

Fluid Updates: Beyond Strong vs. Weak Updates

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Strong and Weak Updates

• In *static analysis*, there is a distinction between two kinds of updates to *abstract* memory locations:

Strong vs weak

Consider points-to edge from abstract location A to B:



• A *strong update* removes existing points-to edges from location A and adds new edge to C.

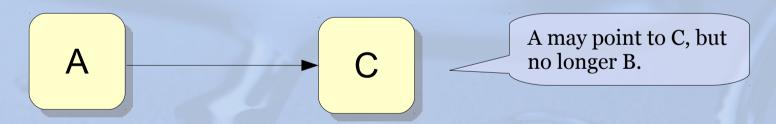


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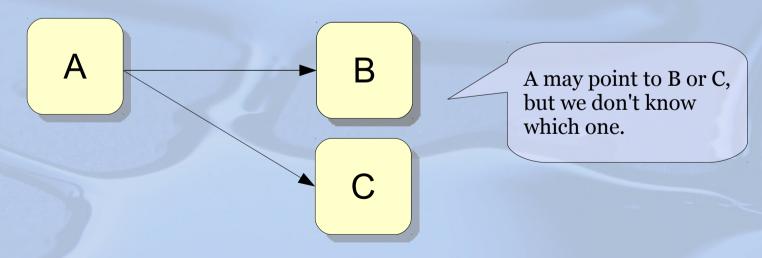


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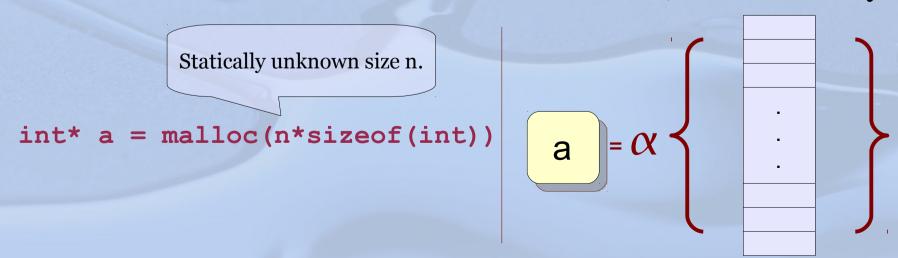


• A *weak update* adds a new edge without removing existing edges.



When are Strong Updates Applicable?

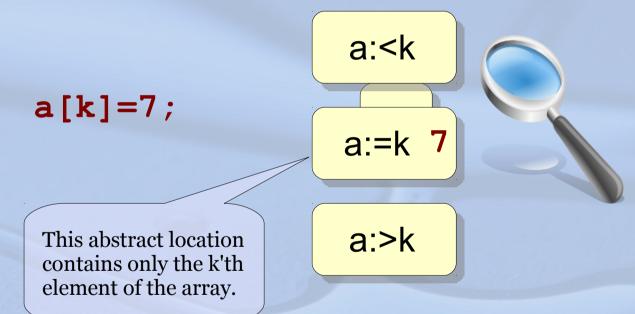
- In general, it is preferable to apply strong updates because it is more precise.
- However, we can only apply strong updates if the abstract location corresponds to one concrete location.
 - This makes it difficult to apply strong updates to elements of unbounded data structures, such as arrays.





Strong Updates to Arrays

Many approaches overcome this difficulty by creating partitions of the array.





Strong Updates to Arrays

Many approaches overcome this difficulty by creating partitions of the array.

```
for (int k=0;
    k < size; k++)
    a[k] = 7;
    a:>k
```

We need to avoid creating an unbounded number of partitions.



Strong Updates to Arrays

 Many approaches overcome this difficulty by creating partitions of the array.

```
for(int k=0;
    k < size; k++)
    a[k] = 7;</pre>
```

a:<=k 7

a:>k

Finitizes the number of partitions.

Drawbacks

- This approach has some drawbacks:
 - Can create large number of explicit partitions
 - Limits scalability
 - *Heuristics* to decide when to focus/blur
 - Can be difficult to automate
 - "Shape" of partitions are fixed a priori
 - Typically contiguous ranges
- The rest of this talk is about a *heap abstraction* and *update mechanism* that does not have these drawbacks.



Symbolic Heap Abstraction

• Key Idea #1:

• Arrays are represented by abstract locations that are qualified by *index variables*.

$$\langle a \rangle_i$$

• These index variables range over *possible indices* of the array.



Symbolic Heap Abstraction

- Key Idea #2:
 - Constraints on index variables select which concrete elements in the source location point to which concrete elements in the target location.

This edge selects all elements of array a whose indices are in the range [0, size).

Each element in array a points to the corresponding element in array b.

$$\begin{array}{c|c} \langle a \rangle_{i_1} & 0 \leq i_1 < \textit{size} \land i_2' = i_1 \\ & \langle b \rangle_{i_2} \\ & \text{By convention, target's index variables are primed.} \end{array}$$

Fluid Updates

To perform a *fluid update* on the symbolic heap:

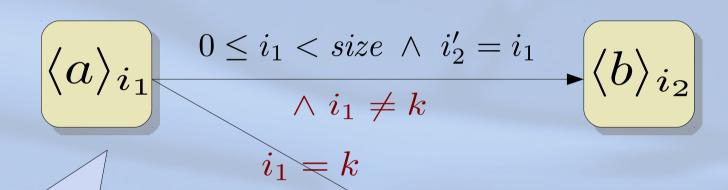
• Compute a constraint Φ specifying which elements in an abstract location A are modified by the update.

• Now, the negation of Φ , $\neg \Phi$, specifies the concrete elements in A *not affected* by the update.

• Hence, a fluid update conjoins $\neg \Phi$ with all existing edges from A, while adding new edge under Φ .



Fluid Update Example



Constraint describing elements not affected by update: $i_1 \neq k$

*c

The constraint describing updated elements is:

$$i_1 = k$$

Consider a statement: a[k] = c



What if we aren't sure which elements are updated?

- In the previous example, we could precisely describe the *exact* set of elements that were written to.
- In general, we cannot always precisely specify which elements are updated.

```
for(i=0; i<size; i++)
{
    if(rand())
        a[i] = NULL;
}</pre>
```

All elements in range [0, size)

may be set to NULL,

but no element

must be set to NULL.

So, an *overapproximation* of the elements updated in the loop is:

```
0 \le i_1 < size
```

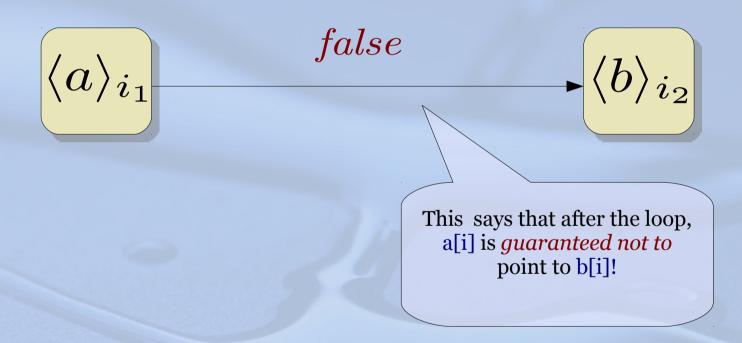


Fluid Update with Overapproximations

Suppose we *erroneously* use the overapproximation for the update



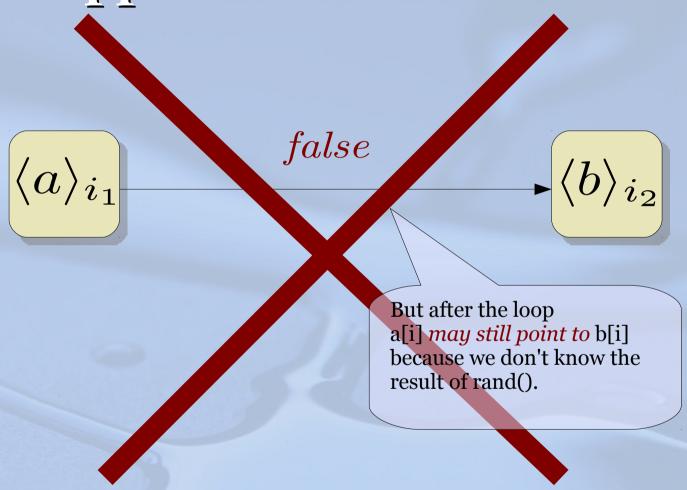
Fluid Update with Overapproximations



Suppose we *erroneously* use the overapproximation for the update



Fluid Update with Overapproximations



Suppose we *erroneously* use the overapproximation for the update



What went wrong?

• The negation of an overapproximation is an underapproximation.

 Hence, if the fluid update uses an overapproximation, we underapproximate the set of elements not affected by the update.

UNSOUND!





Bracketing Constraints

• Solution: Constraints in the symbolic heaps are pairs of constraints, called *bracketing constraints*:

 $\langle \phi_{may}, \phi_{must} \rangle$

This specifies which elements in the source location *may* point to which elements in the target.

This specifies which elements in the source location *must* point to which elements in the target.



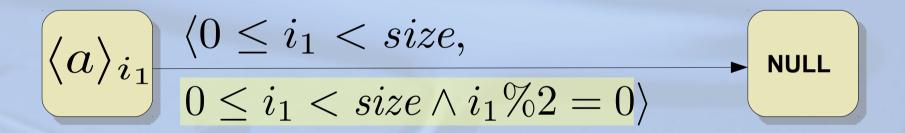
Bracketing Constraint Example

Any element in the range [0, size) *may* point to null.

All **even** elements in the range [0, size) **must** point to null.



Bracketing Constraint Example



But, **odd** elements in the range [0, size) **may or may not** point to null.

- If we use bracketing constraints, the fluid update operation described earlier is sound.
 - This is because negating a bracketing constraint yields correct over- and underapproximations.
- In particular, the negation of a bracketing constraint $\langle \phi_{may}, \phi_{must} \rangle$ is given by:



Consider again the following initial symbolic heap:

If a bracketing constraint $\langle \phi_{may}, \phi_{must} \rangle$ is *precise*, i.e., $\phi_{may} \Leftrightarrow \phi_{must}$ then we write a single constraint instead of a pair.



Consider again the following initial symbolic heap:



From now on, think of a single constraint as a bracketing constraint where *may* and must conditions are the same.



Consider again the following initial symbolic heap:



The negation of this constraint is:

```
\langle \neg false, \ \neg (0 \le i_1 < size) \rangle
```

```
for(i=0; i<size; i++)
{
    if(rand())
        a[i] = NULL;
}</pre>
```

A bracketing constraint describing the update condition is: $\langle 0 < i_1 < size, \ false \rangle$



Consider again the following initial symbolic heap:

The negation of this constraint is:

```
\langle true, i_1 < 0 \lor i_1 \ge size \rangle
```

```
for(i=0; i<size; i++)
{
    if(rand())
        a[i] = NULL;
}</pre>
```

A bracketing constraint describing the update condition is: $\langle 0 < i_1 < size, \ false \rangle$

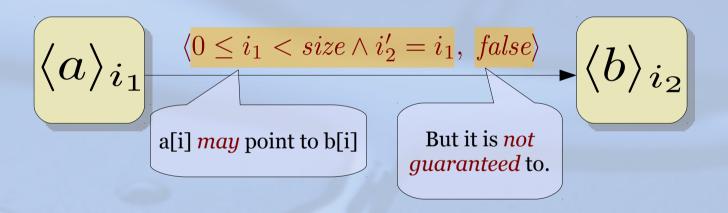
Example Revisited

Consider again the following initial symbolic heap:



Example Revisited

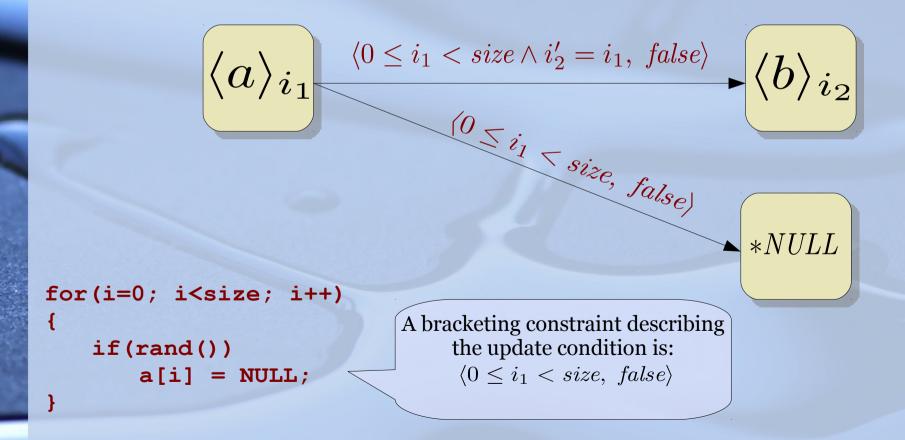
Consider again the following initial symbolic heap:



```
for (i=0; i<size; i++) { A bracketing constraint describing the update condition is: \langle 0 \le i_1 < size, \ false \rangle }
```

Example Revisited

Consider again the following initial symbolic heap:



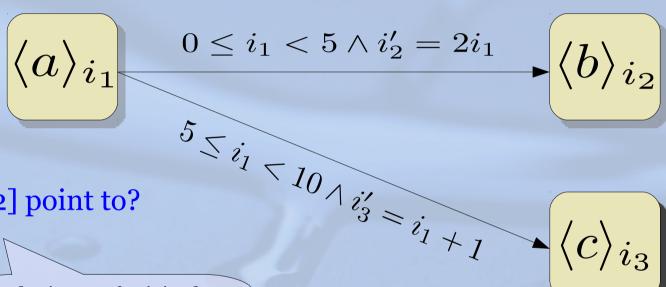


 So far, we described what a symbolic heap looks like and defined a fluid update operation on this heap.

 But how do we interpret/understand the facts that are encoded by the symbolic heap?



Consider the following:

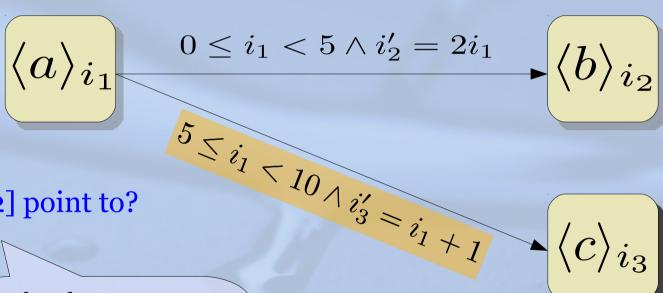


Where does a[2] point to?

Substitute 2 for i₁ in the edge constraints.



Consider the following:



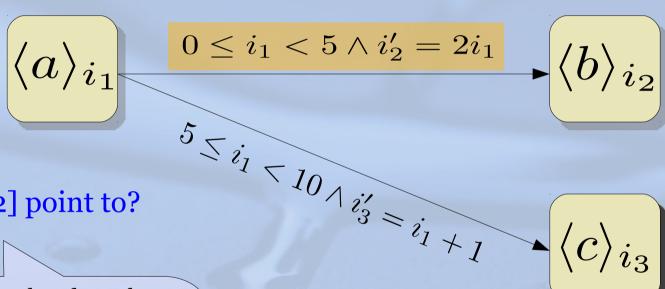
Where does a[2] point to?

For the edge to c, the constraint 5 <= 2 is unsatisfiable, hence, a[2] cannot point to c[3].

$$\langle c \rangle_{i_3}$$



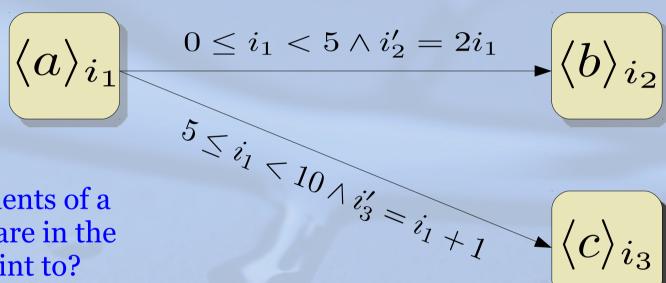
Consider the following:



Where does a[2] point to?

For the edge to b, substituiting 2 yields $i_2' = 4$; so a[2] points to b[4].

Consider the following:

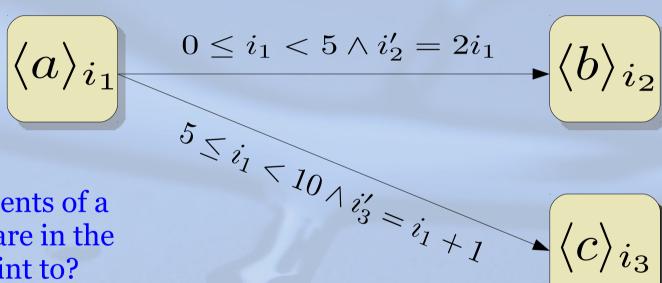


Where do elements of a whose indices are in the range [0, 3] point to?

Here,
$$0 \le i_1 \le 3$$



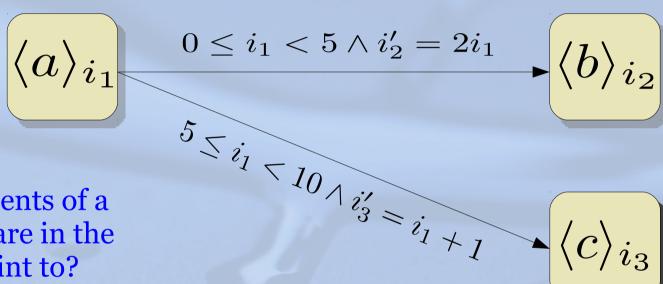
Consider the following:



Where do elements of a whose indices are in the range [0, 3] point to?

So, we need a generalized form of substitution, i.e., existential quantifier elimination.

Consider the following:



Where do elements of a whose indices are in the range [0, 3] point to?

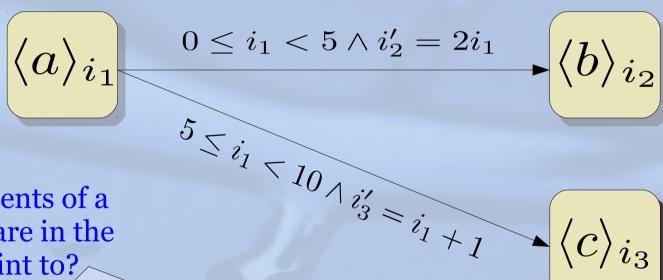
We can determine the points-to targets of elements whose indices are in [0, 3], by *eliminating the quantifier* from the following formula (for the edge to b):

$$\exists i_1. \ 0 \le i_1 \le 3 \ \land (0 \le i_1 < 5 \ \land i'_2 = 2i_1)$$



Interpreting the Symbolic Heap

Consider the following:



Where do elements of a whose indices are in the range [0, 3] point to?

This yields:
$$i_2'\%2 = 0 \land 0 \le i_2' \le 6$$

Elements of a in range [0,3] point to *even* elements of b *in range* [0, 6].



Interpreting the Symbolic Heap

* To summarize, *traversing edges* (i.e., going from one abstract location to its points-to target) requires *existential quantifier elimination*.

 Similar to image computation and computation of strongest post-conditions.



Implementation

- The combination of symbolic heap abstraction and fluid updates forms the core of the *Compass* program verification system for C programs.
 - Compass is path- and context-sensitive
 - Summary-based
 - Used for checking *memory safety properties* (buffer overruns, null errors, uninitialized reads, casting errors, ...) as well as arbitrary *user-provided* assertions.
 - Has successfully been scaled to real programs in the range of a few 10,000 lines of code.

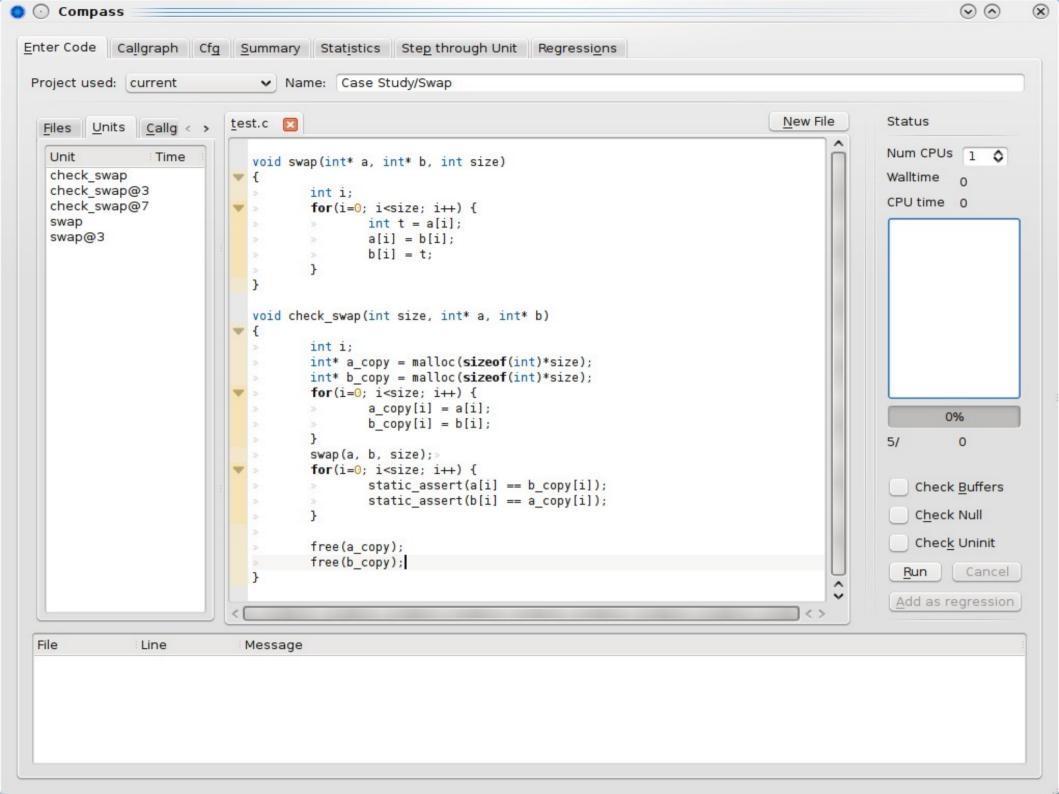


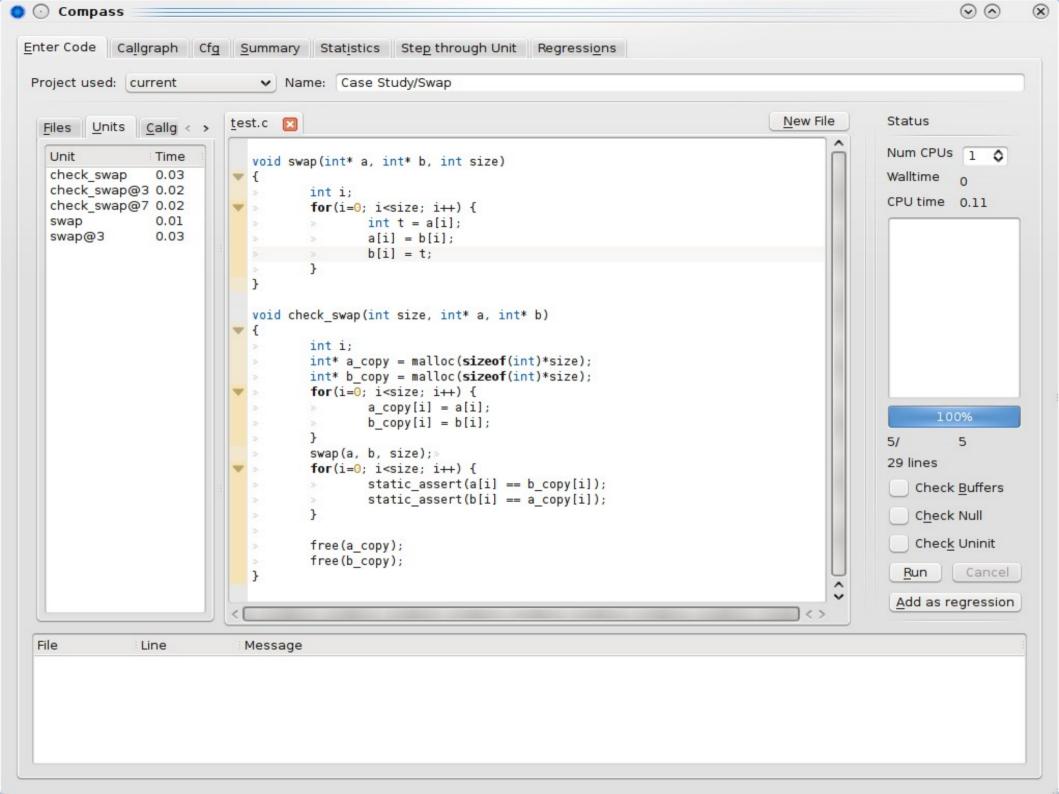
 We first evaluate this technique on a set of challenging array benchmarks.

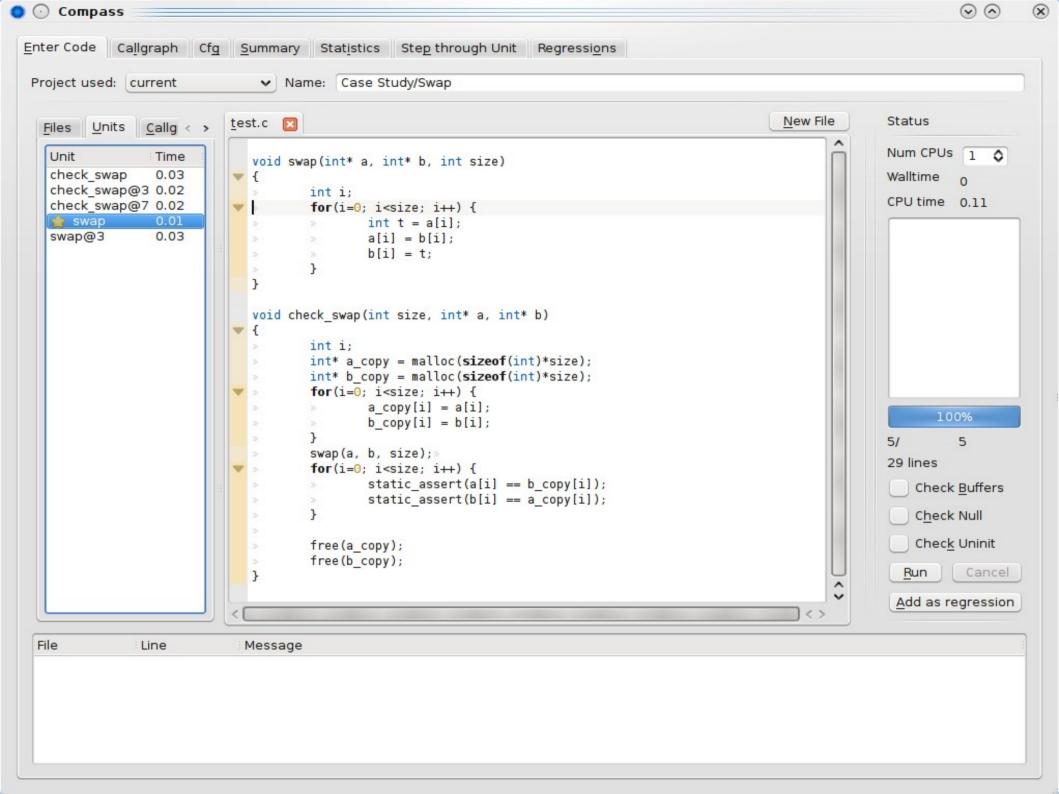
Program	Time	Memory	#Sat	Solve
			queries	time
init	0.01s	< 1 MB	172	0s
init_nonconst	0.02s	< 1 MB	184	0.01s
init_partial	0.01s	< 1MB	166	0.01s
init_partial_buggy	0.02s	< 1 MB	168	0s
init_even	0.04s	< 1 MB	146	0.04s
init_even_buggy	0.04s	< 1 MB	166	0.03s
2D_array_init	0.04s	< 1 MB	311	0.04s
сору	0.01s	< 1 MB	209	0.01s
copy_partial	0.01s	< 1 MB	220	0.01s
copy_odd	0.04s	< 1 MB	243	0.02s
copy_odd_buggy	0.05s	< 1 MB	246	0.05s
reverse	0.03s	< 1 MB	273	0.01s
reverse_buggy	0.04s	< 1 MB	281	0.02s
swap	0.12s	2 MB	590	0.11s
swap_buggy	0.11s	2 MB	557	0.06s
double_swap	0.16s	2 MB	601	0.1s
strcpy	0.07s	< 1 MB	355	0.04s
strlen	0.02s	< 1 MB	165	0.01s
strlen_buggy	0.01s	< 1 MB	89	0.01s
memcpy	0.04s	< 1 MB	225	0.04s
find	0.02s	< 1 MB	119	0.02s
find_first_nonnull	0.02s	< 1 MB	183	0.02s
append	0.02s	< 1 MB	183	0.01s
merge_interleave	0.09s	< 1 MB	296	0.07s
merge_interleave				
_buggy	0.11s	< 1 MB	305	0.09s
alloc_fixed_size	0.02s	< 1 MB	176	0.02s
alloc_fixed_size_buggy	0.02s	< 1 MB	172	0.02s
alloc_nonfixed_size	0.03s	< 1 MB	214	0.02

Also very memory efficient.

Very fast despite being very precise.









Callgraph Cfg Summary Statistics Step through Unit Enter Code

Summary Unit: /home/tdillig/.compass/current/test.c/swap<[nonret] (int*, int*, int)>

-4	*(b)[i4].size
0	*(b)[i4]

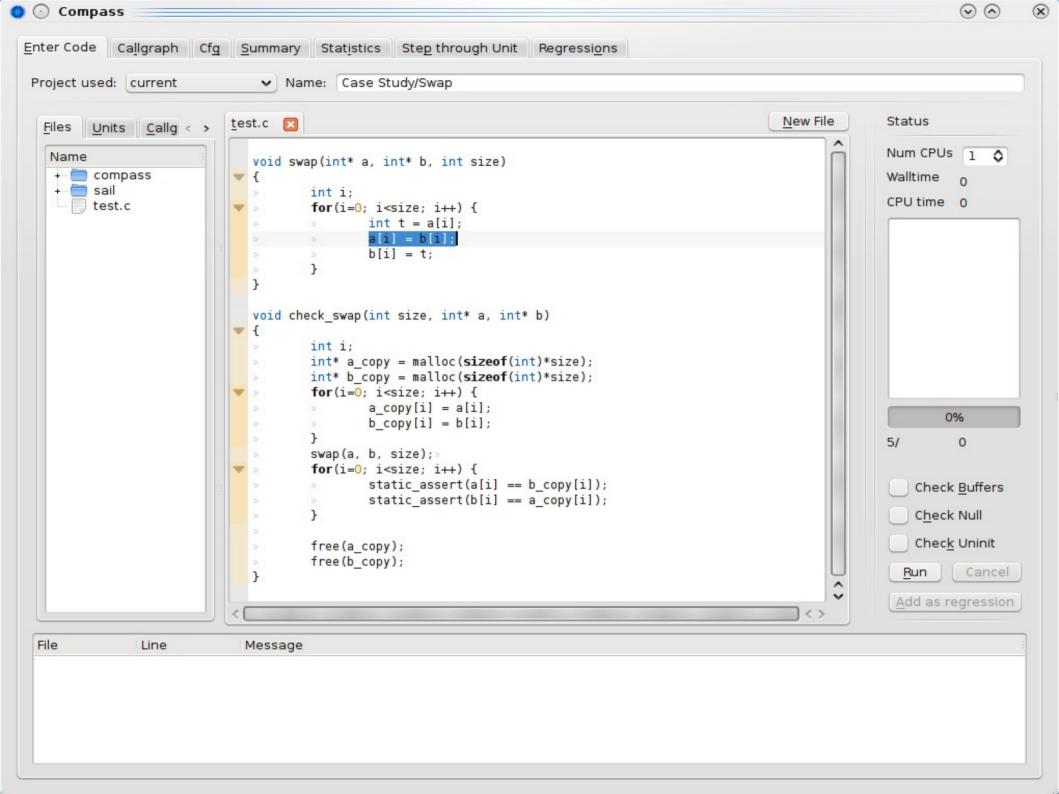
$$(j2 \ge 0 \text{ i} 2 = i4 \text{ j} 2 \le \text{size})[50]$$

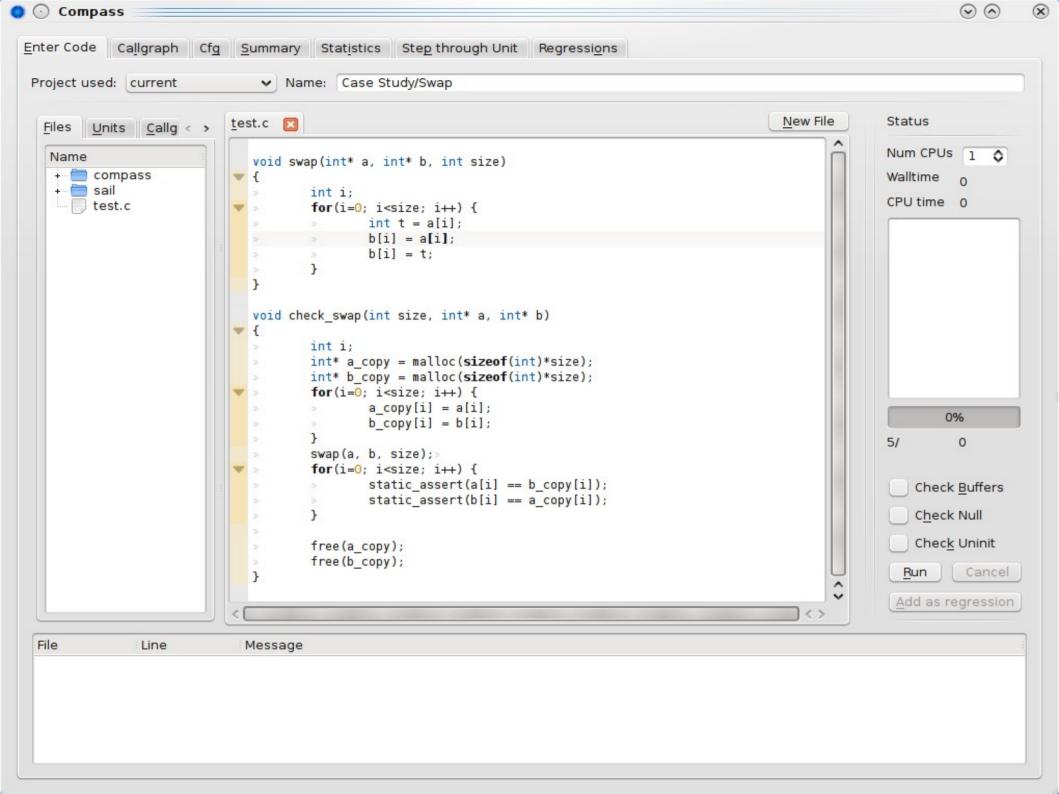
((a)[i2])

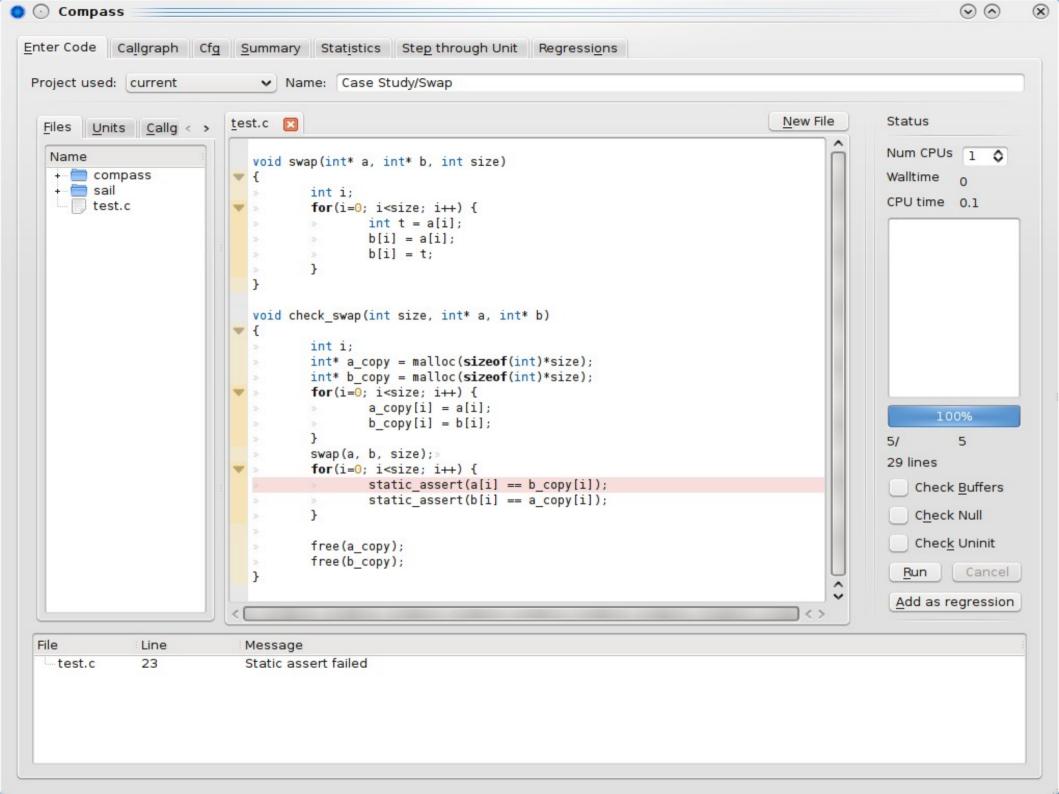
$$(i2 = j4\& i2 > = 0\& size > i2)[49]$$

((b)[i4])

Return Condition: true Loop Counters: Error trace summary:









Unix Coreutils

- A collection of widely used command-line utilities.
- Heavily use arrays, pointers, and string buffers.
- Precise heap analysis necessary for successful verification.

Program	Lines	Total Time	Memory	#Sat queries	Solve Time
$\mathbf{hostname}$	304	0.13s	5 MB	1533	0.12s
${f chroot}$	371	0.13s	3 MB	1821	0.10s
rmdir	483	1.05s	12 MB	3461	1.02s
su	1047	1.86s	32 MB	6088	1.69s
mv	1151	0.70s	21 MB	7427	0.68s
Total	3356	3.87s	73 MB	20330	3.61

Compass verified absence of buffer overruns and null dereferences with *no false positives* and *no annotations*.

Thank You

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