Synthesizing Racy Tests



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Abstract

Subtle concurrency errors in multithreaded libraries that arise because of incorrect or inadequate synchronization are often difficult to pinpoint precisely using only static techniques. On the other hand, the effectiveness of dynamic race detectors is critically dependent on multithreaded test suites whose execution can be used to identify and trigger races. Usually, such multithreaded tests need to invoke a specific combination of methods with objects involved in the invocations being shared appropriately to expose a race. Without *a priori* knowledge of the race, construction of such tests can be challenging.

In this paper, we present a lightweight and scalable technique for synthesizing precisely these kinds of tests. Given a multithreaded library and a sequential test suite, we describe a fully automated analysis that examines sequential execution traces, and produces as its output a concurrent client program that drives shared objects via library method calls to states conducive for triggering a race. Experimental results on a variety of well-tested Java libraries yield 101 synthesized multithreaded tests in less than four minutes. Analyzing the execution of these tests using an off-the-shelf race detector reveals 187 harmful races, including several previously unreported ones. Our implementation, named NARADA, and the results of our experiments are available at

http://www.csa.iisc.ernet.in/~sss/tools/narada.

Categories and Subject Descriptors D.2.5 [*Software Engineering*]: Testing and Debugging—testing tools; F.3.2 [*Logics and Meanings of Programs*]: Semantics of Programming Languages program analysis

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1. Introduction

Ensuring libraries are *thread-safe* [1, 20] is highly desirable because it would alleviate the burden on applications to deal with the complexities of multithreading. However, building thread-safe libraries is a challenging task [10]. Since libraries are intended to be used by multiple clients concurrently, they must be structured to prevent unwanted racy access to shared state. But, performance considerations dictate minimizing unnecessary synchro-

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Copyright © 2015 ACM 978-1-4503-3468-6/15/06...\$15.00. http://dx.doi.org/10.1145/2737924.2737998 nization. Thus, if a library is shown not to be thread-safe, applications may be able to take preventive action to avoid the conditions that trigger a thread-safety violation. For example, if it is known that invoking two methods in a library, with certain kinds of parameters, from distinct threads can result in a data race, applications can be (re)written to acquire suitable locks before invoking these methods, without requiring re-implementation of the library.

There have been many program analysis techniques that have been developed to detect races in shared-memory multithreaded programs. These efforts can be broadly classified into three main categories: (a) static analysis techniques (e.g., [15, 21]), (b) systematic testing (e.g., [13, 14]), and (c) dynamic analysis approaches (e.g., [2, 7, 18, 19, 24–26]). While static analysis techniques are execution agnostic and can be comprehensive, results can oftentimes be imprecise [12], with non-trivial false positive rates. On the other hand, systematic testing and dynamic analysis techniques are *critically* dependent on the availability of *effective* multithreaded tests to detect races precisely.

Designing useful multithreaded tests is difficult because if a race is not known to exist *a priori*, the test essentially devolves into a *blind* search for a *potential* race in the library. For a race to manifest, appropriate library methods need to be invoked from at least two different client threads. Moreover, it is also essential that the objects on which the methods operate need to be in a state conducive to trigger the race, and must be shared among these threads appropriately. The requirement to both carefully understand object sharing properties in the library and drive execution to a proper global state, makes the design of effective multithreaded tests for race detection especially challenging.

```
class Lib {
   Counter c;
   public void synchronized update() { c.inc(); }
   public void synchronized set(Counter x) {
      c = x;
   }
}
class Counter {
   int count;
   public void inc() { count++; }
}
```

Figure 1: Illustrative Example.

Figure 1 presents the implementation of two classes Lib and Counter. Because the method update is synchronized, one may mistakenly assume that invoking it from multiple threads without holding any additional lock will not lead to a race. Upon careful inspection, it is clear that this assumption is not true. For example, consider a scenario in which:

1. objects *p* and *q* of type Lib, and object *r* of type Counter are created,

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- 2. p.set(r) and q.set(r) are invoked, and
- 3. *p.update()* and *q.update()* are invoked from two threads concurrently.

If a multithreaded test performing the aforementioned operations exists, then one of the possible interleavings in the many executions of the test can expose the race on count. Therefore, race detection [7, 13] by analyzing the execution of the multithreaded test is also feasible. However, designing such a test requires creating objects of appropriate types, driving each thread to a suitable state so that concurrent execution of the update method can expose the underlying race. The goal of this paper is to *automatically* synthesize such multithreaded tests.

We propose a novel and scalable technique for automatically synthesizing racy tests to enable race detection in multithreaded libraries. The input to our technique is the implementation of the library and a sequential seed test-suite. The output is a multithreaded test-suite, where each test spawns multiple threads and invokes various methods in the library appropriately. The key insight to our approach is that the properties observed during sequential execution can be leveraged to generate constraints such that a multithreaded execution satisfying the constraints will result in a race.

Our approach executes the sequential seed test-suite to invoke various methods in the library under test. We analyze the sequential execution traces to identify *unprotected* accesses to fields. If a lock on the object before accessing its contents (fields) is not held, then the corresponding access is considered unprotected. We adopt this conservative approach to maximize the effectiveness of our search for feasible locations that can be manipulated to produce a race. Therefore, even if a lock is held before accessing a field, our definition identifies the potential for a race when the lock objects differ on a shared memory access.

We also identify various methods in the library that modify object state based on the parameters supplied by the sequential tests. For a race to manifest, we need to construct a context where the intersection of the held lock objects for any two shared memory accesses is *empty*. Interestingly, while ERASER [24] uses this property to detect races, we apply the same property to *generate* race inducing tests. We identify a composition of method invocations in the library with appropriate parameters to manifest the necessary context.

Building the necessary context requires the presence of objects upon which library operations can be performed. To do this, as part of the synthesized multi-threaded test, we execute the sequential test multiple times and collect the objects, that are used as parameters (including receivers) for different method invocations, by storing references to them. Subsequently, we use the references to drive the objects to the necessary state as required by the context, spawn multiple threads and invoke appropriate methods from these threads concurrently. The resulting execution can result in a race as the held locks on the shared memory accesses do not share a common lock.

We have incorporated our ideas as part of a tool, named NARADA.¹ We have performed a detailed evaluation of NARADA on a number of open-source multithreaded Java libraries and components. Our experimental results show that we are able to synthesize a number of multi-threaded tests that expose many races, including previously undetected ones.² Analyzing nine classes resulted in the synthesis of 101 multithreaded tests leading to the detection of 307 (187 harmful) races; we are able to synthesize the tests in less

```
WriteBehindQueues.java:
27 public final class WriteBehindQueues {
     public static <T> WriteBehindQueue<T>
44
      createSafeWriteBehindQueue(WriteBehindQueue<T> q) {
45
       return new SynchronizedWriteBehindQueue<T>(g);
46
     }
47
48
     public static WriteBehindQueue
        createCoalescedWriteBehindOueue()
49
       return new CoalescedWriteBehindQueue();
50
137 }
SynchronizedWriteBehindOueue.java:
23
   /* Thread safe write behind queue. */
27 class SynchronizedWriteBehindQueue<E> implements
                                  WriteBehindQueue<E> {
29
     private final WriteBehindQueue<E> queue;
31
     private final Object mutex;
32
33
     SynchronizedWriteBehindQueue(WriteBehindQueue<E> q) {
       this.queue = q;
this.mutex = this;
37
38
39
     }
40
63
     public void removeFirst() {
       synchronized (mutex)
64
65
         queue.removeFirst();
66
67
146 }
CoalescedWriteBehindQueue.java:
17 class CoalescedWriteBehindOueue implements
                          WriteBehindQueue<DelayedEntry> {
     protected final Map<Data, DelayedEntry> queue;
19
57
     public void removeFirst() {
58
       final Set<Data> keySet = queue.keySet();
       for (Data key : keySet) {
59
60
         queue.remove(key);
```

```
61 break;
62 }
```

```
63 }
158 }
```

Figure 2: Motivating example.

than four minutes with negligible memory overhead. These results substantially improve upon recent prior work on detecting thread-safety violations [20].

The paper makes the following technical contributions:

- We develop a framework to synthesize multithreaded tests detecting races in library code by using the implementation of the library under consideration and a sequential seed test-suite as input.
- Our approach analyzes sequential execution traces, identifies unprotected accesses, derives the sequence of method invocations that drives objects to states conducive for triggering a race and reuses existing sequential tests to generate the necessary objects for multithreaded execution.
- We provide details about the implementation of our proposed design and give detailed experimental results to demonstrate the efficacy of our approach.
- We also demonstrate the usefulness of our tool, named NARADA, by *seamlessly* integrating it with RACEFUZZER [25] which reports a number of harmful races on multiple benchmarks after analyzing the execution of the tests synthesized by NARADA.

¹ The name of an Indian sage renowned for setting up conflicts to address a greater good.

²https://github.com/hazelcast/hazelcast/issues/4039

2. Motivation

We motivate the problem addressed in the paper by using a real example from hazelcast³, a popular open source in-memory data grid, that is under active development with 12K commits, 68 releases and 82 contributors. According to the documentation, the APIs are used to improve performance of applications, to distribute data across servers, clusters and geographies and to manage very large data sets or very high data ingest rates.

Figure 2 presents partial implementations of three classes from hazelcast-3.3.2. Two method implementations from WriteBehindQueues, a class that provides static factory methods which create write behind queues, are shown. SynchronizedWriteBehindQueue is a *supposedly* threadsafe class based on the comment on line 23. The implementation of the constructor and a method removeFirst is shown in the figure. Finally, the partial implementation of class CoalescedWriteBehindQueue, one of the WriteBehindQueue classes, is shown in the figure where there is no synchronization performed in the implementation of method removeFirst.

```
public void exposeRace() {
    WriteBehindQueue cwbq =
        WriteBehindQueues.createCoalescedWriteBehindQueue();

    WriteBehindQueue<DelayedEntry> swbq1 =
        WriteBehindQueues.createSafeWriteBehindQueue(cwbq);
    WriteBehindQueues.createSafeWriteBehindQueue(cwbq);
    WriteBehindQueues.createSafeWriteBehindQueue(cwbq);
    ...
    Thread t1 = new Thread() {
        void run() { swbq1.removeFirst(); }
    }
    Thread t2 = new Thread() {
        void run() { swbq2.removeFirst(); }
    }
}
```



Based on our analysis, we claim that clients using the library can potentially have races depending upon the invoked methods and the objects on which the methods are invoked. More specifically, executing the multithreaded program shown in Figure 3 can expose a race in the class SynchronizedWriteBehindQueue. This is because the two objects swbq1 and swbq2 wrap one object cwbq as shown in Figure 4. Subsequently, from two threads, removeFirst is invoked where the state of cwbq is updated. However, the updates are performed while holding locks on swbq1 and swbq2 respectively leading to a race. *Ideally*, the update on cwbq should have been performed while holding a lock on cwbq. The developers of the library incorrectly assign the this object as the mutex instead of the queue object in line 38 in SynchronizedWriteBehindQueue.java leading to this problem.

Designing this racy test manually is a non-trivial task. It requires a nuanced understanding of the implementation of the three classes, identification of the shared memory access and the associated lock object correlations, a specific invocation order of methods (in this case, constructors) with appropriate parameters to setup the necessary context, and then an invocation of relevant methods from distinct threads to cause a race. The nesting of the invocations to manifest the race makes the task more challenging. For example, in the above scenario, the race happens when removeFirst invokes remove which updates its internal state.

```
<sup>3</sup>http://www.hazelcast.org
```



Figure 4: Concurrent updates to cwbq.queue.

pub	lic void test() {
I	WriteBehindQueue cwbq =
	<pre>WriteBehindQueues.createCoalescedWriteBehindQueue();</pre>
Ţ	WriteBehindQueue <delayedentry> swbq =</delayedentry>
	WriteBehindQueues.createSafeWriteBehindQueue(cwbq);
:	swbq.removeFirst();
}	

Figure 5: Sequential seed test.

Our analysis is able to automatically synthesize the racy test shown in Figure 3. The input to our analysis is the implementation of the library along with a sequential seed test shown in Figure 5. Apart from the racy test shown in Figure 3, we are able to synthesize many racy tests that expose multiple races in SynchronizedWriteBehindQueue. Even though it may appear that objects will not be intentionally shared as described above, when the library is used as a component in a larger application, the flow of parameters can *unintentionally* result in such sharing. Our analysis can help avoid such scenarios by synthesizing tests to detect potential races when libraries are used.

3. Design

Our analysis is broadly divided into three stages. The first analyzes execution traces derived from executing sequential tests to (a) identify *unprotected* accesses (Section 3.1), and (b) identify various ways to modify library-controlled objects from a client (Section 3.2). The second stage uses this information to build constraints pertaining to library methods, along with appropriate parameters, that need to be invoked to trigger a race (Section 3.3). The third stage synthesizes racy test which executes sequential tests to build object state that can be used to enforce these constraints (Section 3.4). Analyzing the execution of the racy test with suitable dynamic analysis engines can expose underlying races.



Figure 6: Overview of test synthesis.

Figure 6 presents a design overview of our analysis. The first stage of the analysis is accomplished by the Access ANALYZER, the second stage is performed by a combination of PAIR GENERATOR and CONTEXT DERIVER, and TEST SYNTHESIZER addresses the third stage of the analysis. We now describe each stage in detail.

 $\mathcal{H} \in \mathcal{P}((Var + Var \times \overline{Field}) \mapsto (Loc \times \{C, NC\} \times \{L, U\}))$

$$\frac{\mathcal{H} = \mathcal{R}(S, \ell, C, U, x_0, x_1, \dots, x_k) \quad \ell \in Client}{\langle x_0.\mathfrak{m}(x_1, x_2, \dots, x_k)^{\ell}.\overline{T}, \phi, \phi \rangle \to \langle \overline{T}, \mathcal{H}, \phi \rangle}$$
(INVOKE)

 $\frac{\mathcal{H}' = \mathcal{H} \circ \mathcal{R}(\mathcal{S}, \ell, NC, U, x)}{\langle (x := \mathsf{alloc})^{\ell}.\overline{T}, \mathcal{H}, \mathcal{A} \rangle \to \langle \overline{T}, \mathcal{H}', \mathcal{A} \rangle} \text{ (Alloc)}$

$$\begin{aligned} \mathcal{H}' &= \mathsf{bind}(\mathcal{H}, x, y. \mathbf{f}) \\ \mathcal{R}' &= \mathcal{R}[\ell \mapsto (false, \mathcal{H}(y, l, C, U))] \\ \overline{\langle (x := y. \mathbf{f})^{\ell}.\overline{T}, \mathcal{H}, \mathcal{R} \rangle \to \langle \overline{T}, \mathcal{H}', \mathcal{R}' \rangle} \\ \mathcal{H}' &= \mathsf{lbind}(\mathcal{H}, \mathsf{alias}(\mathcal{H}, x), L) \end{aligned}$$

$$\frac{1}{\langle \mathsf{lock}(x)^{\ell}.\overline{T},\mathcal{H},\mathcal{A}\rangle \to \langle \overline{T},\mathcal{H}',\mathcal{A}\rangle}$$
(LOCK)

 $\mathcal{A} \in \mathcal{P}(Label \mapsto (Writeable \times Unprotected))$

$$\frac{\mathcal{H}' = \mathsf{bind}(\mathcal{H}, x, y)}{\langle (x := y)^{\ell}.\overline{T}, \mathcal{H}, \mathcal{A} \rangle \to \langle \overline{T}, \mathcal{H}', \mathcal{A} \rangle}$$
(ASSIGN)

 $\frac{z \in \{x.\mathbf{f}_1, x.\mathbf{f}_1.\mathbf{f}_2, \ldots\} = \mathcal{N}(x) \quad \mathcal{H}(z, l', C, \ldots) = true \quad \mathcal{A}' = \mathcal{A}[\ell \mapsto (true, false)]}{\langle (\text{return}(x))^{\ell}, \overline{T}, \mathcal{H}, \mathcal{A} \rangle \to \langle \overline{T}, \mathcal{H}, \mathcal{A}' \rangle} (\text{Return})$

$$\begin{aligned} & \mathcal{H}' = \mathsf{bind}(\mathcal{H}, \mathsf{alias}(\mathcal{H}, x) \oplus \{\mathbf{f}\}, y) \\ & \mathcal{H}' = \mathcal{H}[\ell \mapsto (\mathcal{H}(x, l, C, _) \land \mathcal{H}(y, l, C, _), \mathcal{H}(x, l, C, U))] \\ & \overline{\langle (x.\mathbf{f} := y)^{\ell}.\overline{T}, \mathcal{H}, \mathcal{R} \rangle \to \langle \overline{T}, \mathcal{H}', \mathcal{R}' \rangle} \ \text{(WRITE)} \end{aligned}$$

$$\frac{\mathcal{H}' = \mathsf{lbind}(\mathcal{H}, \mathsf{alias}(\mathcal{H}, x), U)}{\langle \mathsf{unlock}(x)^{\ell}.\overline{T}, \mathcal{H}, \mathcal{R} \rangle \to \langle \overline{T}, \mathcal{H}', \mathcal{R} \rangle}$$
(unlock)

Figure 7: Inference rules.

3.1 Analysis of Sequential Execution Traces

Our analysis operates over *traces*, a sequence of expressions that comprise the execution of a sequential test; these expressions may consist of code from the client as well as libraries. For the purposes of our discussion, traces are built from the following terms:

$$\begin{array}{rcl} e \in Exp & := & x \mid y \mid x.\mathbf{f} \mid \texttt{alloc} \mid x.\mathfrak{m}(x_1, \dots, x_n) \\ stmt & ::= & x := e \mid x.\mathbf{f} := y \mid \texttt{lock}(x) \mid \texttt{unlock}(x) \mid \texttt{return}(x) \end{array}$$

Expressions include variables (x, y, etc.), field accesses (x.f), object allocations (alloc) and library method calls. Statements are assignments to variables, which may represent objects, and fields of objects, as well as lock, unlock and return statements.

Each element in a trace has a unique label that serves as its dynamic execution index [11, 29]. Given a trace, the first step towards synthesizing a racy test is to analyze it, constructing an abstract structure of the heap and the accesses made by each trace element. Beyond recording basic points-to and aliasing information, these abstractions also maintain information about whether an access to a shared variable is *unprotected* or *writeable*. These concepts, in turn, critically rely on the notion of *controllability* - intuitively, a controllable variable references an object maintained by the library that can be manipulated by the client through library methods. For example, if a field of a library-manipulated object is set to the value passed as a parameter by a client, then the access of the field is considered controllable until the field is subsequently updated to reference an object that is not influenced by any of the method's parameters (e.g., an object allocated locally).

Having information about controllable fields enables the analysis to setup the necessary context for race synthesis. Now, if the access to an object is performed without holding a relevant lock, and the variable being used to perform the access is controllable, we consider that access unprotected, and this access can potentially be influenced in a synthesized test case to construct a race. Second, if both sides of an assignment are controllable, and the access involves a write to a field, then that access is considered writeable; such accesses can also be used to drive execution of a library method to a state that can lead to a race.

Our analysis maintains two structures - \mathcal{H} and \mathcal{A} to maintain this information. \mathcal{H} is a per-trace element abstraction of the program heap, recording the location to which a variable or field (more precisely, sequence of fields rooted at a variable) within an object is bound, whether that variable (or field) is controllable, and whether it is locked or unlocked at any specific point in the execution. In our rules, we overload \mathcal{H} 's definition so that $\mathcal{H}(x, l, Cont, Lock)$ is true if variable x points to a location l where Cont is one of C (controllable) or *NC* (*not* controllable), and *Lock* is one of *L* (locked) or *U* (unlocked). \mathcal{A} is a projection of the heap that records information about accesses, collecting for each point in the trace, writeable and unprotected information that can be used to determine whether the recorded accesses can be involved in a race in a multi-threaded execution. It effectively summarizes heap state relevant for each point in the trace.

Figure 7 defines the rules that evaluate a trace to produce these abstractions. The evaluation relation (\rightarrow) operates over a triple consisting of a trace \overline{T} whose head element is e, and whose current heap abstraction is \mathcal{H} , and whose access projection is \mathcal{A} ; each rule determines how to evaluate e to produce a new heap and a new projection.

The INVOKE rule for a library method invocation from a client⁴ assumes the existence of a bootstrapping function, \mathcal{R} . This function takes as input the source program, \mathcal{S} , the trace element index ℓ , a controllable flag, a lock flag and a set of variables. It initializes the heap \mathcal{H} by allocating new heap locations, and binding variables and fields accessible from the objects denoted by these variables, based on the variable's static type, with these flags. In the antecedent of this rule, since the invocation is initiated by the client and as we ignore the lock acquisitions within the client body, the heap established by \mathcal{R} initializes the receiver object as well as its arguments in the invocation to be controllable and unlocked.

The operator bind in rule ASSIGN (whose definition is straightforward and elided here) performs a *deep* walk over the heap, resetting aliasing properties of its arguments; two variables and/or fields are aliased if they map to the same location. Thus, after an assignment, (x := y), the heap abstraction would identify x to be aliased with y (they would both map to the same heap node, say l), x.f to be aliased with y.f where f is a field in the object referenced by x and y as determined by their static type, and so forth. Note that \mathcal{A} is left unmodified in the rule. This is because an assignment of this form does not modify the value of any field referenced by x or y; while these assignments can be leveraged by subsequent phases of the analysis to break aliasing relationships, they cannot be used to directly induce a race, which arise from storing into and reading from object fields.

When variable x is reassigned to point-to a newly allocated object, its prior aliasing relationship with other variables is broken, and an aliasing relationship with the new object is established (\mathcal{H}'). Furthermore, we initialize the controllability and locking properties of this object using the bootstrapping function, \mathcal{R} , as described

⁴ This is the only operation presumed to be executed by the client; all other rules assume the operation being considered occurs within a library method.

previously; the expression $\mathcal{H} \circ \mathcal{R}(...)$ in the antecedent of the rule ALLOC defines heap concatenation between \mathcal{H} and the heap returned by \mathcal{R} . Because the newly created object is allocated within a library method, its controllability flag is set to *NC*. Moreover, when a variable is assigned to a newly allocated object, there are no changes to the access projection because allocation does not involve writing to an object, only reassigning a reference.

Rule RETURN deals with values returned by a method that may be influenced by the method's parameters. This rule is applicable only on return to the client. Even though an object returned from a library method may have been allocated locally, it could have been assigned (or have one of its fields assigned) to a method parameter. In this case, we mark the return label as being associated with a writeable action, indicating that test synthesis can manipulate this method, or any other library method that takes arguments of the return type, via client actions. The auxiliary operator N returns the set of field access names for the returned object based on its static type. For example, in the following:

void foo (z) {

x := alloc(); y := z; w := x; w.f := y; return w;

even though the return value w is aliased to x, a variable bound to locally-allocated object, the alias of parameter z (here, y), is assigned to one of w's fields. Thus, from this return value, we know that its field f is the parameter passed to foo, thus providing a mechanism to influence any other method that requires an object with that state. To record this ability, the rule marks the return statement as writeable in \mathcal{A} .

Rule READ establishes alias relationships between x and y.f (and recursively from fields found in objects they reference through the use of bind); moreover, the access map is updated to reflect the fact that at this access (at label ℓ in the trace) no field is being written (there is only a read of y.f), and its access is unprotected only if y is controllable, and unlocked.

Rule wRITE is more complicated. Apart from performing a deep walk of the heap with x.f and y as parameters to re-establish aliasing relationships, it is imperative that all aliases of x are extracted (via operator alias) and a field dereference f from each one of them is also aliased to y. We overload the definition of bind so that the expression:

bind(\mathcal{H} ,alias(\mathcal{H} ,x) \oplus {f},y)

establishes aliasing relationships between *every* alias of x.f and y in heap \mathcal{H} ; operator \oplus concatenates its second argument to each element in the set produced by its first. For example, if z.h is an alias of x, then x.f and z.h.f need to be aliased to y. Furthermore, because a write to x happens, we check whether the right-hand side of the assignment (y) and target (x) are controllable from the client. If so, the write access becomes writeable. The detection of unprotectedness is performed as before.

LOCK and UNLOCK rules find all aliases of the parameter x and update their locking state to L and U respectively using the lbind operator that updates the heap state accordingly.

3.1.1 Example

We now explain the application of the inference rules with an illustrative example. Figure 8 presents the implementation of method foo in class A, and the trace obtained by executing the method. Interesting changes to \mathcal{H} after each label are shown in Table 1. When the method is invoked from a client, the bootstrapping of various symbols is done (using \mathcal{R}) and the resulting output is shown after label 1. Subsequently, a lock on this updates the locking state of the appropriate heap element. When this is assigned to b, the two symbols become aliases. Furthermore, as mentioned in the rule for x := y (based on bind's deep walk over the heap), *a.x* and *b.x* are also aliased. The remaining transitions in Table 1 are derived similarly.



Table 1: Interesting changes in aliasing relationships captured by \mathcal{H} at the end of each label from Figure 8. For clarity, we omit including locations and do not repeat unchanged heap elements across execution steps.

1	$a \mapsto (C,U), a.x \mapsto (C,U), a.y \mapsto (C,U), a.x.o \mapsto (C,U),$
	this $\mapsto (C,U)$, this.x $\mapsto (C,U)$, this.y $\mapsto (C,U)$,
	this.x.o $\mapsto (C,U)$, $y \mapsto (C,U)$
2	$a \mapsto (C,L)$, this $\mapsto (C,L)$
3	$b \mapsto (C,L), b.x \mapsto (C,U), b.y \mapsto (C,U), b.x.o \mapsto (C,U)$
4	$t \mapsto (C,U), t \circ \mapsto (C,U)$
5	a.x.o \mapsto (NC,U), b.x.o \mapsto (NC,U), t.o \mapsto (NC,U)
	this.x.o $\mapsto (NC, U)$
6	a.y \mapsto (C,U), b.y \mapsto (C,U), this.y \mapsto (C,U)
7	$a \mapsto (C,U)$, this $\mapsto (C,U)$, $b \mapsto (C,U)$

Apart from \mathcal{H} , we also maintain \mathcal{A} to specify whether an access at a particular point in the trace is writeable and/or unprotected. Upon applying the inference rules, at the end of the method, we observe that:

$$\mathcal{A}: \{4 \mapsto (\texttt{false}, \texttt{false}), 5 \mapsto (\texttt{false}, \texttt{true}), 6 \mapsto (\texttt{true}, \texttt{false})\}$$

The read access at label 4 is neither writeable, since it is a read, nor unprotected because b is locked (L); thus, this access is not manipulable from any client-driven testcase. Similarly, the write access at label 5 is not *writeable* because the right-hand side of the assignment is not controllable (the rand function returns a random object whose contents cannot be controlled by a client). On the other hand, the write access at label 6 is *writeable* because the righthand side of the assignment (y) and target (b) are controllable (C) (see \mathcal{H} after label 5). The access at label 5 is *unprotected* because t is unlocked (U). In contrast, the access at label 6 is protected because b is locked (L).

 \mathcal{A} is leveraged by the subsequent stages of our analysis. For example, the derivation that the access at 5 is *unprotected* suggests the construction of a test which can exploit the access to expose a race. For example, if two threads invoke foo with different objects a_1 and a_2 as the receiver respectively, in a context where $a_1.\mathbf{x}$ and $a_2.\mathbf{x}$ point to the same object, then a race manifests. However, for the fields to point to the same object, it is necessary to set the context appropriately. This is obtained by analyzing the *writeable* bit in \mathcal{A} .

While the above analysis identifies the accesses that are *write-able* and/or *unprotected*, to fully automate racy test synthesis, we still need information connecting the associated accesses to the objects that are accessible from the client; this information includes:

- · the receivers on library method invocations,
- · parameters passed to the invocations, or
- return values from the invocations.

We now describe the process of deriving this data.



Figure 9: Modified inference rules.

3.2 Connecting Client Objects to Interesting Accesses

 \mathcal{A} determines whether an access at a label ℓ is writeable and/or unprotected. However, it does not provide the necessary information to identify the client object on which the access happens. For example, in Figure 8, we know from \mathcal{A} that the access at 5 is unprotected and the access at 6 is writeable. However, we need to additionally identify the client object that corresponds to (i.e., is aliased with) the components comprising these accesses. In other words, if the client invokes $a_1.foo(b_1)$, we need to identify that $a_1.x$ is an unprotected write at 5 and $a_1.y$ is the write at 6. This aliasing information between client-supplied objects and the accesses that read or write them also needs to be established.

To be able to do this effectively, we need additional variables to identify the *source* of an access precisely. For example, in the following:

void foo (z) {
 y := z; z := alloc; x := y.f;
}

when the read of y.f occurs, the object referenced by y is the first parameter to foo and not z since z is re-allocated between y's initial assignment to z and the dereference of y.f. To ensure that we are able to track dataflow of client objects within library methods precisely, we introduce *additional* (local) variables to record parameter values. We rewrite each method with these variables such that each parameter and the receiver is assigned to a new variable (I_i) at the beginning of the method. The inference rules given in Figure 7 are applicable to these assignments, only if the method is directly invoked from the client. Moreover, for subsequent trace elements that are part of the original program, the heap locations for I_i s (and its deep dereferences) in \mathcal{H} will never need to be modified. The above example is rewritten as follows:

Here, when the ASSIGN rule from Figure 7 is applied on the assignment $I_z := z$, then I_z is aliased with z, I_z .f is aliased with z.f and so forth. Subsequently, when z is allocated, it points to a new location in \mathcal{H} . However, I_z points to the old location even though it aliases with z initially. Therefore, when y.f is considered, we can identify that y is an alias of I_z and realize that it is the parameter passed to the method.

The operator $src(x, \mathcal{H})$ returns the additional variable (or its dereference) that aliases with x in \mathcal{H} .

$$\operatorname{src}(x,\mathcal{H}) = \begin{cases} I_{i} \cdot * & \text{if } \exists i, I_{i} \cdot * \in \operatorname{alias}(x,\mathcal{H}) \\ \bot & \text{otherwise} \end{cases}$$

For example, at the end of the method body for the above example, we get $src(y, \mathcal{H}) = I_z$, $src(x, \mathcal{H}) = I_z$.f and $src(z, \mathcal{H}) = \bot$ (where \bot represents an uninteresting value).

To achieve our goal of identifying useful client objects, we modify the inference rules as shown in Figure 9. We maintain another structure \mathcal{D} that records access summaries at a label which may involve zero or more accesses. The operator (\leftrightarrow) defines a relation between the value of its right-hand side and its left-hand side. The definitions of \oplus and \circ are as before. The evaluation relation is now redefined to operate over a quadruple with the triple as defined before, and \mathcal{D} . Rule READ updates \mathcal{D} to denote that the first element of the pair is the additional synthesized variable (or one of its deep dereferences) that aliases with x and the second element is the field f of some synthesized variable (or its deep dereference) that aliases with y. A similar action is performed for the WRITE rule.

The RETURN rule is more involved. We define an update operator that creates a special variable I_r that is associated with the return variable x. For each element z in $\mathcal{N}(x)$, if $x.\overline{f}$ is *controllable*, then the associated state in x becomes writeable. Therefore, we identify the influencing parameter (or receiver) and add an appropriate mapping

$$\ell \mapsto (I_r, \overline{f} \leftrightarrow \operatorname{src}(x, \overline{f}))$$

to \mathcal{D} . Because, this is performed for each z satisfying the above criteria, there can be multiple assignments at the label corresponding to return. Essentially, this means that even if the object being returned was allocated within the library, if one (or more) of its fields are updated within the library with parameters passed from the client, that information can be collected to enable race synthesis. Consider the following code snippet:

The set, $\{I_r.z.f \leftrightarrow I_y, I_r.z \leftrightarrow I_x\}$, is the access summary at the return label. It denotes that a client can invoke the method foo to obtain an object whose fields are determined by the parameters passed by the client. This enables the client to drive the object to a chosen state.

For the example given in Figure 8, introduction of the additional variables results in the code and trace as shown in Figure 11. The generated \mathcal{A} and \mathcal{D} structures are given below:

```
\begin{array}{l} \mathcal{A}: \{4 \mapsto (\textit{false},\textit{false}), 5 \mapsto (\textit{false},\textit{true}), 6 \mapsto (\textit{true},\textit{false}) \} \\ \mathcal{D}: \{4 \mapsto \{ \bot \iff \texttt{I}_1.x\}, 5 \mapsto \texttt{I}_1.x.o \iff \bot \}, 6 \mapsto \texttt{I}_1.y \iff \texttt{I}_2 \} \end{array}
```

The binding at label 6 in \mathcal{D} indicates that the parameter affected by the left-hand side of the assignment (I_1) is the receiver object

$$\frac{e^{\ell} \in m \quad (\mathbf{I}_{i}.\mathbf{f} \nleftrightarrow \mathbf{I}_{j}) \in \mathcal{D}(\ell) \quad \mathsf{type}(\mathbf{I}_{x}) = \mathsf{type}(\mathbf{I}_{i})}{Q(\mathbf{I}_{x}.\mathbf{f}) = m} \quad (\mathsf{SET}) \qquad \frac{Q(\mathbf{I}_{x}.\mathbf{f}) = m_{1} \quad Q(o,g) = m_{2} \quad \mathsf{type}(o) = \mathsf{type}(f)}{Q(\mathbf{I}_{x}.\mathbf{f}.g) = m_{2}.m_{1}} \quad (\mathsf{CONCAT}) \\ \frac{\ell \in \mathit{client} \quad \ell'' \in m \quad (\mathbf{I}_{i}.\mathbf{f}.g \nleftrightarrow \mathbf{I}_{j}) \in \mathcal{D}(\ell'') \quad \mathsf{type}(\mathbf{I}_{x}) = \mathsf{type}(\mathbf{I}_{i})}{Q(\mathbf{I}_{y}.\mathbf{f}.g) = m} \quad (\mathsf{e}^{\ell}.\overline{T}, \ldots, \mathcal{D}) \xrightarrow{*} S_{i} \xrightarrow{*} \langle e^{\ell''}.\overline{T}'', \ldots, \mathcal{D}'' \rangle \land S_{i} \neq \langle e^{\ell''}.\overline{T}', \ldots, (\mathbf{I}_{i}.\mathbf{f} \twoheadleftarrow \ldots) \rangle}{Q(\mathsf{EP-SET})} \quad (\mathsf{DEEP-SET})$$

Figure 10: Deriving method sequences.

Figure 11: Example after introducing additional variables.

(a) and that parameter aliased with the expression on the righthand side of the assignment is the supplied argument (y), denoted by I_2 . The corresponding \mathcal{A} also shows that this is a writeable access. For non-writeable accesses, either the LHS or RHS of the access summary \mathcal{D} at a label is \perp (labels 4, 5). Nevertheless, the unprotected access at these labels can still be derived from \mathcal{D} . For example, the unprotected access at label 5 is $I_1.x.o.$

3.3 Setting the Necessary Context

We use the information describing unprotected accesses to construct racy tests. An unprotected access at a label ℓ in one thread can race with

- a concurrent access at ℓ from a different thread,
- an (un)protected access on the same object at some ℓ' in a different thread.

For each *unprotected* access detected by our analysis, we generate multiple racing pairs as described above. However, the key criterion for a race to manifest is that the object instances under consideration are the same. In general, if the racing pairs are $I_x.f_1.f_2...f_k.f$ and $I_y.f'_1.f'_2...f'_k.f'$, where f and f' have the same type, then the owner of f and f' need to point to the same object instance. In other words, $I_x.f_1.f_2...f_k$ and $I_y.f'_1.f'_2...f'_k$ need to point to the same object instance. Therefore, initially, we need to design a mechanism to set the context such that object instances are shared appropriately.

In the example shown in Figure 11, we derive that the write access at label 5 is unprotected. If we pair the unprotected access $(I_1.x.o \text{ from } D)$ with the access at the same label from a different thread, we need to set a context such that a race on a.x can manifest. This can happen only if $I_1.x$ from the two threads reference the same object. In other words, the fields x of the receivers of foo need to reference the same object. *Importantly*, there is no need to strengthen this constraint further, for example by requiring that the receivers be the same object, because imposing these additional constraints can potentially disable race detection. For instance, if the receivers (represented by I_1) are the same for foo, then the race cannot manifest because of the lock acquisition on the receivers for foo, while requiring that the field x of both these receivers nonetheless refers to the same object. This translates to

identifying method(s) that assign an object passed by the client to the field x of object of type A. The client can then pass the same object to the method(s) to set the field x for two different objects of type A.



Figure 12: Illustrative example for updating object state.

We observe that the overall problem of setting a context can be reduced to the fundamental problem of making an assignment to $x.f_1.f_2...f$ with an object specified by the client. There may not be one method that takes an object as parameter and assigns it to $x.f_1.f_2...f$. A sequence of method invocations may be necessary to accomplish the required assignment. For example, as shown in Figure 12, method *n* may assign *y.g*, *m* may assign *x.f* where *y* and *x.f* have the same types. Sequentially invoking the methods *n* and *m* with the appropriate parameters can modify the state of object *x* by assigning an object that is available from the client to x.f.g even though there is no single setter method to update the associated state. This can be extended to handle the assignment for $x.f_1.f_2...f$.

 \mathcal{D} plays a significant role in deciding the method(s) that need to be invoked. We use it to derive the required sequence of method invocations as shown by the rules in Figure 10. Q is a query operator that takes the field dereference under consideration and outputs either a method or a sequence of methods. The sET rule identifies the method m where I_j can be assigned to the field f of I_i by a client, such that the type of I_i and I_x are equivalent. The concar rule essentially identifies a sequence of two methods m_1 and m_2 such that the first method assigns I_x .f and the second method assigns o.g where the types of o and f match. The rule DEEP-SET identifies a method in which I_x .f.g is assigned. If there is no reassignment to I_i .f until I_i .f.g is assigned in m, then the method mis considered to make an assignment to I_x .f.g.

In the above description, the source of the assignment is always the object that is passed as a parameter. However, in practice, the source of the assignment can be a field of the object that is passed as a parameter. For example, in a trace z := y.g; x.f := z, where y is a parameter passed to a method, x.f is associated with $I_y.g$. To handle such a scenario, we need to repetitively apply the aforementioned rules on $I_y.g$.

We now illustrate setting the context for the example given in Figure 8 by adding method bar to class A and adding class Z as shown in Figure 13. Assume that the methods bar and baz are executed as part of some sequential test(s). Based on our analysis, analyzing the execution trace of bar will detect the presence of a *writeable* assignment to A.x, (i.e) the corresponding \mathcal{D} will have $(I_{this}.x \leftrightarrow I_z.w)$. Similarly, our analysis detects a writeable assignment for Z.w in baz.

```
class A {
 X x; Y y;
  void foo(Y y) {
   synchronized(this) {
     A b = this;
     X t = b.x;
     t.o = rand(); // unprotected access of this.x
                   // protected access of this
     b.y = y;
   }
  }
 void bar(Z z) {
   this.x = z.w; // sets field x of A
  }
class Z {
 X w = null;
  void baz(X x) {
                   // sets field w of Z
   this.w = x;
  }
}
```

Figure 13: Illustrative example for setting context.

For a race to manifest on the field x in class A when method foo is invoked by two threads, it is essential that the field x across the two invocations point to the same object. In other words, when a.foo and a'.foo are invoked by two different threads, then a.x and a'.x need to refer to the same object. Trivially, this is possible by ensuring a and a' refer to the same object. However, this will not help in manifesting the race as a lock is acquired on the receivers (a and a'). Thus, we also need to ensure a and a' are distinct. Initially, we can consider distinct a and a' and identify the possibility of ensuring that their field x can be set appropriately. We can achieve this by invoking method bar on a and a' which will help set the field x. Since the right-hand side of the assignment in bar is a field w of the object z passed as its parameter, we invoke method baz to set the field w appropriately. To summarize, the following context can be derived to manifest the potential race under consideration:

```
z.baz(x); a.bar(z); a'.bar(z); // context
a.foo(...); // thread 1, unprotected access
a'.foo(...); // thread 2, unprotected access
```

Setting the context as shown above accomplishes the task of appropriate object sharing. However, in general, for a successful execution of the synthesized code, we need legal object instances of different types (e.g., x, z, a, a'). In the next section, we address the challenge of creating such object instances to allow us to thereby correctly set the required context and invoke relevant methods concurrently from different threads to expose a race.

3.4 Synthesizing Tests

We now describe our approach for synthesizing an executable test. Primarily, we need to have legal object instances to execute the synthesized code as described in previous sections. For this purpose, we use a simple yet effective approach where we execute the sequential tests, that have been used earlier in our analysis, multiple times. However, instead of running the tests to completion, we suspend the execution before a method is invoked on the objects of interest, i.e., those that have been determined to be relevant for a race. For example, if we need to collect objects of type Z and X that need to be passed to z.baz(x), we execute the sequential test until the invocation of baz and collect the objects pointed by z and x for constructing the racing test. In other words, we store the references to these objects for later use. Similarly, if we need to invoke a method twice such that each invocation is with a different set of parameters, we execute the sequential test twice and collect the objects before the two invocations.

Unless object sharing is explicitly required, we do not share objects across different method invocations. If the objects are shared unnecessarily, it can potentially disable race detection. For instance, in the running example (Figure 13), when foo needs to be invoked separately by two threads, we ensure that two different sets of object instances are collected to be used as receivers to the invocations respectively. Otherwise, if the object referenced by the receivers are the same, the two threads cannot enter the synchronization body in foo concurrently. On the other hand, with the objects being distinct, concurrent access to the synchronized body becomes feasible. Conversely, in the running example, when the context is being set and bar is invoked twice, we ensure the same object of type Z is passed as parameter to both the invocations.

Algorithn	1 Outline of a Generated Test
Input: M	ethod pairs \mathbf{m}_r , $\mathbf{m}_{r'}$ to manifest a race (r, r') .
Se	quence of relevant setter methods Q_r, Q'_r .
1: for (ea	ach m in Q_r) $O_r[m] \leftarrow collectObjects(m)$
2: for (ea	ach m in $Q_{r'}$) $O_{r'}[m] \leftarrow collectObjects(m)$
3: $\mathbf{P}_r \leftarrow$	collectObjects(m _r)
4: $P_{r'} \leftarrow$	collectObjects(m _{r'})
5: share	$Objects(P_r, P_{r'}, 0_r, 0_{r'})$

- 6: for (each m in Q_r) Invoke m with $O_r[m]$ as parameters.
- 7: for (each m in $Q_{r'}$) Invoke m with $O_{r'}[m]$ as parameters.
- 8: Spawn a new thread and invoke m_r with parameters in P_r .
- 9: Spawn a new thread and invoke $m_{r'}$ with parameters in $P_{r'}$.

Algorithm 1 presents the outline of the multi-threaded tests synthesized by our analysis. To synthesize a test that manifests a race between accesses r and r', we use the corresponding method invocations $(m_r, m_{r'})$ as input. We also use a pair of method sequences given by Q_r and $Q_{r'}$ as input to set the relevant context which enables the objects to be driven to a state conducive for manifesting a race. Initially, we collect the objects passed as parameters to the various methods in the sequences using the auxiliary function, collect0bjects (lines 1-2). The auxiliary function invokes the appropriate sequential test, suspends the execution before the method of interest is invoked and stores the references to the objects passed as parameters to the invocation. After collecting the objects for the methods involved in setting the context, we also collect the objects for the pair of methods, m_r and $m_{r'}$, which contain the racy accesses (lines 3-4). We use the auxiliary function, shareObjects to re-arrange objects among $P_r, P_{r'}, O_r$ and $O_{r'}$ so that invoking the methods with the updated object references will help expose the race. Since all the pre-requisites for manifesting a race are satisfied, we invoke the methods in the setter methods with the appropriate objects (lines 6-7). Subsequently, we spawn new threads and invoke the methods that contain the racy accesses concurrently.

Table 2: Application of Algorithm 1 on example.

	Before sharing	After sharing		
O_r	baz: (z_1, x_1) , bar: (a_2, z_2)	baz: (z_1, x_1) , bar: (a_5, z_1)		
$O_{r'}$	baz: (z_3, x_3) , bar: (a_4, z_4)	baz: (z_1, x_1) , bar: (a_6, z_1)		
P_r	foo : (a_5, y_5)	foo: (<i>a</i> ₅ , <i>y</i> ₅)		
$P_{r'}$	foo: (a_6, y_6)	foo : (a_6, y_6)		

For the example presented in Figure 13, the method of interest is foo. The method sequences Q_r and $Q_{r'}$ are {baz, bar} as described previously. When the synthesized multi-threaded test is executed as shown in Algorithm 1, the collected objects for all these methods are shown in Table 2. Due to shareObjects, the parameters of the invocations are re-arranged suitably (shown in boxes). Subsequently, after the methods for setting the context, z_1 .baz (x_1) , a_5 .bar (z_1) and a_6 .bar (z_1) are executed, a_5 .x and a_6 .x will point to x_1 . Finally, the test case invokes $a_5.foo(y_5)$ and $a_6.foo(y_6)$ from two newly spawned threads respectively. The synthesized test can expose a race on $a_5.x.o$ (equivalent to $a_6.x.o$) as it is modified by two threads concurrently.

4. Implementation

We have implemented the analysis described in Section 3 to synthesize racy tests for multithreaded Java libraries. We use soot [28] for instrumenting the bytecode and obtain the execution traces from the sequential test for further analysis. There can be nested calls from a library invocation. We scope the variable names by assigning unique index for each method invocation.

We perform *lazy initialization* to implement the functionality described by \mathcal{R} in Section 3. This is because it is not always possible to assign a separate heap location for every variable by performing a deep walk on the type graph (e.g., linked list). Therefore, when a library method is invoked from the client (the sequential test), we initialize the variables corresponding to the various parameters passed to the invocation and set the controllable and locked flags. Subsequently, for an unseen variable, we assign the flags based on its owner state. For example, the flags for an unseen variable *x*.*f* will be assigned based on the state of *x*.

Our implementation considers the access to some field x.f to be *unprotected* if a lock on x is not held. In practice, there can be scenarios where x.f is always accessed with a lock held on x.g; in other words, there can be a strong correlation between the accesses and some other field. We adopt a conservative approach and consider the access as unprotected and attempt to synthesize a test. The downside of this conservative analysis is the synthesis of tests that will not expose any races.

It is possible to synthesize varied method sequences to set the same context. For example, there can be multiple setters for a field. Our implementation randomly selects one of the possible methods to derive the required method sequence. We treat constructor as any other method to help set the context, but discard unprotected accesses found in them while building the racing pairs. Moreover, in some scenarios, we may not be able to derive the context for assigning the entire field dereference $x.f_1 \dots f$. We attempt to assign the prefixes of the dereference so that the objects at some point of the hierarchy are shared, which can also lead to tests that do not expose races. Methods need not always run to completion to drive an object to a specific state. For example, there can be a strong non-controllable update to a field after the controllable assignment, which can override the earlier update and will not help in setting the required context. We handle it by letting a separate thread invoke the method and suspend its execution at the label corresponding to the writeable assignment or the closest point where all held locks are released.

We integrate the output of our implementation, named NARADA, with RACEFUZZER [25]. The integration is seamless and does not require *any* modifications to the race detector.

5. Experimental Validation

We validate our approach by synthesizing racy tests for multithreaded Java libraries, including thread-safe classes, using our implementation. The experiments are performed on an Ubuntu-14.04 desktop running on a 3.5 Ghz Intel Core i7 processor with 16GB RAM. The information regarding the various libraries used is given in Table 3. hazelcast is an open-source in-memory data grid, openjdk is the Java Development Kit, colt is a high performance scientific computing library, hsqldb is a leading SQL relation database software, hedc is a web-crawler application, h2 is a SQL database engine, and classpath contains core class libraries for use with virtual machines and compilers. The versions of the benchmarks and the classes analyzed for synthesizing races

Table 3: Benchmark Information.

Benchmark	Version	Class name		
hazelcast	3.3.2	SynchronizedWriteBehindQueue (C1)		
openjdk	1.7	SynchronizedCollection (C2)		
		CharArrayWriter (C3)		
colt	1.2.0	DynamicBin1D (C4)		
hsqldb	2.3.2	DoubleIntIndex (C5)		
		Scanner (C6)		
hedc	NA	PooledExecutorWithInvalidate (C7)		
h2	1.4.182	Sequence (C8)		
classpath	0.99	CharArrayReader (C9)		

Table 4: Synthesized test count and synthesis time.

Class	Methods	LoC	Race	Tests	Time
			Pairs		(in secs)
C1	14	104	65	15	12.2
C2	19	85	131	40	13.5
C3	13	92	13	9	2.2
C4	35	313	26	11	33.0
C5	32	508	136	8	7.4
C6	26	1802	85	8	121.7
C7	9	191	4	4	3.6
C8	18	233	4	4	5.8
C9	8	102	2	2	1.9
Total			466	101	201.3

are given in the Table 3. For brevity, we refer to the analyzed classes as C1...C9.

Table 4 presents the number of methods and the lines of code in each class. If a brute-force approach is employed to detect races in a class, not only all possible pairs of methods need to be invoked concurrently (e.g., 35×35 for C4), the objects passed to them should be in a state that can induce a race. Consequently, the overall search space is quite large. We analyze the classes with our implementation by constructing a seed-testsuite for each class where each method in the class is invoked exactly once, without any constraints on the state of the objects involved in the invocations. With this trivial effort, we are able to identify 466 racing pairs across all the classes.⁵ Because our analysis also provides enough information pertaining to the context, it helps prune the search space significantly leading to the overall synthesis of 101 tests. It is not necessary for all racing pairs to have unique tests because there can be multiple unprotected accesses of the same field within a method. For example, 15 tests are synthesized to enable detection of 65 racing pairs in C1. The overall analysis time for all the classes to synthesize the tests is less than four minutes.

Table 5 shows the results of applying RACEFUZZER [25], a dynamic data race detection tool, on the execution of the tests synthesized by NARADA. 307 races were detected in total, of which 259 races were automatically reproduced. We manually analyzed the output of the automatically reproduced races and identified that 187 of these races are indeed harmful. The 62 benign races in C6 are due to a reset method which resets a number of fields to constant values. Moreover, we manually triaged races that could not be reproduced by RACEFUZZER, and discovered that 44 out of these 48 races were true positives (TP), with the remaining four being false positives (FP) that arise due to imprecision in the detector.

⁵ We did not list eight other classes in openjdk because the races were very similar to the races in SynchronizedCollection.

Class	Races	Repro	Manual		
	Detected	Harmful Benign		TP	FP
C1	76	58	2	12	4
C2	84	65	1	18	-
C3	8	7	1	-	-
C4	4	2	0	2	0
C5	36	30	6	-	-
C6	89	15	62	12	-
C7	4	4	-	-	-
C8	4	4	-	-	-
C9	2 2 2		-	-	-
Total	307	187	72	44	4

Table 5: Analysis of synthesized tests by RACEFUZZER. Manually analyzed data for races detected by RACEFUZZER but not reproduced by it is also shown.

There are a few races that are not feasible for which our system produces a racing pair. For example, we construct 26 racing pairs for C4 where only four races are detected. This is because the races on some fields can never manifest as the necessary fields to set a suitable context can never be influenced from clients. The implementation of NARADA and the raw experimental data are publicly available⁶ and we refer the interested reader to it for further details.

Figure 14 presents the distribution of tests as a function of the number of detected races. For C5, C6...C8, each test detects at least one race. For the remaining classes, we also synthesize tests that do not detect any race. As mentioned previously, when we are unable to set a context for a specific field, we try to achieve object sharing to the extent possible by making assignments to its ancestors. For example, in C4 where setting the context is not possible, we still synthesize tests for those racing pairs resulting in a majority of the tests not enabling race detection.



Figure 14: Distribution of tests w.r.t the number of detected races.

We also analyzed these library classes with CONTEGE [20], a tool that performs a *random* search to detect thread safety violations. It was able to detect two thread-safety violations in C5 and one thread-safety violation in C6 by generating 2.9K and 105 tests respectively. The detected thread-safety violations are races, which are also detected by our analysis. For other benchmarks, it generated between 1K- 70K tests, yet was unable to detect any thread-safety violations.

6. Related Work

We are the first to design and implement a *directed* approach for synthesizing tests to enable race detection. In CoNTEGE [20], the authors propose an approach for precisely detecting thread safety violations by generating *random* multithreaded executions. If the multithreaded execution crashes or deadlocks and the corresponding serialized execution executes without a problem, the tool reports a thread safety violation. Because of the randomized nature of executing the tests, any two methods can be invoked concurrently and the objects passed to them need not necessarily be shared. In contrast, our approach analyzes the sequential execution to identify problematic regions and synthesizes a racy test accordingly. Because of the directed nature of our design, we are able to prune the overall search space for races effectively.

In previous work [22], we designed an approach for synthesizing tests to detect deadlocks [23]. We differ from this approach in the intended application (races vs deadlocks) and also the properties that need to be inferred for the purpose of synthesizing tests to enable race detection. In [6], the authors propose an approach that employs concolic execution techniques to cover code regions to enable race detection. In contrast, we propose a technique for synthesizing multithreaded tests. The execution of the synthesized tests can potentially be integrated with their approach for exposing racing schedules.

A number of dynamic analysis techniques have been proposed to enable race detection. FASTTRACK [7] is a dynamic race detector that leverages the *happens-before* relation to detect races precisely. It improves the performance of DJIT+ [19] by employing epochs to minimize the comparison of the vector clock times. RACE-FUZZER [25] presents a mechanism for *fuzzing* the schedule to expose races. ERASER [24] uses the lockset algorithm to detect potential races. All these approaches require defect inducing multithreaded tests to be executed to detect races and can leverage the tests synthesized by our implementation.

CHESS [13] systematically explores the state space to detect concurrency bugs. The priority-based probabilistic concurrency testing (PCT) [3] and parallel PCT [14] propose strategies for quickly detecting concurrency bugs. MAPLE [30] is a tool that exposes untested thread interleavings to enable concurrency bug detection. The availability of multithreaded tests is a pre-requisite for these approaches; all these approaches can benefit from the tests synthesized using our implementation. IMUNIT [9] is a framework for specifying the schedules on multithreaded tests. We can leverage the framework to suitably specify racing schedules on the synthesized tests.

Many static analysis approaches have been proposed to detect races [5, 15, 21]. These approaches require a programmer to verify the correctness of the reported defect. Moreover, because of the static nature of the analysis, there can be many false positives. In contrast, by automatically synthesizing tests and integrating it with dynamic race detectors, we can automatically identify real races and eliminate the need for manual intervention.

Numerous techniques for automatically generating tests have been proposed in the literature [4, 8, 16]. The primary goal for these approaches is to improve code coverage and detect sequential bugs. Apart from the goals being different, we address a different set of challenges including identifying the relevant methods to be invoked concurrently and the appropriate sharing of objects. Our approach uses sequential tests as part of a seed testsuite and can use the output of the sequential test generators for bootstrapping.

In [17], the authors propose a synthesis mechanism inspired by test-driven development where the input/output examples are consumed to synthesize programs. Test-driven repair [27] combines static and dynamic analysis to identify modifications to code that will prevent races. We differ from these approaches in terms of the intended application and the underlying mechanism.

⁶http://www.csa.iisc.ernet.in/~sss/tools/narada

7. Conclusions

Multithreaded tests are necessary to detect races using dynamic analysis engines. However, developing effective multithreaded tests is hard. In this paper, we presented the design and implementation of a novel approach to automatically synthesize racy tests to enable race detection in multithreaded libraries. We demonstrate the efficacy of our approach by applying our implementation, named NARADA, on multiple open-source Java libraries leading to the detection of many harmful races.

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