Finding heap-bounds for hardware synthesis

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Coding hardware in advanced languages

- Use of advanced languages simplifies development process
- Advanced data structures are easy to use (lists, tree etc.)
- Advanced languages dynamically allocate memory

Problem with dynamic allocation

• Hardware has limited amount of memory

 Unbounded heap usage prevents compilation into hardware

Generic heap-bound

- In parameterized design, program has two kinds of input
 - Generic inputs
 - Input signals
- Generic heap-bound is a function over generic inputs that bounds heap usage

Generic inputs are set to a constant during synthesis



If generic heap-bound exists then heap is bounded by a constant during synthesis

Can we infer generic heap bound at compile time?

Outline

An example

• Our solution

• Experimental results & conclusion

Example: priority queue



- Reads infinite series of integers at input channel i
- Sort the inputs in batches of **n**
- Push out sorted batch at output channel o

Example: priority queue

Linked list **b** is initialized to be empty



Example: priority queue



Outline

An example

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Inferring generic heap-bound & compiling

- The following 3 steps can do the job
 - 1. Track heap usage at each possible run of program
 - 2. Estimate maximum heap usage over all runs
 - 3. Translate to non-dynamic allocating heap program

We supply the following solution for above steps

- 1. Numerical heap abstraction (shape analysis)
- 2. Numerical bound analysis (invariant generation)
- 3. Array based heap management





Numerical heap abstraction

Input program is translated into an abstract numerical program using shape analysis



- Each data structure is replaced by a set of integers
- Actions on data structures are replaced by actions on the integers
- A new variable is introduced to represent heap usage

Example: numerical heap abstraction



- shape analysis recognizes **c** as a pointer to a linked list
- An integer k_c is introduced to represent length of the linked list c
- An integer h is introduced to represent the amount of heap used

Abstract numerical program

- Abstract numerical program consists of
 - variables
 - control locations
 - transition relations between locations
 - generic inputs
 - a variable to represent heap usage

Abstract numerical program: priority queue



 \rightarrow k_b is a new variable which represents length of linked list **b**



Numerical bound analysis

 For each location p, we find a heap bound Bnd_p such that h ≤ Bnd_p(generic inputs)



Heap-bound from Invariant

- Invariant is an assertion that is true at all reachable states
- Invariant may imply a heap-bound that bounds heap usage



 For each location p, find an invariant Inv_p such that for some heap-bound Bnd_p

We extend constraint solving method for invariants

Heap-bounds via constraint solving

- A template is substituted for each invariant Inv_p and heap-bound Bnd_p
 - Template = parameterized assertion over program variables
- Build constraints using numerical program and these templates
- Solve the constraints and get the heap-bounds

Template for invariant

- Template for invariant = parameterized assertion over program variables
- Example: template for invariant

a+a_n*n+a_h*h+a_k*k+a_c*k_b+a_c*k_c≤0

- **a**, **a**_n, **a**_h, **a**_k, **a**_b and **a**_c are parameters
- **n**, **h**, **k**, **k**_b, and **k**_c are program variables

A template specifies a space of assertions

• We search for an invariant in this space

Template for heap-bounds

- Template for heap-bound = parameterized function over generic inputs
- Example: template for heap-bound

b_n*n+b

- b_n and b are parameters
- **n** is generic input

Template Maps

- **Inv** = **map** from location to templates for invariants
- **Bnd** = **map** from location to templates for heap-bounds



Constraints

Build constraints that encode two conditions:

- 1. Inductive argument
 - If program state is in invariant and program runs then state remains in invariant
- 2. Invariant implies heap-bound that bounds heap usage



Example:

- For transition location 7 to 13
 Inv₇ ∧ trans(7,13) → Inv'₁₃
- For bound at location 7
 Inv₇ → h ≤ Bnd₇

Solving and getting heap-bound

- The built constraints are solved over the parameters
- Placing the solution of parameters in templates produces the heap-bounds



Example:

- Template for bound at location 7
 Bnd₇: b_n*n+b
- Solution of parameters, b_n=1 and b=0
 Bnd₇: 1*n + 0



Array based heap management

- Following are introduced in the input program
 - an array h of size the heap-bound and initialize it: $\forall i. h[i]=i+1$
 - a variable m and initialize it with 0
- m will act as a head of linked list of available cells
- Example translation
 - $-x = alloc(); \rightarrow x=m; m=h[m];$
 - free(x); \rightarrow m=h[x]; m=x;



Outline

• An example

Our solution

Experimental results & conclusion

Implementation

We have developed a tool-chain from C to gates

- Numerical heap abstraction (shape analysis)
 THOR
- 2. Numerical bound analysis (invariant generation)
 - ARMC and InvGen
- 3. Array based heap management

Experimental results

Computed bounds

Program	Bound	C LOC	VHDL LOC	Bound			
				synthesis time			
merge	$8 * n_1 + 8 * n_2$	80	1927	600m			
prio	8 * n	56	1475	4s			
packet	12 * n + 8	95	2430	6s			
bst_dict	$24 * n_1$	142	2703	80s			

Synthesis and implementation results

Program (ALUTs	Registers	Block Mem	Blocks	Speed	
merge	5,157	4,694	8,192	2	90MHz	
prio	5,859	4,598	4,096	1	83MHz	
packet	9,413	9,158	8,192	2	76MHz	
bst_dict	5,786	5,660	8,192	2	125MHz	

Conclusion

- A new method to compute heap-bounds using
 - shape analysis
 - invariant generation (constraint solving)
- An attempt to bring the following together
 - agility of software development and
 - speed of raw gates

