Type-Based Amortized Resource Analysis with Integers and Arrays

Jan Hoffmann and Zhong Shao, Yale University

Performance Bugs are Common and Expensive

Performance Bugs are Common and Expensive



We're working to resolve the issue as soon as possible. Please try again later.

Please include the reference ID below if you wish to contact us at 1-800-318-2596 Error from: https%3A//www.healthcare.gov/marketplace/global/en_US/registration% Reference ID: 0.cdc7c117.1380633115.2739dce8

HealthCare.gov debacle has been mainly caused by performance issues.

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The System is down at the moment.

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ICE 3 Velaro D delivery delayed by one year because of software performance issues in 2013.



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Toyota's killer firmware: Bad design and its consequences

On Thursday October 24, 2013, an Oklahoma court **ruled against Toyota** in a case of unintended acceleration that lead to the death of one the occupants. Central to the trial was the Engine Control Module's (ECM) firmware.

Stack overflow. Toyota claimed only 41% of the allocated stack space was being used. Barr's investigation showed that 94% was closer to the truth. On top of that, stack-killing, **MISRA-C** rule-violating recursion was found in the code, and the CPU doesn't incorporate memory protection to guard against stack overflow.

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Although Toyota had performed a stack analysis, Barr concluded the automaker had completely botched it. Toyota missed some of the calls made via pointer, missed stack usage by library and assembly functions (about 350 in total), and missed RTOS use during task switching. They also failed to perform run-time stack monitoring.



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Michael Dunn -October 28, 2013 **109 Comments** Share 277 8+1 932 Tweet 724 FLike 3.8k Expert witness found: "Toyota's electronic throttle control system (ETCS) source code is of unreasonable quality."

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Stack overflow was possible because stackbound analysis was faulty.

Power Consumption is Increasingly Important



One of the major cost factors in data centers.



Determines battery life in mobile devices and robots.

This Work: Static Resource Analysis

Given: A program P

Question: What is the worst-case resource consumption of P as a function of the size of its inputs?



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Clock cycles, heap space, power, ...

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Static Resource Analysis		
	ESOP'10	
	APLAS'10	
	POPL'11	
	PhD Thesis	
	CAV'12	
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Not only asymptotic bounds but concrete constant factors.

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5. Show the practicality of the system with an implementation and experiments

Polynomial Amortized Resource Analysis

- Automatic type-based analysis: No annotations required
- Naturally compositional: function types are resource specifications
- Generic in the resource: heap space, clock cycles, energy usage ...
- Precise bounds expressed by multivariate resource polynomials
- Efficient type inference based on linear programming





Bird's Eye View

Type-Based Resource Analysis





















Type-Based Resource Analysis



Type-Based Resource Analysis

Can we transfer the ideas of automatic amortized analysis to C-like programs?

Why Automatic Amortized Analysis for C Code?

- Today's embedded and real-time systems are written in C code
- There are many great techniques for deriving resource bounds on imperative code [Gulwani et al., Albert et al., Brockschmidt et al.]

But: current techniques are not compositional

Why looking a functional programs in the first place?

- Might be used more often in the future
- Clean setting to study and understand the problem (compare: type systems, type inference, higher-order functions, ...)

[PLDI'14]: End-to-End Verification of Stack-Space Bounds for C Programs

- Uses CompCert and a program logic that is based on amortized analysis
- Verified in Coq
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Promising First Results: Stack Bounds

[PLDI'14]: End-to-End Verification of Stack-Space Bounds for C Programs

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Applied to

Automation only for programs without recursion!





	Functional	Imperative
Data structures	Inductive data types	Arrays
Iteration	Recursion	Loops
Control Flow	Pattern matching	Integers

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Upcoming paper.

The General Idea of Amortized Analysis

- Assign potential functions to data structures
 - States are mapped to non-negative numbers
- Potential pays the resource consumption and the potential at the following program point
- Initial potential is an upper bound

 $\Phi(before) \ge \Phi(after) + cost$ $\checkmark telescoping \checkmark$ $\Phi(initial \ state) \ge \sum cost$

 $\Phi(state) \geq 0$

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- Fix a format of potential functions
- Develop type rules that manipulate potential functions

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• / >

Potential is given by type context.

Programs with Unsigned Integers (nat)

Data types:
$$(nat * nat, (q_{(i,j)})_{i,j \in \mathbb{N}})$$

Potential functions: $\Phi((n, m), (q_{(i,j)})) = \sum_{i,j \in \mathbb{N}} q_{(i,j)} {n \choose i} {m \choose j}$

Function types:
$$(A, Q) \rightarrow (B, Q')$$

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mult: (nat,nat) -> nat

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add(n,m) = match n with 0 -> m	<pre>mult(n,m) = match n with 0 -> 0</pre>
n+1 -> 1+add(n,m);	<pre>I n+1 -> add(m,mult(n,m));</pre>

Number of evaluation steps of mult in the worst case: 8nm + 12n + 3

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Number of evaluation steps of mult in the worst case: 8nm + 12n + 3

Possible typing of mult: $(nat * nat, (q_{(i,j)})_{i,j \in \mathbb{N}}) \rightarrow (nat, (p_i)_{i \in \mathbb{N}})$

where

$$q_{(0,0)} = 5$$

 $q_{(1,0)} = 12$
 $q_{(1,1)} = 8$
 $q_{(i,j)} = 0$ otherwise

2

$$p_i = 0$$
 for all i

add : $(nat, nat) \rightarrow nat$ mult: (nat,nat) -> nat add(n,m) =mult(n,m) =match n with | 0 -> mmatch n with | 0 -> 0 $| n+1 -> 1+a \Phi((n, m), (q_{(i,j)})) =$ | n+1 -> add(m, mult(n, m));Number of evaluation $\sum_{i,j\in\mathbb{N}} q_{(i,j)} \binom{n}{i} \binom{m}{j}$ he worst case: 8nm + 12n + 3Possible typing of mult: $(nat * nat, (q_{(i,j)})_{i,j \in \mathbb{N}}) \rightarrow (nat, (p_i)_{i \in \mathbb{N}})$ where $q_{(0,0)} = 3$ $p_i = 0$ for all i $q_{(1,0)} = 12$ $q_{(1,1)} = 8$ $\hat{q}_{(i,i)} = 0$ otherwise

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consumed later.



How to Deal with Multiplications x*y?

Code transformation to recursive function?

- Need to prove soundness (semantic and resource usage equivalence)
- Inefficient: a large constraint set is generated for each multiplication

Better approach: directly describe how to pass potential to the result

$$\Phi((n, m), Q) \ge \Phi(n \cdot m, Q') + \operatorname{cost}(mult)$$

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$$\Phi((n, m), Q) \ge \Phi(n \cdot m, Q') + \operatorname{cost}(mult)$$

Can we express this inequality with a succinct constraint system?

New Type Rule for Multiplication

$$Q = \bigcirc (Q') + M^{\text{mult}}$$

$$\overline{x_1: \text{nat}, x_2: \text{nat}; Q \models M} x_1 * x_2 : (\text{nat}, Q') \text{(T:MULT)}$$

$$\boxdot (Q) = (q'_{(i,j)})_{(i,j) \in \mathcal{I}(\text{nat*nat})} \quad \text{if} \quad q'_{(i,j)} = \sum_k A(i,j,k) q_k$$

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$$\binom{nm}{k} = \sum_{i,j} A(i,j,k) \binom{n}{i} \binom{m}{j}$$

$$A(i,j,k) = \sum_{r,s} (-1)^{i+j+r+s} \binom{i}{r} \binom{j}{s} \binom{rs}{k} = \sum_n \frac{i!j!}{k!} S(n,i) S(n,j) s(k,n)$$

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Riordan and Stein (1972)
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Smaller Constraints Sets Enable Scaling



Number of constraints generated for one multiplication

Other Arithmetic Operations

Treatment of other arithmetic operations is described in the paper

- Operations handled: subst, add, div, mod, mult
- Similar to multiplication

Also in the paper: arrays

- Arrays are treated as non-negative numbers: Array.length() returns a natural number that can be used for iteration
- Potential of data that is stored inside arrays is not tracked

How does it scale?

Dyadic Product of two Arrays

```
dyad : (Arr(int),nat,Arr(int),nat) -> Arr(Arr(int))
```

```
dyad (a,n,b,m) =
  let outerArr = A.make(n,A.make(0,+0)) in
  let _ = fill(a,n,b,m,outerArr) in outerArr;
```

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```

Computed evaluation-step bound:

20nm + 31n + 18

where

n is the value of the second component of the input m is the value of the 4'th component of the input

Dyadic Product with Polynomials

```
matrix : (nat,nat) -> Arr(Arr(int))
matrix (n,m) =
    let size1 = n*n + 9*n + 28 in
    let size2 = m*n + 6*m in
    dyad( A.make(size1,+1),size1, A.make(size2,+1),size2 );
```

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Computed evaluation-step bound:

 $20mn^3 + 300mn^2 + 1641mn + 3366m + 32n^2 + 288n + 942$

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Many Dyadic Products with Polynomials

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Computed evaluation-step bound:

 $1.66n^{6} + 37n^{5} + 334.79n^{4} + 1485.08n^{3} + 2963.54n^{2} + 1789.92n + 3$

where

n is the value of the input
Many Dyadic Products with Polynomials

Computes a ($i^2+9i+28$) x (ij+6j) matrix for every pair (i,j) such that $1 \le j \le i \le n$.

Computed evaluation-step bound:

dyadAllM : nat -> unit

 $1.66n^{6} + 37n^{5} + 334.79n^{4} + 1485.08n^{3} + 2963.54n^{2} + 1789.92n + 3$

where

n is the value of the input



Experimental Evaluation

	Computed Bound	Actual Behavior	Run Time	#Constr.
Dijkstra's Shortest Path	79.5n	O(n	0.1 s	2178
Fast GCD	12m + 7	O(log m)	0.1 s	105
Pascal's Triangle	19n	O(n	0.4 s	998
In-Place Quick Sort	12.25x	O(x	0.7 s	2080
Matrix Multiplication (for a list of matrices)	18nuyx + 31nuy + 38nu + 38n + 3	O(nuyx)	5.6 s	184270
Block Sort	12.25n	O(n	0.4 s	27795
DyadAllM	1.6n 2963.54n	O(n	3.9 s	130236
Matrix-Mult, Flatten, and Sort	12.25u + 19m + 66	O(u	5.9 s	167603

Evaluation-Step Bounds



Conclusion

Directly encoding (non-linear) arithmetic operations in amortized resource analysis lets us track size changes of unsigned integers precisely and efficiently.

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Directly encoding (non-linear) arithmetic operations in amortized resource analysis lets us track size changes of unsigned integers precisely and efficiently.

Ongoing Research: Application of the amortized analysis to C programs

- Bounds are non-negative linear combin. of sizes of intervals [[x,y]]
- Great preliminary results for linear bounds
- Beats already abstract interpretation-based techniques
- Extension to polynomial bounds using the presented techniques