

Abstract Path Testing with PathCrawler

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ABSTRACT

PathCrawler is a tool developed by CEA List for the automatic generation of test inputs to ensure the coverage of all feasible execution paths of a C function. Due to its concolic approach and depth-first exhaustive search strategy implemented in Prolog, PathCrawler is particularly efficient in the generation of tests to cover the fully expanded tree of feasible paths. However, for many tested functions this coverage criterion demands too many tests and a weaker criterion must be used. In order to efficiently generate tests for a new criterion whilst still using a concolic approach, we must modify the search strategy. To facilitate the definition and comparison of different coverage criteria, we propose a new type of tree, trees of abstract paths, and define the different types of abstract node in these trees. We demonstrate how several criteria can be conveniently defined in terms of coverage of these new trees. Moreover, efficient generation of tests to satisfy these criteria using the concolic approach can be designed as different strategies to explore these trees.

Categories and Subject Descriptors

D.2.5 [Software Engineering]: Testing and Debugging – *Testing tools (e.g., data generators, coverage testing).*

General Terms

Design, Performance, Algorithms.

Keywords

structural testing, test generation, coverage criteria.

1. INTRODUCTION

PathCrawler [8][15] was one of the first test input generators to use a combination of concrete data and symbolic execution. In the literature, similar test generation tools are variously referred to as concolic, dynamic-symbolic-execution (DSE) or constraint-based execution tools. Below we will call them *concolic* tools. Unlike some concolic tools, such as [13], PathCrawler does not use concrete values to generate over-approximate path predicates.

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However, PathCrawler is concolic in that, like these tools and [14], [4], [2], PathCrawler recovers the trace of each generated test and uses it to generate a prefix of the path predicate of the next test. This is an efficient way to generate tests for all-feasible-path coverage, the structural coverage criterion PathCrawler was designed to satisfy, because the constraint solver is only called once for the initial test and then once for each node in the tree of feasible execution paths. Indeed, although constraint resolution is very fast most of the time, it is actually NP-complete and it is very difficult to know which constraint problem will take “too long” to resolve. This means that every time the constraint solver is called, there is a risk that it will run for “too long” and have to be interrupted by a timeout condition. This is why it is important to limit calls to the solver in order to speed up test generation. In structural testing, the minimum number of tests is defined by the structural coverage criterion and the tested function, so the only way to limit calls to the constraint solver is to limit the calls which do not contribute to this minimum number of tests. These are either calls which do not result in a test being generated, because the constraint problem is inconsistent (i.e. the path is infeasible), or else calls which generate a test which does not cover anything that has not already been covered by previous tests. PathCrawler limits the first category by always detecting infeasibility in the shortest prefix which is common to several infeasible paths. This paper is about limiting the second category.

PathCrawler generates tests to cover all feasible execution paths of functions coded in ANSI C (except functions containing certain constructions not treated yet, essentially pointer casts). However, for many functions the all-paths coverage criterion demands unmanageable numbers of tests. This phenomenon may be intrinsic to the structure of the tested function, for example if it contains numerous successive conditional instructions with few infeasible combinations (so the number of paths approaches 2 to the power of the number of conditional instructions), or loops containing conditional instructions (so that the number of potentially feasible paths is the number of paths through the loop body to the power of the number of iterations). In such cases, a branch-coverage criterion may be more appropriate.

However the number of paths to cover also depends on how paths are defined. Indeed, structural coverage criteria are not always defined very precisely in the literature, which can pose a problem for the practitioner. For instance, if called functions are treated as though they are in-lined in the code then they may cause a combinatorial explosion in the number of paths whereas coverage of the feasible paths through the tested function itself, without necessarily “covering” the called functions, may only demand a manageable number of tests. PathCrawler decomposes multiple

conditions and then tries to cover the resulting expanded tree of feasible execution paths, which can also cause a combinatorial explosion in the number of “paths” whereas the number of paths through the different decisions of the multiple conditions may be manageable.

We would like to adapt PathCrawler to respect other control-flow-based structural coverage criteria in order to be applied to the programs for which the current criterion demands too many tests. As a first step, the alternative criteria which could be satisfied by a concolic generation strategy must be precisely defined and their interest to the user must be justified.

Moreover, we have also been investigating [16] the use of PathCrawler to generate tests to measure worst-case execution time (WCET). For this purpose, we have devised coverage criteria which exclude paths which, if certain hypotheses are respected, must have shorter execution times than the others. In these criteria, we may cover only the true branch of if-then-else structures with an empty else body or only the maximum number of iterations of loops with no condition in the loop body.

One way to respect coverage criteria other than all-paths would be to use the classic concolic path test generation strategy and just stop test generation when the criterion had been satisfied, e.g. in the case of branch coverage, when all branches had either been covered or proved unreachable. However, this is likely to be inefficient in the following sense. By *inefficient* test generation, we mean that many tests are generated and infeasible partial paths detected which are *redundant* in the sense that although they cover (or, in the case of infeasible partial paths, could have covered) new paths they do not increase coverage (resp. could not have increased coverage) as defined by the criterion in question, for example branch coverage. Redundant tests and redundant infeasible partial paths cost potentially expensive calls to the constraint solver and must be limited. Classic concolic test generation strategies explore the path tree “blindly” and so if, for example, only one branch remains to be covered, they may waste time exploring partial paths which are not even connected to that branch.

```

1      int g(int i, int x){
2          if (i == x)
3              return 2;
4          else
5              return (i*x)+1;
6      }
7
8      int f( int A[2], int e, int x) {
9          int i, res ;
10         res = 0 ;
11         if((x < -1) || (x > 1)) {
12             i = 0;
13             while( (i < 2) && (res == 0)) {
14                 if( e == A[i] )
15                     res = g(i+1,x);
16                 else
17                     i++; } }
18         if(res == 2)
19             return 1;
20         else
21             return 0;
22     }

```

Figure 1.Source code of an example of a tested function, *f*.

In this paper, we present an abstraction of execution paths, “abstract paths”, which provides a conceptual framework to facilitate the definition and comparison of many different control-flow-based structural coverage criteria and of concolic generation strategies to efficiently satisfy these criteria. Structural coverage criteria are often said to be based on the program’s control-flow graph and abstract paths encapsulate parts of the control flow graph. However, the control-flow graph does not treat multiple conditions as we would like to and trees of abstract paths can also be compared to abstract syntax trees. We first used abstract paths for the WCET measurement criteria described in [16]. In the present paper, the idea of abstract paths is revised and generalised so that it can be used for other criteria. In the next section, we will first recall the tree of feasible execution paths explored by concolic test generation tools and introduce an example of a tested function and its feasible paths. In Section 3, we define abstract paths and in Section 4 we show how the criteria mentioned above can be defined in terms of these paths. In Section 5 we consider strategies to explore the abstract path graph in order to efficiently generate tests satisfying each criterion. Finally, we will discuss related work and future directions.

2. THE EXPANDED TREE

The classic concolic test generation strategy is an exhaustive exploration of a tree we will call the fully expanded tree of feasible execution paths (or expanded tree). To generate tests to satisfy other coverage criteria, the exploration of the whole of this tree should not usually be necessary. In order to discuss this further, we start by defining how the source code of the tested function is represented in this tree.

We suppose here that there are no system or library calls or GOTO instructions in the original source code and that it has been simplified so that it only contains if-then-else instructions with simple conditions, sequences of assignments and GOTO instructions added by the simplification. All conditional control instructions such as if-then-else, switch, while,... have been decomposed so that the only conditional instructions left have simple conditions containing no logical connectors (such as &&) or side effects (assignments or function calls). Function calls have been replaced by assignments of the values of the effective parameters to the formal parameters, followed by the source code of the called function and then assignment of the return value.

What we call the *expanded tree* is in fact the tree of feasible execution paths through this simplified source code. It is composed of a root node, leaf nodes, conditional nodes and directed arcs between nodes. The *root node* represents the entry point of the tested function and each *leaf node* represents exit from the function. Each *conditional node* represents an if-then-else instruction with a simple condition. *Arcs* represent a truth value (true or false) and a (possibly empty) sequential block of unconditional instructions. There is a single arc (with value “true”) from the root to the first node and from the last node in each path to a leaf. Each conditional node has one arc entering it and either one or two arcs leaving it. Loops are unrolled. Each *path* from the root through connected arcs and nodes to a leaf represents a feasible execution path. A path which starts at the root and ends with an arc is called a *partial path*. If a conditional node in the expanded tree only has one arc leaving it, then the missing arc would be the final arc in an *infeasible partial path*.

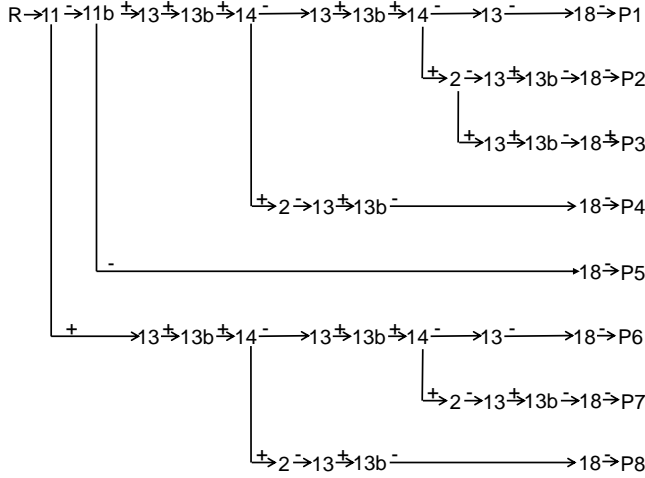


Figure 2. The expanded tree for f

As a running example of a tested function, we use the function f whose source code is displayed in Figure 1. In Figure 2 we show the expanded tree for this example. The root of this tree is labelled R and the leaves are labelled with the unique identifier of the feasible path ($P1, P2, \dots$) leading to this leaf. Each conditional node in this tree is labelled with line number of the conditional instruction in the source code that it represents, followed by the letter “b” if the node represents the second sub-condition in a multiple condition. The arcs are labelled with $+$ for true and $-$ for false but the assignments are not shown. In the following, we will denote the positive arc leaving a node labelled n by $n+$ and the negative arc leaving the same node by $n-$.

The concolic exploration of the expanded tree starts with an arbitrary feasible path and then for each unexplored arc leaving a conditional node in this path it arbitrarily selects a feasible suffix (up to exit from the tested function) unless the partial path up to and including this unexplored arc is infeasible. The feasible suffix is explored in the same way. The arcs can be explored in any order. Figure 2 numbers the paths in an order which illustrates a possible concolic exploration of the arcs in depth-first order, supposing that the first, arbitrarily obtained, path is $P1$.

3. ABSTRACT PATHS

To introduce abstract paths, let us consider a test criterion which requires just the coverage of all feasible paths through the source code seen in a form in which called functions are not in-lined, nor multiple conditions decomposed. We will call this criterion *minimal-all-paths*. To respect minimal-all-paths for our example, it is not necessary to cover both $P6$ and $P1$, which only differ in the sub-conditions ($11-$ and $11b+$ in $P1$ and $11+$ in $P6$) leading to the same positive decision for the multiple condition on line 11. We see that for this criterion, for each path (such as $P1$) in the expanded tree which is covered, there may be a set of feasible paths (such as $P1$ and $P6$) which are equivalent. Below, we will call this set of paths an abstract path for the minimal-all-paths criterion. Similarly, if the test criterion were coverage of all simple conditions, then after covering the first path, $P1$, the condition $14-$ is covered twice but not $14+$. We would then try to cover either of the two partial paths ending in $14+$ which is obtained by modifying one of the two prefixes of $P1$ which ends

in $14-$ (and the next path covered would then be $P2, P3$ or $P4$). In this case, we would consider both partial paths as equivalent in spite of the fact that they have different loop iterations (sequences of $13+$, $13b+$ and $14+$ or $14-$).

Indeed, as shown by Figure 2, the same conditional instruction in the simplified code is usually represented by several different nodes in the expanded tree and two path fragments in this tree can be considered as equivalent, by certain test criteria and under certain conditions, if they start with the same conditional node, s , or different conditional nodes, $s1$ and $s2$, representing the same conditional instruction and end with conditional nodes $t1$ and $t2$ both representing some other conditional instruction. For example, $P1$ and $P6$ in Figure 2 both contain a path fragment starting at the node, s , labelled 11 and ending at nodes $t1$ and $t2$ labelled 13. Such equivalent path fragments are produced when one of the following constructions is present in the un-simplified source code:

- If-then-else structures
- Function calls
- Loops
- Multiple conditions

In order to be able to define, compare and implement criteria such as minimal-all-paths, we propose to use trees which integrate the structural information found in the control flow graph and the abstract syntax tree by grouping certain parts of the expanded tree into *abstract nodes* of the following types:

ITE This abstract node has one arc entering it, one arc leaving it and contains a conditional node and the two alternative path fragments.

CALL This has one arc entering it, one arc leaving it and contains all path fragments through the called function.

LOOP This has one arc entering it, one arc leaving it and contains the conditional node of the loop head and all path fragments for one individual iteration

AND This has one arc entering it and two arcs, with two different truth values, leaving it. It contains the path fragment for the logical conjunction to be satisfied and the two path fragments for it to be false.

OR This also has one arc entering it and two arcs, with two different truth values, leaving it. It contains the two path fragments for the logical disjunction to be satisfied and the path fragment for it to be false.

Note that abstract nodes can be nested so the conditional node in an if-then-else can be a multiple condition and the alternative path fragments can contain other abstract nodes.

Abstract trees are composed of the same root, leaves, arcs and conditional nodes as the expanded tree but they also contain abstract nodes. In Figure 3 we depict an abstract tree for our example which contains all the types of abstract node mentioned above. This tree can be constructed by (a) constructing the control flow graph of the tested function with all called functions in-lined and all multiple conditions decomposed and (b) enclosing the control structures from the abstract syntax tree in abstract nodes. Other abstract trees can be defined for f , as we will see below, by not using all the types of abstract node.

As different path fragments within abstract nodes can lead to a common node, they are not depicted as a tree but as paths through a graph which is similar to a fragment of the control flow graph. Note that within an abstract node, one node, such as the node labelled 14+ in Figure 3, may represent several nodes in the expanded tree (in this case, all the nodes labelled 14+ in the expanded tree).

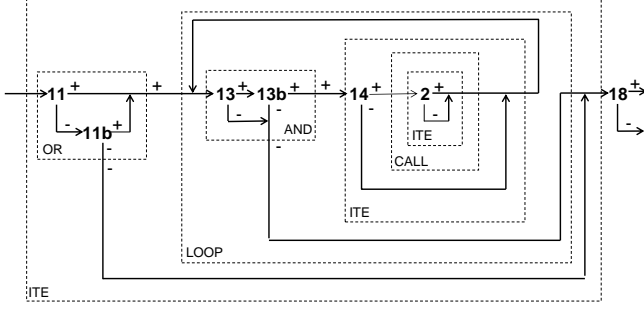


Figure 3. An abstract tree for f

Abstract paths are paths from the root to a leaf of an abstract path tree. Note that when an abstract path goes through a conditional abstract node (i.e. AND or OR), it follows just one arc out (true or false), as in the case of non-abstract conditional nodes in concrete paths. Each abstract path represents a set of concrete paths in the expanded tree. From now on, we will refer to paths in the expanded tree as *concrete paths*. The abstract tree depicted in Figure 3 contains two abstract paths, one representing all feasible paths whose final arc represents the truth of the conditional instruction at line 18, and the other representing all the feasible paths for which this condition is false. As abstract nodes can be nested, an abstract path can contain path fragments which themselves traverse abstract nodes but we will refer to these as *abstract path fragments* and not abstract paths.

Note that the expanded tree only represents feasible paths but the feasibility of a path fragment in an abstract node may depend on the partial path leading to the abstract node if this partial path contains other abstract nodes. We consider that each arc in an abstract tree (including the arcs in the abstract nodes) is reachable, i.e. is present in at least one concrete path.

An *abstract partial path* is a path from the root to an arc which leaves a non-abstract conditional node. Note that if the abstract partial path ends in an arc which is contained in an abstract node, n , (or in nested abstract nodes $n1, n2, \dots$) then n (resp. $n1, n2, \dots$) must be expanded in the abstract partial path.

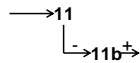


Figure 4. The abstract partial path to arc 11b+

For example, Figure 4 depicts the abstract partial path to the arc labelled 11b+ in the abstract tree depicted in Figure 3, which corresponds to a single concrete partial path because the abstract logical disjunction node has been expanded. In the case of an arc contained in an abstract loop node, the abstract partial path to the arc may represent concrete paths with several different iterations before the final arc is reached, and with different numbers of iterations before the final arc. This is why we depict such abstract

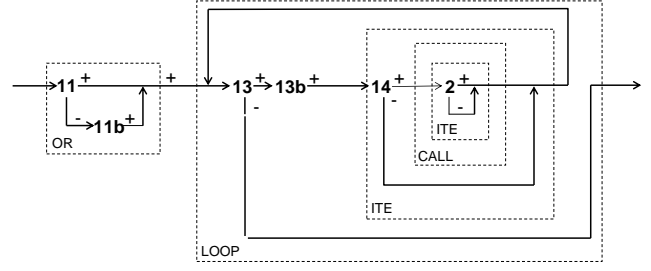


Figure 5. The abstract partial path to arc 13-

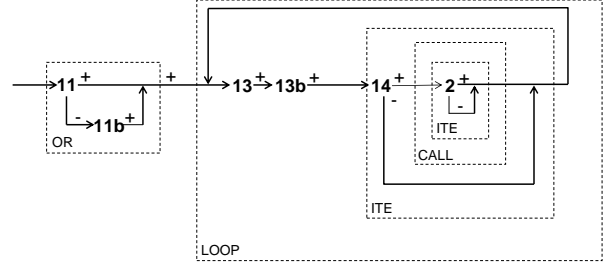


Figure 6. Abstract partial path to arcs 13+, 13b+, 14+, 2+, 2-

partial paths as in Figure 5, which shows the abstract partial path leading to the arc 13- in the abstract tree depicted in Figure 3. The abstract partial paths leading in Figure 3 to the arcs 13+, 13b+, 14+, 14-, 2+ and 2- are all depicted by Figure 6.

4. USE OF ABSTRACT TREES TO DEFINE COVERAGE CRITERIA

Let us now see how different control-flow-based structural coverage criteria can be defined in terms of different abstract trees.

The *minimal-all-paths* criterion was already discussed above. It corresponds to the coverage of all abstract paths in an abstract tree in which only function calls and multiple conditions are encapsulated in abstract nodes. If-then-else structures and loops are left in their expanded form in this tree. Figure 7 depicts this abstract tree for our example.

If the tested function has too many abstract paths even when the function calls and multiple conditions are abstracted then the abstraction of loops can be considered. One criterion commonly used in this case is k -path, in which k is a small integer constant fixed by the user and only the only paths covered are those with up to k iterations of any loop with a variable number of iterations. However, this criterion does not take into account the branches within each iteration. Also, if any paths are only feasible when a path contains more than k iterations of one of these loops, then these paths will not be tested. Abstract loop nodes allow other, more justifiable, criteria to be defined.

If even the abstraction of function calls, multiple conditions and loops leaves too many paths to cover, then the user can decide to cover either all simple conditions, including the sub-conditions of multiple conditions, or just all decisions. We call *minimal-all-*

[illegible]

We will call the next criterion *minimal-all-decisions*. It is coverage of all branches appearing in the tested function's original source code, in which multiple conditions are not decomposed and function calls are not in-lined. This criterion corresponds to the coverage, in the same abstract tree as the previous criterion, of all abstract partial paths in which the last arc leaves either a simple condition which is not contained in an abstract CALL, AND or OR node or else leaves an abstract AND or OR node which is not contained in another abstract CALL, AND or OR node (in our example, the abstract partial paths ending in the arcs 14+, 14-, 18+, 18- or one of the arcs leaving the AND or OR abstract nodes).

single abstract ITEE node represents both the concrete path, p_0 , which takes the empty path fragment of this ITEE and the concrete paths p_1, p_2, \dots that take a non-empty path fragment. p_0 is “shorter than” p_1, p_2, \dots and does not need to be covered. Note that p_1, p_2, \dots are not comparable because the order is only partial. Similarly, if there are several ITEEs in the same abstract path, then all the concrete paths which take a non-empty path fragment in at least one ITEE are not comparable with each other. *Empty-else* and *max-iterations* are satisfied if, for each abstract path in the tree, we cover all concrete paths which are maximal in the corresponding partial order over this abstract path.

Now let us show how abstract trees can be used to define concolic strategies to generate tests without exploring as many feasible paths or infeasible partial paths as the classic depth-first concolic strategy.

Note that our proposed strategies also take into account one reason why PathCrawler is efficient: because it is implemented using constraint logic programming, depth-first search makes the most of Prolog’s built-in backtrack mechanism to store successive states of the constraint store in a stack instead of recalculating them on backtrack.

A concolic strategy consists in first defining the order in which to inspect the arcs in the concrete paths already covered. Then, for each arc, a , which is in a concrete path, P , which has already been covered, the strategy must define whether and how to explore the alternative arc, a' . Let PP be the concrete prefix of a in P and PP' be the concrete partial path formed by adding a' to the end of PP . The strategies defined using abstract trees are based on the following principles:

1. If PP' belongs to an abstract partial path which has not yet been covered, then try to generate any test covering PP'
2. If PP' is only a prefix of concrete (partial) paths which belong to already covered abstract (partial) paths, then there is no need to cover PP' .
3. If PP' is a prefix of at least one concrete (partial) path, PPP' , belonging to an abstract (partial) path that has not yet been covered, then decide how to try and cover PPP' and all other concrete (partial) paths of which PP' is a prefix and which belong to uncovered abstract (partial) paths.

Point (2) above defines when PP' does not need to be covered and points (1) and (3) concern different ways to try and cover PP' .

We propose to divide arc-based strategies into two passes. In a first breadth-first traversal, only the partial paths corresponding to point (1) above are explored. Then the still uncovered partial paths are reviewed and, in a second pass, those corresponding to point (3) above are explored: i.e. all concrete partial paths belonging to each uncovered abstract partial path are systematically explored until either a test to cover the abstract partial path is found or all the concrete partial paths belonging to it have been proved infeasible. The justification for breadth-first search and this two-phase scheme is opportunistic: in covering

PP' , we cover a whole new concrete path, P' , of which PP' is a prefix, and P' may contain other uncovered arcs as well as a' . (or in other words, P' may have other prefixes which belong to different abstract partial paths). If we explore covered paths breadth-first then we first explore the alternatives of arcs at the beginning of paths. These are more likely to have long suffixes, thereby increasing the chances of quickly finding other uncovered arcs. Conversely, only one path can be covered at a time, whether it is abstract or concrete, so for path-based strategies we propose to take advantage of the efficiency of depth-first search.

Now let us see how this general scheme can be instantiated for the first two criteria discussed in Section 4. Note that the precise details of the exploration of the arcs corresponding to point (3) above are just proposed as an example; they could be explored in other ways.

5.1 Minimal-all-paths

This strategy is illustrated on our example in Figure 8. The covered paths are labelled P or R (if the test is redundant), and the infeasible partial paths are labelled I. Each path or partial path is denoted by its successive arcs and each path fragment which is inside an abstract node is enclosed in brackets, as follows: $\langle \rangle$ for OR, $\langle \rangle$ for AND and $\{ \}$ for CALL. In covered paths, the explored

$\langle 11-11b \rangle + \langle 13+13b \rangle + 14-$	$\langle 13+13b \rangle + 14-$	$\langle 13- \rangle - 18- : P1$
$\langle 11-11b \rangle + \langle 13+13b \rangle + 14-$	$\langle 13+13b \rangle + 14-$	$\langle 13- \rangle - 18+ : I1$
$\langle 11-11b \rangle + \langle 13+13b \rangle + 14-$	$\langle 13+13b \rangle + 14-$	$13+ : I2$
$\langle 11-11b \rangle + \langle 13+13b \rangle + 14-$	$\langle 13+13b \rangle + 14+ \{2-\}$	$\langle 13+13b \rangle - 18- : P2$
$\langle 11-11b \rangle + \langle 13+13b \rangle + 14-$	$\langle 13+13b \rangle + 14+ \{2-\}$	$\langle 13+13b \rangle - 18+ : I3$
$\langle 11-11b \rangle + \langle 13+13b \rangle + 14-$	$\langle 13+13b \rangle + 14+ \{2-\}$	$\langle 13+13b \rangle + : I4$
$\langle 11-11b \rangle + \langle 13+13b \rangle + 14-$	$\langle 13+13b \rangle + 14+ \{2-\}$	$\langle 13- \rangle - 18+ : I5$ <i>only explored with suffix 18+</i>
$\langle 11-11b \rangle + \langle 13+13b \rangle + 14-$	$\langle 13+13b \rangle + 14+ \{2+\}$	$\langle 13+13b \rangle - 18+ : P3$
$\langle 11-11b \rangle + \langle 13+13b \rangle + 14-$	$\langle 13+13b \rangle + 14+ \{2+\}$	$\langle 13+13b \rangle - 18- : not explored$
$\langle 11-11b \rangle + \langle 13+13b \rangle + 14-$	$\langle 13+13b \rangle + 14+ \{2+\}$	$\langle 13+13b \rangle + : I6$
$\langle 11-11b \rangle + \langle 13+13b \rangle + 14-$	$\langle 13+13b \rangle + 14+ \{2+\}$	$\langle 13- \rangle - : not explored$
$\langle 11-11b \rangle + \langle 13+13b \rangle + 14-$	$\langle 13+13b \rangle -$	$: I7$
$\langle 11-11b \rangle + \langle 13+13b \rangle + 14-$	$\langle 13- \rangle -$	$: I8$
$\langle 11-11b \rangle + \langle 13+13b \rangle + 14+ \{2-\}$	$\langle 13+13b \rangle -$	$18- : P4$
$\langle 11-11b \rangle + \langle 13+13b \rangle + 14+ \{2-\}$	$\langle 13+13b \rangle -$	$18+ : I9$
$\langle 11-11b \rangle + \langle 13+13b \rangle + 14+ \{2-\}$	$\langle 13+13b \rangle +$	$: I10$
$\langle 11-11b \rangle + \langle 13+13b \rangle + 14+ \{2-\}$	$\langle 13- \rangle -$	$18+ : I11$ <i>only explored with suffix 18+</i>
$\langle 11-11b \rangle + \langle 13+13b \rangle + 14+ \{2+\}$		$: I12$
$\langle 11-11b \rangle + \langle 13+13b \rangle -$		$: I13$
$\langle 11-11b \rangle + \langle 13- \rangle -$		$: I14$
$\langle 11-11b \rangle -$		$18- : P5$
$\langle 11-11b \rangle -$		$18+ : I15$
$\langle 11+ \rangle + \langle 13+13b \rangle + 14-$	$\langle 13+13b \rangle + 14-$	$\langle 13- \rangle - 18- : R1$ <i>same abstract path as P1</i>
$\langle 11+ \rangle + \langle 13+13b \rangle + 14-$	$\langle 13+13b \rangle + 14-$	$\langle 13- \rangle - 18+ : I16$
$\langle 11+ \rangle + \langle 13+13b \rangle + 14-$	$\langle 13+13b \rangle + 14-$	$13+ : I17$
$\langle 11+ \rangle + \langle 13+13b \rangle + 14-$	$\langle 13+13b \rangle + 14+ \{2-\}$	$\langle 13+13b \rangle - 18- : R2$ <i>same abstract path as P2</i>
$\langle 11+ \rangle + \langle 13+13b \rangle + 14-$	$\langle 13+13b \rangle + 14+ \{2-\}$	$\langle 13+13b \rangle - 18+ : not explored$
$\langle 11+ \rangle + \langle 13+13b \rangle + 14-$	$\langle 13+13b \rangle + 14+ \{2-\}$	$\langle 13+13b \rangle + : I18$
$\langle 11+ \rangle + \langle 13+13b \rangle + 14-$	$\langle 13+13b \rangle + 14+ \{2-\}$	$\langle 13- \rangle - : not explored$
$\langle 11+ \rangle + \langle 13+13b \rangle + 14-$	$\langle 13+13b \rangle + 14+ \{2+\}$	$\langle 13+13b \rangle + : I19$ <i>only explored with suffix 13+, 13b+</i>
$\langle 11+ \rangle + \langle 13+13b \rangle + 14-$	$\langle 13+13b \rangle -$	$: I20$
$\langle 11+ \rangle + \langle 13+13b \rangle + 14-$	$\langle 13- \rangle -$	$: I21$
$\langle 11+ \rangle + \langle 13+13b \rangle + 14+ \{2-\}$	$\langle 13+13b \rangle -$	$18- : R3$ <i>same abstract path as P4</i>
$\langle 11+ \rangle + \langle 13+13b \rangle + 14+ \{2-\}$	$\langle 13+13b \rangle -$	$18+ : I22$
$\langle 11+ \rangle + \langle 13+13b \rangle + 14+ \{2-\}$	$\langle 13+13b \rangle +$	$: I23$
$\langle 11+ \rangle + \langle 13+13b \rangle + 14+ \{2-\}$	$\langle 13- \rangle -$	$18+ : I24$ <i>only explored with suffix 18+</i>
$\langle 11+ \rangle + \langle 13+13b \rangle + 14+ \{2+\}$		$: I25$
$\langle 11+ \rangle + \langle 13+13b \rangle -$		$: I26$
$\langle 11+ \rangle + \langle 13- \rangle -$		$: I27$

Figure 8. Minimal-all-paths strategy illustrated on our example

arc a' is underlined. We assume the arbitrary first path is still P1 and maintain the same order of paths as in Figure 2 because this strategy is also based on depth-first search. The infeasible partial paths in Figure 8 are also traversed in the same order as they would be in exhaustive depth-first search of the expanded tree, but in this strategy some partial paths do not need to be explored, whilst some only need to be explored with a particular suffix.

For P1 in this example, a is 18- and PP' is I1, which belongs to a different abstract path from P1 and so corresponds to point (1) above. For I1, a is 13- and PP' is I2, which corresponds to point (3) above because with the suffix 13b+ 18+ it belongs to one uncovered abstract path, but with the suffix 13b+ 18- it belongs to another and with the suffix 13b- 18+ to yet another. In order to explore the concrete suffixes of PP' , we choose here to start by trying to cover the shortest common suffix, which in this case is PP' itself. This choice of implementation risks generating redundant tests, as seen later on in this example, but avoids repeated failures. The PP' corresponding to point (2) above are labelled *not explored* in Figure 8.

At the end of this example, 4 fewer infeasible partial paths have been explored than with an exhaustive exploration of the expanded tree. If we had tried to just stop an exhaustive exploration once this criterion were satisfied, we would not have saved any exploration, because the infeasibility of some of the abstract paths is only proved at the end of the exploration.

5.2 Minimal-all-conditions

Our proposed test generation strategy for this criterion is illustrated on our example in Figure 9, in which LOOP path fragments are enclosed in () and ITE path fragments in []. For this criterion, if a is not contained in a called function and does not already appear negated in one of the paths covered so far, then the strategy always tries to generate a test covering PP' . The first phase of the strategy explores all these cases breadth-first, as illustrated in Figure 9 until the failure to negate arc 18+ in I3. The second phase then considers all arcs a contained in a called function or which already appear negated in one of the paths covered so far but for which a' could be an alternative way to cover an abstract partial path (in this case, the abstract partial path leading to 18+).

At the end of this example, 10 infeasible partial paths and 1 redundant feasible path (a total of 11) have been explored whereas if exhaustive search of the expanded tree had just been stopped once this criterion were satisfied, after covering P6, then 17 infeasible partial paths and 1 redundant feasible path (a total of 18) would have been explored.

6. CONCLUSION

We propose that the fully expanded tree of feasible execution paths currently used by concolic test generation tools be replaced by abstract trees which integrate structural information from the control flow graph and abstract syntax tree. Abstract trees facilitate the precise definition of what is covered or not by each structural criterion, and also the design of test data generation strategies which either prioritise exploration of the tree, by only exploring abstract nodes if necessary, or prioritise certain paths within an abstract node, such as non-empty ITE branches.

Indeed, concolic test generation tools can only become really useful if they can be applied to a reasonably large class of programs. This implies that efficient variants of their classic exploration strategy must be found which still retain the advantages of the concolic approach.

There has been much recent work on this subject, but each time from the point of view of a particular cause of “path explosion”. In [7], it is function calls which are abstracted and the authors also use the term “abstract path”. However, they use it for the path described by a predicate in which the input-output relations of called functions are not known. In our terms, this is equivalent to a path through the tested function in which function calls are encapsulated in an abstract node, but before any paths through the abstract node are known. Their proposed strategy consists in first exploring concolically all “abstract paths” which are feasible when the function calls are replaced by stubs which can return any value. Then the real called functions are put back in place of the stubs and for each of these “abstract paths”, and for each function call, different concrete paths through the real called function are explored concolically until one is found that is consistent with the rest of the “abstract path”. [1] does not use the term “abstract path” but it also manipulates path predicates in which we can consider that the path through called functions is abstracted. However the predicates of [1] only characterise abstract paths for

```
[ <11- 11b+>+ «13+ 13b+»+ [14- ] «13+ 13b+»+ [14- ] «13- »- ] 18- : P1
[ <11+ >+ «13+ 13b+»+ [14- ] «13+ 13b+»+ [14- ] «13- »- ] 18- : P2
[ <11- 11b- >- ] 18- : P3
  <11- 11b+>+ 13+ 13b- : I1
  <11+ >+ 13+ 13b- : I2
[ <11+ >+ («13+ 13b+»+ [14+ {2-}] «13+ 13b- »- )] 18- : P4
[ <11+ >+ («13+ 13b+»+ [14+ {2-}] «13+ 13b- »- )] 18+ : I3
[ <11- 11b- >- ] 18+ : I4
[ <11+ >+ «13+ 13b+»+ [14- ] «13+ 13b+»+ [14- ] «13- »- ] 18+ : I5
[ <11- 11b+>+ «13+ 13b+»+ [14- ] «13+ 13b+»+ [14- ] «13- »- ] 18+ : I6
  <11- 11b+>+ «13+ 13b+»+ [14- ] «13+ 13b+»+ [14- ] 13+ : I7
[ <11- 11b+>+ («13+ 13b+»+ [14- ] «13+ 13b+»+ [14+ {2-}] «13+ 13b- »- )] 18- : R1
[ <11- 11b+>+ («13+ 13b+»+ [14- ] «13+ 13b+»+ [14+ {2-}] «13+ 13b- »- )] 18+ : I8
[ <11- 11b+>+ («13+ 13b+»+ [14- ] «13+ 13b+»+ [14+ {2-}] «13+ 13b+ »+ )] 18+ : I9
[ <11- 11b+>+ «13+ 13b+»+ [14- ] «13+ 13b+»+ [14+ {2-}] «13- »- ] 18+ : I10
[ <11- 11b+>+ («13+ 13b+»+ [14- ] «13+ 13b+»+ [14+ {2+}] «13+ 13b- »- )] 18+ : P5
```

Figure 9. Minimal-all-conditions strategy illustrated on our example

which at least one path through each called function is known. The strategy proposed in [1] is more efficient than that of [7]. It memorises the predicates of all known paths through called functions, along with the calling context predicates. This means it only needs to concolically explore alternative paths through a called function if it fails to construct a feasible predicate by inserting a previously memorised called function path predicate into the predicate of the path through the tested function. In our own previous work based on formal specifications of library functions [11], we also proposed a test generation algorithm to avoid unnecessary exploration of the specifications.

In other work, the focus is on adapting the strategies of particular tools in order to obtain more efficient statement or branch coverage, although their aim may not be to completely satisfy a formally defined criterion and so they may not be concerned, as we are above, about demonstrating the unreachability of the uncovered statements or branches. In [3], heuristics are proposed to decide whether to explore a branch which has already been covered, but with a different prefix. However, the first heuristic proposed is the connection to an uncovered branch, rather as in our minimal-all-conditions strategy of Section 5 above, and [3] discusses the most efficient way to compute this information. In [14] the goal is to quickly cover most reachable statements and the program structure in terms of “building blocks such as methods and loops” is taken into account in the definition of a fair choice between unexplored branches. In [5], when concolic depth-first exploration arrives at a “context-sensitive program point” (such as an exit from an ITE) the exploration of the part of the execution-path tree which is rooted at this program point is used to discover which variables are live at this point. These variables are memorised along with their intersection with the current program state (path constraint and concrete memory state). On the next traversal of the same program point, this memorised dependence data is retrieved and used to decide whether exploration of a different path through the ITE will only result in covering the same suffixes (or in our terms, the same abstract paths).

This previous work may well achieve greater test generation efficiency than the test generation strategies based on abstract trees which we propose here and which just allow control dependencies to be taken into account in order to save some unnecessary constraint resolution. However, we believe that abstract trees will also provide a convenient and unified framework in which to implement the next step, which is to take data dependencies into account to save even more unnecessary exploration. This will be the focus of our future work.

An alternative to the concolic approach is to use search-based techniques to generate structural test data. Similar problems arise, e.g. the treatment of function calls [6], the use of control [12] and data [10] dependencies and the need to explore individual paths to a target [9], which can be compared to the exploration of abstract nodes. It would be interesting to see whether abstract trees could also be used in this setting.

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