Optimal Stateless Model Checking under the Release-Acquire Semantics

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We present a framework for the efficient application of stateless model checking (SMC) to concurrent programs running under the Release-Acquire (RA) fragment of the C/C++11 memory model. Our approach is based on exploring the possible program orders, which define the order in which instructions of a thread are executed, and read-from relations, which specify how reads obtain their values from writes. This is in contrast to previous approaches, which also explore the possible coherence orders, i.e., orderings between conflicting writes. Since unexpected test results such as program crashes or assertion violations depend only on the read-from relation, we avoid a potentially significant source of redundancy. Our framework is based on a novel technique for determining whether a particular read-from relation is feasible under the RA semantics. We define an SMC algorithm which is provably optimal in the sense that it explores each program order and read-from relation exactly once. This optimality result is strictly stronger than previous analogous optimality results, which also take coherence order into account. We have implemented our framework in the tool Tracer. Experiments show that Tracer can be significantly faster than state-of-the-art tools that can handle the RA semantics.

CCS Concepts: • Software and its engineering → Formal software verification;

Additional Key Words and Phrases: software model checking, weak memory models, C/C++11, Release-Acquire

1 INTRODUCTION

Ensuring correctness of concurrent programs is difficult since one must consider all the different ways in which threads can interact. A successful technique for finding concurrency bugs (i.e., defects that arise only under some thread schedulings), and for verifying their absence, is stateless model checking (SMC) [Godefroid 1997]. Given a terminating program, which may be annotated with assertions, SMC systematically explores the set of all thread schedulings that are possible during runs of this program. A special runtime scheduler drives the SMC exploration by making decisions on scheduling whenever such choices may affect the interaction between threads; so that the exploration covers all possible executions and detects any unexpected program results, program crashes, or assertion violations. The technique is entirely automatic, has no false positives, does not consume excessive memory, and can quickly reproduce the concurrency bugs it detects.

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2475-1421/2018/11-ART135
https://doi.org/10.1145/3276505
Initially:

\[
\begin{align*}
th_1: & \quad x := 1 \\
& \quad a := x \\
\quad \text{then}\end{align*}
\]

\[
\begin{align*}
th_2: & \quad x := 2 \\
& \quad b := x \\
\quad \text{then}\end{align*}
\]

(a)

(b) A simple concurrent program and (b) Shasha-Snir traces and weak traces.

SMC has been implemented in tools, such as VeriSoft [Godefroid 2005], CDSChecker [Demsky and Lam 2015; Norris and Dmisky 2016], CHESS [Musuvathi et al. 2008], Concurrorg [Christakis et al. 2013], rinspect [Zhang et al. 2015], and Nidhugg [Abdulla et al. 2015a], and successfully applied to realistic concurrent programs [Godefroid et al. 1998; Kokologiannakis and Sagonas 2017].

SMC faces the problem that the number of possible thread schedulings grows exponentially with the length of program execution, and must therefore be equipped with techniques to reduce the number of explored executions. The most prominent one is partial order reduction [Clarke et al. 1999; Godefroid 1996; Peled 1993; Valmari 1990], adapted to SMC as dynamic partial order reduction (DPOR). DPOR was first developed for concurrent programs that execute under the standard model of Sequential Consistency (SC) [Abdulla et al. 2014; Flanagan and Godefroid 2005; Sen and Agha 2007]. In recent years, DPOR has been adapted to hardware-induced weak memory models, such as TSO and PSO [Abdulla et al. 2015a; Zhang et al. 2015], and language-level concurrency models, such as the C/C++11 memory model [Kokologiannakis et al. 2018; Norris and Dmisky 2016]. DPOR is based on the observation that two executions can be regarded as equivalent if they induce the same ordering between conflicting statement executions (called events), and that it is therefore sufficient to explore at least one execution in each equivalence class. Under SC, such equivalence classes are called Mazurkiewicz traces [Mazurkiewicz 1986]; for weak memory models, the natural generalization of Mazurkiewicz traces are called Shasha-Snir traces [Shasha and Snir 1988]. A Shasha-Snir trace characterizes an execution of a program by three relations between events; (i) po ("program order") totally orders the events of each thread, (ii) co ("coherence") totally orders the writes to each shared variable, and (iii) rf ("read-from") connects each write with the reads that read its value. Under weak memory models, the co and rf relations need not be derived from the global order in which events occur in an execution (as is the case under SC). Each particular weak memory model therefore imposes restrictions on how these relations may be combined.

As an illustration, Figure 1a shows a simple program with two threads, \( th_1 \) and \( th_2 \), that communicate through a shared variable \( x \). Each thread writes to the variable and reads from it into a local register, a resp. b. We would like to explore the possible executions of this program, e.g., to check whether the program can satisfy \( a = 2 \) and \( b = 1 \) upon termination. Under many memory models, including SC, TSO, PSO, RA, and POWER, executions of the program in Figure 1a fall into four equivalence classes, represented by the four possible Shasha-Snir traces \( \tau_1 \), \( \tau_2 \), \( \tau_3 \), and \( \tau_4 \) in
A DPOR algorithm based on Shasha-Snir traces (e.g., [Abdulla et al. 2014, 2016b]) must thus explore at least four executions. However, it is possible to reduce this number further. Namely, a closer inspection reveals that the two traces $\tau_1$ and $\tau_2$ are equivalent, in the sense that each thread goes through the same sequences of local states and computes the same results. This is because $\tau_1$ and $\tau_2$ have the same program order ($po$) and read-from ($rf$) relation. Their only difference is how writes are ordered by $co$, but this is not relevant for the computed results.

The preceding example illustrates that there is a potential for improving the efficiency of DPOR algorithms by using a weaker equivalence induced only by $po$ and $rf$. In this example, the improvement is modest (reducing the number of explored traces from four to three), but it can be significant, sometimes even exponential, for more extensive programs. Several recent DPOR techniques try to exploit the potential offered by such a weaker equivalence [Chalupa et al. 2018; Huang 2015; Huang and Huang 2016; Norris and Demsky 2016]. However, except for the minimal case of an acyclic communication graph [Chalupa et al. 2018], they are far from optimally doing this, since they may still explore a significant number of different executions with the same $rf$ relation. Therefore, the challenge remains to define a more efficient DPOR algorithm, that is optimal in the sense that it explores precisely one execution in each equivalence class induced by $po$ and $rf$.

In this paper, we present a fundamentally new approach to defining DPOR algorithms, which optimally explores only the equivalence classes defined by the program order and read-from relations. Our method is developed for the Release-Acquire (RA) fragment [Lahav et al. 2016] of the C/C++11 memory model. RA is a useful and well-behaved fragment of the C/C++11 memory model, which strikes a good balance between performance and programmability. In the RA semantics, all writes are release accesses, while all reads are acquire accesses. RA allows high-performance implementations, while still providing sufficiently strong guarantees for fundamental concurrent algorithms (such as the read-copy-update mechanism [Lahav et al. 2016]). Our DPOR algorithm is based on the above weakening of Shasha-Snir traces, called weak traces, which are defined by only the program order and read-from relations of an execution. For example, the program in Figure 1a has three weak traces, shown as $\tau_3$, $\tau_4$, and $\tau_5$. Our DPOR algorithm is provably optimal for weak traces, in the sense that it explores precisely one execution for each weak trace that is RA-consistent, i.e., that can be extended with some coherence relation that satisfies the constraints of the RA semantics.

A significant challenge for our DPOR algorithm is to efficiently determine those continuations of a currently explored trace that lead to some RA-consistent trace. E.g., for the program in Figure 1a, letting both threads read from the write of the other thread leads to RA-inconsistency. We solve this problem by defining a saturation operation, which extends a weak trace with a partial coherence relation, which contains precisely those coherence edges that must be present in any corresponding Shasha-Snir trace. During exploration, the DPOR algorithm maintains a saturated version of the currently explored weak trace, and can therefore examine precisely those weak traces that are RA-consistent, without performing useless explorations. When a read event is added, the algorithm determines the set of write events from which it can obtain its value while preserving RA-consistency, and branches into a separate continuation for each such write event. When a write event is added, the algorithm merely adds it to the trace. It can be proven that this preserves RA-consistency, and also keeps the trace saturated (a slight modification is needed for atomic read-modify-write events). The algorithm must also detect if some previous read may read from a newly added write, and then backtrack to allow the write to be performed before that read.

We prove that our DPOR algorithm does not perform any useless work, in the sense that (i) any exploration eventually leads to a terminated RA-consistent execution, i.e., the algorithm never

\[A Shasha-Snir trace also includes events that write initial values, but these can be ignored for this example.\]
blocks because it discovers that it is about to perform redundant or wasted explorations, and (ii) each RA-consistent weak trace is explored precisely once. Our saturation technique and the DPOR algorithm presented in this paper can be extended to cover atomic read-modify-write (RMW) operations and locks (space does not permit inclusion in this paper). This extension also satisfies the same strong optimality results in (i) and (ii).

In summary, we prove a stronger notion of optimality than related SMC approaches: we prove optimality concerning weak traces, whereas others only prove it w.r.t. total traces (i.e., Mazurkiewicz traces or Shasha-Snir traces); moreover, our technique can be extended to cover RMWs.

We have implemented our saturation operation and DPOR algorithm in a tool, called TRACER, and applied it to many challenging benchmarks. We compare our tool with other state-of-the-art stateless model checking tools running under the RA semantics. The experiments show that TRACER always generates optimal numbers of executions w.r.t. weak traces in all benchmarks. On many benchmarks, this number is much smaller than the ones produced by the other tools. The results also show that TRACER has better performance and scales better for more extensive programs, even in the case where it explores the same number of executions as the other tools.

2 BASIC NOTATION

We let \( \mathbb{N} \) denote the set of natural numbers. Fix a set \( A \). If \( A \) is finite then we use \(|A|\) to denote the size of \( A \). For a binary relation \( R \) on \( A \), we write \( a_1 \ [R] \ a_2 \) to denote that \( \langle a_1, a_2 \rangle \in R \). We use \( R^{-1} \) to denote the inverse of \( R \), i.e., \( a_1 \ [R^{-1}] \ a_2 \iff a_2 \ [R] \ a_1 \). We use \( R^* \) and \( R^+ \) to denote the transitive closure and the reflexive transitive closure of \( R \), respectively. We write \( a_1 \ [R]^+ a_2 \) and \( a_1 \ [R]^* a_2 \) to denote that \( a_1 \ [R^+] \ a_2 \) resp. \( a_1 \ [R^*] \ a_2 \). We say that \( R \) is a partial order if it is irreflexive (i.e., \( \neg(a \ [R] a) \) for all \( a \in A \)) and transitive (i.e., if \( a_1 \ [R] a_2 \) and \( a_2 \ [R] a_3 \) then \( a_1 \ [R] a_3 \) for all \( a_1, a_2, a_3 \in A \)). We say that \( R \) is total if, for all \( a_1, a_2 \in A \), either \( a_1 \ [R] a_2 \) or \( a_2 \ [R] a_1 \). We use \( \text{acyclic}(R) \) to denote that \( R \) is acyclic, i.e., there is no \( a \in A \) such that \( a \ [R]^+ a \) for all \( a \in A \). For a set \( B \subseteq A \), we define \( R^1_{|B|} := R \cap (B \times B) \), i.e., it is the restriction of \( R \) to \( B \). For binary relations \( R_1 \) and \( R_2 \) on \( A \), we use \( R_1; R_2 \) to denote the composition of \( R_1 \) and \( R_2 \), i.e., \( a_1 \ [R_1; R_2] \ a_2 \iff \exists a_3 \in A \) such that \( a_1 \ [R_1] \ a_3 \) and \( a_2 \ [R_2] \ a_3 \). For sets \( A \) and \( B \), we use \( f : A \to B \) to denote that \( f \) is a (possibly partial) function from \( A \) to \( B \). We use \( f[a \leftarrow a'] \) to denote the function \( f' \) such that \( f'(a) = a' \) and \( f'(b) = f(b) \) if \( b \neq a \). We use \( A^* \) to denote the set of finite words over \( A \), and use \( e \) to denote the empty word. We use \(|w|\) to denote the length of \( w \), use \( w[i] \) to denote the \( i^{th} \) element of \( w \), and use \( \text{last}(w) \) for \( w[|w|] \). For words \( w_1, w_2 \in A^* \), we use \( w_1 \bullet w_2 \) to denote the concatenation \( w_1 \) and \( w_2 \).

3 MODEL

Programs. We consider a program \( \mathcal{P} \) consisting of a finite set \( \mathcal{T} \) of threads (processes) that share a finite set \( \mathbb{V} \) of (shared) variables, ranging over a domain \( \mathbb{V} \) of values that includes a special value \( 0 \). A thread has a finite set of local registers that store values from \( \mathbb{V} \). Each thread runs a deterministic code, built in a standard way from expressions and atomic commands, using standard control flow constructs (sequential composition, selection, and bounded loop constructs). Throughout the paper, we use \( x, y \) for shared variables, \( a, b, c \) for registers, and \( e \) for expressions. Global statements are either write \( x := e \) to a shared variable or read \( a := x \) from a shared variable. Local statements only access and affect the local state of the thread and include assignments \( a := e \) to registers and conditional control flow constructs. Note that expressions do not contain shared variables, implying that a statement accesses at most one shared variable. For readability reason, we do not consider atomic read-modify-write (RMW) operations and fences. The local state of a thread \( th \in \mathcal{T} \) is defined as usual by its program counter and the contents of its registers.

Proc. ACM Program. Lang., Vol. 2, No. OOPSLA, Article 135. Publication date: November 2018.
**Configurations.** A configuration (global state) of $\mathcal{P}$ is made up of the local states of all the threads. Note that the values of the shared variables are not part of a configuration. The reason is that, under the RA semantics (and many other weak memory models), different threads may have different “local views” of the shared variables, i.e., they may see different values of the shared variables at a given point during the program execution. Therefore, it is not possible to assign a unique value to a shared variable. In existing operational semantics for such weak memory models including the RA semantics (e.g., [Kaiser et al. 2017]), a write instruction does not explicitly modify the values of the shared variables. Instead, a write instruction is added to a “pool” of write instructions that have been issued by the threads. A read instruction can fetch its value from a set of available write instructions in the pool and update its local view accordingly. The sets of write instructions that are available to reading threads depend on the particular memory model.

Following Lahav et al. [2016], we define an operational semantics for $\mathcal{P}$ as a labeled transition relation over configurations. Each transition corresponds to one thread performing a local or global statement. A transition between two configurations $\gamma$ and $\gamma'$ is of form $\gamma \xrightarrow{\ell} \gamma'$, where the label $\ell$ describes the interaction with shared variables. The label $\ell$ is one of three forms: (i) $\langle th, \epsilon \rangle$, indicating a local statement performed by thread $th$, which updates only the local state of $th$; (ii) $\langle th, w, x, v \rangle$, indicating a write of the value $v$ to the variable $x$ by the thread $th$, which also updates the program counter of $th$; and (iii) $\langle th, R, x, v \rangle$ indicating a read of $v$ from $x$ by the thread $th$ into some register, while also updating the program counter of $th$. Observe that since the shared variables are not part of a configuration, the threads do not interact with each other in the transition relation. In particular, there is no constraint on the values that are used in transitions corresponding to read statements. This will obviously make illegal program behaviors possible. We remedy this problem by associating runs with so-called traces, which (among other things) represent how reads obtain their values from writes. A particular memory semantics (such as the RA semantics) is formulated by imposing restrictions on these traces, thereby also restricting the possible runs that are associated with them.

Since local statements are not visible to other threads, we will not represent them explicitly in the transition relation considered in our DPOR algorithm. Instead, we let each transition represent the combined effect of some finite sequence of local statements by a thread followed by a global statement by the same thread. More precisely, for configurations $\gamma$ and $\gamma'$ and a label $\ell$ which is either of the form $\langle th, w, x, v \rangle$ or of the form $\langle th, R, x, v \rangle$, we let $\gamma \xrightarrow{\ell} \gamma'$ denote that we can reach $\gamma'$ from $\gamma$ by performing a sequence of transitions labeled with $\langle th, \epsilon \rangle$ followed by a transition labeled with $\ell$. Defining the relation $\rightarrow$ in this manner ensures that we take the effect of local statements into account while avoiding consideration of interleavings of local statements of different threads in the analysis. Such optimization is common in tools (e.g., Verisoft [Godefroid 1997]).

We introduce some extra notation. We use $\gamma \xrightarrow{(th,R,x,v)\ast} \gamma'$ to denote that $\gamma \xrightarrow{(th,R,x,v)} \gamma'$ for some value $v \in \mathbb{V}$ and configuration $\gamma'$. We use $\gamma \rightarrow \gamma'$ to denote that $\gamma \xrightarrow{\ell} \gamma'$ for some $\ell$ and define $\text{succ}(\gamma) := \{\gamma' \mid \gamma \rightarrow \gamma'\}$, i.e., it is the set of successors of $\gamma$ w.r.t. $\rightarrow$.

A configuration $\gamma$ is said to be terminal if $\text{succ}(\gamma) = \emptyset$, i.e., no thread can execute a global statement from $\gamma$. A run $\rho$ from $\gamma$ is a sequence $\gamma_0 \xrightarrow{\ell_0} \gamma_1 \xrightarrow{\ell_1} \cdots \xrightarrow{\ell_n} \gamma_n$ such that $\gamma_0 = \gamma$. We say that $\rho$ is terminated if $\gamma_n$ is terminal. We let $\text{Runs}(\gamma)$ denote the set of runs from $\gamma$.

**Events.** An event corresponds to a particular execution of a statement in a run of $\mathcal{P}$. A write event $e$ is a tuple $(id, th, w, x, v)$, where $id \in \mathbb{N}$ is an event identifier, $th \in \mathcal{T}$ is a thread, $x \in \mathcal{X}$ is a variable, and $v \in \mathbb{V}$ is a value. This event corresponds to thread $th$ writing the value $v$ to variable $x$. The identifier $id$ denotes that $th$ has executed $id-1$ events before $e$ in the corresponding run. A read
event $e$ is a tuple $⟨id, th, R, x⟩$, where $id$, $th$, and $x$ are as for a write event. This event corresponds to thread $th$ reading some value to $x$. Note that a read event $e$ does not specify the particular value it reads. This value will be defined in a trace by specifying a write event from which $e$ fetches its value. For an event $e$ of form $⟨id, th, W, x, v⟩$ or $⟨id, th, R, x⟩$, we define $e.id := id$, $e.thread := th$, $e.type := t$ where $t \in \{W, R\}$, $e.var := x$, and $e.val := v$ (the latter is not defined for a read event). For each variable $x \in X$, we assume a special write event $\text{init}_x = ⟨−−−, −−−, W, x, 0⟩$, called the initializer event for $x$. This event is not performed by any of the threads in $T$, and writes the value 0 to $x$. We define $E_{\text{init}} := \{\text{init}_x \mid x \in X\}$ as the set of initializer events.

If $E$ is a set of events, we define subsets of $E$ characterized by particular attributes of its events. For instance, for a thread $th$, we let $E_{th}^{\text{init}}$ denote $\{e \in E \mid e.thread = th \land e.type = W\}$.

**Traces.** A trace $τ$ is a tuple $⟨E, po, rf, co⟩$, where $E$ is a set of events which includes the set $E_{\text{init}}$ of initializer events, and where $po$ (program order), $rf$ (read-from), and $co$ (coherence order) are binary relations on $E$ that satisfy:

$$
e \ [po] e' \text{ if } e.thread = e'.thread \text{ and } e.id < e'.id, \text{ i.e., } po \text{ totally orders the events of each individual thread. As mentioned above, each event corresponds to an execution of a program statement. The program order then reflects the order in which the statements of a given thread are executed. Note that } po \text{ does not relate the initializer events.}
$$

$$
e \ [rf] e' \text{ if } e \text{ is a write event and } e' \text{ is a read event on the same variable, which obtains its value from } e. \text{ The inverse of relation rf, denoted } rf^{-1}, \text{ must be a total function on } E^R. \text{ We sometimes view } rf \text{ as the union of a read-from relation } rf^X \text{ for each variable } x \in X.
$$

$co$ is a union $co = \bigcup_{x \in X} co^x$, where $co^x$ is a relation on $E^x$ (including $\text{init}_x$), subject to the constraint that $\text{init}_x$ is before all other write events in $E^x$. Thus $co$ does not relate write events on different variables. The relation $co^x$ reflects how the threads view the order on the write events on $x$. If $e_1 \ [co^x] e_2$ then all threads share the view that $e_1$ has occurred before $e_2$.

Note that (in contrast to $po$), each $co^x$ can be an arbitrary relation; it need not even be transitive. We say that a trace $⟨E, po, rf, co⟩$ is total if $co^x$ is a strict total order on $E^x$ for each $x \in X$. We use $\text{total}(τ)$ to denote that $τ$ is total. A total trace is also called a Shasha-Snir trace. We sometimes use partial trace to denote an arbitrary trace, when we want to emphasize that it need not be total.

As depicted in Figure 1b, we can view $τ = ⟨E, po, rf, co⟩$ as a graph whose nodes are $E$ and whose edges are defined by the relations $po$, $rf$, and $co$. We define $|τ| := |E|$, i.e., it is the number of events in $τ$. We define the empty trace $τ_0 := ⟨E_{\text{init}}, 0, 0, 0⟩$, i.e., it contains only the initializer events, and all the relations are empty.

**Associating Traces with Runs.** We can now define when a trace can be associated with a run. Consider a run $ρ$ of form $γ_0 \overset{ℓ_1}{\longrightarrow} \ldots \overset{ℓ_n}{\longrightarrow} γ_n$, where $ℓ_i = ⟨th_i, t_i, x_i, v_i⟩$, and let $τ = ⟨E, po, rf, co⟩$ be a trace. We write $ρ \models τ$ to denote that the following conditions are satisfied:

- $E \setminus E_{\text{init}} = \{e_1, \ldots, e_n\}$, i.e., each non-initializer event corresponds exactly to one label in $ρ$.
- If $ℓ_i = ⟨th_i, W, x_i, v_i⟩$, then $e_i = ⟨id_i, th_i, W, x_i, v_i⟩$, and if $ℓ_i = ⟨th_i, R, x_i, v_i⟩$, then $e_i = ⟨id_i, th_i, R, x_i⟩$. An event and the corresponding label perform identical operations (write or read) on the same variables. In the case of a write, they also agree on the written value.
- $id_i = \{j \mid (1 \leq j \leq i) \land (th_j = th_i)\}$. The identifier of an event shows how it is ordered relative to the other events performed by the same thread.
- If $e_i \ [rf] e_j$, then $x_1 = x_j$ and $v_1 = v_j$. A read event fetches its value from an event that writes the same value to the same variable.
- If $\text{init}_x \ [rf] e_i$, then $v_1 = 0$, i.e., $e_i$ reads the initial value of $x$ which is assumed to be 0.
Initially: \( x = y = 0 \)

\[
\begin{array}{c|c}
\text{th}_1 & \text{th}_2 \\
\hline
x := 1 & y := 1 \\
y := 2 & x := 2
\end{array}
\]

(a)

\[
\begin{array}{c|c}
\text{th}_1 & \text{th}_2 \\
\hline
x := 1 & a := y \\
y := 1 & x := 2
\end{array}
\]

(b)

Fig. 2. (a) The program 2+2W and (b) a Shasha-Snir trace.

Initially: \( x = y = 0 \)

\[
\begin{array}{c|c}
\text{th}_1 & \text{th}_2 \\
\hline
x := 1 & y := 1 \\
y := 2 & x := 2
\end{array}
\]

(a)

\[
\begin{array}{c|c}
\text{th}_1 & \text{th}_2 \\
\hline
x := 1 & a := y \\
y := 1 & x := 2
\end{array}
\]

(b)

Fig. 3. (a) The program S and (b) a Shasha-Snir trace.

**Release-Acquire Semantics.** Following Lahav et al. [2016], we define the Release-Acquire semantics by defining the set of runs whose associated total traces do not contain certain forbidden cycles.

Given a trace \( \tau = (E, po, rf, co) \), we define the derived relation \( fr \) (from-read) by \( fr := \cup_{x \in X} fr^x \), where \( fr^x := (rf^x)^{-1}; co^x \). Intuitively, if \( e \in fr^x \) \( e' \) then the write \( e' \) overwrites the value read by the read \( e \) (since \( e' \) is coherence-order-after the write event from which \( e \) gets its value).

**Definition 3.1.** For a trace \( \tau \), let \( \models \tau \models RA \) denote that the relation \( po \cup rf \cup co^x \cup fr^x \) is acyclic for each \( x \in X \). We define \( \left\llbracket y \right\rrbracket_{RA}^{\tau} := \{ \tau \mid \exists \rho \in \text{Runs}(\gamma). \models \tau \land \text{total}(\tau) \land \models RA \} \), i.e., \( \left\llbracket y \right\rrbracket_{RA}^{\tau} \) is the set of total traces generated under RA from a given configuration \( \gamma \).

To illustrate the RA semantics, Figure 2a shows a simple program, known as 2+2W [Alglave et al. 2014], with two threads \( th_1 \) and \( th_2 \) that communicate through two shared variables \( x \) and \( y \). Thread \( th_1 \) writes 1 to \( x \) and then 2 to \( y \). Symmetrically, \( th_2 \) writes 1 to \( y \) and then 2 to \( x \). We would like to check whether the writes \( x := 2 \) and \( y := 2 \) can be placed before the writes \( x := 1 \) and \( y := 1 \) in the corresponding coherence order relation (co). Figure 2b gives the corresponding trace \( \tau_1 \). To improve readability, we use a simplified notation for events. More precisely, we represent an event in a trace by the corresponding program instruction. For instance, we write \( x := 1 \) instead of \( \langle 1, th_1, w, x, 1 \rangle \). We observe that the read-from relation \( rf \) and the from-read relation \( fr \) are empty in \( \tau_1 \). Since the relations \( po \cup rf \cup co^x \cup fr^x \) and \( po \cup rf \cup co^y \cup fr^y \) are acyclic, it follows by Definition 3.1, that \( \models \tau_1 \models RA \).

To see the role of read-from relations in the RA semantics, Figure 3a gives another program, known as S [Alglave et al. 2014]. In a similar manner to 2+2W, the program S has two threads \( th_1 \) and \( th_2 \) that communicate through two shared variables \( x \) and \( y \). Thread \( th_1 \) writes 1 to both \( x \) and \( y \). Thread \( th_2 \) reads from \( y \) and stores the value to a local register \( a \) and then writes \( 2 \) to \( x \). We would like to check whether it is possible to place the write \( x := 2 \) before the write \( x := 1 \) in coherence order (co) and make \( a := y \) read from the write \( y := 1 \) in the read-from relation (rf).
Figure 2b gives the corresponding trace $\tau_2$. Since the relation $po \cup rf \cup co^x \cup fr^x$ is cyclic, it follows by Definition 3.1, that $\tau_2 \not RA$.

Since the result computed by a run is uniquely determined by its associated trace, we can analyze a concurrent program under the RA semantics by exploring runs and their associated traces until all traces in $\llbracket Y \rrbracket_{RA}^{Total}$ have been generated. One can even define an algorithm which is optimal in that it explores each trace in $\llbracket Y \rrbracket_{RA}^{Total}$ exactly once [Kokogiannakis et al. 2018]. However, it is possible to be more efficient than this, by exploiting the observation that the results computed in a run, including, e.g., the outcome of assert statements, depend only on the program order ($po$) and read-from ($rf$) relations, irrespective of the $co$ relation. In fact, the sequence of local states of a thread and the values it writes to shared variables are affected only by the sequence of values it reads from shared variables. Hence two runs, whose associated traces have the same $po$ and $rf$ relations, can be regarded as equivalent. An algorithm based on the set $\llbracket Y \rrbracket_{RA}^{Total}$ will in general be redundant, in that it explores several traces that are equivalent in this sense.

**Weak Traces.** The above observation suggests to base the analysis on a weaker notion of trace. Define a weak trace to be a trace whose $co$ relation is empty. For a trace $\tau = \langle E, po, rf, co \rangle$, we define its weakening by weak $(\tau) := \langle E, po, rf, \emptyset \rangle$, obtained by removing all coherence edges. We define the set $\llbracket Y \rrbracket_{RA}^{Weak} := \{ weak (\tau) \mid \tau \in \llbracket Y \rrbracket_{RA}^{Total} \}$. In other words, $\llbracket Y \rrbracket_{RA}^{Weak}$ is the set of weakenings of the traces in the total semantics. Let us introduce a term for such weakenings. For traces $\tau = \langle E, po, rf, co \rangle$ and $\tau' = \langle E', po', rf', co' \rangle$, let $\tau \subseteq \tau'$ denote that $E = E'$, $po = po'$, $rf = rf'$ and $co^x \subseteq (co')^x$ for each $x \in X$. We say that a (partial) trace $\tau$ is RA-consistent if there is a total trace $\tau'$ with $\tau \subseteq \tau'$ such that $\tau' \models RA$. In particular, a total trace $\tau$ is RA-consistent if and only if $\tau \models RA$. Also, $\llbracket Y \rrbracket_{RA}^{Weak}$ is the set of RA-consistent weak traces of $\gamma$.

We can now analyze a program by exploring runs until all weak traces in $\llbracket Y \rrbracket_{RA}^{Weak}$ have been generated. In the next sections, we will present such an analysis algorithm, which is also optimal in the sense that it explores each trace in $\llbracket Y \rrbracket_{RA}^{Weak}$ exactly once. Such an algorithm must overcome the challenges of (i) consistency, i.e., examining only RA-consistent weak traces (note that many weak traces are not RA-consistent; e.g., for the program in Figure 1a, letting both threads read from the write of the other thread is not possible under the RA semantics), and (ii) non-redundancy, i.e., to generate each weak trace only once, thereby avoiding unnecessary explorations. We solve the consistency challenge in §4, by defining a saturated semantics, equivalent with the standard one, in which runs are associated with partial traces that contain precisely those coherence edges that must be present in any corresponding total trace. Based on the saturated semantics, the non-redundancy challenge is solved by the design of our DPOR algorithm in §5.

### 4 SATURATED SEMANTICS

In this section, we address the challenge of exploring only RA-consistent trace by defining a new semantics for RA, called the saturated semantics. The saturated semantics is the basis for our DPOR algorithm in §5, which generates precisely the weak traces in $\llbracket Y \rrbracket_{RA}^{Weak}$ for a given configuration $\gamma$.

The saturated semantics solves the consistency challenge by making it easy to maintain RA-consistency. We define the semantics in two steps. First, in §4.1, we define the notion of a saturated trace as a partial trace which extends a weak trace, whose (partial) coherence relation contains precisely the edges that occur in all RA-consistent total extensions of that weak trace. We show that a saturated trace is RA-consistent iff it does not contain a cycle that is forbidden by the RA semantics (cf. Theorem 4.2). Then, in §4.2, we present two efficient operations that allow adding a new write (resp. read) event to a trace while preserving both saturation and RA-consistency. We use these operations as the basis to define our saturated semantics. Finally, we show a key theorem (cf. Theorem 4.8) that the saturated semantics coincides with the RA semantics on weak traces.
4.1 Saturated Traces

**Definition 4.1.** A trace $\tau$ is saturated if, for all variables $x \in \mathcal{X}$, whenever $e, e' \in \mathcal{E}_x^k$ with $e \neq e'$ such that $e \; [\text{po} \cup \text{rf} \cup \text{co}^x]^+ e''$ and $e' \; [\text{rf}^x] \; e''$, then $e \; [\text{po} \cup \text{rf} \cup \text{co}^x]^+ e'$. 

To motivate this definition, note that whenever $e \; [\text{po} \cup \text{rf} \cup \text{co}^x]^+ e''$ and $e' \; [\text{rf}^x] \; e''$ then (under RA) any coherence edge between $e$ and $e'$ must (if present) be directed from $e$ to $e'$; otherwise $e' \; [\text{co}^x] \; e$ and $e' \; [\text{rf}^x] \; e''$ would imply $e'' \; [\text{fr}^x] \; e$, implying $e \; [\text{po} \cup \text{rf} \cup \text{co}^x \cup \text{fr}^x]^+ e$ (by $e \; [\text{po} \cup \text{rf} \cup \text{co}^x]^+ e'' \; [\text{fr}^x] \; e$), thereby violating the RA semantics.

The following theorem shows an essential property of saturated traces, namely that if such a trace does not contain cycles violating the RA semantics then it is RA-consistent.

**Theorem 4.2.** For a partial trace $\tau$, if $\tau$ is saturated and $\tau \models \text{RA}$, then $\tau$ is RA-consistent.

**Proof.** Assume that $\tau = \langle \mathcal{E}, \text{po}, \text{rf}, \text{co} \rangle$ is saturated and $\tau \models \text{RA}$. We show that there is a total trace $\tau' = \langle \mathcal{E}, \text{po}, \text{rf}, \text{co} \rangle'$ such that $\tau \subseteq \tau'$ and $\tau' \models \text{RA}$. The lemma then follows immediately.

We define a sequence of traces $\tau_0 \subseteq \tau_1 \subseteq \tau_2, \ldots$ inductively such that $\tau_0 = \tau$, $\tau_i$ is saturated and $\tau_i \models \text{RA}$. For $i > 0$, if $\tau_i = \langle \mathcal{E}, \text{po}, \text{rf}, \text{co} \rangle$ is not total then $\tau_{i+1}$ is derived from $\tau_i$ by adding the pair $\langle e_i, e'_i \rangle$ to the coherence order, where $e_i, e'_i \in \mathcal{E}_x^k$ for some $x \in \mathcal{X}$ with $\neg(e'_i \; [\text{po} \cup \text{rf} \cup \text{co}^x]^+ e_i)$ and $\neg(e_i \; [\text{co}^x] \; e'_i)$. Such events $e_i$ and $e'_i$ exist since $\tau_i$ is not total and since $\tau_i \models \text{RA}$ by the induction hypothesis. Also, by the induction hypothesis we know that $\tau_i$ is saturated. It follows that $\tau_{i+1}$ is saturated. Since $\tau_i \models \text{RA}$ it follows that $\tau_{i+1} \models \text{RA}$. Notice that, by construction, we have $|\tau_i| < |\tau_{i+1}|$. It follows that there is a $j$ such that $\tau_j$ is total. Define $\tau' := \tau_j$. This concludes the proof of the lemma.

4.2 Saturated Semantics

Next, we introduce two notions that are relevant when adding a new read event to a trace.

4.2.1 Readability and Visibility. (i) **Readability** identifies the write events $e'$ from which read event $e$ can fetch its value, and **visibility** is used to add new coherence-order edges that are implied by the fact that the new event $e$ reads from $e'$. Below, we fix a trace $\tau = \langle \mathcal{E}, \text{po}, \text{rf}, \text{co} \rangle$.

**Definition 4.3.** For a thread $th \in \mathcal{T}$ and a variable $x \in \mathcal{X}$, the set of readable events $\mathcal{R}(\tau, th, x)$ is defined to be the set of events $e \in \mathcal{E}_x^k$ such that there are no events $e' \in \mathcal{E}_x^k$ and $e'' \in \mathcal{E}_x^k$ with $e \; [\text{po} \cup \text{rf} \cup \text{co}^x]^+ e'$ and $e' \; [\text{po} \cup \text{rf}]^+ e''$. 

![Fig. 4. (a) Adding a read event $e_6$ and (b) a write event $e_5$. The dotted arrows are added due to the new event.](image)

Proc. ACM Program. Lang., Vol. 2, No. OOPSLA, Article 135. Publication date: November 2018.
Intuitively, $R(\tau, \text{th}, x)$ contains all write events on $x$ that are not hidden from thread $\text{th}$ by other write events on $x$. A new read event on $x$ that is added to $\text{th}$ can fetch its value from any write event in $R(\tau, \text{th}, x)$. In fact, if $\text{th}$ is saturated then $R(\tau, \text{th}, x)$ is precisely the set of write events from which $e$ can read without introducing a cycle that is forbidden by the RA semantics.

To illustrate this, let $\tau$ be as in Figure 4a, where the read $e_6$ is about to be added, and let $e_2, e_3,$ and $e_1$ be write events on $x$. The dotted arrows (explained later) represent edges that will be added to the traces due to the new event $e_6$. Let $\text{th}$ be the thread of $e_3$ and $e_6$. Then $e_1, e_3 \in R(\tau, \text{th}, x)$, while $e_2 \notin R(\tau, \text{th}, x)$. Note that reading from $e_2$ would destroy RA-consistency, since any corresponding total trace must have $e_2 \gg [\text{co}^x] e_3$ and hence $e_6 \gg [\text{fr}^x] e_3$, inducing the cycle $e_6 \gg [\text{fr}^x] e_3 \gg [\text{rf}] e_4 \gg [\text{po}] e_6$ thereby violating the RA semantics.

**Definition 4.4.** For a thread $\text{th} \in T$ and a variable $x \in X$, the visible events $V(\tau, \text{th}, x)$ is defined to be the set of events $e$ in $R(\tau, \text{th}, x)$ such that there is an $e' \in E^{\text{th}}$ with $e \gg [\text{po} \cup \text{rf}]^* e'$.

In other words, the set contains all the readable events on $x$ that can reach an event in $\text{th}$ through $\text{po}$ and $\text{rf}$ edges. The point of $V(\tau, \text{th}, x)$ is that if a new read event $e$ on $x$ is added to $\text{th}$, which reads from an event $e'$ then the resulting trace is saturated by adding a coherence-order edge from each $e'' \in V(\tau, \text{th}, x)$ to $e'$. As illustration, in Figure 4a (again before adding $e_6$), we have $e_3 \in V(\tau, \text{th}, x)$ while $e_2 \notin V(\tau, \text{th}, x)$. This means that if we let $e_6$ read from $e_1$ then the saturation of the resulting trace adds a coherence-order edge from $e_3$ to $e_1$; on the other hand, no edge is added from $e_2$ to $e_1$.

### 4.2.2 Saturated Semantics

We define the saturated semantics as a transition relation $\rightarrow_S$ on traces (the first two rules of Figure 5). The transition rules correspond to adding a new write or read event to a trace $\tau$ and obtaining a new trace $\tau'$. Each transition is labeled by an observation $\alpha$ which is either a write event $e$, or a pair $(e, e')$, consisting of a read event $e$ that reads from a write event $e'$. We define $\alpha.event := e$. As we will explain below, an important property of the transition relation is that if the original trace $\tau$ is saturated and has no cycles forbidden by the RA-semantics, then the new trace $\tau'$ will satisfy the same conditions. Therefore, by Theorem 4.2 it follows that all traces generated according to the saturated semantics are RA-consistent. Indeed, the semantics generates precisely those traces that are saturated and contain no cycles violating the RA semantics.

Rule S-TRACE-WRITE describes that a saturated trace is extended with a write event $e$ by (i) adding a new write event $e$, whose identity reflects that it is the most recent event performed by its thread, (ii) making $e$ last in the program order of its thread, and (iii) keeping the read-from and coherence relations. We observe that if $\tau$ is saturated and $\tau \models RA$, then $\tau'$ will be saturated and $\tau' \models RA$. The
The reason is that $e$ does not have any successors w.r.t. $(\text{po} \cup \text{rf} \cup \text{co})^*$ in $\tau'$, and hence the only cycles in $\tau'$ are those that are already in $\tau$. Since no cycle in $\tau$ violates the RA semantics by assumption, no cycle in $\tau'$ will break the RA semantics either. Figure 4b illustrates how to apply the rule to extend a saturated trace with a new read event $e_5$.

Rule $S\text{-TRACE-READ}$ describes that a saturated trace is extended with a read event $e$ that reads from a write event $e'$ by (i) adding the new read event $e$, (ii) ensuring that the write event $e'$ is readable for the thread $th$ w.r.t. the variable $x$ in $\tau$, (iii) extending the program order in the same manner as in $S\text{-TRACE-WRITE}$, (iv) extending the read-from relation to reflect that $e$ reads from $e'$, and (v) extending the coherence order by an edge from each visible event to $e'$. If $\tau$ is saturated then $\tau'$ will also be saturated since we add all the additionally needed coherence-order edges, namely the ones from the events in $\mathcal{V}(\tau, th, x)$ to $e'$. Furthermore, if $\tau$ is saturated and $\tau \models \text{RA}$ then $\tau' \models \text{RA}$.

The reason is that any new cycle violating the RA semantics in $\tau'$ would include an edge from an event $e'' \in \mathcal{V}(\tau, th, x)$ to $e'$. However, such a cycle implies that $e' [\text{po} \cup \text{rf} \cup \text{co}^+] e''$, i.e., $e' \notin \mathcal{R}(\tau, th, x)$ which is a contradiction. Figure 4a illustrates how to apply the rule to extend a saturated trace with a new read event $e_6$.

4.2.3 Properties of the Saturated Transition Relation. We describe three properties of the saturated semantics, namely efficiency, deadlock-freedom, and correctness.

**Efficiency.** The sets $\mathcal{R}(\tau, th, x)$ and $\mathcal{V}(\tau, th, x)$ can both be computed in polynomial time. To see this, suppose we are given a $\tau = (E, \text{po}, \text{rf}, \text{co})$, a thread $th$, and a variable $x$. A polynomial time algorithm for computing the set $\mathcal{R}(\tau, th, x)$ can be defined consisting of the following three steps:

1. Compute the transitive closure of the relation $\text{po} \cup \text{rf} \cup \text{co}^x$ using, e.g., the Floyd-Warshall algorithm [Cormen et al. 2009]. This will cost $O(|E|^3)$ time.
2. Compute the set of events $e'$ such that $e' [\text{po} \cup \text{rf}]^* e''$ for some $e'' \in E^th$. This will cost $O(|E|^2)$ time.
3. For each event $e \in E^{th,x}$, check whether there is an event $e'$ in the set computed in step (ii) with $e [\text{po} \cup \text{rf} \cup \text{co}^x]^* e'$. If not, add the event $e$ to the set $\mathcal{R}(\tau, th, x)$. This will cost $O(|E|^2)$ time.

The set $\mathcal{V}(\tau, th, x)$ can be computed similarly. This gives the following lemma.

**Lemma 4.5.** For a trace $\tau$, a thread $th$, and a variable $x$, we can compute the sets $\mathcal{R}(\tau, th, x)$ and $\mathcal{V}(\tau, th, x)$ in polynomial time.

By Lemma 4.5 it follows that we can compute each step of $\rightarrow_{S}$ in polynomial time. This property is not satisfied by all memory models. For instance, calculating the successors of a trace in the SC semantics amounts to solving an NP-complete problem [Chalupa et al. 2018].

**Deadlock-Freedom.** The saturated semantics is deadlock free in the sense that if a thread can perform a transition from a configuration then there is always a corresponding move in the saturated semantics. This is captured by the following lemma (which follows immediately from the definitions).

**Lemma 4.6.** For a configuration $\gamma$ and trace $\tau$, if $\gamma \models \text{RA}$ and $\text{succ}(\gamma) \neq \emptyset$ then $\text{succ}_{S}(\gamma, \tau) \neq \emptyset$. 
Correctness. The following lemma states the fact that the saturated semantics preserves saturation and RA-consistency.

**Lemma 4.7.** If \( \tau \) is saturated, \( \tau \models RA \), and \( \tau \xrightarrow{_S^\alpha} \tau' \) then \( \tau' \) is saturated and \( \tau' \models RA \).

Define \( \|Y\|_{RA}^S := \{ \tau \mid \exists \gamma'. \langle \gamma, \tau_0 \rangle \xrightarrow{^*_{S}} \langle \gamma', \tau \rangle \} \), i.e., it is the set of traces that can be generated by sequences of \( \rightarrow_{S} \)-transitions, starting from \( \gamma \) and the empty trace. The following theorem states that the RA semantics and the saturated semantics coincide on weak traces, i.e., \( \|Y\|_{RA}^S \) and \( \|Y\|_{RA}^{Weak} \) generate the same set of weak traces.

**Theorem 4.8.** For any configuration \( \gamma \), we have \( \{ \text{weak}(\tau) \mid \tau \in \|Y\|_{RA}^S \} = \|Y\|_{RA}^{Weak} \).

### 4.3 Example

We will give an example (Figure 6) to illustrate why saturation is important in the semantics, and how the semantics preserves saturation of traces. To that end, we will consider the simple program in Figure 1a again. To make the presentation easier to read, we will use a simplified notation for events and observations. First, we represent an event in a trace by the corresponding program instruction. For instance, we will write \( x := 1 \) instead of \( \langle 1, th_1, W, x, 1 \rangle \), and we will write \( a := x \) instead of \( \langle 2, th_1, R, x \rangle \). We write an observation \( \langle e, e' \rangle \) as \( e = 2 \) if \( e \) corresponds to a read statement of the form \( a := x \) and \( e' \) corresponds to a write statement of the form \( x := 2 \). In this particular example, we can use this simplified notation since all the statements in the program are different.

Suppose that we have generated the trace \( \tau_6 \) containing the write events \( x := 1 \) and \( x := 2 \), corresponding to \( th_1 \) and \( th_2 \) running their first instructions respectively. The trace \( \tau_6 \) is trivially saturated since it does not contain any read events. The event \( x := 1 \) is visible to the thread \( th_1 \), and the event \( x := 2 \) is visible to the thread \( th_2 \). Both events are readable by both \( th_1 \) and \( th_2 \).

Suppose that \( th_1 \) executes the read instruction \( a := x \) and it chooses to read from the event \( x := 2 \). This corresponds to performing the observation \( a = 2 \). Since \( x := 1 \) is visible to \( th_1 \) our saturated semantics will add a co-edge from \( x := 1 \) to \( x := 2 \) which means that the resulting trace \( \tau_7 \) is saturated. In \( \tau_7 \), the only readable event by \( th_2 \) is \( x := 2 \) (since \( x := 1 \) is now hidden by \( x := 2 \)). If \( th_2 \) performs the read instruction \( b := x \) then it can only read from the event \( x := 2 \), corresponding to performing the observation \( b = 2 \) and leading to the trace \( \tau_4 \) (the same trace as the one in Figure 1b).
Since the only visible event in $\tau_4$ is $x := 2$, the semantics will not add any new co-edges. Notice that $\tau_4$ is already saturated.

Next, we will show why saturation is crucial for the semantics. Let us return to the scenario where we are in $\tau_0$ and $th_1$ performs its read instruction $a := x$ and it chooses to read from the event $x := 2$. Suppose that we do not add the co-edge from $x := 1$ to $x := 2$ and thus obtain $\tau_5$ which is not saturated. If $th_2$ performs the read instruction $b := x$ then it cannot deduce from $\tau_5$ that it cannot read from $x := 1$. However, if $th_2$ reads from $x := 1$, we obtain $\tau_6$ which is not RA-consistent. The reason is that regardless of whether we put a co-edge from $x := 1$ to $x := 2$ or the opposite, we will obtain a cycle that is forbidden by the RA semantics.

Redundancy. While the saturated semantics generates only RA-consistent traces, it still suffers from the redundancy problem. More precisely, several runs of the program may produce the same weak trace. Consider again the simple program in Figure 1a. Consider two runs, namely $\rho_1$ where first $th_1$ executes $x := 1$ and then $th_2$ executes $x := 2$, and $\rho_2$ where first $th_2$ executes $x := 2$ and then $th_1$ executes $x := 1$. The runs $\rho_1$ and $\rho_2$ have the same trace, namely $\tau_6$ in Figure 6, and hence exploring both of them in the analysis would be wasteful. While this particular scenario is quite simple, redundant explorations may have quite complex forms. The DPOR algorithm in §5 aims to obtain an optimal search, i.e., consider only one run per weak trace of the input program.

5 DPOR ALGORITHM

In this section, we present our DPOR algorithm. For a terminating program, it systematically explores the whole set of weak traces that can be generated according to the saturated semantics. The algorithm is sound, complete, and optimal in the sense that it produces each weak trace corresponding to a terminated run precisely once. We achieve optimality by combining the saturated semantics with an exploration algorithm ensuring that no two generated traces will have same program order and read-from relations. Moreover, the algorithm is deadlock-free in the sense that the exploration only ends at points when the considered program configuration is terminal. We will first give a detailed description of the algorithm in §5.1, then provide a complete example that illustrates the main ingredients of the algorithm in §5.3, and finally state its properties in §5.4.

5.1 Algorithm

The DPOR algorithm explores the program according to the saturated semantics using the main procedure DfVisit, starting with an input configuration, a trace, and a sequence of observations.

For a read event, the algorithm merely adds the event and continues with the next event. For a read event $e$, the algorithm also adds $e$, but it continues in several separate branches each corresponding to a different write event from which $e$ can fetch its value. Besides, the algorithm must handle the case that $e$ may read from a write event $e'$ that will only be added to the trace later in the exploration. We say that $e'$ has been postponed w.r.t. $e$. When $e'$ is eventually generated, the DPOR algorithm will detect it and swap it with $e$, thus making it possible for $e$ to read from $e'$. To swap $e$ and $e'$, we also need to include the sequence of observations that are “necessary” for generating $e'$. We will refer to such a sequence as a schedule (cf. the DECLAREPOSTPONED procedure, §5.1.2). All the generated schedules will eventually be executed thus swapping $e$ with all the write events that are postponed w.r.t. it (cf. the RUNSCHEDULE procedure, §5.1.3).

5.1.1 DfVisit - The main procedure (Algorithm 1). The depth-first exploration is given in Algorithm 1. The DfVisit($\gamma$, $\tau$, $\pi$) procedure explores all RA-consistent weak traces of the program $P$ that are generated from a configuration $\gamma$ and a saturated trace $\tau$, where $\gamma$ and $\tau$ have been generated by performing a sequence of observations $\pi$. As we describe below, the sequence $\pi$ is used for swapping read events with write events that are postponed w.r.t. them. Initially, the
procedure is called with an initial configuration $\gamma$, the empty trace $\tau_0$, and the empty observation sequence $e$.

In a call, if $\gamma$ has no successors, then \textsc{DfVisit} returns to its caller. Otherwise, it considers an enabled write or read statement. If a write statement is enabled, then one such write statement is selected non-deterministically, and the corresponding event $e$ is created (at line 2). This event $e$ is added to the trace according to the saturated semantics and also added to the sequence of observations, whereafter \textsc{DfVisit} is called recursively to continue the exploration (at line 3). After the recursive call has returned, the algorithm calls \textsc{DeclarePostponed}, which finds the read events $e'$ in the input exploration sequence $\pi$, which would be able to read from the write $e$ if $e$ were performed before $e'$. For each such a read event $e'$, \textsc{DeclarePostponed} creates a schedule for $e'$, which is a sequence of observations that can be explored from the point where $e'$ was performed, to let the write $e$ occur before $e'$ so that $e'$ can read from $e$ (see a detailed description in §5.1.2).

If a read statement is enabled, then a read event $e$ is created, and the set $\text{Schedules}(e)$ is initialized to the empty set (at line 7). The set $\text{Schedules}(e)$ will be gradually updated by the \textsc{DeclarePostponed} procedure when subsequent writes are explored. We also associate a Boolean flag Swappable($e$) for each read event $e$. The flag indicates whether $e$ is swappable, i.e., whether following write events should consider $e$ for swapping or not. The reason for including this flag is to prevent \textsc{DeclarePostponed} from swapping read events that occur in a schedule; this would lead to redundant explorations and a violation of the optimality of the algorithm. After that, for each already generated write event $e'$ from which $e$ can read its value, \textsc{DfVisit} is called recursively to continue the exploration (at line 8). After these calls have returned, the set of schedules collected in $\text{Schedules}(e)$ for the read event $e$ is considered. Each such a schedule is explored by the \textsc{RunSchedule} procedure, thereby allowing $e$ to fetch its value from the corresponding write event.

5.1.2 The \textsc{DeclarePostponed} procedure (Algorithm 2). This procedure inputs a trace $\tau$ (the trace that has been built up to this point) and a sequence of observations $\pi$ whose last element is a write event $e$. The algorithm finds the closest $e'$ to $e$ in $\pi$ for which $e$ can be considered as a postponed write. The algorithm then adds a schedule to the set $\text{Schedules}(e')$ to allow $e'$ to read from $e$ in a new exploration. The criterion for considering such a read event $e'$ is that it satisfies four conditions, namely (i) $e'$ reads from the same variable to which $e$ writes, (ii) $e'$ does not precede $e$ w.r.t. the relation $(\text{po} \cup r)^+$, (iii) $e'$ is swappable, and (iv) $e'$ is closest to $e$. The first three conditions are checked at line 4. The condition (iv) is satisfied by the break statement at line 11 which makes the procedure stop after finding the first read event satisfying the first three conditions.

---

**Algorithm 1:** \textsc{DfVisit($\gamma$, $\tau$, $\pi$)}

\begin{itemize}
\item \textbf{Input:} $\gamma$ is a configuration, $\tau = (E, \text{po}, \text{rf}, \text{co})$ is a trace, $\pi$ is an observation sequence.
\item 1. if $\exists \gamma \rightarrow^{(\text{th},w,x,e)} \gamma'$ then \hspace{1cm} // handle a write event
\item \hspace{1cm} let $e$ be $(|E|^t + 1, \text{th}, \text{w}, x, e)$ and $\tau'$ be such that $\gamma, \tau \rightarrow_{S}^{e} (\gamma', \tau')$ \hspace{1cm} // follow S-WRITE
\item \hspace{1cm} \textsc{DfVisit}($\gamma'$, $\tau'$, $\pi$ $\bullet$ $e$)
\item 2. \textsc{DeclarePostponed}($\tau'$, $\pi$ $\bullet$ $e$)
\item 3. else if $\exists \gamma$ $\rightarrow^{(\text{th},r,x,e)}$ $\ast$ then \hspace{1cm} // handle a read event
\item \hspace{1cm} let $e$ be $(|E|^t + 1, \text{th}, R, x)$
\item \hspace{1cm} Schedules($e$) $\leftarrow \emptyset$; Swappable($e$) $\leftarrow$ true
\item \hspace{1cm} for $e', \gamma', \tau' : (\gamma, \tau) \rightarrow S (\gamma', \tau')$ do \textsc{DfVisit}($\gamma'$, $\tau'$, $\pi$ $\bullet$ $\langle e, e' \rangle$) \hspace{1cm} // follow S-READ
\item \hspace{1cm} for $\beta \in$ Schedules($e$) do \textsc{RunSchedule}($\gamma$, $\tau$, $\pi$, $\beta$)
\end{itemize}
would in fact violate the optimality condition. Which by construction is enabled. During this exploration, read events are declared non-swappable using the following facts.

Our DPOR algorithm spends \( Schedules(\gamma) \) time for each explored trace. This complexity is estimated using the following facts.

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**Algorithm 2: DECLAREPOSTPONED(\( \tau, \pi \))**

**Input:** \( \tau = (\langle E, po, rf, co \rangle) \) is a trace and \( \pi \) is an explored observation sequence.

1. let \( e \) be last(\( \pi \)) and \( x \) be \( e.var \)
2. for \( i \leftarrow \lvert \pi \rvert - 1 \) to 1 do
   // look for the closest event \( e' \) that has postponed \( e \)
   3. let \( e' \) be \( \pi[i].event \)
   4. if \( e' \in E^{R,x} \land \neg(e' [po \cup rf]^+ e) \land \text{Swappable}(e') \) then
      \( \beta \leftarrow e \)
   5. for \( j \leftarrow i + 1 \) to \( \lvert \pi \rvert - 1 \) do
      // get all events after \( e' \) in \( \pi \) and precedes \( e \) in \((po\cup rf)^+\)
      6. let \( e'' \) be \( \pi[j].event \)
      7. if \( e'' [po \cup rf]^+ e \) then \( \beta \leftarrow \beta \bullet \pi[j] \)
   8. if \( \not\exists \beta' \in \text{Schedules}(e') \cdot \beta' = \beta \bullet e \bullet \langle e', e \rangle \) then
      9. \( \text{Schedules}(e') \leftarrow \text{Schedules}(e') \cup \{ \beta \bullet e \bullet \langle e', e \rangle \} \) // allow \( e' \) to read from \( e \)
10. break

---

**Algorithm 3: RUNSCHEDULE(\( \gamma, \tau, \pi, \beta \))**

**Input:** \( \gamma \) is a configuration, \( \tau \) is a trace, \( \pi \) is an explored-observation sequence, and \( \beta \) is a schedule.

1. if \( \beta \neq e \) then // explore the sequence of observations one by one
   2. let present \( \beta \) in the form \( \alpha \bullet \beta' \) and \( e \) be \( \alpha.event \)
   3. choose \( \gamma', \tau' : \langle \gamma, \tau \rangle \xrightarrow{\alpha} \langle \gamma', \tau' \rangle \) // follow S-WRITE and S-READ
   4. if \( e.type = R \) then \( \text{Swappable}(e) \leftarrow \text{false} \)
   5. \( \text{RUNSCHEDULE}(\gamma', \tau', \pi \bullet \alpha, \beta') \)
   6. else \( \text{DfVisit}(\gamma, \tau, \pi) \)

---

After finding such a read event \( e' \), a schedule \( \beta \) is created. The schedule consists of all the events following \( e' \) in \( \pi \) and which precede \( e \) w.r.t. the relation \((po \cup rf)^+\) (at line 6). The schedule \( \beta \) has the write event \( e \) and then the pair \( \langle e', e \rangle \) at the end, thereby achieving the goal of making \( e' \) read from \( e \) (at line 10).

The new schedule \( \beta \) is added to the set \( \text{Schedules}(e') \) if the latter does not already contain a schedule \( \beta' \) which is equivalent to \( \beta \). We consider \( \beta \) and \( \beta' \) to be equal, denoted \( \beta \approx \beta' \), if they contain the same set of observations (possibly in different orders). Notice that if \( \beta \approx \beta' \) and we execute \( \beta \) or \( \beta' \) then we reach identical traces which means that \( \beta \) need not be added to the set \( \text{Schedules}(e') \) if \( \beta' \) is already in \( \text{Schedules}(e') \), and that having both \( \beta \) and \( \beta' \) in \( \text{Schedules}(e') \) would in fact violate the optimality condition.

5.1.3 The RUNSCHEDULE procedure (Algorithm 3). The procedure inputs a configuration \( \gamma \), a trace \( \tau \), a sequence of observations \( \pi \), and a schedule \( \beta \). It explores the sequence of scheduled observations in \( \beta \) one by one, by calling itself recursively (at line 5), after which \( \text{DfVisit} \) is called to continue the exploration beyond the end of \( \pi \) (at line 6). Note that each observation corresponds to a statement which by construction is enabled. During this exploration, read events are declared non-swappable by setting the corresponding entry in \( \text{Swappable} \) to false.

5.2 Complexity

Our DPOR algorithm spends polynomial time for each explored trace. This complexity is estimated using the following facts.

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Initially: \( x = y = 0 \)

\[
\begin{align*}
& a := x, x := 2, b := y, x := 3, c := y, x := 4, \\
& \text{th}_1 : \text{th}_2 : \text{th}_3 : \text{th}_4 : \\
& \downarrow \quad \downarrow \quad \downarrow \quad \downarrow \\
& \text{th}_1 \quad \text{th}_2 \quad \text{th}_3 \quad \text{th}_4
\end{align*}
\]

Initially: \( x = y = 0 \)

\[
\begin{align*}
& \text{th}_1 : \text{th}_2 : \text{th}_3 : \text{th}_4 : \\
& \downarrow \quad \downarrow \quad \downarrow \quad \downarrow \\
& \text{th}_1 \quad \text{th}_2 \quad \text{th}_3 \quad \text{th}_4
\end{align*}
\]

\( \text{Fig. 7. Example illustrating the DPOR algorithm.} \)

1. Each single schedule is at most linear (in the size of the program), since its length is bounded by the length of the program.
2. It is indeed possible (in pathological cases) that the exploration will produce an exponential number of schedules. However, each of these schedules will result in a different weak trace; therefore this happens only when the number of weak traces of the program is exponential. Thus (importantly) the effort in constructing these schedules is never wasted.
3. Even if the number of schedules is exponential, it can be checked in polynomial (at most quadratic) time whether a new schedule is equivalent to an existing one. This can be done by organizing the schedules into a tree and systematically compare the new schedule from the beginning to the schedules in the tree (the essential operation is to compare read-from edges).

5.3 Example

Figure 7 illustrates the recursion tree corresponding to a run of the DPOR algorithm on a simple program consisting of four threads \{\text{th}_1, \text{th}_2, \text{th}_3, \text{th}_4\} and two shared variables \{x, y\}. We will use the same simplified notation for events and observations as in the example in §4.3.

The nodes in the tree represent the recursive calls of \text{DfVisit}. The node \( n_0 \) represents the first call with an empty trace and an empty input observation sequence. From \( n_0 \), \text{DfVisit} selects the only available write statement \( x := 2 \) (of \( \text{th}_2 \)), making a recursive call and moving to \( n_1 \) (line 1 of \text{DfVisit}). At node \( n_1 \), one of the three read statements, namely \( a := x \) is selected (line 5 of \text{DfVisit}). There are two possible write events from which the event \( a := x \) can read, namely the initial write event \text{init}_x which gives the value 0 and the event \( x := 2 \) which gives the value 2. Accordingly, when \text{DfVisit} runs the for-loop at line 8, it will eventually create two branches leading to the
nodes \( n_{10} \) and \( n_{20} \). In the first iteration of the for-loop, a recursive call is made leading to the node \( n_{10} \) from where the statement \( b := y \) is selected. There is one write event from which \( b := y \) can read its value, namely \( \text{init}_t \gamma \), so there will be no branching in this case, and the successor node will be \( n_{11} \). At \( n_{11} \) the only available write statement \( x := 3 \) is selected after which \( c := y \) (reading from \( \text{init}_t \gamma \)), and \( x := 4 \) are executed. Notice that the trace obtained at node \( n_{14} \) is \( \tau_1 \) and the observation sequence is \( x := 2 a = 0 b = 0 x := 3 c = 0 x := 4 \). The recursive call to \( \text{DFVisit} \) from \( n_{13} \) to \( n_{14} \) will return immediately to \( n_{13} \) since there are no enabled statements left in the program, after which \( \text{DECLAREPOSTPONED} \) will be called (line 4 of \( \text{DFVisit} \)). The latter will detect that \( x := 4 \) is postponed w.r.t. \( a := x \) (line 4 of \( \text{DECLAREPOSTPONED} \)). The fact that \( x := 4 \) is postponed w.r.t. \( a := x \) is marked by the red arrow in Figure 7. The \( \text{DECLAREPOSTPONED} \) procedure executes the for-loop at line 2 and creates a schedule \( \beta_1 \) given by \( c = 0 x := 4 a = 4 \) (the event \( c = 0 \) precedes \( x := 4 \) w.r.t. \( \text{po} \)). The schedule \( \beta_1 \) is added to the set of schedules of \( a := x \). Similarly, when the recursive call returns to the node \( n_{11} \), the schedule \( \beta_2 \) containing the sequence \( b = 0 x := 3 a = 3 \) will be added to the set of schedules of \( a := x \).

When the calls of the nodes in the left-most branch (nodes \( n_{10} - n_{14} \)) have all returned and we are at node \( n_1 \) again, \( \text{DFVisit} \) will make a recursive call to node \( n_{20} \) corresponding to the second iteration of the for-loop at line 8. From \( n_{20} \), the execution will continue similarly to the branch \( n_{10} - n_{14} \), creating the branch \( n_{20} - n_{24} \) obtaining the trace \( \tau_2 \). In particular, two schedules will be obtained namely \( c = 0 x := 4 a = 4 \) at \( n_{21} \) and \( b = 0 x := 3 a = 3 \) at \( n_{23} \). However, these two schedules are identical (and hence trivially equivalent) to the schedules \( \beta_1 \) and \( \beta_2 \) respectively, and therefore they will not be added to the set of schedules of \( a := x \).

When the recursive call from \( n_{20} \) has returned to \( n_1 \), the for-loop at line 8 will terminate and \( \text{DFVisit} \) moves to line 9 where it considers the set of schedules of \( a := x \) (in some order). In the example, it selects \( \beta_2 \) first and calls \( \text{RUNSCHEDULE} \), which will create the nodes \( n_{30}, n_{31}, \) and \( n_{32} \). In particular, \( \text{RUNSCHEDULE} \) will set \( \text{Swappable}(a := x) \) to false (line 4). From \( n_{32} \) the call returns to \( \text{DFVisit} \) (line 6 of \( \text{RUNSCHEDULE} \)) and the nodes \( n_{33} \) and \( n_{34} \) will be created leading to the trace \( \tau_3 \). Although the write event \( x := 4 \) is potentially postponed w.r.t. \( a := x \), the corresponding schedule will not be added since the \( \text{Swappable} \) flag of \( a := x \) is false (line 4 of \( \text{DECLAREPOSTPONED} \)). This means that the algorithm will not create the dotted branch \( (n_{32} - n_{34}) \) in Figure 7.

When the recursive call returns from \( n_{30} \) to \( n_1 \) the schedule \( \beta_1 \) will be run in a similar manner to \( \beta_2 \) resulting the right-most branch \( (n_{40} - n_{44}) \) and the trace \( \tau_4 \).

**Optimality**. The test at line 9 of \( \text{DECLAREPOSTPONED} \) is necessary. It prevents adding the schedules from \( n_{30} \) and \( n_{31} \). These two schedules would lead to duplicates of the traces \( \tau_3 \) and \( \tau_4 \). Checking the status of \( \text{Swappable} \) is also necessary (line 4 of \( \text{DECLAREPOSTPONED} \)). In our example, this prevents adding the nodes \( n_{12} - n_{14} \). Adding these nodes would result in \( \tau_5 \) which is identical to \( \tau_4 \).

### 5.4 Properties of the DPOR Algorithm

**Soundness** The algorithm is sound in the sense that if we initiate \( \text{DFVisit} \) from a configuration \( \gamma \), the empty trace \( \tau_0 \), and the empty sequence of observations \( \epsilon \), then \( \tau \in \bigcap_{A, \mathcal{R}} A \) for all explored traces \( \tau \). This follows from the fact that the exploration uses the \( \rightarrow_s \) relation (cf. lines 2 and 8 in Algorithm 1 and line 3 in Algorithm 3).

**Optimality** The algorithm is optimal in the sense that, for any two different recursive calls to \( \text{DFVisit} \) with arguments \( \langle \gamma_1, \tau_1, \pi_1 \rangle \) and \( \langle \gamma_2, \tau_2, \pi_2 \rangle \), if \( \gamma_1 \) and \( \gamma_2 \) are terminal then \( \text{weak}(\tau_1) \neq \text{weak}(\tau_2) \). This follows from the following properties: (i) Each time we run the for-loop at line 8 in Algorithm 1, the read event \( \epsilon \) will read from a different write event. (ii) In each schedule \( \beta \in \text{Schedules}(\epsilon) \) at line 9 in Algorithm 1, the event \( \epsilon \) reads from a write event \( \epsilon' \) that is different

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from all the write events from which \( e \) reads at line 8. (iii) Any two schedules added to Schedules(\( e \)) at line 10 in Algorithm 2 will have a read event reading from two different write events.

**Deadlock-Freedom** If DfVisit is called with parameters \( \langle y, \tau, \pi \rangle \) where \( y \) is not terminal then it will produce a subsequent recursive call. This follows directly from Lemma 4.6.

**Completeness** The algorithm is complete in the sense that for any input configuration \( y \) it will produce all weak traces corresponding to terminating runs from \( y \). More precisely, for any configuration \( y \), terminated run \( \rho \in \text{Runs}(y) \), and total trace \( \tau \) with \( \tau \models \rho \), DfVisit(\( y, \tau_0, e \)) will produce a recursive call DfVisit(\( y', \tau', \pi \)) for some terminal \( y', \tau', \pi \) where \( \text{weak}(\tau') = \text{weak}(\tau) \).

6 EXPERIMENTAL EVALUATION

In this section, we describe the implementation of our new techniques, including the saturated semantics in §4 and the new optimal DPOR algorithm w.r.t. weak traces in §5, as a tool called Tracer (weak TRace ACquirE Release). We evaluate the effectiveness of Tracer in practice, by comparing its performance with two other stateless model checking tools running under the Release-Acquire semantics, namely CDSChecker [Norris and Demsky 2016] and Rcmc [Kokologiannakis et al. 2018].

**Tracer tool** Tracer can be used to unit test portions of concurrent code written in the Release-Acquire fragment of the C/C++11 memory model to discover which behaviors the memory model allows. By analyzing the set of weak traces generated during the exploration of programs, Tracer can detect runtime errors such as user-provided assertion violations, deadlocks (with standard definition as in [Silberschatz et al. 2012]), and buggy accesses on uninitialized memory locations. We emphasize that all deadlocks detected by Tracer actually occur in the provided programs and they are not due to our DPOR algorithm.

Tracer is constructed as dynamically-linked shared libraries which implement the C/C++11 acquire and release accesses types in the <atomic> library [ISO 2012] and portions of the other thread-support libraries of C/C++11 (e.g., <threads> and <mutex>). In more detail, it instruments any concurrency-related API calls such as write and read accesses to shared atomic variables, thread creations and thread joinings in input programs into function calls provided by self-implemented utility libraries. At runtime, Tracer determines the next possible transitions and the values returned by atomic memory operations following the saturated semantics in §4. Furthermore, Tracer controls the exploration of the input program until it has explored the whole set of weak traces, using the DPOR algorithm in §5. To verify the correctness of the instrumented libraries in Tracer, we substitute the corresponding headers (e.g., for atomic operations) in GCC and Clang with the headers of the libraries in Tracer. We observe no behavioral differences in our benchmarks when we compile and run them using GCC or Clang with and without the instrumented libraries. Finally, we note that Tracer partly uses some source code from CDSChecker to preprocess and handle input programs at the first step in its model checking progress.

**CDSChecker and Rcmc tools** We compare Tracer with two other state-of-the-art stateless model checking tools running under the RA semantics, namely CDSChecker [Norris and Demsky 2016] and Rcmc [Kokologiannakis et al. 2018]. CDSChecker implements a DPOR algorithm that supports concurrent programs running under the RA semantics, but, as illustrated in §6.1–§6.6, might generate a lot of redundant executions in a less efficient way than Tracer. Rcmc has two options: Rc11 and Wrc11. The Rc11 option generates only consistent executions (w.r.t. RA) by maintaining total coherence orders. Meanwhile, the Wrc11 option maintains partial coherence orders, which may create inconsistent executions which are then not validated (see §6.1). Rcmc/Rc11
is optimal in the sense that, in the absence of RMWs, they explore each consistent Shasha-Snir trace [Shasha and Snir 1988] exactly once. Observe that Rcmc has limited support for thread creation and joining [Personal-com 2018]. Thus, we do some engineering transformations to simulate these operations.

**Experiment setup** We compare Tracer with CDSChecker and Rcmc on six different categories of benchmarks as follows.

1. In §6.1, we apply Tracer, CDSChecker, and Rcmc to a set consisting of thousands of litmus tests (i.e., small programs) taken from [Alglave et al. 2014]. These experiments are mainly used for two purposes: (i) to provide sanity checks of the correctness of the used tools, and (ii) to compare the performance of these tools on small programs running under RA.

2. In §6.2, we run the tools on two standard programs taken from [Flanagan and Godefroid 2005]. These benchmarks have been widely used to compare different stateless model-checking tools (e.g. [Abdulla et al. 2014; Kokologiannakis et al. 2018; Saarikivi et al. 2012]).

3. In §6.3, we run the tools on a collection of concurrent benchmarks taken from the TACAS competition on Software Verification [SV-COMP 2018]. These benchmarks are C/C++ programs with 50-100 lines of code and used by many tools (e.g. [Abdulla et al. 2015a, 2016b; Alglave et al. 2013a; Huang and Huang 2016]).

4. In §6.4, we evaluate the tools on two synthetic programs (one of them is taken from [Norris and Demsky 2016]). We use these benchmarks to show the benefits of using weak traces used by Tracer (see §3) instead of Shasha-Snir traces [Shasha and Snir 1988] used by Rcmc.

5. In §6.5, we apply the tools to three parameterized programs to study their behaviors when increasing the numbers of read and write operations and threads in the programs.

6. In §6.6, we use the tools to analyze more extensive and more challenging benchmarks, containing several implementations of concurrent data structures. Some of them were used in the evaluation of CDSChecker [Demsky and Lam 2015; Norris and Demsky 2016] and Rcmc [Kokologiannakis et al. 2018]. The other benchmarks are high-performance starvation-free algorithms in [Ramalhete and Correia 2016, 2018].

We conduct all experiments on a Debian 4.9.30-2+deb9u2 machine with an Intel(R) Core(TM) i7-3720QM CPU (2.60 GHz) and 16 GB of RAM. Tracer and CDSChecker have been compiled with GCC v6.3.0. We use the artifact-evaluated version of Rcmc [Kokologiannakis et al. 2018] that is compiled with Clang and LLVM v3.8.1. We use the argument `-u 0` in all benchmarks for both Tracer and CDSChecker that provides 0 for release atomic loads from uninitialized memory locations. All these tools were run on deterministic programs with bounded executions. This is achieved by unrolling loops in any tested program up to a specific bound. Therefore, we do not need `-m` and `-y` arguments to control memory liveness and thread yield-based fairness for Tracer and CDSChecker as in [Norris and Demsky 2016] and `-unroll` argument for Rcmc. Finally, all experiments have been set up with a 1-hour timeout.

### 6.1 Evaluation using Litmus Tests

We apply Tracer, CDSChecker, and Rcmc on a set of litmus tests, taken from [Alglave et al. 2014]. Litmus tests are standard benchmarks used by many tools running on weak memory models (e.g. [Abdulla et al. 2016b; Alglave et al. 2014; Kokologiannakis et al. 2018; Sarkar et al. 2011]). Typically, each litmus test contains at most 4 threads with less than 5 shared variables. In a litmus test, threads concurrently execute small sets of instructions. After all threads have finished their executions, an assertion is validated to check whether the set of some behaviors presented by the assertion is allowed. Since litmus tests are small but contain all kinds of dependencies such as address dependency, data dependency, and control dependency (see [Alglave et al. 2014]), they are
Table 1 summarizes the results of different tools on 8064 litmus tests. First of all, all tools, except Rcmc/Wrc11, are consistent w.r.t. the RA semantics. Rcmc/Wrc11 is inconsistent, generating 81 wrong answers. A **wrong answer** denotes that the outcome of the tool in a test is different from the expected outcome. In the above 81 wrong answers, Rcmc/Wrc11 reports unsafe results for the corresponding safe tests. The inconsistency of Rcmc/Wrc11 is due to the imprecise handling of coherence orders. In fact, Rcmc/Wrc11 sometimes does not generate coherence orders that are needed to preserve RA consistency, e.g., the coherence edges in the trace $\tau_7$ in Figure 6. The missing coherence orders help Rcmc/Wrc11 to reduce the number of executions, but can also lead to inconsistent executions. Indeed, Rcmc/Wrc11 allows the inconsistent trace $\tau_9$ in Figure 6 for the program given in Figure 1a. Secondly, Tracer explores fewer executions and has a better performance than the other tools. For example, it generates 22% fewer executions than Rcmc/Wrc11 (that can be observed based on the columns Execs and/or Average Execs in Table 1) which in turn is exploring fewer executions than the remaining tools. Moreover, Tracer is also 4 times faster than Rcmc/Wrc11. Thirdly, we compare three tools on the average running time spent by each tool on the exploration of an execution (the Average Time column in Table 1). The average running time per execution is calculated using the formula

$$\text{AverageTime} := \frac{\text{Time}}{\text{Execs}}$$

i.e. it is the average time a tool needs to explore one execution. We observe that based on AverageTime, Tracer is about 3 times faster than Rcmc but slower than CDSChecker. The reason is that CDSChecker can explore prefixes of some redundant executions before ending them and therefore CDSChecker spends very little time on these small incomplete executions. Meanwhile,

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2Based on our reported results, newer versions of Rcmc can fix some of the 81 wrong answers [Personal-com 2018].

3We generated the expected outcomes using the Herd tool together with the set of RA-axioms provided in [Lahav et al. 2016]. Since Herd is not a stateless model checker and it is significantly slower in these tests, we do not include Herd in our comparison.
Table 2. Comparison of the performance of the three tools when running on two standard benchmarks: Indexer(N) [SV-COMP 2018] and Filesystem(N) [Flanagan and Godefroid 2005]. The LB column indicates how many times a loop is unrolled. If a tool runs out of time, we put t/o. The best number of executions and running time for each benchmark are given in bold font.

| Program      | LB | Executions Explored | Total Running Time |
|--------------|----|----------------------|---------------------|
|              |    | CDSChecker Rcmc Tracer | CDSChecker Rcmc Tracer |
| Indexer(13)  | 20 | 190 64 64             | 0.15s 0.62s 0.04s   |
| Indexer(14)  | 20 | 3,075 512 512         | 2.72s 11.68s 0.38s  |
| Indexer(15)  | 20 | 48,261 4,096 4,096   | 45.32s 4m22s 2.98s  |
| Indexer(16)  | 20 | 740,889 t/o 32,768    | 12m21s t/o 25.78s   |
| Filesystem(16)| 20 | 27 8 8                | 0.04s 0.05s 0.01s   |
| Filesystem(19)| 20 | 728 64 64             | 0.52s 2.02s 0.03s   |
| Filesystem(22)| 20 | 19,678 512 512        | 13.53s 2m29s 0.33s  |
| Filesystem(25)| 20 | 531,415 t/o 4096      | 7m04s t/o 3.12s     |

all explored executions by TRACER and RCMC are complete since all litmus tests only have assertions at the end of programs.

Since RCMC/Wrc11 is inconsistent w.r.t. the RA semantics (cf. Table 1), we exclude it from further comparisons. For the sake of convenience, in the following, we use RCMC for RCMC/Rc11.

6.2 Evaluation using two Standard Benchmarks

We compare TRACER with CDSCHECKER and RCMC on two standard benchmarks, namely Indexer(N) and Filesystem(N), introduced in [Flanagan and Godefroid 2005]. These benchmarks have been widely used to compare different DPOR techniques under the SC semantics, e.g. [Abdulla et al. 2014; Flanagan and Godefroid 2005; Saarikivi et al. 2012]. We parameterize these benchmarks by the number of threads. For Indexer(N), we use 13, 14, 15, and 16 threads. For Filesystem(N), we use 16, 19, 22, and 25 threads. The reason of the parameterization is that more threads generate more conflicts between their write and read operations, and therefore more non-equivalent executions are needed to be explored by different tools. For example, there will be no conflicts between the operations of threads in Indexer(N) with 11 threads and in Filesystem(N) with 13 threads.

In Table 2, we report the number of executions that the three tools explore as well as the time needed by the tools to explore them. We observe that TRACER always examines optimal numbers of executions w.r.t. weak traces and has a better performance than CDSCHECKER and RCMC in all benchmarks. The difference between TRACER and the other tools becomes more evident when we increase the number of threads. In both Indexer(N) and Filesystem(N), RCMC can discover the same numbers of executions as TRACER; however, its total running time and the average time per execution are two orders of magnitude slower than TRACER on average. Observe that the used version of Indexer(N) program contains lock primitives (as it is the case in SV-COMP [2018]). This version is different from the one used in [Kokologiannakis et al. 2018] and this explains the difference in the obtained performance for RCMC between the one reported in Table 2 and the one published in [Kokologiannakis et al. 2018]. Based on our discussion with RCMC’s authors [Personal-com 2018], we conjecture that this performance issue is due to the way RCMC handles lock primitives. Finally, CDSCHECKER explores more executions than TRACER in all benchmarks, and it is about 20 times slower than TRACER.

4 We do experiments with the Indexer(N) benchmark from SV-COMP [2018] since it is similar to the original one and widely used in the verification community. The Indexer(N) in [Kokologiannakis et al. 2018] implements a different version.
Table 3. Comparison of the performance of the three tools when running on the TACAS Competition on Software Verification (SV-COMP) benchmarks. The LB column and the t/o entry have the same meanings as in Table 2.

| Program     | LB | Executions Explored | Total Running Time |
|-------------|----|---------------------|--------------------|
|             |    | CDSChecker | Rcmc | TRACER | CDSChecker | Rcmc | TRACER |
| Pthread_demo| 10 | 184 758    | 184 756 | 184 756 | 24.96s    | 1m03s | 18.00s |
| Gcd         | 8  | 8 814 044 | 1 162 333 | 1 162 333 | 38m20s    | 7m07s | 2m48s  |
| Fibonacci   | 6  | t/o       | 525 630 | 525 630 | t/o       | 31.06s | 57.95s |
| Szymanski   | 6  | t/o       | 26 037 490 | 12 209 410 | t/o       | 44m52s | 14m54s |
| Dekker      | 10 | 7 306 447 | 3 121 870 | 3 121 870 | 15m25s    | 5m12s | 4m52s  |
| Lamport     | 8  | t/o       | 6 580 870 | 3 372 868 | t/o       | 14m40s | 6m58s  |
| Sigma(5)    | 5  | t/o       | 1 279     | 945     | t/o       | 0.09s  | 0.16s  |
| Peterson    | 6  | t/o       | 1 897 228 | 1 897 228 | t/o       | 3m16s  | 3m15s  |
| Stack_true  | 12 | 2 704 157 | 2 704 156 | 2 704 156 | 19m03s    | 5m25s | 10m12s |
| Queue_ok    | 12 | 581 790   | t/o     | 362 880 | t/o       | 33m27s | 12m52s |

6.3 Evaluating using SV-COMP Benchmarks

We compare TRACER with CDSChecker and Rcmc on the set of concurrent benchmarks from SV-COMP [2018] (the TACAS Software Verification competition 2018), cf. Table 3. These benchmarks consist of ten programs that are written in C/C++ with 50-100 lines of code and used by many tools (e.g. [Abdulla et al. 2015a, 2016b; Alglave et al. 2013a; Huang and Huang 2016]). The primary challenge in these benchmarks is to handle a large number of executions that are needed to be explored. For example, Rcmc generates 26 million executions for the Szymanski benchmark. As stated in §6.1, for unsafe benchmarks, Rcmc stops the exploration as soon as it detects the first assertion violation. To fairly compare the efficiency of different DPOR approaches, as it was done in [Abdulla et al. 2014; Kokologiannakis et al. 2018], we remove all the assertions in the benchmarks to let all tools exhaustively explore all possible executions.

From Table 3, we can see that TRACER explores smaller numbers of executions than the two other tools in all cases. As a consequence, TRACER has the best performance in 9 of 10 examples. Meanwhile, CDSChecker times out in 4 of 10 cases. The main reason for these timeouts is that CDSChecker needs to explore a large number of executions. This can be seen in the Gcd example where the number of executions generated by CDSChecker is huge, compared to those produced by Rcmc and TRACER. On two benchmarks, Pthread_demo and Stack_true, where the three tools generate almost the same numbers of executions, CDSChecker is faster than Rcmc but still less efficient than TRACER. The only benchmark in which TRACER does not have the best performance is Fibonacci. In this benchmark, the numbers of executions explored by TRACER and Rcmc are exactly the same, and Rcmc is slightly faster. However, in general TRACER generates 40% fewer number of executions and it is about 3 times faster than Rcmc. Interestingly, TRACER is 2 times faster than Rcmc in the average time per execution, which coincides with the litmus tests in §6.1.

6.4 Evaluation using Synthetic Benchmarks

Next, we expose more differences between the three tools on two synthetic benchmarks given in Figure 8. The first one is N_writers_a_reader(N) benchmark, taken from [Norris and Demsky 2016]. The results, for 7, 8, 9, and 10 threads, are given in Table 4. Since this benchmark does not contain any loops, we do not use loop unrolling and therefore do not show the LB column as in Tables 2 and 3. Since Rcmc is optimal w.r.t. Shasha-Snir traces in the absence of RMWs, the number of executions explored by Rcmc is exactly (factorial) (N + 1)! here. This number corresponds to
/* initially: x=0 */
atomic_int x;

/* N writers*/
void writers(void *arg)
{
  /* tid is from 1 to N */
  int tid = *((int *)arg);
  x.store(tid, release);
}

void reader(void *arg)
{
  int a = x.load(acquire);
  int b = x.load(acquire);
}

/* initially: x=0 */
atomic_int x;

/* 2 writers */
void writers(void *arg)
{
  for (int i=0; i<N; i++)
    x.store(1, release);
}

void reader(void *arg)
{
  int a = x.load(acquire);
  int b = x.load( acquire);
}

Fig. 8. C/C++11 codes of N_writers_a_reader(N) (left) and Redundant_co(N) (right).

| Program                      | Executions Explored | Total Running Time |
|------------------------------|---------------------|--------------------|
| N_writers_a_reader(7)        | 8                   | 0.01s 0.46s 0.00s  |
| N_writers_a_reader(8)        | 9                   | 0.01s 4.19s 0.00s  |
| N_writers_a_reader(9)        | 10                  | 0.01s 46.13s 0.00s |
| N_writers_a_reader(10)       | 11                  | 0.01s 9m35s 0.00s  |
| Redundant_co(5)              | 581                 | 0.03s 0.39s 0.01s  |
| Redundant_co(10)             | 10 631              | 0.64s 23m56s 0.02s |
| Redundant_co(15)             | 59 056              | 4.57s t/o 0.06s  |
| Redundant_co(20)             | 197 231             | 20.27s t/o 0.14s  |

Table 4. Comparison of the performance of the three tools when running on two synthetic programs. The t/o entry has the same meaning as in Table 2.

the number of possible combinations of all feasible po, rf, and total co. However, one can easily see that there are only \(N + 1\) possible values for the read, presented by \(a = 0, a = 1, \ldots, a = N\). In fact, the total coherence order is irrelevant in this benchmark and only the read-from relation is important. Therefore, in contrast to Rcmc, Tracer, which is optimal w.r.t. weak traces, precisely explores \(N + 1\) executions (i.e., linear).

The second benchmark is called Redundant_co(N) and its code is given in Figure 8. We use this example to emphasize more the benefit of using weak traces. As depicted in Table 4, while the number of Shasha-Snir traces and the number of explored traces by Rcmc is \(O(N!)\) (i.e., factorial), Tracer explores only \(O(N^2)\) executions\(^5\). For instance for Redundant_co(10), Tracer finishes the exploration of the program in less than 1 second, while Rcmc examines two orders of magnitude more executions and needs almost half an hour to complete. Since CDSChecker also maintains a partial coherence order as Tracer but not in an optimal way, CDSChecker can explore the same number of executions for N_writers_a_reader(N) as Tracer but not in Redundant_co(N).

6.5 Evaluation using Parameterized Benchmarks

Table 5 reports more experimental results of Tracer, CDSChecker, and Rcmc on three parameterized benchmarks: Sigma(N) (presented in Table 3), Control_flow(N) (used in [Abdulla et al. 2014]), and Exponential_bug(N) (presented in Figure 2 of [Huang 2015]). In Sigma(N)

\[^5\]The exact number of weak traces in Redundant_co(N) is \(3N^2 + 3N + 1\).
Program | LB | Executions Explored | Total Running Time
| | CDSChecker | Rcmc | TRACER | CDSChecker | Rcmc | TRACER |
| Sigma(6) | 6 | 25357 | 10395 | 720 | | 2.20s | 1.96s | 0.04s |
| Sigma(7) | 7 | 605714 | 135135 | 5040 | | 1m02s | 29.06s | 0.40s |
| Sigma(8) | 8 | 16667637 | 2027025 | 40320 | | 33m18s | 8m02s | 3.28s |
| Sigma(9) | 9 | t/o | t/o | | | t/o | t/o | 33.59s |
| Control_flow(6) | 0 | 896 | 55440 | 77 | | 0.09s | 1.97s | 0.01s |
| Control_flow(8) | 0 | 4608 | 11007360 | 273 | | 0.53s | 7m58s | 0.03s |
| Control_flow(10) | 0 | 22528 | t/o | 1045 | | 3.27s | t/o | 0.16s |
| Control_flow(12) | 0 | 106496 | t/o | 4121 | | 19.10s | t/o | 0.79s |
| Exponential_bug(6) | 6 | 983386 | 1203446 | 15601 | 1m18s | 56.75s | 0.96s |
| Exponential_bug(7) | 7 | 2250290 | 2833112 | 22841 | 3m13s | 2m26s | 1.46s |
| Exponential_bug(8) | 8 | 4844378 | 6158718 | 32313 | 7m15s | 5m28s | 2.23s |
| Exponential_bug(9) | 9 | 9896954 | 12526576 | 44428 | 15m29s | 11m48s | 3.22s |

Table 5. Comparison of the performance of the three tools when running on three parameterized benchmarks. The LB column and the t/o entry have the same meanings as in Table 2.

and Control_flow(N), N is the number of threads used in these benchmarks. Meanwhile, in Exponential_bug(N), N is the number of times a thread writes to a specific variable. In a similar way to Tables 2–4, Table 5 shows that TRACER always explores smaller numbers of executions than CDSChecker and Rcmc in all benchmarks, and it has the best performance. The differences between TRACER and the other tools becomes more clear when we increase the number of threads. For example, in Sigma(N) benchmark, the number of executions explored by TRACER increases only 8 times at each step, while it increases by 23 and 15 times for CDSChecker and Rcmc respectively. As a consequence, when there are more than 7 threads, the verification tasks of Sigma(N) are very challenging for CDSChecker and Rcmc but can still be handled by TRACER in less than 1 minute.

### 6.6 Evaluation using Concurrent Data Structure Benchmarks

We apply TRACER to fifteen concurrent data structure algorithms, namely Correia Ramalhete(N), Correia_Ramalhete_turn(N), TideX(N), TideX_nps(N), CLH_c11(N), CLH_rwlock_c11(N), MPSC_c11(N), Ticket_mutex_c11(N), Linux_lock(N), Barrier(N), Seqlock(N), MPMC_queue(N), MCS_lock(N), Cliffc_hashtable(N), Concurrent_hashmap(N), Chase_Lev_dequeue(N), and SPSC_queue(N). The first four algorithms are high-performance starvation-free mutual exclusion locks in [Ramalhete and Correia 2016, 2018]. Meanwhile, CLH_c11(N), CLH_rwlock_c11(N), and MPSC_c11(N) are C/C++11 implementations of CLH queue locks [Magnusson et al. 1994] and Ticket_mutex_c11(N) is a C/C++11 implementation of Ticket Lock [Mellor-Crummey and Scott 1991], taken from [Ramalhete and Correia 2018]. Other algorithms were used in the previous evaluations of CDSChecker [Demsky and Lam 2015; Norris and Demsky 2016] and Rcmc [Kokologiannakis et al. 2018]. Since these data structures can have huge state-spaces, we limit the number N of used threads in these algorithms to four. We also use loop unrolling when it is necessary. Moreover, to run all algorithms under the RA semantics, as it was done in §6.1–§6.5, we consider all write and read operations as acquire and release atomic accesses.

Table 6 gives the results on the four verification tasks derived from Linux_lock(N), Ticket_mutex_c11(N), and Correia_Ramalhete(N) algorithms. We observe that both TRACER and Rcmc have good performance on these benchmarks. TRACER always generates fewer executions than Rcmc and CDSChecker, which coincides with our previous observations in §6.1–§6.5.

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Proc. ACM Program. Lang., Vol. 2, No. OOPSLA, Article 135. Publication date: November 2018.
Table 6. Comparison of the performance of the three tools when running on concurrent data structure benchmarks. The LB column and the t/o entry have the same meanings as in Table 2.

| Program               | LB | Executions Explored | Total Running Time |
|-----------------------|----|---------------------|--------------------|
|                       |    | CDSChecker | Rcmc | Tracer | CDSChecker | Rcmc | Tracer |
| Linux_lock(2)         | 6  | 47          | 21   | 21     | 0.03s      | 0.02s | 0.01s  |
| Linux_lock(3)         | 6  | 14187799    | 412814| 412814 | 16m01s     | 36.36s| 33.21s |
| Ticket_mutex_c11(3)   | 10 | 8054        | 4026  | 4026   | 16m01s     | 50.76s| 0.25s  |
| Correia_Ramalhete(3)  | 5  | 5355        | 5355  | 5355   | 1.10s      | 0.77s | 0.92s  |

Table 7. Comparison of the performance of CDSChecker and Tracer when running on eight concurrent data structure benchmarks. The LB column and the t/o entry have the same meanings as in Table 2.

| Program               | LB | Executions Explored | Total Running Time |
|-----------------------|----|---------------------|--------------------|
|                       |    | CDSChecker | Tracer | CDSChecker | Rcmc | Tracer |
| Barrier(3)            | 10 | 62649       | 31944  | 5.45s      | 1.88s |
| Seqlock(3)            | 5  | 17792       | 14864  | 1.17s      | 0.77s |
| MPMC_queue(3)         | 5  | 621882      | 239254 | 2m11s      | 44.50s |
| MCS_lock(2)           | 5  | 92210       | 70072  | 8.91s      | 5.19s  |
| Cliffc_hashtable(4)   | 0  | 9520        | 4576   | 2.68s      | 0.94s  |
| Concurrent_hashmap(4) | 0  | 110         | 42     | 0.01s      | 0.00s  |
| Chase_Lev_deque(2) (deadlock) | 0 | 162306     | 20852  | 20.65s     | 2.10s  |
| SPSC_queue(2) (deadlock) | 3 | 754         | 57     | 0.18s      | 0.01s  |

more precise, Tracer and Rcmc explore the same number of executions in the four benchmarks and Tracer has a slightly better performance than Rcmc in 3 of 4 cases. In Correia_Ramalhete(3), the numbers of executions explored by CDSChecker, Rcmc, and Tracer are exactly the same, and Rcmc is slightly faster than CDSChecker and Tracer. Moreover, when we increase the number N of threads in Linux_lock(N) from 2 to 3, CDSChecker significantly increases the number of explored executions and the total running time. We note that the reported results on these benchmarks for Rcmc are similar to the results published in [Kokologiannakis et al. 2018]. The reported results for CDSChecker are not the same as the ones reported in [Norris and Demsky 2016] because all benchmarks in Table 6 are loop-unrolled and therefore we need not use the -m and -y arguments to control memory liveness and thread yield-based fairness for CDSChecker as in [Norris and Demsky 2016].

In the following, we show the performance of Tracer and CDSChecker on the remaining algorithms. We exclude Rcmc from the comparison since the tool currently cannot handle these algorithms [Personal-com 2018]. As it was done in §6.3, we let Tracer and CDSChecker exhaustively explore all possible executions even in the cases where they detect some errors.

Table 7 compares Tracer and CDSChecker on eight algorithms taken from [Norris and Demsky 2016]. We observe that both Tracer and CDSChecker handle these benchmarks quite well. In all examples, Tracer generates fewer executions than CDSChecker and has a better performance, which is also consistent with our previous observations. Interestingly, both tools can detect two deadlock errors in Chase_Lev_deque(2) and SPSC_queue(2). The former is due to our weakening of seq-cst atomic accesses used in the original benchmark to acquire and release atomic accesses. The later is a well-known known bug in [SPSC-bug 2008].

Table 8 compares Tracer and CDSChecker on the remaining six algorithms taken from [Ramalhete and Correia 2018]. In a similar way to Table 7, both tools handle the benchmarks quite
Table 8. Comparison of the performance of CDSChecker and Tracer when running on six concurrent data structure benchmarks. The LB column and the t/o entry have the same meanings as in Table 2.

| Program                                      | LB | CDSChecker Executions Explored | Tracer Executions Explored | Total Running Time |
|----------------------------------------------|----|--------------------------------|----------------------------|--------------------|
| Tidex(3)                                     | 5  | 4 567                          | 748                        | 0.37s 0.04s        |
| Tidex_nps(2)                                 | 10 | 10 257                         | 10 254                     | 6m15s 22.58s       |
| CLH_c11(3)                                   | 10 | 2 562                          | 732                        | 0.22s 0.06s        |
| CLH_rwlock_c11(3)                             | 10 | 20                             | 6                          | 0.01s 0.00s        |
| MPSC_c11(3)                                  | 10 | 12 937                         | 4 824                      | 1.20s 0.36s        |
| Correia_Ramalhete_turn(3) (mutex broken)      | 2  | 441 494                        | 96 184                     | 1m17s 12.78s       |

well. Moreover, we also observe that Tracer explores fewer executions than CDSChecker and has a better performance. Interestingly, although both tools generate almost the same number of executions in Tidex_nps(2), Tracer is significantly faster than CDSChecker. Finally, both tools can efficiently detect a violation of the mutual exclusion property in Correia_Ramalhete_turn(3). In the same way as the two deadlocks in Table 7, this violation is due to our weakening of seq-cst atomic accesses used in the original benchmark to acquire and release atomic accesses.

6.7 Conclusions of the Experiments

As expected, the experiments show that, for many benchmarks, the number of weak traces is significantly smaller than the number of total traces. On these examples, Tracer (which generates optimal numbers of executions w.r.t. weak traces), is much more efficient than CDSChecker and Rcmc. The results also show that Tracer has better performance and scales better to more extensive programs, even on benchmarks where the numbers of total and weak traces are equal, in which case Tracer explores the same number of executions as the other tools.

7 RELATED WORK

Since the pioneering work of Verisoft [Godefroid 1997, 2005] and CHESS [Musuvathi et al. 2008], stateless model checking (SMC), coupled with (dynamic) partial order reduction techniques (e.g. [Abdulla et al. 2014; Flanagan and Godefroid 2005; Rodriguez et al. 2015]) has been successfully applied to real life programs [Godefroid et al. 1998; Kokologiannakis and Sagonas 2017]. The method of [Abdulla et al. 2014] is optimal w.r.t. Mazurkiewicz traces (i.e., Shasha-Snir traces) under SC while our algorithm is designed for RA and it is optimal w.r.t. weak traces. With modifications, SMC techniques have been subsequently applied to the TSO and PSO weak memory models [Abdulla et al. 2015a; Demsky and Lam 2015; Zhang et al. 2015], and POWER [Abdulla et al. 2016b]. Common to all these approaches is that they explore at least one execution for each Mazurkiewicz [Mazurkiewicz 1986] or Shasha-Snir [Shasha and Snir 1988] traces. This induces a limit on the efficiency of the corresponding tools.

Several recent DPOR techniques try to exploit the potential offered by a weaker equivalence than Mazurkiewicz and Shasha-Snir traces [Chalupa et al. 2018; Huang 2015; Huang and Huang 2016; Norris and Demsky 2016]. Maximal causality reduction (MCR) is a technique based on exploring the possible values that reads can see, instead of the possible value-producing writes, as in our approach. MCR has been developed for SC [Huang 2015] and TSO and PSO [Huang and Huang 2016]. It may in some cases explore fewer traces than our approach, but it relies on potentially costly calls to an SMT solver to find new executions. In practice, MCR [Huang 2015] may not be optimal w.r.t. its partitioning (see [Chalupa et al. 2018]) while our algorithm is provably optimal w.r.t. weak traces. Moreover, it remains to be seen whether MCR can be adapted to the RA semantics, and how
it would compare. Chalupa et al. [2018] proposes a DPOR algorithm using a similar equivalence (that is based on \(po\) and \(rf\) relations) as our weak trace but under SC. Except for the minimal case of an acyclic communication graph [Chalupa et al. 2018], they may still explore a significant number of different executions with the same \(rf\) relation. Furthermore, checking the consistency of a trace under SC is an NP-complete problem [Chalupa et al. 2018]. As shown in our results, checking the consistency of a trace under the RA semantics can be efficiently done by polynomial time algorithms (cf. §4).

Recently, SMC was also adapted to (variants of) the C/C++11 memory model, which includes RA, producing the tools CDSCHECKER [Norris and Demsky 2016] and Rcmc [Kokologiannakis et al. 2018]. CDSCHECKER maintains a coherence order that need not be total, but not in an optimal way. It can generate inconsistent executions, which must afterward be validated. Rcmc has two options: Rc11 and Wrc11. Under Rc11, it maintains only total coherence orders; under Wrc11 it does not, which may generate RA-inconsistent executions (which are then not validated). Rcmc is optimal under the criterion of total coherence order, but only in the absence of RMW operations. In contrast, our technique is provably optimal w.r.t. weak traces, and including RMWs. On the other hand, CDSCHECKER and Rcmc also cover the different access modes of the C/C++11 memory model.

Bounded model checking can adapt to various weak memory models (e.g. [Alglave et al. 2013b,a; Torlak et al. 2010]). There is no report on using them for RA, but experiments for POWER [Abdulla et al. 2016b] concluded that Nidhugg is at least as efficient as [Alglave et al. 2013b].

Beyond SMC techniques for weak memory models, there have been many works related to the verification of programs running under weak memory models (e.g., [Abdulla et al. 2016a, 2017, 2015b; Atig et al. 2010; Burckhardt et al. 2007; Kuperstein et al. 2011; Liu et al. 2012]). Some works propose algorithms and tools for monitoring and testing programs running under weak memory models (e.g., [Burckhardt and Musuvathi 2008; Burnim et al. 2011; Liu et al. 2012]).

8 CONCLUSIONS AND FUTURE WORK

We have presented a new approach to defining DPOR algorithms which is optimal in the sense that it generates at most one trace with a given program order and read-from relation. We have instantiated the approach for the RA fragment of C/C++11. Our tool demonstrates that our method is substantially more efficient than state-of-the-art tools that handle the same fragment.

Although we only consider the RA semantics in this paper, we believe that our approach is general and can be extended to other memory models. For instance, we can extend the approach to the SRA (Strong RA) semantics [Lahav et al. 2016] by a modification of the sets of readable and visible events (cf. §4). It is interesting to see whether we can also handle the relaxed fragment of C/C++11 by employing a swapping mechanism for event speculations similar to the one we have proposed in this paper for treating postponed write events (cf. §5). For other models such as SC, our saturation scheme is necessarily not complete, since saturation for such models amounts to solving an NP-complete problem [Chalupa et al. 2018]. However, we believe that by maintaining saturated traces and only running the costly operations mentioned in [Chalupa et al. 2018] “by demand”, we can substantially improve efficiency even under the SC semantics.

ACKNOWLEDGMENTS

We thank Brian Demsky, Peizhao Ou, Konstantinos Sagonas, Michalis Kokologiannakis, Carl Leonardsson, Magnus Lång, and the OOPSLA’18 reviewers for their helpful feedback. This work was carried out within the UPMARC Linnaeus centre of excellence (Uppsala Programming for Multicore Architectures Research Center).
REFERENCES

P.A. Abdulla, S. Aronis, M. Faouzi Atig, B. Jonsson, C. Leonardsson, and K. Sagonas. 2015a. Stateless Model Checking for TSO and PSO. In *Tools and Algorithms for the Construction and Analysis of Systems, (TACAS)* (LNCS), Vol. 9035. Springer, London, UK, 353–367.

Parosh Abdulla, Stavros Aronis, Bengt Jonsson, and Konstantinos Sagonas. 2014. Optimal Dynamic Partial Order Reduction. In *Symposium on Principles of Programming Languages, (POPL)*. ACM, San Diego, CA, USA, 373–384.

Parosh Aziz Abdulla, Mohamed Faouzi Atig, Ahmed Bouajjani, and Tuan Phong Ngo. 2016a. The Benefits of Duality in Verifying Concurrent Programs under TSO. In *CONCUR 2016*. 5:1–5:15.

Parosh Aziz Abdulla, Mohamed Faouzi Atig, Ahmed Bouajjani, and Tuan Phong Ngo. 2017. Context-Bounded Analysis for POWER. In *TACAS 2017*. 56–74.

Parosh Aziz Abdulla, Mohamed Faouzi Atig, Bengt Jonsson, and Carl Leonardsson. 2016b. Stateless Model Checking for POWER. In *Computer Aided Verification, (CAV)* (LNCS), Vol. 9780. Springer, Toronto, ON, Canada, 134–156.

Parosh Abdulla, Mohamed Faouzi Atig, and Ngo Tuan Phong. 2015b. The Best of Both Worlds: Trading Efficiency and Optimality in Fence Insertion for TSO. In *ESOP 2015*. 308–332.

Jade Alglave, Daniel Kroening, Vincent Nimal, and Michael Tautschnig. 2013b. Software Verification for Weak Memory via Program Transformation. In *European Symposium on Programming, (ESOP)* (LNCS), Vol. 7792. Springer, Rome, Italy, 512–532.

Jade Alglave, Daniel Kroening, and Michael Tautschnig. 2013a. Partial Orders for Efficient Bounded Model Checking of Concurrent Software. In *Computer Aided Verification, (CAV)* (LNCS), Vol. 8044. Springer, Saint Petersburg, Russia, 141–157.

J. Alglave, L. Maranget, and M. Tautschnig. 2014. Herding Cats: Modelling, Simulation, Testing, and Data Mining for Weak Memory. *ACM Trans. Program. Lang. Syst.* Vol. 36, 2 (2014), 7:1–7:74.

Mohamed Faouzi Atig, Ahmed Bouajjani, Sebastian Burckhardt, and Madanlal Musuvathi. 2010. On the verification problem for weak memory models. In *POPL 2010*. 7–18.

S. Burckhardt, R. Alur, and M.M.K. Martin. 2007. CheckFence: checking consistency of concurrent data types on relaxed memory models. In *Programming Language Design and Implementation, (PLDI)*. ACM, San Diego, California, USA, 12–21.

S. Burckhardt and M. Musuvathi. 2008. Effective Program Verification for Relaxed Memory Models. In *Computer Aided Verification, (CAV)* (LNCS), Vol. 5123. Springer, Princeton, NJ, USA, 107–120.

J. Burnim, K. Sen, and C. Stergiou. 2011. Testing concurrent programs on relaxed memory models. In *International Symposium on Software Testing and Analysis, (ISSTA)*. ACM, Toronto, ON, Canada, 122–132.

Marek Chalupa, Krishnendu Chatterjee, Andreas PAVlogiannis, Nishant Sinha, and Kapil Vaidya. 2018. Data-centric dynamic partial order reduction. *PACMPL* 2, POPL (2018), 31:1–31:30.

M. Christakis, A. Gotovos, and K. Sagonas. 2013. Systematic Testing for Detecting Concurrency Errors in Erlang Programs. In *International Conference on Software Testing, Verification and Validation, (ICST)*. IEEE, Luxembourg, Luxembourg, 154–163.

Edmund M. Clarke, Orna Grumberg, Marius Minea, and Doron A. Peled. 1999. State Space Reduction Using Partial Order Techniques. *STTT* 2, 3 (1999), 279–287.

Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest, and Clifford Stein. 2009. *Introduction to Algorithms*. The MIT Press.

Brian Demsky and Patrick Lam. 2015. SATCheck: SAT-directed stateless model checking for SC and TSO. In *Object-Oriented Programming, Systems, Languages, and Applications, (OOPSLA)*. ACM, Pittsburgh, PA, USA, 20–36.

C. Planagan and P. Godefroid. 2005. Dynamic partial-order reduction for model checking software. In *Principles of Programming Languages, (POPL)*, ACM, Long Beach, California, USA, 110–121.

P. Godefroid. 1996. Partial-Order Methods for the Verification of Concurrent Systems: An Approach to the State-Explosion Problem. Ph.D. Dissertation. University of Liège. Also, volume 1032 of LNCS, Springer.

P. Godefroid. 1997. Model Checking for Programming Languages using VeriSoft. In *Principles of Programming Languages, (POPL)*. ACM Press, Paris, France, 174–186.

Patrice Godefroid. 2005. Software Model Checking: The VeriSoft Approach. *Formal Methods in System Design* 26, 2 (2005), 77–101.

P. Godefroid, B. Hammer, and L. Jagodeesan. 1998. Model Checking Without a Model: An Analysis of the Heart-Beat Monitor of a Telephone Switch using VeriSoft. In *Proc. ACM SIGSOFT International Symposium on Software Testing and Analysis*. 124–133.

Jeff Huang. 2015. Stateless model checking concurrent programs with maximal causality reduction. In *Programming Language Design and Implementation, (PLDI)*. ACM, Portland, OR, USA, 165–174.

Shiyou Huang and Jeff Huang. 2016. Maximal causality reduction for TSO and PSO. In *Object-Oriented Programming, Systems, Languages, and Applications, (OOPSLA)*. ACM, Amsterdam, The Netherlands, 447–461.

ISO. 2012. *C++ Specification Standard*. International Organization for Standardization. 1338 (est.) pages.
