Here is provide a complete treatment of the relevant results.

We begin with some lemmas that are necessary for the derivation of the final result.

**Lemma II.1.** The characteristic function $\chi_{FD}$ that is obtained from the symbolic operations performed in Lines 1-4 of Algorithm 2 identifies correctly the feasible deadlock states in the transition set $\Delta_E$ w.r.t. the process-advancing events of this set.

**Proof.** First we notice that, by its construction, the set $\chi_D$ contains only states that are deadlocks w.r.t. the process-advancing events of $\Delta_E$, and therefore, the same is true for its subset $\chi_{FD}$. Furthermore, the states contained in $\chi_{FD}$ are feasible deadlocks since $\chi_{FD}$ is obtained from the filtration of $\chi_D$ with the characteristic function $\chi_F$.

Next we show that $\chi_{FD}$ contains all the feasible deadlock states w.r.t. the process-advancing events of $\Delta_E$. This part can be proved by contradiction. Let us assume that there exists a transition $(s, s')$ in $\Delta_E$ where $s'$ is a feasible deadlock state that is not identified by the computation that is performed in Lines 1-4, i.e., $s' \notin FD$. Since $s'$ is the target state of $(s, s')$, which is a transition in $\Delta_E = \Delta_A \cup \Delta_L$, state $s'$ must be in the set $T$. By the working assumption, state $s' \notin D$ after the computation at Line 3, since in that case it would have been retained in $FD$ (being a feasible state). Knowing that $s' \notin D$, $s' \in T$ and $s' \neq s_0$, state $s'$ can only be in the state set $E$. But this contradicts the definition of deadlock states, since states in $E$ enable process-advancing events. \hfill $\square$

**Lemma II.2.** For every transition $(s, s')$ of the EFA $\Delta_E$, feasibility of the target state $s'$ implies also the feasibility of the source state $s$.

**Proof.** We prove the contrapositive of the above statement, i.e., every transition $(s, s')$ of the EFA $\Delta_E$ with an infeasible source state $s$ has also an infeasible target state $s'$. Infeasibility of state $s$ implies that there exists some resource $R_i$ with

$$vR_i + \sum_{j=1}^{n} \sum_{k=1}^{l(j)-1} A_{jk[i]} * v_{jk} = d \neq C_i,$$

for the values of the variables $vR_i$ and $v_{jk}$, $j = 1, \ldots, n$, $k = 1, \ldots, l(j)-1$ that define state $s$. But it can be easily checked that every forward-advancing transition from state $s$ preserves the invariant

$$vR_i + \sum_{j=1}^{n} \sum_{k=1}^{l(j)-1} A_{jk[i]} * v_{jk} = d,$$

and therefore, it cannot restore feasibility w.r.t. to the implied allocation of resource $R_i$. \hfill $\square$

**Lemma II.3.** All states entering the sets $U$ and $\chi_{FB}$ during the execution of Algorithm 2 are feasible.

**Proof.** This lemma is an immediate implication of Lemmas II.1 and II.2, and of the fact that all the elements of these two sets are obtained by starting from some feasible deadlock state in $\chi_{FD}$ and backtracing upon some transitions in $\Delta_E$. \hfill $\square$
Lemma II.4. The set $U$ that is computed by Algorithm 2 contains all the feasible unsafe states in $\Delta_E$.

Proof. By Lemma II.1 and Line 5 of Algorithm 2, $U$ contains all the feasible deadlocks. Next we will show that $U$ also contains all the feasible deadlock-free unsafe states of the considered RAS.

Let us consider any such feasible deadlock-free unsafe state $\hat{u}$. The finite and acyclic nature of the paths that define the execution logic of the various process types in the considered RAS class, imply that the subspace that is reached from state $\hat{u}$ following only transitions in $\Delta_A$ has a finite, acyclic structure. This remark, when combined with the presumed unsafety of state $\hat{u}$, further implies that every path in $\Delta_A$ that emanates from state $\hat{u}$ is an acyclic path that terminates at some feasible deadlock state. Let $\zeta$ denote the longest length of these paths, where the length of a path is defined by the number of the involved transitions. Next, we will show, by induction on $\zeta$, that state $\hat{u}$ will enter the state set $U$ that is maintained by Algorithm 2 before the termination of the iteration in Lines 6-15.

First we consider the base case of $\zeta = 1$. Then, all the transitions emanating from $\hat{u}$ lead to a feasible deadlock state in $\chi_{FD}$, and therefore, they will be contained in the set $\Delta_U$ computed during the first iteration of the algorithm. Hence, state $\hat{u}$ will not be in the set $\chi_{NU}$ that is computed at that iteration, and therefore, it will be correctly included in the set $\chi_{cur}$ and, eventually, in $\chi_U$.

Next, let us suppose that all the feasible unsafe states with a maximal path of length $\zeta - 1$ from the feasible deadlock states of $\chi_{FD}$ will be correctly identified and entered in set $U$ by Algorithm 2. Since the target state of each process-advancing transition that emanates from state $\hat{u}$ has a maximal path leading to $\chi_{FD}$ of length less than or equal to $\zeta - 1$, each of these states will be eventually identified by the algorithm. Let us consider, in particular, the iteration where the last of these states, let’s say $u_l$, enters $U$. In the next iteration, $u_l$ will be in $\chi_{U_{new}}$ and, therefore, $\hat{u}$ will be in $\chi_{SU}$. Furthermore, $\hat{u}$ will not be in $\chi_{NU}$ since all of its emanated transitions will be either in $\Delta_U$ or in $\Delta_{U_{pre}}$. Hence, $\hat{u}$ will be included in $\chi_{U_{cur}}$ and eventually into $\chi_U$.

Lemma II.5. The set $U$ that is computed by Algorithm 2 contains no feasible safe states of $\Delta_E$.

Proof. We prove this result by induction on the number of iterations performed by the algorithm. The base case of zero iteration is covered by Lemma II.1. Next, suppose that the statement of Lemma II.5 is true for the set $U$ constructed during the first $n$ iterations. This assumption when combined with Lines 7 and 14 of Algorithm 2, further imply that the transition set $\Delta_{U_{pre}}$ contains only transitions with target states in the already constructed state set $U$. But then, Lines 7-10 of the algorithm implies that the state set $\chi_{NU}$ that is constructed at iteration $n + 1$ contains all the safe states that can be reached from the current set $\chi_{U_{new}}$ by backtracing on some transition of $\Delta_A$. Hence, the set $\chi_{cur}$ constructed at the $n + 1$ iteration, at Line 11, contains no safe states, and the addition of this set to state set $U$, at Line 13, does not introduce in $U$ any safe states either.

A complete proof of Theorem IV.1

Proof. Property 1 was established in Lemma II.3.

In view of Lemma II.4, to prove Property 2 we just need to show that the construction of the set $FB$ in Lines 16-18 of Algorithm 2 retains all the boundary feasible unsafe states in $U$. But this can be easily checked from the facts that (a) the transition set $\Delta_B$ contains all the transitions with target states in set $U$, while (b) the transition set $\Delta_{SB}$ is obtained from $\Delta_B$ by removing only transitions with source states in $U$ (and therefore, unsafe, according to Lemma II.5).

Property 3 is also inferred from the construction of the set $\Delta_{SB}$ from the set $\Delta_B$ through the removal of all those transitions with source states in $U$, upon noticing that $U$ contains all the feasible unsafe states of $\Delta_E$ (according to Lemma II.4).

Finally, Property 4 results from Lemma II.5 and the fact that all the transitions in the set $\Delta_E$ have target states in $U$.  —
Algorithm 3: Symbolic computation of the minimal boundary unsafe states

Input: $\chi_{FB}$, \{\(\Delta_=(\tilde{v}_1,v_1),\ldots,\Delta_=(\tilde{v}_K,v_K)\) and \{\(\Delta_{\geq}(\tilde{v}_1,v_1),\ldots,\Delta_{\geq}(\tilde{v}_K,v_K)\}\)

Output: $\chi_{FB}$

1. \(\Delta_{EQ} := \Delta_=(\tilde{v}_1,v_1) \wedge \ldots \wedge \Delta_=(\tilde{v}_K,v_K)\)  // \(\Delta_{EQ}\) represents the set of pairs, which represents each state \((v_1,\ldots,v_K)\)
   // by respectively using the Boolean variable sets \(\bar{X}^D\) and \(X^D\);
2. \(\Delta_{GE} := \Delta_{\geq}(\tilde{v}_1,v_1) \wedge \ldots \wedge \Delta_{\geq}(\tilde{v}_K,v_K)\)  // \(\Delta_{GE}\) represents the set of pairs, which associate each state \((v_1,\ldots,v_K)\)
   // represented \(\bar{X}^D\) with its equal and dominant states, represented by \(X^D\);
3. \(\Delta_{GT} := \Delta_{GE} \wedge \neg \Delta_{EQ}\)  // \(\Delta_{GT}\) is the set of pairs where each state is associated with its dominant states;
4. \(\Delta_{BGT} := \chi_{FB}[X^D \rightarrow \bar{X}^D] \wedge \Delta_{GT}\)  // \(\Delta_{BGT}\) collects the pairs in \(\Delta_{GT}\) with the first elements as
   // the feasible boundary unsafe states;
5. \(\chi_{GB} := \exists \bar{X}^D. \Delta_{BGT}\)  // \(\chi_{GB}\) collects the states that are larger (component wise) than the states in \(\chi_{FB}\):
6. \(\chi_{FB} := \chi_{FB} \wedge \neg \chi_{BG}\)  // remove from \(\chi_{FB}\) all the non-minimal states, which belong to the set \(\chi_{BG}\):

Algorithm 3 presents a way to compute the characteristic function of the set \(\bar{F}B\) from the characteristic function \(\chi_{FB}\) obtained in Eq. (8) of the main document. Before we proceed with the discussion of this algorithm, we need to introduce two auxiliary BDD sets, collectively denoted by \{\(\Delta_=(\tilde{v}_1,v_1),\ldots,\Delta_=(\tilde{v}_K,v_K)\) and \{\(\Delta_{\geq}(\tilde{v}_1,v_1),\ldots,\Delta_{\geq}(\tilde{v}_K,v_K)\}\}, which will be useful for identifying state dominances, according to Eq. (9) of the main document, by the proposed algorithm. Each pair \(\Delta_=(\tilde{v}_k,v_k)\) and \(\Delta_{\geq}(\tilde{v}_k,v_k)\) pertain to the corresponding instance variable \(v_k\), and it can be constructed as follows:

\[
\Delta_=(\tilde{v}_k,v_k) := \bigvee_{v_k \in D_k} \left(\bar{X}^{D_k}(v_k) \wedge X^{D_k}(v_k)\right)
\]

\[
\Delta_{\geq}(\tilde{v}_k,v_k) := \bigvee_{v_k \in D_k} \left(\bar{X}^{D_k}(v_k) \wedge \bigvee_{v_k' \geq v_k} X^{D_k}(v_k')\right)
\]

In (1) and (2), \(\bar{X}^{D_k}(v_k)\) denotes the symbolic representation of the value of \(k\)-th variable \(v_k\) using a new set of Boolean variables denoted by \(\bar{X}^D\), while \(X^{D_k}(v_k)\) and \(X^{D_k}(v_k')\) denote the symbolic representations of the values \(v_k\) and \(v_k'\), of the same instance variable, using the set of the Boolean variables \(X^D\) that represent the instance variable \(v_k\) in the original BDD \(\Delta_E\). From a conceptual standpoint, \(\Delta_{\geq}(\tilde{v}_k,v_k)\) associates each value \(v_k\) with all those values \(v'_k \in D_k\) that are greater than or equal to \(v_k\) while \(\Delta_=(\tilde{v}_k,v_k)\) merely associates each value \(v_k\) with itself.

Taking as input the feasible boundary unsafe state set \(\chi_{FB}\) and the aforementioned auxiliary BDDs, the symbolic computation of the minimal feasible boundary unsafe states is formally expressed by Algorithm 3. Specifically, in Lines 1-2, Algorithm 3 constructs two BDDs, respectively denoted by \(\Delta_{EQ}\) and \(\Delta_{GE}\), by performing the conjunction operation on \{\(\Delta_=(\tilde{v}_1,v_1),\ldots,\Delta_=(\tilde{v}_K,v_K)\) and \{\(\Delta_{\geq}(\tilde{v}_1,v_1),\ldots,\Delta_{\geq}(\tilde{v}_K,v_K)\}\}. The characteristic function \(\Delta_{EQ}\) associates each state \((v_1,\ldots,v_K)\) with two different symbolic representations using the Boolean variable sets \(\bar{X}^D\) and \(X^D\), while \(\Delta_{GE}\) associates each state \((v_1,\ldots,v_K)\), represented by \(\bar{X}^D\), with a set of states, represented by \(X^D\), which are larger than or equal to \((v_1,\ldots,v_K)\). Subsequently, the symbolic computation performed at Line 3 of Algorithm 3 removes all the associations of \(\Delta_{EQ}\) from \(\Delta_{GE}\) and the resulting set is denoted by \(\Delta_{GT}\). Line 4 of Algorithm 3 computes the characteristic function \(\Delta_{BGT}\) which associates each state in \(\chi_{FB}\) with the corresponding dominant states, and, subsequently, Line 5 extracts all these dominant states into the set \(\chi_{GB}\). Finally, the set of minimal feasible boundary unsafe states, \(\chi_{FB}^\Pi\), is obtained in Line 6 by removing from \(\chi_{FB}\) the states in \(\chi_{GB}\).

Next, we prove the correctness of Algorithm 3.

**Theorem III.1.** The characteristic function \(\chi_{FB}^\Pi\) returned by Algorithm 3 recognizes correctly the minimal elements of \(\chi_{FB}\).
Proof. First we show that Algorithm 3 does not miss any minimal element of $FB$. For this, let us assume that there exists a minimal feasible boundary unsafe state $u \in FB$ that is not identified by Algorithm 3, i.e., $u \notin \overline{FB}$. Hence, state $u$ is contained in the state set $GB$ that is removed from $FB$. By the computation performed at Line 5, we know that the states of $GB$ are the second elements of the relation that is encoded by set $BGT$, which is a subset of $GT$. According to Lines 1-3, $GT$ associates each possible state $\langle v_1, \ldots, v_K \rangle$ with its dominant states (as the second elements). Then, it can be inferred from Lines 4-5 that, since $u \in GB$, there exists a state $u' \in FB$ such that the pair $(u', u)$ is in set $BGT$. Therefore, we have $u' < u$, which contradicts the working assumption of the minimality of $u$.

Next, we show that Algorithm 3 does not include any non-minimal element of $FB$. For this, let us assume that there exists a non-minimal boundary unsafe state $u'' \in \overline{FB}$. Hence, by the working assumption, there exists a minimal boundary unsafe state $u \in \overline{FB}$ such that $u'' > u$. But then, state $u''$ must be in the set $GB$ that is computed in Lines 3-5, and it should be one of the states removed from $FB$ through the computation of Line 6.

Based on the above arguments, it is concluded Algorithm 3 retains in set $\overline{FB}$ all the minimal elements of $FB$, and it excludes from this set any non-minimal elements of $FB$. \qed