Understanding and Detecting Concurrency Attacks

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The University of Hong Kong
@cs.hku.hk

Abstract

1. Introduction

Multi-threaded program is hard to be correct. Concurrency bugs are common in modern multi-threaded programs including atomic violation, ..., and especially data race[38, 39, 71, 72]. Extant work well explores interleaving that causes concurrency bugs, and efficiently detects explicit concurrency bugs that direct to severe consequences such as execution order violation, wrong output and program crash [38, 42, 57, 64, 71, 72].

Recent studies[55, 68] show rise of concerns about *concurrency attacks*. Just like sequential bugs lead to attacks, concurrency bugs also lead to concurrency attacks. By triggering concurrency bugs and employing subtle inputs, hackers may leverage the corrupted memory to conduct attacks including privilege escalations[8, 10], hijacking code execution[9], bypassing security checks[3–5], and breaking database integrity[55]. These vulnerabilities often hide in large amount of concurrency bug reports. For example, bug information leveraged by a xxx attack is hidden in 1000 race reports produced by TSAN[57], a famous and widely used data race detector. ***Also, despite threads conducting concurrency bugs, behaviors of another thread may also be affected when data race bugs infecting shared memory (e.g.heap overflow)[1, 6]. ***

Unfortunately, although great progress has been made to detect and replay severe bugs(e.g.ConMem[72]), extant work still lacks exploration of concurrency attacks from enormous concurrency bugs. Our study over several known concurrency bugs[1, 2] shows that even concurrency bugs have been successfully detected and reported, professionals may still lack knowledge about how severe consequences these bugs may cause. For instance, *apache-25520*[1] has been reported over years and well studied by researchers[39]. We are the first to exploit a new heap overflow attack leveraging on this bug and break the HTML integrity.

We studied 26 attacks and find two major challenges for exploring concurrency attacks from concurrency bugs. First, concurrency attacks may be implicit and hidden in common concurrency bugs. Existing concurrent bug detectors focus on bug happening itself. ConMem[72] first propose to consider concurrency bugs that may cause severe consequences (e.g.program crash) and ConSeq[71] uses severe consequence report to help diagnose concurrency bugs. However, our observation shows that explicit error (e.g.program crash) is not necessary for concurrency attacks. Some concurrency bugs (e.g.*xxx-xxx*) may not cause program crash or interrupted, but attackers may still leverage them and conduct attacks with crafted input. ???-Worse still, crash bugs may even lead to more severe vulnerabilities. In *CVE-2017-7533* conducted by our team, although the data race primarily causes kernel crash, we crafted the input and successfully conduct a privilege escalation attack without crashing the kernel. Current concurrency bug detecting tools are not designed to analyze this kind of latent vulnerabilities, and hence may direct wrong level of warnings towards the bugs.

Second, extant work ignores indicating the *victim thread* of concurrency attacks. A concurrency bug may become much more vulnerable when attackers employee another thread to construct their attacks. In CVE-2017-7533, we do not only leverage two threads to trigger a data race and construct kernel heap overflow, but also require another *victim thread* to lay the target structure on the same heap. By crafting inputs and corrupting the target structure, we finally achieve arbitrary code execution and get a root shell. Automatically indicating the victim thread would be of vital helpful for developers to better understand the latent vulnerabilities. For example, in the Apache-25520 case, after knowing about information of victim thread, we successfully increased the severity of the bug and conducted an integrity violation.

To address the two challenges, we introduced a new model(§3.1) that explains most concurrency attacks we studied. The model breaks down concurrency attacks into three stages: bug happening, bug-to-attack propagation, and attack

happening. In this model, the two key things for exploring concurrency attacks is the analysis of bug-to-attack propagation and definition of attack happening, while bug happening has been well studied and detected. Our studies show propagation includes data flow and

Leveraging this model, we designed a two-phase framework(§3.2), XXX, for detecting concurrency attacks. The first phase is *concurrency analyzer* to analyze bug-to-attack propagations. Our study found that most vulnerable races are already included in the race detectors reports, and concurrency attacks sites are often explicit in program code. Therefore, we can perform static analysis on only the data and control flow propagations between the bug reports and the potential attack sites, then we can collect relevant call stacks and branch statements as the potentially vulnerable input hints.

The second phase is *concurrency fuzzer?* to ... .

We implemented XXX on Linux, supporting both user space and kernel space attack detection.

We evaluated OWL on 6 diverse, widely used programs, including Apache, Chrome, Libsafe, Linux kernel, MySQL, and SSDB. OWLs benign schedule hints and runtime verifiers reduced 94.3% of the race reports, and it did not miss the evaluated concurrency attacks. With the greatly reduced reports, OWLs vulnerable input hints helped us identify subtle vulnerable inputs, leading to the detection of 7 known concurrency attacks as well as 3 previous unknown, severe ones in SSDB and Apache. The analysis performance of OWL was reasonable for in-house testing.

This paper makes two major contributions:

1. **A general model explains happening and exploiting of concurrency attacks.** This model explains most concurrency attacks in wild and providing two major direction for detecting concurrency attacks.
2. **A general concurrency attack detection framework and its implementation, XXX.** XXX can easily employ existing concurrent bug detectors and vulnerability analyzer to improve the accuracy and ??? of detection.

The rest of this paper is structured as follows. §2 introduces the background of concurrency attacks. §3 gives an overview on the concurrency attack model and architecture of XXX. §**??**...

2. Background

2.1 Concurrency Bug

2.2 Concurrency Attack

Extant studies [55, 68] concurrency attacks into .. concurrency bugs and made three major findings on concurrency attacks. First, concurrency attacks make much more severe threats: 35 of the bugs can corrupt critical memory and cause four types of severe security consequences, including privilege escalations [7, 10], malicious code injections [8], and bypassing security authentications [4, 3, 5]. Second, concurrency bugs and attacks can often be easily triggered via subtle program inputs. For instance, attackers can use inputs to control the physical timings of disk IO and program loops and trigger concurrency bugs with a small number of re-executions. Third, compared to traditional TOCTOU attacks, which stem from corrupted file accesses, handling concurrency attacks is much more difficult because they stem from corrupted, miscellaneous memory accesses.

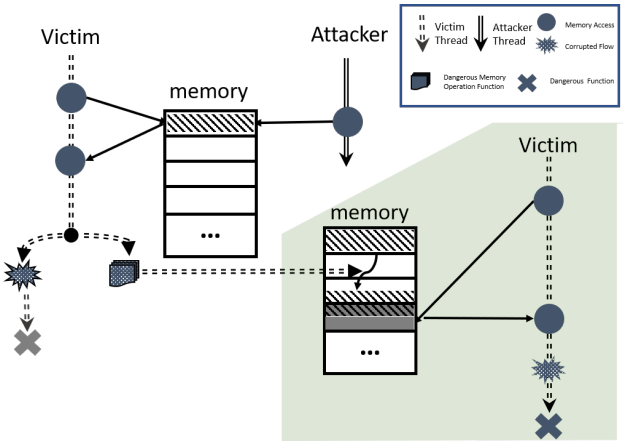

Figure 1: ***Concurrency Attack Model***

2.3 Bug-to-Attack Propagation

3. Overview

We introduce a model that explains most concurrency attacks we studied. Leveraging the model, we design a framework XXX to detect concurrency attacks. This section gives an overview of the model and architecture of XXX.

3.1 Concurrency Attack Model

3.2 XXX's Architecture

4. XXX's Framework

5. Implementation

6. Discussions

7. Evaluation

We evaluated XXX on 6 widely used C/C++ programs, including three server applications (`Apache` [12] web server, `MySQL` [11] database server, and SSDB [54] key-value store server), one library (`Libsafe` [36]), the 4.11.9 Linux kernel, and one web browser (`Chrome`). We used the programs' common performance benchmarks as workloads. Our evaluation was done on XXX.

We focused our evaluation on four key questions:

1. How many false reports from concurrency error detection tools can XXX reduce (§**??**)?
2. How many potential victim threads can XXX indicate(§3)?
3. Can XXX detect known concurrency attacks in the real-world (§**??**)?

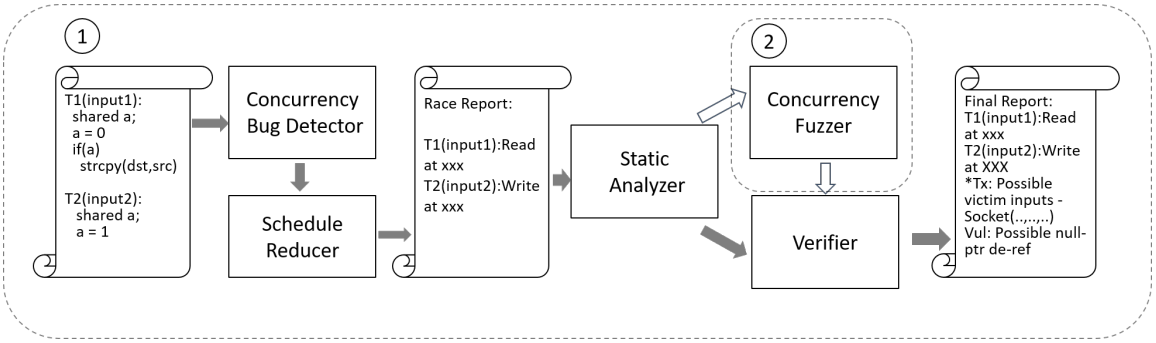

Figure 2: ***Architecture***

4. Can XXX detect previously unknown concurrency attacks in the real-world (§**??**)?

Name	LoC	# r.r.	# XXX's r.	# atks	# atks found
Apache	290K	715	10	3	3
Chrome	3.4M	1715	115	1	1
Libsafe	3.4K	3	3	1	1
Linux	2.8M	24641	34	2	2
MySQL	1.5M	1123	16	2	2
SSDB	67K	12	2	1	1
Total	-	-	-	-	-

Table 1: XXX ***race report reduction and concurrency attack detection results.*** Description

Name	Type	# Fed Inputs	# Victim Inputs	# atks
Apache	apr_palloc	-	-	-
Linux	kmalloc	180	6	1
Total	-	-	-	-

Table 2: XXX*'s Concurrency Fuzzer.* Description

8. Related Work

TOCTOU attacks. Time-Of-Check-to-Time-Of-Use attacks [16, 56, 58, 61] target mainly the file interface, and leverage atomicity violation on time-of-check (access()) and time-of-use (open()) of a file to gain illegal file access.

A prior concurrency attack study [66] elaborates that concurrency attacks are much broader and more difficult to track than TOCTOU attacks for two main reasons. First, TOCTOU mainly causes illegal file access, while concurrency attacks can cause a much broader range of security vulnerabilities, ranging from gaining root privileges [8] , injecting malicious code [7], to corrupting critical memory [1]. Second, concurrency attacks stem from miscellaneous kinds of memory access, and TOCTOU stem from file accesses, thus handling concurrency attacks is much more difficult than TOCTOU.

Sequential security techniques. Defense techniques for sequential programs are well studied, including taint tracking [24, 41, 42, 45], anomaly detection [23, 49], address space randomization [52], and static analysis [15, 17, 26, 31, 59].

However, with the presence of multithreading, most existing sequential defense tools can be largely weakened or even completely bypassed [67]. For instance, concurrency bugs in global memory may corrupt metadata tags in metadata tracking techniques. Anomaly detection is lack of a concurrency model to reason about concurrency bugs and attacks.

Concurrency reliability tools. Various prior systems work on concurrency bug detection [25, 35, 38, 40, 48, 62, 70–72], diagnosis [14, 34, 43, 44, 50], and correction [32, 33, 60, 63]. They focus on concurrency bugs themselves, while OWL focuses on security related consequences of concurrency bugs. Therefore, these systems are complementary to OWL.

Conseq [71] detects harmful concurrency bugs by analyzing their failure consequence. Its key observation is that concurrency bugs and those bugs' failure sites are usually within a short control and data flow propagation distance (e.g., within the same function). Concurrency attacks(targets of OWL) usually exploit corrupted memory that resides in different functions, thus Conseq is inadequate for concurrency attacks. Though, Conseqs proactive harmful schedule exploration technique will be useful for OWL to trigger more vulnerable schedules.

Static vulnerability detection tools. There are already a variety of static vulnerability detection approaches [27, 27, 37, 53, 65, 73], which fall into two categories based on whether they target general or specific programs.

The first category [37, 65] targets general programs and these approaches have been shown to find severe vulnerabilities in large code. However, these pure static analyses may not be adequate to cope with concurrency attacks. Benjamin et al. [37] leverage pointer analysis to detect data flows from unchecked inputs to sensitive sites. This approach ignores control flow and thus it is not suitable to track concurrency attacks like the Libsafe one in 4.3**[TOCHECK]**. Yamaguchi et al. [65] did not incorporate inter-procedural analysis and thus is not suitable to track concurrency attacks either. Moreover, these general approaches are not designed to reason about concurrent behaviors (e.g., [65] can not detect data races).

OWL belongs to the first category because it targets general programs. Unlike the prior approaches in this category, OWL incorporates concurrency bug detectors to reason about concurrent behaviors, and OWLs consequence analyzer integrates critical dynamic information (i.e., call

stacks) into static analysis to enable comprehensive dataflow, control- flow, and inter-procedural analysis features.

The second category [13, 27, 53, 73] makes static analysis focused on specific behaviors (e.g., APIs) in specific programs to achieve scalability and accuracy. These approaches check web application logic [27], Android applications [13], cross checking security APIs [53], and Linux Security Module [73]. OWLs analysis is complementary to these approaches; OWL can be further integrated with these approaches to track concurrency attacks.

Symbolic execution. Symbolic execution is an advanced program analysis technique that can systematically explore a programs execution paths to find bugs. Researchers have built scalable and effective symbolic execution systems to detect software bugs [17–19, 21, 28–30, 46, 51, 69], block malicious inputs [47], preserve privacy in error reports [20], and detect programming rule violations [22]. Specifically, UCKLEE [46] has been shown to effectively detect hundreds of security vulnerabilities in widely used programs. Symbolic execution is orthogonal to OWL; it can augment OWLs input hints by automatically generating concrete vulnerable inputs.

9. Conclusion

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