The Complexity of Dynamic Data Race Prediction

Umang Mathur University of Illinois, Urbana Champaign USA umathur3@illinois.edu Andreas Pavlogiannis Aarhus University Denmark pavlogiannis@cs.au.dk Mahesh Viswanathan University of Illinois, Urbana Champaign USA vmahesh@illinois.edu

Abstract

Writing concurrent programs is notoriously hard due to scheduling non-determinism. The most common concurrency bugs are data races, which are accesses to a shared resource that can be executed concurrently. Dynamic data-race prediction is the most standard technique for detecting data races: given an observed, data-race-free trace t, the task is to determine whether t can be reordered to a trace t^* that exposes a data-race. Although the problem has received significant practical attention for over three decades, its complexity has remained elusive. In this work, we address this lacuna, identifying sources of intractability and conditions under which the problem is efficiently solvable. Given a trace t of size n over k threads, our main results are as follows.

First, we establish a general $O(k \cdot n^{2 \cdot (k-1)})$ upper-bound, as well as an $O(n^k)$ upper-bound when certain parameters of t are constant. In addition, we show that the problem is NP-hard and even W[1]-hard parameterized by k, and thus unlikely to be fixed-parameter tractable. Second, we study the problem over acyclic communication topologies, such as server-clients hierarchies. We establish an $O(k^2 \cdot d \cdot n^2 \cdot \log n)$ upper-bound, where d is the number of shared variables accessed in t. In addition, we show that even for traces with k = 2 threads, the problem has no $O(n^{2-\epsilon})$ algorithm under the Orthogonal Vectors conjecture. Since any trace with 2 threads defines an acyclic topology, our upper-bound for this case is optimal wrt polynomial improvements for up to moderate values of k and d. Finally, motivated by existing heuristics, we study a distance-bounded version of the problem, where the task is to expose a data race by a witness trace that is similar to t. We develop an algorithm that works in O(n) time when certain parameters of *t* are constant.

1 Introduction

Keywords Data Race Prediction, Complexity

A concurrent program is said to have a data race if it can exhibit an execution in which two conflicting accesses¹ to the same memory location are "concurrent". Data races in concurrent programs are often symptomatic of bugs in software like data corruption [6, 19, 26], pose challenges in defining the semantics of programming languages, and have led to serious problems in the past [43]; it is no surprise that data races have been deemed pure evil [7]. Automatically finding data races in programs remains a widely studied problem because of its critical importance in building correct concurrent software. Data-race detection techniques can broadly be classified into static and dynamic. Given that the race-detection problem in programs is undecidable, static race detection approaches [25, 33] are typically conservative, produce false alarms, and do not scale to large software. On the other hand, since dynamic approaches [15, 23, 32, 36] have the more modest goal of discovering data races by analyzing a single trace, they are lightweight, and can often scale to production-level software. Moreover, many dynamic approaches are sound, i.e., do not raise false race reports. The effectiveness and scalability of dynamic approaches has lead to many practical advances on the topic. Despite a wide-spread interest on the problem, characterizing its complexity has remained elusive.

Informally, the *dynamic race prediction* problem is the following: given an observed trace t of a multi-threaded program, determine if t demonstrates the presence of a data race in the program that generates t. This means that either t has two conflicting data accesses that are concurrent, or a different trace resulting from scheduling the threads of t in a different order, witnesses such a race. Additional traces that result from alternate thread schedules are captured by the notion of a *correct reordering* of t, that characterizes a set of traces that can be exhibited by *any* program that can generate t; a precise definition of correct reordering is given in Section 2.1. So formally, the data race prediction problem is, given a trace t, determine if there is a correct reordering of t in which a pair of conflicting data accesses are concurrent.

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¹Two accesses are conflicting if they access the same memory location, with one of them being a write access.

While the data race prediction problem is clearly in NP guess a correct reordering and check if it demonstrates a data race – its precise complexity has not been identified. Evidence based on prior work, suggests a belief that the problem might be NP-complete. First, related problems, like datarace detection for programs with strong synchronization primitives [27–29], or verifying sequential consistency [16], are known to be NP-hard. Second, all known "complete" algorithms run in worst-case exponential time. These approaches either rely on an explicit enumeration of all correct reorderings [10, 37], or they are symbolic approaches that reduce the race prediction problem to a constraint satisfaction problem [18, 35, 41]. On the other hand, a slew of "partial order"-based methods have been proposed, whose goal is to predict data races in polynomial time, but at the cost of being incomplete and failing to detect data races in some traces. These include algorithms based on the classical happens-before partial order [15, 21-23, 38], and those based on newer partial orders that improve the prediction of data races over happens-before [20, 31, 34, 40].

In this paper we study the problem of data-race prediction from a complexity-theoretic perspective. Our goal is to understand whether the problem is intractable, the causes for intractability, and conditions under it can can be solved efficiently. We provide partial answers to all these questions, and in some cases characterize the tractability/intractability landscape precisely in the form of optimality results.

Contributions. Consider an input trace t of size n over k threads. Our main contributions are as follows. We refer to Section 2.3 for a formal summary.

Our first result shows that the data-race prediction problem is solvable in $O(k \cdot n^{2 \cdot (k-1)})$ time, and can be improved to $O(n^k)$ when certain additional parameters of t are constant. We note that most benchmarks used in practice have a constant number of threads [15, 20, 22, 31, 34, 40], and in such cases our upper-bound is polynomial.

The observation that data race predication is in polynomial time for constantly many threads naturally leads to two follow-up questions. Does the problem remain tractable for any k? And if not, is it fixed parameter tractable (FPT) wrt k, i.e., is there an algorithm with running time of the form $O(f(k) \cdot n^{O(1)})$? Our second result answers both these questions in the negative, by showing that the problem is W[1]-hard. This formally establishes the NP-hardness of the problem, and excludes efficient algorithms when k is even moderately large (e.g., $k = \Omega(\log n)$).

We then investigate whether there are practically relevant contexts where data-race prediction is more efficiently solvable, i.e., the degree of the polynomial is fixed and independent of k. We consider the case of traces over acyclic communication topologies, such as pipelines, server-clients hierarchies and divide-and-conquer parallelism. Our third result shows that, perhaps surprisingly, over such topologies data-race prediction can be solved in $O(k^2 \cdot d \cdot n^2 \cdot \log n)$ time, where *d* is the total number of synchronization variables (locks) and global memory locations.

In practice, the size *n* of the trace is by far the dominating parameter, while *k* and *d* are many orders of magnitude smaller. Hence, given the above upper-bound, the relevant question is whether the complexity on *n* can be improved further. Our fourth result shows that this is unlikely: we show that, under the Orthogonal Vectors conjecture, there is no $O(n^{2-\epsilon})$ algorithm even for traces restricted to only 2 threads. As any trace with 2 threads induces an acyclic topology, our upper-bound is (conditionally) optimal wrt polynomial improvements.

Finally, the majority of practical data-race prediction heuristics search for a data race witness among correct reorderings that are very similar to the observed trace t, i.e., by only attempting a few event reorderings on t. Motivated by these approaches, we investigate the complexity of a distancebounded version of data-race predication, where the goal is to expose a data race by only looking at correct reorderings of t that are a small distance away. Here, distance between traces is measured by the number of critical sections and write events whose order is reversed from t. Our fifth result is a linear-time (and thus, optimal) algorithm for this problem, when certain parameters of the trace t are constant. This result gives a solid basis for the principled development of fast heuristics for dynamic data-race prediction.

Technical contributions. Towards our main results, we make several technical contributions that might be of independent interest. We summarize some of them below.

- We improve the lower-bound of the well-known problem on verifying sequential consistency with read-mapping (VSC-rm) [16] from the long-lasting NP-hardness to W[1]-hard.
- 2. We show that VSC-rm can be solved efficiently on tree communication topologies of any number of threads, which improves a recent result of [31] for only 2 threads, as well as a result of [9] for more than 2 threads.
- 3. The first challenge in data-race prediction given a trace *t* is to choose the set *X* of events of *t* over which to attempt to construct a correct reordering. Identifying such choices for *X* is a significant challenge [18, 31, 34]. We establish non-trivial upper-bounds on the number of choices for *X*, and show that they are constantly many when certain parameters of *t* are constant.
- 4. Particularly for tree communication topologies, we show that a single choice for such *X* suffices.

Finally, we note that our notion of a predictable data race in a trace t requires as a witness a reordering t^* of t in which every read event reads from the same write event as in t. This guarantees that t^* is valid in any program that produced t.

More permissive reorderings, e.g., requiring that every read event reads the same value, are also possible, and can capture potentially more races. Our notion of witness reflects the most common practice in race-detection literature, where trace logging typically does not track the values.

Related Work. Antoni Mazurkiewicz [1, 24] used the notion of traces to mathematically model executions of concurrent programs. Bertoni et. al. [4] studied various languagetheoretic questions about Mazurkiewicz traces. The folklore results about the NP-hardness of race detection are often attributed to Netzer and Miller [27-29]. However, the problem considered in their work differs in significant ways from the problem of data-race prediction. First, the notion of feasible executions in [27] (the counterpart of the notion of correct reorderings) requires that any two conflicting events be ordered in the same way as the observed execution, and hence, are less permissive. Next, the NP-hardness arises from the use of complex synchronization primitives like wait and signal, which are more powerful than the primitives we study here (release/acquire of locks and read/write of registers). The results due to Netzer and Miller, thus, do not apply to the problem of data-race prediction. Gibbons and Korach [16] establish NP-hardness for a closely related problem in distributed computing - verifying sequential consistency with a read mapping (VSC-rm). Yet again, the problem is different than the problem of race prediction. Complexity theoretic investigations have also been undertaken for other problems in distributed computing like linearizability [13, 16, 17], serializability [30] and transactional consistency [5]. Hence, although there have been many theoretical results on related problems in concurrency, none of them addresses dynamic data-race prediction. Our work fills this gap.

Some proofs are relegated to the appendix.

2 Preliminaries

2.1 Model

General notation. Given a natural number k, let $[k] = \{1, \ldots, k\}$. Given a function $f : X \to Y$, we let dom(f) = X and img(f) = Y. Given two functions f, g, we write $f \subseteq g$ to denote that dom $(f) \subseteq \text{dom}(g)$ and for every $x \in \text{dom}(f)$ we have f(x) = g(x). Given a set $X' \subseteq \text{dom}(f)$, we denote by f|X' the function with dom(f|X') = X' and $f|X' \subseteq f$.

Concurrent program. We consider a shared-memory concurrent program \mathcal{P} that consists of k threads $\{p_i\}_{i \in [k]}$, under sequential consistency semantics [39]. For simplicity of presentation we assume no thread is created dynamically and the set $\{p_i\}_{i \in [k]}$ is known a-priori. Communication between threads occurs over a set of global variables \mathcal{G} , and synchronization over a set of locks \mathcal{L} such that $\mathcal{G} \cap \mathcal{L} = \emptyset$. We let $\mathcal{V} = \mathcal{G} \cup \mathcal{L}$ be the set of all variables of \mathcal{P} . Each thread is deterministic, and performs a sequence of operations. We are

only interested in the operations that access a global variable or a lock, which are called *events*. In particular, the allowed events are the following.

- Given a global variable x ∈ G, a thread can either write to x via an event w(x) or *read* from x via an event r(x).
- 2. Given a lock $\ell \in \mathcal{L}$, a thread can either *acquire* ℓ via an event $acq(\ell)$ or *release* ℓ via an event $rel(\ell)$.

Each event is atomic, represented by a tuple (a, b, c, d), where

- *a* ∈ {w, r, acq, rel} represents the type of the event (i.e., write, read, lock-acquire or lock-release event),
- 2. *b* represents the variable or lock that the event accesses,
- 3. *c* is the thread of the event, and
- 4. *d* is a unique identifier of the event.

Given an event *e*, we let loc(*e*) denote the global variable or lock that *e* accesses. We occasionally write e(x) to denote an event *e* with loc(*e*) = *x*, while the thread and event id is often implied by the context. We denote by W_p (resp. \mathcal{R}_p , \mathcal{L}_p^A , \mathcal{L}_p^R) the set of all write (resp. read, lock-acquire, lock-release) events that can be performed by thread *p*. We let $\mathcal{E}_p = \mathcal{W}_p \cup \mathcal{R}_p \cup \mathcal{L}_p^A \cup \mathcal{L}_p^R$. We denote by $\mathcal{E} = \bigcup_p \mathcal{E}_p$, $\mathcal{W} = \bigcup_p \mathcal{W}_p$, $\mathcal{R} = \bigcup_p \mathcal{R}_p$, $\mathcal{L}^A = \bigcup_p \mathcal{L}_p^A$, $\mathcal{L}^R = \bigcup_p \mathcal{L}_p^R$ the events, write, read, lock-acquire and lock-release events of the program \mathcal{P} , respectively. Given an event $e \in \mathcal{E}$, we denote by p(e) the thread of *e*. Finally, given a set of events $X \subseteq \mathcal{E}$, we denote by $\mathcal{R}(X)$ (resp., $\mathcal{W}(X)$, $\mathcal{L}^A(X)$, $\mathcal{L}^R(X)$) the set of read (resp., write, lock-acquire, lock-release) events of *X*. For succinctness, we let $\mathcal{W}\mathcal{R}(X) = \mathcal{W}(X) \cup \mathcal{R}(X)$, $\mathcal{R}\mathcal{L}(X) = \mathcal{R}(X) \cup \mathcal{L}^R(X)$ and $\mathcal{W}\mathcal{L}(X) = \mathcal{W}(X) \cup \mathcal{L}^A(X)$. The semantics of \mathcal{P} are the standard for sequential consistency [39].

Conflicting events. Given two distinct events $e_1, e_2 \in \mathcal{E}$, we say that e_1 and e_2 are *conflicting*, denoted by $e_1 \bowtie e_2$, if (i) $loc(e_1) = loc(e_2)$ (i.e., both events access the same global variable or the same lock) and (ii) $\{e_1, e_2\} \cap \mathcal{W} \neq \emptyset$ or $\{e_1, e_2\} \cap \mathcal{L}^A \neq \emptyset$ i.e., at least one of them is either a write event or a lock-acquire event. We extend the notion of conflict to sets of events in the natural way: two sets of events $X_1, X_2 \subseteq \mathcal{E}$ are called *conflicting*, denoted by $X_1 \bowtie X_2$ if $\exists (e_1, e_2) \in (X_1 \times X_2)$ such that $e_1 \bowtie e_2$.

Event sequences. Let *t* be a sequence of events. We denote by $\mathcal{E}(t)$ the set of events, by $\mathcal{L}(t)$ the set of locks, and by $\mathcal{G}(t)$ the set of global variables in *t*. We let $\mathcal{W}(t)$ (resp., $\mathcal{R}(t), \mathcal{L}^A(t), \mathcal{L}^R(t)$) denote the set $\mathcal{W}(\mathcal{E}(t))$ (resp., $\mathcal{R}(\mathcal{E}(t))$), $\mathcal{L}^A(\mathcal{E}(t)), \mathcal{L}^R(\mathcal{E}(t))$), i.e., it is the set of write (resp., read, lock-acquire, lock-release) events of *t*. Given two distinct events $e_1, e_2 \in \mathcal{E}(t)$, we say that e_1 is earlier than e_2 in *t*, denoted by $e_1 <_t e_2$ iff e_1 appears before e_2 in *t*. We say that e_1 is thread-ordered earlier than e_2 , denoted $e_1 <_{TO(t)} e_2$, when $e_1 <_t e_2$ and $p(e_1) = p(e_2)$. For events $e_1, e_2 \in \mathcal{E}(t)$, we say $e_1 \leq_t e_2$ (resp. $e_1 \leq_{TO(t)} e_2$) if either $e_1 = e_2$ or $e_1 <_t e_2$ (resp. $e_1 <_{TO(t)} e_2$). We will often use $<_{TO}$ (resp. \leq_{TO}) in place of

 $<_{TO(t)}$ (resp. $\leq_{TO(t)}$) when the trace *t* is clear from context. Given a set of events $X \subseteq \mathcal{E}$, we denote by t|X the *projection* of *t* onto *X*. Given a thread p_i , we let $t|p_i = t|\mathcal{E}_{p_i}$. Given two sequences t_1, t_2 , we denote by $t_1 \circ t_2$ their concatenation.

Lock events. Given a sequence of events t and a lockacquire event acq $\in \mathcal{L}^{A}(t)$, we denote by match_t(acq) the earliest lock-release event rel $\in \mathcal{L}^{R}(t)$ such that rel \bowtie acq and acq $<_{TO}$ rel, and let match_t(acq) = \perp if no such lockrelease event exists. If match_t(acq) $\neq \perp$, we require that $p(acq) = p(match_t(acq))$, i.e., the two lock events belong to the same thread. Similarly, given a lock-release event rel $\in \mathcal{L}^{R}(t)$, we denote by match_t(rel) the latest acquire event $acq \in \mathcal{L}^A(t)$ such that $match_t(acq) = rel$ and require that such a lock-acquire event always exists. Given a lockacquire event acq, the *critical section* $CS_t(acq)$ is the set of events *e* such that (i) acq $<_{TO}$ *e* and (ii) if match_t(acq) $\neq \perp$, then $e <_{TO}$ match_t(acq). For simplicity of presentation, we assume that locks are not re-entrant. That is, for any two lock-acquire events acq_1, acq_2 with $acq_1 \bowtie acq_2$ and $acq_1 <_{TO} acq_2$, we must have $match_t(acq_1) <_{TO} acq_2$. The *lock-nesting depth* of *t* is the maximum number ℓ such that there exist distinct lock-acquire events $\{acq_i\}_{i=1}^{\ell}$ with (i) $\operatorname{acq}_1 <_{\operatorname{TO}} \operatorname{acq}_2 <_{\operatorname{TO}} \cdots <_{\operatorname{TO}} \operatorname{acq}_{\ell}$, and (ii) for all $i \in [\ell]$, if $\operatorname{match}_{t}(\operatorname{acq}_{i}) \in \mathcal{E}(t)$ then $\operatorname{acq}_{\ell} <_{\operatorname{TO}} \operatorname{match}_{t}(\operatorname{acq}_{i})$.

Traces and reads-from functions. An event sequence *t* is called a *trace* if for any two lock-acquire events $acq_1, acq_2 \in \mathcal{L}^A(t)$, if $loc(acq_1) = loc(acq_2)$ and $acq_1 <_t acq_2$, then $rel_1 = match_t(acq_1) \in \mathcal{L}^R(t)$ and $rel_1 <_t acq_2$. A trace therefore ensures that locks obey mutual exclusion, i.e., critical sections over the same lock cannot overlap.

Given a trace *t*, we define its *reads-from function* RF_t : $\mathcal{R}(t) \rightarrow \mathcal{W}(t)$ as follows: RF_t(r) = w iff w <_t r and $\forall w' \in \mathcal{W}(t)$ with w $\bowtie w'$, we have w' <_t r \Rightarrow w' <_t w. That is, RF_t maps every read event r to the write event w that r observes in *t*. For simplicity, we assume that *t* starts with a write event to every location, hence RF_t is well-defined. For notational convenience, we extend the reads-from function RF_t to lock-release events, such that, for any lock-release event rel $\in \mathcal{L}^{R}(t)$, we have RF_t(rel) = match_t(rel), i.e., rel observes its matching lock acquire event.

Correct reordering, enabled events and predictable data races. A trace t^* is a correct reordering of trace t if (i) $\mathcal{E}(t^*) \subseteq \mathcal{E}(t)$, (ii) for every thread p_i , we have that $t^*|p_i$ is a prefix of $t|p_i$, and (iii) $\operatorname{RF}_{t^*} \subseteq \operatorname{RF}_t$, i.e., the reads-from functions of t^* and t agree on their common read and lock-release events. Given a trace t, an event $e \in \mathcal{E}(t)$ and a correct reordering t^* of t, we say that e is *enabled* in t^* if $e \notin \mathcal{E}(t^*)$ and for every $e' \in \mathcal{E}(t)$ such that $e' <_{\operatorname{TO}} e$, we have that $e' \in \mathcal{E}(t^*)$. Given two conflicting events $e_1, e_2 \in \mathcal{E}(t)$ with $\operatorname{loc}(e_1) = \operatorname{loc}(e_2) \in \mathcal{G}$, we say the pair (e_1, e_2) is a predictable data race of trace t if there is a correct reordering t^* of t such that both e_1 and e_2 are enabled in t^* . Finally, we say t has

a predictable data race if there is a pair (e_1, e_2) which is a predictable data race of t.

Note that predictability of a race is defined with respect to a correct reordering in which every read event observes the same write event. This requirement guarantees that the correct reordering is a valid trace of any concurrent program that produced the initial trace. Hence, every such program is racy. More permissive notions of predictability can also be defined, e.g., by requiring that, in a correct reordering, every read event reads the same value (possibly from a different write event). This alternative definition would capture potentially more predictable races. Our definition of correct reorderings reflects the most common practice in race-detection literature, where trace logging typically does not track the values [20, 22, 31, 34, 40].

The communication topology. The trace *t* naturally induces a *communication topology* graph G = (V, E) where (i) $V = \{p_i\}_i$ and (ii) $E = \{(p_i, p_j) | i \neq j \text{ and } \mathcal{E}(p_i) \bowtie \mathcal{E}(p_j)\}$. In words, we have one node in *G* per thread, and there is an edge between two distinct nodes if the corresponding threads execute conflicting events (note that *G* is undirected). For simplicity, we assume that *G* is connected. In later sections, we will make a distinction between tree topologies (i.e., that do not contain cycles) and general topologies (that might contain cycles). Common examples of tree topologies include stars (e.g., server-clients), pipelines, divide-and-conquer parallelism, and the special case of two threads.

2.2 Problem Statement

In the dynamic data-race prediction problem, we are given an observed trace t, and the task is to identify whether thas a predictable data race. In this work we focus on the following decision problem — given a trace t and two (read or write) conflicting events $e_1, e_2 \in \mathcal{E}(t)$, the task is to decide whether (e_1, e_2) is a predictable data race of t. Clearly, having established the complexity of the decision problem, the general problem can be solved by answering the decision problem for all $O(n^2)$ pairs of conflicting variable access events of t. In the other direction, as the following lemma observes, detecting whether t has some predictable data race is no easier than detecting whether a given event pair of tconstitutes a predictable data race. We refer to Appendix A for the proof.

Lemma 2.1. Given a trace t of length n and two events $e_1, e_2 \in \mathcal{E}(t)$, we can construct a trace t' in O(n) time so that t' has a predictable data race iff (e_1, e_2) is a predictable data race of t.

To make the presentation simpler, we assume w.l.o.g that there are no open critical sections in t, i.e., every lockacquire event acq is followed by a matching lock-release event match_t(acq). Motivated by practical applications, we also study the complexity of dynamic data-race prediction parameterized by a notion of distance between the input trace t and the witness t^* that reveals the data race.

Trace distances. Consider a trace t and a correct reordering t' of t. The set of *reversals* between t and t' is defined as

$$Rv(t, t') = \{ (w_1, w_2) \in \mathcal{WL}(t') \times \mathcal{WL}(t') \mid w_1 \bowtie w_2 \text{ and } w_1 <_t w_2 \text{ and } w_2 <_{t'} w_1 \}.$$

In words, $\operatorname{Rv}(t, t')$ contains the pairs of conflicting write events or lock-acquire events, the order of which has been *reversed* in t' when compared to t. The *distance* of t' from t is defined as $\delta(t, t') = |\operatorname{Rv}(t, t')|$. Our notion of distance, thus, only counts the number of reversals of conflicting write or lock-acquire events instead of counting reversals over all events (or even conflicting write-read events).

Distance-bounded dynamic data race prediction. Consider a trace *t* and two events e_1 , e_2 of *t*. Given an integer $\ell \ge 0$, the ℓ -distance-bounded dynamic data-race prediction problem is the promise problem² that allows for any answer (True/False) if (e_1, e_2) is a predictable data race of *t* and every witness correct reordering t^* is such that $\delta(t, t^*) > \ell$.

2.3 Summary of Main Results

Here we state the main results of this work, and present the technical details in the later parts of the paper.

2.3.1 The General Case

First, we study the complexity of the problem with respect to various parameters of the input trace. These parameters are the number of threads, the number of variables, the locknesting depth, as well as the lock-dependence factor, which, intuitively, measures the amount of data flow between critical sections. In the following, $<_{\text{TRF}}$ (formal definition in Section 3), is the smallest partial order that contains $<_{\text{TO}}$, and also orders read events after their corresponding observed write event (i.e., RF(r) $<_{\text{TRF}}$ r for every $r \in \mathcal{RL}(\mathcal{E}(t))$).

The lock-dependence factor. The *lock-dependence graph* of a trace *t* is the graph $G_t = (V_t, E_t)$ defined as follows.

- 1. The set of vertices is $V_t = \mathcal{L}^A(t)$, i.e., it is the set of lock-acquire events of t.
- 2. The set of edges is such that $(acq_1, acq_2) \in E_t$ if (i) $acq_1 \not<_{TRF} acq_2$, (ii) $acq_1 <_{TRF} match_t(acq_2)$, and (iii) $match_t(acq_1) \not<_{TRF} match_t(acq_2)$.

Given a lock-acquire event $\operatorname{acq} \in V_t$, let A_{acq} be the set of lock-acquire events that can reach acq in G_t . We define the *lock dependence factor* of *t* as $\max_{\operatorname{acq} \in V_t} |A_{\operatorname{acq}}|$. We show the following theorem.

Theorem 2.2. Consider a trace t of length n, k threads, locknesting depth γ , and lock-dependence factor ζ . The dynamic data-race prediction problem on t can be solved in $O(\alpha \cdot \beta)$ time, where $\alpha = \min(n, k \cdot \gamma \cdot \zeta)^{k-2}$ and $\beta = k \cdot n^k$.

In particular, the problem is polynomial-time solvable for a fixed number of threads k. In practice, the parameters k, γ and ζ behave as constants, and in such cases our upperbound becomes $O(n^k)$. Theorem 2.2 naturally leads to two questions, namely (i) whether there is a polynomial-time algorithm for any k, and (ii) if not, whether the problem is FPT with respect to the parameter k, i.e., can be solved in $O(f(k) \cdot n^{O(1)})$ time, for some function f. Question (ii) is very relevant, as typically k is several orders of magnitude smaller than n. We complement Theorem 2.3 with the following lower-bound, which answers both questions in negative.

Theorem 2.3. The dynamic data-race prediction problem is W[1]-hard parameterized by the number of threads.

2.3.2 Tree Communication Topologies

Next, we study the problem for tree communication topologies, such as pipelines and server-clients architectures. We show the following theorem.

Theorem 2.4. Let t be a trace over a tree communication topology with n events, k threads and d variables. The dynamic data-race prediction problem for t can be solved in $O(k^2 \cdot d \cdot n^2 \cdot \log n)$ time.

Perhaps surprisingly, in sharp contrast to Theorem 2.3, for tree topologies there exists an efficient algorithm where the degree of the polynomial is fixed and does not depend on any input parameter (e.g., number of threads). Note that the dominating factor in this complexity is n^2 , while k and d are typically much smaller. Hence, the relevant theoretical question is whether the dependency on n can be improved further. We show that this is unlikely, by complementing Theorem 2.4 with the following conditional lower-bound, based on the Orthogonal Vectors conjecture [8].

Theorem 2.5. Let t be a trace with n events, $k \ge 2$ threads and $d \ge 9$ shared global variables with at least one lock. There is no algorithm that solves the decision problem of dynamic data-race prediction for t in time $O(n^{2-\epsilon})$, for any $\epsilon > 0$, unless the Orthogonal Vectors conjecture fails.

Since k = 2 implies a tree communication topology, the result of Theorem 2.4 is conditionally optimal, up-to polylogarithmic factors, for a reasonable number of threads and variables (e.g., when $k, d = \log^{O(1)}(n)$).

2.3.3 Witnesses in Small Distance

Finally, we study the problem in more practical settings, namely, when (i) the number of threads, lock-nesting depth,

² The promise problem ([14]) given languages L_{True} and L_{False} is to design an algorithm *A* such that A(x) = True for every $x \in L_{\text{True}}$, A(x) = False for every $x \in L_{\text{False}}$, and all for all other inputs $x \notin L_{\text{True}} \cup L_{\text{False}}$, the output A(x) of the algorithm is allowed to be any of True or False.

lock-dependence factor of t are bounded, and (ii) we are searching for a witness at a small distance from t.

Theorem 2.6. Fix a reversal bound $\ell \ge 0$. Consider a trace t of length n and constant number of threads, lock-nesting depth and lock-dependence factor. The ℓ -distance-bounded dynamic data-race prediction problem for t can be solved in O(n) time.

3 Trace Ideals

3.1 Partial Orders

Partially ordered sets. A partially ordered set (or poset) is a pair (X, P) where X is a set of (write, read, lock-acquire, lock-release) events and P is a reflexive, antisymmetric and transitive relation over *X*. We will often write $e_1 \leq_P e_2$ to denote $(e_1, e_2) \in P$. Given two events $e_1, e_2 \in X$ we write $e_1 <_P e_2$ to denote that $e_1 \leq_P e_2$ and $e_1 \neq e_2$, and write $e_1 \ll_P e_2$ to denote that $e_1 <_P e_2$ and there exists no event esuch that $e_1 <_P e <_P e_2$. Given two distinct events $e_1, e_2 \in X$, we say that e_1 and e_2 are *unordered* by *P*, denoted by $e_1 \parallel_P e_2$, if neither $e_1 <_P e_2$ nor $e_2 <_P e_1$. We call an event $e \in X$ *maximal* if there exists no $e' \in X$ such that $e <_P e'$. Given a set $Y \subseteq X$, we denote by P|Y the *projection* of *P* on *Y*, i.e., we have $P|Y \subseteq Y \times Y$, and for all $e_1, e_2 \in Y$, we have $e_1 \leq_{P|Y} e_2$ iff $e_1 \leq_P e_2$. Given two posets (X, P) and (X, Q), we say that the partial order *Q* refines *P*, denoted by $Q \sqsubseteq P$, if for every two events $e_1, e_2 \in X$, if $e_1 \leq_P e_2$ then $e_1 \leq_O e_2$. If Q refines *P*, we say that *P* is *weaker* than *Q*. We denote by $Q \sqsubset P$ the fact that $Q \sqsubseteq P$ and $P \not\sqsubseteq Q$. A *linearization* of (X, P) is a total order over X that refines P. An order ideal (or simply ideal) of a poset (X, P) is subset $Y \subseteq X$ such that for every two events $e_1 \in Y$ and $e_2 \in X$ with $e_2 \leq_P e_1$, we have $e_2 \in Y$. An event *e* is *executable* in ideal *Y* if $Y \cup \{e\}$ is also an ideal of (X, P). The number of threads and variables of a poset (X, P)is the number of threads and variables of the events of *X*.

Partially ordered sets with reads-from functions. A *poset with a reads-from function* (or *rf-poset*) is a tuple (X, P, RF) where (i) $RF \colon \mathcal{RL}(X) \to \mathcal{WL}(X)$ is a reads-from function such that for all $r \in \mathcal{RL}(X)$, we have $RF(r) \in \mathcal{W}(X)$ iff $r \in \mathcal{R}(X)$, and (ii) (X, P) is a poset where for all $r \in \mathcal{RL}(X)$ we have $RF(r) <_P r$. Notation from posets is naturally lifted to *rf*-posets, e.g., an ideal of \mathcal{P} is an ideal of (X, P).

Thread-reads-from order and trace ideals. Given a trace *t*, the thread-reads-from order TRF(t) (or simply TRF when *t* is clear from context) is the weakest partial order over the set $\mathcal{E}(t)$ such that (i) TRF \sqsubseteq TO, and (ii) ($\mathcal{E}(t)$, TRF, RF_t) is an rf-poset. In particular, TRF is the transitive closure of $(\text{TO} \cup \{\text{RF}_t(\mathbf{r}) < \mathbf{r} \mid \mathbf{r} \in \mathcal{R}(t)\})$. A *trace ideal* of *t* is an ideal *X* of the poset ($\mathcal{E}(t)$, TRF). We say an event $e \in \mathcal{E}(t) \setminus X$ is *enabled* in X if for every $e' <_{\text{TO}} e$, we have $e' \in X$. We call *X lock-feasible* if for every two lock-acquire events acq_1 , $\text{acq}_2 \in \mathcal{L}^A(X)$ with $\text{acq}_1 \bowtie \text{acq}_2$, we have $\text{match}_t(\text{acq}_i) \in X$ for some $i \in [2]$. We call *X feasible* if it is lock-feasible, and there exists a partial order *P* over *X* such that (i) $P \sqsubseteq \text{TRF}|X$ and

(ii) for every pair of lock-acquire events $\operatorname{acq}_1, \operatorname{acq}_2 \in \mathcal{L}^A(X)$ with $\operatorname{acq}_1 \bowtie \operatorname{acq}_2$, and $\operatorname{match}_t(\operatorname{acq}_1) \notin X$, we have $\operatorname{rel}_2 <_P$ acq_1 , where $\operatorname{rel}_2 = \operatorname{match}_t(\operatorname{acq}_2)$. If X is feasible, we define the *canonical rf-poset* of X as $(X, Q, \operatorname{RF}_t | X)$, where Q is the weakest among all such partial orders P. It is easy to see that Q is well-defined, i.e., there exists at most one weakest partial order among all such partial orders P.

The realizability problem of feasible trace ideals. The realizability problem for an rf-poset $\mathcal{P} = (X, P, RF)$ asks whether there exists a linearization t^* of P such that $RF_{t^*} = RF$. Given a trace t and a feasible trace ideal X of t, the realizability problem for X is the realizability problem of the canonical rf-poset (X, P, RF) of X. The following remark relates the decision problem of dynamic race prediction in t with the realizability of trace ideals of t.

Remark 1. If t^* is a witness of the realizability of X, then t^* is a correct reordering of t. Two conflicting events $e_1, e_2 \in \mathcal{E}(t)$ are a predictable data race of t iff there exists a realizable trace ideal X of t such that e_1, e_2 are enabled in X.

Read pairs and triplets. For notational convenience, we introduce the notion of read pairs and read triplets. Given an rf-poset $\mathcal{P} = (X, P, RF)$, a *read pair* (or *pair* for short) of \mathcal{P} is a pair (w, r) such that $r \in \mathcal{RL}(X)$ and w = RF(r) (note that $w \in X$). A *read triplet* (or *triplet* for short) is a triplet (w, r, w') such that (i) (w, r) is a pair of \mathcal{P} , (ii) $w' \in X$, and (iii) $w' \neq w$ and $r \bowtie w'$. We denote by Pairs(\mathcal{P}) and Triplets(\mathcal{P}) the set of pairs and triplets of \mathcal{P} , respectively.

Closed rf-posets. We call an rf-poset $\mathcal{P} = (X, P, \text{RF})$ *closed* if for every triplet $(w, r, w') \in \text{Triplets}(\mathcal{P})$, we have (i) if $w' <_P r$ then $w' <_P w$, and (ii) if $w <_P w'$ then $r <_P$ w'. Given, an rf-poset $\mathcal{P} = (X, P, \text{RF})$, the *closure* of \mathcal{P} is an rf-poset $\mathcal{Q} = (X, Q, \text{RF})$ where Q is the weakest partial order over X such that $Q \sqsubseteq P$ and Q is closed. If no such Q exists, we let the closure of \mathcal{P} be \bot . The closure is welldefined [31]. The associated Closure problem is, given an rf-poset \mathcal{P} , decide whether the closure of \mathcal{P} is not \bot .

Remark 2. An rf-poset is realizable only if its closure exists and is realizable.

3.2 Bounds on the Number of Feasible Trace Ideals

Remark 1 suggests that the dynamic data-race prediction problem for a trace t is reducible to deciding whether t has some realizable trace ideal. In general, if t has length n and kthreads, there exist n^k possible trace ideals to test for realizability. Here we derive another upper-bound on the number of such ideals that are sufficient to test, based on the number of threads of t, its lock-nesting depth and its lock-dependence factor. These parameters typically behave as constants in practice, and thus understanding the complexity of dynamic data race prediction in terms of these parameters is crucial. **Causal cones.** Given an event $e \in \mathcal{E}(t)$, the *causal cone* Cone_t(e) of $e \in \mathcal{E}(t)$ is the smallest trace ideal X of t so that e is enabled in X. In words, we construct Cone_t(e) by taking the TRF-downwards closure of the thread-local predecessor e' of e (i.e., e' $\ll_{TO} e$). Given a non-empty set of events $S \subseteq \mathcal{E}(t)$, we define the causal cone of S as Cone_t(S) = $\bigcup_{e \in S} \text{Cone}_t(e)$; notice that Cone_t(S) is a trace ideal.

Candidate ideal set. Given a set of events *X*, we denote by OpenAcqs(*X*) the set of lock-acquire events acq such that match_t(acq) $\notin X$. Given two events $e_1, e_2 \in \mathcal{E}(t)$, the *candidate ideal set* CIS_t(e_1, e_2) of e_1, e_2 is the smallest set of trace ideals of *t* such that the following hold.

- 1. $\text{Cone}_t(\{e_1, e_2\}) \in \text{CIS}_t(e_1, e_2).$
- 2. Let $Y \in ClS_t(e_1, e_2)$, acq $\in OpenAcqs(Y)$, rel = match_t(acq), and $Y' = Cone_t(Y \cup \{rel\}) \cup \{rel\}$. If $e_1, e_2 \notin Y'$, then $Y' \in ClS_t(e_1, e_2)$.

In light of Remark 1, we will decide whether (e_1, e_2) is a predictable data race by deciding the realizability of ideals in the candidate ideal set. Item 2 states that, as long as there is some ideal Y in the candidate ideal set such that Y leaves some critical section open, we construct another ideal $Y' \supset Y$ by choosing one such open critical section and closing it, and add Y' in the candidate set as well. Intuitively, the open critical section of Y might deem Y not realizable, while closing that critical section might make Y' realizable. Clearly, if $e_1 \in Y'$ or $e_2 \in Y'$, then the realizability of Y' does not imply a data race on e_1, e_2 as one of the two events is not enabled in Y' (Remark 1). As the following lemma shows, in order to decide whether (e_1, e_2) is a predictable data race of t, it suffices to test for realizability all the ideals in $\text{ClS}_t(e_1, e_2)$.

Lemma 3.1. (e_1, e_2) is a predictable data race of t iff there exists a realizable ideal $X \in \text{CIS}_t(e_1, e_2)$ such that $e_1, e_2 \notin X$.

The following lemma gives an upper-bound on $|CIS_t(e_1, e_2)|$, i.e., on the number of ideals we need to test for realizability. **Lemma 3.2.** We have $|CIS_t(e_1, e_2)| \leq \min(n, \alpha)^{k-2}$, where $\alpha = k \cdot \gamma \cdot \zeta$, and k is the number of threads, γ is the lock-nesting depth, and ζ is the lock-dependence factor of t.

4 The General Case

In this section we address the general case of dynamic datarace prediction. The section is organized in two parts, which present the formal details of Theorem 2.2 and Theorem 2.3.

4.1 Upper Bound

In this section we establish Theorem 2.2. Recall that, by Lemma 3.1, the problem is reducible to detecting a realizable rf-poset in the candidate ideal set of the two events that are tested for a data-race. Rf-poset realizability is known to be NP-complete [16], and solvable in polynomial time when the number of threads is bounded [2]. Here we establish more precise upper-bounds, based on the number of threads. In particular, we show the following.

Lemma 4.1. *Rf*-poset realizability can be solved in $O(k \cdot n^k)$ time for an *rf*-poset of size *n* and *k* threads.

Frontiers and extensions. Let $\mathcal{P} = (X, P, RF)$ be an rfposet, and consider an ideal *Y* of \mathcal{P} . The *frontier* of *Y*, denoted Frontier $\mathcal{P}(Y)$, is the set of pairs $(w, r) \in \text{Pairs}(\mathcal{P})$ such that $w \in Y$ and $r \notin Y$. An event *e* executable in *Y* is said to *extend Y* if for every triplet $(w, r, e) \in \text{Triplets}(\mathcal{P})$, we have $(w, r) \notin \text{Frontier}_{\mathcal{P}}(Y)$. In this case, we say that $Y \cup \{e\}$ is an *extension* of *Y* via *e*.

Ideal graphs and canonical traces. Let $\mathcal{P} = (X, P, RF)$ be an rf-poset. The *ideal graph* of \mathcal{P} , denoted $G_{\mathcal{P}} = (V_{\mathcal{P}}, E_{\mathcal{P}})$ is a directed graph defined as follows.

- 1. $V_{\mathcal{P}}$ is the set of ideals of \mathcal{P} .
- 2. We have $(Y_1, Y_2) \in E_{\mathcal{P}}$ iff Y_2 is an extension of Y_1 .

The *ideal tree* of \mathcal{P} , denoted $T_{\mathcal{P}} = (I_{\mathcal{P}}, \mathcal{R}_{\mathcal{P}})$ is a (arbitrary) spanning tree of $G_{\mathcal{P}}$ when restricted to nodes reachable from \emptyset . We let \emptyset be the root of $T_{\mathcal{P}}$. Given an ideal $Y \in I_{\mathcal{P}}$, we define the *canonical trace* t_Y of Y inductively, as follows. If $Y = \emptyset$ then $t_Y = \epsilon$. Otherwise, Y has a parent Y' in $T_{\mathcal{P}}$ such that $Y = Y' \cup \{e\}$ for some event $e \in X$. We define $t_Y = t_{Y'} \circ e$. Lemma 4.1 relies on the following lemmas. We refer to Appendix C.1 for the proofs.

Lemma 4.2. We have $X \in I_{\mathcal{P}}$ iff \mathcal{P} is realizable.

Lemma 4.3. The ideal graph $G_{\mathcal{P}}$ has $O(n^k)$ nodes.

Proof of Theorem 2.2. Consider a trace *t* and two conflicting events *e*₁, *e*₂ ∈ *WR*(*t*). By Lemma 3.1, to decide whether (*e*₁, *e*₂) is a predictable data race of *t*, it suffices to iterate over all feasible trace ideals *X* in the candidate ideal set CIS_{*t*}(*e*₁, *e*₂), and test whether *X* is realizable. By Lemma 3.2, we have |CIS_{*t*}(*e*₁, *e*₂)| = *O*(*α*), where *α* = min(*n*, *k* · *γ* · *ζ*)^{*k*-2}. Finally, due to Lemma 4.1, the realizability of every such ideal can be performed in *O*(*k* · *n^k*) = *O*(*β*) time. □

4.2 Hardness of Data Race Prediction

Here we establish that the problem of dynamic data-race prediction is W[1]-hard when parameterized by the number of threads k. Our proof is established in two steps. In the first step, we show the following lemma.

Lemma 4.4. *Rf*-poset realizability parameterized by the number of threads k is W[1]-hard.

Rf-poset realizability is known to be NP-hard [16, Theorem 4.1], and Lemma 4.4 strengthens that result by showing that the problem is even unlikely to be FPT. In the second step, we show how the class of W[1]-hard instances of in Lemma 4.4 can be reduced to dynamic data-race prediction.

Hardness of rf-poset realizability. Our reduction is from the INDEPENDENT-SET(c) problem, which takes as input

an undirected graph G = (V, E) and asks whether G has an independent set of size c. INDEPENDENT-SET(c) parameterized by *c* is one of the canonical W[1]-hard problems [12]. Given an input G = (V, E) of INDEPENDENT-SET(c) with n = |V|, we construct an rf-poset $\mathcal{P}_G = (X, P, RF)$ of size $O(c \cdot n)$ and O(c) threads such that \mathcal{P}_G is realizable iff *G* has an independent set of size c. We assume wlog that every node in G has at least one neighbor, otherwise, we can remove all s such nodes and solve the problem for parameter c' = c - s. The rf-poset \mathcal{P}_G consists of $k = 2 \cdot c + 2$ total orders (X_i, τ_i) . Figure 1 provides an illustration. In high level, for each $i \in [c]$, τ_i and τ_{c+i} are used to encode the *i*-th copy of *G*, whereas the last two total orders are auxiliary. Superscripts on the events and/or their variables refer to the node of G that is encoded by those events. Below we describe the events and certain orderings between them. The partial order P is the transitive closure of these orderings.

1. For $i = 2 \cdot c + 1$, τ_i consists of a single event $\tau_i = w(x)$. 2. For $i = 2 \cdot c + 2$, we have $\tau_i = \sigma \circ \sigma$, where

$$\sigma = \mathsf{r}(s_1), \dots, \mathsf{r}(s_c), \operatorname{acq}(\ell_1), \dots, \operatorname{acq}(\ell_c) \quad \text{and}$$
$$\upsilon = \mathsf{r}(x), \operatorname{rel}(\ell_c), \dots, \operatorname{rel}(\ell_1).$$

3. For each $i \in [c]$, we have $\tau_i = \tau_i^1 \circ \tau_i^2 \circ \ldots \circ \tau_i^n$, where each τ_i^j encodes node *j* of *G* and is defined as follows. Let $\overline{\tau}_i^j = \sigma_i^j \circ \overline{\upsilon}_i^j$, where

$$\sigma_i^j = \operatorname{acq}_i(\ell_{\{j,l_1\}}), \dots, \operatorname{acq}_i(\ell_{\{j,l_m\}}) \quad \text{and} \\ \sigma_i^j = \operatorname{rel}_i(\ell_{\{j,l_m\}}), \dots, \operatorname{rel}_i(\ell_{\{j,l_1\}})$$

where l_1, \ldots, l_m are the neighbors of j in G. For each $j \in [n] \setminus \{1, n\}$, the sequence τ_i^j is identical to $\overline{\tau}_i^j$, with the addition that the innermost critical section (i.e., between $\operatorname{acq}_i(\ell_{\{j, l_m\}})$ and $\operatorname{rel}_i(\ell_{\{j, l_m\}})$) contains the sequence $w(y_i^j), r(z_i^j)$. The sequence τ_i^1 is defined similarly, except that the innermost critical section contains the sequence $w(s_i), r(z_i^1)$. Finally, the sequence τ_i^n is defined similarly, except that the innermost critical section contains the sequence $w(y_i^n), r(z_i^1)$. Finally, the sequence τ_i^n is defined similarly, except that the innermost critical section contains the sequence $w(y_i^n), r_i(x)$.

4. For each $i \in [c]$, we have $\tau_{c+i} = \tau_{c+i}^1 \circ \tau_{c+i}^2 \circ \ldots \circ \tau_{c+i}^{n-1}$, where $\tau_{c+i}^j = \operatorname{acq}^j(\ell_i), \operatorname{w}(z_i^j), \operatorname{r}(y_i^{j+1}), \operatorname{rel}^j(\ell_i)$.

Note that every memory location is written exactly once, hence the reads-from function RF is defined implicitly. In addition, for every read event r, we have RF(r) $<_P$ r, as well as $r(x) <_P r_i(x)$ for each $i \in [c]$.

Correctness. We now sketch the correctness of the construction, while we refer to Appendix C.2 for the proof. Assume that \mathcal{P} is realizable by a witness *t*. We say that r(x) *separates* a critical section in *t* if the lock-acquire (resp., lock-release) event of that critical section appears before (resp., after) r(x) in *t*. The construction guarantees that, for each $i \in [c]$, r(x) separates the critical sections of τ_i that encode some node l_i of *G*. By construction, these critical sections

are on locks $\ell_{\{l_i,v\}}$, where v ranges over the neighbors of l_i in *G*. Hence, for any $i' \neq i$, the node $l_{i'}$ cannot be a neighbor of l_i , as this would imply that both critical sections on lock $\ell_{\{l_i,l_{i'}\}}$ are opened before r(x) and closed after r(x) in *t*, which clearly violates lock semantics. Thus, an independent set $A = \{l_1, \ldots, l_c\}$ of *G* is formed by taking each l_i to be the node of *G*, the critical sections of which belong to thread τ_i and are separated by r(x) in *t*. On the other hand, if *G* has an independent set $A = \{l_1, \ldots, l_c\}$, a witness *t* that realizes \mathcal{P} can be constructed by separating the critical sections of the node l_i in τ_i , for each $i \in [c]$.

Hardness of dynamic data-race prediction. Finally, we turn our attention to the hardness of dynamic data-race prediction. Consider the INDEPENDENT-SET(*c*) problem on a graph *G* and the associated rf-poset $\mathcal{P}_G = (X, P, \text{RF})$ defined above. We construct a trace *t* with $\mathcal{E}(t) = X$ and RF_t = RF such that (w(x), r(x)) is a predictable data-race of *t* iff \mathcal{P}_G is realizable. In particular, *t* consists of $2 \cdot c + 2$ threads p_i , one for each total order τ_i of \mathcal{P}_G . We obtain *t* as

$$t = \tau_{2 \cdot c+1} \circ t_1 \circ \ldots \circ t_c \circ \tau_{2 \cdot c+2} ,$$

where each t_i is an appropriate interleaving of the total orders τ_i and τ_{c+i} that respects the reads-from function RF. In Appendix C.2, we conclude the proof of Theorem 2.3 by showing that *G* has an independent set of size *c* iff (w(*x*), r(*x*)) is a predictable data-race of *t*.

Remark 3. It is known that INDEPENDENT-SET(c) cannot be solved in $f(c) \cdot n^{o(c)}$ time under ETH [11]. As our reduction to rf-poset realizability and dynamic data-race prediction uses k = O(c) threads, each of these problems does not have a $f(k) \cdot n^{o(k)}$ -time algorithm under ETH.

5 Tree Communication Topologies

In this section we focus on the case where the input trace t constitutes a tree communication topology. The section is organized in two parts, which present the formal details of Theorem 2.4 and Theorem 2.5.

5.1 An Efficient Algorithm for Tree Topologies

In this section we present the formal details of Theorem 2.4. Recall that the Theorem 2.2 states an $O(\alpha \cdot \beta)$ upper-bound for dynamic data-race prediction, where β is the complexity of deciding rf-poset realizability, and α is an upper-bound on the number of candidate ideals whose realizability we need to check. For tree communication topologies, we obtain Theorem 2.4: (i) we show an improved upper-bound β on the complexity of the realizability of trace ideals over tree topologies, and (ii) we show that it suffices to check the realizability of a single trace ideal (i.e., $\alpha = 1$). We start with point (i), and then proceed with (ii).

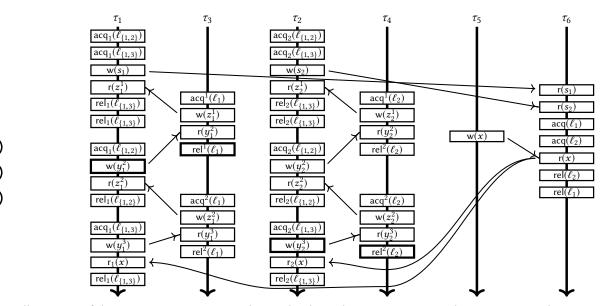


Figure 1. Illustration of the o-poset \mathcal{P}_G given a graph *G* and independent-set size c = 2. Edges represent orderings in *P*.

Tree-inducible rf-posets. Let (X, P) be a poset where $X \subseteq$ $\mathcal{E}(t)$. We call (X, P) tree-inducible if X can be partitioned into k sets $\{X_i\}_{1 \le i \le k}$ such that the following conditions hold.

- 1. The graph $T = ([k], \{(i, j) | X_i \bowtie X_i\})$ is a tree.
- 2. $P|X_i$ is a total order for each $i \in [k]$.

G

3. For every node $\ell \in [k]$ such that ℓ is an internal node in *T* and for every two connected components C_1 , C_2 of T that are created after removing ℓ from T, we have the following. Consider two nodes $i \in C_1$ and $j \in C_2$ and two events $e_1 \in X_i$ and $e_2 \in X_j$ such that $e_1 <_P e_2$, there exists some event $e \in X_{\ell}$ such that $e_1 <_P e <_P e_2$.

We call an rf-poset (X, P, RF) tree-inducible if (X, P) is treeinducible. The insight is that traces from tree communication topologies yield tree-inducible trace ideals. Our motivation behind tree inducibility comes from the following lemma.

Lemma 5.1. *Rf-poset realizability of tree-inducible rf-posets* can be solved in $O(k^2 \cdot d \cdot n^2 \cdot \log n)$ time, for an rf-poset of size n, k threads and d variables.

The proof of Lemma 5.1 is in two steps. Recall the definition of closed rf-posets from Section 3.1. First, we show that a treeinducible, closed rf-poset is realizable (Lemma 5.2). Second, we show that the closure of a tree-inducible rf-poset is also tree-inducible (Lemma 5.3).

Lemma 5.2. Every closed, tree-inducible rf-poset is realizable.

Indeed, consider a tree-inducible, closed rf-poset \mathcal{P} = (X, P, RF). The witness *t* realizing \mathcal{P} is obtained in two steps.

1. We construct a poset (X, Q) with $Q \sqsubseteq P$ as follows. Initially, we let Q be identical to P. Let \mathcal{P} be tree-inducible to a tree $T = ([k], \{(i, j) | X_i \bowtie X_i\})$. We traverse T topdown, and for every node *i* and child *j* of *i*, for every two

- events $e_1 \in X_i$ and $e_2 \in X_j$ with $e_1 \bowtie e_2$ and $e_2 \not\leq_P e_1$, we order $e_1 <_O e_2$. Finally, we transitively close Q. 2. We construct *t* by linearizing (X, Q) arbitrarily.

In Appendix D.1 we show that *t* is well-defined and realizes \mathcal{P} . The next lemma shows that tree-inducibility is preserved under taking closures (if the closure exists).

Lemma 5.3. Consider an rf-poset $\mathcal{P} = (X, P, RF)$ and let Q = (X, Q, RF) be the closure of \mathcal{P} . If \mathcal{P} is tree-inducible then Q is also tree-inducible.

Since, by Remark 2, an rf-poset is realizable only if its closure exists and is realizable, Lemma 5.2 and Lemma 5.3 allow to decide the realizability of an rf-poset by computing its closure. The complexity of the algorithm comes from the complexity of deciding whether the closure exists. We refer to Appendix D.1 for the full proof of Lemma 5.1.

Recall that Lemma 3.2 provides an upper-bound on the number of trace ideals of t that we need to examine for realizability in order to decide data-race prediction. We now proceed with point (ii) towards Theorem 2.4, i.e., we show that for tree communication topologies, a single ideal suffices. Our proof is based on the notion of lock causal cones below.

Lock causal cones. Consider a trace *t* that defines a tree communication topology $G_t = (V_t, E_t)$. Given an event $e \in$ $\mathcal{E}(t)$ the lock causal cone LCone_t(e) of e is the set X defined by the following process. Consider that G_t is rooted in p(e).

- 1. Initially X contains all predecessors of e in $(\mathcal{E}(t), TO)$. We perform a top-down traversal of G_t , and consider a current thread p_1 visited by the traversal.
- 2. Let p_2 be the parent of p_1 in the traversal tree, and e_2 be the unique maximal event in $(X|p_2, TO)$, i.e., e_2 is the last

event of thread p_2 that appears in *X*. We insert in *X* all events $e_1 \in \mathcal{E}(t)|p_1$ such that $e_1 <_{\mathsf{TRF}} e_2$.

3. While there exists some lock-acquire event $acq_1 \in X|p_1$ and there exists another lock-acquire event $acq_2 \in$ OpenAcqs(X) with $acq_1 \bowtie acq_2$ and $p(acq_2) = p_2$, we insert in X all predecessors of rel₁ in ($\mathcal{E}(t)$, TO) (including rel₁), where rel₁ = match_t(acq₁).

Observe that, by construction, $\text{LCone}_t(e)$ is a lock-feasible trace ideal of *t*. In addition, for any two events $e_1, e_2 \in \mathcal{E}(t)$, the set $\text{LCone}_t(e_1) \cup \text{LCone}_t(e_2)$ is an ideal of *t*, though not necessarily lock-feasible. Our motivation behind lock causal cones comes from the following lemma. Intuitively, we can decide a predictable data-race by deciding the realizability of the ideal that is the union of the two lock-causal cones.

Lemma 5.4. Let $X = \text{LCone}_t(e_1) \cup \text{LCone}_t(e_2)$. We have that (e_1, e_2) is a predictable data race of t iff $(i) \{e_1, e_2\} \cap X = \emptyset$, and (ii) X is a realizable trace ideal of t.

The (\Leftarrow) direction of the lemma is straightforward. We refer to Appendix D.1 for the (\Rightarrow) direction. Finally, Theorem 2.4 follows immediately from Lemma 5.1 and Lemma 5.4.

Proof of Theorem 2.4. By Lemma 5.4, we have that (e_1, e_2) is a predictable data race of t iff $\{e_1, e_2\} \cap X = \emptyset$ and X is realizable. By Lemma 5.1, deciding the realizability of X is done in $O(k^2 \cdot d \cdot n^2 \cdot \log n)$ time. The desired result follows. \Box

5.2 A Lower Bound for Two Threads

In this section we prove a conditional quadratic lower bound for dynamic data-race prediction for two threads. Our proof is via a reduction from the Orthogonal Vectors problem. To make it conceptually simpler, we present our reduction in two steps. First, we show a fine-grained reduction from Orthogonal Vectors to the realizability of an rf-poset with 2 threads and 7 variables. Afterwards, we show how the realizability problem for the rf-posets of the first step can be reduced to the decision problem of dynamic data race prediction with 2 threads, 9 variables and 1 lock.

The Orthogonal Vectors problem (OV). An instance of Orthogonal Vectors consists of two sets *A*, *B*, where each set contains n/2 binary vectors in *D* dimensions. The task is to determine whether there exists a pair of vectors $(a, b) \in A \times B$ that is orthogonal, i.e., for all $i \in [D]$ we have $a[i] \cdot b[i] = 0$. There exist algorithms that solve the problem in $O(n^2 \cdot D)$ and $O(2^D \cdot n)$ time, simply by computing the inner product of each pair $(a, b) \in A \times B$ and following a classic Four-Russians technique, respectively. It is conjectured that there is no truly sub-quadratic algorithm for OV [8].

Conjecture 5.5 (Orthogonal Vectors). There is no algorithm for OV that operates in $O(n^{2-\epsilon} \cdot D^{O(1)})$ time, for any $\epsilon > 0$.

It is also known that SETH implies the OV conjecture [42]. We first relate OV with rf-poset realizability. **Lemma 5.6.** *Rf*-poset realizability for an *rf*-poset with 2 threads and 7 variables has no $O(n^{2-\epsilon})$ -time algorithm for any $\epsilon > 0$, under the Orthogonal Vectors conjecture.

Reduction from OV to **rf-poset realizability.** For a finegrained reduction from OV to rf-poset realizability, consider an OV instance (A, B), where $A = (a_j)_{1 \le j \le n/2}$, $B = (b_l)_{1 \le l \le n/2}$, and each $a_j, b_l \in \{0, 1\}^D$. We will construct an rf-poset $\mathcal{P} = (X, P, RF)$ with 2 threads and 7 variables such that the closure of \mathcal{P} exists iff there exists a pair of orthogonal vectors $(a, b) \in A \times B$. Since 2 threads define a tree-inducible rf-poset, Remark 2 and Lemma 5.2 imply that \mathcal{P} is realizable iff OV has a positive answer. The set *X* consists of two disjoint sets X_A, X_B , so that each is totally ordered in *P*. For ease of presentation, we denote by τ_A, τ_B the linear orders $(X_A, P|X_A)$ and $(X_B, P|X_B)$, respectively. To develop some insight, we start with a high-level view of the construction, and then proceed with the details.

Overview of the construction. The linear orders τ_A and τ_B encode the vectors of *A* and *B*, respectively. Each of τ_A and τ_B consists of n/2 segments, so that the *i*-th (resp. (n/2-i+1)th) segment of τ_A (resp. τ_B) encodes the contents of the *i*-th vector of A (resp., B). The two total orders are constructed with a closure computation in mind, which inserts event orderings in *P* one-by-one. In high-level, an ordering $e_1 < e_2$ encodes the test of whether the bits in a specific coordinate *i* of two vectors $a_i \in A$ and $b_l \in B$ have product 0. If yes, and moreover, i < D, then the closure conditions enforce a new ordering $e'_1 < e'_2$, which encodes the test of the bits in coordinate i + 1. Otherwise i = D, and the closure has been computed and an orthogonal pair has been found. On the other hand, if the bits in coordinate *i* have product 1, the two current vectors are not orthogonal, and the closure conditions enforce a new ordering $e_1'' < e_2''$, which encodes the test of the first coordinate of the next pair of vectors. The above is achieved using 7 variables $\{x_i\}_{i \in [7]}$.

Formal construction. We now present the formal details of the construction (illustrated in Figure 2). The construction creates various events which have the form e^{a_j} and e^{b_l} when they are used at the vector level, and have the form $e_i^{a_j}$ and $e_i^{b_l}$, where $i \in [D]$, when they are used at the coordinate level. As a general rule, for each $j, l \in [n/2 - 1]$, we have $e^{a_j} <_{\tau_A} e^{a_{j+1}}$ and $e^{b_{l+1}} <_{\tau_B} e^{b_l}$, both for events at the vector and at the coordinate level. At the coordinate level, we also have $e_{i+1}^{a_j} <_{\tau_A} e_i^{a_j}$ and $e_{i+1}^{b_l} <_{\tau_B} e_i^{b_l}$. For succinctness, we often write $e_1, e_2 < e_3$ to denote $e_1 < e_3$ and $e_2 < e_3$. We next describe the events and orderings between them. The partial order P is the transitive closure of these orderings.

Events on x_1 *and* x_2 . For every vector $a_j \in A$ and coordinate $i \in [D]$, we create three events $w_i^{a_j}(x_1)$, $w_i^{a_j}(x_2)$ and $r_i^{a_j}(x_2)$. We make $\text{RF}(r_i^{a_j}(x_2)) = w_i^{a_j}(x_2)$, and order

$$w_i^{a_j}(x_1), w_i^{a_j}(x_2) <_{\tau_A} r_i^{a_j}(x_2) .$$
 (1)

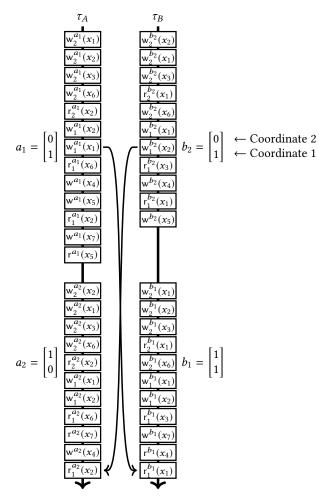


Figure 2. Illustration of the reduction of an OV instance $(A = \{a_1, a_2\}, B = \{b_1, b_2\})$ to the Closure problem of an \mathcal{P} .

For every vector $b_l \in B$ and coordinate $i \in [D]$, we create three events $w_i^{b_l}(x_1)$, $r_i^{b_l}(x_1)$ and $w_i^{b_l}(x_2)$. We make $RF(r_i^{b_l}(x_1)) = w_i^{b_l}(x_1)$, and order

$$w_i^{b_l}(x_1), w_i^{b_l}(x_2) <_{\tau_B} r_i^{b_l}(x_1)$$
 (2)

In addition, we order

$$w_{i}^{a_{j}}(x_{2}) <_{\tau_{A}} w_{i}^{a_{j}}(x_{1}) \quad \text{iff} \quad a_{j}[i] = 1 \qquad \text{and} \\ w_{i}^{b_{l}}(x_{1}) <_{\tau_{B}} w_{i}^{b_{l}}(x_{2}) \quad \text{iff} \quad b_{l}[i] = 1 .$$
(3)

Observe that if $a_j[i] \cdot b_l[i] = 1$, and if we order $w_i^{a_j}(x_1) < w_i^{b_l}(x_1)$ then transitively $w_i^{a_j}(x_2) < w_i^{b_l}(x_2)$ and hence by closure $r_i^{a_j}(x_2) < w_i^{b_l}(x_2)$.

Events on x_3 . Let $i \in [D-1]$ be a coordinate. For every vector $a_j \in A$, we create an event $w_{i+1}^{a_j}(x_3)$, and order

$$\mathsf{w}_{i+1}^{a_j}(x_1), \mathsf{w}_{i+1}^{a_j}(x_2) <_{\tau_A} \mathsf{w}_{i+1}^{a_j}(x_3) <_{\tau_A} \mathsf{w}_i^{a_j}(x_1), \mathsf{w}_i^{a_j}(x_2) .$$
(4)

For every vector $b_l \in B$, we create two events $w_{i+1}^{b_l}(x_3)$ and $r_i^{b_l}(x_3)$, and make $\operatorname{RF}(r_i^{b_l}(x_3)) = w_{i+1}^{b_l}(x_3)$. We order

$$w_{i+1}^{b_l}(x_1), w_{i+1}^{b_l}(x_2) <_{\tau_B} w_{i+1}^{b_l}(x_3) <_{\tau_B} r_{i+1}^{b_l}(x_1) \quad \text{and} \\ w_i^{b_l}(x_1), w_i^{b_l}(x_2) <_{\tau_B} r_i^{b_l}(x_3) <_{\tau_B} r_i^{b_l}(x_1) .$$
(5)

Observe that if we order $w_i^{a_j}(x_1) < w_i^{b_l}(x_1)$ then we also have $w_{i+1}^{a_j}(x_3) < r_i^{b_l}(x_3)$, hence by closure $w_{i+1}^{a_j}(x_3) < w_{i+1}^{b_l}(x_3)$ and thus $w_{i+1}^{a_j}(x_1) < r_i^{b_l}(x_1)$.

Events on x_6 . For every coordinate $i \in [D-1]$, we do as follows. For every vector $a_j \in A$, we create two events $w_{i+1}^{a_j}(x_6)$ and $r_i^{a_j}(x_6)$, and make $RF(r_i^{a_j}(x_6)) = w_{i+1}^{a_j}(x_6)$. We order

$$\begin{split} & \mathsf{w}_{i+1}^{a_j}(x_3) <_{\tau_A} \mathsf{w}_{i+1}^{a_j}(x_6) <_{\tau_A} \mathsf{r}_{i+1}^{a_j}(x_2) \quad \text{and} \\ & \mathsf{w}_i^{a_j}(x_1), \mathsf{w}_i^{a_j}(x_2) <_{\tau_A} \mathsf{r}_i^{a_j}(x_6) <_{\tau_A} \mathsf{r}_i^{a_j}(x_2) \;. \end{split}$$

For every vector $b_l \in B$, we create one event $w_{i+1}^{b_l}(x_6)$, and order

$$\mathsf{r}_{i+1}^{b_l}(x_1) <_{\tau_B} \mathsf{w}_{i+1}^{b_l}(x_6) <_{\tau_B} \mathsf{w}_i^{b_l}(x_1), \mathsf{w}_i^{b_l}(x_2)$$
. (7)

Observe that if we order $r_{i+1}^{a_j}(x_2) < w_{i+1}^{b_l}(x_2)$, since $w_{i+1}^{b_l}(x_2) <_{\tau_B} r_{i+1}^{b_l}(x_1)$, we also have $w_{i+1}^{a_j}(x_6) < w_{i+1}^{b_l}(x_6)$, hence by closure $r_i^{a_j}(x_6) < w_{i+1}^{b_l}(x_6)$ and thus $w_i^{a_j}(x_2) < w_i^{b_l}(x_2)$.

Events on x_4 . For every vector $a_j \in A$, we create one event $w^{a_j}(x_4)$, and order

$$\mathsf{r}_{1}^{a_{j}}(x_{6}) <_{\tau_{A}} \mathsf{w}^{a_{j}}(x_{4}) <_{\tau_{A}} \mathsf{r}_{1}^{a_{j}}(x_{2})$$
 (8)

For every vector $b_l \in B$, with $l \in [n/2 - 1]$, we create two events $w^{b_{l+1}}(x_4)$ and $r^{b_l}(x_4)$, and make $RF(r^{b_l}(x_4)) = w^{b_{l+1}}(x_4)$. We order

$$r_1^{b_l}(x_3) <_{\tau_B} r^{b_l}(x_4) <_{\tau_B} r_1^{b_l}(x_1) \text{ and } r_1^{b_{l+1}}(x_3) <_{\tau_B} w^{b_{l+1}}(x_4) <_{\tau_B} r_1^{b_{l+1}}(x_1) .$$
 (9)

Observe that if we order $r_1^{a_j}(x_2) < w_1^{b_l}(x_2)$ then we also have $w^{a_j}(x_4) < r^{b_l}(x_4)$ (since $w_1^{b_l}(x_2) <_{\tau_B} r_1^{b_l}(x_3)$ by Eq. (5)) and thus by closure $w^{a_j}(x_4) < w^{b_{l+1}}(x_4)$.

Events on x_5 *and* x_7 . For every vector $a_j \in A$, with $j \in [n/2-1]$, we create two events $w^{a_j}(x_5)$ and $r^{a_j}(x_5)$, and make $RF(r^{a_j}(x_5)) = w^{a_j}(x_5)$. We also create two events $w^{a_j}(x_7)$ and $r^{a_{j+1}}(x_7)$, and make $RF(r^{a_{j+1}}(x_7)) = w^{a_j}(x_7)$. We order

$$w^{a_{j}}(x_{4}) <_{\tau_{A}} w^{a_{j}}(x_{5}) <_{\tau_{A}} r_{1}^{a_{j}}(x_{2}) <_{\tau_{A}} w^{a_{j}}(x_{7}) <_{\tau_{A}} r^{a_{j}}(x_{5}) \text{ and} r_{1}^{a_{j+1}}(x_{6}) <_{\tau_{A}} r^{a_{j+1}}(x_{7}) <_{\tau_{A}} w^{a_{j+1}}(x_{4}) .$$
(10)

We also create two events $w^{b_{n/2}}(x_5)$ and $w^{b_1}(x_7)$, and order

$$r_{1}^{b_{n/2}}(x_{1}) <_{\tau_{B}} w^{b_{n/2}}(x_{5}) <_{\tau_{B}} w_{D}^{b_{n/2}-1}(x_{1}), w_{D}^{b_{n/2}-1}(x_{2}) \quad \text{and} \\ r_{1}^{b_{1}}(x_{3}) <_{\tau_{B}} w^{b_{1}}(x_{7}) <_{\tau_{B}} r^{b_{1}}(x_{4}) .$$

$$(11)$$

Observe that if we order $r_1^{a_j}(x_2) < w_1^{b_{n/2}}(x_2)$ then we also have $w^{a_j}(x_5) < w^{b_{n/2}}(x_5)$ (since $w_1^{b_{n/2}}(x_2) <_{\tau_B} r_1^{b_{n/2}}(x_1)$ by a

previous item) and thus by closure $r^{a_j}(x_5) < w^{b_{n/2}}(x_5)$. But then also $w^{a_j}(x_7) < w^{b_1}(x_7)$ (since $w^{b_{n/2}}(x_5) <_{\tau_B} w^{b_1}(x_7)$) and thus by closure $r^{a_{j+1}}(x_7) < w^{b_1}(x_7)$.

Final orderings. Finally, we order

$$\mathbf{w}^{b_{n/2}}(x_5) <_{\tau_B} \mathbf{w}_D^{b_{n/2-1}}(x_1), \mathbf{w}_D^{b_{n/2-1}}(x_2)$$
 (12)

For each $j \in [n/2 - 1]$, we order

$$a_{j}(x_{5}) <_{\tau_{A}} \mathsf{w}_{D}^{a_{j+1}}(x_{1}), \mathsf{w}_{D}^{a_{j+1}}(x_{2})$$

For each $j \in [n/2 - 2]$, we order

$$\mathbf{r}^{b_{j+1}}(x_1) <_{\tau_B} \mathbf{w}_D^{b_j}(x_1), \mathbf{w}_D^{b_j}(x_2)$$
 (13)

We make two orderings across τ_A and τ_B , namely

$$w_1^{a_1}(x_1) <_P r_1^{b_1}(x_1)$$
 and $w_1^{b_{n/2}}(x_2) <_P r_1^{a_{n/2}}(x_2)$. (14)

Correctness. Observe that we have used 7 variables, while $|X_A| + |X_B| = O(n \cdot D)$, and the reduction can be easily computed in linear time. We refer to Appendix D.2 for the full proofs of the correctness of the above construction. This concludes Lemma 5.6, as any algorithm for rf-poset realizability on the above instances that runs in $O((n \cdot D)^{2-\epsilon})$ time also solves OV in $O(n^{2-\epsilon} \cdot D^{O(1)})$ time. Although the full proof is rather technical, the correctness is conceptually straightforward. We illustrate the key idea on the example of Figure 2, where we perform closure operations by inserting new orderings in a partial order <.

By construction, we have $w_1^{a_1}(x_1) < r_1^{b_1}(x_1)$, which signifies testing the first coordinate of vectors a_1 and b_1 . Note that $a_1[1] \cdot b_1[1] = 1$, for which our encoding guarantees that eventually $r_1^{a_1}(x_2) < w_1^{b_1}(x_2)$. Indeed, since $w_1^{a_1}(x_1) < r_1^{b_1}(x_1)$, by closure we also have $w_1^{a_1}(x_1) < w_1^{b_1}(x_1)$. In turn, this leads to $w_1^{a_1}(x_2) < w_1^{b_1}(x_2)$, and by closure, we also have $r_1^{a_1}(x_2) < w_1^{b_1}(x_2)$. This leads to $w^{a_1}(x_4) < r^{b_1}(x_4)$, and by closure, we have $w^{a_1}(x_4) < w^{b_2}(x_4)$. This last ordering leads to $w_1^{a_1}(x_1) < r_1^{b_2}(x_1)$, which signifies testing the first coordinate of vectors a_1 and b_2 , i.e., moving with the next vector of *B*.

The process for a_1 and b_2 is similar to a_1 and b_1 , as the two vectors are found not orthogonal already in the first coordinate. As previously, we eventually arrive at $r_1^{a_1}(x_2) < w_1^{b_2}(x_2)$. Note that this leads to $w^{a_1}(x_5) < w^{b_2}(x_5)$, and by closure, we have $r^{a_1}(x_5) < w^{b_2}(x_5)$. In turn, this leads to $w^{a_1}(x_7) < w^{b_1}(x_7)$, and by closure, we have $r^{a_2}(x_7) < w^{b_1}(x_7)$. This last ordering leads to $w_1^{a_2}(x_1) < r_1^{b_1}(x_1)$, which signifies testing the first coordinate of vectors a_2 and b_1 , i.e., moving with the next vector of A and the first vector of B.

The process for a_2 and b_1 is initially different than before, as $a_2[1] \cdot b_1[1] = 0$, i.e., the test on the first coordinate does not deem a_2 and b_1 not orthogonal. By closure, the ordering $w_1^{a_2}(x_1) < r_1^{b_1}(x_1)$ leads to $w_1^{a_2}(x_1) < w_1^{b_1}(x_1)$. This leads to $w_2^{a_2}(x_3) < r_1^{b_1}(x_3)$, and by closure, we have $w_2^{a_2}(x_3) < w_2^{b_1}(x_3)$. This leads to $w_2^{a_2}(x_1) < r_2^{b_1}(x_1)$, which signifies testing the second coordinate of vectors a_2 and b_1 . As $a_2[2] \cdot b_2[2] = 1$, the two vectors are discovered as non-orthogonal, which is captured by an eventual ordering $r_2^{a_2}(x_2) < w_2^{b_1}(x_2)$. This ordering which witnesses non-orthogonality is propagated downwards to the first coordinate, i.e., $r_1^{a_2}(x_2) < w_2^{b_1}(x_2)$. This propagation is made by events on variable x_6 . Indeed, first note that, as $r_2^{a_2}(x_2) < w_2^{b_1}(x_2)$, we also have $w_2^{a_2}(x_6) < w_2^{b_1}(x_6)$, and by closure, we have $r_1^{a_2}(x_6) < w_2^{b_1}(x_6)$. This leads to $w_1^{a_2}(x_2) < w_1^{b_1}(x_2)$, and by closure, $r_1^{a_2}(x_2) < w_2^{b_1}(x_2)$, which marks the two vectors as non-orthogonal. This leads to $w^{a_2}(x_4) < r^{b_1}(x_4)$, and by closure, we have $w^{a_2}(x_4) < w^{b_2}(x_4)$. This last ordering leads to $w_1^{a_2}(x_1) < r_1^{b_2}(x_1)$, which signifies testing the first coordinate of vectors a_2 and b_2 , i.e., moving with the next vector of B.

The process for a_2 and b_2 is initially similar to the previous case, as $a_2[1] \cdot b_2[1] = 0$. However, because we also have $a_2[2] \cdot b_2[2] = 0$, we will *not* order $r_2^{a_2}(x_2) < w_2^{b_2}(x_2)$, and the closure will terminate after ordering $w_2^{a_2}(x_1) < w_2^{b_2}(x_1)$. Since no cyclical orderings were introduce, the closure of \mathcal{P} exists, and by Lemma 5.2, \mathcal{P} is realizable. Finally, observe that *if* we eventually had $r_1^{a_2}(x_2) < w_1^{b_2}(x_2)$ (signifying that a_2 and b_2 are not orthogonal, hence there is no orthogonal pair in $A \times B$), this would create a cycle with the ordering $w_1^{b_2}(x_2) <_P r_1^{a_2}(x_2)$, and by Remark 2, \mathcal{P} would *not* be realizable.

Reduction to dynamic data-race prediction. Consider an instance of the rf-poset $\mathcal{P} = (X, P, RF)$ constructed in the above reduction, and we construct a trace t and two events $e_1, e_2 \in \mathcal{E}(t)$ such that \mathcal{P} is realizable iff (e_1, e_2) is a predictable data race of t. The trace t consists of two threads p_A, p_B and two local traces τ'_A and τ'_B such that τ'_A and τ'_B contain the events of p_A and p_B , respectively. Each of τ'_A and τ'_B is identical to τ_A and τ_B of \mathcal{P} , respectively, with some additional events inserted in it. In particular, besides the variables $x_i, i \in [7]$ that appear in the events of X, we introduce one variable y and one lock ℓ . For the event set, we have

$$\begin{split} \mathcal{E}(t) = & X \cup \{\mathsf{w}(y),\mathsf{r}(y)\} \cup \{\mathsf{acq}_A(\ell),\mathsf{rel}_A(\ell)\} \cup \\ \{\mathsf{acq}_B(\ell),\mathsf{rel}_B(\ell)\} \cup \{\mathsf{w}(z),\mathsf{r}(z)\} \;. \end{split}$$

The local traces τ'_A and τ'_B are constructed as follows.

- 1. For τ'_A , we insert an empty critical section $\operatorname{acq}_A(\ell)$, $\operatorname{rel}_A(\ell)$ right after $w_1^{a_1}(x_1)$. Additionally, we insert the read event r(y) right before $r_1^{a_{n/2}}(x_2)$, and the event r(z) as the last event of τ'_A .
- 2. For τ'_B , we insert the write event w(y) right after w₁^{$b_{n/2}$}(x_2). Additionally, we insert w(z) right after r₁^{b_1}(x_1), and surround these two events with acq_B(ℓ), rel_B(ℓ).

Finally, we obtain *t* as $t = \tau'_B \circ \tau'_A$, i.e., the two local traces are executed sequentially and there is no context switching. The task is to decide whether (w(z), r(z)) is a predictable data

6 Witnesses in Small Distance

The results in the previous sections neglect information provided by the input trace t about constructing a correct reordering that witnesses the data race. Indeed, our hardness results show that, in the worst case, the orderings in t provide no help. However, in practice when a data race exists, a witness trace t^* can be constructed that is similar to t. In fact, virtually all practical techniques predict data races by constructing t^* to be very similar to t (e.g., [15, 20, 31, 34, 40]).

The distance-bounded realizability problem of feasi**ble trace ideals.** Given a natural number ℓ , a trace *t* and a feasible trace ideal X of t, the solution to the ℓ -distancebounded realizability problem is False if X is not realizable, True if there is a witness t^* that realizes X such that $\delta(t, t^*) \leq \ell$, and can be any answer (True/False) if *X* is realizable but any witness t^* that realizes X is such that $\delta(t, t^*) > \ell$. We remark that this formulation is that of a promise prob*lem* [14]. We are interested in the case where $\ell = O(1)$. There exists a straightforward algorithm that operates in $O(|X|^{2 \cdot \ell})$ time. The algorithm iterates over all possible subsets of pairs of conflicting write and lock-acquire events that have size at most ℓ , and tries all possible combinations of conflicting-write reversals in that set. Theorem 2.6 is based on the following lemma, which states that the problem can be solved much faster when k is also constant.

Lemma 6.1. Consider a natural number ℓ , a trace t over n events and k threads, and a feasible trace ideal X of t. The ℓ -distance-bounded realizability problem for X can be solved in $O(k^{\ell+O(1)} \cdot n)$ time.

Proof of Theorem 2.6. By Lemma 6.1, given a trace ideal *X* of *t*, we can solve the ℓ -distance-bounded realizability problem for *X* in *O*(*n*) time. The proof then follows by Lemma 3.1 and Lemma 3.2, as to decide whether (e_1, e_2) is a predictable data race of *t*, it suffices to examine *O*(1) trace ideals of *t*.

In the remaining of this section we prove Lemma 6.1. We first define the notion of read extensions of graphs. Afterwards, we present the algorithm for the lemma, and show its correctness and complexity.

Read extensions. Consider a digraph G = (X, E) where X is a set of events. Given two events $e_1, e_2 \in G$, we write $e_1 \rightsquigarrow e_2$ to denote that e_2 is reachable from e_1 . We call G write-ordered if for every two distinct conflicting write or lock-acquire events $w_1, w_2 \in W\mathcal{L}(X)$, we have $w_1 \rightsquigarrow w_2$ or $w_2 \rightsquigarrow w_1$ in G. Given an acyclic write-ordered graph $G_1 = (X, E_1)$, the *read extension* of G_1 is the digraph $G_2 = (X, E_2)$ where
$$\begin{split} E_2 &= E_1 \cup A \cup B, \text{ where the sets } A \text{ and } B \text{ are defined as follows.} \\ A &= \{(\mathsf{r},\mathsf{w}) \in \mathcal{RL}(X) \times \mathcal{WL}(X) \mid \mathsf{r} \bowtie \mathsf{w} \text{ and } (\mathrm{RF}_t(\mathsf{r}),\mathsf{w}) \in E_1\} \ , \\ B &= \{(\mathsf{w},\mathsf{r}) \in \mathcal{WL}(X) \times \mathcal{RL}(X) \mid \mathsf{r} \bowtie \mathsf{w} \text{ and } (\mathsf{w},\mathrm{RF}_t(\mathsf{r})) \in E_1\} \ . \end{split}$$

A fast algorithm for distance-bounded rf-poset realizability. Let $\mathcal{P} = (X, P, RF)$ be the canonical rf-poset of X, and the task is to decide the realizability of \mathcal{P} with ℓ reversals. We describe a recursive algorithm for solving the problem for some rf-poset Q = (X, Q, RF) with ℓ' reversals, for some $\ell' \leq \ell$, where initially Q = P and $\ell' = \ell$.

Algorithm and correctness. Consider the set

$$C = \{ (\mathbf{w}_1, \mathbf{w}_2) \in \mathcal{WL}(X) \times \mathcal{WL}(X) \mid \mathbf{w}_1 \bowtie \mathbf{w}_2 \text{ and } \mathbf{w}_1 \mid |_Q w_2 \text{ and } \mathbf{w}_1 <_t \mathbf{w}_2 \}.$$

We construct a graph $G_1 = (X, E_1)$, where $E_1 = (\text{TRF}|X) \cup C$. Note that G_1 is write-ordered. If it is acyclic, we construct the read extension G_2 of G_1 . Observe that if G_2 is acyclic then any linearization t^* of G realizes Q, hence we are done. Now consider that either G_1 or G_2 is not acyclic, and let $G = G_1$ if G_1 is not acyclic, otherwise $G = G_2$. Given a cycle C of G, represented as a collection of edges, define the set of *crossedges* of C as $C \setminus Q$. Note that, since there are k threads, Ghas a cycle with $\leq k$ cross edges. In addition, any trace t^* that realizes Q must linearize an rf-poset (X, Q_a, RF) where $a = (e_1, e_2)$ ranges over the cross-edges of C. In particular, we take $Q_a = Q \cup \{b\}$, where

$$b = \begin{cases} (e_2, e_1), & \text{if } a \in \mathcal{WL}(X) \times \mathcal{WL}(X) \\ (\text{RF}(e_2), e_1), & \text{if } a \in \mathcal{WL}(X) \times \mathcal{RL}(X) \\ (e_2, \text{RF}(e_1)), & \text{if } a \in \mathcal{RL}(X) \times \mathcal{WL}(X) . \end{cases}$$

Observe that any such choice of *b* reverses the order of two conflicting write events or lock-acquire events in *t*. Since there are $\leq k$ cross edges in \mathcal{C} , there are $\leq k$ such choices for Q_a . Repeating the same process recursively for the rf-poset (X, Q_a, RF) for $\ell' - 1$ levels solves the ℓ' -distance-bounded realizability problem for Q. Since initially $\ell' = \ell$ and Q = P, this process solves the same problem for \mathcal{P} and thus for X.

Complexity. The recursion tree above has branching $\leq k$ and depth $\leq \ell$, hence there will be at most k^{ℓ} recursive instances. In Appendix E, we provide some lower-level algorithmic details which show that each instance can be solved in $O(k^{O(1)} \cdot n)$ time. The main idea is that each of the graphs G_1 and G_2 have a sparse transitive reduction [3] of size $O(k \cdot n)$, and thus each graph can be analyzed in $O(k \cdot n)$ time.

7 Conclusion

In this work, we have studied the complexity of dynamic data-race prediction, and have drawn a rich complexity landscape depending on various parameters of the input trace. Our main results indicate that the problem is in polynomial time when the number of threads is bounded, however, it is unlikely to be FPT wrt this parameter. On the other hand, we have shown that the problem can be solved in, essentially, quadratic time, when the communication topology is acyclic. We have also proved a quadratic lower bound for this case, which shows that our algorithm for tree communication topologies is optimal. Finally, motivated by practical techniques, we have shown that a distance-bounded version of data-race prediction can be solved in linear time under mild assumptions on the input parameters.

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A Details of Section 2

In this section we provide the proof of Lemma 2.1.

Lemma 2.1. Given a trace t of length n and two events $e_1, e_2 \in \mathcal{E}(t)$, we can construct a trace t' in O(n) time so that t' has a predictable data race iff (e_1, e_2) is a predictable data race of t.

Proof. We outline the construction of t'. We introduce two locks ℓ_1, ℓ_2 , and for each $i \in [2]$, we surround e_i with the lock ℓ_i , i.e., we replace e_i with $acq(\ell_i), e_i, rel(\ell_i)$. For every other event $e \in \mathcal{E}(t) \cap \mathcal{WR}(\mathcal{E}(t)) \setminus \{e_1, e_2\}$, we replace e with the following sequence

$$\operatorname{acq}(\ell_1), \operatorname{acq}(\ell_2), e, \operatorname{rel}(\ell_2), \operatorname{rel}(\ell_1)$$

Observe that the resulting sequence t' is a valid trace. It is easy to see that (e_1, e_2) can be the only predictable data race of t', and any correct reordering of t that witnesses the data race (e_1, e_2) can be transformed to a correct reordering of t'that witnesses the same data race, and vice versa.

The desired result follows.

B Details of Section 3

Here we prove Lemma 3.1 and Lemma 3.2.

Lemma 3.1. (e_1, e_2) is a predictable data race of t iff there exists a realizable ideal $X \in CIS_t(e_1, e_2)$ such that $e_1, e_2 \notin X$.

Proof. The (\Leftarrow) direction of the statement is straightforward, and here we focus on the (\Rightarrow) direction. Let t^* be a correct reordering that witnesses (e_1, e_2) , and $X^* = \mathcal{E}(t^*)$. We show that there exists an ideal $X \in \text{CIS}_t(e_1, e_2)$ such that (i) $X \subseteq X^*$, and (ii) OpenAcqs(X) \subseteq OpenAcqs(X^*). Observe that the two conditions imply the lemma: (i) since $X \subseteq X^*$, we have that $e_1, e_2 \notin X$, while both events are enabled in X, and (ii) since $X \subseteq X^*$ and OpenAcqs(X) \subseteq OpenAcqs(X^*), we have that $t^* | X$ is a correct reordering of t, and hence X is realizable.

Consider any ideal $Y \in \text{CIS}_t(e_1, e_2)$ such that $Y \subseteq X^*$. Clearly, at least one such Y exists, by taking $Y = \text{Cone}_t(\{e_1, e_2\})$ and noticing that $Y \subseteq X^*$. If $\text{OpenAcqs}(Y) \nsubseteq \text{OpenAcqs}(X^*)$, there exists some lock-acquire event acq $\in \text{OpenAcqs}(Y)$ such that $\text{rel} \in X^*$, where $\text{rel} = \text{match}_t(\text{acq})$. But then $Y' \in \text{CIS}_t(e_1, e_2)$, where $Y' = Y \cup \text{Cone}_t(\{\text{rel}\}) \cup \{\text{rel}\}$. Note that $Y' \subseteq X^*$, and repeat the process for Y = Y'. Since $Y \subset Y'$, this process can be repeated at most n times, thus at some point we have chosen an ideal $Y \in \text{CIS}_t(e_1, e_2)$ with the desired properties of X.

The desired result follows.

Lemma 3.2. We have $|CIS_t(e_1, e_2)| \leq \min(n, \alpha)^{k-2}$, where $\alpha = k \cdot \gamma \cdot \zeta$, and k is the number of threads, γ is the lock-nesting depth, and ζ is the lock-dependence factor of t.

Proof. Let $Z = \text{ClS}_t(e_1, e_2)$ and $X = \text{Cone}_t(\{e_1, e_2\})$. For an ideal $Y \in Z \setminus \{X\}$, let rel_Y be the lock-release event that lead to $Y \in Z$ according to Item 2 of the definition of $\text{ClS}_t(e_1, e_2)$, and $\text{acq}_Y = \text{match}_t(\text{rel}_Y)$. In addition, we call the ideal $Y' \in Z$ with $\text{acq}_Y \in \text{OpenAcqs}(Y')$ that lead to adding $Y \in Z$ the parent of Y. We define inductively $A_X = \emptyset$, and $A_Y = A_{Y'} \cup \{\text{acq}_Y\}$, where Y' is the parent of Y. Note that every ideal $Y \in Z$ is uniquely characterized by A_Y .

Let G_t be the lock-dependence graph of t. We show by induction that for every $Y \in Z \setminus \{X\}$ there exists a lock-acquire event acq \in OpenAcqs(X) such that acq is reachable from acq_Y in G_t . Let Y_1 be the smallest (wrt set inclusion) ancestor of Y such that acq_Y \in OpenAcqs(Y_1). The statement holds if $Y_1 = X$, by taking acq = acq_Y. Otherwise, let Y_2 the parent of Y_1 , and we have acq_{Y1} \in OpenAcqs(Y_2). Note that (i) acq_Y $\not\leq_{\mathsf{TRF}}$ acq_{Y1} (since acq_Y \notin Y_2), (ii) acq_Y $<_{\mathsf{TRF}}$ rel_{Y1} (since rel_Y \notin Y_1). It follows that (acq_Y, acq_{Y1}) is an edge in G_t . By the induction hypothesis, we have that there exists some lock-acquire event acq \in OpenAcqs(X) that is reachable from acq_{Y1} in G_t . Hence acq is reachable by acq_Y in G_t , as desired.

Now, let $A = \bigcup_{Y \in \mathbb{Z}} A_Y$, and by the previous paragraph, for every lock-acquire event $\operatorname{acq}' \in A$, there exists a lockacquire event $\operatorname{acq} \in \operatorname{OpenAcqs}(X)$ that is reachable from acq' in G_t . Hence, $|A| \leq \operatorname{OpenAcqs}(X) \cdot \zeta$. In addition, since there are *k* threads and the lock-nesting depth is γ , we have $|\operatorname{OpenAcqs}(X)| \leq k \cdot \gamma$, thus $|A| \leq k \cdot \gamma \cdot \zeta = \alpha$. Moreover, we trivially have $|A| \leq n$.

Finally, since (i) we have k threads, and (ii) in each ideal $Y \in Z$ the events e_1 and e_2 are enabled, we have $|Z| \le |A|^{k-2} \le \min(n, \alpha)^{k-2}$ such ideals.

The desired result follows.

C Details of Section 4

C.1 Details of Section 4.1

Here we provide formal proofs to Lemma 4.2, Lemma 4.3 and, using these, Lemma 4.1.

Lemma 4.2. We have $X \in I_{\mathcal{P}}$ iff \mathcal{P} is realizable.

Proof. We prove each direction separately.

(⇒). We prove by induction that every canonical trace t_Y of $T_{\mathcal{P}}$ realizes *Y*. The statement clearly holds if $Y = \epsilon$. Otherwise, let $t_Y = t_{Y'} \circ e$, i.e., *Y* is the extension of *Y'* by the event *e*, and by the induction hypothesis we have that $t_{Y'}$ realizes *Y'*. The statement clearly holds if $e \in \mathcal{WL}(t)$, so we focus on the case where $e \in \mathcal{RL}(t)$. Consider the pair (w, e) ∈ Pairs(*X*), and note that w <_{TRF} *e* and thus w ∈ *Y'*. It remains to argue that for every triplet (w, e, w') ∈ Triplets(\mathcal{P}), if w' ∈ *Y'* then w' <_{tv'} w. Assume towards contradiction otherwise,

thus there exist two ancestors $Y_1, Y_2 \in V_{\varphi}^T$ of Y' such that (i) $Y_2 = Y_1 \cup \{w'\}$ and (ii) $w \in Y_1$. In that case we have (w, r) \in Frontier_{\mathcal{P}}(Y₁), hence w' could not have been executable in Y_1 , a contradiction.

(⇐). Let t^* be a witness trace that realizes \mathcal{P} . We argue that for every prefix t' of t^* , there exists a node $X' \in I_{\mathcal{P}}$ such that $X' = \mathcal{E}(t')$. The proof is by induction on the prefixes t'. The statement clearly holds if $t' = \epsilon$, by taking X' to be the root of $T_{\mathcal{P}}$. Otherwise, let t' = t'', e, for some event $e \in X$, and by the induction hypothesis, there exists $X'' \in I_{\mathcal{P}}$ such that X'' = $\mathcal{E}(t'')$. It suffices to argue that e extends X''. The statement is trivial if $e \in \mathcal{RL}(X)$, hence we focus on the case where $e \in W\mathcal{L}(X)$. It suffices to argue that for every pair (w, r) ∈ Frontier $_{\mathcal{P}}(X'')$ we have (w, r, $e) \notin$ Triplets(\mathcal{P}). Consider any such triplet, and since t^* is a witness of the realizability of X, we have $r \in \mathcal{E}(t^*)$. By the induction hypothesis, we have $w \in \mathcal{E}(t'')$ and $r \notin \mathcal{E}(t'')$. But then $\operatorname{RF}_{t^*}(r) \neq w$, a contradiction.

The desired result follows.

Lemma 4.3. The ideal graph $G_{\mathcal{P}}$ has $O(n^k)$ nodes.

Proof. The statement follows directly from the definition of poset ideals and the fact that the poset (X, P) has width at most k.

We are now ready to prove Lemma 4.1.

Lemma 4.1. *Rf*-poset realizability can be solved in $O(k \cdot n^k)$ time for an rf-poset of size n and k threads.

Proof. Consider an rf-poset \mathcal{P} of size n, k threads and d variables. By Lemma 4.2, to decide the realizability of \mathcal{P} it suffices to construct the ideal graph $G_{\mathcal{P}}$ and test whether $X \in V_{\mathcal{P}}$. By Lemma 4.3, $G_{\mathcal{P}}$ has $O(n^k)$ nodes, and since \mathcal{P} has k threads, there each node of G_X has $\leq k$ outgoing edges. Hence G_X can be constructed in $O(k \cdot n^k) = O(\beta)$ time.

The desired result follows.

C.2 Details of Section 4.2

Here we prove formally Lemma 4.4 and, using this, Theorem 2.3.

Lemma 4.4. *Rf*-poset realizability parameterized by the number of threads k is W[1]-hard.

Proof. We show that \mathcal{P}_G is realizable iff *G* has an independent set of size *c*. We prove each direction separately.

(⇒) Let *t* be a trace that realizes \mathcal{P}_G . For each $i \in [c]$, let m_i be the maximum integer *j* such that the event $r(z_i^j)$ has a predecessor e_i in τ_i with $e_i <_t r(x)$. Note that for each such *i* we have $w(s_i) <_t r(x)$, and since $w(s_i)$ is the predecessor of $r(z_i^1)$ in τ_i , the index m_i is well-defined. We argue that $A = \{l_i\}_{i \in [c]}$ is an independent set of *G*, where $l_i = m_i$ if

 $r(x) <_t r(z_i^{m_i})$ and $l_i = m_i + 1$ otherwise. We note that in the second case, $l_i = n$, as, otherwise, we would also have $acq_{l_i}(\ell_i) <_t r(x)$, and since r(x) is protected by lock ℓ_i , we would also have that $rel_{l_i}(\ell_i) <_t r(x)$. But then, $w(y_1^{l_i+1}) <_t r(x)$, which would contradict our choice of m_i .

Indeed, consider any distinct $i_1, i_2 \in [c]$ and any neighbor v_1 and v_2 of m_{i_1} and m_{i_2} , respectively. Note that our choice of $A = \{l_i\}_{i \in [c]}$ implies that both $\operatorname{acq}_{i_1}(\ell_{\{l_{i_1}, v_1\}}) <_t r(x)$ and $\operatorname{acq}_{i_2}(\ell_{\{l_{i_2}, v_2\}}) <_t r(x)$. It suffices to argue that both $r(x) <_t \operatorname{rel}_{i_1}(\ell_{\{l_{i_1}, v_1\}})$ and $r(x) <_t \operatorname{rel}_{i_2}(\ell_{\{l_{i_2}, v_2\}})$, as then we have that $(l_{i_1}, l_{i_2}) \notin E$, which concludes that A is an independent set of G.

Assume towards contradiction that $\operatorname{rel}_{i_1}(\ell_{\{l_{i_1},v_1\}}) <_t r(x)$. Clearly $l_{i_1} < n$, as $r(x) <_P \operatorname{rel}_{i_1}(\ell_{\{n,v_1\}})$. But then, $\operatorname{acq}^{l_{i_1}}(\ell_{i_1}) <_t \operatorname{rel}_{i_1}(\ell_{\{l_{i_1},v_1\}})$ and thus $\operatorname{acq}^{l_{i_1}}(\ell_{i_1}) <_t r(x)$. Since r(x) appears in a critical section on lock ℓ_{i_1} , we have $\operatorname{rel}^{l_{i_1}}(\ell_1) <_t r(x)$. But then $e_{i_1} <_t r(x)$, where e_{i_1} is the predecessor of $r(z_i^{l_1+1})$ in τ_{i_1} , which contradicts our definition of l_1 . Hence, $r(x) <_t \operatorname{rel}_{i_1}(\ell_{\{l_{i_1},v_1\}})$, $\operatorname{rel}_{i_2}(\ell_{\{l_{i_2},v_2\}})$, and thus *A* is an independent set of *G*.

(\Leftarrow) Let *A* be an independent set of *G* of size *c*, and l_1, \ldots, l_c some arbitrary ordering of *A*. For each $i \in [c]$ let Y_i be the following events of threads *i* and c + i.

- 1. All strict predecessors of e_i in τ_i , where $e_i = r(z_i^{l_i})$ if $l_i < n$ and $e_i = r_i(x)$ otherwise.
- 2. If $l_i > 1$, all predecessors of rel^{l_i-1}(ℓ_i) in τ_{c+i} (including rel^{l_i-1}(ℓ_i)).

For each $i \in [c]$, let

 $Z_i = \{e \in Y_i \mid \exists acq \in OpenAcqs(Y_i) \text{ s.t. } acq \leq_P e\},\$

i.e., Z_i contains all events of τ_i and τ_{c+i} that succeed some lock-acquire event that is open in Y_i . We argue that for any two distinct $i_1, i_2 \in [c]$, any two lock-acquire events $\operatorname{acq}_1 \in Z_{i_1}$ and $\operatorname{acq}_2 \in Z_{i_2}$ access a different lock. Note that the statement follows easily if one acq_1 or acq_2 belongs to τ_{c+i} , for some $i \in [c]$, as the locks accessed by each such total ordered are only also accessed by $\tau_{2\cdot c+2}$. Hence, we focus on the case where each of acq_1 and acq_2 belong to τ_i , for some $i \in [c]$. Assume towards contradiction otherwise, hence there exist distinct $i_1, i_2 \in [c]$ such that (i) $\operatorname{acq}_{i_1}(\ell_{\{l_{i_1}, l_{i_2}\}}) \in Z_{i_1}$ and $\operatorname{acq}_i_2(\ell_{\{l_{i_1}, l_{i_2}\}}) \in Z_{i_2}$, while (ii) $\operatorname{rel}_{i_1}(\ell_{\{l_{i_1}, l_{i_2}\}}) \notin Z_{i_1}$ and $\operatorname{rel}_{i_2}(\ell_{\{l_{i_1}, l_{i_2}\}}) \notin Z_{i_2}$. It follows that $(l_{i_1}, l_{i_2}) \in E$, which contradicts the fact that A is an independent set of G.

We now construct a trace *t* that realizes \mathcal{P}_G in five phases, where initially, we have $t = \epsilon$.

- 1. *Phase 1:* For each $i \in [c]$, we linearize the partial order $P|(Y_i \setminus Z_i)$ arbitrarily, and append it to *t*.
- 2. *Phase 2:* For each $i \in [c]$, we linearize the partial order $P|Z_i$ arbitrarily, and append it to *t*.
- 3. *Phase 3:* We append to *t* the sequence $t_1 \circ w(x)$, $r(x) \circ t_2$, where t_1 and t_2 are sequences over the events of $\tau_{2 \cdot c+2}$,

as follows.

$$t_1 = r(s_1), \dots, r(s_c), \operatorname{acq}(\ell_1), \dots \operatorname{acq}(\ell_c) \quad \text{and} \\ t_2 = \operatorname{rel}(\ell_c), \dots, \operatorname{rel}(\ell_1)$$

- 4. *Phase 4:* For each $i \in [c]$, we linearize the partial order $P|(S_i \setminus Y_i)$, where S_i is the smallest ideal of (X, P) that contains the matching lock-release events of all lock-acquire events that are open in Y_i
- 5. *Phase* 5: For each $i \in [c]$ we linearize *P* over the remaining events of τ_i and τ_{c+i} arbitrarily, and append them to *t*.

Finally, we argue that *t* is a valid witness trace. It is straightforward to verify that *t* is a linearization of *P*. Moreover, since every memory location is written exactly once in *X*, for every read event $r \in \mathcal{R}(X)$ we have $RF_t(r) = RF(r)$. It remains to argue that *t* respects the critical sections of \mathcal{P}_G . Observe that the only phase in which we interleave open critical sections between total orders τ_i that access the same lock is in Phase 2. However, as we have shown, for every two distinct $i_1, i_2 \in [c]$, any two lock-acquire events $acq_1 \in Z_{i_1}$ and $acq_2 \in Z_{i_2}$ access a different lock. It follows that *t* respects the critical sections of \mathcal{P}_G .

The desired result follows. \Box

We are now ready to prove Theorem 2.3.

Theorem 2.3. The dynamic data-race prediction problem is W[1]-hard parameterized by the number of threads.

Proof. We show that (w(x), r(x)) is a predictable data race of t iff \mathcal{P}_G is realizable. If \mathcal{P}_G is realizable, then (w(x), r(x)) is a predictable data-race of t witnessed by the witness t^* of the realizability of \mathcal{P}_G , as constructed in the proof of Lemma 4.4 (direction \Leftarrow), restricted to the events { $e \in X : e <_{t^*} w(x)$ }.

For the inverse direction, let t^* be a correct reordering of t that witnesses the data race (w(x), r(x)). We construct a trace t' that realizes \mathcal{P}_G as

$$t' = t^* \circ w(x), r(x), rel(\ell_c), \ldots, rel(\ell_1) \circ t_4 \circ t_5$$

where t_4 and t_5 are analogous to the linearizations in phase 4 and Phase 5, respectively, of the construction in the proof of Lemma 4.4. In particular, let t'' be the prefix of t' until the event rel(ℓ_1).

- 1. We construct t_4 as follows. For each $i \in [c]$, we linearize the set $S_i \setminus \mathcal{E}(t'')$ and append it to t_4 , where S_i is the smallest ideal of X that contains the matching lock-release events of all lock-acquire events that are open in t''. It is easy to see that each S_i contains the event $r_i(x)$, hence t' respects the orderings $r(x) <_P r_i(x)$. Indeed, as $RF(r_i(x)) = w(x)$ and $w(x) \notin t^*$, we have that $r_i(x) \notin t^*$, and hence by definition, $r_i(x) \in S_i$.
- 2. We construct t_5 as follows. For each $i \in [c]$ we linearize the remaining events of τ_i and τ_{c+i} arbitrarily, and append them to t_5 .

The correctness is established similarly to the proof of Lemma 4.4.

The desired result follows. \Box

D Details of Section 5

D.1 Details of Section 5.1

Here we provide details of Section 5.1. We start with the proof of Lemma 5.2. First, consider the construction of witness trace t, by linearizing the poset (X, Q). We have the following lemma, which states that (X, Q) is well-defined.

Lemma D.1. (X, Q) is a poset.

Proof. Assume towards contradiction otherwise. Consider the process in which we insert the orderings in Q in sequence, and examine the first pair of conflicting events (e_1, e_2) such that we try to order $e_1 <_Q e_2$ whereas it already holds that $e_2 <_Q e_1$. For $j \in [2]$, let i_j be such that $e_j \in X_{i_j}$, and it follows that i_1 is the parent of i_2 in T. Observe that, by the construction of Q, we have $e_1 \parallel_P e_2$. Consider the sequence of orderings

$$e_2 = \overline{e}_1 \ll_Q \overline{e}_2 \ll_Q \ldots \ll_Q \overline{e}_\ell = e_1$$

that witnesses (transitively) that $e_2 <_Q e_1$. Since \mathcal{P} is treeinducible, there exists some $j \in [\ell]$ such that $\overline{e}_j \in X_{i_1}$ and $e_2 <_P \overline{e}_j$. Since X_{i_1} is totally ordered in P, the events \overline{e}_j and e_1 are ordered in P. Clearly $e_1 <_P \overline{e}_j$, otherwise we would have $e_2 <_P e_1$. But then we already have $\overline{e}_j <_Q e_1$, and hence a cycle already exists in Q, contradicting our assumption about (e_1, e_2) being the first pair where a cycle is encountered. The desired result follows.

As the above lemma establishes that t is well-defined, we can prove Lemma 5.2 by showing that t is a witness of the realizability of \mathcal{P} .

Lemma 5.2. Every closed, tree-inducible rf-poset is realizable.

Proof. Consider any triplet (w, r, w') ∈ Triplets(\mathcal{P}) and we argue that (i) w <_t r and (ii) if w <_t w' then r <_t w'. For (i), by the definition of rf-posets, we have w <_P r and thus w <_t r. We now turn our attention to (ii). If w' \nexists_P r or w' \nexists_P w, since \mathcal{P} is closed, we have either w' <_P w or r <_P w', and hence w' <_Q w or r <_Q w'. Since t is a linearization of Q, the first case leads to a contradiction, whereas the second case leads to r <_t w', as desired. Now assume that w' \parallel_P r and w' \parallel_P w, and let $i \in [k]$ be such that $r \in X_i$. Since \mathcal{P} is tree-inducible, it follows that w ∈ X_i and w' ∈ X_j for some $i \neq j$, and such that (i, j) is an edge of T. Since w <_t w', we have w <_Q w', and thus, by construction, t *j* is a child of *i* in T. Again, by construction, we have r <_Q w. Since *t* is a linearization of Q, we have that RF_t(r) ≠ w'.

The desired result follows.

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Next, we prove Lemma 5.3, which states that tree-inducibility is preserved under taking closures.

Lemma 5.3. Consider an rf-poset $\mathcal{P} = (X, P, RF)$ and let Q = (X, Q, RF) be the closure of \mathcal{P} . If \mathcal{P} is tree-inducible then Q is also tree-inducible.

Proof. Let $T = ([k], \{(i, j) | X_i \bowtie X_j\})$ be a rooted tree such that \mathcal{P} is tree-inducible to T, and we argue that Q is also tree-inducible to T. Since $Q \sqsubseteq P$, clearly the condition 1 and condition 2 of tree-inducibility are met. Condition 3 follows from the fact that (i) Q is the transitive closure of $P \cup Q'$, where Q' is a relation between conflicting events, and (ii) since T is a tree, for any two conflicting events e_1, e_2 we have either $e_1, e_2 \in X_i$ for some $i \in [k]$, or $e_1 \in X_i$ and $e_2 \in X_j$ for some $i, j \in [k]$ such that (i, j) is an edge of T.

The desired result follows.

Now we can conclude the proof of Lemma 5.1, which is step (i) towards the proof of Theorem 2.4.

Lemma 5.1. *Rf*-poset realizability of tree-inducible rf-posets can be solved in $O(k^2 \cdot d \cdot n^2 \cdot \log n)$ time, for an rf-poset of size *n*, *k* threads and *d* variables.

Proof. Consider a tree-inducible rf-poset \mathcal{P} . In $O(k^2 \cdot d \cdot n^2 \cdot \log n)$ time, we can decide whether the closure of $\mathcal{P} = (X, P, RF)$ exists [31]. By Remark 2, if the closure does not exist, \mathcal{P} is not realizable. On the other hand, if the closure exists, denote it by $\mathcal{Q} = (X, Q, RF)$. By Lemma 5.3, \mathcal{Q} is tree-inducible, and by Lemma 5.2, it is realizable by a witness t^* . Since $Q \sqsubseteq P$, we have that $t^* \sqsubseteq P$, and thus t^* also realizes \mathcal{P} .

The desired result follows.

We now turn our attention to Lemma 5.4 which forms step (ii) towards Theorem 2.4. We start with the following lemma, which, in high level, guarantees that if t^* is a trace in which e is enabled, then $t^*|LCone_t(e)$ also has this property.

Lemma D.2. Let $X = \text{LCone}_t(e)$, and t^* be any correct reordering of t in which e is enabled. The following assertions hold.

- 1. $X \subseteq \mathcal{E}(t^*)$.
- 2. Consider any thread $p_1 \neq p(e)$, and thread p_2 such that (i) $X|p_1 \neq \emptyset$, and (ii) p_2 is a parent of p_1 in the tree topology G_t rooted at p(e). Let e_2 be the unique maximal event of $X|p_2$ in t^* , and e_1 any event in $X|p_1$. We have that $e_1 <_{t^*} e_2$.

Proof. The proof is by induction on the steps of the process that constructs $LCone_t(e)$. Clearly, both statements hold after Item 1 of the process has been executed. Similarly, both statements hold easily after each time Item 2 of the process has been executed. We now proceed with Item 3 of the process. For each $i \in [2]$, consider the lock-acquire events

 $\operatorname{acq}_i \in X$ as identified in this step, and $\operatorname{rel}_i = \operatorname{match}_t(\operatorname{acq}_i)$. Observe that e_2 appears in the critical section of acq_2 , while by the induction hypothesis, we have $\operatorname{acq}_1, \operatorname{acq}_2 \in \mathcal{E}(t^*)$. We distinguish the step that led to $\operatorname{acq}_1 \in X$.

- Step 2. Then $acq_1 <_{TRF} e_2$, and thus (i) all predecessors of rel₁ (including rel₁) appear in t^* , and (ii) rel₁ $<_{t^*} acq_2$, hence rel₁ $<_{t^*} e_2$, and thus $e_1 <_{t^*} e_2$ for all predecessors of rel₁ inserted to *X*.
- Step 3. For each $i \in [2]$, consider the lock-acquire events acq_{i}' as identified in that step, and $\operatorname{rel}_{i}' = \operatorname{match}_{t}(\operatorname{acq}_{i}')$. Note that $\operatorname{acq}_{1} <_{\operatorname{TO}} \operatorname{rel}_{1}'$, and since, by the induction hypothesis, we have $\operatorname{rel}_{1}' <_{t^{*}} e_{2}$, we obtain $\operatorname{acq}_{1} <_{t^{*}} e_{2}$. Thus, again, (i) all predecessors of rel_{1} (including rel_{1}) appear in t^{*} , and (ii) $\operatorname{rel}_{1} <_{t^{*}} \operatorname{acq}_{2}$, hence $\operatorname{rel}_{1} <_{t^{*}} e_{2}$, and thus $e_{1} <_{t^{*}} e_{2}$ for all predecessors of rel_{1} inserted to X.

The desired result follows.

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Next we have a technical lemma, which will allow us to conclude that if (e_1, e_2) is a predictable data race of *t*, the ideal LCone_t $(e_1) \cup$ LCone_t (e_2) is lock-feasible.

Lemma D.3. For any two conflicting events e_1 , e_2 , consider the set $X = \text{LCone}_t(e_1) \cup \text{LCone}_t(e_2)$. For any two conflicting events $e'_1, e'_2 \in X$ such that $p(e'_i) \neq p(e_1)$ for each $i \in [2]$, we have that $e'_1, e'_2 \in \text{LCone}_t(e_i)$, for some $i \in [2]$.

Proof. We assume that $p(e_1) \neq p(e_2)$, as the statement clearly holds otherwise. Assume w.l.o.g. that $e'_i \in \text{LCone}_t(e_i)$ for each $i \in [2]$, and consider the tree $T = ([k], \{(i, j) | \mathcal{E}(t)| p_i \bowtie \mathcal{E}(t)| p_j\})$. Let $i_1, i_2, i'_1, i'_2 \in [k]$ be such that $p(e_1) = p_{i_1}$, $p(e_2) = p_{i_2}, p(e'_1) = p_{i'_1}$ and $p(e'_2) = p_{i'_2}$, and since $e_1 \bowtie e_2$, we have that (i_1, i_2) is an edge of T. Consider the two components C_1, C_2 that are created in T by removing the edge (i_1, i_2) , such that $i_1 \in C_1$ and $i_2 \in C_2$. Since $e'_1 \bowtie e'_2$ and $i'_1, i'_2 \neq i_1$, we have that $i'_1, i'_2 \in C_1$ or $i'_1, i'_2 \in C_2$. We only consider $i'_1, i'_2 \in C_2$, as the other case is similar.

Since $e'_1 \in \text{LCone}_t(e_1)$, there exists event $e \in \text{LCone}_t(e_1)$ such that $p(e) = p_{i_2}$ and either $e'_1 = e$ or $e'_1 \in \text{LCone}_t(e)$. If $e <_{\text{TO}} e_2$, then $\text{LCone}_t(e) \subseteq \text{LCone}_t(e_2)$ and thus $e'_1 \in \text{LCone}_t(e_2)$. Otherwise, $e_2 \in \text{LCone}_t(e_1)$, hence $\text{LCone}_t(e_2) \subseteq \text{LCone}_t(e_1)$ and thus $e'_2 \in \text{LCone}_t(e_1)$.

The desired result follows.

We are now ready to prove Lemma 5.4.

Lemma 5.4. Let $X = \text{LCone}_t(e_1) \cup \text{LCone}_t(e_2)$. We have that (e_1, e_2) is a predictable data race of t iff $(i) \{e_1, e_2\} \cap X = \emptyset$, and (ii) X is a realizable trace ideal of t.

Proof. For the (\Leftarrow) direction, notice that e_1 and e_2 are enabled in *X*, and since *X* is realizable, we have that (e_1, e_2) is a predictable data race of *t*. We now focus on the (\Rightarrow) direction.

Let t^* be a witness of the data race (e_1, e_2) . As e_1 and e_2 are enabled in t^* , we have $\{e_1, e_2\} \cap \mathcal{E}(t^*) = \emptyset$. By Lemma D.2, we have $X \subseteq \mathcal{E}(t^*)$, and thus $\{e_1, e_2\} \cap X = \emptyset$.

We now argue that $t' = t^* | X$ realizes X. Since $X \subseteq$ $\mathcal{E}(t^*)$ and X is a trace ideal, it only remains to show that t' respects the critical sections. Consider any two lockacquire events $acq_1, acq_2 \in X$ such that $acq_1 \bowtie acq_2$ and $\operatorname{acq}_1 <_{t^*} \operatorname{acq}_2$, and we argue that $\operatorname{rel}_1 \in X$. where $\operatorname{rel}_1 = \operatorname{match}_t(\operatorname{acq}_1)$. Let $Y = \mathcal{E}(t) | \{ p(e_1), p(e_2) \}$, and observe that $X|\{p(e_1), p(e_2)\} = Y$, thus the statement is true if $\{p(acq_1), p(acq_2)\} \subseteq \{p(e_1), p(e_2)\}$. Otherwise, we have $p(acq_i) \neq p(e_i)$ for some $i \in [2]$ and each $j \in [2]$. By Lemma D.3, we have that $acq_1, acq_2 \in LCone_t(e_l)$ for some $l \in [2]$. Assume towards contradiction that $rel_1 \notin X$. By the definition of $LCone_t(e_l)$, we have that $p(acq_1)$ is a parent of $p(acq_2)$ in the tree G_t rooted at $p(e_l)$. By Lemma D.2, we have $acq_2 <_{t^*} e$, where *e* is the last event of $X|p(acq_1)$ in t^* . Note that *e* belongs to the critical section of acq_1 , hence we must have $acq_2 <_{t^*} acq_1$, a contradiction.

The desired result follows.

D.2 Details of Section 5.2

Here we provide the proof of Lemma 5.6. We first develop some helpful notation.

Given integers $j, l \in [n/2]$, we denote by X_j^l the events of X_A and X_B that correspond to vectors a_j and b_l , that is, events that have superscript a_j or b_l , with the following exception: if the observation RF(r) of a read event r is not in X_j^l then r is also not in X_j^l . The notation carries over to partial orders Q_j^l and reads-from functions RF $_j^l$. We also use inequalities for the subscripts and superscripts of X (e.g., $X_j^{\leq n/2}$) to denote events that corresponds to vectors $a_{j'}$ and $b_{l'}$ for j' and l'that satisfy the inequalities.

Consider a partial order Q' over a subset $Y \subseteq X$ such that $Q' \subseteq \tau_A | Y$ and $Q' \subseteq \tau_B | Y$. Given two events $e_1, e_2 \in Y$ such that $\{e_1, e_2\} \not\subseteq X_A$ and $\{e_1, e_2\} \not\subseteq X_B$ (i.e., the events belong to different sets among X_A and X_B) we say that a Q' has a *cross edge* $e_1 <_{Q'} e_2$ to mean that it has the ordering $e_1 <_{Q'} e_2$ and the orderings that are introduced transitively through it.

Before the final proof of Lemma 5.6, we present some technical, but conceptually simple, lemmas. Consider the two total orders τ_A and τ_B without any cross edges. The first lemma reasons at the coordinate level of two vectors a_j and b_l . It states that if we start with a cross edge $w_1^{a_j}(x_1) < r_1^{b_l}(x_1)$, the closure rules eventually lead to an ordering $r_1^{a_j}(x_2) < w_1^{b_l}(x_2)$ iff a_j and b_l are not orthogonal.

Lemma D.4. For any $j, l \in [n/2]$, consider the rf-poset $Q_j^l = (X_j^l, Q_j^l, RF_j^l)$, where Q_j^l has a single cross edge $w_1^{a_j}(x_1) <_{Q_j^i}$

 $r_1^{b_l}(x_1)$. Let $S_j^l = (X_j^l, S_j^l, \mathrm{RF}_j^l)$ be the closure of Q_j^l . We have $r_1^{a_j}(x_2) <_{S_j^l} w_1^{b_l}(x_2)$ iff a_j and b_l are not orthogonal.

Proof (sketch). First, observe that since $w_1^{a_j}(x_1) <_{Q_j^l} r_1^{b_l}(x_1)$, by closure, we will have $w_1^{a_j}(x_1) <_{S_j^l} w_1^{b_l}(x_1)$. By a simple induction on the events on variables x_1 and x_3 , we have that $w_i^{a_j}(x_1) <_{S_j^l} w_i^{b_l}(x_1)$ for each $i \in [D]$. In turn, one of these orderings leads to $w_i^{a_j}(x_2) <_{S_j^l} w_i^{b_l}(x_2)$ and thus $r_i^{a_j}(x_2) <_{S_j^l}$ $w_i^{b_l}(x_2)$ iff $a_j[i] \cdot b_l[i] = 1$. By a simple induction on the events on variables x_2 and x_6 , we have that $r_1^{a_j}(x_2) <_{S_j^l} w_1^{b_l}(x_2)$ iff a_j and b_l are not orthogonal.

The next lemma reasons at the vector *B* level. Consider any fixed $j \in [n/2]$. The lemma states that if we order $w_1^{a_1}(x_1) < r_1^{b_1}(x_1)$, if a_j is not orthogonal to $b_{l'}$ for any $l' \leq n/2 - 1$, the closure rules eventually lead to $w_1^{a_j}(x_1) < r_1^{b_{n/2}}(x_1)$. On the other hand, if there exists a smallest l_1 such that a_j is orthogonal to b_{l_1} , the closure rules will stop inserting cross edges between events of vectors a_j and all $b_{l'}$, for $l' \geq l_1 + 1$. **Lemma D.5.** For any $j \in [n/2]$ and $l \leq n/2 - 1$, consider the *rf*-poset $Q_j^{\leq l+1} = (X_j^{\leq l+1}, Q_j^{\leq l+1}, RF_j^{\leq l+1})$, where $Q_j^{\leq l+1}$ has a single cross edge $w_1^{a_j}(x_1) <_{Q_j^{\leq l+1}} r_1^{b_1}(x_1)$. Let $S_j^{\leq l+1} = (X_j^{\leq l+1}, S_j^{\leq l+1}, RF_j^{\leq l+1})$ be the closure of $Q_j^{\leq l+1}$. The following assertions hold.

- 1. If a_j is not orthogonal to $b_{l'}$ for any $l' \in [l]$, then $w_1^{a_j}(x_1) <_{S_{i}^{\leq l+1}} r_1^{b_{l+1}}(x_1)$.
- 2. If a_j is orthogonal to $b_{l'}$, for some $l' \in [l]$, then there are no cross edges in $(X_j^{l_1+1 \leq l' \leq l+1}, S_j^{l_1+1 \leq l' \leq l+1})$, where l_1 is the smallest l' such that a_j is orthogonal to $b_{l'}$.

Proof (sketch). The proof is by induction on *l*. For the base case (l = 1), consider the rf-poset Q_j^1 , and by Lemma D.4, we have $r_1^{a_j}(x_2) <_{S_j^1} w_1^{b_1}(x_2)$ iff a_j and b_1 are not orthogonal. Observe that if $r_1^{a_j}(x_2) <_{S_j^1} w_1^{b_1}(x_2)$, we also have $r_1^{a_j}(x_2) <_{S_j^{\leq 2}} w_1^{b_1}(x_2)$, and thus $w^{a_j}(x_4) <_{S_j^{\leq 2}} r^{b_1}(x_4)$. Then, by closure we have $w^{a_j}(x_4) <_{S_j^{\leq 2}} w^{b_2}(x_4)$ and thus transitively $w_2^{a_j}(x_1) <_{S_j^{\leq 2}} r_1^{b_2}(x_1)$. On the other hand, it can be easily seen that if $r_1^{a_j}(x_2) \not<_{S_j^1} w_1^{b_1}(x_2)$ then the cross edges in $(X_j^{\leq 2}, S_j^{\leq 2})$ are precisely the cross edges in $(X_j^{\leq 1}, S_j^{\leq 1})$, and thus there are no cross edges in (X_j^2, S_j^2) .

Now assume that the claim holds for l, and we argue that it holds for l+1. By the induction hypothesis, we have that if a_j is not orthogonal to $b_{l'}$ for any $l' \in [l]$, then $w_1^{a_j}(x_1) <_{S_j^{\leq l+1}} r_1^{b_{l+1}}(x_1)$ and thus $w_1^{a_j}(x_1) <_{S_j^{\leq l+2}} r_1^{b_{l+1}}(x_1)$. A similar analysis to the base case shows that if a_j is not orthogonal to b_{l+1}

then $w_1^{a_j}(x_1) <_{S_j^{\leq l+2}} r_1^{b_{l+2}}(x_1)$. On the other hand, if a_j is orthogonal to b_{l+1} , it can be easily seen that the cross edges in $(X_j^{\leq l+2}, S_j^{\leq l+2})$ are precisely the cross edges in $(X_j^{\leq l+1}, S_j^{\leq l+1})$, and thus there are no cross edges in $(X_j^{l_{1+1}}, S_j^{l_{1+1}})$, where $l_1 = l + 1$. Finally, if a_j is orthogonal to $b_{l'}$ for some $l' \in [l]$, the statement follows easily by the induction hypothesis. The desired result follows.

The next lemma reasons at the vector A level. The lemma states that if we order $w_1^{a_1}(x_1) < r_1^{b_1}(x_1)$, if $a_{j'}$ is not orthogonal to $b_{l'}$ for any $j' \leq n/2 - 1$ and $l' \leq n/2$, the closure rules eventually lead to $w_1^{a_{n/2}}(x_1) < r_1^{b_{n/2}}(x_1)$. On the other hand, if there exists a smallest j_1 such that a_{j_1} is orthogonal to $b_{l'}$, for some $l' \in [n/2]$, the closure rules will stop inserting cross edges between events of all vectors $a_{j'}$ and all vectors b_l , for $j' \geq j_1 + 1$ and $l \in [n/2]$.

Lemma D.6. For any $j \in [n/2 - 1]$, consider the rfposet $Q_{\leq j+1}^{\leq n/2} = (X_{\leq j+1}^{\leq n/2}, Q_{\leq j+1}^{\leq n/2}, \operatorname{RF}_{\leq j+1}^{\leq n/2})$, where $Q_{\leq j+1}^{\leq n/2}$ has a single cross edge $w_1^{a_1}(x_1) <_{Q_{\leq j+1}^{\leq n/2}} r_1^{b_1}(x_1)$. Let $S_{\leq j+1}^{\leq n/2} = (X_{\leq j+1}^{\leq n/2}, S_{\leq j+1}^{\leq n/2}, \operatorname{RF}_{\leq j+1}^{\leq n/2})$ be the closure of $Q_{\leq j+1}^{\leq n/2}$. The following assertions hold.

- 1. If $a_{j'}$ is not orthogonal to $b_{l'}$ for any $j' \in [j]$ and $l' \in [n/2]$, then $w_1^{a_{j+1}}(x_1) <_{S_{\leq j+1}^{\leq n/2}} w_1^{b_1}(x_1)$.
- 2. If $a_{j'}$ is orthogonal to $b_{l'}$ for some $j' \in [j]$ and $l' \in [n/2]$, then there are no cross edges in $(X_{j_1+1 \leq j' \leq j+1}^{\leq n/2}, S_{j_1+1 \leq j' \leq j+1}^{\leq n/2})$, where j_1 is the smallest j' such that $a_{j'}$ is orthogonal to $b_{l'}$, for some $l' \in [n/2]$.

Proof (sketch). The proof is by induction on *j*. For the base case (*j* = 1), consider the rf-poset $Q_1^{\leq n/2} = (X_1^{\leq n/2}, Q_1^{\leq n/2}, \operatorname{RF}_1^{\leq n/2})$, and by Lemma D.5, we have that if *a*₁ is not orthogonal to *b*_{*l'*} for any *l'* ∈ [*n*/2 − 1], then $w_1^{a_j}(x_1) <_{S_j^{\leq n/2}} r_1^{b_{n/2}}(x_1)$. Consider the rf-poset $Q_1^{n/2}$, and by Lemma D.4, we have $r_1^{a_1}(x_2) <_{S_1^{n/2}} w_1^{b_{n/2}}(x_2)$ iff *a*₁ and *b*_{*n*/2} are not orthogonal. Observe that if $r_1^{a_1}(x_2) <_{S_1^{n/2}} w_1^{b_{n/2}}(x_2)$, we also have $r_1^{a_1}(x_2) <_{S_1^{\leq n/2}} w_1^{b_{n/2}}(x_2)$, and thus $r_1^{a_1}(x_2) <_{S_{\leq 2}^{\leq n/2}} w_1^{b_{n/2}}(x_2)$, hence, transitively, $w_1^{a_1}(x_5) <_{S_{\leq 2}^{\leq n/2}} w_{n/2}^{b_{n/2}}(x_5)$. Hence, by closure, we have $r^{a_1}(x_5) <_{S_{\leq 2}^{\leq n/2}} w_{n/2}^{b_{n/2}}(x_5)$. Then, by closure we have $r^{a_1}(x_5) <_{S_{\leq 2}^{\leq n/2}} w_1^{b_{n/2}}(x_5)$. Then, by closure we have $r^{a_1}(x_5) <_{S_{\leq 2}^{\leq n/2}} w_1^{a_1}(x_1) <_{S_{\leq 2}^{\leq n/2}} w_{n/2}^{b_{n/2}}(x_7)$ and thus transitively $w_1^{a_2}(x_1) <_{S_{\leq 2}^{\leq n/2}} r_1^{b_1}(x_1)$. On the other hand, it can be easily seen that if $r_1^{a_1}(x_2) <_{S_{1}^{a_{1/2}}} r_1^{b_{n/2}}(x_2)$ then the cross edges in $(X_{\leq 2}^{\leq n/2}, S_{\leq 2}^{\leq n/2})$, and thus there are no cross edges in $(X_2^{\leq n/2}, S_2^{\leq n/2})$.

Now assume that the claim holds for j, and we argue that it holds for j + 1. By the induction hypothesis, we have that if $a_{j'}$ is not orthogonal to $b_{l'}$ for any $j' \in [j]$ and $l' \in [n/2]$, then $r_1^{a_{j+1}}(x_2) <_{S_{\leq j+1}^{\leq n/2}} w_1^{b_1}(x_2)$ and thus $r_1^{a_{j+1}}(x_2) <_{S_{\leq j+2}^{\leq n/2}} w_1^{b_1}(x_2)$. A similar analysis to the base case shows that if a_{j+1} is not orthogonal to $b_{l'}$ for any $l' \in [n/2]$, then $r_1^{a_{j+2}}(x_2) <_{S_{\leq j+2}^{\leq n/2}} w_1^{b_1}(x_2)$. A similar analysis to the base case shows that if a_{j+1} is not orthogonal to $b_{l'}$ for any $l' \in [n/2]$, then $r_1^{a_{j+2}}(x_2) <_{S_{\leq j+2}^{\leq n/2}} w_1^{b_1}(x_2)$. On the other hand, if a_{j+1} is orthogonal to some $b_{l'}$, it can be easily seen that the cross edges in $(X_{\leq j+2}^{\leq n/2}, S_{\leq j+2}^{n/2})$ are precisely the cross edges in $(X_{\leq j+1}^{\leq n/2}, S_{\leq j+1}^{n/2})$. Thus, there are no cross edges in $(X_{j+1}^{\leq n/2}, S_{j+1}^{s/2})$, where $j_1 = j' + 1$. Finally, if $a_{j'}$ is orthogonal to $b_{l'}$ for some $j' \in [j]$ and $l' \in [n/2]$, the statement follows easily.

The desired result follows.

We next have three lemmas that each is symmetric to Lemma D.4, Item 2 of Lemma D.5, and Item 2 of Lemma D.6, respectively. The proof of each lemma is analogous to its symmetric lemma, and is omitted here.

Lemma D.7. For any $j, l \in [n/2]$, consider the rf-poset $\overline{Q}_j^l = (X_j^l, \overline{Q}_j^l, \operatorname{RF}_j^l)$, where \overline{Q}_j^l has a single cross edge $w_1^{b_l}(x_2) <_{\overline{Q}_i^j} r_1^{a_j}(x_2)$. Let $\overline{S}_j^l = (X_j^l, \overline{S}_j^l, \operatorname{RF}_j^l)$ be the closure of \overline{Q}_j^l . We have $r_1^{b_l}(x_1) <_{\overline{S}_j^l} w_1^{a_j}(x_1)$ iff a_j and b_l are not orthogonal.

Lemma D.8. For any $j \in [n/2]$ and $l \ge 2$, consider the rf-poset $\overline{Q}_j^{\ge l-1} = (X_j^{\ge l-1}, \overline{Q}_j^{\ge l-1}, \operatorname{RF}_j^{\ge l-1})$, where $\overline{Q}_j^{\ge l-1}$ has a single cross edge $w_l^{b_{n/2}}(x_2) <_{\overline{Q}_j^{\ge l-1}} r_1^{a_j}(x_2)$. Let $\overline{S}_j^{\ge l-1} = (X_j^{\ge l-1}, \overline{S}_j^{\ge l-1}, \operatorname{RF}_j^{\ge l-1})$ be the closure of $\overline{Q}_j^{\ge l-1}$. If a_j is orthogonal to $b_{l'}$, for some $l' \in [l]$, then there are no cross edges in $(X_j^{l_2-1\ge l'\ge l-1}, S_j^{l_2-1\ge l'\ge l-1})$, where l_2 is the largest l' such that a_j is orthogonal to b_{l_2} .

Lemma D.9. For any $j \geq 2$ with j > 1, consider the rfposet $\overline{Q}_{\geq j-1}^{\geq 1} = (X_{\geq j-1}^{\geq 1}, \overline{Q}_{\geq j-1}^{\geq 1}, \operatorname{RF}_{\geq j-1}^{\geq 1})$, where $\overline{Q}_{\geq j-1}^{\geq 1}$ has a single cross edge $w_1^{b_{n/2}}(x_2) <_{\overline{Q}_{\geq j-1}^{\geq 1}} r_1^{a_{n/2}}(x_2)$. Let $\overline{S}_{\geq j-1}^{\geq 1} =$ $(X_{\geq j-1}^{\geq 1}, \overline{S}_{\geq j-1}^{\geq 1}, \operatorname{RF}_{\geq j-1}^{\geq 1})$ be the closure of $\overline{Q}_{\geq j-1}^{\geq 1}$. If $a_{j'}$ is orthogonal to $b_{l'}$ for some $j' \in [j]$ and $l' \in [n/2]$, then there are no cross edges in $(X_{j_2-1\geq j'\geq j-1}^{\geq 1}, \overline{S}_{j_2-1\geq j'\geq j-1}^{\geq 1})$, where j_2 is the largest j' such that $a_{j'}$ is orthogonal to $b_{l'}$, for some $l' \in [n/2]$.

Using Lemma D.4, Lemma D.5, Lemma D.6 we can now prove Lemma 5.6.

Lemma 5.6. *Rf*-poset realizability for an *rf*-poset with 2 threads and 7 variables has no $O(n^{2-\epsilon})$ -time algorithm for any $\epsilon > 0$, under the Orthogonal Vectors conjecture.

Proof. We show that \mathcal{P} is realizable iff there exist $j, l \in [n/2]$ such that a_j is orthogonal to b_l . We prove each direction separately.

(⇒). Since \mathcal{P} is realizable, by Remark 2, we have that the closure of \mathcal{P} exists. Let $\mathcal{K} = (X, K, RF)$ be the closure of \mathcal{P} . Let Q = (X, Q, RF) where Q has a single cross edge $w_1^{a_1}(x_1) <_Q r_1^{b_1}(x_1)$, and S = (X, S, RF) be the closure of Q. Note that $P \sqsubseteq Q$ and thus $K \sqsubseteq S$, and we have that $w_1^{a_{n/2}}(x_2) \not\leq_S r_1^{b_{n/2}}(x_2)$. If $w_1^{a_{n/2}}(x_1) <_S w_1^{b_{n/2}}(x_1)$, by Lemma D.4, we have that $a_{n/2}$ is orthogonal to $b_{n/2}$. Otherwise, if $w_1^{a_{n/2}}(x_1) <_S r_1^{b_1}(x_1)$, by Item 1 of Lemma D.5, we have that $a_{n/2}$ is orthogonal to b_l , for some $l \in [n/2]$. Finally, if $w_1^{a_{n/2}}(x_1) \not\leq_S r_1^{b_1}(x_1)$, by Item 1 of Lemma D.6, we have that a_j is orthogonal to b_l , for some $j \leq n/2 - 1$ and $l \in [n/2]$. (⇐). Let Q = (X, Q, RF) where Q has a single cross edge $w_1^{a_1}(x_1) <_{Q \leq j+1}^{\leq n/2} r_1^{b_1}(x_1)$, and $\overline{Q} = (X, \overline{Q}, RF)$ where \overline{Q} has a single cross edge $w_1^{b_{n/2}}(x_2) <_{\overline{Q}} r_1^{a_{n/2}}(x_2)$. Let S = (X, S, RF) and $\overline{S} = (X, \overline{S}, RF)$ be the closures of Q and \overline{Q} , respectively. We argue that $(X, S \cup \overline{S})$ is a poset. Note that this implies the lemma, as $(X, S \cup \overline{S}, RF)$ is the closure of \mathcal{P} .

Let (j_1, l_1) (resp., (j_2, l_2)) be the lexicographically smallest (resp., largest) pair of integers in $[n/2] \times [n/2]$ such that a_{j_1} is orthogonal to b_{l_1} (resp., a_{j_2} is orthogonal to b_{l_2}). We assume wlog that $j_1 = j_2 = j$ and $l_1 = l_2 = l$, as in any other case, the relation $S \cup \overline{S}$ is strictly smaller. First, assume that $(X_j^l, S_j^l \cup \overline{S}_j^l)$ is a poset. By Item 2 of Lemma D.5, we have that there are no cross edges in $S_l^{l+1 \le l' \le n/2}$, while by Lemma D.8, we have that there are no cross edges in $\overline{S}_j^{l-1 \ge l' \ge 1}$. Similarly, by Item 2 of Lemma D.6, we have that there are no cross edges in $S_{j+1 \le j'}^{\le n/2}$, while by Lemma D.9, we have that there are no cross edges in $\overline{S}_{j-1 \ge j'}^{\ge 1}$. It follows that $(X, S \cup \overline{S})$ is a poset.

It remains to argue that $(X_j^l, S_j^l \cup \overline{S}_j^l)$ is a poset. The relation $S_j^l \cup \overline{S}_j^l$ is the transitive closure of a relation that consists of the total orders $t_A | X_j^l$ and $t_B | X_j^l$, together with the following relations, for each $i \in [D]$.

$$w_{i}^{a_{j}}(x_{1}) <_{S_{j}^{l}} w_{i}^{b_{l}}(x_{1}) \text{ and } w_{i}^{a_{j}}(x_{3}) <_{S_{j}^{l}} w_{i}^{b_{l}}(x_{3}) \text{ and} w_{i}^{b_{l}}(x_{2}) <_{\overline{S}_{j}^{l}} w_{i}^{a_{j}}(x_{2}) \text{ and } w_{i}^{b_{l}}(x_{6}) <_{\overline{S}_{j}^{l}} w_{i}^{a_{j}}(x_{6})$$

It is straightforward to verify that if $S_j^l \cup \overline{S}_j^l$ has a cycle then, for some $i \in [D]$, we have

$$w_i^{a_j}(x_2) <_{\tau_A} w_i^{a_j}(x_1)$$
 and $w_i^{b_l}(x_1) <_{\tau_B} w_i^{b_l}(x_2)$.

Thus by construction, $a_j[i] = 1$ and $b_l[i] = 1$, which contradicts the fact that a_j is orthogonal to b_l .

The desired result follows.

Finally, we are ready to prove Theorem 2.5.

Theorem 2.5. Let t be a trace with n events, $k \ge 2$ threads and $d \ge 9$ shared global variables with at least one lock. There

is no algorithm that solves the decision problem of dynamic data-race prediction for t in time $O(n^{2-\epsilon})$, for any $\epsilon > 0$, unless the Orthogonal Vectors conjecture fails.

Proof. We complete the proof of Theorem 2.5 by showing that, in the above construction, \mathcal{P} has a closure iff (w(z), r(z)) is a data race of t, witnessed by a correct reordering t^* . Observe that Q has 2 threads, and is thus tree-inducible. Due to Lemma 5.2, it suffices to show that \mathcal{P} is realizable iff (w(z), r(z)) is a data race of t.

(⇒). Assume that \mathcal{P} is realizable and t' is a witness trace. We construct the correct reordering t^* as follows. We have $\mathcal{E}(t^*) = \mathcal{E}(t) \setminus \{\operatorname{rel}_2(\ell), w(z), r(z)\}$. We make t^* identical to t', i.e., $t^*|X = t'$. For the events $\mathcal{E}(t^*) \setminus X$, we make (i) w(y) appear right after $w_1^{b_{n/2}}(x_2)$, (ii) r(y) appear right before $r_1^{a_{n/2}}(x_2)$, (iii) $\operatorname{acq}_A(\ell)$, $\operatorname{rel}_A(\ell)$ appear right after $w^{a_1}(x_1)$, and (iv) $\operatorname{acq}_B(\ell)$ appear last in t^* . Observe that t^* is a correct reordering in which w(z) and r(z) are enabled, hence t^* is a witness of the data race (w(z), r(z)).

(⇐). Assume that (w(z), r(z)) is a predictable data race of t, and let t^* be a correct reordering of t witnessing the data race. We construct a trace t' as $t' = t^*|X$, and argue that t' realizes \mathcal{P} . Clearly RF_{t'} = RF. To see that t' is a linearization of (X, P), it suffices to argue that

$$w_1^{b_{n/2}}(x_2) <_{t'} r_1^{a_{n/2}}(x_2)$$
 and $w_1^{a_1}(x_1) <_{t'} r_1^{b_1}(x_1)$

The first ordering follows from the fact that

$$w_1^{b_{n/2}}(x_2) <_{t^*} w(y) <_{t^*} r_y <_{t^*} r_1^{a_{n/2}}(x_2)$$
.

For the second ordering, observe that $\mathcal{E}(t^*) = \mathcal{E}(t) \setminus \{\operatorname{rel}_2(\ell), w(z), r(z)\}$. Hence,

$$w^{a_1}(x_1) <_{t^*} \operatorname{rel}_1(\ell) <_{t^*} \operatorname{acq}_2(\ell) <_{t^*} r_a^{b_1}(x_1)$$
.

The desired result follows.

E Details of Section 6

In this section we present the full proof of Lemma 6.1.

Lemma 6.1. Consider a natural number ℓ , a trace t over n events and k threads, and a feasible trace ideal X of t. The ℓ -distance-bounded realizability problem for X can be solved in $O(k^{\ell+O(1)} \cdot n)$ time.

Proof. Let $\mathcal{P} = (X, P, RF)$ be the canonical rf-poset of X, and the task is to decide the realizability of \mathcal{P} with ℓ reversals. We describe a recursive algorithm for solving the problem for some rf-poset Q = (X, Q, RF) with ℓ' reversals, for some $\ell' \leq \ell$, where initially Q = P and $\ell' = \ell$. We first give a high-level description and argue about its correctness. Afterwards, we describe some low-level details that allow us to reason about the complexity.

Algorithm and correctness. Consider the set

$$C = \{ (\mathbf{w}_1, \mathbf{w}_2) \in \mathcal{WL}(X) \times \mathcal{WL}(X) \colon \mathbf{w}_1 \bowtie \mathbf{w}_2 \text{ and} \\ \mathbf{w}_1 \parallel_O \mathbf{w}_2 \text{ and } \mathbf{w}_1 <_t \mathbf{w}_2 \}.$$

We construct a graph $G_1 = (X, E_1)$, where $E_1 = (\mathsf{TRF}|X) \cup C$. Note that G_1 is write-ordered. If it is acyclic, we construct the read extension G_2 of G_1 . Observe that if G_2 is acyclic then any linearization t^* of *G* realizes *Q*, hence we are done. Now consider that either G_1 or G_2 is not acyclic, and let $G = G_1$ if G_1 is not acyclic, otherwise $G = G_2$. Given a cycle \mathcal{C} of *G*, represented as a collection of edges, define the set of cross-edges of \mathcal{C} as $\mathcal{C} \setminus Q$. Observe that since there are k threads, G has a cycle with $\leq k$ cross edges. Indeed, if there are more than k cross edges in \mathcal{C} , then there are two cross edges that go out of two nodes that belong to the same thread and hence are connected by a path in G. Then, we can simply remove one of these edges and G will still have a cycle. Hence, we can repeat this process until we end up with a cycle that has $\leq k$ cross edges. In addition, any trace t^* that realizes Q must linearize an rf-poset (X, Q_a, RF) where $a = (e_1, e_2)$ ranges over the cross-edges of \mathcal{C} . In particular, we take $Q_a = Q \cup \{b\}$, where

$$b = \begin{cases} (e_2, e_1), & \text{if } a \in \mathcal{WL}(X) \times \mathcal{WL}(X) \\ (\text{RF}(e_2), e_1), & \text{if } a \in \mathcal{WL}(X) \times \mathcal{RL}(X) \\ (e_2, \text{RF}(e_1)), & \text{if } a \in \mathcal{RL}(X) \times \mathcal{WL}(X) . \end{cases}$$

Observe that any such choice of *b* reverses the order of two conflicting write events or lock-acquire events in *t*. Since there are $\leq k$ cross edges in \mathcal{C} , there are $\leq k$ such choices for Q_a . Repeating the same process recursively for the rf-poset (X, Q_a, RF) for $\ell' - 1$ levels solves the ℓ' -distance-bounded realizability problem for Q. Since initially $\ell' = \ell$ and Q = P, this process solves the same problem for \mathcal{P} and thus for X.

Complexity. The recursion tree above has branching $\leq k$ and depth $\leq \ell$, hence there will be at most k^{ℓ} recursive instances. We now provide some lower-level algorithmic details which show that each instance can be solved in $O(k^{O(1)} \cdot n)$ time. The main idea is that each of the graphs G_1 and G_2 have a sparse transitive reduction [3] of size $O(k \cdot n)$, and thus we can represent G_1 and G_2 with few edges.

For the graph G_1 , we construct a sparse graph G'_1 that preserves the reachability relationships of G_1 as follows. We traverse the trace t top-down. For the current event w such that $w \in \mathcal{WL}(X)$, for every $i \in [k]$, let w_1 be the last event of thread i which precedes w in t, and such that $w_1 \in \mathcal{WL}(X)$ and $w \bowtie w_1$. Let w'_1 be the last event of t such that $w'_1 \leq_{TO} w_1$ and $w \not\prec_Q w'_1$ (note that possibly $w'_1 = w_1$). If such w'_1 exists, since X is an ideal, we have $w'_1 \in X$. We introduce the edge $w'_1 \rightarrow w$ in G'_1 .

Similarly, for the graph G_2 , we construct a sparse graph G'_2 that preserves the reachability relationships of G_2 as follows. We iterate over all read and lock-release events $r \in X$. For each such r and thread $i \in [k]$, we insert two edges (w_2, r) and (r, w_3) in G'_2 , where w_2 (resp., w_3) is the latest predecessor (resp., earliest successor) of RF(r) in thread *i* that conflicts with r.

We now argue that all events w_1 , w'_1 , w_2 and w_3 above can be identified in O(n) total time. For every $i \in [k]$ and memory location x, we construct a total order τ_i^x of all write events or lock-acquire events on location x of thread p_i . Clearly all such total orders can be constructed in O(n) time.

- Events w₁: as we traverse *t* top-down, we simply remember for each memory location *x*, the last write event or lock-acquire event on location *x* for each thread. Hence, the total time for identifying all such events is *O*(*n*).
- 2. Events w'_1 : Given the event w_1 on memory location x, we simply traverse the total order τ_i^x from w'_1 backwards until we find either (i) an event w'_1 with the desired properties, or (ii) an event w''_1 that has been examined before when inserting edges to thread p_i . In the first case, we simply add the edge $(w'_1 \rightarrow w)$ as described above. In the second case we do nothing, as the desired ordering is already present due to transitivity through a write event w' of thread p_i that has been examined before (thus $w' <_{TO} w$). Hence, every event in every total order τ_i^x is examined O(1) times, thus the total time for identifying all w'_1 events is O(n).
- 3. Events w_2 and w_3 : after we have constructed G'_1 , each pair of such events can be retrieved from G'_1 in O(1) time. Hence the total time for identifying all such events is O(n).

Finally, the above process creates graphs G'_1 and G'_2 that have $O(k^{O(1)} \cdot n)$ edges. Detecting a cycle \mathcal{C}' in either G'_1 or G'_2 is done by a simple DFS, which takes linear time in the number of edges. Converting \mathcal{C}' to a cycle \mathcal{C} such that \mathcal{C} has at most k cross edges can be done in $O(|\mathcal{C}'|) = O(k^{O(1)} \cdot n)$ time, by removing multiple edges whose endpoints are TO-ordered. The desired result follows.