Control and Data Abstraction: The Cornerstones of Practical Formal Verification.

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\textbf{Abstract.} In spite of the impressive progress in the development of the two main methods for formal verification of reactive systems -- Symbolic Model Checking and Deductive Verification, they are still limited in their ability to handle large systems. It is generally recognized that the only way these methods can ever scale up is by the extensive use of abstraction and modularization, which break the task of verifying a large system into several smaller tasks of verifying simpler systems.

In this paper, we review the two main tools of compositionality and abstraction in the framework of linear temporal logic. We illustrate the application of these two methods for the reduction of an infinite-state system into a finite-state system that can then be verified using model checking.

The technical contributions contained in this paper are a full formulation of abstraction when applied to a system with both weak and strong fairness requirements and to a general temporal formula, and a presentation of a compositional framework for shared variables and its application for forming network invariants.

\textbf{Key words:} Formal verification - Linear temporal logic - Data abstraction - Control abstraction - Network invariant - Model checking - Safety and liveness property - Weak and strong fairness

\section{Introduction}

In spite of the impressive progress in the development of the two main methods for formal verification of reactive systems -- Model Checking (in particular symbolic) and Deductive Verification, they are still limited in their ability to handle large systems. It is generally recognized that the only way these methods can ever scale up to handle industrial-size designs is by the extensive use of abstraction and modularization, which break the task of verifying a large system into several smaller tasks of verifying simpler systems.

In this paper, we review the two main tools of compositionality and abstraction in the framework of linear temporal logic. We illustrate the application of these two methods for the reduction of an infinite-state system into a finite-state system that can then be verified using model checking.

To simplify matters, we have considered two special classes of infinite-state systems for which the combination of compositionality and abstraction can effectively simplify the systems into finite-state ones. The first class is where the unboundedness of the system results from its structure. These are parameterized systems consisting of a parallel composition of finite-state processes, whose number is a varying parameter. For such systems, the source of complexity is the control or the architectural structure. We describe the techniques useful for such systems as control abstraction, since it is the control component that we try to simplify. Another source for state complexity is having data variables which range over infinite domains such as the integers. We refer to the techniques appropriate for simplifying such systems as data abstraction.

Many methods have been proposed for the uniform verification of parameterized systems, which is the subject of our control abstraction. These include explicit induction ([12], [33]) network invariants, which can be viewed as implicit induction ([20], [35], [15], [23]), methods that can be viewed as abstraction and approximation of network invariants ([4], [32], [6]), and other methods that can be viewed as based on abstraction ([16], [13]). The approach described here is based on the idea of net-
work invariants as introduced in [35], and elaborated in [20] into a working method.

There has been extensive study of the use of data abstraction techniques, mostly based on the notions of abstract interpretation ([9], [10]). Most of the previous work was done in a branching context which complicates the problem if one wishes to preserve both existential and universal properties. On the other hand, if we restrict ourselves to a universal fragment of the logic, e.g., ACTL*, then the conclusions reached are similar to our main result for the restricted case that the property \( \psi \) contains negations only within assertions.

The paper [7] obtains a similar result for the fragment ACTL*. However, instead of starting with a concrete property \( \psi \) and abstracting it into an appropriate \( \psi^c \), they start with an abstract ACTL* formula \( \Psi \) evaluated over the abstract system \( D^a \) and show how to translate (concretize) it into a concrete formula \( \psi = C(\Psi) \). The concretization is such that \( \alpha^{-1}(\psi) = \Psi \).

The survey in [8] considers an even simpler case in which the abstraction does not concern the variables on which the property \( \psi \) depends. Consequently, this is the case in which \( \psi^c = \psi \).

A more elaborate study in [11] considers a more complex specification language \(-L_\mu\), which is a positive version of the \( \mu \)-calculus.

None of these three articles considers explicitly the question of fairness requirements and how they are affected by the abstraction process.

Approaches based on simulation and studies of the properties they preserve are considered in [24] and [14].

A linear-time application of abstract interpretation is proposed in [8], applying the abstractions directly to the computational model of fair transition systems which is very close to the fds model considered here. However, the method is only applied for the verification of safety properties. Liveness, and therefore fairness, are not considered.

2 A Computational Model: Fair Discrete Structure

As a computational model for reactive systems, we take the model of fair discrete system (FDS), which is a slight variation on the model of fair transition system [28]. The FDS model was first introduced in [19] under the name "Fair Kripke Structure".

An FDS \( (V, W, C, \Theta, \rho, J, \mathcal{C}) \) consists of the following components.

- \( V = \{ u_1, \ldots, u_n \} \): A finite set of typed system variables, containing data and control variables. The set of states (interpretation) over \( V \) is denoted by \( \Sigma \). Note that \( \Sigma \) can be both finite or infinite, depending on the domains of \( V \).

- \( W = \{ w_1, \ldots, w_n \} \subseteq V \): A finite set of owned variables. These are the variables that only the system itself can modify. All other variables can also be modified by the environment. A system is said to be closed if \( W = V \).

- \( C = \{ c_1, \ldots, c_n \} \subseteq V \): A finite set of observable variables. These are the variables which the environment can observe.

It is required that \( V = W \cup C \), i.e., for every system variable \( u \in V \), \( u \) is owned, observable, or both.

- \( \Theta \): The initial condition – an assertion (first-order state formula) characterizing the initial states.

- \( \rho : A transition relation – an assertion \( \rho(V, V') \), relating the values \( V \) of the variables in state \( s \in \Sigma \) to the values \( V' \) in a \( D \)-successor state \( s' \in \Sigma \).

- \( J : \{ J_1, \ldots, J_k \} \): A set of justice (weak fairness) requirements. The justice requirement \( J \in J \) is an assertion, intended to guarantee that every computation contains infinitely many \( J \)-states (states satisfying \( J \)).

- \( C : \{ (p_1, q_1), \ldots, (p_n, q_n) \} \): A set of compassion (strong fairness) requirements. The compassion requirement \( (p, q) \in C \) is a pair of assertions, intended to guarantee that every computation containing infinitely many \( p \)-states also contains infinitely many \( q \)-states.

We require that every state \( s \in \Sigma \) has at least one \( D \)-successor. This is often ensured by including in \( \rho \) the idling disjunct \( V \neq V' \) (also called the stuttering step).

In such cases, every state \( s \) is its own \( D \)-successor.

Let \( \sigma : s_0, s_1, s_2, \ldots \), be an infinite sequence of states, \( \phi \) be an assertion, and let \( j \geq 0 \) be a natural number.

We say that \( j \) is a \( \phi \)-position of \( \sigma \) if \( s_j \) is a \( \phi \)-state.

Let \( D \) be an FDS for which the above components have been identified. We define a computation of \( D \) to be an infinite sequence of states \( \sigma : s_0, s_1, s_2, \ldots \), satisfying the following requirements:

- **Initiality:** \( s_0 \) is initial, i.e., \( s_0 \models \Theta \).

- **Consecution:** For each \( j = 0, 1, \ldots \), the state \( s_{j+1} \) is a \( D \)-successor of the state \( s_j \).

- **Justice:** For each \( J \in J \), \( \sigma \) contains infinitely many \( J \)-positions.

- **Compassion:** For each \( (p, q) \in C \), if \( \sigma \) contains infinitely many \( p \)-positions, it must also contain infinitely many \( q \)-positions.

For an FDS \( D \), we denote by \( \text{Comp}(D) \) the set of all computations of \( D \). An FDS \( D \) is called feasible if \( \text{Comp}(D) \neq \emptyset \), namely, if \( D \) has at least one computation.

An infinite state sequence \( \sigma \) is called a run of \( D \) if it satisfies the requirements of initiality and consecution but not, necessarily, any of the fairness requirements. System \( D \) is said to be viable if every finite run can be extended into a computation. One of the differences between the model of fair transition systems and the FDS model is that every FTS is viable by construction, while it is easy to define an FDS which is not viable, e.g., by having the justice list include the assertion \text{false}. On the
other hand, every FDS which is derived from a program is viable.

All our concrete examples are given in SPL (Simple Programming Language), which is used to represent concurrent programs (e.g., [28], [26]). Every SPL program can be compiled into an FDS in a straightforward manner. In particular, every statement in an SPL program contributes a disjunct to the transition relation. For example, the assignment statement

$$\ell_0 : y := x + 1; \ell_1 :$$
can be executed when control is at location $\ell_0$. When executed, it assigns $x + 1$ to $y$ while control moves from $\ell_0$ to $\ell_1$. This transition contributes to $\rho$ the disjunct

$$\rho_{\ell_0} : \text{at}_{\ell_0} \wedge \text{at}_{\ell_1} \wedge y' = x + 1 \wedge x' = x,$$

The predicates $\text{at}_{\ell_0}$ and $\text{at}_{\ell_1}$ stand, respectively, for the assertions $\pi_0 = 0$ and $\pi_1 = 1$, where $\pi_i$ is the control variable denoting the current location within the process to which the statement belongs.

Every variable declared in an SPL program is specified as having one of the modes $in$, $out$, $in-out$, or $local$. This specification determines whether the variable is considered to be owned or observable or both according to the following table

<table>
<thead>
<tr>
<th>Mode</th>
<th>Owned?</th>
<th>Observable?</th>
</tr>
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<tbody>
<tr>
<td>in</td>
<td>N</td>
<td>Y</td>
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<tr>
<td>out</td>
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<td>Y</td>
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<tr>
<td>in-out</td>
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<td>Y</td>
</tr>
<tr>
<td>local</td>
<td>Y</td>
<td>N</td>
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</tbody>
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3 Operations on FDS's

There are several important operations, one may wish to apply to FDS's.

The first useful set of operations on programs and systems is forming their parallel composition, implying that the two systems execute concurrently. Consider the two fair discrete systems $D_1 = (V_1, W_1, \Theta_1, \rho_1, J_1, C_1)$ and $D_2 = (V_2, W_2, \Theta_2, \rho_2, J_2, C_2)$. We consider two ways of forming the parallel composition of $D_1$ and $D_2$.

3.1 Asynchronous Parallel Composition

The systems $D_1$ and $D_2$ are said to be composable if $W_1 \cap W_2 = \emptyset$, $V_1 \cap V_2 = \emptyset$, and neither system modifies the variables owned by the other, i.e.,

$$\rho_1 \rightarrow \text{pres}(W_2 \cap V_1) \quad \text{and} \quad \rho_2 \rightarrow \text{pres}(W_1 \cap V_2).$$

The first condition requires that a variable can only be owned by one of the systems. The second condition requires that variables known to both systems must be observable in both.

For composable systems $D_1$ and $D_2$, we define their asynchronous parallel composition, denoted by $D_1 || D_2$, to be the system $\mathcal{D} = (V, W, \Theta, \rho, J, \mathcal{C})$, where

$$V = V_1 \cup V_2 \quad W = W_1 \cup W_2 \quad \Theta = \Theta_1 \wedge \Theta_2 \quad J = J_1 \cup J_2 \quad \mathcal{C} = C_1 \cup C_2 \quad \rho = \rho_1 \wedge \rho_2.$$

For a set of variables $U \subseteq V$, the predicate $\text{pres}(U)$ stands for the assertion $U' = U$, implying that all the variables in $U$ are preserved by the transition.

Obviously, the basic actions of the composed system $\mathcal{D}$ are chosen from the basic actions of its components, i.e., $D_1$ and $D_2$. Thus, we can view the execution of $\mathcal{D}$ as the interleaved execution of $D_1$ and $D_2$.

As seen from the definition, $D_1$ and $D_2$ may have disjoint as well as common system variables, and the variables of $\mathcal{D}$ are the union of all of these variables. The initial condition of $\mathcal{D}$ is the conjunction of the initial conditions of $D_1$ and $D_2$. The transition relation of $\mathcal{D}$ states that at any step, we may choose to perform a step of $D_1$ or a step of $D_2$. However, when we select one of the two systems, we should also take care to preserve the private variables of the other system. For example, choosing to execute a step of $D_1$, we should preserve all variables in $V_2 \setminus V_1$ and all the variables owned by $D_2$.

The justice and compression sets of $\mathcal{D}$ are formed as the respective unions of the justice and compression sets of the component systems.

Asynchronous parallel composition corresponds to the SPL parallel operator $||$ constructing a program out of concurrent processes.

3.2 Synchronous Parallel Composition

We define the synchronous parallel composition of $D_1$ and $D_2$, denoted by $D_1 ||| D_2$, to be the system

$$\mathcal{D} : (V, W, \Theta, \rho, J, \mathcal{C}),$$

where

$$V = V_1 \cup V_2 \quad W = W_1 \cup W_2 \quad \Theta = \Theta_1 \wedge \Theta_2 \quad J = J_1 \cup J_2 \quad \mathcal{C} = C_1 \cup C_2 \quad \rho = \rho_1 \wedge \rho_2.$$
3.3 Modularization of an FDS

Let $P$ be an SPL program and $D$ its corresponding FDS. The standard compilation of a program into an FDS views the program as a closed system which has no interaction with its environment. In the context of compositional verification, we need an open system view of an FDS, which takes into account not only actions performed by the system but also actions (in particular, variable changes) performed by the environment.

Let $D: \langle V, W, O, \Theta, \rho, J, C \rangle$ be an FDS and $s \notin V$ be a fresh boolean variable. The modular version of $D$ is given by $D_m: \langle V_m, W, O_m, \Theta_m, \rho_m, J, C \rangle$, where,

$$V_m = V \cup \{s\}, \quad O_m = O \cup \{s\}, \quad \rho_m = (\rho \land s') \lor (W' = W \land \neg s').$$

That is, $D_m$, the modular version of $D$, admits as an additional action a transition which preserves the values of all variables owned by $D$ but allows all other shared variables to change in an arbitrary way. This provides the most general representation of an environment action. The scheduling variable $s$ is used to ensure interleaving between the module and its environment. We refer to $D_m$ as the modular or open version of system $D$.

We define a modular computation of $D$ to be any computation of $D_m$.

3.4 Restricting an Open Shared Variable

When constructing a system out of smaller components, it is often the case that all processes within the system are allowed to access a certain shared variable, but only a subset of the processes is allowed to modify it's value. For example, we may have a system

$$D = D_1 \parallel D_2 \parallel D_3,$$

in which all processes are allowed to access variable $x$, but only processes $D_1$ and $D_2$ are allowed to modify its value.

We provide a special restriction (sealing-off) operation, which moves one of the system variables to the category of owned variables, thereby disallowing its modification by the environment.

Let $D: \langle V, W, O, \Theta, \rho, J, C \rangle$ be an FDS and let $U \subseteq V - W$ be a set of variables which are not owned by $D$.

The result of restricting $U$ in $D$, denoted by $D \setminus U$, is the FDS $D_r: \langle V_r, W_r, O, \Theta, \rho, J, C \rangle$, where $W_r = W \cup U$.

Thus, to represent a system consisting of sub-systems $D_1$, $D_2$, and $D_3$, in which $D_3$ is not allowed to modify variable $x$, we may write

$$D = ((D_1 \parallel D_2)[x]) \parallel D_3.$$

To represent the closing off of an entire system $D$, we write $D_{\text{sm}}$, which is an abbreviation for $D \setminus (V - W)$.

This restricts the environment from writing on any of the system variables. A system such that $W = V$ is often described as a closed system, because it can have no interaction with its environment.

4 Specification Language: Temporal Logic

As a requirement specification language for reactive systems we take temporal logic (TL) [27]. For simplicity, we consider only the future fragment of TL. Extending the approach to the full logic is straightforward.

We assume an underlying assertion language $\mathcal{L}$ which contains the predicate calculus and interpreted symbols for expressing the standard operations and relations over some concrete domains. A temporal formula is constructed out of state formulas (assertions) to which we apply the boolean operators $\neg$ and $\lor$ (the other boolean operators can be defined from these), and the basic temporal operators $\bigwedge$ (next) and $U$ (until).

A model for a temporal formula $p$ is an infinite sequence of states $\sigma : s_0, s_1, \ldots$, where each state $s_j$ provides an interpretation for the variables mentioned in $p$.

Given a model $\sigma$, we present an inductive definition for the notion of a temporal formula $p$ holding at a position $j \geq 0$ in $\sigma$, denoted by $(\sigma, j) \models p$.

- For a state formula $p_j$, $(\sigma, j) \models p_j \iff s_j \models p_j$

  That is, we evaluate $p$ locally, using the interpretation given by $s_j$,

  - $(\sigma, j) \models \neg p_j \iff (\sigma, j) \not\models p_j$
  - $(\sigma, j) \models p_j \lor q \iff (\sigma, j) \models p_j \lor (\sigma, j) \models q$
  - $(\sigma, j) \models \bigwedge p \iff (\sigma, j + 1) \models p$
  - $(\sigma, j) \models p U q \iff$ for some $k \geq j, (\sigma, k) \models q$, and for every $i$ such that $j \leq i < k, (\sigma, i) \models p$

Additional temporal operators can be defined by

$$\diamond p = \text{true \ U\ p \ (eventually)}$$

$$\square p = \neg \diamond \neg p \ (\text{henceforth})$$

For a temporal formula $p$ and a position $j \geq 0$ such that $(\sigma, j) \models p$, we say that $j$ is a $p$-position (in $\sigma$). If $(\sigma, 0) \models p$, we say that $p$ holds on $\sigma$, and denote it by $\sigma \models p$. A formula $p$ is called satisfiable if $p$ holds on some model. A formula $p$ is called valid, denoted by $\models p$, if $p$ holds on all models.

Given an FDS $D$ and a temporal formula $p$, we say that $p$ is $D$-valid, denoted by $D \models p$, if $p$ holds on all models which are computations of $D$. A property $\varphi$ is said to be modularly valid over FDS $D$, denoted $D \models_M \varphi$, if $\varphi$ is $D_{\text{sm}}$-valid, i.e., $D_{\text{sm}} \models \varphi$.

An algorithm for model checking whether a temporal formula $p$ is valid over a finite-state FDS $D$ is presented in [19]. The paper presents a version of the algorithm using explicit state enumeration methods as well as a symbolic version. Based on the ideas developed in [22] and [5], the approach calls for the construction of a tester for the negation of $p$. This is an FDS $D_{\neg p}$ whose computations are all the sequences which satisfy the negated formula $\neg p$. Then, we form the synchronous parallel composition $D_{\text{comp}} = D || D_{\neg p}$ and check for feasibility. If $D_{\text{comp}}$ is found to be feasible, this implies that $D$ has a computation which violates $p$ and therefore $p$ is not valid over
\section{Control Abstraction}

Let $U \subseteq V$ be a subset of the system variables. For a $V$-state $s$, we denote by $s\downarrow_U$ the $U$-state obtained by projecting $s$ onto $U$. That is, the interpretation $s$ restricted to the domain $U$.

The state sequence $\sigma : s_0, s_1, \ldots$ is defined to be an observation of the FDS $D : \langle V, W, O, F, \rho, J, C \rangle$ if $\sigma$ is a stuttering variant of the $O$-projection

$$\sigma\downarrow_O = s_0\downarrow_O, s_1\downarrow_O, \ldots,$$

where $\sigma : s_0, s_1, \ldots$ is a computation of $D$. Let $\text{Obs}(D)$ denote the set of all observations of system $D$.

The two FDS's $D_A : \langle V_A, W_A, O_A, F_A, \rho_A, J_A, C_A \rangle$ and $D_C : \langle V_C, W_C, O_C, F_C, \rho_C, J_C, C_C \rangle$ are defined to be comparable, if $O_A = O_C$ and $O_A \cap W_A = O_C \cap W_C$. The FDS $D_A$ is an abstraction of the comparable $D_C$, denoted by $D_C \subseteq D_A$, if $\text{Obs}(D_C) \subseteq \text{Obs}(D_A)$, i.e., every observation of $D_C$ is also an observation of $D_A$. We refer to $D_C$ and $D_A$ as the concrete and abstract systems, respectively. The abstraction relation is obviously reflexive and transitive.

It would have been very useful if the abstraction relation defined above, would have been compositional with respect to (asynchronous) parallel composition. That is, if $D_C \subseteq D_A$, would have implied $(D_C \parallel Q) \subseteq (D_A \parallel Q)$ for every FDS $Q$. Unfortunately, this is not the case.

Consider, for example, the FDS's corresponding to programs \textsc{incx} and \textsc{incy} presented in Fig. 1. Up to stuttering and idling, both of these FDS's have the unique observation

$$\langle x : 0, y : 0 \rangle, \langle x : 1, y : 1 \rangle, \langle x : 2, y : 2 \rangle, \ldots$$

It follows that \textsc{incx} and \textsc{incy} have the same set of observations. In particular, this implies that $\text{incx} \equiv \text{incy}$.

However, when we consider the program $Q$ given by

\begin{verbatim}
in-out x : integer
  loop forever do
    m0 : x := 0
  end
end
\end{verbatim}

we find out that the observation

$$\langle x : 0, y : 0 \rangle, \langle x : 1, y : 1 \rangle, \langle x : 0, y : 1 \rangle, \langle x : 1, y : 1 \rangle, \ldots$$

belongs to $\text{Obs}(\text{incx} \parallel Q)$ but does not belong to the set $\text{Obs}(\text{incy} \parallel Q)$. Consequently, while $\text{incx} \equiv \text{incy}$,

$$\text{incx} \parallel Q \not\equiv \text{incy} \parallel Q,$$

which shows that the abstraction relation $\equiv$ is not compositional.

Obviously, the problem lies in the fact that the relation $\equiv$ is based on the set of observations of the closed-system FDS semantics of programs. The difference between programs \textsc{incx} and \textsc{incy} can be observed only when we take into account actions of the environment, such as resetting variable $x$ to 0. In the definition of the computations (and therefore observations) of the FDS assigned to these programs, such actions are not represented.

Once we diagnose the malady, the remedy is quite straightforward. We say that FDS $D_A$ is a modular abstraction of the comparable $D_C$, denoted by $D_C \equiv_m D_A$, if $\text{Obs}(D_C) \subseteq \text{Obs}(D_A)$, i.e., every observation of $(D_C)_M$ the modularized version of $D_C$ is also an observation of $(D_A)_M$, the modularized version of $D_A$.

Note that, while \textsc{incy} is a plain abstraction of \textsc{incx}, it is not a modular abstraction of \textsc{incx}. To see this we point to

$$\langle s : 0, x : 0, y : 0 \rangle, \langle s : 1, x : 1, y : 1 \rangle, \langle s : 0, x : 0, y : 1 \rangle, \langle s : 1, x : 1, y : 1 \rangle, \ldots,$$

which is an observation of $(\text{incx})_M$ but not of $(\text{incy})_M$.

When we upgrade from plain abstraction to modular abstraction, we obtain the desired property of compositionality of the abstraction relation with respect to the operations of parallel composition and restriction, as stated by the following claim:

\textbf{Claim 1.} Let $D_C$ and $D_A$ be two comparable FDS's such that $D_C \equiv_m D_A$. Then, for every FDS $Q$, and temporal formula $\varphi$,

1. $(D_C \parallel Q) \equiv_m (D_A \parallel Q)$
2. $(D_C)_M \equiv_m (D_A)_M$
3. $D_A \models \varphi$ implies $D_C \models \varphi$

We describe these compositional properties by saying that the operations of parallel composition and restriction are monotonic with respect to modular abstraction, while temporal validity is anti-monotonic.

This indicates how we propose to use abstraction in order to simplify the verification task. Namely, given a property $p$ to be verified over a complex system $D_C$, we use modular abstraction in order to derive a simpler system $D_A$ and then verify that $p$ is $D_A$-valid. Note that the implication is still in one direction. Namely, validity over the abstract system implies concrete validity but not, necessarily, vice versa. The most striking applications of this strategy are when $D_C$ is an infinite-state system, while its abstraction $D_A$ is finite-state and thus amenable to verification by model checking.

The comparable FDS's $P$ and $Q$ are defined to be modularly equivalent, denoted $P \sim_m Q$, if both $P \equiv_m Q$ and $Q \equiv_m P$.

\section{Verification by Abstract Network Invariants}

In this section, we concentrate on cases in which the system is a parallel composition $P(n) : (P_1 \parallel \cdots \parallel P_n)_n$,
where each \( P_i \) is a finite-state system. The final restriction of the parallel composition guarantees that no further interference from the environment is possible. The unbounded number of states for system \( P(n) \) comes from the fact that we consider an infinite family of systems, and yet wish to verify uniformly (i.e., for every value of \( n \geq 1 \)) that the property \( p \) is valid.

The general principles of the method and one of the examples presented in this section are shared with [20]. The main differences between the two presentations are that, while [20] considers processes communicating by synchronous message passing, we focus here on communication by shared variables, and we find the abstraction we use somewhat simpler to comprehend, perhaps due to the different communication mechanisms.

For simplicity, assume that the property \( p \) only refers to the observable variables of \( P_1 \) and that processes

\[ P_2, \ldots, P_n \]

are identical (up to renaming). The strategy we propose can be summarized as follows:

**Verification by Abstract Network Invariants**

1. Devise a network invariant \( I \), which is an FDS intended to provide a modular abstraction for the parallel composition \( P_2 \parallel \cdots \parallel P_n \) for any \( n \).
2. Confirm that \( I \) is indeed a network invariant, by model checking that \( P_2 \parallel \cdots \parallel P_n \models M I \) and that \( (I \parallel I) \models M I \). The technique of model checking a modular abstraction is presented in Subsection 6.3.
3. Model check \( D_R \models p \), where \( D_R \) is the restricted system \( (P_1 \parallel D_R) \).

We argue that this strategy is sound. Namely, if \( D_R \models p \) then \( P(n) \models p \) for every \( n > 1 \). Step 2 of the strategy establishes \( P_2 \parallel \cdots \parallel P_n \models M I \). By monotonicity of the parallel composition (Claim 1), it follows that

\[ P_1 \parallel \cdots \parallel P_n \models M (P_1 \parallel I) \]

By monotonicity of the restriction operation, we conclude that

\[ P(n) = (P_1 \parallel \cdots \parallel P_n)_{M} \models M (P_1 \parallel I)_{R} = D_R. \]

Due to the anti-monotonicity of the validity relation, it follows that \( D_R \models p \) implies \( P(n) \models p \), establishing that the proposed strategy is sound.

**Fig. 1.** Programs INCX and INCY

**Fig. 2.** Program MUX-SEM.

Step 1 in the strategy is the only one requiring ingenuity and which cannot be fully mechanized. However, while presenting the examples, we will provide some explanations and clues for the choices we made.

### 6.1 Mutual Exclusion by Semaphores

As our first running example, we use program MUX-SEM presented in Fig. 2. The program consists of \( n \) processes. Each process \( P[i] \) cycles through three possible locations: \( N_i, T_i, \) and \( C_i \). Location \( N_i \) represents the non-critical activity which the process can perform without coordination with the other processes. Location \( T_i \) is the “trying” location, at which a process decides it needs to access its critical location. At the trying location, the process waits for the semaphore variable \( y \) to become 1. On entering the critical section \( C_i \), the process sets \( y \) to 0. Finally, \( C_i \) is the critical location which should be reachable only exclusively by one process at a time. On exit from the critical section, variable \( y \) is reset to 1.

The mode \( \text{local} \) specified for variable \( y \) identifies \( y \) as being owned by the entire system but not by any of the individual processes. This specification ensures that the variable \( y \) cannot be modified by an environment agent external to the program. By the standard compilation of SPL programs, each process \( P[i] \) is associated with a justice requirement \( J_i : \neg C_i \) and a compassion requirement \( C_i : (T_i \land y > 0) \lor C_i \). The justice requirement ensures that process \( P[i] \) does not remain stuck forever at location \( C_i \). The compassion requirement ensures that process \( P[i] \) does not remain stuck forever at location \( T_i \) while \( y \) turns positive infinitely many times. Note that a process may choose to stay forever at \( N_i \) or may get stuck at \( T_i \) if \( y \) turns positive only finitely many times and then remains zero forever. The latter behavior cannot occur in program MUX-SEM but this can be established only by a global analysis of the complete system.
In Fig. 3, we present process INV-CAND, which is our first candidate for the network invariant abstracting $P[2] \parallel \cdots \parallel P[\eta]$. In this section we choose to represent FDS's by transition diagrams, in which we explicitly list the fairness requirements. Process INV-CAND can be obtained by simplifying a single copy of the concrete $P[2]$. The simplification consists of merging locations $N$ and $T$ into a single location $N$ and relaxing the fairness requirements associated with this combined location. This simplification is suggested by noting that the main liveness requirement of accessibility is studied only for process $P[1]$. The only liveness properties we require from the environment processes $P[2], \ldots, P[\eta]$, is that they eventually exit their critical sections and release the semaphore. Thus, while being at location $N$, process INV-CAND may choose non-deterministically to stay at $N$ or move to $C$ if $y$ equals 1. There is no justice requirement associated with location $N$, due the possibility that the process may choose to remain there. On the other hand, with location $C$, we associate the justice requirement $\neg C$ which excludes behaviors in which INV-CAND gets stuck at $C$. Let us denote by $C_a$ the FDS corresponding to INV-CAND.

A useful heuristic that often leads to the generation of network invariants is forming the sequence of FDS's $I_1 = C_a, I_2 = C_a \parallel C_a, I_3 = I_2 \parallel C_a, \ldots$, and comparing every two successive $I_i$'s, hoping that the sequence will converge. Convergence means that we identify an index $j \geq 0$ such that $I_j \approx_m I_{j+1}$. Trying this approach with the FDS $C_a$ fails. Comparing $I_2: C_a \parallel C_a$ with $I_1: C_a$, we find that $(I_2)_m$ can generate the observation

$$\langle y: 1, s: 0 \rangle, \langle y: 0, s: 1 \rangle, \langle y: 1, s: 0 \rangle, \langle y: 0, s: 1 \rangle, \langle y: 1, s: 1 \rangle, \langle y: 0, s: 0 \rangle, \langle y: 1, s: 1 \rangle, \cdots,$$

which cannot be generated by $(I_1)_m$. Such a behavior can be explained as a scenario in the behavior of $P[2] \parallel P[3]$ under an unrestricted environment. First, $P[2]$ enters its critical section according to the step $\langle y: 1, s: 0 \rangle \rightarrow \langle y: 0, s: 1 \rangle$. Then, while $P[2]$ is still in its critical section, the environment raises $y$ to 1, according to the step $\langle y: 0, s: 1 \rangle \rightarrow \langle y: 1, s: 0 \rangle$ (we know that this is an environment step because $s' = 0$). Then $P[3]$ enters its own critical section, as recorded in $\langle y: 1, s: 0 \rangle \rightarrow \langle y: 0, s: 1 \rangle$. Following that, $P[2]$ exits its critical section $(\langle y: 0, s: 1 \rangle \rightarrow \langle y: 1, s: 1 \rangle)$, the environment resets $y$ to 1 $(\langle y: 1, s: 1 \rangle \rightarrow \langle y: 0, s: 0 \rangle)$, and finally $P[3]$ exits $(\langle y: 0, s: 0 \rangle, \langle y: 1, s: 1 \rangle)$. What is special about this behavior is that $I_2$ exits twice in succession without an observable entry between these two exits. In all behaviors of $I_1: C_a$, which has only one copy of $P[2]$, every two exits must be separated by an observable entry.

In a similar way, we find that $I_3$ can exit its critical sections three times in succession, if the environment cooperates, which cannot be done by $I_2$. This shows that the sequence $I_1, I_2, \ldots$ will never converge. Looking closer at this example, we realize that the factor that differentiates between $I_1$ and $I_2$ and between $I_2$ and $I_3$ is their response to a behavior of the environment which will never be realized in the closed system, namely raising the semaphore variable to 1 while one of the processes is in its critical section. This leads us to the next (and final) abstraction $I_{max}$, presented in Fig. 4.

The system $I_{max}$ behaves as $C_a$ as long as the environment behaves properly. However, once it detects that the environment raised the value of $y$ from 0 to 1 while the system was in the critical section, it goes into a chaos control state in which "anything goes". That is, all arbitrary sequences of values for the observable variables will be accepted from this point on. It is obvious that $I_{max}$ is an abstraction of $C_a$ because it differs from $C_a$ in all the additional behaviors it is ready to generate once it reached the chaos state.

It is not difficult to verify that $I_{max}$ is a network invariant. We model checked that $C_a \subseteq_m I_{max}$ and that $(I_{max} \parallel I_{max}) \subseteq_m I_{max}$.

It only remains to perform step 3 in the abstraction strategy presented in the beginning of the section. We form the closed system FDS $D = (P[2] \parallel P[3])$ and use model checking to verify the liveness property $D \models \square(N_1 \rightarrow \Diamond C_1)$. This has been done and established that process $P[1]$ of program MUX-SEM has the property of accessibility for any number of processes.

6.2 The Dining Philosophers Problem

As a more advanced example, we applied the technique described above to the problem of the dining philosophers. As originally described by Dijkstra, $n$ philosophers are seated at a round table. Each philosopher alternates between a thinking phase and a phase in which he becomes hungry and wishes to eat. There are $n$ chopsticks placed around the table, one chopstick between
every two philosophers. In order to eat, each philosopher needs to acquire the chop-sticks on both sides. A chop-stick can be possessed by only one philosopher at a time.

A solution to the dining philosophers problem, using semaphores, is presented by program DINE-CONTR of Fig. 5.

In this program, philosophers $P[2], \ldots, P[n]$ reach first for the chop-stick on their left, represented by semaphore variable $c[j]$ for philosopher $j$, and then for their right chop-stick (semaphore $c[j] \oplus 1$). Philosopher $P[1]$ behaves differently, reaching first for his right chop-stick ($c[2]$) and only later for his left chop-stick ($c[1]$). We wish to prove the liveness property of accessibility for each of the philosophers, which can be specified by the temporal formula

$$\psi_{\text{acc}}: \square (at \cdot L_j \rightarrow \Diamond (at \cdot L_{j+1})), $$

for every $j = 1, \ldots, n$. This property ensures that every hungry philosopher eventually gets to eat.

Proceeding through a sequence of abstraction steps similar to the previous example, we finally wind up with the FDS $I_{\text{contr}}$ presented in Fig. 6.

The diagram of Fig. 6 consists of two components that operate in parallel, one taking care of the left semaphore $L$ and the other handling the right semaphore $R$. Whenever an environment fault is detected, i.e. the environment raises a semaphore that has been lowered by the system, both components escape to the chaos state after which all behaviors are possible. By the graphical conventions, the transitions to a chaos state have priority over internal transitions such as the one connecting $(N_R, R)$ to $(C_R, \neg R)$.

Since $I_{\text{contr}}$ is intended to abstract behaviors of a string of consecutive philosophers

$$P[i] || P[i+1] || \cdots || P[j],$$

we should not be surprised that the behavior of the left semaphore $L$ is only loosely coupled with that of the right semaphore $R$. This is because $L$ stands for $c[j]$ (assuming $j > 1$) the left semaphore of process $P[j]$ the leftmost process in the string, while $R$ stands for $c[j+1]$ the right semaphore of $P[j]$, the rightmost philosopher. There is still a weak coupling which is expressed through the fairness requirement. For ordinary philosophers, who take the right chop-stick last, the obligation to release semaphore $R$ once it is taken, can be guaranteed locally, independently of the environment. This is expressed by the justice requirement $\neg C_R$ forbidding the system to remain forever in $C_R$ with the semaphore $R$ occupied. The situation is different with the left semaphore $L$. No sub-system (modeled by $I_{\text{contr}}$) can unconditionally guarantee release of $L$ once it is taken. Consequently, the fairness requirement guaranteeing the release of $L$ is formulated as the composition requirement $(R, \neg C_L)$ making the release of $L$ (as implied by being at $N_L$) conditional on the infinite recurrence of an available $R$. Already at the level of a single philosopher, after acquiring $L$ the system proceeds to acquire $R$. If $R$ is not available with sufficient frequency, the system will fail in obtaining it, and will keep $L$ occupied forever.

It is straightforward to verify (using model checking) that $I_{\text{contr}}$ modularly abstracts any of the processes $P[2], \ldots, P[n]$ and that $(I_{\text{contr}} \ || \ I_{\text{contra}}) \subseteq_M I_{\text{contr}}$.

It follows that $I_{\text{contr}}$ is a network invariant for any sequence of regular philosophers. We can combine $I_{\text{contr}}$ with $P[1]$ to establish the accessibility properties of the contrary philosopher $P[1]$.

We can also verify the accessibility property for all ordinary philosophers. To do so, we consider the combination $P[1] || I_{\text{contr}} || P[\text{ordinary}] || I_{\text{contra}}$ in which we use the network invariant $I_{\text{contr}}$ as an abstraction for the sequence of philosophers separating $P[1]$ from $P[\text{ordinary}]$ and then again as an abstraction for the sequence of philosophers separating $P[\text{ordinary}]$ from $P[1]$ in the other direction.

In all of these combinations, we should remember to close the ring by identifying the leftmost semaphore of the combination with the rightmost semaphore.

### 6.3 Model Checking Modular Abstraction

When carrying out the abstraction process as described in this section, we are repeatedly required to verify that one FDS modularly abstracts another. Most of the available computer-aided verification (CAV) tools for LTL (e.g., STEP [2] and TLM-BASIC[31]) are designed to support verification tasks. That is, they accept as inputs a system description, equivalent to an FDS $D$, and a temporal formula $\varphi$ and attempt to establish (or refute) that $D \models \varphi$.

In this subsection, we show how the modular abstraction problem $D_\alpha \subseteq_M D_\beta$ can be reduced into a verification problem. This reduction can be used in order to establish the modular abstraction relation between systems while using the available LTL verification tools.
The idea of proving abstraction (equivalently refinement) by forming a superposition of the abstract and concrete systems, as we do here, has been proposed in [17]. The underlying theory of proving abstraction by simulation relations is thoroughly discussed in [1] and applied in [25, 34, 29, 21].

In theory, for the case that both $D_C$ and $D_A$ are finite-state, the modular abstraction problem is algorithmically solvable. All that is required is to convert the two systems into $\omega$-automata, compute the complement of the $D_A$-automaton, and check that the languages of $D_C$ and $\overline{D_A}$ have an empty intersection. However, very few symbolic model checkers provide that capability of complementing a system and, even when they do, this operation could be exponentially expensive.

Instead, we base our approach on the simple observation that when $D_A$ is deterministic, it is possible to construct a combined system which will try to emulate the joint computation of the two systems. For the case that $D_A$ is non-deterministic, we rely on the user to provide an additional restriction on the possible actions of $D_A$, reducing them to a single possible action. Thus, we trade computational complexity for full automation, and our approach may require user interaction.

Consider two comparable FDS’s:

$$D_C : \langle V_C, W_C, \Theta_C, \Theta_C, \rho_C, J_C, C_C \rangle$$

and its proposed abstraction

$$D_A : \langle V_A, W_A, \Theta_A, \Theta_A, \rho_A, J_A, C_A \rangle,$$

and assume we wish to establish that $D_C \subseteq_M D_A$. Without loss of generality, we can assume that $V_C \cap V_A = \emptyset$, but that there exists a 1-1 correspondence between the variables of $\Theta_C$ and those of $\Theta_A$.

We say that the FDS $D_S : \langle V_S, W_S, \Theta_S, \Theta_S, \rho_S, J_S, C_S \rangle$ is a superposition of $D_C$ and $D_A$ if it has the following form:

$$V_S = V_C \cup V_A$$

$$W_S = W_C \cup W_A$$

$$\Theta_S = \Theta_C \cup \Theta_A$$

$$\Theta_S = \Theta_C \land \Theta_A$$

$$\wedge (\forall V_A : \Theta_A \land \Theta_d \land \Theta_A = \Theta_C) \rightarrow \Theta_A = \Theta_C$$

$$\rho_S = (\forall \rho_P \land (\exists V_A : \rho_P \land \Theta_A = \Theta_C) \rightarrow \Theta_A = \Theta_C)$$

$$\vee (\exists \rho_E : \rho_E \land \Theta_A = \Theta_C \rightarrow \Theta_A = \Theta_C)$$

where

$$\rho_P = \rho_C \land \rho_A \land \rho_d$$

$$\rho_E = \rho_{pres(W_C)} \land \rho_{pres(W_A)}$$

$$J_S = J_C$$

and $C_S = C_C$.

The general idea in the construction of the superposition system $D_S$ is that every computation of $D_S$ induces a computation of $D_C$ (when projected on $V_C$) and a run of $D_A$ (when projected $V_A$). Thus, a computation of $D_S$ can be viewed as a joint computation of the two systems $D_C$ and $D_A$. There are two desired features a successful superposition of $D_C$ and $D_A$ should satisfy.

1. Every computation of $D_C$ is induced by some computation of $D_S$. Thus, the additional conjunctions in $\Theta_S$ and $\rho_S$ should not restrict the behavior of $D_C$.

2. To the best of its ability, $D_S$ should attempt to maintain the correspondence $\Theta_S = \Theta_A$. This explains the role of the implications conjuncted into $\Theta_S$ and $\rho_S$.

These implications require that, if it is possible to choose abstract variables which are consistent with the constraints of $D_A$ and maintain $\Theta_S = \Theta_A$, then such a choice should be made.

Note also that the system $D_S$ has already been modularized by defining $\rho_S$ as the choice between a system step $\rho_P$ which is compatible with $\rho_C \land \rho_A$ and an environment step $\rho_E$ which only guarantees the preservation of $W_S = W_C \cup W_A$.

The system $D_S$ has $\Theta_d$ and $\rho_d$ as open parameters, which should be provided by the user. Once they are specified, $D_S$ can be automatically constructed from $D_C$ and $D_A$, and this is what has been implemented in the current TLY-BASIC implementation of the modular abstraction checker within TLY.
A very simple choice is to take $\Theta_d = \rho_d = 1$, namely, take them both as being identically true. This choice is adequate in all cases that the abstract system $D_A$ is deterministic. Determinism in the abstraction context means that, for every $D_A$-state $s$ and a set of specified values $U$ for the observable variables $O_A$, there exists at most one $s'$ a $p_A$-successor of $s$, such that $s'[O_A] = U$. All the examples presented in this section such as the network invariants presented in Fig. 4 and Fig. 6 are deterministic. In fact, one of the reasons for eliminating the trying location $T$ in process INV-CAND and the other network invariants was to make them deterministic. In view of this, all the modular abstractions mentioned in this section were resolved by superposition systems in which we have taken $\Theta_d = \rho_d = 1$.

The following claim makes precise the relation between computations of $D_{s}$, computations of $D_{c}$, and runs of $D_{A}$.

**Claim 2.** If $\sigma$ is a computation of $D_{s}$, then $\sigma|_{V_{c}}$ is a computation of $D_{c}$ and $\sigma|_{V_{A}}$ is a run of $D_{A}$.

In the preceding discussion, we listed two features which are desirable in a good superposition. However, these features are not automatically guaranteed. In Fig. 7, we present a proof rule whose premises guarantee that the system $D_{s}$ has the desired features.

Premise A1 guarantees that for every value assignment to the concrete variables $V_{c}$ satisfying $\Theta_{c}$, there exists a value assignment to the abstract variables $V_{A}$ satisfying $\Theta_{c} \land \Theta_{d}$. Thus, $\Theta_{A} \land \Theta_{d}$ does not restrict the choice of values for $V_{c}$.

Premise A2 stipulates a similar non-restriction requirement for $\rho_{d}$. It requires that, for every value assignments to $V_{c}$, $V_{c}'$, and $V_{A}$, which make $\rho_{d}(V_{c},V_{c}')$ true, there exists a value assignment to $V_{A}'$ which satisfies $\rho_{d}(V_{A},V_{A}') \land \rho_{d}(V_{c},V_{c}',V_{A},V_{A}')$. Thus, $\rho_{A} \land \rho_{d}$ does not restrict the choice of values for $V_{c}$, $V_{c}'$, and $V_{A}$.

Finally, premise A3 requires that every computation $\sigma$ of $D_{s}$ maintains the invariant $\Box(O_{c} = O_{d})$, and $\sigma|_{V_{A}}$, the projection of $\sigma$ on the abstract variables $V_{A}$ yields a computation of $D_{A}$. According to Claim 2, $\sigma|_{V_{A}}$ is a run of $D_{A}$. Adding to it the fact that $\sigma$ satisfies all the fairness requirements of $D_{A}$, as established by A3, we can conclude that $\sigma|_{V_{A}}$ is also a computation of $D_{A}$.

The following claim states that rule MOD-ABST is sound.

**Claim 3.** If the premises of rule MOD-ABST are valid for some choice of $\Theta_{d}$ and $\rho_{d}$, then $D_{A}$ is a modular abstraction of $D_{c}$.

### 7 Data Abstraction

In this section, we present a general methodology for data abstraction, strongly inspired by the notion of abstract interpretation [9]. Since in this case we do not deal with compositionality and modularization, we use a slightly simpler FDS model, in which system variables are not classified into $W$ and $\Theta$.

Let $D = (V, \Theta, \rho, \Sigma, C)$ be an FDS, and $\Sigma$ denote the set of states of $D$, the concrete states. Let $\alpha : \Sigma \mapsto \Sigma_{A}$ be a mapping of concrete states into abstract states. We say that $\alpha$ is a finitary abstraction mapping, if $\Sigma_{A}$ is a finite set. The strategy of verification by data abstraction can be summarized as follows:

**Verification by Data Abstraction**

1. Define a finitary abstraction mapping $\alpha$ to abstract the (possibly infinite) concrete FDS $D$ into a finite, abstract FDS $D^{\alpha}$.
2. Abstract the concrete temporal property $\psi$ into a finitary abstract temporal property $\psi^{\alpha}$.
3. Verify $D^{\alpha} \models \psi^{\alpha}$.
4. Infer $D \models \psi$.

An implementation of this general strategy which specifies a recipe for defining the abstractions $D^{\alpha}$ and $\psi^{\alpha}$ for a given $\alpha$ is called a data abstraction method.

A data abstraction method is said to be safe (equivalently, sound) if, for every FDS $D$, temporal formula $\psi$, and a state abstraction mapping $\alpha$ (not necessarily finitary), $\models \psi^{\alpha}$ implies $\models \psi$, and $D^{\alpha} \models \psi^{\alpha}$ implies $D \models \psi$.

#### 7.1 Safe Abstraction of Temporal Formulas

To provide a syntactic representation of the abstraction mapping, we assume a set of abstract variables $V_{A}$ and a set of expressions $E^{\alpha}$, such that the equality $V_{A} = E^{\alpha}(V)$ syntactically represents the semantic mapping $\alpha$.

Let $p(V)$ be an assertion. We wish to define the abstraction $p^{\alpha}(V_{A})$ such that $\models p^{\alpha}(V_{A})$ implies $\models p(V)$. We introduce the operator $\alpha^{-}$, defined by

$$\alpha^{-}(p(V)) : \forall V \left( V_{A} = E^{\alpha}(V) \to p(V) \right).$$

The assertion $\alpha^{-}(p)$ holds for an abstract state $S \in \Sigma_{A}$ iff the assertion $p$ holds for all concrete states $s \in \Sigma$ such that $s \in \alpha^{-1}(S)$, i.e., all states $s$ such that $s = \alpha(s)$. Alternatively, $\alpha^{-}(p)$ is the largest set of states $X \subset \Sigma_{A}$ such that $\alpha^{-1}(X) \subseteq \|p\|$, where $\|p\|$ represents the set of states which satisfy the assertion $p$. If $\alpha^{-}(p)$ is valid, then $\|\alpha^{-}(p)\| = \Sigma_{A}$ implying $\alpha^{-1}(\|\alpha^{-}(p)\|) = \Sigma_{A}$ which, by the above inclusion, leads to $\|p\| = \Sigma$ establishing the validity of $p$. 

For complex formulas, we have to consider assertions which are nested within an odd number of negations. To abstract an assertion under such a context, we define the operator \( \alpha^+ \), dual to \( \alpha^- \), as follows

\[
\alpha^+(p(V)) := \exists V \left(V^\wedge = \xi^a(V) \land p(V)\right).
\]

The assertion \( \alpha^+(p) \) holds for an abstract state \( S \in \Sigma^a \) if the assertion \( p \) holds for some concrete state \( s \in \Sigma \) such that \( s \in \alpha^+(S) \), i.e., some state \( s \) such that \( S = \alpha(s) \). Alternatively, \( \alpha^+(p) \) is the smallest set \( X \subseteq \Sigma^a \) such that \( |\phi| \subseteq \alpha^{-1}(X) \).

An abstraction \( \alpha \) is said to be precise with respect to an assertion \( p \) if \( \alpha^+(p) \sim \alpha^-(p) \). A sufficient condition for \( \alpha \) to be precise w.r.t. \( p \) is that the abstract variables include a boolean variable \( B \), with the definition \( B \equiv p \).

Having defined the abstractions \( \alpha^- \) and \( \alpha^+ \) which operate on assertions, we lift them to the abstractions \( \alpha_x^- \) and \( \alpha_x^+ \), which can be applied to temporal formulas. These temporal abstractions are defined inductively, as presented in Fig. 8.

We respectively refer to \( \alpha^-_x(p) \) and \( \alpha^+_x(p) \) as the universal (or contracting) and existential (or expanding) abstraction of the formula \( p \).

Note that equivalent temporal formulas may have different abstractions. For example, the contracting abstractions of the equivalent formulas

\[
p \lor (q \lor \Diamond r) \quad \text{and} \quad (p \lor q) \lor \Diamond r,
\]

where \( p, q, \) and \( r \) are assertions (state formulas) are respectively given by the formulas

\[
\alpha^-(p) \lor \alpha^-(q) \lor \Diamond \alpha^-(r) \quad \text{and} \quad \alpha^-(p \lor q) \lor \Diamond \alpha^-(r),
\]

which may be inequivalent. Similarly, the respective abstractions of

\[
p \land (q \land \Box \top) \quad \text{and} \quad p \land q
\]

are

\[
\alpha^+(p) \land \alpha^+(q) \quad \text{and} \quad \alpha^+(p \land q).
\]

**Claim 4.** Let \( \psi \) be a temporal formula and \( \alpha \) be an abstraction mapping. Then

\[
\models \alpha^-_x(\psi) \quad \text{implies} \quad \models \psi \quad \text{and} \quad \models \psi \quad \text{implies} \quad \models \alpha^+_x(\psi)
\]

The proof of this claim appears in [18].

In the following sections, we denote by \( \psi^\alpha \) the contracting abstraction \( \alpha^-_x(\psi) \) of the temporal formula \( \psi \).

### 7.2 Safe Abstraction of FDS’s

In the previous subsection, we established that the abstraction of the temporal formula \( \psi \) into \( \psi^\alpha = \alpha^-_x(\psi) \) is safe (equivalently sound) in the sense that if \( \psi^\alpha \) is valid, then so is \( \psi \).

Here we will establish sufficient conditions for the joint abstraction of the FDS \( D \) and the temporal formula \( \psi \) to be safe (sound) in the sense that \( D^\alpha \models \psi^\alpha \) implies \( D \models \psi \). To do so, we reduce the problem of the safe joint abstraction of an FDS and a temporal property into the problem of safe abstraction of a single temporal property, a problem that has been solved in the preceding subsection.

Given an FDS \( D = \langle V, \Theta, \rho, J, \xi, C \rangle \), there exists a temporal formula \( \text{Sem}(D) \), called the temporal semantics of \( D \) [30], such that, for every infinite state sequence \( \sigma \), it holds that \( \sigma \models \text{Sem}(D) \) iff \( \sigma \in \text{Comp}(D) \). The temporal semantics of an FDS \( D \) is given by

\[
\text{Sem}(D) := \left( \Theta(V) \land \Box \rho(V) \lor \Box V \land \bigwedge_{J \in J} \Box (\rho(J(X)) \land \bigwedge_{(p,q) \in C} (\Box \rho(V) \to \Box \Box q(V)) \right),
\]

where we use the temporal expression \( \Box V \) to denote the next values of the system variables \( V \). Given a verification problem \( D \models \psi \), we construct the temporal formula

\[
\text{Ver}(D, \psi) := \text{Sem}(D) \rightarrow \psi.
\]

It is not difficult to establish that \( D \models \psi \) iff \( \text{Ver}(D, \psi) \) is valid.

Applying a safe \( \alpha \)-abstraction to \( \text{Ver}(D, \psi) \), we obtain

\[
\alpha^-_x(\text{Ver}(D, \psi)) = \left( \alpha^+(\Theta) \land \Box \alpha^+(\rho) \land \bigwedge_{J \in J} \Box (\alpha^+(J) \land \bigwedge_{(p,q) \in C} (\Box \alpha^+(p) \to \Box \Box (\alpha^+(q)) \right)
\]

where

\[
\alpha^+(p) := \exists V : \Box V : \left( V^\wedge = \xi^a(V) \land \Box V^\wedge = \xi^a(\Box V) \land (\rho(V) \lor \Box V) \right)
\]

Based on the way \( \alpha^-_x(\text{Ver}(D, \psi)) \) abstracts the different components of \( D \), we define the \( \alpha^- \)-abstraction version of \( D \) to be the FDS \( D^\alpha = \langle V^\wedge, \Theta^\alpha, \rho^\alpha, J^\alpha, \xi^\alpha \rangle \), where

\[
\Theta^\alpha = \alpha^+(\Theta), \quad J^\alpha = \{ \alpha^+(J) \mid J \in J \},
\]

\[
\rho^\alpha = \{ \alpha^+(p) \mid (p, q) \in C \}
\]

The following claim defines our recipe for verification by data abstraction and states its soundness (safety).

**Claim 5 (Soundness).** The abstraction method which, for a given \( \alpha \), abstracts \( \psi \) into \( \alpha^-_x(\psi) \) and abstracts \( D \) into \( D^\alpha = \{ V^\wedge, \Theta^\alpha, \rho^\alpha, J^\alpha, \xi^\alpha \} \), is safe. That is,

\[
D^\alpha \models \psi^\alpha \quad \text{implies} \quad D \models \psi.
\]

**Proof:** An immediate consequence of claim 4 and the definitions of \( D^\alpha \) and \( \psi^\alpha \). □

As an example, we consider program BAKERY-2, presented in Fig. 9.

Program BAKERY-2 is obviously an infinite-state system, since the variables \( y_1 \) and \( y_2 \) can assume arbitrarily large values.
The temporal properties we wish to establish are given by

\[
\begin{align*}
\psi_{\text{exc}} & : \Box \neg (at_\ell \land at_m) \\
\psi_{\text{acc}} & : \ Diamond (at_\ell \rightarrow \ Box \ Diamond at_\ell),
\end{align*}
\]

The safety property \(\psi_{\text{exc}}\) requires mutual exclusion, guaranteeing that the two processes never co-reside in their respective critical section at the same time. The liveness property \(\psi_{\text{acc}}\) requires accessibility for process \(P_1\), guaranteeing that, whenever \(P_1\) reaches location \(\ell_2\) it will eventually reach location \(\ell_4\).

Following [3], we define abstract boolean variables \(B_{p_1}, B_{p_2}, \ldots, B_{p_k}\), one for each atomic data formula, where the atomic data formulas for BAKERY-2 are \(y_1 = 0\), \(y_2 = 0\), and \(y_1 < y_2\). Note that the formula \(y_2 \leq y_1\) is equivalent to the negation of \(y_1 < y_2\) and needs not be included as an independent atomic formula.

The abstract system variables consist of the concrete control variables, which are left unchanged, and a set of abstract boolean variables \(B_{p_1}, B_{p_2}, \ldots, B_{p_k}\). The abstraction mapping \(\alpha\) is defined by

\[
\alpha: \{B_{p_1} = p_1, B_{p_2} = p_2, \ldots, B_{p_k} = p_k\}
\]

That is, the boolean variable \(B_{p_i}\) has the value \(true\) in the abstract state iff the assertion \(p_i\) holds at the corresponding concrete state.

It is straightforward to compute the \(\alpha\)-induced abstractions of the initial condition \(\Theta^\alpha\) and the transition relation \(\rho^\alpha\). In Fig. 10, we present program BAKERY-2 (with a capital B), the \(\alpha\)-induced abstraction of program BAKERY-2.

Since the properties we wish to verify refer only to the control variables (through the \(at_\ell\) and \(at_m\) expressions), they are not affected by the abstraction. Program BAKERY-2 is a finite-state program, and we can apply model checking to verify that it satisfies the two properties of mutual exclusion and accessibility. By Claim 5, we can infer that the original program BAKERY-2 also satisfies these two temporal properties.

7.3 Augmentation by Progress Monitors

Program BAKERY-2 is an example of successful data abstraction. However, there are cases when abstraction alone is inadequate for transforming an infinite-state system satisfying a property into a finite-state abstraction which maintain the property. In the following we illustrate the problem and the proposed solution on a simple example. For the treatment of the general case, see [18]. In Fig. 11, we present a simple looping program. The assignment at statement \(\ell_2\) assigns to \(y\) non-deterministically the values \(y + 1\) or \(y\). The property we wish to verify is that program \(\text{SUM-ADD}\) always terminates, independently of the initial value of the natural variable \(y\).

A natural abstraction for the variable \(y\) is defined by
\begin{center}
\begin{tabular}{|c|c|}
\hline
local $B_{y1}=0, B_{y2}=0, B_{y1} < y_2$ & boolean initially $B_{y1} = 0, B_{y2} = 0, 1, B_{y1} < y_2 = 0$\\
\hline
$\ell_0 : \text{loop forever do}$ & $m_0 : \text{loop forever do}$
\hline
$\ell_1 : \text{NonCritical}$ & $m_1 : \text{NonCritical}$
\hline
$\ell_2 : (B_{y1}=0, B_{y1} < y_2) \Rightarrow (0,0)$ & $m_2 : (B_{y2}=0, B_{y1} < y_2) \Rightarrow (0,1)$
\hline
$\ell_3 : \text{wait} B_{y2} = 0 \lor B_{y1} < y_2$ & $m_3 : \text{wait} B_{y1} = 0 \lor -B_{y1} < y_2$
\hline
$\ell_4 : \text{Critical}$ & $m_4 : \text{Critical}$
\hline
$\ell_5 : (B_{y1}=0, B_{y1} < y_2) \Rightarrow (1, -B_{y2}=0)$ & $m_5 : (B_{y2}=0, B_{y1} < y_2) \Rightarrow (1,0)$
\hline
\end{tabular}
\end{center}

\textbf{Fig. 10.} Program \textsc{Bakery-2}: the Bakery algorithm for two processes.

\begin{center}
\begin{tabular}{|c|c|}
\hline
$y : \text{natural}$ & \\
\hline
$\ell_0 : \text{while } y > 1 \text{ do}$ & \\
$\ell_1 : y := y - 2$ & \\
$\ell_2 : y := \{y + 1, y\}$ & \\
$\ell_3 : \text{skip}$ & \\
$\ell_4 :$ & \\
\hline
\end{tabular}
\end{center}

\textbf{Fig. 11.} Program \textsc{sub-add}.

\begin{center}
\begin{tabular}{|c|c|}
\hline
$Y : \{\text{zero, one, large}\}$ & \\
\hline
$\ell_0 : \text{while } Y = \text{large do}$ & \\
$\ell_1 : Y := \text{sub2}(Y)$ & \\
$\ell_2 : Y := \{\text{add1}(Y), Y\}$ & \\
$\ell_3 : \text{skip}$ & \\
$\ell_4 :$ & \\
\hline
\end{tabular}
\end{center}

\textbf{Fig. 12.} Program \textsc{sub-add-abs-1} abstracting program \textsc{sub-add}.

\[
Y = \begin{cases} 
\text{if } y = 0 \text{ then zero} \\
\text{else if } y = 1 \text{ then one} \\
\text{else large}
\end{cases}
\]

where $y$ is abstracted into the three-valued domain
\[
\{\text{zero, one, large}\}.
\]

However, applying this abstraction yields the abstract program \textsc{sub-add-abs-1}, presented in Fig. 12, where the abstract functions \text{sub2} and \text{add1} are defined by

\[
\text{sub2}(Y) = \begin{cases} 
\text{if } Y = \{\text{zero, one}\} \text{ then zero} \\
\text{else } \{\text{zero, one, large}\}
\end{cases}
\]

\[
\text{add1}(Y) = \begin{cases} 
\text{if } Y = \text{zero} \text{ then one} \\
\text{else large}
\end{cases}
\]

Unfortunately, program \textsc{sub-add-abs-1} needs not terminate, because the function \text{sub2} can always choose to yield \text{large} as a result.

Termination of programs like program \textsc{sub-add} can always be established by identification of a \textit{progress measure} that never increases and sometimes is guaranteed to decrease. In this case, for example, we can use the progress measure $\delta : y + at_\ell \leq 2$ which never increases and always decreases on the execution of statement $\ell_1$. To obtain a working abstraction, we first compose program \textsc{sub-add} with an additional module, to which we refer as the \textit{progress monitor} for the progress measure $\delta$, as shown in Fig. 13.

The construct \textbf{always do} appearing in monitor $M_8$ means that the assignment which is the body of this construct is executed at every step. The comparison function $\text{comp}(\delta, \delta')$ is defined by

\[
\text{comp}(\delta, \delta') = \begin{cases} 
\text{if } \delta < \delta' \text{ then 1} \\
\text{else if } \delta = \delta' \text{ then 0} \\
\text{else -1.}
\end{cases}
\]

Note that the expressions on the right-hand-side of the assignments in the monitor allow references to the \textit{new} values of $\delta$ as computed in the same step by the monitored program.

The presentation of the monitor module $M_8$ in Fig. 13 is only for illustration purposes. The precise definition of this module is given by the following FDS:

\[
V = V_D \cup \{\text{inc} : \{-1, 0, 1\}\}
\]

\[
\Theta : \text{t}
\]

\[
\rho : \text{inc} = \text{comp}(\delta, \delta')
\]

\[
\mathcal{I} : \emptyset, \quad C : \{(\text{inc} < 0, \text{inc} > 0)\}
\]

where $V_D$ are the system variables of the monitored FDS $D$. Thus, at every step of the computation, module $M_8$ compares the new value of $\delta$ ($\delta'$) with the current value, and sets variable \text{inc} to -1, 0, or 1, according to whether the value of $\delta$ has decreased, stayed the same, or increased, respectively. This FDS has no justice requirements but has the single compassion requirement ($\text{inc} < 0, \text{inc} > 0$) stating that $\delta$ cannot decrease infinitely many times without also increasing infinitely many times. This requirement is a direct consequence of the fact that $\delta$ ranges over the well-founded domain of the natural numbers, which does not allow an infinitely decreasing sequence.
Fig. 13. Program SUB-ADD composed with a monitor.

Fig. 14. A sequential equivalent of the monitored program.

Fig. 15. Abstracted version of the monitored Program SUB-ADD-ABS-2.

It is possible to represent this composition as (almost) equivalent to the sequential program presented in Fig. 14, where we have conjoined the repeated assignment of module $M_6$ with every assignment of process SUB-ADD. The “almost” qualification admits that we did not conjoin this assignment with the transition associated with location $\ell_0$ which tests the value of $y$ and decides when to terminate. In a fully formal treatment of this example, the assignment will also be conjoined to this testing transition.

The abstraction of the program of Fig. 14 will abstract $y$ into a variable $Y$ ranging over $\{\text{zero, one, large}\}$. The variable $\text{inc}$, ranging over the finite domain $\{-1,0,1\}$, is not abstracted. The resulting abstraction is presented in Fig. 15.

The program SUB-ADD-ABS-2 (Fig. 15) differs from program SUB-ADD-ABS-1 (Fig. 12) by the additional compassion requirement $(\text{inc} < 0, \text{inc} > 0)$. However, it is this additional requirement which forces program SUB-ADD-ABS-2 to terminate. This is because a run in which $\text{sub1}$ always yields large as a result is a run in which $\text{inc}$ is negative infinitely many times (on every visit to $\ell_1$) and is never positive beyond the first state. The fact that SUB-ADD-ABS-2 always terminates can now be successfully model-checked.

The extension to the case that the progress measure ranges not over the Naturals but over lexicographic tuples of naturals is straightforward.

7.4 The Data Abstraction Method is Complete

In a separate work [18], concentrating on the data abstraction method, we have established that this method is relatively complete. Completeness in this context means that for every (possibly infinite) system $D$ and a temporal property $\psi$, such that $D \models \psi$, there exists a (progress) monitor $M_6$ whose composition with $D$ does not constrain the computations of $D$, and a finitary state abstraction mapping $\alpha$, such that $(D || M_6)^\alpha \models \psi^\alpha$. This implies that whenever $\psi$ is a property valid for $D$, we can apply the method of data abstraction described in this section in order to formally verify that $\psi$ is $D$-valid.

8 Conclusions

The paper presented two central techniques for reducing a big verification task into several smaller ones. These techniques are specially impressive when they reduce an infinite-state system into a finite-state one.

The first technique is based on control abstraction and reduces an unbounded environment for a single module into an abstract environment model which preserves the relevant features of the environment. Often, the unbounded environment represents a set of brother processes and the derived abstract model represents a network-invariant which is independent of the size of this set.

The second technique is that of data abstraction in which variables ranging over infinite domains are abstracted into variables ranging over finite domains. The method presents a general recipe for computing such an abstraction for every user-provided state mapping, such that the abstraction preserves all counter-examples to any temporal property. This means that if the property has been verified to be valid on the abstract level, its concrete version is guaranteed to be valid (no false positives).
An important feature of our formulations of these two methods is that they are not restricted to the verification of safety properties as are many previous formulations of similar approaches, but deal quite effectively with liveness properties, in fact with all temporally expressible properties. On the system side, they take full account of both weak (justice) and strong (compassion) fairness requirements.

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References