Allocation Removal by Partial Evaluation in a Tracing JIT

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- Interpretation overhead
- Type dispatching
- Boxing of primitive types

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; $y = x + c$ evaluated in an interpreter:

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What to do?

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- Add an optimization that can deal with heap operations

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- Use a JIT compiler
- Add an optimization that can deal with heap operations
- optimize short-lived objects
- remove some of the redundant type checks

Overview

Experimental Context

Proposed Optimization

Benchmarks

Our Experimental Context is the PyPy Project

A general environment for implementing dynamic languages

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Approach

- write an interpreter for the language in RPython
- compilable to an efficient C-based VM

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A general environment for implementing dynamic languages

Approach

- write an interpreter for the language in RPython
- compilable to an efficient C-based VM
- (RPython is a restricted subset of Python)

PyPy Uses a Novel Tracing JIT

the feature that makes PyPy interesting:

- a meta-JIT, applicable to many languages
- needs a few source-code hints (or user annotations) in the interpreter
- JIT is a tracing JIT compiler

Tracing JITs Compile by Observing an Interpreter

- VM contains both an interpreter and the tracing JIT compiler
- JIT works by observing and logging what the interpreter does
 - for interesting, commonly executed code paths
 - produces a linear list of operations (trace)
- trace is optimized and then turned into machine code

The Advantages of Tracing JITs

- Traces are interesting linear pieces of code
- most of the time correspond to loops
- everything called in the trace is inlined
- can perform good optimizations on the trace
- rarer paths run by the interpreter

Example Trace

```
Trace of x = a + b; y = x + c:
guard_class(a, Integer)
guard_class(b, Integer)
i1 = get(a, intval)
i2 = get(b, intval)
i3 = int_add(i1, i2)
x = new(Integer)
set(x, intval, i3)
```

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guard_class(x, Integer)
guard_class(c, Integer)
i4 = get(x, intval)
i5 = get(c, intval)
i6 = int_add(i4, i5)
y = new(Integer)
set(y, intval, i6)
return(y)
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Optimized Example Trace

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Contribution of our Paper

- a simple, efficient and effective optimization of heap operations in a trace
- using online partial evaluation
- fully implemented and in use in large-scale interpreters

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Ingredients

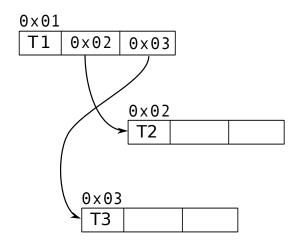
- a slightly simplified model for objects on the heap
- operational semantics of trace operations that manipulate the heap
- optimization rules for those operations, following the operational semantics



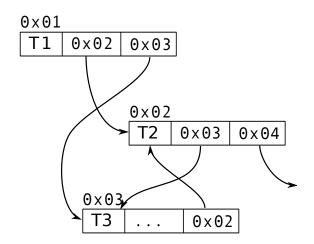
Heap Model

0×01 T1

Heap Model



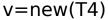
Heap Model

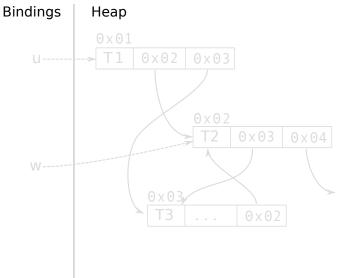


Operations Manipulating Heap Objects

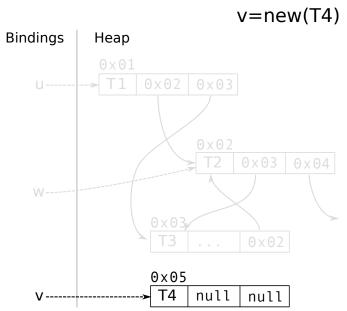
- v = new(T) makes a new object
- u = get(w, F) reads a field out of an object
- set(v, F, w) writes a field of an object
- guard(v, T) checks the type of an object

Operations: New

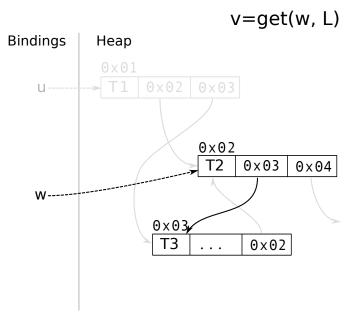




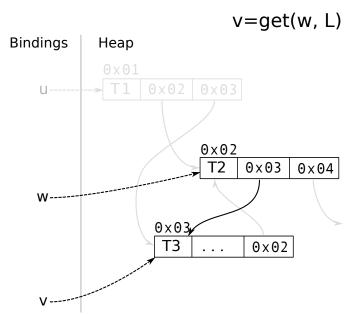
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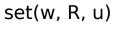
Operations: Get

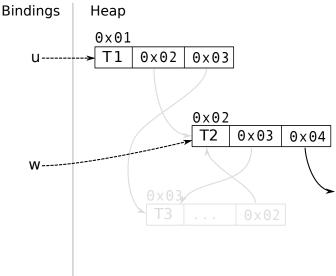


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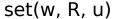


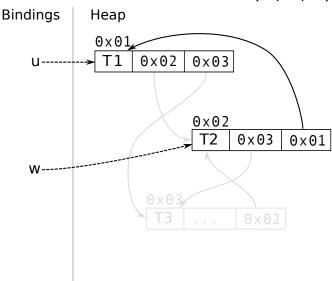
Operations: Set



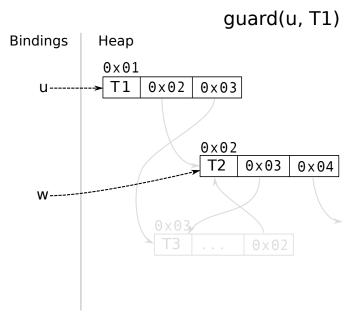


Operations: Set

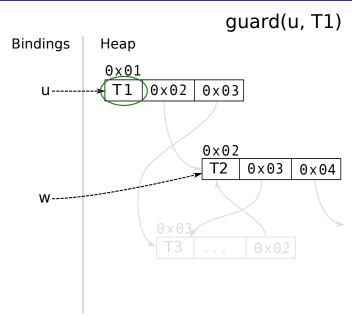




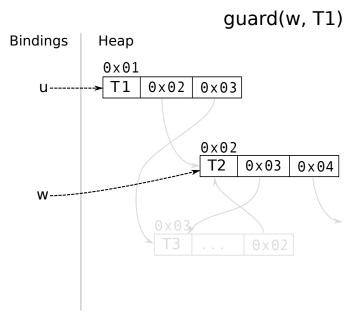
Operations: Guard



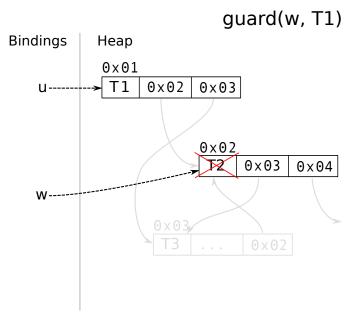
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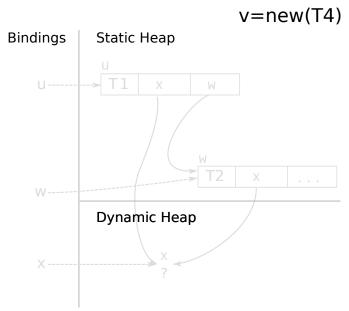
Optimization by Online Partial Evaluation

- Trace is optimized using online partial evaluation
- part of the runtime heap is modelled in the static heap
- static heap contains objects that are allocated within the trace
- (as opposed to before the trace is executed)

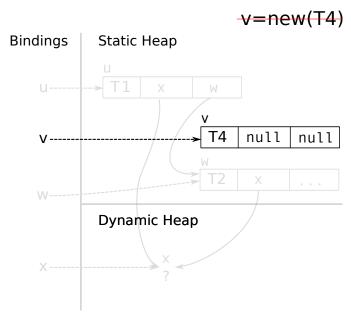
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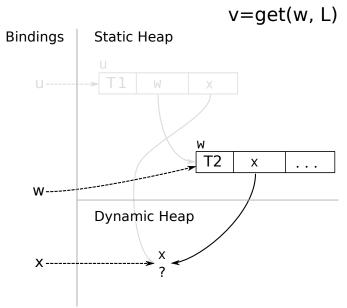
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- part of the runtime heap is modelled in the static heap
- static heap contains objects that are allocated within the trace
- (as opposed to before the trace is executed)
- operations acting on the static heap can be executed
- follows operational semantics
- all others need to be residualized
- all fairly straightforward

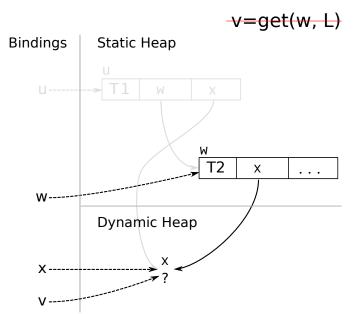
Optimizing New

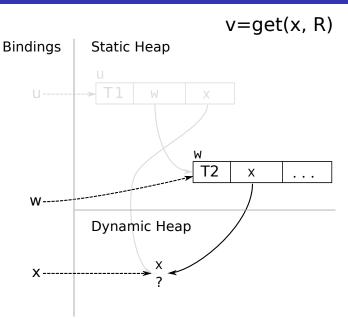


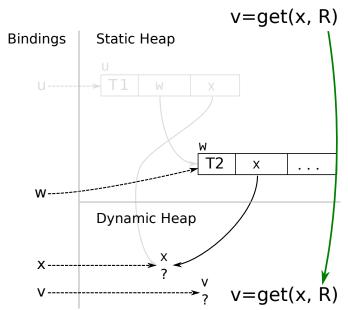
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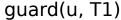


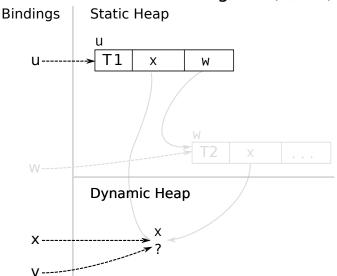


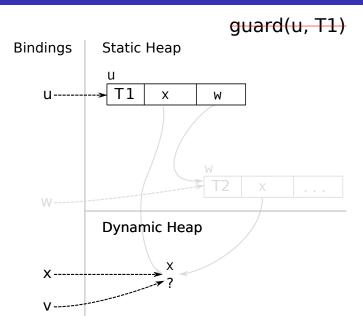


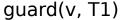


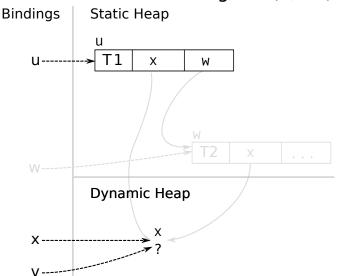


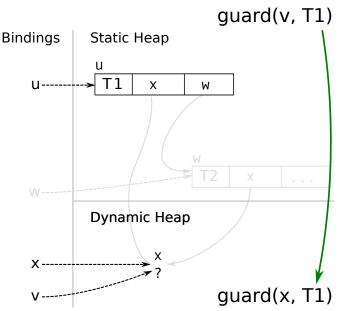


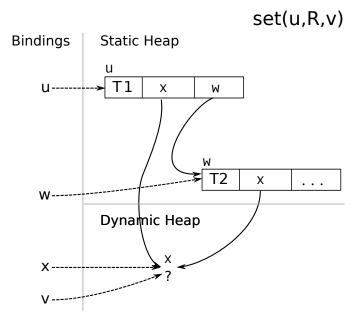


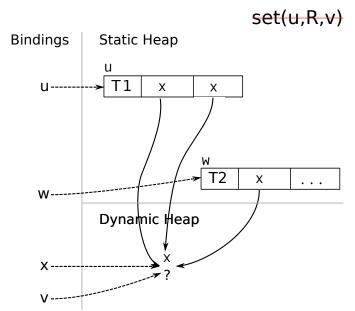












Lifting

Problem: What happens if we write a static object into a dynamic one?

Lifting

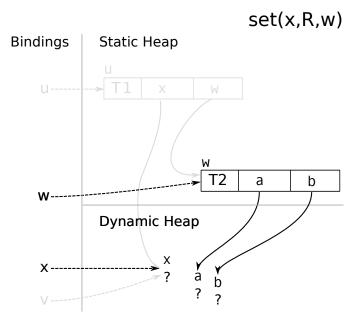
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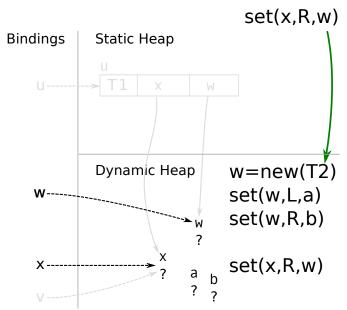
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- lose track of the static object because of possibility of aliasing
- need to lift the static object
- lifting produces operations that recreate the static object
- needs to be careful due to recursive structures





Properties of the Optimization

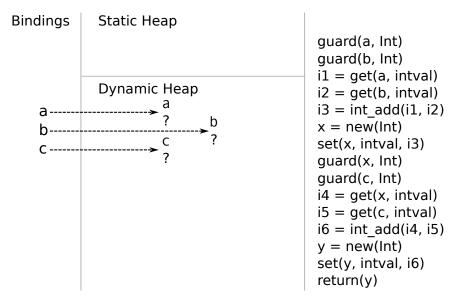
- output trace is never longer than input trace
- runtime linear in the length of the trace

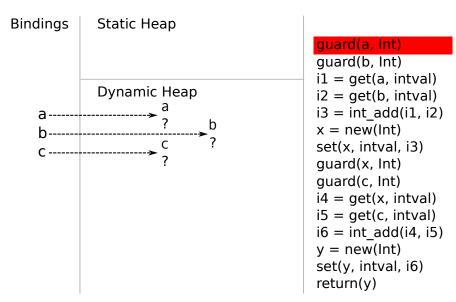
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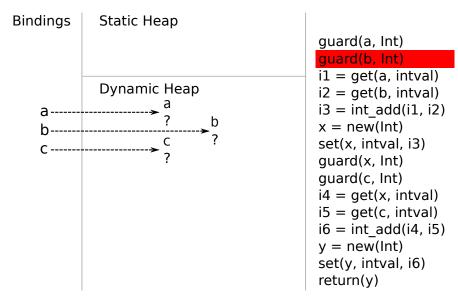
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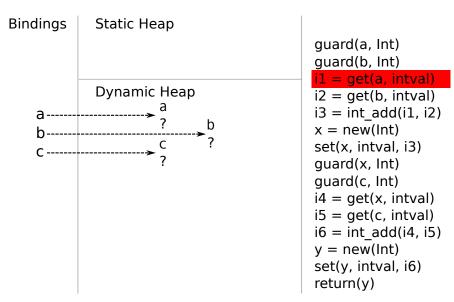
Implementation

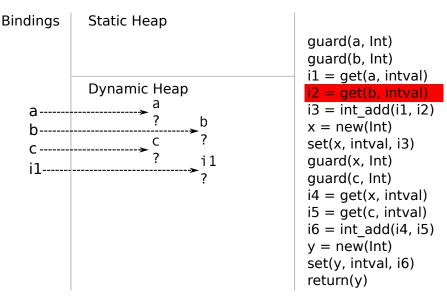
- about 400 lines of code
- some added complexity over presentation
- objects with arbitrary numbers of fields
- array support

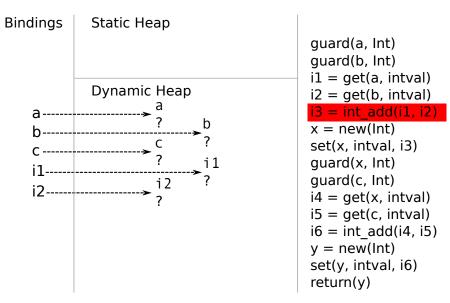


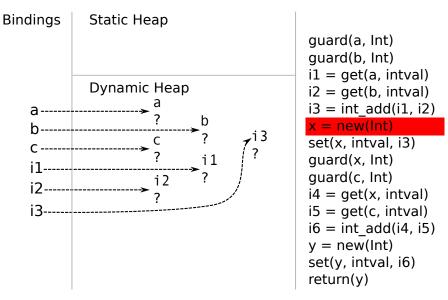


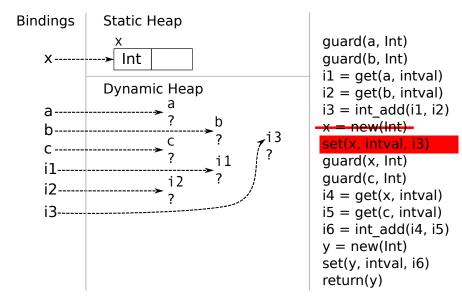


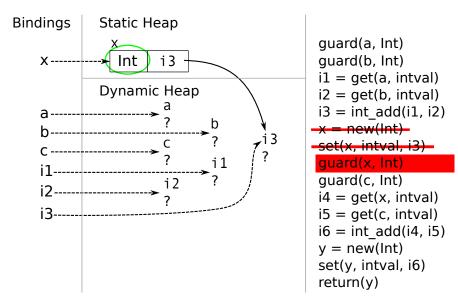


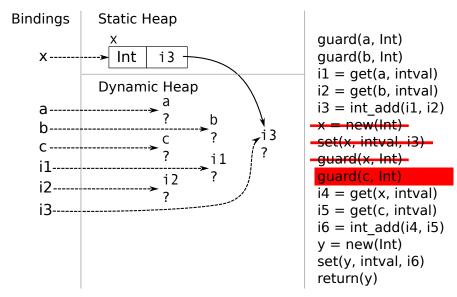


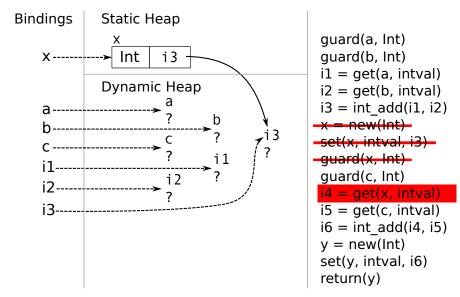


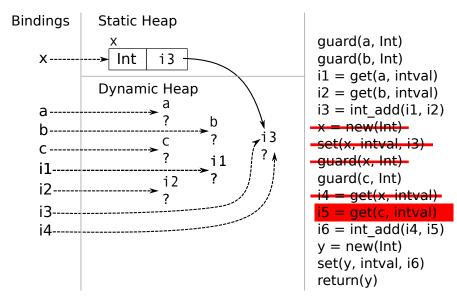


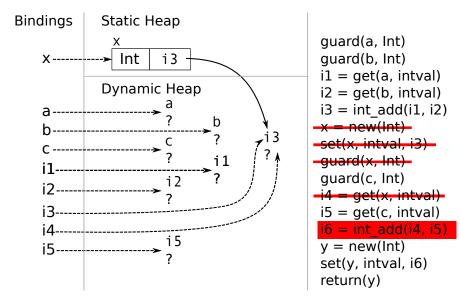


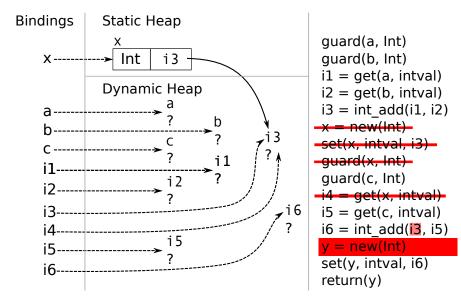


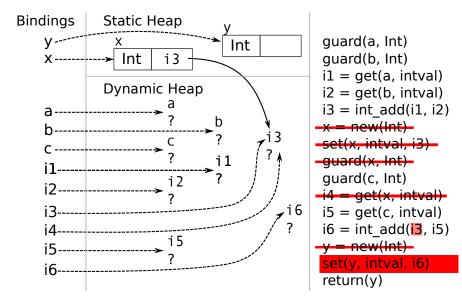


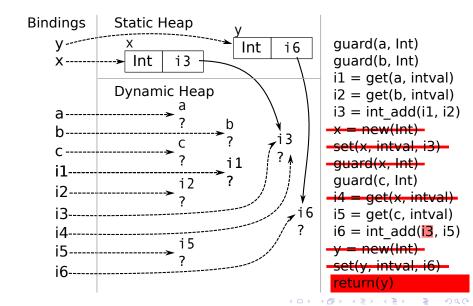


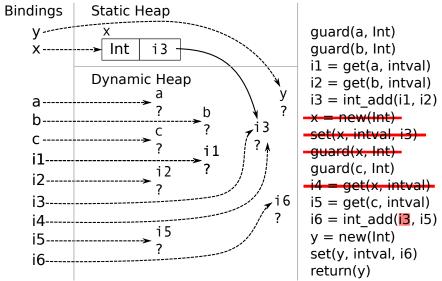












Benchmark Results

- to evaluate the optimization we used PyPy's Python interpreter with real-world programs
 - interpreter is about 30'000 lines of code

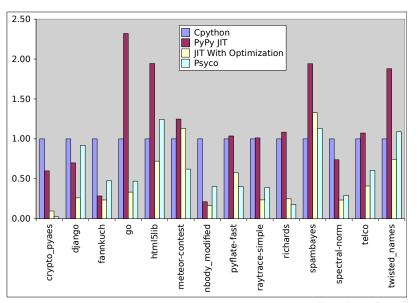
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 - 70% of all new operations
 - 14% of all get/set operations
 - 93% of all guard operations

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- optimization can remove
 - 70% of all new operations
 - 14% of all get/set operations
 - 93% of all guard operations
- Timings improve by a factor between 1.1 and 6.95
- outperforming standard Python on all benchmarks but two

Benchmark



Conclusion

- We propose a very simple partial-evaluation-based optimization for tracing JITs of dynamic languages that:
 - can remove a lot of allocations and type checks in practical programs.
 - is efficient and effective.
 - has no control issues because all control decisions are made by the tracing JIT.

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 - can remove a lot of allocations and type checks in practical programs.
 - is efficient and effective.
 - has no control issues because all control decisions are made by the tracing JIT.
- We claim that this is a general strategy to get rid of control problems by simply observing the runtime behaviour of the program.

Backup Slides

What About Correctness?

- We haven't proven correctness yet
- should not be too hard
- lifting needs to be carefully handled

Comparison to Escape Analysis

- Effect very similar to escape analysis
- Escape analysis needs a complex upfront analysis
- our optimization automatically has a lot of context, due to the inlining tracing does
- our optimization can optimize operations on objects even if they escape later

Comparison to "Dynamic Typing"/Boxing Analysis

- those optimizations work ahead of time
- don't work for many dynamic languages, where the source simply does not contain enough information

Python Example:

```
def sum(container, initial):
    result = initial
    for element in container:
        result = result + element
    return result
```