# Maximal Causal Models for Sequentially Consistent Systems

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#### Abstract

This paper shows that it is possible to build a maximal and sound causal model for concurrent computations from a given execution trace. It is sound, in the sense that any program which can generate a trace can also generate all traces in its causal model. It is maximal (among sound models), in the sense that by extending the causal model of an observed trace with a new trace, the model becomes unsound: there exists a program generating the original trace which cannot generate the newly introduced trace. Thus, the maximal sound model has the property that it comprises all traces which all programs that can generate the original trace can also generate. The existence of such a model is of great theoretical value. First, it can be used to prove the soundness of non-maximal, and thus smaller, causal models. Second, since it is maximal, the proposed model allows for natural and causal-model-independent definitions of trace-based properties; this paper proposes maximal definitions for causal dataraces and causal atomicity. Finally, although defined axiomatically, the set of traces comprised by the proposed model are shown to be effectively constructed from the original trace. Thus, maximal causal models are also amenable for developing practical analysis tools.

#### 1 Introduction

Traces of events describing concurrent computations have been employed in a plethora of methods for testing and analyzing concurrent systems. A common characteristic of all these methods is that one uses an abstraction of a trace, i.e., a *model*, to "predict" (problematic) event patterns occurring in other traces abstracted by the model. Consider, for example, the conventional happens-before causality: if two conflicting accesses to an object are not causally ordered, then a data-race is reported [22]. But is this the best one can do? Of course, not. A series of papers propose more relaxed happens-before causal models where one can also permute blocks protected by the same lock, provided that they access disjoint variables [18], thus discovering new concurrency bugs not observable with plain happens-before. But is this the best one can do? Of course, not.

Thread 1	Thread 2	Execution	,	Thread 1	Thread 2	$\boldsymbol{E}$	xecutio
$sync(l)$ $\{$		1:		$sync(l)$ {		1:	
y = 1; x = 1; if $(x == 2)$ z = 1;		$y \leftarrow 1 \\ x \leftarrow 1 \\ x \rightarrow 1$					$x \leftarrow y \leftarrow$
z=1,		2:		x = 1;			$x \leftarrow$
	$sync(l)\{$		$\neg$	}		2:	
	x = 2;	$x \leftarrow 2$			$\operatorname{sync}(l)$ { if $(x > 0)$		$x \rightarrow 0$
	y=2;	$y \leftarrow 2$			y=2;		$y \leftarrow$
	(a)				} (b)		

Figure 1: Motivating examples.

Other papers propose models where one can also permute semantic blocks (whose actions are possibly generated by different threads) provided that each read access continues to correspond to the same write access [23, 25]. Others could go even further: Section 5 discusses a series of existing causal models (we only study sound models here, i.e., ones which only report real problems in the analyzed systems, allowing developers to focus on fixing those real problems and not on additionally sorting them out from false positives). We would naturally like to know whether there is an end to the question "Is this the best we can do?", that is, whether there is any causal model that can be associated to a given execution trace which comprises the maximum number of causally equivalent traces.

Although most runtime analysis techniques are built upon some underlying sound causal model, possibly relaxed for efficiency reasons, each effort seems to focus more on how to capture it efficiently rather than proving its soundness (often implicitly assumed) or studying its relationship to existing models (other than empirically comparing the number of found bugs). Moreover, since such approaches attempt to extract information from one observed trace and to find property violations, they actually deal with causal properties (e.g., causal datarace, causal atomicity), which are instances of desired system-wise properties that can be detected using only the causal information gathered from the observed trace. Since what can be inferred from a trace intrinsically depends on the chosen causal model, definitions of causal properties differ from technique to technique, with the undesirable effect that a causal property (e.g., a datarace) in one model might not be recognized by another model.

### 1.1 Motivating Examples

Each example in Figure 1(a) and (b) shows a two-threaded program, together with one of its possible executions, in which Thread 1 is executed completely before Thread 2 starts. In this representation of executions, synchronized blocks are boxed, while write and read operations on shared locations are denoted by

 $\leftarrow$  (receiving a value), and  $\rightarrow$  (yielding a value), respectively. Both *programs* exhibit a race condition between the two write operations on y. However, are the observed executions also exhibiting a causal datarace?

When analyzing the observed execution in Figure 1(a), a simple happens-before approach ordering all accesses to concurrent objects [22] cannot observe a causal datarace: the release operation of the lock in Thread 1 is required to happen-before the acquire of the lock in Thread 2. Happens-before with lock atomicity [18] is not able to infer a causal datarace either: although the lock atomicity would allow for the two lock-blocks to be permuted, the read of x in Thread 1 is still required to happen-before the write of x in Thread 2. Yet, the race condition can be captured as a causal datarace of the observed execution by weaker happens-before models [23, 25], since in those models, one can additionally permute a write before a read of same location, as long as it is permuted before the write corresponding to that read. Thus, the trace generated by the program in Figure 1(a) has or does not have a causal datarace, depending upon the particular causal model employed.

However, we were not able to find any existing (sound) causal model able to detect the race condition in Figure 1(b) as a causal datarace for the observed execution. The reason for this is that all models enforce at least the read-after-write dependency (i.e., a read should always follow the *latest write event* of the same variable), and therefore would not allow the permutation of the last two lock-blocks of the execution, since the read of x in Thread 2 must follow the last write of x in Thread 1. Nevertheless, there is enough information in the observed execution to be able to detect the race: since both writes of x in Thread 1 write the same value, it is actually possible to permute the last two lock blocks, and thus detect the race. Moreover, since one could conceive a technique specialized for finding such cases, it can be rightfully claimed that the observed execution has in fact a causal datarace, although not captured by any existing definition!

Given this ever increasing (regarding coverage) sequence of causal models and definitions for causal properties, it is only natural to ask the following questions:

Is there any causal model that generalizes all existing models, and which cannot be surpassed? Also, is there a unified definition for a causal property, which all present and future causal models can relate to?

We give positive answers to these questions in the context of sequential consistency [14]. While we believe the presented approach can be applied to other memory models, we chose sequential consistency here for three reasons: (1) it is broadly accepted, popular and intuitive; (2) it is subsumed by other memory models: errors detected under sequential consistency are also errors for other memory models; (3) recent research in computer architecture (e.g., [4]) shows that it actually can be efficiently supported and implemented in multiprocessor hardware, strengthening the applicability of our approach.

**Contributions.** Our first contribution is a novel axiomatization for concurrent computations, based on consistency and feasibility axioms, which yields sound (by

definition) causal models for observed executions. We then show the theoretical significance of a provable sound maximal model, as a means to unify existing definitions of causal properties, but also as a tool for proving soundness of runtime analysis techniques. We exemplify the latter by using our causal model to easily (re)prove the soundness for a series of existing causal models. The main result of the paper is that the obtained model is indeed the maximal causal model for the observed execution, in the sense that it comprises precisely all traces which can be generated by all programs which can generate the observed trace. Concretely, we show that: (1) all programs which can produce the observed execution can generate all traces in the model; and (2) for any trace not in the model there exists a program generating the observed trace which cannot generate it.

Comparison with past work. There has been a considerable amount of research on models and techniques to abstract executions for the purpose of inferring causally equivalent executions satisfying/violating particular but important properties, such as dataraces or atomicity/serializability [3, 9, 12, 17, 18, 21–24]. Section 5 discusses the relation between our model and the sound causal models upon which the above mentioned techniques were based [11, 18, 22, 23, 25]. Our axiomatic approach is closest in spirit to that of Netzer and Miller [17], which proposes an axiomatization of a happens-before causal order between memory accesses and semaphore operations. Instead, we directly axiomatize legal sequentially consistent execution traces.

Although our focus here is rather foundational, attempting to unify existing causal models and causal definitions of execution-dependent properties by building a provable maximally sound model to support them, this should not imply that axiomatic approaches to concurrent executions have an exclusively theoretical value. For example, Ganai and Gupta [10] apply a similar technique for softawre model checking, attempting to reduce the state space to be explored using sequential consistency constraints. Moreover, building on a previous draft of this paper [1], Said et al. [20] encoded the axioms of our proposed model (extended with constructs for thread creation and wait/notify) into an SMT solver and used that to effectively search the model for potential dataraces in Java programs.

Another interesting and productive line of research attempts to use information about the actual program code to either statically detect potential bad behaviors [7, 16], or to use information about the program and about the property to be checked to further relax the models of executions [5, 6]. Our approach is complementary to these, establishing a foundation on which code-based techniques can be developed.

Paper structure. Section 2 introduces some notation and discusses sequential consistency. Section 3 axiomatizes consistent concurrent systems and defines our proposed causal models. Section 4 uses these models to give uniform semantic definitions of trace-related properties, such as causal dataraces and atomicity. Section 5 shows how existing models are included in ours, thus proving their soundness. Section 6 presents a constructive characterization of the maximal

model. Section 7 formally defines the maximality claim and proves our model maximal among sound models, and Section 8 concludes.

### 2 Execution Model

Assume a machine that can execute arbitrarily many threads in parallel. The execution environment contains a set of *concurrent objects* (shared memory locations, locks, ...), which are accessed by threads to share data and synchronize. *Threads*, which can only interact through the execution environment, are abstracted as *sequences of operations on concurrent objects*. The only source of thread non-determinism is the execution environment, that is, if the interaction between a thread and the environment is the same across executions, the thread will execute the same operations, in the same order. To simplify the presentation, we assume no dynamic creation of threads (this presents no technical difficulty).

#### 2.1 Concurrent Objects, Serial Specification

We adopt the Herlihy and Wing [13] definition of concurrent objects and serial specifications. A concurrent object is behaviorally defined through a set of atomic operations, which any thread can perform on it, and a serial specification of its legal behavior in isolation. The serial specification describes the valid sequences of operations which can be performed on the object. We next describe two common types of concurrent objects.

Shared memory locations. Each shared memory location can be regarded as a shared object with read and write operations, whose serial specification states that each read yields the same value as the one of the previous write. Moreover, to avoid non-determinism due to the initial state of the memory, we will further require that all memory locations are initialized, that is, the first operation for each location is a write.

**Mutexes.** Each mutex can be regarded as a concurrent object providing *acquire* and *release* operations. Their mutual exclusion property is achieved through the serial specification which accepts only those sequences in which the difference between the number of *acquire* and *release* operations is either 0 or 1 for each prefix, and all consecutive pairs of *acquire-release* share the same thread.

To keep the proofs simple and the concepts clear we refrain here from adding more concurrency constructs (such as spawn/join, wait/notify, or semaphores). Note, however, that this would not introduce additional complexity, but just constrain further the notion of consistency.

#### 2.2 Events and Traces

Operations performed by threads on concurrent objects are recorded as *events*. We consider events to be abstract entities from an infinite "collection" *Events*, and

describe them as tuples of attribute-value pairs. The only attributes considered here are: thread—the unique id of the thread generating the event, op—the operation performed (e.g., write, read, acquire, or release), target—the concurrent object accessed by the event, and data—the value sent/received by the current event, if such exists (e.g., for the write/read operations). For example,  $(thread=t_1, op=write, target=x, data=1)$  describes an event recording a write operation by thread  $t_1$  to memory location x with value 1. When there is no confusion, we only list the attribute values in an event, e.g.,  $(t_1, write, x, 1)$ . For any event e and attribute attr, attr(e) denotes the value corresponding to the attribute attr in e, and e[v/attr] to denote the event obtained from e by replacing the value of attribute attr by v. An  $execution\ trace$  is abstracted as a sequence of events. Given a trace  $\tau$ , a concurrent object e0 and a thread e1, let e2 and e3 and e4 denote the restriction of e4 to events involving only e6, and only e7, respectively. Let e8 be the latest event of e9 having the e9 attribute e6. If e6 is omitted, it simply means the latest event in e7.

Sequential consistency can be now elegantly defined:

#### Definition 1 (Attiya and Welch [2]) Let $\tau$ be any trace.

- (1)  $\tau$  is **legal** if and only if  $\tau \upharpoonright_o$  satisfies o's serial specification for any object o;
- (2) An interleaving of  $\tau$  is a trace  $\sigma$  such that  $\sigma \upharpoonright_t = \tau \upharpoonright_t$  for each thread t.
- (3) A trace  $\sigma$  is (sequentially) consistent if it admits a legal interleaving.

Since we restrict ourselves to sequential consistency, from here on when we say that a trace is sequentially consistent we automatically mean that it is also legal.

### 3 Feasibility Model

This section introduces a novel axiomatization for a machine producing consistent executions, and uses it to associate a sound-by-definition causal model to any observed execution, comprising all executions which can potentially be inferred from that execution alone, without additional knowledge of the system generating it.

Figure 2 highlights the two major concepts underlying our approach, namely consistent traces and feasible executions. A consistent trace (Definition 1) disallows "wrong" behaviors, such as reading a value different from the one which was written, or proceeding when a lock cannot be acquired. Our novel notion, that of feasible executions, refers to sets of execution traces and aims at capturing all the behaviors that a given system or program can manifest. No matter what task a concurrent system or program accomplishes, its set of traces must obey some basic properties. First, feasible traces are generate-able, meaning that any prefix of any feasible trace is also feasible; this is captured by our first axiom of feasible traces, prefix closedness. Second, we assume that thread interleaving is the only source of non-determinism in producing traces; this is captured by our second axiom of feasible traces, local determinism.

Each particular multithreaded system or programming environment, say  $\mathcal{S}$ , has its own notion of feasible execution, given by its specific intended semantics. Let us call all (possibly incomplete) traces that  $\mathcal{S}$  can yield  $\mathcal{S}$ -feasible, and let

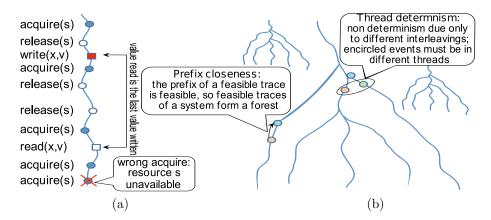


Figure 2: Feasibility model: (a) trace consistency; and (b) feasible executions.

feasible(S) be their set. Instead of defining feasible(S), which requires a formal definition of S and is therefore S-specific (and tedious), we here axiomatize it by what we believe are its crucial properties:

<u>Prefix Closedness:</u> Events are indivisible and generated in execution order; hence, feasible(S) must be prefix closed: if  $\tau_1\tau_2$  is S-feasible, then  $\tau_1$  is S-feasible.

**Local Determinism:** The execution of a concurrent operation is determined by the previous events in the same thread, and can happen at any consistent moment after them. Formally, if  $\tau e, \tau' \in feasible(\mathcal{S})$  and  $\tau \upharpoonright_{thread(e)} = \tau' \upharpoonright_{thread(e)}$  then: if  $\tau'e$  is consistent then  $\tau'e \in feasible(\mathcal{S})$ ; moreover, if op(e) = read and there exists an event e' such that e = e'[data(e)/data] and  $\tau'e'$  is consistent, then  $\tau'e \in feasible(\mathcal{S})$ . The second part says that if a read operation is enabled, i.e., all previous events have been generated, then it can be executed at any consistent time (despite the fact that the value it receives might be different from that observed in the original trace).

**Definition 2** S is consistent if and only if feasible (S) satisfies the axioms above.

A major goal of trace-based analysis is to infer/analyze as many traces as possible using a recorded trace. When one does not know (or does not want to use) the source code of the multithreaded program being executed, one can only infer potential traces of the system resembling the observed trace. Let us now define the proposed causal model, termed *feasibility closure*, as the set of executions which can be inferred from an observed execution—they correspond to the traces obtainable from  $\tau$  using the feasibility axioms.

**Definition 3** The **feasibility closure** of a consistent trace  $\tau$ , written feasible( $\tau$ ), is the smallest set of traces containing  $\tau$  which is prefix-closed and satisfies the local determinism property. A trace in feasible( $\tau$ ) is called  $\tau$ -feasible.

The following result formalizes the soundness of the proposed model, showing how closure properties guarantee that all traces in our causal model are feasible. Moreover, it shows that *any* system/program which can generate one trace, can also generate *all* traces comprised by its causal model.

**Proposition 1** If S consistent and  $\tau \in feasible(S)$  then  $feasible(\tau) \subseteq feasible(S)$ . Moreover, if  $\sigma$  is consistent and  $\tau \in feasible(\sigma)$ , then  $feasible(\tau) \subseteq feasible(\sigma)$ .

PROOF. Both feasible(S) and  $feasible(\sigma)$  are closed under the feasibility axioms. Since  $\tau$  belongs to both of them, and  $feasible(\tau)$  is the smallest set closed under the same axioms, it follows that it must be included in both.

The intuition for  $\tau \in feasible(\sigma)$  is that if a run of any program executed on  $\mathcal{S}$  can produce  $\sigma$ , then there is also some run of the same program executed also on  $\mathcal{S}$  that can produce  $\tau$ . Since  $feasible(\sigma)$  was chosen to be the smallest set of traces closed under the axioms above, it follows, also intuitively, that if  $\tau \notin feasible(\sigma)$  then there is some program that yields  $\sigma$  but which cannot yield  $\tau$ . This maximality claim will be formalized and proved in Section 7.

Therefore, observing an execution trace  $\tau$ ,  $feasible(\tau)$  comprises all traces that can be obtained by all programs which can produce  $\tau$ .

As shown later, Proposition 1 can be further strengthened using the characterization of the feasibility closure, in the sense that if  $\tau$  is a consistent interleaving of  $\sigma$  then  $feasible(\sigma) = feasible(\tau)$  (Corollary 1). This basically means that the feasibility closure does not depend on the representative legal trace chosen for an observed sequentially consistent trace  $\sigma$  in Definition 1, and thus it can be rightfully called the *legal feasibility closure* of the sequentially consistent trace  $\sigma$ .

### 4 Formal Definitions for Causal Properties

In this section we revisit some definitions of causal properties for traces, namely dataraces and atomicity, and claim that the proposed model make those uniform and consistent for models. Indeed, having a maximal model comprising all possible sound models allows for unique semantical definitions which can be shared among all such models.

#### 4.1 Dataraces

Two events have a data conflict if they belong to different threads, both access the same memory location, and at least one access is a write. In our notation, events  $e_1$  and  $e_2$  have a data conflict if  $thread(e_1) \neq thread(e_2)$ ,  $target(e_1) = target(e_2)$ , and  $write \in \{op(e_1), op(e_2)\}$ . A datarace occurs when an execution contains two events having a data-conflict, without proper synchronization between them [22]. An obvious datarace between events  $e_1$  and  $e_2$  in a consistent trace  $\tau$  can be observed when the two data-conflicting events  $(e_1$  and  $e_2)$  are consecutively generated (i.e.,  $\tau = \tau_1 e_1 e_2 \tau_2$ ), so the second part of the definition above is trivially satisfied. However, this definition, although "model independent", is rather restrictive, since the chances of noticing the two accesses occurring consecutively

are really low. For this reason, the notion of *causal datarace* is more appropriate. Informally, an execution admits a causal datarace between two memory accesses if the two accesses could have been executed concurrently under an alternative scheduling, inferable from the observed execution.

As previously discussed in Section 1.1, many techniques have been proposed for finding causal dataraces. However, the formal definition of a datarace for an execution in such a model is typically operational, that is, constrained by the model itself. For example, in the techniques based on happens-before, the datarace is defined as two events having a data conflict which are not ordered by the happens-before causal order induced by the observed execution [22]; if considering happens before with locksets [18], the happens-before ordering is relaxed to only order memory location accesses, with the additional requirement that the lock-protected blocks be maintained atomic. Therefore, each causal model encountered in the literature defines its own model-dependent definition for a causal datarace, to take full advantage of its particularities.

Since, as shown in Section 7, our causal model is maximal, we can precisely give a definition of causal datarace which only depends on the observed execution:

**Definition 4** A trace  $\tau = \tau_1 e_1 \tau_2 e_2$  admits a **causal datarace** on data-conflicting events  $e_1$  and  $e_2$  if and only if there exists a  $\tau$ -feasible trace  $\sigma$  such that  $\sigma \upharpoonright_{thread(e_1)} = \tau_1 \upharpoonright_{thread(e_1)}$  and  $\sigma \upharpoonright_{thread(e_2)} = (\tau_1 e_1 \tau_2) \upharpoonright_{thread(e_2)}$ .

The above definition states that we can predict a data race from an observed trace  $\tau$  if there exists a  $\tau$ -feasible trace which makes the data race apparent. We chose as a witness a trace stopped at the moment when both threads are about to execute the events in a race. This is indeed a clear witness for the race, since, by the local determinism axiom, the execution of the conflicting operations is allowed to proceed in any order from this point; moreover this saves us the trouble of specifying that the value of an event involved in the trace might change if it corresponds to a *read* operation. Nevertheless, one should note that, unlike in other causal models, the witness traces containing the events in an observable race, in both orderings, are also part of the feasibility closure of  $\tau$ .

With our causal datarace definition, the race in Figure 1(b) is finally captured, having (1, acquire, l) (1, write, x, 1) (1, release, l) (2, acquire, l) (2, read, x, 1) as a witness belonging to the feasibility closure of the observed execution.

#### 4.2 Atomicity

Assume the existence of an additional concurrent object, named transaction monitor, with two operations begin and end and the serial specification requiring that for each thread the first transaction monitor operation is a begin and, for each thread, there are no two transaction begin operations without a transaction end between them. That is, transaction monitors are similar to but weaker than locks, in the sense that the mutual exclusion is not enforced, although desired. A transaction of a consistent trace  $\tau$  is then a subsequence of events  $\sigma$  of  $\tau$  having the same thread, starting with a transaction begin operation and ending with the next transaction end operation.

Within this framework, one can either define global atomicity, which amounts to serializability of transactions [25], or local atomicity [8], which requires each transaction be serializable, but not necessarily the entire execution.

**Definition 5** A transaction  $\sigma$  of  $\tau$  is **atomic** for consistent trace  $\tau$  if there exists a  $\tau$ -feasible trace  $\tau_1 \sigma \tau_2$ .  $\tau$  is **locally atomic** if each of its transactions are atomic for  $\tau$ .  $\tau$  is **(globally) atomic**, or **serializable**, if there exists a  $\tau$ -feasible trace  $\tau'$  such that each transaction  $\sigma$  of  $\tau$  is a contiguous subsequence of  $\tau'$ .

Although these definitions are similar to those found in the above mentioned papers, they are now model-independent. Being defined using the maximal causal model, they become universal, applicable to all conceivable sound causal models.

### 5 Relationship with Existing Models

In this section we analyze the relationships between our model and other existing (sound) models for (consistent) multithreaded computations. Proving that existing models are faithfully captured by our model shows that these rather ad hoc (from a theoretical perspective) models are indeed sound. We start with the following result, which can be regarded as a sufficient criterion for feasibility:

**Theorem 1** Any consistent prefix  $\sigma_1$  of an interleaving  $\sigma_1\sigma_2$  of  $\tau$  is  $\tau$ -feasible.

PROOF. Induction on the length of the interleaving prefix. The base case is trivial. Let  $\tau'e$  be a consistent interleaving prefix of  $\tau$ , and assume that  $\tau'$  is  $\tau$ -feasible. Let t = thread(e), and let  $\tau_1, \tau_2$  be such that  $\tau = \tau_1 e \tau_2$ . By prefix closedness, it follows that  $\tau_1 e$  is feasible. Moreover, since  $(\tau'e)|_t = \tau'|_t e$  is a prefix of  $\tau|_t$ , it follows that  $\tau'|_t = \tau_1|_t$ . Using the local determinism for  $\tau_1 e$  and  $\tau'$ , we obtain that  $\tau'e$  is  $\tau$ -feasible (since it is consistent).

The remainder of this section shows that existing sound causal models are captured by the feasibility closure as simple instances of Theorem 1.

#### 5.1 Happens Before Relation on Mazurkiewicz Traces

One elegant way to capture the happens-before trace equivalence is the Mazurkiewicz trace associated to the dependence given by the happens-before relation [11].

The happens-before dependence is a set  $T \cup D$ , where  $T = \bigcup_t \{(e_1, e_2) : \tau|_t = \tau_1 e_1 e_2 \tau_2\}$  is the intra-thread sequential dependence relation and  $D = \bigcup_x \{(e_1, e_2) : \tau|_x = \tau_1 e_1 e_2 \tau_2 \text{ such that } e_1 \text{ or } e_2 \text{ is a write of } x\}$  is the sequential memory dependence relation. Given this happens-before dependence, the Mazurkiewicz trace associated with  $\tau$  is defined as the least set  $[\tau]$  of traces containing  $\tau$  and being closed under permutation of consecutive independent events [15]: if  $\tau_1 e_1 e_2 \tau_2 \in [\tau]$  and  $(e_1, e_2) \not\in T \cup D$ , then  $\tau_1 e_2 e_1 \tau_2 \in [\tau]$ .

The following result shows that the feasibility closure is closed under the equivalence relation generated by happens-before, that is, happens-before is captured by our model, and thus re-shown sound for consistent executions:

**Proposition 2** If  $\tau_1 e_1 e_2 \tau_2$  is  $\tau$ -feasible and  $(e_1, e_2) \notin T \cup D$ , then  $\tau_1 e_2 e_1 \tau_2$  is  $\tau$ -feasible. Given any  $\tau$ -feasible trace  $\tau'$ ,  $[\tau'] \subseteq feasible(\tau)$ . Hence,  $[\tau] \subseteq feasible(\tau)$ .

PROOF. Let  $\tau_1e_1e_2\tau_2$  be a  $\tau$ -feasible trace such that  $(e_1,e_2) \notin T \cup D$ . We will show that  $\tau_1e_2e_1\tau_2$  is also  $\tau$ -feasible. First, all prefixes of  $\tau_1e_1e_2\tau_2$ , including  $\tau_1$ ,  $\tau_1e_1$ ,  $\tau_1e_1e_2$ ,  $\tau_1e_1e_2\tau_2'e_2'$  (for any prefix  $\tau_2'e_2'$  of  $\tau_2$ ), are  $\tau$ -feasible, since  $feasible(\tau)$  is prefix closed. Now, we can iteratively use closedness under local determinism (1) for  $\tau_1e_1e_2$  and  $\tau_1$ , to derive that  $\tau_1e_2$  is  $\tau$ -feasible; (2) for  $\tau_1e_1$  and  $\tau_1e_2$  to derive that  $\tau_1e_2e_1$  is also  $\tau$ -feasible; (3) by finitary induction for each prefix  $\tau_2'e_2'$  of  $\tau_2$ , for  $\tau_1e_1e_2\tau_2'e_2'$  and  $\tau_1e_2e_1\tau_2'$  to derive that  $\tau_1e_2e_1\tau_2'e_2'$  is also  $\tau$ -feasible.

Therefore, for any  $\tau$ -feasible trace  $\tau'$ ,  $feasible(\tau')$  is closed under permutation of consecutive independent events; hence,  $[\tau'] \subseteq feasible(\tau') \subseteq feasible(\tau)$ .

#### 5.2 Weak Happens Before

Several more recent trace analysis techniques [23, 25] argue that the happensbefore model can be further relaxed, noticing that the only purpose of the write-after-read happens-before order is to guarantee that a read event always reads the same write event as before in any feasible interleaving of the original trace. Therefore, one only needs to preserve the read-after-write dependence:

**Definition 6** Suppose  $\tau = \tau_1 e_1 \tau_2 e_2 \tau_3$ . Then  $e_2$  write-read depends on  $e_1$  in  $\tau$ , written  $e_1 <_{\tau}^{wr} e_2$ , if  $target(e_1) = target(e_2)$ ,  $op(e_1) = write$ ,  $op(e_2) = read$ , and for all  $e \in \mathcal{E}_{\tau_2}$ , either  $target(e) \neq target(e_1)$ , or  $op(e) \neq write$ .

That is,  $e_1 <_{\tau}^{wr} e_2$  iff the value read by  $e_2$  is the value written by  $e_1$ .

Sen et al. [23] introduce the notion of atomic sets associated to each write event, containing itself and all read events which write-read depend on it, accepting as feasible executions all linearizations of the transitive closure of the combined  $<_{\tau}^{wr}$  and thread ordering, satisfying the additional requirement that the atomic sets are preserved. However, as noticed by Wang and Stoller [25], this can be simply restated as follows:

**Definition 7**  $\tau \sim \sigma$  if  $\tau$  is an interleaving of  $\sigma$  and  $<_{\tau}^{wr} = <_{\sigma}^{wr}$ .

That is, the  $\sim$ -equivalence class of  $\tau$  contains all interleavings of  $\tau$  which have exactly the same write-read dependence relation. Next result shows that this model is also captured by our model.

**Proposition 3** If  $\sigma_1$  is  $\tau$ -feasible, and  $\sigma_1 \sim \sigma_2$ , then  $\sigma_2$  is also  $\tau$ -feasible.

PROOF. We show that we are in the conditions of Theorem 1: Since  $\sigma_1$  is consistent, and  $<_{\sigma_2}^{wr} = <_{\sigma_1}^{wr}$ , it follows that  $\sigma_2$  must also be consistent, since all read events follow the same write events as in the  $\sigma_1$ , which, by the consistency of  $\sigma_1$ , precisely implies that each read event returns the value of the previous write event.

#### 5.3 Happens-Before with synchronization

A conservative, sound, and requiring no implementation changes approach to handling locks in happens-before-based trace analysis techniques is to assume that acquire and release operations on the same lock yield the same happens-before dependence as if they were particular write and read operations (on the lock variable) [22]. However, this prevents synchronized blocks from being permuted, and thus imposes coverage limitations. The lock-set approaches, also called hybrid happens-before [18], propose to handle locks separately, associating to each event the set of locks [21] protecting them, hereby not enforcing any particular order between synchronized blocks.

We here group the events protected by locks in atomic blocks. Events  $e_1$  and  $e_2$  from a consistent trace  $\tau$ , both generated by thread t, are l-atomic in  $\tau$ , written  $e_1 \updownarrow_l^{\tau} e_2$ , if and only if there is some acquire event e on lock l generated by t before both  $e_1$  and  $e_2$ , and there is no release event e' on l generated by t between e and either of  $e_1, e_2$ . For each lock l, let  $[e]_l$  denote the l-atomic equivalence class of e. A trace  $\tau'$  is consistent with the lock atomicity of  $\tau$  if there exists no lock l and decomposition  $\tau_1 e_1 \tau_2 e_2 \tau_3 e_3 \tau_4 e_4 \tau_5$  such that  $e_1 \updownarrow_l^{\tau} e_3$  and  $e_2 \updownarrow_l^{\tau} e_4$  and  $[e_1]_l \neq [e_2]_l$ . Let  $\prec_{hb}^{\tau}$  be the transitive closure of the union between happens-before and thread orderings of  $\tau$ . The following holds:

**Proposition 4** Let  $\sigma$  be a  $\tau$ -feasible trace. Any linearization of  $\prec_{hb}^{\sigma}$  consistent with the lock atomicity of  $\sigma$  is  $\tau$ -feasible.

PROOF. Again, we reduce our proof to Theorem 1. First, any linearization of  $\prec_{hb}^{\sigma}$  is an interleaving of  $\sigma$ . Moreover, since  $\sigma$  is consistent, preservation of happens-before ensures that the serial specification of the memory locations is satisfied. Finally, consistency with lock atomicity implies that the serial specification for mutexes is also satisfied. Therefore, any linearization of  $\prec_{hb}^{\sigma}$  consistent with the lock atomicity of  $\sigma$ , is a consistent interleaving of  $\sigma$ , thus  $\sigma$ -feasible.

#### 5.4 Weak-Happens-Before with synchronization

We next present two approaches to handling synchronization in weak-happensbefore models and show they are both embeddable in our model.

Lock atomicity via write-read atomicity [23]. Since the notion of write-read atomicity already allows atomic sets to be permuted, it seems reasonable to use the conservative idea from standard happens-before methods, and treat acquire as a write event and release as a read event. Formally, given the consistent trace  $\tau$ , one could additionally introduce an atomic dependence relation  $<^{\pi}_{\tau}$  given by  $e_1 <^a_{\tau} e_2$  if  $\tau = \tau_1 e_2 \tau_2 e_2 \tau_3$ ,  $target(e_1) = target(e_2)$ ,  $op(e_1) = acquire$ ,  $op(e_2) = release$ , and there is no event e in  $\tau_2$  such that  $target(e) = target(e_1)$ , and op(e) = acquire. With this definition, equivalent traces to an observed trace  $\tau$  are those interleavings of  $\tau$  having the same write-read and atomic dependencies.

However, this definition needs a careful approach. Consider the example in Figure 1(b), and suppose that we observe a similar execution, but that the program is stopped after the read of x in Thread 2. Since no release event has been generated, the acquire in Thread 2 has no event depending on it, and thus it can be permuted (without the read event on x it was supposed to protect) before the last lock-block of Thread 1. Then, the final read of x itself can be permuted past the final release of t in Thread 1, exhibiting a spurious causal datarace.

Nevertheless, these models are sound for *synchronization complete* traces, that is, traces in which each acquired lock is eventually released.

**Proposition 5** Let  $\sigma$  be a synchronization complete  $\tau$ -feasible trace. Any interleaving  $\sigma'$  of  $\sigma$  satisfying that  $<_{\sigma'}^{wr} = <_{\sigma}^{wr}$  and  $<_{\sigma'}^{a} = <_{\sigma}^{a}$  is  $\tau$ -feasible.

PROOF. Since we already shown that  $<_{sigma'}^{wr} = <_{\sigma}^{wr}$  implies that the serial specification of memory locations is verified, we only need to show that  $<_{\sigma'}^a = <_{\sigma}^a$  implies the satisfaction of the mutex specification for synchronization complete traces, that is, that any prefix of  $\sigma'$  has at most one more acquire operations than release operations, and all consecutive pairs of acquire-release have the same thread. The second part is easily guaranteed by the fact that  $<_{\sigma'}^a = <_{\sigma}^a$ , since  $<^a$  enforces the acquire-release in relation are consecutive, and, since  $\sigma$  is consistent, this definition additionally implies that they have the same thread. The first part comes from the fact that, since  $\sigma$  is synchronization complete, every acquire has a corresponding release, with whom it is in the  $<_{\sigma}^a$  relation.

Lock atomicity via locksets. Wang and Stoller [25] propose a weak-happens-before model based on write-read dependence, while using locksets to handle locks as individual objects. In this model, a trace  $\tau'$  is equivalent with a consistent trace  $\tau$  if  $\tau'$  is an interleaving of  $\tau$  having the same write-read dependence relation and being consistent with the lock atomicity of  $\tau$ .

**Proposition 6** Let  $\sigma$  be a  $\tau$ -feasible trace. Any interleaving  $\sigma'$  of  $\sigma$ , consistent with the lock atomicity of  $\sigma$  and satisfying that  $<_{\sigma'}^{wr} = <_{\sigma}^{wr}$  is  $\tau$ -feasible.

PROOF. From the proof of Proposition 3,  $<_{\sigma'}^{wr} = <_{\sigma}^{wr}$  implies the serial specification of memory locations is obeyed in  $\sigma'$ . Additionally, from the proof of Proposition 4, consistency with the lock atomicity of a consistent trace implies that the serial specification of mutexes is obeyed. We can therefore apply Theorem 1.

## 6 Characterizing the Feasibility Closure

Section 3 showed how our causal model can be naturally defined by characterizing feasibility axiomatically rather than constructively. Closure axioms guarantee that all equivalent traces which can be derived based on the consistency axioms are considered. This section presents a constructive characterization of the feasibility closure.

As might have been suggested by Theorem 1, consistent interleaving prefixes cover all possibilities of generating  $\tau$ -feasible traces using only the events in  $\tau$ . However, the definition of interleaving (prefix) overlooks the final part of the local determinism axiom, that is, the one regarding operations which might receive different values from their objects. To achieve a complete constructive characterization of the feasibility closure, we have to go beyond prefixes of interleavings, more exactly, one read operation per thread beyond. This is because, as guaranteed by local determinism, whenever all events before an event have been generated in a thread, the operation on the concurrent object specified in that event can also take place, but its data attribute might now retrieve a value different from the one it had in the observed trace. However, once such an event whose data is different from the one in the original trace is derived, the execution cannot be continued for that thread, because that event might influence/prevent the generation of the following events. An extended interleaving prefix is a (partial) trace which behaves similarly to the observed trace up to its final event for each thread, which might have a different value:

**Definition 8** Trace  $\tau' = \tau_1 e'$  is an **extended prefix** of  $\tau = \tau_1 e \tau_2$  if either e = e', or op(e) = read and e = e'[data(e)/data].  $\tau'$  is an **extended interleaving prefix** of  $\tau$  if  $\tau'|_{t}$  is an extended prefix of  $\tau|_{t}$  for any thread t.

We can now give a complete constructive characterization for  $\tau$ -feasible traces:

**Theorem 2** Given  $\tau$  consistent, a trace  $\tau'$  is  $\tau$ -feasible iff it is a consistent extended interleaving prefix of  $\tau$ .

PROOF. Proving that any consistent extended interleaving prefix of  $\tau$  is  $\tau$ -feasible proceeds similarly to the proof of Theorem 1. For the reverse, one needs to show that the set of consistent extended interleaving prefixes of  $\tau$  contains  $\tau$ , is prefix closed, and closed under local determinism. First two are obvious:  $\tau$  is an interleaving prefix of itself, and any prefix of an extended interleaving prefix of  $\tau$  is an extended interleaving prefix of  $\tau$  by the definition. Now let  $\tau_1 e$  and  $\tau_2$  be consistent interleaving prefixes of  $\tau$  such that thread(e) = t, and  $\tau_1 \upharpoonright_t = \tau_2 \upharpoonright_t$ . Since  $(\tau_1 e) \upharpoonright_t$  is an extended prefix of  $\tau \upharpoonright_t$ , then either  $(\tau_1 e) \upharpoonright_t$  is a prefix of  $\tau$ , or op(e) = read, and there exists e', such that thread(e') = n, op(e') = read, target(e') = target(e), and target(e') is a prefix of target(e') = target(e), and target(e') is a prefix of target(e') = target(e), and target(e') is a prefix of target(e') = target(e), and target(e') is a prefix of target(e') = target(e).

Let e'' be e, if  $op(e) \neq read$ , or  $e'' = e[data(e^w)/data]$ , if  $e^w = latest_{write}(\tau_2 \upharpoonright_{target(e)})$ . Then  $\tau_2 e''$  is an extended interleaving prefix. If  $op(e) \neq read$ , and  $\tau_2 e$  is consistent, then it also is a consistent extended interleaving prefix. If op(e) = read, then, since  $\tau_2 e''$  is consistent (by the choice of e''), the property follows.

An important corollary of Theorem 2 is that the feasibility closure does not depend on the legal trace chosen to represent a sequentially consistent trace, as it contains all the representative legal traces.

Corollary 1 If  $\tau$  is a consistent interleaving of consistent trace  $\sigma$  then feasible( $\tau$ ) = feasible( $\sigma$ ).

```
Input: Trace \tau_0 of size n over k threads.
    Maps: thread : \{1, \ldots, k\} \rightarrow Stack
                \sigma: Locations \rightarrow \mathit{Int}
    Initial:
                \sigma[x] \leftarrow \perp, for all variables and mutexes
                thread[t] \leftarrow \tau_0 \upharpoonright_t, for all threads
                 Advanceable \leftarrow \{1, \dots, k\}
 1 \tau \leftarrow \epsilon; t \leftarrow 0;
 2 while t < k do
         t \leftarrow t + 1;
 3
         if t \in Advanceable then
 4
               e \leftarrow \mathsf{top}(\mathsf{thread}[t]);
 5
               if (op(e) \neq acquire \lor \sigma[target(e)] = \bot) \land (op(e) \neq read \lor )
 6
               \sigma[target(e)] \neq \bot) then
                                                                                                // advance
                    l \leftarrow target(e);
 7
                    if op(e) = read \wedge data(e) \neq \sigma[l] then
                                                                                 // extended prefix
 8
                         Advanceable \leftarrow Advanceable \setminus \{t\};
 9
                         data(e) \leftarrow \sigma[l];
10
11
                    else
                                                                                       // update state
                         pop(thread[t]);
12
                         if op(e) = write then \sigma[l] \leftarrow data(e);
13
                         if op(e) = acquire then \sigma[l] \leftarrow \sigma[l] - 1;
14
                         if op(e) = release then \sigma[l] \leftarrow \sigma[l] + 1;
15
16
                    push(\tau,e); check \tau against \varphi;
17
                    t \leftarrow 0;
18
              end
19
20
         end
         while t = \mathbf{k} \wedge \tau \neq \epsilon do
                                                                                            // backtrack
21
               e \leftarrow pop(\tau); t \leftarrow thread(e); l \leftarrow target(e);
22
               if t \notin Advanceable then
                                                                                  // extended prefix
23
                    Advanceable \leftarrow Advanceable \cup \{t\};
\mathbf{24}
25
               else
                                                                                     // restore state
                    push(thread[t],e);
26
                    if op(e) = write then \sigma[l] \leftarrow data(latest(\tau \upharpoonright_l));
27
                    if op(e) = acquire then \sigma[l] \leftarrow t;
28
                    if op(e) = release then \sigma[l] \leftarrow \perp;
29
              end
30
         end
31
32 end
```

Algorithm 1: Model checking the feasibility closure

Model checking the feasibility closure. Algorithm 1 can be used to explore (and check properties against) the feasibility closure of a given trace. It takes as input a trace  $\tau_0$  and a procedure  $\varphi$  saying whether a property is satisfied by a (partial) trace (and state), and checks whether all traces in the feasibility closure of  $\tau_0$  (and their corresponding states) satisfy the property of  $\varphi$ . In the initialization phase, the original trace is split into threads and each thread projection is loaded into a stack, with first events in the thread at top of the stack, and the store initialized with  $\perp$  for both variables and mutexes. We additionally maintain a set of enabled threads (Advanceable), that is, threads for which all events generated had the same state as in the original execution, therefore they can still be advanced. The trace created,  $\tau$ , is also maintained as a stack, but with first events at bottom of the stack; it is initially empty. Variable t keeps track of the index of the thread which should be advanced next. The main loop is a backtracking loop, exiting only when the entire space has been explored. Inside the loop, the first part (lines 3–6) checks whether the next thread can be advanced. If a thread is found, the state is modified accordingly (lines 12–15), disabling further advances to the thread if the state of the added event differs from the one in the observed trace (lines 8-10); note that in the latter case, the top event in the corresponding thread needs not be removed, since the thread is disabled. Then,  $\tau$  is advanced and added to the result set, property  $\varphi$  is checked (line 17), and the search for the next advance-able thread is restarted (line 18). If no additional thread can be advanced from this state, the algorithm backtracks, undoing the effects of previous advances (lines 21–31).

A simple amortized analysis of Algorithm 1 shows that, without any additional knowledge about the property to check  $\varphi$ , it essentially performs a minimal amount of work: it generates and checks against  $\varphi$  each consistent extended interleaving prefix of  $\tau_0$ , searching for each next event through the tops of at most k thread stacks. Supposing that  $\varphi$  is a simple safety property taking constant time and memory to evaluate in any given state  $\sigma$ , which is frequently the case in many situations, the time complexity of our algorithm is  $\mathcal{O}(|feasible(\tau_0)| \times k)$  and its memory complexity is  $\mathcal{O}(|\tau_0|)$ ; recall that  $feasible(\tau_0)$  is prefix-closed.

Happens-before based model checkers can exploit the property being checked or the structure of the program to gain efficiency (but not coverage). We envision similar techniques could potentially be applied to our models. However, here we are not trying to propose an *optimal* model checker but rather to show that the maximum model is algorithmically analyzable, not just an existential entity, and it can be the basis on which other analysis techniques can be built.

### 7 Maximality

Proposition 1 showed that our causal model is sound in the sense that the same program which generated a trace can also generate its entire feasibility closure. Section 5 further showed that the causal model is quite comprehensive, being able to capture the existing sound causal models.

In this section we show that the model is also maximal among sound models,

Proc ::= Proc || Proc

$$| \quad Int : Stmt \qquad \frac{\langle s, \sigma, \delta, \rho, t \rangle \xrightarrow{\tau} \langle s', \sigma', \delta', \rho', t \rangle}{\langle t : s, \sigma, \delta, \rho \rangle \xrightarrow{\tau} \langle t : s', \sigma', \delta', \rho', t \rangle} \\ \langle t : s, \sigma, \delta, \rho \rangle \xrightarrow{\tau} \langle t : s', \sigma', \delta', \rho', t \rangle} \\ | \quad Stmt \ ::= \quad Stmt \ ; Stmt \qquad \frac{\langle s_1, \sigma', \delta', \rho', t \rangle \xrightarrow{\tau} \langle s'_1, \sigma', \delta', \rho', t \rangle}{\langle s_1 \ ; s_2, \sigma, \delta, \rho, t \rangle \xrightarrow{\tau} \langle s'_1 \ ; s_2, \sigma', \delta', \rho', t \rangle} \\ | \quad \text{nop} \qquad \overline{\langle (\text{nop} \ ; s, \sigma, \delta, \rho, t) \xrightarrow{\tau} \langle s_1 \ ; s_2, \sigma', \delta', \rho', t \rangle}} \\ | \quad \text{if } Int \ \text{then } Stmt \qquad \overline{\langle (\text{if } i \ \text{then } s, \sigma, \delta, \rho, t) \xrightarrow{\epsilon} \langle (\text{nop}, \sigma, \delta, \rho, t) \rangle}} \\ | \quad \text{if } i \ \text{then } s, \sigma, \delta, \rho, t \rangle \xrightarrow{\epsilon} \langle (\text{nop}, \sigma, \delta, \rho, t) \rangle} \\ | \quad \text{load } Loc \qquad \overline{\langle (\text{load } x, \sigma, \delta, \rho, t) \xrightarrow{(t, read, x, i)} \langle (\text{nop}, \sigma, \delta, \rho, t) \rangle}} \\ | \quad \text{load } Loc \qquad \overline{\langle (\text{load } x, \sigma, \delta, \rho, t) \xrightarrow{(t, read, x, i)} \langle (\text{nop}, \sigma, \delta, \rho, t) \rangle}} \\ | \quad \text{where } i = 0 \qquad \overline{\langle (\text{load } x, \sigma, \delta, \rho, t) \xrightarrow{(t, read, x, i)} \langle (\text{nop}, \sigma, \delta, \rho, t) \rangle}} \\ | \quad \text{acquire } Loc \qquad \overline{\langle (\text{acquire } x, \sigma, \delta, \rho, t) \xrightarrow{(t, sequire, x)} \langle (\text{nop}, \sigma, \delta, t) \times (t, t), \rho, t \rangle}} \\ | \quad \text{release } Loc \qquad \overline{\langle (\text{release } x, \sigma, \delta, \rho, t) \xrightarrow{(t, release, x)} \langle (\text{nop}, \sigma, \delta, t) \times (t, t), \rho, t \rangle}} \\ | \quad \text{if } \delta(s, t) \Rightarrow \overline{\langle (\text{release } x, \sigma, \delta, \rho, t) \xrightarrow{(t, release, x)} \langle (\text{nop}, \sigma, \delta, t) \times (t, t), \rho, t \rangle}} \\ | \quad \text{if } \delta(s, t) \Rightarrow \overline{\langle (\text{release } x, \sigma, \delta, \rho, t) \xrightarrow{(t, release, x)} \langle (\text{nop}, \sigma, \delta, t) \times (t, t), \rho, t \rangle}} \\ | \quad \text{if } \delta(s, t) \Rightarrow \overline{\langle (\text{release } x, \sigma, \delta, \rho, t) \xrightarrow{(t, release, x)} \langle (\text{nop}, \sigma, \delta, t) \times (t, t), \rho, t \rangle}} \\ | \quad \text{if } \delta(s, t) \Rightarrow \overline{\langle (\text{release } x, \sigma, \delta, \rho, t) \xrightarrow{(t, release, x)} \langle (\text{nop}, \sigma, \delta, t), t \times (t, t), \rho, t \rangle}} \\ | \quad \text{if } \delta(s, t) \Rightarrow \overline{\langle (\text{release } x, \sigma, \delta, \rho, t) \xrightarrow{(t, release, x)} \langle (\text{nop}, \sigma, \delta, t), t \times (t, t), \rho, t \rangle}} \\ | \quad \text{if } \delta(s, t) \Rightarrow \overline{\langle (\text{release } x, \sigma, \delta, \rho, t) \xrightarrow{(t, release, x)} \langle (\text{nop}, \sigma, \delta, t), t \times (t, t), \rho, t \rangle}} \\ | \quad \text{if } \delta(s, t) \Rightarrow \overline{\langle (\text{release } x, \sigma, \delta, \rho, t) \xrightarrow{(t, release, x)} \langle (\text{nop}, \sigma, \delta, t), t \times (t, t), \rho, t \rangle}} \\ | \quad \text{if } \delta(s, t) \Rightarrow \overline{\langle (\text{release } x, \sigma, \delta, \rho, t) \xrightarrow{(t, release, x)} \langle (\text{release } x, \tau, \delta, \rho, t), t \times (t, t), \rho, t \rangle}} \\ | \quad$$

 $\frac{\langle p_1, \sigma, \delta, \rho \rangle \xrightarrow{\tau} \langle p'_1, \sigma', \delta', \rho' \rangle}{\langle p_1 \mid \mid p_2, \sigma, \delta, \rho \rangle \xrightarrow{\tau} \langle p'_1 \mid \mid p_2, \sigma', \delta', \rho' \rangle} (Par_1) \qquad \frac{\langle p_2, \sigma, \delta, \rho \rangle \xrightarrow{\tau} \langle p'_2, \sigma', \delta', \rho' \rangle}{\langle p_1 \mid \mid p_2, \sigma, \delta, \rho \rangle \xrightarrow{\tau} \langle p_1 \mid \mid p'_2, \sigma', \delta', \rho' \rangle}$ 

Figure 3: Syntax and SOS semantics for the CONC language

in the sense that any extension to it is done at the expense of soundness. We will prove therefore that given a trace  $\tau'$  which is not in the feasibility closure of a trace  $\tau$ , there exists a program p which can generate  $\tau$  but not  $\tau'$ ; therefore, if the model were extended to include  $\tau'$  and used  $\tau'$  as a witness that a property is satisfied/invalidated by a program generating  $\tau$ , this would be a false witness if the program which generated  $\tau$  was p.

To prove our claim, we propose CONC, a very simple (not even Turing complete) concurrent language which can conceivably be simulated in any real language. Figure 3 presents the grammar and SOS semantics of CONC. The grammar specifies a parallel composition of named threads. Each thread is a succession of statements and uses one internal register to load data from the shared memory. load x loads the value at location x into the internal register of the thread, x := i stores integer i at location x, acquire and release have the straight-forward semantics, and if i then s executes s only if the internal register has value i. A running configuration of CONC is a tuple  $\langle p, \sigma, \delta, \rho \rangle$  where p is the remainder of the program being executed,  $\sigma$  maps variables to values,  $\delta$  maps each lock to the id of the thread holding it, and  $\rho$  gives for each thread the value of its internal register. Assuming p has p threads, the initial configuration of the system is  $START(p) = \langle p, \sigma_{\epsilon}, \delta_{\epsilon}, \rho_{\epsilon} \rangle$  where  $\sigma_{\epsilon}$ ,  $\delta_{\epsilon}$ , and  $\rho_{\epsilon}^{n}$ , initialize all locations, locks, and registers for the p threads with p respectively.

It is straightforward to associate to each event an instruction producing it. Let code be the mapping defined on events as follows:

$$code(e) = \left\{ \begin{array}{ll} \mathsf{load} \ x & \ \mathbf{if} \ e = (t, read, x, i) \\ x := i & \ \mathbf{if} \ e = (t, write, x, i) \\ \mathsf{acquire} \ x & \ \mathbf{if} \ e = (t, acquire, x) \\ \mathsf{release} \ x & \ \mathbf{if} \ e = (t, release, x) \end{array} \right.$$

Given a program p, let  $p 
vert_t$  be its projection on thread t, that is, the statement labeled by t in the parallel composition.

The following result shows that, except for the code, the running configuration is completely determined by the trace generated up to that point, basically allowing us to work just with code and traces in the sequel.

**Proposition 7** ( $\tau$ -configurations) Suppose that n is the number of threads of p and

CONC 
$$\vdash START(p) \xrightarrow{\tau}^* \langle p', \sigma_{\tau}, \delta_{\tau}, \rho_{\tau}^n \rangle$$
.

Then (1) 
$$\sigma_{\tau}(x) = data(latest_{write}(\tau|_{x}));$$
  
(2)  $\delta_{\tau}(x) = \begin{cases} thread(latest(\tau|_{x})) \text{ if } op(latest(\tau|_{x})) = acquire \\ \bot \text{ otherwise } \end{cases}$   
and (3)  $\rho_{\tau}^{n}(t) = data(latest_{read}(\tau|_{t})).$ 

PROOF. First note that  $\epsilon$ -transitions only affect the code-part of configurations. Therefore, the base case, when  $\tau = \epsilon$ , is trivial. Suppose now that

$$START(p) \xrightarrow{\tau}^* \langle p', \sigma_{\tau}, \delta_{\tau}, \rho_{\tau}^n \rangle \xrightarrow{e} \langle p'', \sigma, \delta, \rho \rangle.$$

Suppose that  $thread(e) = t_i$ . Then in order for e to be generated,  $p' \upharpoonright_{t_i}$  must be code(e); p''' and  $p'' = \mathsf{nop}$ ; p'''.

The proof tree is built by applying in order i-1 steps of  $Par_2$  and then a step of  $(Par_1)$  (or none, if i=n), then (Thread), (Seq), and finally one of the (Read), (Write), (Acq), or (Rel); we therefore need to show that

$$\langle code(e), \sigma_{\tau}, \delta_{\tau}, \rho_{\tau}^{n}, t_{i} \rangle \xrightarrow{e} \langle \mathsf{nop}, \sigma_{\tau e}, \delta_{\tau e}, \rho_{\tau e}^{n}, t_{i} \rangle$$

If  $e = (t_i, read, x, i)$ , then code(e) = load x so we can apply the (*Read*) rule, which updates only  $\rho_{\tau}^n$  to  $\rho_{\tau}^n[r \leftarrow i] = \rho_{\tau e}^n$ .

If  $e = (t_i, write, x, i)$ , then code(e) = x := i, and rule Write applies updating only  $\sigma$  to  $\sigma_{\tau}[x \leftarrow i] = \sigma_{\tau e}$ .

If  $e = (t_i, acquire, x)$ , then  $code(e) = \mathsf{acquire}\ x$  and only  $\delta$  is updated to  $\delta_{\tau}[x \leftarrow t] = \delta_{\tau e}$ .

If  $e=(t_i, release, x)$ , then code(e)= release x and only  $\delta$  is updated to  $\delta_{\tau}[x \to \bot] = \delta_{\tau e}$ .

Therefore, in the sequel we will use CONC  $\vdash p \xrightarrow{\tau^*} p'$  instead of CONC  $\vdash START(p) \xrightarrow{\tau^*} \langle p', \sigma_{\tau}, \delta_{\tau}, \rho_{\tau}^n \rangle$ 

Now, let us prove that the semantics of CONC does indeed satisfy the sequential consistency axioms. Let p be a CONC program and let feasible(p) be the set of all p-feasible traces; that is  $\tau$  is p-feasible if there exists a program p' such that CONC  $\vdash p \xrightarrow{\tau}^* p'$ . We sill show that feasible(p) satisfies the  $strong\ local\ determinism\ property$ , namely, not only that an enabled event can be generated at any point by its thread, but that it also must be the next event generated by that thread (ignoring the data attribute for read events). Formally, feasible(p) satisfies strong local determinism if it satisfies local determinism and if  $\tau_1e_1$  and  $\tau_2e_2$  are p-feasible,  $thread(e_1) = thread(e_2) = t$ , and  $\tau_1 \upharpoonright_{t} = \tau_2 \upharpoonright_{t}$ , then  $op(e_1) = op(e_2)$ ,  $target(e_1) = target(e_2)$ ; if additionally  $op(e_i) = write$ , then also  $data(e_1) = data(e_2)$ .

The restriction of a derivation CONC  $\vdash p_0 \xrightarrow{\tau_1} p_1 \xrightarrow{\tau_2} \cdots \xrightarrow{\tau_n} p_n$  to thread t is the maximal subsequence of statements  $p_0 \upharpoonright_t, p_1 \upharpoonright_t, \ldots, p_n \upharpoonright_t$  such that any two consecutive statements in the sequence are distinct, and we'll denote it by  $\left[ \text{CONC} \vdash p_0 \xrightarrow{\tau_1 \cdots \tau_n}^* p_n \right] \upharpoonright_t$ .

The following result shows that every CONC program p is a consistent system in the sense of Definition 2.

**Proposition 8 (CONC consistency)** feasible(p) satisfies prefix closedness and <math>strong local determinism.

PROOF. Prefix closedness is obvious, since the semantics can emit at most one event for each execution step.

Now, notice that every transition in CONC modifies the code for precisely one of the threads. Let us prove the following stronger local result which basically shows that the evolution of a thread does not depend on events which are not directly related to it.

Suppose CONC  $\vdash p \xrightarrow{\tau_1}^* p_1 \xrightarrow{\tau_1'} p_1'$  (where  $\tau_1'$  is either  $\epsilon$  or an event) and CONC  $\vdash p \xrightarrow{\tau_2}^* p_2$ , such that  $\tau_1 \upharpoonright_t = \tau_2 \upharpoonright_t$ ,  $p_1 \upharpoonright_t = p_2 \upharpoonright_t$ ,  $p_1' \upharpoonright_t \neq p_1 \upharpoonright_t$ , and either  $\tau_2 \tau_1'$  is consistent or  $\tau_1'$  is a read event. Then there exist an unique  $p_2'$  such that  $p_2' \upharpoonright_t \neq p_2 \upharpoonright_t$  and CONC  $\vdash p \xrightarrow{\tau_2}^* p_2 \xrightarrow{\tau_2'} p_2'$ , and, moreover  $p_1' \upharpoonright_t = p_2' \upharpoonright_t$ , and either  $\tau_1' = \tau_2'$  or both of them are reading from the same target but with different values.

Note that uniqueness holds trivially, as once a thread was chosen, at most one rule can apply. Now, for the existence part, if the transition for  $\tau'_1$  is

- (Nop), then it can be applied on  $p_2$  with the same effect;
- $(If_{true})$  or  $(If_{false})$ , then since  $\rho_{\tau}^{n}(t) = data(latest_{read}(\tau \upharpoonright_{t}))$  and  $\tau_{1} \upharpoonright_{t} = \tau_{2} \upharpoonright_{t}$ , it follows that  $\rho_{\tau_{1}}^{n}(t) = \rho_{\tau_{2}}^{n}(t)$ , therefore the exact same transition can be applied on  $p_{2}$ ;
- (*Read*), then  $\tau'_1 = (t, read, x, i)$ , which must be generated by an instruction load x, whence the same rule can be applied on  $p_2$ , generating event (t, read, x, i'), where  $i' = \sigma_{\tau_2}(x)$ .
- (Write), then  $\tau'_1 = (t, write, x, i)$ , which must be generated by an instruction x := i, whence and same rule can be applied on  $p_2$ , generating the same event:
- (Acq), then  $\tau_1' = (t, acquire, x)$ , which must be generated by acquire x; since  $\tau_2\tau_1'$  is consistent, it must be that  $\delta_{\tau_2}(x) = \bot$ , hence the rule can be applied on  $p_2$  generating the same event;
- (Rel), then  $\tau_1' = (t, release, x)$ , which must be generated by release x; since  $\tau_2 \tau_1'$  is consistent, it must be that  $\delta_{\tau_2}(x) = t$ , hence the rule can be applied on  $p_2$  generating the same event.

Let us now use the above lemma to prove the local determinism property. Assume  $\tau_1 e$  and  $\tau_2$  are p-feasible traces such that thread(e) = t and  $\tau_1 \upharpoonright_t = \tau_2 \upharpoonright_t$ . Let  $p_1, p_2$  be such that  $CONC \vdash p \xrightarrow{\tau_1 e} p_1$  and  $CONC \vdash p \xrightarrow{\tau_2} p_2$ . Then we can use the lemma proved above to show that  $\left[CONC \vdash p \xrightarrow{\tau_2}^* p_2\right] \upharpoonright_t$  is a prefix of  $\left[CONC \vdash p \xrightarrow{\tau_1 e}^* p_2\right] \upharpoonright_t$ , by induction on the length of  $\left[CONC \vdash p \xrightarrow{\tau_2}^* p_2\right] \upharpoonright_t$ .

Moreover, using the same lemma, we can (uniquely) continue the execution of  $p_2$  on thread t using the same steps from the execution of  $p_1$  and the fact that either  $\tau_2 e$  is consistent or e is a read event.

For strong local determinism, if  $\tau_1 e_1$  and  $\tau_2 e_2$  are *p*-feasible,  $thread(e_1) = thread(e_2) = t$ , and  $\tau_1 \upharpoonright_t = \tau_2 \upharpoonright_t$ , then we can use an argument similar to the one above to match the statements corresponding to thread *t* from the beginning of the execution and until  $e_1$  and  $e_2$  are generated, and for this step, applying the part of the lemma referring to the generated events.

Now, given a trace  $\tau$ , let us build the canonical CONC program generating it. code can be naturally extended on traces by  $code(e\tau) = code(e)$ ;  $code(\tau)$ . Let  $\{t_1, t_2, \ldots, t_n\}$  be the set of thread ids appearing in  $\tau$ . Then the program  $program(\tau)$  associated to trace  $\tau$  is defined by

$$program(\tau) = t_1 : code(\tau \upharpoonright_{t_1}) \mid\mid \cdots \mid\mid t_n : code(\tau \upharpoonright_{t_n})$$

Let us also define the empty program with n threads as  $program^n(\epsilon) = t_1$ : nop  $|| \cdots || t_n :$  nop. The following result shows that the program corresponding to a consistent trace can indeed generate that trace.

**Proposition 9** If  $\tau$  is a consistent trace with n threads, then CONC  $\vdash$  program $(\tau) \xrightarrow{\tau}^* program^n(\epsilon)$ .

PROOF. We prove by induction on the length of  $\tau$  that CONC  $\vdash program(\tau\tau') \xrightarrow{\tau}^* program^n(\tau')$  where

$$program^{n}(\tau) = t_{1} : \overline{code}(\tau|_{t_{1}}) \mid\mid \cdots \mid\mid t_{n} : \overline{code}(\tau|_{t_{n}}), \overline{code}(\tau) = \begin{cases} code(\tau) & \text{if } \tau \neq \epsilon \\ \text{nop} & \text{otherwise} \end{cases}.$$

The base case, when  $\tau = \epsilon$ , is trivial:  $program^n(\tau') = program(\tau)$ . Suppose now that CONC  $\vdash program(\tau e \tau') \xrightarrow{\tau}^* program^n(e \tau')$ . We need to show that

$$\langle program^n(e\tau'), \sigma_{\tau}, \delta_{\tau}, \rho_{\tau}^n \rangle \xrightarrow{e}^{*} \langle program^n(\tau'), \sigma_{\tau e}, \delta_{\tau e}, \rho_{\tau e}^n \rangle.$$

Suppose that  $thread(e) = t_i$ . Then:

$$\mathit{program}^n(e\tau')\!\!\upharpoonright_{t_j} = \left\{ \begin{array}{l} \mathit{program}^n(\tau')\!\!\upharpoonright_{t_j}, & \textbf{if } j \neq i \\ \mathit{code}(e) \ ; \mathit{program}^n(\tau')\!\!\upharpoonright_{t_i}, \textbf{otherwise} \end{array} \right.$$

We can build a proof for the above assertion in two execution steps: one generating the event and affecting the configuration by turning the instruction code(e) into nop, and the other dissolving the nop and obtaining the desired configuration. For the first step, the proof tree is build by applying in order i-1 steps of  $Par_2$  and then a step of  $(Par_1)$  (or none, if i=n), then (Thread), (Seq), and finally one of the (Read), (Write), (Acq), or (Rel) to deduce

$$\langle code(e), \sigma_{\tau}, \delta_{\tau}, \rho_{\tau}^{n}, t \rangle \xrightarrow{e} \langle \mathsf{nop}, \sigma_{\tau e}, \delta_{\tau e}, \rho_{\tau e}^{n}, t \rangle$$

If e = (t, read, x, i), then code(e) = load x so we can apply the (Read) rule, which updates the register for t of  $\rho_{\tau}^{n}$  to  $\sigma_{\tau}(x) = data(latest_{write}(\tau \upharpoonright_{x}))$ . However, due to consistency constraints,  $data(e) = data(latest_{write}(\tau \upharpoonright_{x}))$ ,

whence  $\rho_{\tau}^{n}[r \leftarrow i] = \rho_{\tau e}^{n}$  and the rule generates e. Moreover,  $\sigma_{\tau e} = \sigma_{\tau}$  and  $\delta_{\tau e} = \delta_{\tau}$ , whence our claim is proven.

If e = (t, write, x, i), then code(e) = x := i, and rule Write applies generating e; we have that  $\sigma_{\tau}[x \leftarrow i] = \sigma_{\tau e}$ ,  $\delta_{\tau e} = \delta_{\tau}$ , and  $\rho_{\tau e}^{n} = \rho_{\tau}^{n}$ .

If e=(t, acquire, x), then code(e)= acquire x and because of the consistency requirements for mutexes (the difference between the number of acquire and release operation is either 0 or 1 for each prefix), it must be that  $op(latest(\tau)_x) \neq acquire$ , whence  $\delta_{\tau}(x) = \bot$  and rule Acq can apply generating e; we have that  $\delta_{\tau}[x \leftarrow t] = \delta_{\tau e}$ ,  $\sigma_{\tau} = \sigma_{\tau e}$ , and  $\rho_{\tau}^{n} = \rho_{\tau e}^{n}$ .

If e = (t, release, x), then code(e) = release x and because of the consistency requirements for mutexes, it must be that  $op(latest(\tau \upharpoonright_x)) = acquire$ , whence  $\delta_{\tau}(x) = thread(latest(\tau \upharpoonright_x))$ . Again, from consistency requirements, since all consecutive acquire-release pairs share the same thread, it follows that  $\delta_{\tau}(x) = t$  and so the rule Rel can apply generating e; we have that  $\delta_{\tau}[x \leftarrow \bot] = \delta_{\tau e}$ ,  $\sigma_{\tau} = \sigma_{\tau e}$ , and  $\rho_{\tau}^n = \rho_{\tau e}^n$ .

The following theorem justifies the maximality claims for the proposed causal model.

**Theorem 3 (Maximality)** For any  $\tau'$  consistent but not  $\tau$ -feasible there exists a program generating  $\tau$  but not  $\tau'$ .

PROOF. We can assume, without loss of generality, that  $\tau' = \tau_1 e'$  such that  $\tau_1$  is  $\tau$ -feasible (potentially empty).

Let t = thread(e').

(1) Suppose  $\tau_1 \upharpoonright_t$  is a prefix of  $\tau \upharpoonright_t$ . If  $\tau \upharpoonright_t = \tau_1 \upharpoonright_t$  then  $\tau'$  cannot be  $\operatorname{program}(\tau)$ -feasible because that would mean there exists a derivation CONC  $\vdash \operatorname{program}(\tau) \xrightarrow{\tau_1}^* p_1 \xrightarrow{e'} p'_1$ ; however since CONC  $\vdash \operatorname{program}(\tau) \xrightarrow{\tau}^* p$  with  $p = \operatorname{program}^n(\epsilon)$ , from the proof of Proposition 8,  $[\operatorname{CONC} \vdash \operatorname{program}(\tau) \xrightarrow{\tau}^* p] \upharpoonright_t$  must be a proper prefix of  $[\operatorname{CONC} \vdash \operatorname{program}(\tau) \xrightarrow{\tau_1}^* p_1 \xrightarrow{e'} p'_1] \upharpoonright_t$ , which is not possible as  $p \upharpoonright_t = \mathsf{nop}$ .

Otherwise it must be that  $\tau \upharpoonright_t = \tau_1 \upharpoonright_t e\tau_2$ , and we can apply the strong local determinism to deduce that if  $\tau'$  is  $program(\tau)$ -feasible, then either e = e' or their type is read and their states are different. But both these lead to contradiction because they imply that  $\tau_1 e' \in feasible(\tau)$ .

(2) If  $\tau_1 \upharpoonright_t = \tau_0 e_0'$  is not a prefix of  $\tau \upharpoonright_t$ , then, because  $\tau_1$  is  $\tau$ -feasible, it must be that  $\tau \upharpoonright_t = \tau_0 e_0 e \tau_2$  such that  $op(e_0) = op(e_0') = read$ ,  $target(e_0) = target(e_0')$ , and  $data(e_0) \neq data(e_0')$ .

Let us consider a special event  $?^t = (t, ?)$  (with the meaning that  $thread(?^t) = t$  and  $op(?^t) = ?$ ) and for each statement s, an extension  $code^s$  of code with  $code^s(?^t) = s$ , and  $program^s(\tau) = code^s(\tau \upharpoonright_{t_1}) \mid | \cdots | | code^s(\tau \upharpoonright_{t_n})$ .

Let  $x = target(e_0)$ ,  $i = data(e_0)$  and let

$$j = \begin{cases} data(e') - 1, & \text{if } data(e') \text{ defined} \\ 0, & \text{otherwise} \end{cases}$$

Let  $\tau_l$ ,  $\tau_r$  be such that  $\tau = \tau_l e \tau_r$  and  $\tau_l \upharpoonright_t = \tau_0 e_0$ , and let  $p' = program^s(\tau_l?^t e \tau_r)$ , obtained by inserting s = if i then x := j in the code for thread t, between

the code for  $e_0$  and that for e. This new program can still generate  $\tau$ , as the state read by  $e_0$  is different than that read by  $e'_0$  and thus the conditional is not executed, but it cannot generate  $\tau'$ , as the next event for thread t upon generating  $\tau_0 e'_0$  must be (t, write, x, j) which is guaranteed to be distinct from e'

#### 8 Conclusion and Future Work

We have shown that, by axiomatizing basic properties of (sequentially consistent) concurrent systems, one can obtain maximally sound causal models for concurrent executions, which can be naturally associated to each observed trace, capturing all feasible traces which could be inferred from it. The maximality result has two important theoretical implications. First, verifying the soundness claims for any causal model is reduced to proving that it is a submodel of the maximal one. Second, since the maximal model captures all causally equivalent traces, it allows for universal, model-independent definitions for causal properties. Finally, the model can be explored for finding concurrency anomalies either directly through its axioms (as shown practical by Said et al. [20]) or through its constructive characterization presented in this paper.

We strongly believe that the ideas developed here in the context of sequential consistency can be generalized to more relaxed memory models. However, we expect this task to be more complex, as the axiomatization of these memory models is also more complex. For example, one of the simplest relaxed memory models, the x86-TSO [19], has no less than nine axioms interrelating the behavior of concurrent objects. To model it in our framework, one would have to adjust both the consistency criteria—to account for the usage of write buffers—and the local determinism axiom—to account for the fact that commits from the write buffer to memory happen relatively independent of the thread execution. The main difficulty introduced by this new paradigm is that a concurrent object's behavior is no longer independent, but constrained by other concurrent objects: for example, committing a buffered write into memory now depends on the fact that the previously buffered writes were already committed, while acquiring a lock depends on the fact that the write buffer is empty. Nevertheless, we believe these issues are not insurmountable, but rather promising avenues for further research.

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