

Research Proposal for Ph.D. Thesis:
DIFFERENTIAL PROGRAM ANALYSIS

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1 Introduction

Our research presents a novel technique for identifying and describing semantic change in programs while relying on abstract interpretation and program analysis methods [6]. We define the problem of *semantics differencing* as identifying observable difference in behavior for executions originating from *the same input*. The notion of observable difference can be trivially defined as difference in output values in the final state [21, 8, 22, 17], but our research extends this notions to detect other important behavioral differences such as array access patterns, assertion failures, etc.

Differential analysis has received attention in classical work (e.g., [10, 11, 13]) and recently been the subject of research in program analysis [21, 8, 22, 17]. Hoare [9] defined program equivalence and differencing as a "grand challenge" to the verification community and recently suggested [18] several applications where static analysis can be easily adopted as an effective solution for tasks such as equivalence checking, semantic diffing, differential contract checking, patch debugging, etc. Specifying semantic difference between program versions has seen growing interest for various applications ranging from testing of concurrent programs [4], understanding software upgrades [14], to automatic generation of security exploits [2].

In oppose to existing recent techniques for semantic differencing, that rely on variants of symbolic execution and SMT solving [21, 8, 22, 17], our technique employs abstract interpretation towards identifying difference. This has the potential to scale better and reach better coverage than existing techniques (and specifically provide a solution for handling loops). Using abstract interpretation for differential analysis means reasoning over two programs. As this problem is trivially solved by symbolic execution techniques by simply analyzing program sequentially, we assume that one of the reasons no previous abstract interpretation work exists for this topic, is that separately (or sequentially) abstracting two programs, cannot be directly used towards calculating equivalence or difference. Equality of two abstract program states does not ensure concrete equality and furthermore, the notion of difference between abstract program states is vague. Our work overcomes these obstacles by:

- Presenting a method for abstract interpretation of a pair of programs (P, P') for *sound* semantic equivalence and differencing by abstracting direct relationships between (P, P') variables in a partially disjunctive domain. We describe a partitioning technique for state reduction and scaling. We define

a widening operator for abstracting unbound paths in our domain.

- Phrasing a new technique for syntactically interleaving a pair of programs (P, P') for the creation of a *correlating program* $P \bowtie P'$ which contains the semantics of both programs. We propose an analysis over the program for characterizing program equivalence and difference, based on the aforementioned abstraction, given the properties of the correlating program which aligns (P, P') executions.

We have implemented our approach in a tool based on the LLVM compiler infrastructure and the APRON numerical abstract domain library, and evaluated it using select patches from open-source software including GNU core utilities, Mozilla Firefox, and the Linux Kernel. Our evaluation shows that the tool often manages to establish equivalence, reports useful approximation of semantic differences when differences exists, and reports only a few false differences.

We plan further research for refining and adjusting our technique to achieve better results on a wider range of programs by extending our analysis to be interprocedural, using several abstract domains to capture differences in more complex data structures (pointers, heaps) and differencing concurrent programs.

We do not limit ourselves purely to abstract interpretation techniques and one major future direction of our work will include applying existing dynamic analysis and symbolic execution techniques on correlated programs. The interleaved structure offered by the correlating program holds great potential as it alleviates the need to "delay" differencing until both programs have been analyzed completely (this will be further explained in Section 5). Another major research direction will be different compositions of the correlating program that will allow for better differencing. Specifically, an equivalence-guided correlation refinement technique can be used i.e. the way the programs are interleaved will be directed by the analysis according to equivalence criteria.

As our results so far have limited use as they basically offer an abstraction of program difference at a certain program location, we also plan to produce techniques for using our results towards practical applications such as regression test generation, patch installation debugging and synthesis of attacks from a patch.

2 Overview of Existing Research

In this section, we provide an informal overview of our existing research using a simple example.

```

int sign(int x) {
  int sgn;
  if (x < 0)
    sgn = -1;
  else
    sgn = 1;
  return sgn;
}

int sign'(int x) {
  int sgn';
  if (x' < 0)
    sgn' = -1;
  else
    sgn' = 1;
  if (x' == 0)
    sgn' = 0;
  return sgn';
}

```

Figure 1: Two simple implementations of the *sign* operation.

Consider the two programs of Fig. 1, inspired by an example from [23]. For these programs, we would like to establish that the output of *sign* and *sign'* only differs in the case where $x = 0$ and that the difference is $sgn = 1 \neq sgn' = 0$ as generally described in Fig. 2.

$x.x'$ constraints	sgn	sgn'
$x < 0$	$sgn \mapsto -1$	$sgn' \mapsto -1$
$x = 0$	$sgn \mapsto 1$	$sgn' \mapsto 0$
$x > 0$	$sgn \mapsto 1$	$sgn' \mapsto 1$

Figure 2: behavior of *sign* and *sign'*.

2.1 Equivalence Under Abstraction

As a first naive attempt to achieve this, one could try to analyze each version of the program separately and compare the (abstract) results. However, this is clearly unsound, as equivalence under abstraction does not entail concrete equivalence. For example, using an interval analysis [7] would yield that in both programs the value of variable `sgn` ranges in the same interval $[-1, 1]$, missing the fact that *sign* never returns the value 0.

To establish equivalence under abstraction, we need to abstract *relationships between the values of variables* in *sign* and *sign'*. Specifically, we need to track the relationship between the values of `sgn` in both versions. This requires a joint representation in which these relationships can be tracked.

2.2 Correlating Program

As a first step, our goal is to construct a program where we can track relationships between variables of *sign* and *sign'*. A naive solution would be to construct a

program $P; P'$ by sequentially composing the two program versions (as in [22, 8]). However, establishing equivalence between variables in such program requires an analysis to precisely track values of variables among paths across P so they can be later compared to values in *corresponding paths* in P' . Fig. 3 informally illustrates the paths that have to be correlated through $sign; sign'$ to track the relationship between sgn and sgn' (the notation \equiv_x means $x = x'$). In order to establish equivalence for sgn and sgn' over $sign; sign'$, the analysis essentially needs to separately track all paths, withholding over-approximation, since abstracting together paths (for instance those right after the first branch) will result in a non-restorable loss of equivalence (we could never restore \equiv_{sgn} if we abstract the $sgn = 1$ and $sgn = -1$ paths).

Intuitively, establishing equivalence using the sequential composition $P; P'$ requires full path sensitivity, leading to an inherently non-scalable solution. Further, in the presence of loops and widening, applying widening separately to the loops of P and to those of P' does not allow maintaining variable relationships under abstraction.

To address these challenges, we construct a *correlating program* $P \bowtie P'$ where operations of P and P' are carefully interleaved to achieve optimal correlation throughout the analysis. Fig. 4 shows the correlating program for the programs of Fig. 1. The programs were transformed to a guarded command language form to allow for interleaving. A key feature of the correlating program for closely related program versions is the ability to keep matched instructions, that appear in both versions, closely interleaved. This allows the analysis to better maintain relationships as the program executions are better aligned. Using the correlating program, we can directly track the relationship between sgn in $sign$ and its corresponding variable sgn' in $sign'$.

2.3 Correlating Abstract Domain

To analyze a given correlating program, we introduce a *correlating abstract domain* that tracks relationships between corresponding variables in P and P' by tracking relationship in the correlating program $P \bowtie P'$. Unfortunately, any domain with convex constraints will still fail to capture the precise relationship between sgn and sgn' . For example, using the polyhedra abstract domain [7], the relationship between sgn and sgn' in the correlating program would be lost, leaving only the trivial $\langle 1 \geq sgn \geq -1, 1 \geq sgn' \geq -1 \rangle$ constraint. Although the result soundly reports a difference, as we do not explicitly know that \equiv_{sgn} , we still know nothing of the difference between the programs.

An obvious, but prohibitively expensive, solution to the problem is to use disjunctive completion, moving to a powerset domain in which every abstract state is a set of convex objects (e.g., set of polyhedra). A state in such domain is a set of convex abstract representations (e.g., polyhedra [7] or octagon [19]). For example, analyzing $sign \bowtie sign'$ using a powerset domain would yield:

$$\begin{aligned}\sigma_1 &= \{x = x' < 0, sgn = sgn' \mapsto -1\} \\ \sigma_2 &= \{x = x' = 0, sgn \mapsto 1, sgn' \mapsto -1\} \\ \sigma_3 &= \{x = x' > 0, sgn = sgn' \mapsto 1\}\end{aligned}$$

However, using such domain would significantly limit the applicability of the approach. The desirable solution is a partially disjunctive domain, in which only certain disjunctions are kept separate during the analysis, while others are merged. The challenge in our setting is in keeping the partition fine enough such that equivalence could be preserved, without reaching exponential blowup.

2.3.1 Partitioning

As the goal of this work is to distinguish equivalent from differencing behaviors, using equivalence as criteria for merging paths is apt. The partitioning will abstract together paths that hold equivalence for the same set of variables, allowing for a maximum of $2^{|VC|}$ disjunctions in the abstract state, where VC is the set of correlated variables. So far we have implicitly defined VC as a correlation between P, P' input and outputs, but our approach is in fact parameterized by this matching, allowing for any P variable to be matched with any of P' which has the potential to provide a more precise result (in the cost of scaling) or alternatively provide a more coarse, scalable result by allowing less variables or only certain equivalence classes of 2^{VC} .

For example partitioning the result of Fig. 3 according to our criteria would abstract behaviors σ_1 and σ_3 together, as they hold equivalence for sgn . The merge would abstract away data regarding x and represent sgn as the $[-1, 1]$ interval, losing precision but gaining reduction in state size. This loss of precision is acceptable as it is complemented by the offending state σ_2 .

$$\begin{aligned}\sigma_1 &= \{x = x', sgn = sgn' \mapsto [-1, 1]\} \\ \sigma_2 &= \{x' = 0, sgn \mapsto 1, sgn' \mapsto -1\}\end{aligned}$$

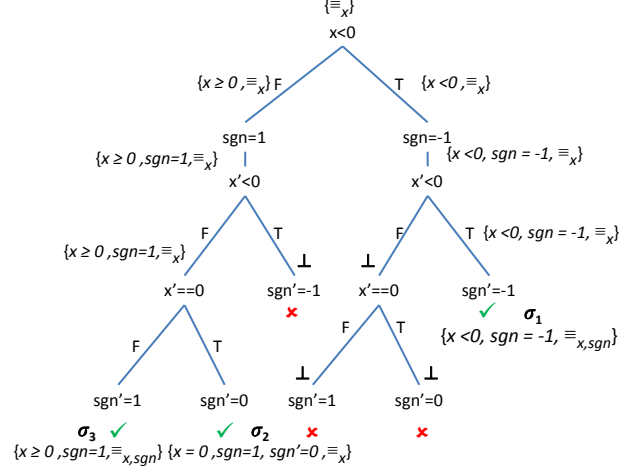


Figure 3: Joint $sign; sign'$ analysis

To gain a reduction of state size, we must perform partitioning dynamically during analysis. This cannot be achieved using a sequential composition $P; P'$. Looking at Fig. 3 we see that equivalence holds only at final states. Intuitively, this is because an operation in P has to "wait" for its equivalent operation to occur in P' . To overcome this, our correlating program $P \bowtie P'$ interleaves P and P' commands in an optimized manner, and informs the analysis when programs have reached a point where correlation may be established. The choice of *correlation points* (denoted CP) is done during the construction of the correlating program. We do not go into the specifics of creating $P \bowtie P'$ but only briefly note that the interleaving is chosen according to a syntactic diff process over a guarded command language version of the programs.

```

int sign(int x) {
  int x' = x;
  guard g1 = (x < 0);
  guard g1' = (x' < 0);
  int sgn;
  int sgn';
  if (g1) sgn = -1;
  if (g1') sgn' = -1;
  if (!g1) sgn = 1;
  if (!g1) sgn' = 1;
  guard g2' = (x' == 0);
  if (g2') sgn' = 0;
}

```

Figure 4: Correlating program $sign \bowtie sign'$.

2.3.2 Windening

Although we achieved a reduction in state size using partitioning, we have yet to account for programs with loops. Handling loops is where most previous approaches fall short [8, 15, 21, 22]. To overcome this, we define a widening operator for our domain, based on the convex sub-domain widening operator. The main challenge here, as our state is a set of convex objects, is finding an optimal pairwise matching between objects for a precise widened result. Ideally, we would like to pair objects that adhere to the same “looping path” meaning we would like to match a path π_i ’s abstraction with a path π_{i+1} that results from taking another step in the loop. This requires encoding path information along with the sub-state abstraction. This information is acquired by keeping *guard values* explicitly, as they appear in our correlating program, inside the state. As guard values (true or false) reflect branch outcomes, they can be used to match sub-states that advanced on the loop by matching their guard values.

```

int sum(int arr[], unsigned len) {
    int result = 0;
    for (unsigned i = 1; i < len; i+=2)
        result += arr[i];
    return result;
}
-----
int sum'(int arr[], unsigned len) {
    int result = 0;
    unsigned i = 0;
    while (i + 1 < len) {
        i++;
        result += arr[i];
        i++;
    }
    return result;
}

```

Figure 5: Two equivalent versions of a simple looping program for partial array summation.

We note that the correlating program is crucial to maintaining equivalence over loops. To demonstrate this we perform the simple exercise of checking equivalence of a small looping program with itself. Consider the array summation program in Fig. 5. Equivalence for these two small programs cannot be established soundly by approached based on under approximation. To emphasize the importance of the correlating program, we will first show the result of an analysis of $sum; sum'$ which will be:

$$\begin{aligned}
\sigma_1 &= \{len = len' \leq 1, result = result' \mapsto 0\} \\
\sigma_2 &= \{len = len' > 1\}
\end{aligned}$$

```

int sum(int arr[], unsigned len) {
    unsigned len' = len;
    int arr'[] = arr;
    int result = 0;
    int result' = 0;
    {
        unsigned i = 1;
        unsigned i' = 0;
    l:  guard g = (i < len);
    l': guard g' = (i' + 1 < len');
        if (g') i'++;
        if (g) result += arr[i];
        if (g') result' += arr'[i'];
        if (g') i'++;
        if (g) i+=2;
        if (g) goto l;
        if (g') goto l';
    }
}

```

Figure 6: $sum \bowtie sum'$

This loss of equivalence occurred due to the inability precisely track the relationship of $result$ and $result'$ over $sum; sum'$: as we widened the first loop to converge, all paths passing through that loop were merged together, losing the ability to be "matched" with the second loop waiting further down the road. Performing the same analysis on $sum \bowtie sum'$ instead as seen in Fig. 6, allows maintaining equivalence, as the loops are interleaved correctly to allow establishing $result = result'$ as a loop invariant, surviving the widening process to prove equivalence at the end as the result would be: $\sigma_1 = \{\equiv_{result}\}$. We note that we explicitly assume equivalence in array content for sum and sum' .

2.4 Definition of Difference

So far, we defined difference in programs as difference in output variable values at the program final state. We extend the notion of difference beyond that, allowing for several program locations to be identified as viable for differentiation (we name these locations *differencing points* denoted DP). Other than the programs endpoints, we include every location which: (i) emits an output value (using a system function or through a global variable) (ii) accesses array. This means that procedures may now differ, even if they hold equivalence on the final state. Thus we define difference also as producing a different output value in mid-execution or accessing an array through a different index. This means that programs which

differ in array access *patterns* will now be flagged as different. This is a sound approach for difference, as this may include access to illegal bounds (an example for this can be seen in Fig. 11). We note that this definition of description cannot be achieved by instrumenting output or array access through temporary variables and deferring checking their equivalence to the end, as the amount of temporaries needed may be unbound. We therefore require procedures to agree on these locations, in case we wish to verify using the extended definition of difference.

3 Evaluation of Existing Work

We have implemented our approach in a tool called DIZY based on the LLVM compiler infrastructure and the APRON numerical abstract domain library, and evaluated our approach on a number of challenging real world programs where the patches affect numerical variables. As benchmarks, we used several programs from the GNU core utilities (differencing versions 6.10 and 6.11), as well as a few handpicked patches taken from the Linux kernel and the Mozilla Firefox web browser. In most programs, we were able to precisely describe the difference.

3.1 Results

```
size_t rpc_uaddr2sockaddr (const size_t uaddr_len, ...) {
    char buf[ RPCBIND_MAXUADDRLEN ];
    ...
-   if ( uaddr_len > sizeof ( buf ))
+   if ( uaddr_len > sizeof ( buf ) - 2:)
        return 0;
    ...
    (*)1
    buf [ uaddr_len ] = '\n';
    buf [ uaddr_len + 1] = '\0';
    ...
}
```

Figure 7: Linux kernel `addr.c` v2.6.32-rc6 module with patch.

Producing delta from abstract state Fig. 7 shows a patch made to the `net/sunrpc/addr.c` module in the Linux kernel SUNRPC implementation v2.6.32-rc6 which is aimed at removing an off-by-two array access out of bounds violation in the original program. Although simple, this example shows two advantages DIZY: (i) the ability to analyze programs without the need to run them and (ii) the ability to capture

fine-grained differences. The result of analyzing the correlated program is shown in Fig. 8.

$$\frac{\sigma_1:}{\text{returned}' = \text{true} \quad \text{returned} = \text{false} \quad \text{uaddr_len}' \leq \text{RPCBIND_MAXUADDRLEN} - 2 \quad \text{uaddr_len}' \geq \text{RPCBIND_MAXUADDRLEN}}$$

Figure 8: Difference for `rpc_uaddr2sockaddr`

The difference is captured by one sub-state where the execution has ended in the patched program but continues in the original. We instrument the correlating program with a return flag to capture the difference as otherwise equivalence holds (none of the original variables change).

```
int input_position;

bool char_to_clump(char c) {
    int width;
    ...
+ if (width < 0 && input_position == 0) {
+     chars = 0;
+     input_position = 0;
+ } else if (width < 0 && input_position <= -width) {
+     input_position = 0;
+ } else {
+     input_position += width;
+ }
    (*)1
    ...
    return chars;
}
```

Figure 9: Original and patched version of `coreutils pr.c`'s `char_to_clump` procedure

Capturing complex delta Fig. 9 shows a patch made to the `char_to_clump` function in version 6.11 of `coreutils`. The patch replacing the execution of the line `input_position += width`, which originally executed unconditionally, with a conditional structure that in the new version, allows the line to execute only under certain complex conditions. Since the variables handled in this patch (the global `input_position` and return value `chars`) emit output, describing how their values changed and under which terms is important, especially as the patch cannot be easily parsed by a programmer to understand its meaning. The

result of our analysis at the return point is show in Fig. 10. These results include information regarding initial values of parameters for improved precision.

σ_1 :	σ_2 :	σ_3 :
<code>input_position₀ = 0</code>	<code>input_position₀ < -width</code>	<code>input_position₀ < -width</code>
<code>chars' = 0</code>	<code>input_position₀ < 0</code>	<code>input_position₀ > 0</code>
<code>input_position = width</code>	<code>input_position' = 0</code>	<code>input_position' = 0</code>
<code>input_position < 0</code>	<code>input_position < width</code>	<code>input_position > width</code>
<code>input_position' = 0</code>	<code>width < 0</code>	<code>input_position <= 0</code>

Figure 10: Difference for coreutils pr.c `char_to_clump`

The result convey the difference in the output variable values alongside some of conditions under which the difference occurs. The result is composed of three sub-states featuring difference and adhere to two added paths in the patched program. The first sub-state belongs to the first branch in the added conditional: the difference is comprised of (i) the new value of `input_position` is 0 as opposed to it being `width` in the former version (the analysis took the `input_position += width` line into account and incorporated knowing that `input_position = 0` from the branch condition). The analysis also deduced that the old `input_position` is negative under the same input as the branch condition dictates that `width` is negative. (ii) `chars` in the new program is 0 under this path. The two other sub-states adhere to the second added path in the conditional and track a difference for `input_position` alone, basically stating that `input_position` under this path used to assume values in ranges $[-\text{inf}, \text{width}]$ and $[\text{width}, 0]$ but now is simply 0. The splitting of this path into two cases is a result of expressing the non-convex `input_position \neq 0` condition from the first branch conditional using two sub-states. The result also describes constraints on the procedure's input under which the difference exist. Another product of the analysis, which we do not show here, are sub-states describing paths which the patch did not affect.

```

bool bsd_split_3 (char *s, size_t s_len,...) {
    int i = s_len;
    i--;
+   if (s_len == 0) return false;
    while (i && s[i] != '\0') { (*)1
        i--;
    }
    ...
    (*)2
}

```

Figure 11: Original and patched version of coreutils md5sum.c's bsd_split_3 procedure

Maintaining Equivalence and Reporting Difference in Loops We explore a challenging loop scenario where all paths in the programs contain loops and only some of them maintain equivalence. Fig. 11 shows part of coreutils md5sum.c bsd_split_3 function that was patched in version version 6.11 to disallow 0-length inputs. Although this patch seems trivial, analyzing it is challenging as it affects the behavior of loops i.e. unbound path lengths. The main challenge in this example, is separating the path where s_len is 0, which results in the loop index i ranging within negative values (producing an array access out of bounds fault), from the rest of the behaviors that maintain equivalence, throughout the widening process which is required for the analysis to reach a fixed point. The result of our analysis with partitioning is shown in Fig. 12 (a) (per differencing points $(*)_1, (*)_2$).

(*) ₁ :		(*) ₂ :	
σ_1 :	σ_2 (equivalent):	σ_1 (equivalent):	
$s_len = 0$	$s_len' = s_len$	$s_len' = s_len$	
$s_len' = 0$	$i' = i$	$i' = i$	
$i \leq -1$	$s_len' - 1 \geq i'$	$s_len' - 1 \geq i'$	

(a)

(*) ₁ :			(*) ₂ :	
σ_1 :	σ_2 (equivalent):	σ_3 (equivalent):	σ_1 (equivalent):	
$s_len = 0$	$s_len' = s_len$	$s_len' = s_len$	$s_len' = s_len$	
$s_len' = 0$	$i' = i$	$i = 0$	$i' = i$	
$i \leq -1$	$s_len' - 1 \geq i'$	$s_len' \geq 1$	$i = 0$	
			$s_len' \geq 1$	

(b)

Figure 12: Difference for bsd_split_3

We can see the analysis successfully reports a difference for the singularity

point $s_len = 0$ inside the loop, precisely describing the scenario where i' is negative. We can also see the other equivalent state existing within the loop which depicts the results of the widened analysis for all other paths (the $s_len \neq 0$ constraint is not existing there due to partitioning as we will soon show). The differencing sub-state will be omitted once we move past the loop as the $i \leq -1$ constraint will not allow it to exist beyond the loop body thus we are left with the equivalent state alone after the loop which correctly expresses the fact that the programs are equivalent at this point (since both i 's converged at 0). We can see that the result at the second differencing point has lost precision since it does not reflect the $i = 0$ constraint. The loss of this constraint is, again, due to partitioning as both sub-states that describe exiting the loop and the one describing entering the loop, hold equivalence for all variables and are joined together and lose the extra constraint information. If we analyze the same example with no partitioning we get the result of Fig. 12 (b).

Which further separates the paths in the program, allowing for a different sub-state for the $i = 0$ and $i \neq 0$ substates (again, the $i \neq 0$ constraints was lost when joining together the $i > 0$ and $i < 0$ states as they both adhere to the same path and hold the same guard values). This extra precision is beneficial, but we still managed to supply a satisfactory result using the more scalable partitioning by equivalence technique.

4 Related Work

Our work has been mainly inspired by recent work identifying program differencing as having vast security implications [2, 24] as well as advancements made in the field of under-approximations of program equivalence [8, 15, 21, 22].

The problem of program differencing is fundamental [9] and early work mainly focused on computing syntactical difference [12]. These solutions are an important stepping stone and we used syntactical diff as a means to achieve interleaving of programs in our correlating program for better analysis results. Another possibility for creating this program is to rely on the editing sequence that creates the new version from the original program [10]. Horwitz was one of the first to present a technique for comparing two versions of a program, using a partitioning operation on a graph representation of programs, and classifying each change as either semantic or textual. This is an important preliminary work that can be used for better correlation between programs. The results however did not supply a semantic description of the change. Lahiri, Vaswani and Hoare [18] presented a

high-level overview of the opportunities, applications and challenges of differential static analysis and point out many possible benefits to be gained from such work. This also positions differential analysis as an important emerging topic.

Jackson and Ladd [13] proposed a tool for computing data dependencies between input and output variables and comparing these dependencies along versions of a program for discovering difference. This method may falsely report difference as semantic difference may occur even if data dependencies have not changed. Furthermore, data dependencies offer little insight as to the meaning of difference i.e. input and output values. Nevertheless, this was an important first step in employing program analysis as a means for semantic differencing.

Several works on the problem of equivalence of combinatorial circuits [16, 20, 5] made important contributions in establishing the problem of equivalence as feasible, producing practical solutions for hardware verification.

We rely on classic methods of abstract interpretation [6] for presenting an over approximating solution for semantic differencing and equivalence. To achieve this we devised a static analysis over a correlating program. The idea of a correlating program is similar to that of self-composition [25] except that we compose two different programs in a interleaving designed to maintain a close correlation between them. The use of a correlating construct for differencing is novel as previous methods mainly use sequential composition [8, 21, 22], disregarding possible program correlation.

We base our analysis on numerical abstractions [7, 19] that allow us to reason about variables of different programs. The abstraction is further refined in a way similar to trace partitioning [23] with an equivalence-based partitioning criteria.

Symbolic execution based methods [21, 22] offer practical equivalence verification techniques for loop and recursion free programs with small state space. These works complement each other in regards to reporting difference as one [21] presents an over approximating description of difference they call differential summaries and the other [22] presents an under approximating description including concrete inputs for test cases demonstrating difference in behavior. An interesting question is how could these methods be combined iteratively to achieve better precision. Also, this work can be used to complement our work in cases where equivalence could not be proven and the description of difference can be leveraged for the extraction of concrete input that leads to offending states.

Bounded model checking based work [8] presents the notion of partial equivalence which allows checking for equivalence under specific conditions, supplied by the user but are bound by loops. They employ a technique based on theorem provers for proving an equivalence formula which embeds program logic (in

SSA form) alongside the requirement for input and output equivalence and user provided constraints.

[1] introduced a correlating heap semantics for verifying linearizability of concurrent programs. In their work, a correlating heap semantics is used to establish correspondence between a concurrent program and a sequential version of the program at specific linearization points.

5 Future Work

In this section we will first outline the different aspects of the differential analysis problem and discuss questions and research directions which derive from them. Second, we will propose a time line for exploring these directions throughout the research. As the problem of differential analysis is vast, we do not plan to explore all proposed directions immediately (some will not appear in the research plan and will be considered as the research progresses).

5.1 Problem Dimensions

5.1.1 Program Correlation

The problem of correlating executions of different programs for differentiation is only partially addressed in our work so far. We chose to use a standard semantics over a correlating program in order to leverage existing analysis techniques (that were adapted to maintain information about variable correlations). We constructed our correlating program using a simple syntactic diff algorithm over imperative commands of normalized programs. This method performs well on many examples, but may still produce a result that defies successful analysis, when encountered with complex language features or more drastic program changes.

Correlating Program Caveats A more careful examination of the correlating program, reveals that branching may break program semantics. By observing any of the correlating programs examples that contain a branching `goto` command, one sees that if the branch destination of one program conflicts with that of the second program, the second program semantics will change (and vice versa). This has not been a problem for the set of examples we handled so far, since every branch in one program was immediately followed by a branch in the other

- to the same destination. However, this needs adjustment if we strive to handle more complex, larger examples. Correct composition of programs requires instrumenting the programs with an explicit program counter, and executing instructions according to the value of the counter. This can produce a more dynamic composition that allows more control over which program advances at each step. An example of such composition for the program from Fig. 1 is shown in Fig. 13.

```

int sign(int x) {
  int x' = x;
  int pc = 0, pc' = 0;
  int turn = 1;
  while (pc < 3 || pc' < 5) {
    if (turn) {
      switch (pc) {
        case 0: guard g1 = (x < 0); pc++;    turn = 0;    break;
        case 1: if (g1) sgn = -1;    pc++;    turn = 0;    break;
        case 2: if (!g1) sgn = 1;    pc++;    turn = 0;    break;
        default:                                turn = 0;    break;
      }
    } else {
      switch (pc') {
        case 0: guard g1' = (x' < 0);    pc'++;    turn = 1;    break;
        case 1: if (g1') sgn' = -1;    pc'++;    turn = 1;    break;
        case 2: if (!g1') sgn' = 1;    pc'++;    turn = 1;    break;
        case 3: guard g2' = (x' == 0);    pc'++;    break;
        case 4: if (g2') sgn' = 0;    pc'++;    break;
        default:                                break;
      }
    }
  }
}

```

Figure 13: Correlating program $sign \bowtie^* sign'$ according to the new format.

The program counters are instrumented as the variables pc and pc' . The variable $turn$ is used to determine which of the programs will advance next. This decision can be altered during the analysis process to advance further over one program, this can also be determined dynamically (for dynamic analysis methods).

Program Graph Correlation Our method for composing programs so far has used the textual representation sequential order of commands as a basis for composition. We took the vectors of (guarded) commands from both programs and tried to match it using a syntactic diff. There is much potential in applying other techniques for finding this matching. Previous work addressing syntactical methods for program equivalence using graph representation [11, 10] can be used to find a matching for differentiation. This matching, unlike ours, considers all sort

of instructions in the program (i.e. branch conditionals, etc. as well as imperative commands).

Correlation Refinement The correlating program is an input for the analysis process and effectively determines the order in which each of the programs will execute alternately. We raise the question: does this "scheduling" need be predetermined? The analysis can defy the constant interleaving supplied by the correlating program and choose to execute more commands of one program instead of alternating to the other program to try and explore a better result. The immediate criteria for defining this "better" result would be equivalence, i.e. the analysis would choose to advance further in the execution of one program, until it reached a more equivalent abstract state (more variables now maintain equivalence). An example where such a technique is useful can be seen in Fig. 14.

```

static void
print_numbers (long first, long step, long last, ...)
{
    long i;
    for (i = 0; /* empty */; i++) {
        long x = first + i * step;
        if (step < 0 ? x < last : last < x) break;
        if (i) fputs (separator, stdout);
        printf (fmt, x);
    }
    if (i)
        fputs (terminator, stdout);
}

```

coreutils seq.c v6.9

```

static void
print_numbers (long first, long step, long last, ...)
{
    bool out_of_range = (step < 0 ? first < last : last < first);
    if (! out_of_range) {
        long x = first;
        long i;
        for (i = 1; /* empty */ ; i++) {
            printf (fmt, x);
            if (out_of_range) break;
            x = first + i * step;
            out_of_range = (step < 0 ? x < last : last < x);
            if (out_of_range){
                bool print_extra_number = false;
                ... // print_extra_number is decided here
                if (! print_extra_number) break;
            }
            fputs (separator, stdout);
        }
        fputs (terminator, stdout);
    }
}

```

coreutils seq.c v6.10

Figure 14: Original and patched version of coreutils seq.c's print_numbers procedure

Trying to find a textual correlation between these two version will fail, and existing approaches will not be able to capture the fact that these version only differ slightly, and only one extra output will be emitted (for the case where *out_of_range* is true and *print_extra_number* is true). In this example, if we allow the analysis to choose the order of alternation between each programs, until it reaches a point of output (this is an equivalence-in-output criteria), it will produce a result stating that the programs are equivalence up to the point where the extra number is printed.

This is a sort of equivalence-guided correlation refinement analysis. We note that equivalence is only one option for guiding the correlation. To practically produce such an analysis, we will either need to extend the correlating program to the format discussed in the previous paragraph or use a dual analysis over two programs, as discussed in Section 5.1.3.

5.1.2 Definition of Difference

Extending the Notion of Difference Previous work regarding semantic differencing [21, 8, 22, 17] define program difference as difference between outputs in final state. This notion of difference may be insufficient, as observable behavioral changes may appear before the program terminates (as explained in Section 2.4. We explore extending the notion of difference towards (i) array access indexes (ii) output functions (iii) assertion violations. This means that our analysis will compute difference at these locations - specifically for variables which values affect access index, output value and assertion correctness - as well as the final state. This task is not trivial for all programs, as different program versions can emit a difference number of output, or access different arrays. These scenarios require further study.

Specification Guided Differential Analysis Proposing to compute program difference at assertion locations introduces specifications into the domain of differential analysis. One may wish to ignore output equivalence and concentrate on the question of *how does patching a certain program affect its specification?* This has the potential to be a simpler problem and a lower hanging fruit as we have seen no previous work addressing this particular question. One can further leverage the program specification, by assuming it as part of the analysis process for a sort of "delta verification" technique which relies on the hypothesis that the previous version holds up to its specification, and tries to verify that the patched version does so as well.

5.1.3 Extending Existing Analysis

So far we have explored using only intra-procedural static analysis over a correlated program with numerical abstract domains. This can be vastly extended as follows:

Dual Program Analysis As mentioned, performing the analysis over the correlating program has disadvantages as it requires a complex canonization and composition of programs. We consider porting our analysis framework to work over two program CFGs. This will alleviate the need for performing complex transformation and composition and will also allow us to implement correlation refinement as we can independently choose on which of the program graphs we wish to advance (conceptually, this could also be performed on the correlated program but with much less ease). This will also allow us to easily use graph correlation techniques for choosing the order of advancement over the two programs. The disadvantage of this method is that it does not create a single, runnable correlating program (that can be used by other techniques as described in).

Inter-procedural Analysis We aim to achieve a better inter-procedural analysis technique as we now rely on modular analysis of function calls to prove equivalence or inlining in case equivalence cannot be proven.

Floats, Pointers, Arrays and Heap Domains Further research is required for adding abstract domains to support representation of other variable types including floating point, pointers, arrays and complex heap data structures.

5.1.4 Analysis Techniques

We do not limit ourselves purely to static methods and abstract interpretation. The experience and techniques gained in our research so far can be used to apply other analysis techniques successfully.

Dynamic Analysis. We plan to employ dynamic analysis techniques, for finding differences by running both versions of the program with the same input and check for equivalence between output states. This method is lacking as it is not able to detect differences that occur during execution such as array accesses, pointer de-referencing etc. Using our *correlating program* however, we can detect these differences as we can compare values at differencing points.

Symbolic Execution. This method has become very popular in achieving better test coverage for programs [3] and there have been attempts at employing this method for comparing program versions [22, 17, 21]. Although this method is not sound, it offers results with zero false positives. We intend to explore the

use of symbolic execution methods for differencing, starting by using our correlating program as an input to existing symbolic execution tools. This will allow producing actual input values which generate different outputs in the compared programs.

Concolic Execution. A method combining testing (**concrete**) and symbolic execution ((**symbolic**) which is more scalable as only part of the execution is symbolically represented and the rest runs with concrete values. An interesting question is how can we direct this method towards focusing on program differences (using the correlating program) and getting more differencing inputs.

Statically Directed Symbolic Execution. One of the pitfalls of symbolic execution methods is coverage. Due to an exponential number of paths, pruning techniques are required to achieve better coverage. We intend to the results from our static analysis towards limiting variable ranges and number of paths to be explored for finding difference. We believe that using the analysis data (for instance to ignore parts of the programs that were proven equivalent) can immensely improve performance.

5.1.5 Future Evaluation

Finally we will discuss the method we plan on evaluating our work with.

1. **Contrived Challenge Programs:** We will test on a set of short program, each presenting a different challenge for differential analysis, including: variable and location correlation, false differences, false equivalence, loops, etc. to evaluate the advantages and pitfalls of each technique.
2. **GNU Coreutils:** We will test on the coreutils benchmark of medium length C programs. These will give us a clue as to how we perform on real world code.
3. **Patch Survey:** We will perform a survey of patching to determine how patches look in large scale programs (OS, browsers, etc.) and which features they include and evaluate how our tool will perform there.

5.2 Research Timeline

The following are the immediate research topics we plan to address. We will consider other topics as mentioned in Section 1 as our research advances and according to the results of our immediate plans.

1. Better our existing tool to achieve compelling results on a larger set of real numerical programs. Our tool will be able to process large code examples and produce a precise and useful delta when such exists or prove no delta exists. – 2 months.
2. Explore applying dynamic analysis methods such as fuzz testing, concolic and symbolic execution to correlating programs in order to evaluate how these methods perform and scale in regards to our static approach as well as extract differentiating input towards test case generation applications – 8 months.
3. Combine static and dynamic analysis methods towards achieving better precision and extracting more differencing program input for automatic test case generation. Use static analysis directed symbolic execution for maximum scalability and precision – 6 months

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Abstract Semantic Differencing for Numerical Programs

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Abstract

We address the problem of computing semantic differences between a program and a patched version of the program. Our goal is to obtain a precise characterization of the difference between program versions, or establish their equivalence when no difference exists.

We focus on computing semantic differences in numerical programs where the values of variables have no a-priori bounds, and use abstract interpretation to compute an over-approximation of program differences. Computing differences and establishing equivalence under abstraction requires abstracting relationships between variables in the two programs. Towards that end, we first construct a *correlating program* in which these relationships can be tracked, and then use a *correlating abstract domain* to compute a sound approximation of these relationships. To better establish equivalence between correlated variables and precisely capture differences, our domain has to represent non-convex information. To balance precision and cost of this representation, our domain may over-approximate numerical information as long as equivalence between correlated variables is preserved.

We have implemented our approach in a tool called DIZY, built on the LLVM compiler infrastructure and the APRON numerical abstract domain library, and applied it to a number of challenging real-world examples, including programs from the GNU core utilities, Mozilla Firefox and the Linux Kernel. Our evaluation shows that DIZY often manages to establish equivalence, describes precise approximation of semantic differences when difference exists, and reports only a few false differences.

1. Introduction

Understanding the semantic difference between two versions of a program is invaluable in the process of software development. A developer making changes to a program is often interested in answering questions like: (i) did the patch add/remove the desired functionality? (ii) does the patch introduce other, *unexpected*, behaviors? (iii) which regression tests should be run? Answering these questions manually is difficult and time consuming.

Semantic differencing has received much attention in classical work (e.g., [9–11]) and has recently seen growing interest for various applications ranging from testing of concurrent programs [4], understanding software upgrades [14], to automatic generation of security exploits [2].

Existing Techniques Existing techniques mostly offer under-approximating solutions, the prominent of which is regression testing which provides limited assurance of behavior equivalence while consuming significant time and compute resources. Other approaches for computing semantics differences [20, 21] rely on symbolic execution techniques, may miss differences, and generally unable to prove equivalence. Previous work for equivalent checking [8] rely on unsound bounded model checking techniques

to prove (input-output) equivalence of two closely related numerical programs, under certain conditions.

We present an approach based on abstract program interpretation [6] for a *sound*, succinct representation of changed program behaviors and proving equivalence. Our method focuses on abstracting relationships between variables, and therefore behaviors, in both versions allowing us to achieve a precise description of difference and prove equivalence while ignoring other program information which may encumber a traditional analysis but is less relevant in our setting

Problem Definition We define the problem of *semantics differencing* as follows: Given a pair of programs (P, P') that agree on the number and type of inputs (i.e. have the same prototype), for every execution π of P that originate from an input i and a corresponding execution π' of P' that originates from the *same input* i our goal is:

- Check whether π and π' agree on output i.e., are output-equivalent.
- In case of difference in behavior, provide a description of difference.

We intentionally define the notion of input and output equivalence loosely at this point, and we discuss several realizations of these in later sections. Take notice that this definition alleviates the need to provide detailed description of *equivalent behaviors*, allowing for a more scalable solution.

To answer the question of semantic differencing for infinite-state programs, we employ abstract interpretation. Though the notion of difference is well defined in the concrete case, defining and soundly computing it under abstraction is challenging as:

- Differencing requires correlation of *different program executions* meaning the abstraction must be able to capture and compare only the input-equivalent executions, and avoid comparing ones that are not input-equivalent.
- Equivalence of abstract output values does not entail concrete value equivalence.

To address these challenges, we introduce two new concepts: (i) *correlating program* - a single program $P \bowtie P'$ that captures the behaviors of both P and P' in a way that facilitates abstract interpretation; (ii) *correlating abstract domain* - a domain for tracking relationships between variables in P and variables in P' using $P \bowtie P'$.

Correlating Program Abstracting relationships allows us to maintain focus on difference while over-approximating (whenever necessary for scalability) equivalent behaviors. In order to monitor these relationships we created a *correlating program* which captures the behavior of both the original program and its patched version. Instead of designing a correlating semantics that is capable of co-executing two programs, we chose to automatically construct the correlating program such that we can benefit from the use of standard analysis frameworks for analyzing the resulting program.

Another advantage of this new construct, is that the correlation allows for a finer-grained equivalence checking (between local variables in mid-execution and not only output). This may be used to extend the notion of difference past that of the trivial output at final state, to consider difference in array access bounds, intermediate emitting of output, etc. Our work allows the selection of several program points where difference will be calculated (denoted DP as defined in Section 4) enabling us to capture more differences. Thus, other methods for equivalence checking [8, 20, 21] can also be applied directly on the correlating program to find more differences.

Correlating Abstraction Our abstraction holds data of both sets of variables, joined together, and is initialized to hold equality over all matched variables. This means we can reflect relationships without necessarily knowing the actual value of a variables (we can know that $x = x'$ even though actual values are unknown). We ran our analysis over the correlating program while updated the domain to reflect program behavior. Since some updates may result in non-convex information (e.g. taking a condition of the form $x \neq 0$ into account), our domain has to represent non-convex information, at least temporarily. We address this by working with a powerset domain of convex sub-states, and we partitioned (abstracted together) sub-states according to equivalence criteria to avoid exponential blowup. Our domain may over-approximate numerical information as long as equivalence between correlated variables is preserved.

1.1 Main Contributions

The main contributions of this paper are as follows:

- we present a method for abstract interpretation of a pair of programs (P, P') for *sound* semantic equivalence and differencing by abstracting direct relationships between (P, P') variables in a partially disjunctive domain. We describe a partitioning technique for state reduction and scaling. We define a widening operator for abstracting unbound paths in our domain.
- we phrase a new technique for syntactically interleaving a pair of programs (P, P') for the creation of a *correlating program* $P \bowtie P'$ which contains the semantics of both programs. We propose an analysis over the program for characterizing program equivalence and difference, based on the aforementioned abstraction, given the properties of the correlating program which aligns (P, P') executions.
- We have implemented our approach in a tool based on the LLVM compiler infrastructure and the APRON numerical abstract domain library, and evaluated it using select patches from open-source software including GNU core utilities, Mozilla Firefox, and the Linux Kernel. Our evaluation shows that the tool often manages to establish equivalence, reports useful approximation of semantic differences when differences exists, and reports only a few false differences.

2. Overview

In this section, we provide an informal overview of our approach using a simple example.

Consider the two programs of Fig. 1, inspired by an example from [22]. For these programs, we would like to establish that the output of *sign* and *sign'* only differs in the case where $x = 0$ and that the difference is $sgn = 1 \neq sgn' = 0$ as generally described in Fig. 2.

Separate Analysis is Unsound As a first naive attempt to achieve this, one could try to analyze each version of the program separately and compare the (abstract) results. However, this is clearly unsound, as equivalence under abstraction does not entail concrete equivalence. For example, using an interval analysis [7] would

```

int sign(int x) {
  int sgn;
  if (x < 0)
    sgn = -1
  else
    sgn = 1
  return sgn
}

int sign'(int x) {
  int sgn';
  if (x' < 0)
    sgn' = -1
  else
    sgn' = 1
  if (x' == 0)
    sgn' = 0
  return sgn'
}

```

Figure 1: Two simple implementations of the *sign* operation.

$x.x'$ constraints	sgn	sgn'
$x < 0$	$sgn \mapsto -1$	$sgn' \mapsto -1$
$x = 0$	$sgn \mapsto 1$	$sgn' \mapsto 0$
$x > 0$	$sgn \mapsto 1$	$sgn' \mapsto 1$

Figure 2: behavior of *sign* and *sign'*.

yield that in both programs the value of variable *sgn* ranges in the same interval $[-1, 1]$, missing the fact that *sign* never returns the value 0.

Establishing Equivalence under Abstraction To establish equivalence under abstraction, we need to abstract *relationships between the values of variables in sign and sign'*. Specifically, we need to track the relationship between the values of *sgn* in both versions. This requires a joint representation in which these relationships can be tracked.

Correlating Program As a first step, our goal is to construct a program where we can track relationships between variables of *sign* and *sign'*. A naive solution would be to construct a program $P; P'$ by sequentially composing the two program versions (as in [8, 21]). However, establishing equivalence between variables in such program requires an analysis to precisely track values of variables among paths across P so they can be later compared to values in *corresponding paths* in P' . Fig. 3 informally illustrates the paths that have to be correlated through *sign; sign'* to track the relationship between *sgn* and *sgn'* (the notation \equiv_x means $x = x'$). In order to establish equivalence for *sgn* and *sgn'* over *sign; sign'*, the analysis essentially needs to separately track all paths, withholding over-approximation, since abstracting together paths (for instance those right after the first branch) will result in a non-restorable loss of equivalence (we could never restore \equiv_{sgn} if we abstract the $sgn = 1$ and $sgn = -1$ paths).

Intuitively, establishing equivalence using the sequential composition $P; P'$ requires full path sensitivity, leading to an inherently non-scalable solution. Further, in the presence of loops and widening, applying widening separately to the loops of P and to those of P' does not allow maintaining variable relationships under abstraction.

To address these challenges, we construct a *correlating program* $P \bowtie P'$ where operations of P and P' are carefully interleaved to achieve optimal correlation throughout the analysis. Fig. 4 shows the correlating program for the programs of Fig. 1. The programs were transformed to a guarded command language form to allow for interleaving. A key feature of the correlating program for closely related program versions is the ability to keep matched instructions, that appear in both versions, closely interleaved. This allows the analysis to better maintain relationships as the program executions are better aligned. Using the correlating program, we can directly track the relationship between *sgn* in *sign* and its corresponding variable *sgn'* in *sign'*. Section 6 provides more details on the construction of a correlating program.

Correlating Abstract Domain To analyze a given correlating program, we introduce a *correlating abstract domain* that tracks relationships between corresponding variables in P and P' by tracking relationship in the correlating program $P \bowtie P'$. Unfortunately, any domain with convex constraints will still fail to capture the precise relationship between sgn and sgn' . For example, using the polyhedra abstract domain [7], the relationship between sgn and sgn' in the correlating program would be lost, leaving only the trivial $\langle 1 \geq sgn \geq -1, 1 \geq sgn' \geq -1 \rangle$ constraint. Although the result soundly reports a difference, as we do not explicitly know that \equiv_{sgn} , we still know nothing of the difference between the programs.

An obvious, but prohibitively expensive, solution to the problem is to use disjunctive completion, moving to a powerset domain in which every abstract state is a set of convex objects (e.g., set of polyhedra). A state in such domain is a set of convex abstract representations (e.g., polyhedra [7] or octagon [17]). For example, analyzing $sgn \bowtie sgn'$ using a powerset domain would yield:

$$\begin{aligned}\sigma_1 &= \{x = x' < 0, sgn = sgn' \mapsto -1\} \\ \sigma_2 &= \{x = x' = 0, sgn \mapsto 1, sgn' \mapsto -1\} \\ \sigma_3 &= \{x = x' > 0, sgn = sgn' \mapsto 1\}\end{aligned}$$

However, using such domain would significantly limit the applicability of the approach. The desirable solution is a partially disjunctive domain, in which only certain disjunctions are kept separate during the analysis, while others are merged. The challenge in our setting is in keeping the partition fine enough such that equivalence could be preserved, without reaching exponential blowup.

As the goal of this work is to distinguish equivalent from differing behaviors, using equivalence as criteria for merging paths is apt. The partitioning will abstract together paths that hold equivalence for the same set of variables, allowing for a maximum of $2^{|VC|}$ disjunctions in the abstract state, where VC is the set of correlated variables as defined in Section 4. So far we have implicitly defined VC as a correlation between P, P' input and outputs, but our approach is in fact parameterized by this matching, allowing for any P variable to be matched with any of P' which has the potential to provide a more precise result (in the cost of scaling) or alternatively provide a more coarse, scalable result by allowing less variables or only certain equivalence classes of $2^{|VC|}$. A formal definition and discussion of VC is found in Section 4.

For example partitioning the result of Fig. 3 according to our criteria would abstract behaviors σ_1 and σ_3 together, as they hold equivalence for sgn . The merge would abstract away data regarding x and represent sgn as the $[-1, 1]$ interval, losing precision but gaining reduction in state size. This loss of precision is acceptable as it is complemented by the offending state σ_2 .

$$\begin{aligned}\sigma_1 &= \{x = x', sgn = sgn' \mapsto [-1, 1]\} \\ \sigma_2 &= \{x' = 0, sgn \mapsto 1, sgn' \mapsto -1\}\end{aligned}$$

To gain a reduction of state size, we must perform partitioning dynamically during analysis. This cannot be achieved using a sequential composition $P; P'$. Looking at Fig. 3 we see that equivalence holds only at final states. Intuitively, this is because an operation in P has to "wait" for its equivalent operation to occur in P' . To overcome this, our correlating program $P \bowtie P'$ interleaves P and P' commands in an optimized manner, and informs the analysis when programs have reached a point where correlation may be established. The choice of *correlation points* (denoted CP) is done during the construction of the correlating program. We describe the specifics of creating $P \bowtie P'$ in Section 6 and only briefly note that the interleaving is chosen according to a syntactic diff process over a guarded command language version of the programs.

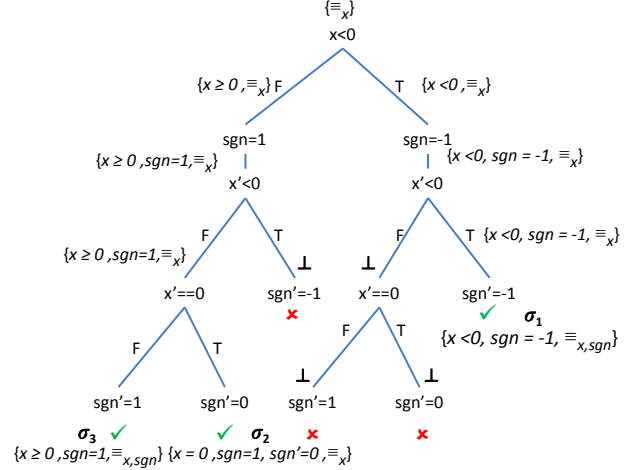


Figure 3: Joint $sign; sign'$ analysis

```
int sign(int x) {
  int x' = x;
  guard g1 = (x < 0);
  guard g1' = (x' < 0);
  int sgn;
  int sgn';
  if (g1) sgn = -1;
  if (g1') sgn' = -1;
  if (!g1) sgn = 1;
  if (!g1') sgn' = 1;
  guard g2' = (x' == 0);
  if (g2') sgn' = 0;
}
```

Figure 4: Correlating program $sign \bowtie sign'$.

Widening Although we achieved a reduction in state size using partitioning, we have yet to account for programs with loops. Handling loops is where most previous approaches fall short [8, 15, 20, 21]. To overcome this, we define a widening operator for our domain, based on the convex sub-domain widening operator. The main challenge here, as our state is a set of convex objects, is finding an optimal pairwise matching between objects for a precise widened result. Ideally, we would like to pair objects that adhere to the same "looping path" meaning we would like to match a path π_i 's abstraction with a path π_{i+1} that results from taking another step in the loop. This requires encoding path information along with the sub-state abstraction. This information is acquired by keeping *guard values* explicitly, as they appear in our correlating program, inside the state. As guard values (true or false) reflect branch outcomes, they can be used to match sub-states that advanced on the loop by matching their guard values.

We note that the correlating program is crucial to maintaining equivalence over loops. To demonstrate this we perform the simple exercise of checking equivalence of a small looping program with itself. Consider the array summation program in Fig. 5. Equivalence for these two small programs cannot be established soundly by approached based on under approximation. To emphasize the importance of the correlating program, we will first show the result of an analysis of $sum; sum'$ which will be:

$$\begin{aligned}\sigma_1 &= \{len = len' \leq 1, result = result' \mapsto 0\} \\ \sigma_2 &= \{len = len' > 1\}\end{aligned}$$

```

int sum(int arr[], unsigned len) {
    int result = 0;
    for (unsigned i = 1; i < len; i+=2)
        result += arr[i];
    return result;
}

int sum'(int arr[], unsigned len) {
    int result = 0;
    unsigned i = 0;
    while (i + 1 < len) {
        i++;
        result += arr[i];
        i++;
    }
    return result;
}

```

Figure 5: Two equivalent versions of a simple looping program for partial array summation.

```

int sum(int arr[], unsigned len) {
    unsigned len' = len;
    int arr'[] = arr;
    int result = 0;
    int result' = 0;
    {
        unsigned i = 1;
        unsigned i' = 0;
        1: guard g = (i < len);
        1': guard g' = (i' + 1 < len');
        if (g') i'++;
        if (g) result += arr[i];
        if (g') result' += arr'[i'];
        if (g') i'++;
        if (g) i+=2;
        if (g) goto 1;
        if (g') goto 1';
    }
}

```

Figure 6: $sum \bowtie sum'$

This loss of equivalence occurred due to the inability precisely track the relationship of $result$ and $result'$ over $sum; sum'$: as we widened the first loop to converge, all paths passing through that loop were merged together, losing the ability to be "matched" with the second loop waiting further down the road. Performing the same analysis on $sum \bowtie sum'$ instead as seen in Fig. 6, allows maintaining equivalence, as the loops are interleaved correctly to allow establishing $result = result'$ as a loop invariant, surviving the widening process to prove equivalence at the end as the result would be: $\sigma_1 = \{\equiv_{result}\}$. We note that we explicitly assume equivalence in array content for sum and sum' .

3. Preliminaries

We use the following standard concrete semantics definitions for a program:

A program location $loc \in Loc$, also referred to as label denoted lab , is a unique identifier for a certain location in a program corresponding to the value of the program counter at a certain point in the execution of the program. We also define two special labels for the start and exit locations of the program as *begin* and *fin* respectively.

Given a set of variables Var , a set of possible values for these variables Val and the set of locations Loc , a *concrete program state* is a tuple $\sigma \triangleq \langle loc, values \rangle \in \Sigma$ mapping the set of program variables to their concrete value at a certain program location loc i.e. $values : Var \rightarrow Val$. The set of all possible states of a program P is denoted Σ_P .

We describe an imperative program P , as a tuple $(Val, Var, \rightarrow, \Sigma_0)$ where $\rightarrow : \Sigma_P \times \Sigma_P$ is a transition system which given a concrete program state returns the following state in the program and Σ_0 is a set of initial states of the program. Our formal semantics need not deal with errors states therefore we ignore crash states of the programs, as well as inter-procedural programs since our work deals with function calls by either ignoring them and only assuming equivalence for the output (if their input was indeed equivalent) when equivalence was proven or by inlining them (we exclude recursion for now).

A program trace $\pi \in \Sigma_P^*$, is a sequence of states $\langle \sigma_0, \sigma_1, \dots \rangle$ describing a single execution of the program. Every program can be described by the set of all possible traces for its run $\llbracket P \rrbracket \subseteq \Sigma^*$. We refer to these semantics as concrete state semantics. We also define the following standard operations on traces:

- $label : \Sigma_P \rightarrow Lab$ returns the program label of a state.
- $last : \Sigma_P^* \rightarrow \Sigma_P$ returns the last state in a trace.
- $pre : \llbracket P \rrbracket \rightarrow 2^{\Sigma_P^*}$ returns the set of all prefixes of a trace.
- $states : \llbracket P \rrbracket \rightarrow 2^{\Sigma_P}$ returns the set of states the trace is composed of.

As we are addressing the semantics of two programs, we define the *product state* $\sigma_{\times} \in \Sigma_{P \times P'}$ as a pair of states $\langle \sigma, \sigma' \rangle$, the *product program* $P \times P'$ as a product of the transition systems of the underlying programs and *product trace* as a sequence of product states.

4. Concrete Semantics

In this section, we define the notion of concrete difference between programs, based on a standard concrete semantics.

4.1 Concrete State Differencing

Comparing two programs P and P' under concrete semantics means comparing their *traces*, but only those that originates from the same input. Towards that end, we first define the difference between two concrete states.

Intuitively, given two concrete states, the difference between them is the set of variables (and their values) where the two states map corresponding variables to different values. As variable names may differ between programs, we parameterize the definition with a mapping that establishes a correspondence between variables in P and P' . Thus concrete state differencing is restricted to comparing values of corresponding variables only.

Variable Correspondence A *variable correspondence* $VC \subseteq Var \times Var'$, is a partial mapping between two sets of program variables. The variable correspondence mapping can be taken as input from the user however, our evaluation indicates that is often sufficient to use a standard name-based mapping for patched versions of programs:

$$VC_{EQ} \triangleq \{(v, v') | v \in Var \wedge v' \in Var' \wedge name(v) = name(v')\}$$

Concrete State Delta Next we define the concrete state delta:

DEFINITION 1 (Concrete State Delta). *Given two concrete states $\sigma \in \Sigma_P$, $\sigma' \in \Sigma_{P'}$, and a correspondence VC , the concrete state delta is defined as:*

$$\Delta_S(\sigma, \sigma') \triangleq \{(v, val) | (v, v') \in VC \wedge \sigma(v) = val \neq \sigma'(v')\}$$

Informally, Δ_S means the "part of the state σ where corresponding variables do not agree on values (with respect to σ')". Note that Δ_S is not symmetric. In fact, the direction in which Δ_S is used has meaning in the context of a program P and a patched version of it P' . We define $\Delta_S^- = \Delta_S(\sigma, \sigma')$ which means the values of the

state that was "removed" in P' and $\Delta_S^+ = \Delta_S(\sigma', \sigma)$ which stands for the values "added" in P' . When there is no observable difference between the states we get that $\Delta_S^+(\sigma, \sigma') = \Delta_S^-(\sigma, \sigma') = \emptyset$, and say that the states are *equivalent* denoted $\sigma \equiv \sigma'$.

EXAMPLE 1. Consider two concrete states $\sigma = (x \mapsto 1, y \mapsto 2, z \mapsto 3)$ and $\sigma' = (x' \mapsto 0, y' \mapsto 2, w' \mapsto 4)$ and using VC_{EQ} then $\Delta_S^- = (x \mapsto 1)$ since x and x' match and do not agree on value, y and y' agree (thus are not in delta) and z is not in VC_{EQ} . Similarly, $\Delta_S^+ = (x' \mapsto 0)$.

We now use our notion of concrete state difference to define the difference between concrete program traces. Our goal is to compare traces that are input-equivalent i.e. originate from the same input, and check whether they are output-equivalent. To differentiate traces, we need to compare the states along each trace, but which states should we compare? this is not a trivial question since traces can vary in length and order. The comparison requires a mapping for picking pairs of states to be differentiated within the two traces.

Trace Diff Points Given two traces $\pi \in \llbracket P \rrbracket$ and $\pi' \in \llbracket P' \rrbracket$ that originate from equivalent input states, we define a trace index correspondence relation named *trace diff points* denoted DP_π as a matching of indexes specifying states where concrete state delta should be computed. Formally, $DP_\pi \subseteq \{(i, i') \mid 0 \leq i \leq |\pi|, 0 \leq i' \leq |\pi'|\}$. The question of supplying this matching, in a way that results in meaningful delta, is not a trivial one, we delay this discussion until we define the trace delta.

Trace Delta Now that we have a way of matching states to be compared between two traces, we define the notion of trace difference:

DEFINITION 2 (Trace Delta). Given two traces $\pi \in \llbracket P \rrbracket$ and $\pi' \in \llbracket P' \rrbracket$ that originate from equivalent input states, and a trace index correspondence DP_π we define the trace delta $\Delta_T(\pi, \pi') : \llbracket P \rrbracket \times \llbracket P' \rrbracket \rightarrow (\mathbb{N} \rightarrow 2^{Val})$ as state differentiations between all corresponding states in π and π' :

$$\Delta_T(\pi, \pi') = \{(i, \Delta_S(\sigma_i, \sigma'_{i'})) \mid (i, i') \in DP_\pi, \Delta_S(\sigma_i, \sigma'_{i'}) \neq \emptyset\}$$

That is, for every pair of indexes $(i, i') \in DP_\pi$ that adhere to states $(\sigma_i, \sigma'_{i'})$ that differ $(\Delta_S(\sigma_i, \sigma'_{i'}) \neq \emptyset)$, $\Delta_T(\pi, \pi')$ will contain the mapping $i \mapsto \Delta_S(\sigma_i, \sigma'_{i'})$, thus the result will map certain states in π to their state delta with the corresponding state in π' (as matched by DP_π). Since $\Delta_T(\pi, \pi')$ is based on state difference, we define Δ_T^+ and Δ_T^- similarly to their underlying states difference operations.

One possible choice for DP_π is the endpoints of the two traces $\{(n, n')\}$ (assuming they are finite) meaning differentiating the final states of the executions or formally: $\Delta_{T_n}^- \triangleq \Delta_{T_n}(\pi, \pi') = \{n \mapsto \Delta_S(\sigma_n, \sigma'_{n'})\}$ (we will also be interested in $\Delta_{T_n}^+$). The final state delta may not always be sufficient for truly describing the difference between traces as according to our definition of difference, we would be interested in checking difference at intermediate locations in the program (that emit output, check assertions or access arrays). This can perhaps be achieved by instrumenting the semantics such that the state contains all "temporary" values for a variable along with the trace index (program location is not sufficient here as a trace can loop over a certain location). This solution encumbers the analysis and substantially complicates the selection of VC as it requires relating all of these temporary, indexed, variables. Such a correspondence may be extremely hard to produce. Also the number of variables here can range up to the length of the trace (which may be unbound). Finding a DP which allows correct differentiation is a daunting task as traces of separate (although similar) programs can vastly differ.

```
void foo(unsigned x) {      void foo'(unsigned x) {
    unsigned i = 0;          unsigned i = 0;
    lab:guard g = (i >= x);  lab:guard g = (i >= 2*x);
    if (g) return;          if (g) return;
    ...                     ...
    i++;                    i++;
    goto lab;                goto lab;
}
```

Figure 7: Two simple guarded versions of a loop procedure

EXAMPLE 2. Consider the two program versions shown in Fig. 7 and the following traces generated from the input $x = 2$: $\pi = \langle (x \mapsto 2, i \mapsto 0, g \mapsto 0), (x \mapsto 2, i \mapsto 1, g \mapsto 0), (x \mapsto 2, i \mapsto 2, g \mapsto 1) \rangle$ and $\pi' = \langle (x \mapsto 2, i \mapsto 0, g \mapsto 0), (x \mapsto 2, i \mapsto 1, g \mapsto 0), (x \mapsto 2, i \mapsto 2, g \mapsto 0), (x \mapsto 2, i \mapsto 3, g \mapsto 0), (x \mapsto 2, i \mapsto 4, g \mapsto 1) \rangle$, we see that even in this simple program, finding a correlation based on traces alone is hard. Instead, if one uses program location as a means of correlation, one can produce a result that describes how values of i range differently in the new version P . If we look at all the possible values for i at label lab and differentiate them (as a set) from the values in the patched version (in the same location), we get a meaningful result that i in the patched version can differ by ranging from $x + 1$ up to $2x$.

4.2 Differencing at Program Labels

Here, we will formally describe the choosing of DP based on program locations.

Trace Delta using Program Labels Given two traces (π, π') and two program labels (l, l') we define a trace delta based on all states that are labeled l in P and l' in P' . First we define π_l as a subsequence of π where only states that are labeled l were chosen ($\pi'_{l'}$ is defined similarly). Next, we denote $\Delta_L(\pi_l, \pi'_{l'})$ as a means for comparing these sequences. As $\pi_l, \pi'_{l'}$ may vary in length and order, we cannot simply define it as applying Δ_S on each pair of states in $(\pi_l, \pi'_{l'})$ by order. In fact, Δ_L can be defined in different ways to reflect different concepts of difference, for instance, it can be defined as the differentiating the last states of π_l and $\pi'_{l'}$ (assuming they are both finite) to reflect we are only interested in the final values in that location. We chose to define Δ_L as the difference between the set of states which appear in π_l against the set of those in $\pi'_{l'}$.

DEFINITION 3 (Label based Trace Delta). Given two traces π, π' and two program labels l, l' we define the trace delta based on labels as:

$$\Delta_L(\pi_l, \pi'_{l'}) \triangleq \{\sigma \in \text{states}(\pi_l) \mid \neg \exists \sigma' \in \text{states}(\pi'_{l'}) \cdot \sigma \equiv \sigma'\}.$$

Meaning, all states that exist at label l in P but cannot be matched with any state existing at l' in P' .

EXAMPLE 3. Consider Fig. 7, for π, π' that originate from $x = 2$ then $\Delta_L(\pi_{lab}, \pi'_{lab'}) = \emptyset$ and $\Delta_L(\pi'_{lab'}, \pi_{lab}) = \{(i' \mapsto 3), (i' \mapsto 4)\}$. We see that this notion of Δ indeed captures a useful description of difference.

The problem of choosing DP is now reduced to the matching of labels as the trace indexing correspondence DP_π is induced by the definition over labels. Since we need to differentiate sets of states belonging to a certain program label, we require a correspondence of labels and therefore we define the label diff points correspondence.

Label Diff Points Given two programs (P, P') and their sets of program labels (Lab, Lab') , we define a label correspondence relation named *label diff points* denoted $DP_{Lab} \subseteq Lab \times Lab'$ as a matching of labels between programs. From this point on

any mention of the diff-points correspondence DP will refer to label diff-points DP_{Lab} . As discussed in Section 2, we allow a broad selection of differencing points, including exit points, output locations and array accesses thus capturing differences beyond return value.

Now, we will move past the concrete semantics towards *abstract semantics*. This is required as it is unfeasible to describe difference based on traces. Before doing so, we must adjust our concrete semantics since a concrete semantics based on individual traces will not allow us to correlate traces that originate from the same input. This is the first formal indication of how a separate abstraction, that considers each of the programs by itself, cannot succeed.

4.3 Concrete Correlating Semantics

We define the correlating state and trace which bind the executions of both programs, P and P' , together and define the notion of delta in this setting. Essentially, these will be states and traces of the product program $P \times P'$ but only traces that originate from equivalent input states are considered. This allows us to define the *correlating abstract semantics* which is key for successful differencing.

DEFINITION 4 (Correlating Concrete State). A *correlating concrete state* $\sigma_{\bowtie} : Var \cup Var' \rightarrow Val$ is a unified concrete state, mapping variables from both programs (P, P') to their values. The set of all possible correlating states is denoted $\Sigma_{P \bowtie P'}$.

DEFINITION 5 (Correlating Concrete Trace). A *correlating trace* π_{\bowtie} , is a sequence of correlating states $\dots, \sigma_{\bowtie_i}, \dots$ describing an execution of $P \times P'$. We restrict to traces that originate from equivalent input states i.e., $\sigma_{\bowtie_0} \equiv \sigma'_{\bowtie_0}$. The label l_{\bowtie} , $last_{\bowtie}$ and pre_{\bowtie} operations are defined similarly.

We must remember however, that the number of traces to be compared is potentially unbounded which means that the delta we compute may be unbounded too. Therefore we must use an abstraction over the concrete semantics that will allow us to represent executions in a bounded way.

5. Abstract Correlating Semantics

In this section, we introduce our correlating abstract domain which allows bounded representation of product-program state while maintaining equivalence between correlated variables.

5.1 Abstract Correlating State

We represent variable information using standard relational abstract domains. As our analysis is path sensitive, we allow for a set of abstract sub-states, each adhering to a certain path in the product program. This abstraction is similar to the trace partitioning domain as described in [22].

Our power-set domain records precise state information but does not scale due to exponential blowup. As a first means of reducing state size, we define a special join operation that dynamically partitions the abstract state according to the set of equivalences maintained in each sub-state and joins all sub-states in the same partition together (using the sub-domain join operation). This join criteria allows separation of equivalence preserving paths thus achieving better precision. Second, to allow a feasible bound abstraction for programs with infinite number of paths, we define a widening operator which utilizes the sub-domain's widening but cleverly chooses which sub-states are to be widened, according to path information encoded in state. We start off by abstracting the correlating trace semantics in Sec. 4.3.

In the following, we assume an abstract relational domain $(D^\#, \sqsubseteq_{D^\#})$ equipped with operations $\sqcap_{D^\#}$, $\sqcup_{D^\#}$ and $\nabla_{D^\#}$, for representing sets of concrete states in $\Sigma_{P \bowtie P'}$. We separate the set

of program variables into original program variables denoted Var (which also include a special added variable for return value, if such exists) and the added guard variables denoted $Guard$ that are used for storing conditional values alone ($Guard$ also include a special added guard for return flag). We assume the abstract values in $D^\#$ are constraints over the variables and guards (we denote $D^\#_{Guard}$ for sub-domain abstraction of guards and $D^\#_{Var}$ for original variables), and do not go into further details regarding the particular abstract domain as it is a parameter of the analysis. We also assume that the sub-domain $D^\#$ allows for a sound over-approximation of the concrete semantics (given a sound interpretation of program operations). In our experiments, we use the polyhedra abstract domain [7] and the octagon abstract domain [17].

Correlating Abstract State A correlating abstract program state $\sigma^\# \in Lab_\times \rightarrow 2^{D^\#_{Guard} \times D^\#_{Var}}$, is a set of pairs $\langle ctx, data \rangle$ mapped to a product program label l_\times , where $ctx \in D^\#_{Guard}$ is the execution context i.e. an abstraction of guards values via the relational numerical domain and $data \in D^\#_{Var}$ is an abstraction of the variables. We separate abstractions over guard variables added by the transformation to GCL format from original program variables there need not be any relationships between guard and regular variables.

5.2 Abstract Correlating Semantics

Tab. 1 describes the abstract transformers. The table shows the effect of each statement on a given abstract state $\sigma^\# = l_\times \mapsto S$. The abstract transformers are defined using the abstract transformers of the underlying abstract domain $D^\#$. We assume that any program P can be transformed such that it contains the operations described in Tab. 1 alone (this is achieved by the GCL format). We also assume that for $\llbracket g := e \rrbracket^\#$ operations, e is a logical operation with binary value.

Next, we define the abstraction function $\alpha : 2^{\Sigma_{P \bowtie P'}} \rightarrow 2^{D^\# \times D^\#}$ for a set of concrete traces $T \subseteq \Sigma_{P \bowtie P'}$. As in our domain traces are abstracted together if they share the exact same path, we first define an operation $path : \Sigma_{P \bowtie P'}^* \rightarrow Lab^*$ which returns a sequence of labels for a trace's states i.e. what is the path taken by that trace. We also allow applying $path$ on a set of traces to denote the set of paths resulting by applying the function of each of the traces. Finally we define the trace abstraction as following:

$$\alpha(T) \triangleq \{\sqcup_{path(\pi)=p} \beta(last(\pi)) \mid p \in path(T)\}$$

where $\beta(\sigma) = \langle \beta_{D^\#}(\sigma|_{Guard}), \beta_{D^\#}(\sigma|_{Var}) \rangle$ i.e. applying the the abstraction function of the abstract sub-domain $\beta_{D^\#}$ on parts of the concrete state applying to $Guards$ (denoted $\sigma|_{Guard}$) and $Vars$ (denoted $\sigma|_{Var}$) separately. Our abstraction partitions trace prefixes π by path and abstracts together the concrete states reached by the prefix - $last(\pi)$, using the sub-domain.

Every path in the product program will be represented by a single sub-state of the sub-domain. As a result, all *traces prefixes* that follow the same path to l_\times will be abstracted into a single sub-state of the underlying domain. This abstraction fits semantics differencing well, as inputs that follow the same path display the same behavior and will usually either keep or break equivalence together, allowing us to separate them from other behaviors (it is possible for a path to display both behaviors as in Fig. 8 and we will discuss how we are able to manipulate the abstract state and separate equivalent behaviors from ones that offend equivalence). Another issue to be addressed is the fact that our state is still potentially unbound as the number of paths in the program may be exponential and even infinite (due to loops).

$\llbracket v := e \rrbracket^\sharp$	$l_x \mapsto \{ \langle ctx, \llbracket v := e \rrbracket^\sharp_{D^\sharp}(data) \rangle \mid \langle ctx, data \rangle \in S \}$
$\llbracket g := e \rrbracket^\sharp$	$l_x \mapsto \{ \langle \llbracket g := true \rrbracket^\sharp_{D^\sharp}(ctx), \llbracket e \rrbracket^\sharp_{D^\sharp}(data) \rangle \mid \langle ctx, data \rangle \in S \} \cup \{ \langle \llbracket g := false \rrbracket^\sharp_{D^\sharp}(ctx), \llbracket \neg e \rrbracket^\sharp_{D^\sharp}(data) \rangle \mid \langle ctx, data \rangle \in S \}$
$\llbracket \text{if } (g) \{ s_0 \} \text{ else } \{ s_1 \} \rrbracket^\sharp$	$l_x \mapsto \{ \langle \llbracket g = true \rrbracket^\sharp_{D^\sharp}(ctx), \llbracket s_0 \rrbracket^\sharp_{D^\sharp}(data) \rangle \mid \langle ctx, data \rangle \in S \} \cup \{ \langle \llbracket g = false \rrbracket^\sharp_{D^\sharp}(ctx), \llbracket s_1 \rrbracket^\sharp_{D^\sharp}(data) \rangle \mid \langle ctx, data \rangle \in S \}$
$\llbracket \text{goto lab} \rrbracket^\sharp$	σ^\sharp

Table 1: Abstract transformers using abstract transformers of the underlying domain D^\sharp . The table describe the effect of each statement on an abstract state $\sigma^\sharp = l_x \mapsto S$.

```

int f(int x) {
    return x;
}

int f'(int x) {
    return 2*x;
}

```

Figure 8: Single path differentiation candidates

5.3 Dynamic Partitioning

Performing analysis with the powerset domain does not scale as the number of paths in the correlated program may be exponential (we defer the case of unbound paths to widening of loops). We must allow for reduction of state $\sigma^\sharp = l_x \mapsto S$ with acceptable loss of precision. This reduction via partitioning can be achieved by joining the abstract sub-states in S (using the standard precision losing join of the sub-domain). However this can only be accomplished after first deciding which of the sub-states shall be joined together and then choosing the program locations for the partitioning to occur. A trivial partitioning strategy is simply reverting back to the sub-domain by applying the join on all sub-states which may result in unacceptable precision loss as exemplified in Section 2 by the attempt to perform convex polyhedra analysis for $sign \bowtie sign'$ from Fig. 4. However, by taking a closer look at the final state of a fully disjunctive analysis of the same example (\equiv_v means v and v' are equivalent in the state) :

$$\begin{aligned}
\sigma^\sharp(fin) = & \{ \langle (g1, \neg g2', \equiv_{g1}), (x > 0, sgn = 1, \equiv_{x,sgn}) \rangle, \\
& \langle (\neg g1, \neg g2', \equiv_{g1}), (x < 0, sgn = -1, \equiv_{x,sgn}) \rangle, \\
& \langle (\neg g1, g2', \equiv_{g1}), (x = 0, sgn = 0, sgn' = 1, \equiv_x) \rangle \}
\end{aligned}$$

One may observe that were we to join the two sub-states that maintain equivalence on $\{x, sgn, g1\}$, it would result in an acceptable loss of precision (of losing the x related constraints). This is achieved by partitioning sub-states according to the set of variables which they preserve equivalence for. This bounds the state size at $2^{|VC|}$, where VC is the set of correlating variables we wish to track. As mentioned, another key factor in preserving equivalence and maintaining precision is the program location at which the partitioning occurs. The first possibility, which is somewhat symmetric to the first proposed partitioning strategy, is to partition at every join point i.e. after every branch converges. Let us examine $sign \bowtie sign'$ state after processing the first guarded instruction `if (G1) sgn = -1;` (we ignored $g2'$ effect at this point for brevity):

$$\begin{aligned}
\sigma^\sharp = & \{ \langle (g1, \equiv_{g1}), (x \geq 0, \equiv_{x,sgn}) \rangle, \\
& \langle (g1, \equiv_{g1}), (x < 0, sgn' = -1, \equiv_x) \rangle \}
\end{aligned}$$

This suggests that partitioning at join points will perform badly in many scenarios, specifically here as we will lose all data regarding sgn . However if we could delay the partitioning to a point where the two programs "converge" (after the following `if (G1') sgn' = -1;` line), we will get a more precise temporary result which preserves equivalence. To accomplish this, we define special program locations we name *correlating points* which present places where programs have likely converged. These are a sub-product of the correlating program construction process described in Section 6.

Other viable program locations for partitioning are our *differencing points*. Partitioning at these locations is essentially more precise than at correlation points. We remind that diff-points are

```

unsigned max = ...;
int sum'(int arr[], unsigned len) {
    int result = 0;
    if (len > max)
        return -1;
    for (unsigned i = 1; i < len; i+=2)
        result += arr[i];
    return result;
}

```

```

unsigned max' = ...;
int sum(int arr[], unsigned len) {
    unsigned len' = len;
    int arr'[] = arr;
    int result = 0;
    int result' = 0;
    guard r' = (len' > max');
    if (r') retval' = -1;
    if (r') r' = 0;
    {
        unsigned i = 1;
        unsigned i' = 1;
    l: guard g = (i < len);
    l': guard g' = 0;
        if (r') g' = (i' < len');
        if (g) result += arr[i];
        if (r') if (g') result' += arr'[i'];
        if (g) i+=2;
        if (r') if (g') i'+=2;
        if (g) goto l;
        if (r') if (g') goto l';
    }
}

```

Figure 9: Patched sum'' and correlating $sum \bowtie sum''$

product program locations where both programs conceptually converge as they are about to be checked for equivalence. Therefore if equivalence exists - it must exist now, making differencing points a prime candidate for a partitioning point.

5.4 Widening

In order for our analysis to handle loops we require a means for reaching a fixed point. As our analysis iterates over a loop, sub-states may be added or transformed continuously, never converging. We therefore need to define a widening operator for our new domain, which will further over-approximate the looping state and arrive at a fixed point. We have the widening operator of our sub-domain at our disposal, but we are faced with the question of how to apply this operator, i.e. which pairs of sub-states $\langle ctx, data \rangle$ from σ^\sharp should be widened with which. A first viable strategy, similar to the first partitioning strategy, is to perform an overall join operation on all pairs which will result in a single pair of sub-states and then simply apply the widening to this sub-state using the sub-domain's ∇ operator. If we examine applying this strategy to $sum \bowtie sum'$ from Fig. 6, we get that it will successfully arrive at a fixed point that also maintains equivalence as all sub-states maintain equivalence at loop back-edges. Now let us try applying the strategy to the more complex $sum \bowtie sum''$ as seen in Fig. 9. First we mention that as sum' introduces a return statement under the $len > max$ condition, the example shows an extra r' guard and

$retval'$ variable for representing a return (this exists in all GCL programs but we omitted it so far for brevity). While analyzing, once we pass that first conditional, our state is split to reflect the return effect:

$$\sigma^\# = [d_1 = \langle (\neg r'), (len \leq max, result = 0, \equiv_{len, result}) \rangle, \\ d_2 = \langle (r'), (len > max, retval' = -1, result = 0, \equiv_{len, result}) \rangle]$$

As we further advance into the loop, d_1 will maintain equivalence but d_2 will continue to update the part of the state regarding untagged variables (since r' is *false*), specifically it will change $result$ continuously, preventing the analysis from reaching fixed point. We would require widening here but using the naive strategy of a complete join will result in aggressive loss of precision, specifically losing all information regarding $result$. The problem originates from the fact that prior to widening, we joined sub-states which adhere to two different loop behaviors: one where both sum and sum' loop together (that originated from $len < max$) and the other where sum' has exited but sum continues to loop ($len \geq max$). Ideally, we would like to match these two behaviors and widen them accordingly. We devised a widening strategy that allows us to do this as it basically matches sub-states that adhere to the same behavior, or loop-paths. This strategy dictates using *guards* for the matching. If two sub-states agree on their set of guards, it means they represent the same loop path and can be widened as the latter originated from the former (widening operates on subsequent iterations). In our example, using this strategy will allow the correct matching of states after consequent $k, k + 1$ loop iterations:

$$\sigma_k^\# = [d_1 = \langle (\neg r', g, \equiv_g), (len \leq max, i = 2k + 1, \equiv_{i, len, result}) \rangle, \\ d_2 = \langle (r', \neg g, g'), \\ (len > max, retval' = -1, result' = 0, i' = 2k + 1, i = 1, \equiv_{len}) \rangle] \\ \text{And:} \\ \sigma_{k+1}^\# = [d_1 = \langle (\neg r', g, \equiv_g), (len \leq max, i = 2k + 3, \equiv_{i, len, result}) \rangle, \\ d_2 = \langle (r', \neg g, g'), \\ (len > max, retval' = -1, result' = 0, i' = 2k + 3, i = 1, \equiv_{len}) \rangle]$$

As we can identify the states predecessors by simply matching the guards. d_1 will be widened for a precise description of the difference shown as $\langle len = len' > max', retval' = -1, retval' = \top \rangle$.

5.5 Differencing for Abstract Correlating States

Given a state in our correlating domain, we want to determine whether equivalence is kept and if so under which conditions it is kept (for partial equivalence) or determine there is difference and characterize it. As our state may hold several pairs of sub-states, each holding different equivalence data, we can provide a verbose answer regarding whether equivalence holds. We partition our sub-states according to the set of variables they hold equivalence for and report the state for each equivalence partition class. Since we instrument our correlating program to preserve initial input values, for some of these states we will also be able to report input constraints thus informing the user of the input ranges that maintain equivalence. In the cases where equivalence could not be proved, we report the offending states and apply a differencing algorithm for extraction of the delta. Fig. 8 shows an example of where our analysis is unable to prove equivalence (as it is sound), although part of the state does maintain equivalence (specifically for $x = 0$). This is due to the abstraction being too coarse. We describe an algorithm that given a sub-state $d \in D^\#$, computes the differentiating part of the sub-state (where correlated variables disagree on values) by splitting it into parts according to equivalence. This is done by treating the relational constraints in our domain as geometrical objects and formulating delta based on that.

Correlating Abstract State Delta

DEFINITION 6 (Correlating Abstract State Delta). *Given a sub-state d and a correspondence VC , the correlating state delta $\Delta_A(d)$, computes abstract state differentiation over d . The result is an abstract state $\sqsubseteq d$ approximating all concrete values for variables correlated by VC , that differ between P and P' . Formally, the delta is simply the abstraction of the concrete trace deltas:*

$$\Delta_A(d)^+ \triangleq \alpha(\cup_{path} \Delta_T^+)$$

$$\Delta_A(d)^- \triangleq \alpha(\cup_{path} \Delta_T^-)$$

where deltas are grouped together by path and then abstracted.

The algorithm for the extraction of delta from a correlating state, is as follows:

1. d_\equiv is a state abstracting the concrete states *shared* by the original and patched program. It is achieved by computing: $d_\equiv \triangleq d|_{V=V'} \equiv d \sqcap \bigwedge \{v = v' | (v, v') \in VC\}$.
2. $\overline{d_\equiv}$ is the negated state i.e. $D^\# \setminus d_\equiv$ and it is computed by negating d_\equiv (as mentioned before, all logical operations, including negation, are defined on our representation of an abstract state).
3. Eventually: $\Delta_A(d) \triangleq d \sqcap \overline{d_\equiv}$ abstracts all states in $P \times P'$ where correlated variables values do not match.
4. $\Delta_A(d)^+ = \Delta_A(d)|_{V'}$ is a projection of the differentiation to display values of P' alone i.e. "added values".
5. $\Delta_A(d)^- = \Delta_A(d)|_V$ is a projection of the differentiation to display values of P alone i.e. "removed values".

EXAMPLE 4. Applying the algorithm on Fig. 8's P and P' where $d = \{retval' = 2retval\}$ will result in the following:

1. $d_\equiv = \langle retval' = 0, retval = 0 \rangle$.
2. $\overline{d_\equiv} = [\langle retval' > 0 \rangle, \langle retval' < 0 \rangle, \langle retval > 0 \rangle, \langle retval < 0 \rangle]$
3. $\Delta_A(d) = [\langle retval' = 2retval, retval' > 0 \rangle, \langle retval' = 2retval, retval' < 0 \rangle, \langle retval' = 2retval, retval > 0 \rangle, \langle retval' = 2retval, retval < 0 \rangle]$
4. $\Delta_A(d)^+ = [\langle retval' > 0 \rangle, \langle retval' < 0 \rangle]$
5. $\Delta_A(d)^- = [\langle retval > 0 \rangle, \langle retval < 0 \rangle]$

We note that as a sub-state is basically a conjunction of constraints, negating it by splitting to constraints and negating each individually reflects correctly the effect of negating a conjunction as we are left with a disjunction of negations, as seen in step 2. We also see that displaying the result in the form of projections is ill-advised as in some states differentiation data is represented by relationships on correlated variables alone, thus projecting will lose all data and we will be left with a less informative result. A geometrical representation of Δ_A calculation can be seen in Fig. 10.

From this point forward any mention of 'delta' (denoted Δ) will refer to the correlating abstract state delta (denoted Δ_A). We claim that Δ is a correct abstraction for the concrete state delta which allows for a scalable representation of difference we aim to capture.

6. Correlating Program

In this section we will describe how we construct our correlating program $P \bowtie P'$. The process of correlating attempts to find a better interleaving of programs for a more precise differentiation. The building process also instruments $P \bowtie P'$ with the required differencing points DP (marked $(*)$ in figures) which allow reporting of difference as well as correlation points CP which define the locations for out partitioning. We also allow a user defined selection of DP and CP .

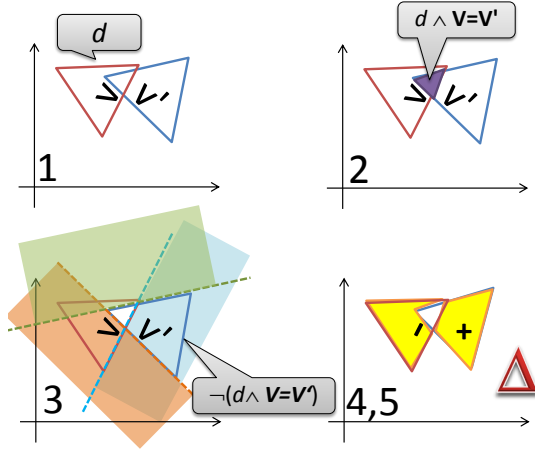


Figure 10: Delta computation geometrical representation.

6.1 Construction of $P \bowtie P'$

The idea of a correlating program is similar to that of self-composition [24], but the way in which statements in the correlating program are combined is carefully designed to keep the steps of the two programs close to each other. Rather than having the patched program sequentially composed after the original program, our correlating program interleaves the two versions. Analysis of the correlating program can then recover equivalence between values of correlated variables even when equivalence is *temporarily* violated by an update in one version, as the corresponding update in the other version follows shortly thereafter. To adhere to our formal definitions, we note that the correlating program is a reduction of the product program, allowing a single interleaving of commands, where in $P \times P'$ all interleaving are possible (similarly to $P || P'$), thus all definitions apply to $P \bowtie P'$ as it is a single case of the generalized definitions.

We will generally describe the process of constructing the correlating program. The correlating program is an optimized reduction over $P \times P'$ where not all pairs of $(\sigma^{\#}, \sigma'^{\#})$ are considered, but only pairs that result from a controlled execution, where correlating instructions in P and P' will execute adjacently. This will allow for superior precision.

The input for the correlation process are two program (P, P') in C. The first step involves transforming both programs to a normalized guarded instruction form (P_G, P'_G) . Next, a vector of *imperative commands* I (and I' respectively) is extracted from each program for the purposes of performing the syntactic diff. An imperative command in our GCL format is defined to be either one of $v := e \mid \text{goto } l \mid f(\dots)$ as they effectively change the program state (variable values, excluding guards) and control. Function calls are either inlined, in case equivalence could not be proven for them, or left as is, in case they are equivalent or are external system calls. Continuing the construction process, a syntactical diff [12] is computed over the vectors (I, I') . One of the inputs to the diff process is VC as it is needed to identify correlated variables and the diff comparison will regard commands differing by variable names which are correlated by VC as equal. The result of the last step will be a vector I_{Δ} specifying for each command in I, I' whether it's an added command in P' (for I') marked $+$, a deleted command from P (for I) marked $-$, or a command existing in both versions marked $=$. This diff determines the order in which the commands will be interleaved in the resulting $P \bowtie P'$ as we will iterate over the result vector I_{Δ} and use it to construct

the correlating program. We remind that since I, I' contain only the imperative commands, we cannot use it directly as $P \bowtie P'$. Instead we will use the imperative commands as markers, specifying which chunk of program from P_G or P'_G should be taken next and put in the result. The construction goes as follows: iterate over I_{Δ} and for every command c (c') labeled l_c ($l_{c'}$):

- read P_G (P'_G) up to label l_c ($l_{c'}$) including into block B_c ($B'_{c'}$)
- for $B'_{c'}$, tag all variables in the block.
- emit the block to the output.
- delete B_c ($B'_{c'}$) from P_G (P'_G).

The construction is now complete. We only add that at the start of the process, we strip P'_G of its prototype and add declarations for the tagged input variables, initializing them to the untagged version. As mentioned CP is also a product of the construction, and it's defined using $=$ commands: after two $=$ commands are emitted to the output, we add an instrumentation line, telling the analysis of the correlation point. One final observation regarding the correlating program is that it is a legitimate program that can be run to achieve the effect of running both versions. We plan to leverage this ability to use dynamic analysis and testing techniques such as fuzzing [19] and directed automated testing [3] on the correlating program in our future work.

7. Evaluation

We evaluated DIZY on a number of challenging real world programs where the patches affect numerical variables. As benchmarks, we used several programs from the GNU core utilities (differencing versions 6.10 and 6.11), as well as a few handpicked patches taken from the Linux kernel and the Mozilla Firefox web browser. For these programs, DIZY was able to precisely describe the difference.

7.1 Prototype Implementation

We implemented a correlating compiler named CCC which creates correlating programs from any two C programs. We also implemented a differencing analysis for analyzing correlated programs. Both tools are based on LLVM and CLANG compiler infrastructure. We analyze C code directly since it is more structured, has type information and keeps a low number of variables, as opposed to intermediate representation. We also benefit from our delta being computed over original variables. As mentioned in Section 6, we normalize the input programs before unifying them for a simpler analysis. Our analysis is intra-procedural and we handle function calls by either modularly proving their equivalence and assuming it once encountered or, in case equivalence could not be proved, by inlining. Calls to external system functions do not change local state in our examples and thus were ignored. We used the APRON abstract numerical domain library and conducted our experiments using several domains including octagon [17] and polyhedra [7]. All of our experiments were conducted running on a Intel(R) Core-i7(TM) processor with 4GB.

For brevity, when presenting results we only show the relevant code fragments, however, our analysis has been applied to the full original program. A complete version of the results is available at <http://www.cs.technion.ac.il/~nimi/dizy>.

7.2 Results

Producing delta from abstract state Fig. 11 shows a patch made to the `net/sunrpc/addr.c` module in the Linux kernel SUNRPC implementation v2.6.32-rc6 which is aimed at removing an off-by-two array access out of bounds violation in the original program. Although simple, this example shows two advantages of DIZY: (i) the ability to analyze programs without the need to run them

```

size_t rpc_uaddr2sockaddr (const size_t uaddr_len, ...) {
    char buf[ RPCBIND_MAXUADDRLEN ];
    ...
    - if ( uaddr_len > sizeof ( buf ))
    + if ( uaddr_len > sizeof ( buf ) - 2:)
        return 0;
    ...
    (*)1
    buf [ uaddr_len ] = '\n';
    buf [ uaddr_len + 1] = '\0';
    ...
}

```

Figure 11: Linux kernel `addr.c` v2.6.32-rc6 module with patch.

σ_1 :
returned' = true
returned = false
uaddr_len' ≤ RPCBIND_MAXUADDRLEN - 2
uaddr_len' ≥ RPCBIND_MAXUADDRLEN

Figure 12: Difference for SUNRPC implementation

```

nsresult SetTextInternal (int textLength, int aCount,
                        int aLength, int aOffset,
                        PRUnichar * aBuffer) {
    PRInt32 newLength = textLength - aCount + aLength ;
    PRUnichar * to;
    ...
    + if ((unsigned)newLength > (1 << 29))
    +   return NS_ERROR_DOM_DOMSTRING_SIZE_ERR;
    (*)1
    memcpy (to + aOffset , aBuffer ,
            aLength * sizeof ( PRUnichar ));
    ...
}

```

Figure 13: Firefox `nsGenericDOMDataNode` module with patch.

and (ii) the ability to capture differences beyond output values. The result of analyzing the correlated program is shown in Fig. 12.

The difference is captured by one sub-state where the execution has ended in the patched program but continues in the original. We instrument the correlating program with a return flag to capture the difference as otherwise equivalence holds (none of the original variables change as we do not track arrays).

Another example, taken from CVE-2010-1196 advisory regarding Firefox’s heap buffer overflow on 64-bit systems is shown in Fig. 13 (vulnerable part of the function only). Firefox 3.5 and 3.6 (up to 3.6.4) contain a heap buffer overflow vulnerability which is caused by an integer overflow. Due to the amount of data needed to trigger the vulnerability (> 8GB), this is only exploitable on 64-bit systems. The vulnerable code is found in `/content/base/src/nsGenericDOMDataNode.cpp` of the Mozilla code base and was adapted to C for analysis purposes.

Here, we need to describe a more complex and non-convex constraint that leads to difference. Running `DIZY` with partitioning produces the result shown in Fig. 14 (a) (only the part of the state which breaks equivalence is shown).

The difference in state is described correctly as indeed the only change in values for the patch scenario would be the return value and the early return of the patched version. However, we did not preserve the conditional constraints as they are non-convex. Running the same analysis with no partitioning (this is feasible as the procedure does not loop) produces the result shown in Fig. 14 (b).

σ_1 :
returned' = true
returned = false
return value' = NS_ERROR
(a)

σ_1 :	σ_2 :
returned' = true	returned' = true
returned = false	returned = false
newLength > 536870912	newLength > -3758096384
	newLength < 0
return value' = NS_ERROR	return value' = NS_ERROR
(b)	

Figure 14: Difference for Firefox `nsGenericDOMDataNode`, (a) with a single polyhedra; (b) set of polyhedra.

```

int input_position;

bool char_to_clump(char c) {
    int width;
    ...
    + if (width < 0 && input_position == 0) {
    +   chars = 0;
    +   input_position = 0;
    + } else if (width < 0 && input_position <= -width) {
    +   input_position = 0;
    + } else {
    +   input_position += width;
    + }
    (*)1
    ...
    return chars;
}

```

Figure 15: Original and patched version of `coreutils pr.c`’s `char_to_clump` procedure

σ_1 :	σ_2 :	σ_3 :
input_position ₀ = 0	input_position ₀ < -width	input_position ₀ < -width
chars' = 0	input_position ₀ < 0	input_position ₀ > 0
input_position = width	input_position' = 0	input_position' = 0
input_position < 0	input_position < width	input_position > width
input_position' = 0	width < 0	input_position <= 0

Figure 16: Difference for `coreutils pr.c` `char_to_clump`

Now we see that the `(unsigned)newLength > (1 << 29)` constraint has been successfully encoded in two offending states, each holding a part of the problematic range.

Capturing complex delta Fig. 15 shows a patch made to the `char_to_clump` function in version 6.11 of `coreutils`. The patch replacing the execution of the line `input_position += width`, which originally executed unconditionally, with a conditional structure that in the new version, allows the line to execute only under certain complex conditions. Since the variables handled in this patch (the global `input_position` and return value `chars`) emit output, describing how their values changed and under which terms is important, especially as the patch cannot be easily parsed by a programmer to understand its meaning. The result of our analysis at the return point is shown in Fig. 16. These results include information regarding initial values of parameters for improved precision (this is one of `DIZY`’s features).

The result convey the difference in the output variable values alongside some of conditions under which the difference occurs. The result is composed of three sub-states featuring difference and adhere to two added paths in the patched program. The first sub-state belongs to the first branch in the added conditional: the difference is comprised of (i) the new value of `input_position` is 0 as

```

int logicalValue(int t) {
    if (!(curr - t >= 100)) {
        return old;
    } else {
        int val = 0;
        for (int i = 0;
             i < data.length; i++)
            val = val + data[i];
        old = val;
        return val;
    }
}

const int THRESHOLD = 100;
int logicalValue(int t) {
    int elapsed = curr - t;
    int val = 0;
    if (elapsed < THRESHOLD) {
        val = 1;
    } else {
        for (int i = 0;
             i < data.length; i++)
            val = val + data[i];
        old = val;
        return val;
    }
}

```

Figure 17: Two versions of the `logicalValue` procedure taken from [20].

σ_1 :
$\text{curr} - t < 100$
$\text{return value} = \text{old}$
$\text{return value}' = 1$

Figure 18: Difference for `logicalValue`

opposed to it being `width` in the former version (the analysis took the `input_position += width` line into account and incorporated knowing that `input_position = 0` from the branch condition). The analysis also deduced that the old `input_position` is negative under the same input as the branch condition dictates that `width` is negative. (ii) `chars` in the new program is 0 under this path. The two other sub-states adhere to the second added path in the conditional and track a difference for `input_position` alone, basically stating that `input_position` under this path used to assume values in ranges $[-\infty, \text{width}]$ and $[\text{width}, 0]$ but now is simply 0. The splitting of this path into two cases is a result of expressing the non-convex `input_position \neq 0` condition from the first branch conditional using two sub-states. The result also describes constraints on the procedure's input under which the difference exist. Another product of the analysis, which we do not show here, are sub-states describing paths which the patch did not affect.

Maintaining Equivalence and Reporting Difference in Loops

Fig. 17 shows two version of the java `logicalValue()` method taken from [20], adapted to C. This example features semantic preserving refactoring modification (introducing the `elapsed` variable and `THRESHOLD` constant, simplifying a conditional and moving the return statement out of branch block) and one semantic change where 1 is returned instead of `old` in case `curr - t < 100`. The challenge in this example is proving equivalence over the loop branch and reporting difference for the negated path. Using a separate analysis, we would have to deduce at the following loop invariant: $\text{val} = \sum_{i=0}^{\text{data.length}} \text{data}[i]$ in order to show equivalence. However, as our abstraction focuses on variable relationships and our correlating program allows us to interleave the two loops in lock-step, all our analysis needs to deduce is the $\text{val} = \text{val}'$ constraint. As we apply widening to converge, the constraint will be kept, allowing us to establish equivalence for the looping path. DIZY reports the state shown in Fig. 18 state for the exit point of `logicalValue()`. We note that in [20] the example was run by unrolling 2 steps of the loop.

Next we shall explore a different loop scenario where all paths in the programs contain loops and only some of them maintain equivalence. Fig. 19 shows part of coreutils `md5sum.c` `bsd_split_3` function that was patched in version 6.11 to disallow 0-length inputs. Although this patch seems trivial, analyzing it is challenging as it affects the behavior of loops i.e. unbound path lengths. The main challenge in this example, is separating the

```

bool bsd_split_3 (char *s, size_t s_len, ...) {
    int i = s_len;
    i--;
    + if (s_len == 0) return false;
    while (i && s[i] != '\0') { (*)_1
        i--;
    }
    ...
    (*)_2
}

```

Figure 19: Original and patched version of coreutils `md5sum.c`'s `bsd_split_3` procedure

$(*)_1$:	σ_2 (equivalent):	$(*)_2$:	σ_1 (equivalent):
σ_1 :	$s_len' = s_len$	σ_1 (equivalent):	$s_len' = s_len$
$s_len = 0$	$i' = i$		$i' = i$
$s_len' = 0$	$s_len' - 1 \geq i'$		$s_len' - 1 \geq i'$
$i \leq -1$			

(a)

$(*)_1$:	σ_2 (equivalent):	σ_3 (equivalent):	$(*)_2$:	σ_1 (equivalent):
σ_1 :	$s_len' = s_len$	$s_len' = s_len$	σ_1 (equivalent):	$s_len' = s_len$
$s_len = 0$	$i' = i$	$i = 0$		$i' = i$
$s_len' = 0$	$s_len' - 1 \geq i'$	$s_len' \geq 1$		$i = 0$
$i \leq -1$				$s_len' \geq 1$

(b)

Figure 20: Difference for `bsd_split_3`

path where `s_len` is 0, which results in the loop index `i` ranging within negative values (producing an array access out of bounds fault), from the rest of the behaviors that maintain equivalence, throughout the widening process which is required for the analysis to reach a fixed point. The result of our analysis with partitioning is shown in Fig. 20 (a) (per differencing points $(*)_1, (*)_2$).

We can see the analysis successfully reports a difference for the singularity point $s_len = 0$ inside the loop, precisely describing the scenario where i' is negative. We can also see the other equivalent state existing within the loop which depicts the results of the widened analysis for all other paths (the $s_len \neq 0$ constraint is not existing there due to partitioning as we will soon show). The differencing sub-state will be omitted once we move past the loop as the $i \leq -1$ constraint will not allow it to exist beyond the loop body thus we are left with the equivalent state alone after the loop which correctly expresses the fact that the programs are equivalent at this point (since both i 's converged at 0). We can see that the result at the second differencing point has lost precision since it does not reflect the $i = 0$ constraint. The loss of this constraint is, again, due to partitioning as both sub-states that describe exiting the loop and the one describing entering the loop, hold equivalence for all variables and are joined to together and lose the extra constraint information. If we analyze the same example with no partitioning we get the result of Fig. 20 (b).

Which further separates the paths in the program, allowing for a different sub-state for the $i = 0$ and $i \neq 0$ substates (again, the $i \neq 0$ constraints was lost when joining together the $i > 0$ and $i < 0$ states as they both adhere to the same path and hold the same guard values). This extra precision is beneficial, but we still managed to supply a satisfactory result using the more scalable partitioning by equivalence technique.

8. Related Work

Our work has been mainly inspired by recent work identifying program differencing as having vast security implications [2, 23]

as well as advancements made in the field of under-approximations of program equivalence [8, 15, 20, 21].

The problem of program differencing is fundamental [9] and early work mainly focused on computing syntactical difference [12]. These solutions are an important stepping stone and we used syntactical diff as a means to achieve interleaving of programs in our correlating program for better analysis results. Another possibility for creating this program is to rely on the editing sequence that creates the new version from the original program [10].

Jackson and Ladd [13] proposed a tool for computing data dependencies between input and output variables and comparing these dependencies along versions of a program for discovering difference. This method may falsely report difference as semantic difference may occur even if data dependencies have not changed. Furthermore, data dependencies offer little insight as to the meaning of difference i.e. input and output values. Nevertheless, this was an important first step in employing program analysis as a means for semantic differencing.

Several works on the problem of equivalence of combinatorial circuits [5, 16, 18] made important contributions in establishing the problem of equivalence as feasible, producing practical solutions for hardware verification.

We rely on classic methods of abstract interpretation [6] for presenting an over approximating solution for semantic differencing and equivalence. To achieve this we devised a static analysis over a correlating program. The idea of a correlating program is similar to that of self-composition [24] except that we compose two different programs in a interleaving designed to maintain a close correlation between them. The use of a correlating construct for differencing is novel as previous methods mainly use sequential composition [8, 20, 21], disregarding possible program correlation.

We base our analysis on numerical abstractions [7, 17] that allow us to reason about variables of different programs. The abstraction is further refined in a way similar to trace partitioning [22] with an equivalence-based partitioning criteria.

Symbolic execution based methods [20, 21] offer practical equivalence verification techniques for loop and recursion free programs with small state space. These works complement each other in regards to reporting difference as one [20] presents an over approximating description of difference they call differential summaries and the other [21] presents an under approximating description including concrete inputs for test cases demonstrating difference in behavior. An interesting question is how could these methods be combined iteratively to achieve better precision. Also, this work can be used to complement our work in cases where equivalence could not be proven and the description of difference can be leveraged for the extraction of concrete input that leads to offending states.

Bounded model checking based work [8] presents the notion of partial equivalence which allows checking for equivalence under specific conditions, supplied by the user but are bound by loops. They employ a technique based on theorem provers for proving an equivalence formula which embeds program logic (in SSA form) alongside the requirement for input and output equivalence and user provided constraints.

[1] introduced a correlating heap semantics for verifying linearizability of concurrent programs. In their work, a correlating heap semantics is used to establish correspondence between a concurrent program and a sequential version of the program at specific linearization points.

9. Conclusions

This work presented an abstract interpretation approach for program equivalence and differencing. We defined a correlating program construct, a reduction of the the product program, that al-

lowed mutual reasoning and establishing of equivalence. For the purpose of this mutual analysis, we defined a partially disjunctive correlating abstract domain, that allowed us to over approximate variable relationships along paths according to equivalence criteria. We also defined a widening operator which over approximates looping paths correctly. We showed that this approach is feasible and can be applied successfully to challenging real world patches.

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