

Master's Thesis

Automated Memory Error Repair
Based on Hybrid Program Analysis

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Abstract

Automated program repair is a technology that aims to fix program errors and vulnerabilities automatically. In the field of memory error repair, with the development of bug detection tools, we can easily detect memory errors in programs. However, fixing those errors is time-consuming and error-prone. [The program's heap-related behaviors, such as heap object, error source and sink, play a critical role in memory error repair, so how to detect the errors and collect the relevant information are the main tasks.](#) The existing techniques are mainly based on static analysis, where the static analyzer is used to detect program memory errors and then repair tools collect the essential information via static analysis. But static analyzer may give wrong alarms which will affect the performance of the repair tools, and [heap-related behavior analyzing](#) often requires high overhead.

We present **HAMER**, a hybrid automated memory error repair tool that aims to address those shortcomings by using hybrid analysis. HAMER first uses fuzzer to check the alarms given by the static analyzer and extracts the real errors from those alarms. Then it tries to fix those errors by using hybrid analysis. HAMER is [an](#) automated memory error repair technique based on fuzzing results. HAMER does not waste time resolving alarms that are incorrect because the errors reported by fuzzer are real errors, and HAMER also utilizes fuzzer to verify the generated patches to ensure they are correct. For the necessary information for synthesizing patches, HAMER uses [existing](#) lightweight static analysis techniques to collect it. [Our preliminary experiment suggests](#) that HAMER can fix the complicated memory errors effectively, which the existing static memory error repair technique can not repair.

Contents

Abstract	ii
1 Introduction	1
2 Background	3
2.1 Infer	3
2.2 LibFuzzer	4
2.3 Component-Based Program Synthesis	5
3 Overview	6
3.1 Motivating Example 1	8
3.2 Motivating Example 2	9
4 Approach	10
4.1 Patch Template	10
4.2 Error Detection	10
4.3 Dependency Collection	11
4.4 Source Instrumentation	12
4.5 Patch Generation	12
4.5.1 simp-CBPS	12
4.5.2 Repair Algorithm	14
4.5.3 Function Fix	14
4.5.4 Temporary Variable	16
5 Evaluation	17
5.1 Implementation	17
5.2 Experimental Setup	17
5.3 Experimental Results	18
5.3.1 Research Question 1	18
5.3.2 Research Question 2	20
5.3.3 Research Question 3	22
6 Related Work	24
7 Conclusion	26
7.1 Conclusion	26
7.2 Limitations	26
7.3 Future Work	26

List of Figures

2.1	Example of Infer	3
3.1	HAMER pipeline	6
3.2	Motivating Example 1: Infer false-negative alarm	7
3.3	Motivating Example 2: Infer false-positive alarm	7
5.1	test5	20
5.2	test10	21
5.3	test11	22
5.4	test8	23

List of Tables

3.1	Instrumentation result of o1	8
4.1	Test suite	13
5.1	Characteristics of synthesizing code	18
5.2	Evaluation result	19
6.1	Related works	25

Chapter 1

Introduction

Memory errors, such as memory leaks, can have catastrophic effects, thus detecting and fixing them has always been a critical task for developers. Memory error detection performance is improving with the development of memory error detection technologies, however resolving these problems takes a lot of time and work for developers, and erroneous patches might lead to more significant effects.

Many existing leak detection approaches only report the stack where the allocation occurs. To fix the leak, we must add a deallocation statement, but the suitable fix location is usually far away from the allocation point. Furthermore, these pointers to allocated memory are passed between functions and it often becomes difficult for the developer to find the suitable points where to deallocate the memory. If the developer inserts a wrong patch or inserts a patch at a wrong location, it may lead to more significant errors which can lead to several exploits by attackers. Hence it is important to track the use of allocated memory (read/write) along different program paths for suggesting a fix.

Existing memory error repair techniques [11, 16] are mainly based on static analysis. This is because fixing errors like memory leaks necessitates an understanding of heap-related behavior, such as error source and sink. Collecting heap-related behavior information needs a high time and space overhead. Because wrong patches or fixing locations might lead to more significant errors, and different heap objects can interact with each other, all of this must be fully considered when generating patches. The state-of-the-art automated memory error repair technology SAVER [11] stores these heap-related behaviors by constructing an object flow graph, which has high time and space complexity. However, because static analysis techniques are not good at dealing with problems such as indirect calls and alias, the repair tools based on static analysis might generate the wrong patches. While dynamic analysis can detect memory errors efficiently by instrumenting the critical program points to observe the legality of program behavior during execution, it does not provide enough information to generate patches.

In this paper, we present **HAMER**, a hybrid analysis-based memory error repair technique. We use a static analyzer to detect the program first, then use a fuzzer to detect the alarms and extract the real errors. Fuzzer triggers runtime errors, thus it will not give false-positive alarms, allowing HAMER to avoid trying to fix wrong alarms. After that, we collect program variables that can be utilized to synthesize patches by variable dependency analysis. During this procedure, we also collect the return locations of each function as the candidate fix locations. With this lightweight static analysis, HAMER is able to obtain enough information to fix the buggy program. We then gather the test cases generated by the fuzzer which trigger or do not trigger the errors and use the component-based program synthesis [12] to try to generate patches from these variables and test cases. Finally, we utilize the fuzzer to verify the current fixed program, and if the repair is erroneous, we collect the test case that triggers the errors and repeat our repair method until the error

is fixed or timeout. This strategy ensures that the patches generated by HAMER can repair current errors without introducing new ones. Our preliminary experiment suggests that HAMER can fix the complicated memory errors effectively, which the existing static memory error repair technique cannot repair. It also shows that our lightweight static analysis can collect enough information in a short time to assist HAMER to generate the correct patches.

Contributions. This paper makes the following contributions:

- We present a new technique for repairing memory errors based on hybrid analysis. We present an efficient repair algorithm that only needs lightweight static analysis to collect relevant information and makes flexible use of the fuzzer’s result to generate and verify patches. Also, our technique can generate and insert conditional deallocation patches correctly and can fix multiple memory leaks at the same time.
- We present HAMER ¹, a memory error repair tool that implements the proposed approach.

¹<https://github.com/QIANZECHANG/HAMER>

Chapter 2

Background

HAMER detects bugs using existing bug detection tools. In this section, we will introduce the static analyzer (Infer [3] ¹) and fuzzer (LibFuzzer ²) that HAMER uses. We also introduce the component-based program synthesis (CBPS) [12] which we use to synthesize the patches.

2.1 Infer

Infer is a static program analyzer for Java, C, and Objective-C, written in OCaml, developed by Facebook. At present Infer is tracking problems caused by null pointer dereferences and resource and memory leaks, which cause some of the more important problems. Infer is a high-performance static analyzer with high scalability and efficiency, and it is widely used by programmers and researchers. Infer automatically discovers specifications for the functions that can be proven to be memory safe using the theoretical notion of bi-abductive inference [4]. It consists in solving the following extension of the entailment problem: $H * X \vdash H' * Y$, where H and H' are given formulas in separation logic and X and Y needs to be deduced. The efficiency of bi-abductive inference derives from the fact that it allows discovering the part of the heap missing for using a specification within a particular calling context of another procedure. For example, in Figure 2.1, the *foo* function dynamically allocates memory through the *alloc* function. Infer uses bi-abductive inference to approximate the procedures' heap behavior, so Infer does not determine that *alloc* occurs memory leak because *alloc* returns *p*, meaning that *p* is not unreachable and the responsibility for freeing *p* is transferred to *foo*. Infer will construct the specification of *alloc* using these heap features and use it to analyze other procedures when *alloc* is called. Therefore, in *foo*, *p* gets the return of *alloc* and has not been freed until last, Infer determines that *foo* occurs memory leak.

```
1 int* alloc() {  
2     int* p=malloc(sizeof(int));  
3     return p;  
4 }  
5  
6 void foo() {  
7     int* p=alloc();  
8 }
```

Figure 2.1: Example of Infer

¹<https://fbinfer.com/>

²<https://llvm.org/docs/LibFuzzer.html>

However, Infer is hard to handle some issues, such as indirect call and alias, which causes Infer to provide false-negative and false-positive alarms. As a result of this shortcoming, automated repair tools may waste time on the wrong alarms (false-positive) and have no opportunity to fix the errors that are not discovered (false-negative).

2.2 LibFuzzer

LibFuzzer is an in-process, coverage-guided, evolutionary fuzzing engine. LibFuzzer is linked with the library under test, and feeds fuzzed inputs to the library via a specific fuzzing entry point (target function); the fuzzer then tracks which areas of the code are reached, and generates mutations on the corpus of input data in order to maximize the code coverage. The code coverage information for libFuzzer is provided by LLVM’s SanitizerCoverage³ instrumentation.

LibFuzzer uses the information provided by the AddressSanitizer [23]⁴ and the LeakSanitizer⁵ to determine if an error has happened within the detected code coverage when detecting memory errors. AddressSanitizer uses a shadow memory to encode the allocated space (address and size), and the behavior is verified for legality each time the space is accessed. AddressSanitizer instruments every reads and writes to the memory during compile-time and checks verifies that the memory being read/written to is actually valid, by crosschecking the shadow memory’s encoding during program execution. LeakSanitizer stores the allocation thread ID and the allocation stack itself each time memory is allocated and records the situation when memory is deallocated. At the very end of program execution, LeakSanitizer will go over the heap to find out such objects that have not been freed.

LibFuzzer relies on a corpus of sample inputs for the code under test. This corpus should ideally be seeded with a varied collection of valid and invalid inputs for the code under test. The fuzzer generates random mutations based on the sample inputs in the current corpus. If a mutation triggers the error in a previously-uncovered path in the code under test, then that mutation is saved to the corpus for future variations. LibFuzzer will work without any initial seeds but will be less efficient if the library under test needs complex, structured inputs.

Because LibFuzzer can only fuzz test one function argument at a time, we must test each argument independently for functions with multiple arguments. LibFuzzer may take a long time or fail to mutate the inputs that can go into each path for the functions that need complex inputs. Furthermore, LibFuzzer will stop and report errors anytime it triggers the error, thus if a function has several errors in different paths, LibFuzzer cannot ensure that all of them will be detected at the same time. The code below, for example, has two memory leaks.

```
p1=malloc(1);  
if(a<5)p2=malloc(1);
```

The ideal situation would be for LibFuzzer to input the value smaller than 5 and trigger both errors. But in most cases, LibFuzzer inputs other values only triggers the memory leak of *p1* and then exits. As a result, LibFuzzer may not be able to detect all of the errors in a function at the same time (although it has the ability to do). Although LibFuzzer can detect memory errors efficiently, it cannot provide enough information for repair tools to generate patches. For example, even though LibFuzzer inputs *a=3* and triggers both errors, we can not figure out what is the correct conditionals of the patches and where to

³<https://clang.llvm.org/docs/SanitizerCoverage.html>

⁴<https://clang.llvm.org/docs/AddressSanitizer.html>

⁵<https://clang.llvm.org/docs/LeakSanitizer.html>

insert the patches.

2.3 Component-Based Program Synthesis

Given a set of input-output pairs (test suite), component-based program synthesis generates a program that satisfies all the given input-output pairs, which means, if (a, b) is one of the input-output pairs, then the synthesized program must output b when its input is a . In component-based program synthesis, we use a set of basic components (constant, operators, program variables) to synthesize the expected expression. For example, to synthesize a linear expression, $(constant, minus, plus)$ and program variables (x, y) are given as the basic components, so we can synthesize the following expressions.

$x+y, x-y, y-x, x+c, x-c, c-x, y+c, y-c, c-y$

We use the components to synthesize the candidate expression and construct the constraint using the test suite. The constraint is in first-order logic and solved by an SMT solver. If it is satisfiable, the expression can be constructed.

$$\theta = \bigwedge_{i,o \in T} \varphi(\psi(e), i, o) \quad (2.1)$$

Formula 2.1 denotes the formal definition of the CBPS constraint. $\psi(e)$ denotes that the synthesized expression e should be a well-formed expression that satisfied the specification of the operator and i, o is an input-output pair of test suite T . $\varphi(\psi(e), i, o)$ denotes that when e inputted i , the output must be o . Finally, the constraint θ represents that e must be a well-formed expression and given test suite T , e should satisfy all input-output pairs in it.

Chapter 3

Overview

We illustrate key features of HAMER and how it works. Figure 3.1 depicts a high-level overview of the HAMER pipeline.

First, we check the program with the static analyzer Infer, and then we manually collect the functions that contain Infer alarms. Following that, these functions are fuzzed by LibFuzzer to detect true errors. For those functions on the error path, we collect the variables that are dependent on the function arguments and stub them. Dependent variables will be used to synthesize the conditional of the patch. After that, we instrument all of the dependent variables and run the program via LibFuzzer to collect the tests that do or do not trigger the error. Finally, we use a simplified component-based program synthesis (simp-CBPS) to generate patches. Because the quality of the test suite affects the quality of the patch, we use LibFuzzer to check the current fixed code again, and if it has not been fixed or if a new error occurs, we collect the tests that cause the error or insert the patch in a different location. This strategy ensures that the patches generated by HAMER are correct and do not introduce new errors. In the following paragraphs of this section, we will use two motivating examples to demonstrate the workflow and characteristics of HAMER.

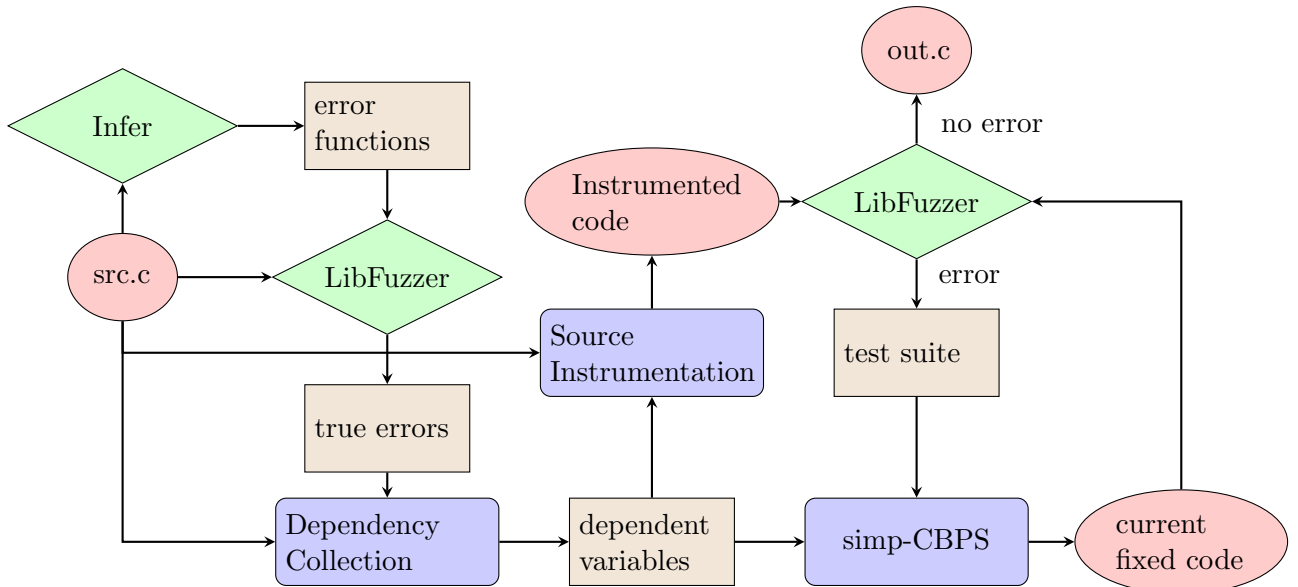


Figure 3.1: HAMER pipeline

```

1  typedef struct N{
2      int v;
3  }node;
4
5  node *new_node1(int a){
6      node *n=(node*) malloc( sizeof( node)
7      );
8      n->v=a;
9      return n;
10 }
11 node *new_node2(int a){
12     node *n=(node*) malloc( sizeof( node)
13     );
14     n->v=a*a;
15     return n;
16 }
17 int func(int a){
18     node* (*p[]) ()={new_node1 ,
19     new_node2 };
20     node *x;
21     node *y=(node*) malloc( sizeof( node)
22     ); //o2
23     x=(*p[0]) (a) ; //o0
24     if (a<5){
25         x=(*p[1]) (a) ; //o1
26     }
27     x->v=10;
28     return 0;
29 }

```

(a) o0, o1, o2 occur memory leak

```

17 int func(int a){
18     node* tmp_o0;
19     node* tmp_o2;
20     int tmp_a = a;
21     node* tmp_o1;
22     node* (*p[]) ()={new_node1 ,
23     new_node2 };
24     node *x;
25     node *y=(node*) malloc( sizeof(
26     node));
27     tmp_o2 = y;
28     x=(*p[0]) (a) ;
29     tmp_o0 = x;
30     if (a<5){
31         x=(*p[1]) (a) ;
32         tmp_o1 = x;
33     }
34     x->v=10;
35     if (tmp_a<=4)free (tmp_o1) ;
36     free (tmp_o2);
37     free (tmp_o0);
38     return 0;
39 }

```

(b) HAMER-generated patch

Figure 3.2: Motivating Example 1: Infer false-negative alarm

```

1  typedef struct N{
2      int* p1;
3      int* p2;
4  }node;
5
6  int func(int a){
7      node x;
8      x.p1=(int*) malloc(4) ; //o0
9      x.p2=(int*) malloc(4) ; //o1
10     free (&x.p1+1);
11     free (x.p1);
12
13     return 1;
14 }

```

(a) No memory leak

```

1  typedef struct N{
2      int* p1;
3      int* p2;
4  }node;
5
6  int func(int a){
7      node x;
8      x.p1=(int*) malloc(4) ; //o0
9      x.p2=(int*) malloc(4) ; //o1
10     free (&x.p1+1);
11     free (x.p1);
12     free (x.p2);
13     return 1;
14 }

```

(b) SAVER-generated patch

Figure 3.3: Motivating Example 2: Infer false-positive alarm

3.1 Motivating Example 1

This buggy code has three error heap objects, denoted by *o0*, *o1*, and *o2*, as shown in Figure 3.2a. First, we use Infer to detect this code, obtaining the following result:

Object allocated at line 20 is unreachable at line 20.

Because static analyzers like Infer have a difficult time resolving issues like indirect calls, they can only detect the memory leak of *o2*. After we received the Infer results, the error function was detected again by LibFuzzer, and all errors were successfully detected. For example, for *o1*, we can get the fuzzing result shown below:

```
in malloc ../a.out
in new_node2 ../src.c:12:18
in func ../src.c:23:7
```

We can get all of the functions on the error path using LibFuzzer. Obviously, the correct fix location could be in any of the functions, so we collect the variables that are dependent on the function argument, and we also collect both the heap object information and the return location of each function during this static analysis.

Table 3.1: Instrumentation result of *o1*

func	new_node2		error
a	a	n->v	
0	0	0	1
5	5	25	0
6	6	36	0
8	8	64	0

Following that, we instrument all of the dependent variables and run the source instrumented code through Libfuzzer to collect dynamic values for each dependent variable. Table 3.1 displays the results of *o1*'s collection. The *error* column indicates whether or not memory leak occurred at the current value, with 1 indicating that it did and 0 indicating that it did not. Table 3.1 shows that *o1* leaks when the variable *a* in the function *func* is less than or equal to 4. We can synthesize the ideal patch using simp-CBPS, which is:

```
if(a<=4)free(o1);
```

Finally, we use LibFuzzer to detect the patched code, and if it fixes the current bug, we keep the patch and fix other bugs until all bugs are fixed or time out. If the current patch does not fix the bug or causes a new bug, we try to insert the patch into another location or collect new tests to synthesize a new patch. For example, we can also synthesize patches like `if(a<=3)free(o1);` using the Table 3.1 results. With Libfuzzer, we can see that when `a=4`, *o1* still occurs memory leak. We will start by trying alternative fix locations, however, because the function *func* only has one return place, we can only collect new tests to synthesize a new patch. We add `(a : 4, error : 1)` to the test suite and then use the synthesizer to generate a new patch. It's self-evident that the improved test suite enabled us to obtain the correct patch. We will utilize temporary variables to save the heap object and variables in the conditional when we apply the patch. This step is necessary to prevent these variables from changing between the allocation location and patch insertion location.

Although it is possible to fix the *o1* memory leak by inserting a patch into function *new_node2*, the use-after-free problem will occur if the memory is freed too early due to the use of `x->v` at line 25 of the function *func*.

3.2 Motivating Example 2

We briefly discussed how HAMER uses fuzzing (LibFuzzer) to detect and fix errors that are not noticed by the static analyzer (Infer) in Motivating Example 1. Similarly, HAMER can avoid attempting to resolve false alarms. In Motivating Example 2, line 10 frees *o1* in some other way, but Infer misses it, so it assumes *o1* occurs memory leak. SAVER [11] also ignores the fact that this is a false alarm and generates the incorrect patch, resulting in *double-free*. HAMER utilizes LibFuzzer to dynamically detect code, and LibFuzzer does not report issues for Motivating Example 2, therefore HAMER saves time trying to generate a fix for the wrong alarm. However, depending on the complexity of the function argument and the fuzzer’s mutation strategy, the fuzzer may fail to visit the error path.

Chapter 4

Approach

In this section, we describe our approach in detail, explaining what technical issues arise and how we address them. There are three major issues to consider:

- It is difficult for LibFuzzer to directly collect the high-quality test suite. How can HAMER synthesize the correct conditional?
- How does HAMER choose the correct fix location(s) when a function has multiple return places?
- How does HAMER deal with several memory errors in a single function?

HAMER will solve these issues by using LibFuzzer to constantly check the patched code. In section 4.5, we will go over our repair algorithm in detail. Until then, we will describe how HAMER gathers the data required to fix the errors.

4.1 Patch Template

The purpose of this research is to fix temporal memory errors such as memory leaks, not spatial memory errors such as buffer overflow. The most common solution for temporal memory errors is to free the memory at the correct location. As a result, the following patch template can fix most of the temporal memory errors.

```
if(cond)free(ob);
```

We can free the specified heap object under the specific condition by inserting a conditional deallocator. Hence, in order to generate the correct patch, we should know the following three details:

- (1) conditional *cond*
- (2) error heap object *ob*
- (3) fix location

We will present in detail how we generated the correct patch in the following subsection.

4.2 Error Detection

HAMER uses a static analyzer to find functions that may have memory errors and then uses fuzzer to detect the real errors in those functions. The purpose of applying the static analyzer is to improve HAMER's scalability, as the fuzzer always takes a long time to detect errors. Using the static analyzer to pick out candidate functions prevents fuzzer

from wasting time in locations where it is not necessary. But obviously, HAMER will not be able to fix errors in functions that are not provided by the static analyzer. So for shortcode, we can just simply use fuzzer to detect errors. After error detection, we can obtain a set E that contains the details of each error reported by fuzzer, similar to the report shown in Motivating Example 1.

4.3 Dependency Collection

We collect the error paths and error heap objects for each element in the set E separately after obtaining the error report E . We can organize the information of each error from the error report to produce the set EP using the error paths and coordinates provided by LibFuzzer.

$$EP = \{get_path(e) | e \in E\} \quad (4.1)$$

In Motivating Example 1 (Figure 3.2), LibFuzzer gives the error report of $o1$, and we can obtain the error path of $o1$:

```
{funcname: func, coord: src.c:23:7, next:
 {funcname: new_node2, coord: src.c:12:18, next:
  {funcname: malloc}}}
```

It is worthy to note that the *coord* of each function is the error location inside the function, not the function’s coordinate. All error paths, obviously, will end with an allocation function, and the coordinates of the allocation function will be kept in the previous node’s *coord*, which we will use to localize the error heap object.

After organizing the error report, we collect the dependent variables in each function on the error path, which are utilized to synthesize the conditional. Fuzzer tries to trigger the function’s error by inputting different data, so all the variables in the function that are dependent on the argument could be utilized to synthesize the patch’s conditional. These dependent variables are collected using *def-use* chain, and their names, types, and coordinates are saved. During this static analysis, we also gather the return location of each function, as well as the name and type of the error heap object.

As we present in the Background, fixing the temporal memory error requires inserting the deallocation at the correct location. A memory leak will occur if allocated memory space is not freed at the end of the program, hence it is critical to free allocated memory before it is unreachable. Most of the existing memory error repair techniques [11, 16] are based on static analysis that collects heap-related behavior, so they can insert the patch at the right location. However, the tradeoff is that it requires a high overhead, and static analysis is tough to deal with some problems such as indirect call, which may make repair tools fail to generate a patch or generate a wrong patch. HAMER aims to collect the essential information for fixing errors via lightweight static analysis. There are three major reasons why allocated memory cannot be accessed: (1) it is freed, (2) no pointer points to it, and (3) the current function exits. Memory leaks will not occur if it is properly freed. For the second reason, we will use a temporary variable to save the pointer to the error memory when it is allocated, as we will explain in detail in subsection 4.5. Hence, we simply need to think about the third reason. A function may exit in two ways: by *return* or by exiting automatically at the end of the function. As a result, we just need to collect all of each function’s *return* locations, as well as the location of the function tail, and use them as the candidate fixing location.

We extract the name and type of error heap objects at the location detected by the fuzzer. Its type is mainly determined by three factors: (1) declaration type, (2) casting

type, and (3) type in the *sizeof*. For instance, the following code demonstrates how developers use *malloc* for memory allocation:

```
int* p=(int*)malloc(sizeof(int));
```

We only need one of these to determine the type of the heap object, but there are indeed cases where all three are not written, such as when type is defined in a structure and then cast and *sizeof* are not used:

```
x->v=malloc (4);
```

In such cases, HAMER is unable to obtain their type, however, this is not a weakness of our approach. HAMER employs lightweight static analysis that can handle most cases. If we want to achieve higher performance, we can employ a more precise type analysis, but the tradeoff is higher overhead.

Finally, we formalized the dependency information *Dep* as follows:

$$Dep = \{(get_dep(path), get_ret(path), get_ob(path)) \mid path \in EP\} \quad (4.2)$$

4.4 Source Instrumentation

We apply source instrumentation on the dependent variables after obtaining them in order to collect their dynamic values and build a test suite. A common test suite consists of several *input* – *output* pairs; however, because the goal of our technique is to fix memory leaks, each *input* corresponds to an *output* that indicates whether or not the *input* will trigger the error; if it does, the output is 1, otherwise, it is 0. For example, our instrumented code for the variable *a* of the *func* function in Figure 3.2a is as below:

```
fprintf(stderr,"instrument:(line:17) a:%d\n",a);
```

We record the locations of the variables to identify their values because the same variables will have different values at different locations. Because the results of LibFuzzer are sent to *stderr*, we also send the results of the instrumentation to *stderr* and combine them with the results of LibFuzzer.

Our approach’s overall concept of source instrumentation and dynamic values collection is shown in Algorithm 1. We create a new set *SynInf* to contain the information about fixing each error. We first define a set *dep.var* to store all the dependent variables in *Dep* and for different errors, we save the dependent variables *Var*, return location *Ret*, and error heap object information *ob* to *SynInf*. To get enough data, we use LibFuzzer to run the instrumented code 10 times and collect data for different errors. If the current execution results trigger an error, we set the *output* of the triggered error’s dependent variable to 1, otherwise, we set it to 0.

4.5 Patch Generation

In this section, we will present our repair algorithm and demonstrate how our repair algorithm solves the major issues we mentioned before.

4.5.1 simp-CBPS

Since we only need to synthesize the conditional of deallocation, we use a simplified version of component-based program synthesis [12] (simp-CBPS). In simp-CBPS, a component is a variable, a constant, or an operator. simp-CBPS uses these user-given components to generate code that satisfies the test suite.

For example, we use the following components to synthesize code that satisfies the test suite in Table 4.1.

Algorithm 1 Source Instrumentation Result Collection

Input: src, Dep **Output:** $SynInf$

```
1:  $dep\_var \leftarrow \emptyset$ 
2:  $SynInf \leftarrow \emptyset$ 
3:  $err\_id \leftarrow 0$ 
4: for  $(Var, Ret, ob) \in Dep$  do
5:    $dep\_var \leftarrow dep\_var \cup Var$ 
6:    $SynInf[err\_id].add((Var, Ret, ob))$ 
7:    $err\_id \leftarrow err\_id + 1$ 
8: end for
9:  $inst\_code \leftarrow \text{Instrument}(src, dep\_var)$ 
10:  $i \leftarrow 0$ 
11: while  $i < 10$  do
12:    $res \leftarrow \text{Fuzz}(inst\_code)$ 
13:   for  $err\_id = 0$  to  $|Dep| - 1$  do
14:     for  $var \in dep\_var$  do
15:       if memory leak happen then
16:          $SynInf[err\_id][var].add((res[var], 1))$ 
17:       else
18:          $SynInf[err\_id][var].add((res[var], 0))$ 
19:       end if
20:     end for
21:   end for
22:    $i \leftarrow i + 1$ 
23: end while
```

variable: x constant: c operator: $*_1 < *_2$

Since $<$ is a binary operator, we can construct the expression $x < c$ and $c < x$ using the variable x and the constant c . Then we assign the value from Table 4.1 to get the logical formula below:

$$\begin{aligned} x < c : & (4 < c) \wedge (5 < c) \wedge \neg(6 < c) \wedge \neg(7 < c) \\ c < x : & (c < 4) \wedge (c < 5) \wedge \neg(c < 6) \wedge \neg(c < 7) \end{aligned}$$

We have turned the program synthesis problem into *Satisfiability Modulo Theories* (SMT) by doing the above action. We solve the logical formula via the SMT solver. If a logical formula is unsatisfiable, it means that the present synthesized expression does not pass the test suite, indicating that it is not the expected expression. For example, the second logical formula is unsatisfiable, hence $c < x$ is not the correct expression. The first logical formula is satisfiable and the result is $c=6$, so we can get the expected expression $x < 6$.

Table 4.1: Test suite

x	output
4	True
5	True
6	False
7	False

The quality of the test suite is the most critical part of using CBPS to synthesize expressions. CBPS will synthesize incorrect expressions if the test suite provided by the user is of poor quality. If Table 4.1 does not have $(x = 6, output = False)$, for example, we might obtain $x < 7$. Obviously, CBPS can only synthesize numeric expressions.

4.5.2 Repair Algorithm

The input of our repair algorithm (Algorithm 2) is the source code and the collected information *SynInf*, and the output is the fixed code generated by HAMER. We use a *queue* to keep the number of each error and repair them one by one. When repairing, we first pop an error number from the *queue* and then get the corresponding information from *SynInf*. After that, we synthesize the conditional of the patch using simp-CBPS. Because all the functions on the error path have the potential to become the fix location, simp-CBPS synthesizes all satisfiable patches, and the fix location and heap object information corresponding to the patch are gathered and saved to *cur_patches*.

After that, we synthesize the patch and use the function *Fix* (Algorithm 3) to insert it in the right location. If the currently generated patch fails to fix the error, we gather the test that triggers the error and append it to the test suite. For example, when HAMER fixes the memory leak of *o1* in Figure 3.2a of Motivating Example1, it's difficult to collect enough tests directly, thus the next two incorrect patches could be synthesized:

- (1) `if (a<=3)free(o1);`
- (2) `if (a<=5)free(o1);`

If simp-CBPS does not get the test $(a : 4, error : 1)$, it will synthesize the conditional of the first wrong patch, and if it does not get the test $(a : 5, error : 0)$, it will synthesize the conditional of the second wrong patch. For the first patch, we can get that memory leak will happen when $a=4$ via LibFuzzer, so we add $(a : 4, error : 1)$ to the test suite. For the second patch, when $a=5$, since *o1* is not defined yet, free it is undefined behavior, so we add $(a : 5, error : 0)$ to the test suite. Similarly, if a test triggers a double free, we will add this test and $(error : 0)$ to the test suite, indicating that this heap object has already been deallocated at this test.

We keep the patch of the current error if this error is fixed, and we clear the test suite in *SynInf* if there are still have errors to be fixed. The goal of this step is to save space and speed up the SMT solver computation. *SynInf* takes up a lot of space because the test suite is constantly updated. The quality of the gathered test suite will also be affected if a function has several errors. We can recollect the test suite when an error is fixed to receive a higher-quality test suite and save space. If the current error is not fixed, we add the error number to the queue and try to fix it later.

To ensure that our algorithm terminates at the right time, we use two methods. The first occurs in line 9. If simp-CBPS synthesizes the same patch as last time, indicating that this error is difficult to fix in the current situation. In most cases, we are unable to collect a high-quality test suite due to multiple errors in the function, so we give up trying to fix this error for the time being and return to it after fixing other, much simpler errors. The second is that we create an *unfixederror* variable to keep track of errors that we try to fix but can not. Because there may sometimes be several errors that HAMER cannot solve, we stop HAMER if none of the remaining errors are fixed after HAMER tries to fix them.

4.5.3 Function Fix

It is critical to insert the patch in the correct location in order to fix temporal memory errors. Function *Fix* to determine if the fix location is correct. We test the patch and the fix location of each function on the error path one by one. Fuzzer checks a patch when it is

Algorithm 2 Repair Algorithm

Input: *src*, *SynInf***Output:** *fixed_code*

```
1: fixed_code  $\leftarrow$  src
2: queue  $\leftarrow$   $\{0, 1, \dots, |SynInf| - 1\}$ 
3: unfixederror = 0
4: while  $|queue| \neq 0$  and unfixederror  $\neq |queue|$  do
5:   err_id  $\leftarrow$  queue.pop
6:   err_inf  $\leftarrow$  SynInf[err_id]
7:   repeat
8:     cur_patches  $\leftarrow$  simp-CBPS(err_inf)
9:     if cur_patches same as last time then
10:      break
11:    end if
12:    cur_code  $\leftarrow$  Fix(cur_patches, fixed_code)
13:    if not fixed then
14:      err_inf  $\leftarrow$  Update(err_inf)
15:    end if
16:  until timeout or error fixed
17:  if current error fixed then
18:    fixed_code  $\leftarrow$  cur_code
19:    unfixederror  $\leftarrow$  0
20:    if queue is not empty then
21:      SynInf  $\leftarrow$  Clean(SynInf)
22:      SynInf  $\leftarrow$  Update(SynInf)
23:    end if
24:  else
25:    queue.add(err_id)
26:    unfixederror  $\leftarrow$  unfixederror + 1
27:  end if
28: end while
```

inserted into a fix location. If the patch introduces a new error in the existing fix location, it is clear that the fix location is incorrect, and we should try a different fix location. If no new errors are raised but the existing error remains unfixed, it is possible that the current fix location only partially fixed the error; in that case, we save the patch inserted at the current fix location and try other fix locations. The following code, for example, has two returns, and we must apply the patch to both of them to fix the error.

```
p=malloc(1);
if(c){
  use(p);
  return 0;
}else{
  use(p);
  return 1;
}
```

The memory leak was not fixed when we inserted the patch before the first *return*, so we kept the current patch and inserted the patch before the second *return*, and then the error was correctly fixed.

Algorithm 3 Function Fix

Input: *cur_patches*, *code***Output:** *fixed_code*

```
1: test_code  $\leftarrow$  code
2: for (cur_patch, Ret)  $\in$  cur_patches do
3:   for retloc  $\in$  Ret do
4:     test_code  $\leftarrow$  InsertPatch((cur_patch, retloc))
5:     Fuzz(test_code)
6:     if new error occurs then
7:       test_code  $\leftarrow$  code
8:       continue
9:     else if same error occurs then
10:      continue
11:    else
12:      return test_code
13:    end if
14:  end for
15: end for
16: return  $\emptyset$ 
```

4.5.4 Temporary Variable

We use temporary variables to store the variables used in the patch conditional and the error heap object while inserting patches. Because the values of these significant variables may modify between the error source and the fix location, analyzing these changes requires complex static analysis. To address this issue without increasing the complexity of our algorithm, we use temporary variables to store the values of these variables, so that even if their values changed, the patches generated by HAMER are still correct (e.g., Figure 3.2b).

Chapter 5

Evaluation

We evaluate the effectiveness and efficiency of HAMER and answer the following research questions.

RQ1 Compared with state-of-the-art automated memory error repair tool SAVER, what is the overall effectiveness of HAMER?

RQ2 Can HAMER address the three major issues we mentioned in Chapter 4?

RQ3 How efficient is HAMER in using lightweight static analysis?

5.1 Implementation

We have implemented our approach in a tool named ExtractFix, whose pipeline is shown in Figure 3.1. HAMER detects the memory leak in the program using existing vulnerability detection tools. To detect the real errors, HAMER first uses a static analyzer to detect the program, then uses a fuzzer to detect static analyzer alarms. The main role of the static analyzer is to filter out the problematic functions in the program so that the fuzzer does not waste time on other parts. We directly use a fuzzer to verify the code in our experiments because we utilize short code examples to test the efficiency of HAMER. We use LibFuzzer to perform the fuzz testing. Since LibFuzzer may not detect all of the errors, we fuzz code 10 times each time. If LibFuzzer does not detect an error within 5 seconds for one execution, the code is judged no error (or fixed). The input data format is set to an integer of two digits or less. HAMER is written in 1000 lines of python3 code. For syntactic analysis of *c*, we use the pycparser module ¹. In our implementation of simp-CBPS, we use the Z3 [6] python module ² as the SMT solver and implemented six operators (`!=`, `==`, `<`, `<=`, `>`, `>=`) as components.

5.2 Experimental Setup

To analyze HAMER, we synthesized 11 codes. Because HAMER’s scalability is currently limited, and memory leaks in real-world projects can not properly reflect HAMER’s properties, we gave up utilizing real-world projects to evaluate HAMER. These codes were synthesized from six different directions, as shown in Table 5.1. To fuzz the code, LibFuzzer requires an entry point, which we manually added to the code as shown below.

¹<https://github.com/eliben/pycparser>

²<https://github.com/Z3Prover/z3>

Table 5.1: Characteristics of synthesizing code. FN and FP denote that the code contains Infer’s false-negative and false-positive alarms. EP denotes that the code contains the error with a long error path. ME denotes that the code contains multiple error heap objects. MR denotes that the code has multiple returns. CF denotes that the code has a complicated control flow.

Syn	LoC	Feature
test1	12	CF
test2	9	MR
test3	23	EP, FN
test4	11	FP, FN
test5	15	EP, ME, FN
test6	24	EP, ME, MR, FN
test7	9	MR, CF, FN
test8	26	EP, ME, FN
test9	24	EP, ME, FN
test10	15	ME, CF
test11	15	EP, CF, FP

```

int LLVMFuzzerTestOneInput(char *data, int size) {
    ConvertCharToInt(data);
    EntryPoint(data);
    return 0;
}

```

HAMER’s static analysis and fuzzing time were also observed during the experiment. Since the repair process of HAMER was not absolutely the same every time (depending on the test generated by LibFuzzer), we run HAMER ten times for each code and then calculate the average repair time. We installed SAVER directly from their github ³ (Infer version: v0.15.0-821a8db). Because SAVER relies on Infer’s static analysis results, we must consider Infer’s execution time when calculating SAVER’s execution time. Table 5.2 shows the final experimental results.

5.3 Experimental Results

5.3.1 Compared with state-of-the-art automated memory error repair tool SAVER, what is the overall effectiveness of HAMER?

There are 23 memory leaks in the 11 test codes we synthesized. Infer found 7 memory leaks successfully, missed 16 memory leaks, and reported 2 false alarms. Because SAVER relies on infer’s alarms to repair the code, it has no opportunity to fix the 16 unreported errors (false-negative). SAVER generated 4 correct patches and 2 wrong patches for the 7 alarms correctly reported by Infer, and it failed to identify Infer’s false alarm and generated 2 wrong patches.

All 23 memory leaks were detected via LibFuzzer. For test8, LibFuzzer may not be able to detect all of the errors, also making it difficult to verify whether the patch generated by HAMER is correct or not. In **RQ2**, we will go over the details of test8. HAMER can generate all correct patches for other test codes and does not provide any erroneous patches. HAMER’s average repair time is 9.10 seconds, although it is clear that the majority of that time is spent by LibFuzzer checking whether the patches are correct. We

³https://github.com/kupl/SAVER_public

Table 5.2: Evaluation result of HAMER and SAVER. ML denotes the number of memory leaks. T, FP, and FN denote the number of true, false-positive, and false-negative alarms detected by Infer. Fuzzer denotes the number of memory leaks detected by LibFuzzer. ✓ and × report the correct and wrong patches generated by HAMER and SAVER. SA, Fuzz, and Total report the static analysis time, fuzzing time, and the total fix time.

Syn	ML	Infer			SAVER			HAMER			
		T	FP/FN	sec	✓	×	sec	Fuzzer	✓	×	SA/Fuzz/Total
test1	1	1	0/0	0.68	0	0	0.05	1	1	0	0.01/7.01/7.94
test2	2	2	0/0	0.13	2	0	0.07	2	2	0	0.01/7.27/8.42
test3	2	0	0/2	0.36	-	-	-	2	2	0	0.04/7.71/9.14
test4	1	0	1/1	0.15	0	1	0.03	1	1	0	0.01/7.06/7.90
test5	2	0	0/2	0.14	-	-	-	2	2	0	0.03/7.71/9.10
test6	3	0	0/3	0.28	-	-	-	3	3	0	0.03/8.03/9.60
test7	2	1	0/1	0.14	0	1	0.05	2	2	0	0.01/7.43/8.61
test8	4	0	0/4	0.49	-	-	-	3-4	0-3	0-1	0.05/4.17/7.17
								4	4	0	0.05/9.94/13.17
test9	3	1	0/2	0.42	1	0	0.05	3	3	0	0.04/8.91/11.29
test10	2	2	0/0	0.42	1	1	0.07	2	2	0	0.02/7.67/9.04
test11	1	0	1/0	0.14	0	1	0.05	1	1	0	0.02/7.02/7.89
Total/Ave	23	7	2/16	0.32	4	4	0.05	22-23	19-23	0-1	0.03/7.49/9.10

also set the fuzzing timeout to 5 seconds, which implies that any properly repaired test codes will take 5 seconds to confirm whether or not the patch is correct via LibFuzzer. We have already discussed the issues of indirect call and alias in Chapter 3, so we will not go over it again here.

Infer failed to detect two memory leaks in test5 (shown in Figure 5.1) because it cannot effectively deal with the alias. It made SAVER have no chance to fix errors (although this is easy for SAVER). Similar problems appear in test3,4,6,8,9 when memory is dynamically allocated at another function and then returned to the current function’s pointer. LibFuzzer, on the other hand, tracks the memory situation via AddressSanitizer, which instruments all dynamic memory allocation, allowing it to quickly detect any memory leaks in codes with simple control flow. Because *o1* uses *o0*, if *o0* is freed before *o1*, use-after-free will occur. HAMER stores error heap objects via temporary variables, so we do not need to worry about the order of deallocation.

In test10 (shown in Figure 5.2), although Infer detected two memory leaks, SAVER generated a correct patch and a wrong patch. SAVER can not fix multiple errors at the same time. If only fix *o0* without considering *o1*, the patch generated by SAVER at line10 is correct, but if the patch of *o1* at line11 is added, double free will occur. HAMER can fix multiple errors at the same time and check whether the generated patches will cause new errors, so it can fix test10 correctly

In test11 (shown in Figure 5.3), *o0* allocated, deallocated, and leaked at 3 different functions. The error heap object was identified by Infer, but the leak location was incorrectly analyzed, and Infer reported the following alarm.

Object allocated at line 12 is unreachable at line 13.

Since SAVER is unable to follow heap-related behavior across multiple functions, it generated the erroneous patch, as shown in Figure 5.3b. HAMER quickly detected the error heap object using LibFuzzer, collected the key tests, and successfully generated the conditional deallocation statement patch (Figure 5.3c).


```

1 typedef struct N{
2     struct N *next;
3     int v;
4 }node;
5
6 node *new_node(int a){
7     node *n=(node*) malloc( sizeof(node));
8     n->next=NULL;
9     n->v=a;
10    return n;
11 }
12
13 int func(int a){
14     node *x=new_node(a); //o0
15     x->next=new_node(a+1); //o1
16     return 0;
17 }

```

(a) o0, o1 occur memory leak

```

13 int func(int a){
14     node* tmp_o0;
15     node* tmp_o1;
16     node *x=new_node(a);
17     tmp_o0 = x;
18     x->next=new_node(a+1);
19     tmp_o1 = x->next;
20     free(tmp_o1);
21     free(tmp_o0);
22     return 0;
23 }

```

(b) HAMER-generated patch

Figure 5.1: test5

HAMER has a higher repairability than SAVER. HAMER has more opportunities to repair more memory leaks, as well as better handling of issues like multiple errors, indirect calls, and alias, and is less likely to generate erroneous patches.

5.3.2 Can HAMER address the three major issues we mentioned in Chapter 4?

To synthesize correct conditional, HAMER continuously updates the test suite until the correct patch is generated or timeout. For *o1* in test10, sometimes LibFuzzer can not directly provide a significant test (e.g., ($a : 5, error : True$)), causing HAMER to generate the wrong conditional $a \geq 4$. However, HAMER checked the patch via LibFuzzer and found that the memory leak of *o1* still existed when $a=5$, so HAMER added this test to the test suite and generated the correct patch at last.

HAMER tries to insert the patch to each candidate fix location and verifies the patch via LibFuzzer. Test2, 6, 7 all have multiple returns, and HAMER successfully fixed them all, so HAMER can deal with the case of multiple returns in a function.

HAMER will repair each error one by one via HAMER's repair algorithm shown in Chapter 4. HAMER is able to reach 100% repairability for the function which has two error heap objects (e.g., tests 5, 6, 9, 10). Due to the feature of LibFuzzer, LibFuzzer stops and outputs results when it detects the error on the current explored path, so if multiple error heap objects exist on multiple different paths, LibFuzzer will be unable to detect all the errors at the same time, causing HAMER failed to generate patches. However, we can solve this problem by improving our repair algorithm, which we will discuss in the Future Work in Chapter 7.

Figure 5.4 shows the correct and incorrect patches generated by HAMER for test8. Test8 has four error heap objects (*o0*, *o1*, *o2*, *o3*). In most cases, LibFuzzer can detect all of the errors and give the necessary tests, enabling HAMER to successfully fix test8 as shown in Figure 5.4b. However, due to a large number of errors, LibFuzzer may not be able to detect all of them. For example, test8 has two paths($a < 5$, $a \geq 5$), as shown in Figure 5.4c, and executing either path will trigger the memory leaks of *o0*, *o1*, *o3*. Although we initially set LibFuzzer to run 10 times in order to detect all errors, there were sometimes that the produced tests were all ≥ 5 , so that the path of $a < 5$ has never

```

1 typedef struct N{
2     int* p1;
3     int* p2;
4 }node;
5
6 int func(int a){
7     node x;
8     x.p1=(int*) malloc(4); //o0
9     int* n;
10    if(a<5){
11        n=x.p1;
12    } else {
13        n=malloc(4); //o1
14    }
15    return 1;
16 }

```

(a) o0, o1 occur memory leak

```

6 int func(int a){
7     node x;
8     x.p1=(int*) malloc(4);
9     int* n;
10    if(a<5){
11        n=x.p1;
12    } else {
13        n=malloc(4);
14    }
15    free(x.p1);
16    free(n);
17    return 1;
18 }

```

(b) SAVER-generated patch

```

6 int func(int a){
7     int* tmp_o0;
8     int tmp_a = a;
9     int* tmp_o1;
10    node x;
11    x.p1=(int*) malloc(4);
12    tmp_o0 = x.p1;
13    int* n;
14    if(a<5){
15        n=x.p1;
16    } else {
17        n=malloc(4);
18        tmp_o1 = n;
19    }
20    if(tmp_a>=5) free(tmp_o1);
21    free(tmp_o0);
22    return 1;
23 }

```

(c) HAMER-generated patch

Figure 5.2: test10

been entered and the memory leak of *o2* has never been triggered. Although HAMER could generate the correct patch for *o0*, *o1*, *o3*, when checked by LibFuzzer, each patch will cause the memory leak of *o2* and HAMER will determine that the current patch is not correct since it will cause a new error.

The instance of HAMER generating an error patch is also shown in Figure 5.4d. In beginning, HAMER generated the right patches of *o0* and *o1*. HAMER synthesized conditional $a \geq 2$ with the current test suite when trying to fix *o2*. Then, using LibFuzzer, we can determine that the current patch was erroneous, and for example, we can collect the test of $(a : 4, error : True)$, and synthesize conditional $a \geq 4$. However, because *o3* has not been repaired and whatever was input memory leak for *o3* will be triggered, LibFuzzer failed to input the critical test $a=5$ to trigger *o2*, causing HAMER to incorrectly judge this patch as the correct one and keep it, even LibFuzzer executed 10 times. When fixing *o3*, since *o3* was fixed correctly, LibFuzzer explored all paths and triggered the memory leak of *o2*, causing HAMER to incorrectly judge the patch of *o3* as the wrong patch. Finally, HAMER generated an incorrect patch and deleted a correct patch.

```

1  int* f() {
2      int *a=(int*) malloc ( sizeof (int));
3      return a;
4  }
5
6  void g(int*p){
7      *p=1;
8      free(p);
9  }
10
11 int func(int a){
12     int *p=f(); //o0
13     if(a>5){
14         g(p);
15     }
16     return 0;
17 }

```

(a) o0 occurs memory leak

```

11 int func(int a){
12     int *p=f();
13     free(p);
14     if(a>5){
15         g(p);
16     }
17     return 0;
18 }

```

(b) SAVER-generated patch

```

11 int func(int a){
12     int tmp_a = a;
13     int* tmp_o0;
14     int *p=f();
15     tmp_o0 = p;
16     if(a>5){
17         g(p);
18     }
19     if(tmp_a<=5)free(tmp_o0);
20     return 0;
21 }

```

(c) HAMER-generated patch

Figure 5.3: test11

HAMER can handle multiple returns, multiple errors, and collect critical tests, although performance will be affected when the number of errors increases.

5.3.3 How efficient is HAMER in using lightweight static analysis?

The repair times for HAMER and SAVER are shown in Table 5.2. HAMER is a memory error repair tool based on hybrid analysis, which collects information for repair using lightweight static analysis. SAVER is a static memory error repair tool that uses static analysis from detection to repair. SAVER's repair time is quite short, but it can not guarantee the correctness of the patches it generated, and its repairability is not as good as HAMER's. HAMER's repair time is longer, but it is obvious that most of the time is spent on fuzzing, and due to our settings, HAMER spends 5 seconds at the end of the repair checking the correctness of the patches using LibFuzzer. The lightweight static analysis utilized by HAMER can collect the essential information to assist HAMER in generating the correct patches in a very short time. The HAMER repair algorithm does not require time to collect a large amount of heap-related behavior through static analysis, and experimental results also show that HAMER has not failed to repair errors due to our lightweight static analysis missing essential information.

Our lightweight static analysis can collect enough information in a short time to assist HAMER to generate the correct patches.

```

1  typedef struct N{
2      struct M *m;
3      int v;
4  }node;
5
6  typedef struct M{
7      struct M *next;
8      int k;
9  }field;
10
11 node *new_node(int a){
12     node *n=(node*) malloc(sizeof(node));
13     n->m=(field*) malloc(sizeof(field));
14     n->m->next=NULL;
15     return n;
16 }
17
18 int func(int a){
19     node *x=new_node(a); //o0, o2
20     node *y=new_node(a); //o1, o3
21     x->v=a;
22     x->m->k=a+1;
23     y->v=a;
24     y->m->k=a+1;
25     if(a<5){
26         free(x);
27     }
28     return 0;
29 }

```

(a) o0, o1, o2, o3 occur memory leak

```

18 int func(int a){
19     node *x=new_node(a);
20     node *y=new_node(a);
21     x->v=a;
22     x->m->k=a+1;
23     y->v=a;
24     y->m->k=a+1;
25     if(a<5){
26         free(x);
27     }
28     return 0;
29 }

```

(c) LibFuzzer only detect 3 errors at first, so HAMER failed to generate patch

```

18 int func(int a){
19     field* tmp_o0;
20     node* tmp_o1;
21     int tmp_a = a;
22     node* tmp_o2;
23     field* tmp_o3;
24     node *x=new_node(a);
25     tmp_o0 = x->m; tmp_o2 = x;
26     node *y=new_node(a);
27     tmp_o1 = y; tmp_o3 = y->m;
28     x->v=a;
29     x->m->k=a+1;
30     y->v=a;
31     y->m->k=a+1;
32     if(a<5){
33         free(x);
34     }
35     free(tmp_o3);
36     if(tmp_a>=5) free(tmp_o2);
37     free(tmp_o1);
38     free(tmp_o0);
39     return 0;
40 }

```

(b) HAMER-generated correct patch

```

18 int func(int a){
19     field* tmp_o0;
20     node* tmp_o1;
21     int tmp_a = a;
22     node* tmp_o2;
23     node *x=new_node(a);
24     tmp_o0 = x->m; tmp_o2 = x;
25     node *y=new_node(a);
26     tmp_o1 = y;
27     x->v=a;
28     x->m->k=a+1;
29     y->v=a;
30     y->m->k=a+1;
31     if(a<5){
32         free(x);
33     }
34     if(tmp_a>=4) free(tmp_o2);
35     free(tmp_o1);
36     free(tmp_o0);
37     return 0;
38 }

```

(d) LibFuzzer failed to verify the wrong patch, make HAMER generate wrong patch and failed to generate the patch of o3

Figure 5.4: test8

Chapter 6

Related Work

Existing repair techniques can be categorized as general-purpose or special-purpose, depending on whether they are intended to fix all types of errors or only specific types of errors. HAMER is a special-purpose technique that focuses on fixing temporal memory errors such as memory leaks. Other specialized techniques focus on buffer overflows [25], null dereferences [1, 7], etc.. We compared the features of several related techniques shown in Table 6.1.

There are many studies focused on fixing memory errors [2, 5, 8, 10, 11, 22, 27–30, 32, 33]. Among this, the most directly related to our research is MemFix [16] and SAVER [11]. MemFix starts by removing all deallocation statements from the program, then uses static analysis to trace and collect all heap objects and their candidate fix locations, and finally finds a set of patches that can free all dynamically allocated memory without introducing memory errors. However, MemFix cannot generate conditional deallocation statements and has limited scalability and repairability. SAVER is the state-of-the-art memory error repair technique, with high scalability and repairability. SAVER first uses Infer to detect the memory error and then collects the heap-related behavior of the program through static analysis and constructs the object flow graph. SAVER’s repair algorithm converts the memory error repair problem into the relabeling problem of the object flow graph and can avoid generating wrong patches by considering the relationship between heap objects. However, due to the limitations of static analysis, SAVER cannot understand the dynamic information of the program, which makes it fail to fix some errors or generate erroneous patches. HAMER uses fuzzer to detect programs, allowing HAMER has more opportunities to fix errors (static analyzer’s false-negative) and do not spend time on wrong alarms (static analyzer’s false-positive). HAMER also uses fuzzer to check generated patches, preventing HAMER from generating wrong patches. HAMER’s extensive repair strategy allows it to handle various kinds of memory errors. DEF_LEAK [33] is a dynamic approach to detecting, eliminating, and fixing memory leaks. DEF_LEAK first instruments the program for collecting memory information and then exposes memory leaks through dynamic symbolic execution. DEF_LEAK can generate the deallocation statement patch to eliminate the leaks at run time. However, DEF_LEAK’s repair strategy is very simple (just add deallocation statement at suitable location) and can not fix complicated memory errors, so it does not have high repairability. HAMER uses a conditional deallocation statement to fix the error, so it can fix a leak in a typical path. HAMER can also deallocate a heap object at multiple leak locations.

General-purpose program repair techniques can be classified as semantics-based techniques [9, 14, 19–21, 24, 31] and search-based techniques [13, 15, 17, 18, 26]. Search-based techniques search for the correct patch in a certain search space, so although it has high scalability, the quality of the generated patches is often poor and the overhead is high. Getafix [1] fixes errors by learning from developers’ past fixing history. Getafix has high

Table 6.1: Related works

Target	Technique	Fix location	Test suite	Patch	Verification
General purpose	Angelix	statistical	necessary	static	manual
	CPR	user input	unnecessary	dynamic	manual
	ExtractFix	static	crash input	static	manual
	Getafix	static	training data	static	manual
Memory error	SAVER	static	unnecessary	static	static
	DEF_LEAK	dynamic	unnecessary	static	manual
	HAMER	hybrid	unnecessary	hybrid	dynamic

scalability and repairability for fixing errors with simple repair patterns such as null dereference. Although fixing memory errors often only requires inserting a deallocation statement, the fix location is usually too far away from the error source, which Getafix can not handle, and Getafix is a static technique, so it cannot guarantee the correctness of the generated patch. Semantics-based techniques collect path constraints through symbolic execution and then synthesize patches through CBPS. Angelix [20] has high scalability and can fix multi-location bugs, but like other semantics-based techniques, Angelix is prone to generating overfitting patches, therefore its repairability entirely depends on the quality of the test suite. HAMER can collect the tests triggered errors generated by the fuzzer, which are essential to synthesize the correct patches. CPR [24] collects essential tests for patch synthesis by exploring input and patch space through concolic execution. Extract-Fix [9] improves the performance of CPBS by collecting the sanitizer’s constraint and the tests triggered sanitizer. ExtractFix also uses the crash location as the starting point and performs backward control and data-dependency analysis along with crashing path to collect candidate fix locations, so it has the ability to fix the error kinds which have the different crash and fix locations. The main reason these techniques do not perform well in fixing memory errors is that they cannot identify the correct fix location and it is difficult for them to guarantee the correctness of the patches they generated. HAMER collects essential tests via fuzzer, so it does not need the test suite at the beginning. HAMER collects the candidate fix location via lightweight static analysis and uses a fuzzer to find the correct fix location and verify the generated patches. HAMER is a memory error repair technique that can be fully automated among the processes of detection, repair, and verification.

Chapter 7

Conclusion

7.1 Conclusion

Fixing memory errors is difficult because new and more dangerous errors are easily introduced. In this research, we presented a new automated memory error repair technique based on hybrid analysis. HAMER dynamically detects memory errors and verifies generated patches via fuzzer and collects essential information for repair via lightweight static analysis. Experiments show that HAMER can fix the complicated memory errors effectively, which the existing static memory error repair technique can not repair. In the future, HAMER is expected to get better scalability and repairability.

7.2 Limitations

Since we use CBPS to synthesize the conditional of the patch and CBPS can only synthesize the numeric conditionals, HAMER can not synthesize the conditional that includes other types of elements, such as strings. HAMER is also unable to repair memory errors for parallelism. Finally, HAMER's current repair algorithm cannot handle the loop. If a variable is in a loop, for example, it will yield multiple values for a fuzzer input, making it impossible for HAMER to build an expected test suite. Similarly, if the program does dynamic memory allocation in a loop, HAMER can only fix the memory leaks that occur in each loop one by one, but the correct patch may just require deallocation in the loop.

7.3 Future Work

HAMER is an automated memory error repair tool. Although HAMER can only fix memory leaks at the present, we can simply extend our repair algorithm to fix use-after-free and double-free errors. We can fix use-after-free by first removing the free function and then treating it as a memory leak. Similarly, deleting the redundant free function for double free may either fix the error or introduce a memory leak, which we can also fix using the current approach.

We can also improve our repair algorithm to prevent patches from failing to generate patches or generating wrong patches due to fuzzer faults, such as test8. To avoid this issue, HAMER's current strategy is to run fuzzer multiple times to decrease the probability of an error being missed. We have two ways of improving HAMER's performance.

- If a new dynamic memory leak is detected during the repair process, we compare it to the previously detected errors and add it to our repair list if it is a new error.
- If we trigger a previously repaired error when repairing the current error, we add that error back to the repair list.

At this time, our implementation of simp-CBPS can only synthesize the conditional of a few templates, thus we need to expand the types of conditional by adding more conditional template specifications. We also need to increase HAMER's scalability and figure out how to deal with the loop.

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