# 國立交通大學

# 資訊科學與工程研究所

# 碩 士 論 文

# 操作符號位址以產生異常執行路徑

Exploiting Symbolic Locations for Abnormal Execution Paths

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#### 摘要

程式開發者不能完全避免因疏忽造成的漏洞,因而如今軟體安全是一個重要 的議題。擬真測試是一項典型的自動化軟體測試技術,藉由實體測試與符號測試 之間的結合交互作用,擬真測試可以達到較高且較精確的程式碼檢測率。但在擬 真執行時,若發現路徑條件表示式含有符號型態的位址,它將無法掌握在相反路 徑條件下所代表的實體數值,因而無法找到該相反路徑。本論文針對擬真符號測 試,提出一個能增加程式碼檢測率的擬真位址模組,為了確保符號測試正常執行, 我們暫時將含有符號位址的部分取代掉,並將替換的資訊記錄在符號位址表,最 後我們透過路徑條件和符號位址表找出可能的位址解。我們目的為透過求解出來 的符號位址,進入我們之前無法執行的路徑,如此一來我們將能提升程式碼檢測 率,並找到更多程式錯誤。

#### Exploiting Symbolic Locations for Abnormal Execution Paths

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#### ABSTRACT

The vulnerability caused by the negligence of the programmer is unavoidable. Software security is an important issue today. Concolic testing is a typical technique in automatic software testing. It achieves high coverage and precise analysis by combining concrete and symbolic execution in a co-operative way. But it cannot handle the situation when the address is symbolic in the path condition, so concolic executer may not find a concrete value which represents the test case of another negated path. This thesis proposes symbolic address module for enhancing the coverage of concolic testing. We use a substitute method to ensure symbolic executor running correctly and construct a symbolic address map to record symbolic address information. According to map information and path conditions, we generate a possible answer for symbolic addresses. We aim to find symbolic address solutions to enter abnormal paths we had never executed before. Then we can find more bugs by improving the code coverage.

<span id="page-4-0"></span>

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#### <span id="page-8-0"></span>**1. Introduction**

Software testing is the process that assuring program quality is identical with our expectation. In the development of complicated software, humanly programmer may miss some requirement or implement redundant functions. Both behaviors will lead to bugs or security problems.

The process of software testing is tedious and labor-consuming so manual testing is unfeasible. In recent years automatic software testing technique is mature gradually, there are many researches proposed to resolve the issues [\[17,](#page-41-0) [24,](#page-42-0) [23,](#page-42-1) [6,](#page-41-1) [16\]](#page-41-2). A typical testing technique named concolic testing [\[15,](#page-41-3) [20,](#page-42-2) [7\]](#page-41-4); it tests the software by combining concrete and symbolic execution[\[12\]](#page-41-5) in a co-operative way. This method is feasibility on real program unit. But if the path constraint has the address which is symbolic, concolic executor cannot find the suitable real address solution for negated path constraint and it will abort this negated path.

Our work is base on concolic testing; we propose a new testing feature named symbolic address module. We aim to exploit symbolic address solutions and improve the path coverage. Then we can enter abnormal paths that we had never entered before.

#### <span id="page-8-1"></span>**1.1 Background**

In the recent years, software security is a serious issue. Because of humanly software testing is not efficient, it's important to use the automatic tool to inspect software for vulnerability likes buffer overflow[\[10,](#page-41-6) [14,](#page-41-7) [22\]](#page-42-3). We focus on tainted base vulnerability; the overwritten data may cause unexpected behaviors. We describe major researches about our work below.

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#### <span id="page-9-0"></span>**1.1.1 Common Vulnerabilities**

Stack-based Overflow

A buffer overflow occurring in the stack memory is referred to as a stack overflow. A common case is that a local variable near the buffer but the program doesn't exam the buffer size. When we manipulate the buffer memory out of range, the local variable will be covered by our input. Not only local variable but we also possibly overwrite function pointer, exception handler or return address. Attacker may use those overwritten data to crash program or executing an unexpected instruction.

#### Heap-based Overflow

A buffer overflow occurring in the heap data area is heap overflow. Heap memory is dynamically allocated by program at run-time. Exploitation is to cause the program overwrite the memory management information which associated with heap memory such as dynamic memory allocation linkage. For example in BSD Phkmalloc, we can overflow metadata of malloc and then overwrite GOT entries or return address.

#### Uninitialized Variables

Uninitialized variable is a new declared variable which the program didn't set an initial value before using it. The value of uninitialized variable cannot be expected but it may tainted by other variable when two variable allocated in the same address range. Attacker can find a specific path to control the uninitialized variable and it may cause the vulnerability.

#### <span id="page-9-1"></span>**1.1.2 Program analysis policy**

• Static analysis

Static analysis is performed without actually executing programs. Instead,

static analysis just scans the source code to gather information about the possible set of values, parses execution states of the program. It is usually implemented in formal methods such as data-flow analysis, model checking.

Static analysis tool used to detect vulnerability such as buffer overflow. We can check if there are dangerous standard library functions in source code such as *strcpy* and *fgets*. Unfortunately, the drawback of static analysis is high false positive; it cannot promise that all the found vulnerability will occur in actually executing programs.

#### • Dynamic analysis

Dynamic analysis actually executes the program and detects vulnerabilities at run-time. Such as valgrind[\[18\]](#page-41-8), a tool for memory debugging and memory leaking analysis tool. It usually needs a large number of test cases and a software testing technique: code coverage observer to explore paths. Dynamic analysis can promise that all the found vulnerability will occur during executing program. It is more precise than static analysis, but it also

needs more time in executing analysis.

#### <span id="page-10-0"></span>**1.1.3 Program testing mechanism**

#### Random testing

Random testing is also named fuzzing; it is commonly used to test program security. It selects random inputs for target program and monitors if there is exceptions such as crashes occurred.

Fuzzing explores random paths very fast, but it wastes a large amount of time to enter the same path. The tool zzuf perform fuzzing testing on target program.

Symbolic execution

Symbolic execution is useful for software validation because it can prove if the errors may occur. The main idea is to use the tracking symbolic for the input variable. It executes the program symbolically on symbolic variable. It collect symbolic path constrains and then uses constraint solver to explore execution paths. In the result, the explored paths represent by mathematical expressions. The drawback of symbolic execution is it does not scale for large programs because of there is too many execution paths.

#### Concolic testing

Concolic testing combines random testing and symbolic execution in a co-operative way. It initializes the input variable with the symbolic variable. As program runs, it first chooses a random value to determine a path and collect the path constraints. In the next run, it negates the last path condition and feeds this new path conditions to the solver, and gets another concrete value which represent the test case of new path. This counterexample technique can be used to find next path conditions and available test case until all the paths is explored.

Concolic testing is focus on finding bugs in the real program. It has higher branch coverage than random testing and has no false positives or scalability problem like in symbolic execution.

#### <span id="page-11-0"></span>**1.1.4 Control the branch by symbolic address**

In normal concolic execution, the symbolic variable refers to the tainted value. We explore paths with branch conditions which including symbolic variable.

Considering about the path condition has the tainted address, following are two specific types:

i. Constant\_buf[symbolic index]=constant\_value

#### ii. Symbolic\_pointer[constant value] = constant\_value

In type **i**, the base address is constant, but the array index is symbolic. The left of equation is a symbolic array index dereference. If we want to satisfy the condition, we have to find a specific address which it's dereference value is exact equal to right equation constant\_value. In type ii, the pointer is symbolic. The left of equation is a symbolic pointer dereference. We can also find a specific address which it's dereference value is equal to right equation constant\_value.

Above the first type we called symbolic array index, the second type we called symbolic pointer. Because of their address are symbolic variable, we say that's the symbolic address. Our thesis interest in symbolic address solution, we try to exploit symbolic locations for abnormal execution paths.

**WWW** 

#### <span id="page-12-0"></span>**1.2 Motivation**

Concolic testing is a popular software verification technique, it explore program paths as many as possible and find bugs. In figure 1, the buffer overflow occurs at *Line 6*. The address of pointer *p* is tainted by the standard input. In cocolic testing, the address of pointer *p* became a symbolic variable and it cannot determine the value of *p[0]*. In the situation we may miss the true path and we cannot find the vulnerability at *Line 8*.

> *1 #include <stdio.h> 2 void main() 3 { 4 int \*p; 5 char buf[4]; 6 fgets(buf,10,stdin); 7 if(p[0]==5) 8 vulnerability; 9 return 0; 10 }*

Figure 1 the address of pointer **p** is tainded by **stdin**

If we can find a address for p that let the dereference of pointer p is properly

5, then we can enter an abnormal execution path and find the vulnerability.

#### <span id="page-13-0"></span>**1.3 Problem Description**

For concolic testing, if the pointer or the array index is symbolic in branch condition, the executor doesn't know where to get the proper value from memory. The executor will execute incorrectly and then give up the path. But in some cases if we choose a suitable address for symbolic pointer or symbolic array index, the condition will be satisfied. Then we can enter this execution path.

In order to perform better coverage, we should construct a Symbolic Address Map for recording information of symbolic pointer and symbolic array index; we should add relation constraints into branch condition for exploring abnormal paths.

#### <span id="page-13-1"></span>**1.4 Objective**

We focus on handling symbolic address to enter abnormal paths, and we can trigger more vulnerability in those paths which we had never entered before. To achieve these goals, we will try to implement two major objects on  $S^2E[5, 4]$  $S^2E[5, 4]$  $S^2E[5, 4]$ :

- 1. Symbolic Address Map: A table records symbolic addresses information and relationship between each symbolic address.
- 2. Symbolic Address constraints generator: A generator which generates relation constrains base on Symbolic Address Map for satisfying the abnormal path condition.

#### <span id="page-13-2"></span>**2. Related Work**

Following tools specify and track symbolic variables and constraints; they fully

explore program paths and find bugs. CRED[\[19,](#page-41-11) [11\]](#page-41-12) (C Range Error Detector) directly checks the bound of memory accesses; it's just a bound checking tool. DART[\[9\]](#page-41-13) (Directed Automated Random Testing) dynamic analysis of how the program behaves under random testing and automatic generation of new test inputs. It mainly handles the integer constraints and invokes random testing with symbolic pointer. CUTE[\[21\]](#page-42-4) is the first concolic testing tool which splintered from DART. It simulates the pointer into array, it can handle some symbolic pointer cases, but it cannot handle symbolic array index. CREST[\[1\]](#page-40-2) is a concolic testing tool for C, it combine concolic testing with heuristic search strategies to perform high coverage on large software systems. It doesn't handle symbolic array index or symbolic pointer. EXE[\[3\]](#page-40-3) can handle the more complex pointer access than CUTE. But it cannot handle multi-dimension dereference. It can handle symbolic array index but just in-bound related. SAGE[\[8\]](#page-41-14) implements a new memory model for handling symbolic array index but just a bound checking tool. Catchconv is a symbolic execution and run-time integer conversion testing tool. It is a module of Valgrind and only focus on testing integer conversion error. SecTAC[\[25\]](#page-42-5) is Trace-based security testing tool. Each trace is symbolically executed to produce program constrains and security constraints. Its trace can handle neither symbolic array index nor symbolic pointer. KLEE[\[2\]](#page-40-4) is redesigned from EXE. It is a symbolic virtual machine built on LLVM[\[13\]](#page-41-15) compiler infrastructure and uses search heuristics to reach high coverage in program. KLEE cannot handle symbolic array index or symbolic pointer. S2E is redesigned from KLEE, it provides the illusion of symbolic execution of an entire software stack, including applications, libraries, OS kernel, device drivers, and even firmware. It has guessing steps for symbolic address but not enough. Alert is developed by our laboratory; it used the memory model of EXE and the execution model of CUTE. It handles both symbolic array index and symbolic pointer, but it cannot handle out-of-bounds array index.

Our research base on S2E executor, we implement a plug-in for handling symbolic address. The comparison of above tools is shown as table 1.



# <span id="page-15-0"></span>**3. Method and Steps**

We provide a plugin out of box for enhancing path coverage by handling symbolic address problem, including symbolic array index dereference and symbolic pointer dereference.

 $S<sup>2</sup>E$  executor inherited KLEE symbolic executor. In the original edition, when  $S<sup>2</sup>E$  executor found an address is symbolic in constraints, it didn't know where to get the suitable value in the memory.

Figure 2 shows actual examples for symbolic array index and symbolic pointer for  $S^2E$ .  $S^2E$  executor will transfer the concrete value to the symbolic variable at *line 5*. At *Line* 7, because of the buf's address is symbolic,  $S^2E$  executor will try to assign concrete values to buf's address and fork states to solve the constraints. But in the most of the cases, the number of forking states always reached the maximum number of states to fork when concretizing symbolic value. Unfortunately the original edition failed to solve the symbolic address problem and can't reach *line 8*.



Figure 2 symbolic pointer and symbolic array index

But in the left half side of figure 2, if we assign the address of *buf[0]* equal to address of *a*, we can pass the true branch and reach *line 8:"GOAL"*. In the right half side of figure 2, if address of *buf[i]* equal to address of *a (i=4)*, the program can also reach *line 8:"GOAL"*.

In our research, we add a new plug-in named  $SymbolicAddress$  for  $S<sup>2</sup>E$ . We use a substitute method to ensure  $S^2E$  running correctly and construct a symbolic address

map to record every symbolic address. When  $S<sup>2</sup>E$  executor state terminating, we will check the state constraints and symbolic address map and then generate a possible answer. The following is detailed conception.

#### <span id="page-17-0"></span>**3.1 Symbolic Address classifications**

Before solving the symbolic address problem, we have to define what is symbolic address? In our thesis, Symbolic address is classed as two main parts: symbolic pointer and symbolic array index.

Figure 3 shows the possible symbolic pointer classifications.

Example **i** shows a trivial symbolic pointer. Address of pointer *p* is symbolic and "*p[0] == concrete value"* is a condition in branch.

Example **ii** shows multiple symbolic pointers. We have to consider those different symbolic pointers dereference are adding together in a condition.

Example **iii** shows the pointer may have offset. The same pointer but different offset can be bind to different addresses in the memory. It means two different offset of the same pointer have different dereference values, but they have a fixed distance between them in the memory.

Example **iv** shows the pointer to pointer case. Of course, not only pointer to pointer, but also we have to consider triple or more. ex. \*\*\*  $p,$  \*\*\*\*  $p$ …



#### Figure 3 symbolic pointer classifications

Figure 4 shows the possible symbolic array index.

Example **i** shows a trivial symbolic array index. Because of integer *i* is symbolic, then address of *buf[i]* is also symbolic. "*buf[i] == concrete value"* is a condition in branch.

Example **ii** shows multiple symbolic array indexes, different symbolic array indexes dereference are adding together in a condition.

Example **iii** shows two different base addresses have the same symbolic array index. They bind to different addresses in the memory and may have different values, but they have a fixed distance between them in the memory. Furthermore, The case maybe (*bufA[i] + bufB[i] + bufC[i])* or more.

Example **iv** shows the multi-level symbolic array index. The entire size of *buf* is *(i*  $\times j \times k$ *)*. It means the different *(i, j, k)* may cause the *buf[i][j][k]* have the same address in the memory.



Figure 4 symbolic array index classifications

#### <span id="page-19-0"></span>**3.2 Symbolic Address variable Substituting**

 $S<sup>2</sup>E$  executor should fork a new state and add a negate constraint to it when execute a branch condition. According to left half side of figure 2, we can get a state constraints diagram figure 5.



Figure 5 origin state constrains diagram

In the fact, KLEE executor cannot execute correctly when the branch condition has the symbolic address. Although  $S^2E$  executor inherited KLEE symbolic executor and has guessing steps to handle symbolic address, it still cannot allocate appropriate address in most of case. In this case,  $S^2E$  executor will fork states until reached the maximum number of states to fork when concretizing symbolic value, then fail to find a available address for *buf[0]*.

We have an idea for  $S^2E$  executor when found a symbolic address in state condition. We declare a new symbolic variable and then substitute the symbolic

address dereference value. Figure 6 shows the state constraints diagram after the substitution.  $S^2E$  executor now can execute successfully.



Only substitution is not enough, we have to construct the relation sheep between our new made symbolic variables. We will explain the symbolic address map structure in next section.

#### <span id="page-20-0"></span>**3.3 Symbolic Address Map**

Symbolic address map contains four basic elements: *Origin Expression, Substituted Expression, Related Address and Target Address*.

i. *Origin Expression* is a symbolic address expression, symbolic executor doesn't know where to read it in the memory.

- ii. *Substituted Expression* is a new declared expression used to substitute *Origin Expression*. In addition, we add it to Symbolic Table and then Symbolic executor believes it is a symbolic variable. Now symbolic executor can execute it continually.
- iii. *Related Address* is the same with the address of *Origin Expression's* symbolic variable in symbolic table. If two symbolic addresses have the same *Related Address*, one of them may another one's dereference.
- iv. *Target Address* is a blank space now. It used to store a concrete address which suits with state constraints.

Figure 7 shows how to construct the symbolic address map. If the branch conditions still have symbolic addresses, it adds symbolic address into the map recursively until there is no symbolic address.



Figure 7 Flowchart of Constructing symbolic address map

Figure 8 is a symbolic address map example. Symbolic executor found a symbolic address in branch condition *(Line 13)*. We declared a new expression named *buf1*, its *origin expression* was *buf*, and the *related address* was the same with the address of *buf* in Symbolic Table. Besides, we added *buf1* into Symbolic Table. Until now, *buf1* represented the value at address *buf[1]* in memory. Because of *buf[1]* was symbolic, we had to declared a new expression named *buf2* and substitute *buf1*. In the end, *buf3* represented a concrete value at address *buf[1][3][5]* in memory. We finished substituting all the symbolic address in branch condition.



Figure 8 symbolic address map example

#### <span id="page-22-0"></span>**3.4 Symbolic Address Constrains Generator**

Our goal is finding *Target Address* for symbolic address map. We could use *Target Address* stored in symbolic address map to generate symbolic address constrains. In the end, we add symbolic address constrains into symbolic execution state, and then symbolic executor will automatically generate a test case for every symbolic address.

Figure 9 shows how to generate symbolic address solutions. If state constraints

have any symbolic addresses before symbolic execution state terminated, we choose a combination of addresses from Symbolics Table and pass them into symbolic address map *Target Address* field. We use STP solver to identify if the relationships in Symbolic Address Map is satisfy all constrains in symbolic execution state. If answer is yes then we obtain a solution, otherwise we choose next combination addresses from Symbolic Table and use STP solver to identify again.

If all combination of addresses in Symbolic Table is not the solution, then we try to find it in the actual memory. As before, but we choose a combination of addresses from actual memory. If there is no solution in Symbolics Table or actual memory, we say that this path maybe is impossible in the program.



Figure 9 Flowchart of symbolic address solution

### <span id="page-23-0"></span>**4 Implementation**

 $S<sup>2</sup>E$  provides the core symbolic execution engine. All the analysis is done by

various plug-in. In this thesis, we write a plug-in named *SymbolicAddress* that uses features of the  $S^2E$  plug-in infrastructure.

We substitute symbolic pointer dereference and symbolic array index dereference during symbolic execution, and we add them into Symbolic Address Map. Symbolic Address Map describes what expression to be substitute and where address to be substitute.

Before  $S^2E$  execution state terminated, according to Symbolic Address Map we search available addresses in Symbolics Table or actual memory for every symbolic address. Finally, we add those available address relation constrains into  $S^2E$  execution state conditions, and the S 2 E plug-in named *TestCaseGenerater* will generate an available test case for every symbolic variable automatically.

#### <span id="page-24-0"></span>**4.1 Symbolic Address Map & Class**

As Figure 10 shows, Symbolic Address Map construct from symbolic addresses during symbolic executing.

*Struct SApoint{ uint64\_t tempAddress; ref<Expr> tempExpr; const Array\* tempArray; ref<Expr> targetExpr; const Array\* targetArray; uint64\_t targetAddress; ref<Expr> targetValueExpr; };*

Figure 10 Symbolic Address structure

#### *tempAddress : Related Address*.

#### *tempExpr : Substituted Expression*.

*tempArray :* the Aarray object used to store major variable name of *tempExpr.*

*targetExpr : Origin Expression*.

*targetArray :* the Aarray object used to store major variable name of *targetExpr.*

*targetAddress : Target Address*.

*targetValueExpr :* the content of *targetaddress* in actual memory.



Figure 11 shows our implementation. *SAcounter* used to calculate how many symbolic addresses in SAmap. Function *adjust* doing the substitute stage when symbolic executor found the symbolic address. Function *solutionGEN* doing the solution searching stage and constrains generating stage at symbolic execution state terminating.

#### <span id="page-26-0"></span>**4.2 Symbolic Address Plug-in for S2E**

S 2 E symbolic executor will call the function *handleForkAndConcretize* when instructions have expression. If the expression is constant or state->forkDisabled is on, it will simply pick one possible value and return. Otherwise, if the expression has symbolic address, it will run the guessing steps.

As figure 12, we instrument our function *adjust* to Instead the guessing steps. We pick the moment to substitute symbolic address when *handleForkAndConcretize* find the symbolic address.

```
S : the current Execution State pointer
E : the current Expr in the branch
1 S2EExecutor::handleForkAndConcretize ( S , E , …){
2 …
3 If E is NOT a symbolic address{
4 …
5 return
6 }
7 …
8 guessing steps
9 adjust( S , E )
6 }
```
Figure 12 instrumenting function *adjust*

Before every symbolic execution state terminating,  $S^2E$  handler will call the function *processTestCase.* This function will call the original author plug-in named *TestCaseGenerator*, it will generate an available value for each symbolic variable.

As figure 13, we instrument our function *solutionGEN* before calling the plug-in *TestCaseGenerator*. We search available address for each symbolic address and add related constrains between each other into  $S^2E$  state constrains. In the end, plug-in

*TestCaseGenerator* will also generate an available value for each origin symbolic variable and our new made symbolic variable.

```
S : the current Execution State pointer
1 S2EHandler::processTestCase ( S ){
2 …
3 solutionGEN ( S ) // add constrains before running s2e::plugins::TestCaseGenerator
4 …
5 getPlugin("TestCaseGenerator")
6 }
```
Figure 13 instrumenting function *solutionGEN*

#### <span id="page-27-0"></span>**4.3 Symbolic Address adjust function**

Figure 14 shows the pseudo code of function *adjust*. We use integer *N* to count how many symbolic addresses now. We also use the String*"SA*"+ *IntToStr*( *N* ) as the new name for our new declared expression. *Line 8* to Line 12 will fill the new symbolic address with the substitute information. Now the struct member *P.targetAddress* and *P. targetValueExpr* will be blank, we will use them in other

steps.

```
S : the current Execution State pointer
E : the current Expr in the branch
N : the number of symbolic addresses
A : the origin symbolic variable's address in memory
P : a SApoint struct used to store symbolic address information
NAME : a new symbolic variable used to Substitute symbolic address value
SAMAP : the vector used to store every SApoint information in the program
 1 SymbolicAddress:: adjust(S , E ){
 2 N ← N + 1
 3 A ← symbolic variable's address in E
 4 NAME \leftarrow String "SA" + IntToStr(N)5 S->createSymbolicArray( A , 4 , NAME )
 6
 7 P ← new SApoint
 8 P.tempAddress \leftarrow A
 9 P.targetArray ← findSymbolicObjects( E )->second
10 P.targetExpr \leftarrow E
11 P.tempArray ← findSymbolicObjects( NAME )->second
12 P.tempExpr ← Expr:: createTempRead( P.tempArray , 32 )
13
14 SAMAP.push_back( P )
15 }
```
#### Figure 14 function *adjust*

Figure 15 is a Symbolic Address Map example of figure 2. Both programs we

only consider State 0 when executing true branch, and each state has one symbolic address.



<span id="page-29-0"></span>Figure 15 example of Symbolic Address Map after adjusting **4.4 Symbolic Address solution generator function**

Figure 16 shows the pseudo code of function *solutionGEN*. We have two searching algorithm: *searchSymbolicMap* and *searchMemoryMap*. First we search the Symbolics Table for symbolic address solution. If we didn't find an available solution, then we search the entire memory from address *BM* to address *EM*. In our research, we define the region from *BM* to *EM* as addresses near the symbolic address in stack memory.

When we get an available solution for Symbolic Address Map, we add two constrains (*line 13*  $\sim$  *16*) into S<sup>2</sup>E execution state. At *Line 13*, symbolic address expression *P ->targetExpr* should equal to concrete address *P ->targetAddress*. At *Line 14 ~ 15*, our new declared expression  $\overline{P}$  ->tempExpr should equal to expression *P ->targetValueExpr* which address is *P ->targetAddress*.

```
S : the current Execution State pointer
B : a boolean variable will be TRUE if the state have a symbolic solution
SAMAP : the vector used to store every SApoint information in the program
BM : the begin of the memory
EM : the end of the memory
P : a SApoint pointer used to store symbolic address information
E : the current symbolic address constrains that should be add to the state
1 SymbolicAddress:: solutionGEN ( S ){
2 If searchSymbolicMap( S , SAMAP ) is true{
3 B ← True
4 }else if searchMemoryMap(S , SAMAP , BM , EM) is true{
5 \qquad \qquad \mathcal{B} \leftarrow \text{True}6 }else{
7 No solution!!
8 B ← false
 9 }
10
11 If B is true{
12 for each P \in SAMAP (
13 E ← EqExpr::create(\mathcal{P} ->targetExpr, ConstantExpr::create(\mathcal{P} ->targetAddress, 32))
14 E ← =AndExpr::create(E ,
15 EqExpr::create(P ->tempExpr, P ->targetValueExpr))
16 S ->constraints.addConstraint(E )
17 }
18 }
19 }
```
#### Figure 16 function *solutionGEN*

#### <span id="page-30-0"></span>**4.4.1 Searching solutions from Symbolics Table**

Figure 18 shows the pseudo code of function *searchSymbolicMap*. At *Line 2*, function *chooseNextSymbolicAddress* chooses a combination of addresses from Symbolics Table, and those addresses are set in each object *SApoint* member *targetValueExpr* where in Symbolic Address Map. We loop the function until found an available combination or all the combinations are unavailable.

At *Line 9~13*, we handle the case which object *SApoint* member *targetArray* is the same. We have two major cases:

i. Symbolic pointer: multiple offsets of the same pointer address.

*if(*  $p[0] + p[1] ==$  *concrete value), p is a symbolic pointer.* 

ii. Symbolic array index: the same symbolic array index with multiple bases.

*if( bufA[i] + bufB[i] == concrete value ), i ,j are symbolic array index*

We can solve the problem by adding a constraint: the offset between two object *SApoint* member *targetArray* should equal to offset between two object *SApoint* member *targetExpr*.

At Line  $14~18$ , we add a constraint to handle the pointer to pointer problem:

*int* \*\*p, p is a symbolic pointer, if  $p[1][2] ==$  concrete value )

If X and Y are both the symbolic addresses and Y adjust from X. Figure 17 shows the dereference relationship between X and Y. The value at *X->targetAddress* in memory should equal to which subtract the offset of *Y->targetExpr* from *Y->targetAddress.* Moreover, this method can handle not only the pointer to pointer, but also \*\*\*p, \*\*\*\*p… or more situation.



Figure  $17 \text{ p}[1][2] ==$  concrete value

```
S : the current Execution State pointer
SAMAP : the vector used to store every SApoint information in the program
\mathcal P \mathpunct{:} a SApoint pointer used to store symbolic address information
P' : a SApoint pointer used to store symbolic address information
E : the current address constraints used for solver
SUC : a Boolean variable used to determin 
 1 searchSymbolicMap( S , SAMAP ){
 2 while chooseNextSymbolicAddress( S , SAMAP ) is true{
 3 E ← NULL
 4 for each P \in SAMAP /
 \mathcal{F} \leftarrowAndExpr::create (\mathcal{F},
 6 EqExpr::create(P ->targetExpr, ConstantExpr::create(P ->targetAddress , 32) ))
 7 \mathbf{E} \leftarrow \text{AndExpr::create}(\mathbf{E},8 EqExpr::create(P ->tempExpr, P ->targetValueExpr))
 9 for each P' \in SAMAP && P' ->targetArray->name is equal P ->targetArray->name{
10 \qquad E \leftarrow \text{AndExpr::create}(E)11 EqExpr::create(ConstantExpr::create((P ->targetAddress)-( P' ->targetAddress) , 32),
12 SubExpr::create(P ->targetExpr, P' ->targetExpr)))
13 }
14 for each P' \in SAMAP && P'' is adjust from P' {
15 \mathbf{E} \leftarrow \text{AndExpr::create}(\mathbf{E}),16 EqExpr::create( S ->readMemory( P'->targetAddress , 32 ), 
17 ConstantExpr::create( (P'' ->targetAddress)-(offset of P'' ->targetExpr), 32))
18 }
19 }
20 solver->mayBeTrue(Query( S ->constraints , E ) , SUC )
21 if SUC is true{
22 for each P \in SAMAP /
23 P ->targetValueExpr ← S->readMemory(P->targetAddress, 32)
24 }
25 return TRUE
26 }
27 }
28 return FALSE
29 }
```
Figure 18 function *searchSymbolicMap*

#### <span id="page-33-0"></span>**4.4.2 Searching solutions from actual Memory**

The function *searchMemoryMap* is almost the same with *searchSymbolicMap.* The only different is *Line 2* in figure 18; instead of *chooseNextSymbolicAddress* we use the function *chooseNextTargetAddress*. We search a range of actual memory, including concrete value and symbolic variable. In the fact, if we define the region as entire memory including text, data, heap and stack, it will take a large amount of searching time. In our implement, we search near the memory region in stack which near the symbolic address. We have the best opportunity to find available addresses in this region.

In addition, if the number of origin symbolic variables is more than the number of symbolic address we new declared. We do not need to search anywhere. We set each origin symbolic variable address to each *SApointer* member *targetAddress*. It will be a symbolic address solution.

# <span id="page-33-1"></span>**5 Result and Experiment**

We present results of experiments and prove symbolic address module in this section. We use the example of SAGE and our made programs to illustrate the solution of symbolic array index and symbolic pointer. Next, we discuss the efficiency of two searching algorisms when execution state has remaining symbolic variables. In the end, we illustrate the enhancement on path coverage with real programs.

#### <span id="page-33-2"></span>**5.1 A simple Example of SAGE**

As shown in figure 19, we test the example of SAGE. The program has two symbolic variable x and y. Our goal is to reach the Line 15.

When reached line 14, we executed the true branch first. We found 2 symbolic

array indexes in the state constraint. Then we new declared symbolic variables x1 and y2, x1 is the dereference value of buf[x], y2 is the dereference value of buf[y]. We updated our symbolic address map and adjusted the state constraint to

*(Eq (ReadLSB w32 0 x1)*

*(Add w32 2 (ReadLSB w32 0 y2)))*

For generating symbolic address solution, we searched a combination of addresses in stack memory. Later, we discovered when x equal 3 and y equal 1, then x1 equal 2 and y2 equal 0, and this was a combination of addresses solution which satisfied the state condition. According to symbolic address map, we added related constrains into the state constraint. In the end, plug-in *TestCaseGenerater* automatically generated a test case for symbolic variables which including our new declared. The false branch symbolic address solution is also generated in the same way.

 $\mathcal{S}$  and  $\mathcal{S}$ 



Figure 19 SAGE

#### <span id="page-35-0"></span>**5.1.1 More Complicated Symbolic Array Index**

In this sub section, we evaluate the test case as shown in figure 20, which has more complicated symbolic array index. This experiment focus on two major classifications for branch condition:

*if ( bufA[i][j][k] + bufB[k] + bufC[l] == 10 )*

- 1. Multi-dimension symbolic array index
	- $\rightarrow$  bufA[i][j][k] is a 3-dimension symbolic array index
- 2. Different base addresses have the same symbolic array index
	- $\rightarrow$  bufA[i][j][k] and bufB[k] have the same symbolic array index k

The program has four symbolic variable i, j, k and l. Our goal is to reach the Line 16. As our expectation, the true branch symbolic address solution:

*bufA[1][0][2] + bufB[2] + bufC[1] = i1+ k2 + l3 = 5 + 2 + 3 = 10*

In the false branch, because of bufA[0][0][0] is an uninitialed value, so its value is 0xb7ffhf68.



Figure 20 program 1

#### <span id="page-36-0"></span>**5.1.2 Symbolic Pointer classifications**

We evaluate the test case as shown in figure 21, which includes all symbolic pointer classifications. This experiment focus on two major classifications for branch condition:

 $if$  ( $pA[3]$  +  $pA[4]$  +  $pB[0][1][2]$  == 10)

1. Multi-dimension symbolicpointer

 $\rightarrow$  pB[0][1][2] is a 3-dimension symbolic pointer

- 2. different offsets of the same pointer address
	- $\rightarrow$  pA[3] and pA[4] is the same pointer but different with offset

The program has a single-dimension pointer pA and a 3-dimention pointer pB. Our

goal is to reach the Line 14.

In the true branch, pA1 is the dereference value of pA[3] and pA2 is the dereference value of pA[4]. Because of pB[0][1][2] is a 3-dimension symbolic pointer, pB3 is the first-dimension dereference value, pB34 is the second-dimension dereference value which adjust from pB3, pB345 is the third-dimension dereference value which adjust from pB34, so pB includes three symbolic addresses. There are total five symbolic addresses in the path condition. The final state constraint is:

*(Eq 10*

 *(Add w32 (ReadLSB w32 0 pB345) (Add w32 (ReadLSB w32 0 pA2) (ReadLSB w32 0 pA1))))*

Figure 22 shows the multi-dimension graph for symbolic pointer pA and pB. The true  $\equiv$   $\equiv$   $\mid$   $\equiv$   $\mid$ branch symbolic address solution:

*pA[3] + pA[4] + pB[0][1][2] = pA1+ pA2 + pB345 = 3 + 5 + 2 = 10*

In the end, we exploit this abnormal path success.





Figure 22 multi-dimension graph for symbolic pointer

#### <span id="page-38-0"></span>**5.2 Searching Algorisms analysis**

In our thesis, we had implemented 2 algorisms for symbolic address solution generator: searching memory and searching Symbolic Table. Searching memory is a basic searching algorism; we search for the solution in memory. If searching range is too large, it may get stuck in searching step for a long time; if the range is too small, it may miss possible solutions. According to our experiment, we usually set the range to near the symbolic address location. Searching Symbolics Table is an opportunistic algorism. We map the address solutions to Symbolics Table and check if it is a solution. This algorism is useful when remaining symbolic variables which have no

relationship with symbolic address exist.

Table 2 shows the number of compares on two algorisms; we test 3 programs which in previous section. N/A means it has no solution in Symbolics Table. Because of there is no any remaining symbolic variable, searching Symbolics Table only improve the program1's false branch.

		$\mathbf{u}$ <b>Searching memory</b>		<b>Searching Symbolics Table</b>	
Program	Num. of symbolic address	true branch	false branch	true branch	false branch
<b>SAGE</b>		23		N/A	
Program1		1312	74	N/A	34
Program2		12476	1640	N/A	N/A

Table 2 analysis of searching algorism

In the fact, when the buffer overflow happened, it always not only covers the symbolic address. It also covers other variable in most of case. Table 3 shows the situations when remaining symbolic variables are exist. Searching Symbolic Table is a powerful method to reduce the number of compares; in progam2 it reduces the number of compares from 12476 to 92 when remaining variable when there are two remaining symbolic variables. In addition, the compares will increase a little when there are more than 3 remaining symbolic variables. It's because of the Symbolics Table size will increase with remaining symbolic variables. It has to waste some compares on initialization.

**Program** 1 **remaining**  $\begin{bmatrix} 2 & \text{remaining} \\ 3 & \text{remaining} \\ 4 & \text{ remaining} \\ 1 & 4 \end{bmatrix}$ **true branch false branch true branch false branch true branch false branch true branch false branch SAGE** | 2 | 1 | 2 | 1 | 2 | 1 | 2 | 1 **Program1** 34 34 48 48 64 64 82 82 **Program2** 8274 N/A 92 83 103 102 124 123

Table 3 remaining symbolic variable

### <span id="page-40-0"></span>**6 Conclusion**

We propose a new symbolic address module in this thesis. We construct a new symbolic address map on  $S^2E$ . We trace and adjust the symbolic address during the symbolic execution step, and then we handle the concrete execution step by our symbolic address solution generator. Two execution steps work in a co-operative way and exploit abnormal paths.

Our objective is to exploit the path condition which has the symbolic address like (*\*tainted-pointer == concret-value)*. If the program has another remaining tainted variable, we can directly assign *concret-value* to this tainted variable and assign tainted variable address to *tainted-pointer*, and the branch is always be true. In other words, if there is a symbolic address with remaining tainted variable, we can say this branch is completely controllable. If there is no remaining tainted variable, we still possibly find a solution in concrete memory; in the situation we say this branch is possibly controllable.

In future works, it is possibly find the statement (*\*tainted-pointer = tainted-value)* in the program. If found the situation then we can modify any memory content and fully control the target program followed my inclination.

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