Amortised memory analysis using the depth of data structures

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Overview

Interested in efficient memory bounds inference.

- ► Hofmann-Jost [POPL'03] amortised analysis gives **linear heap** space bounds in **total size** of the input.
- Such bounds are a poor fit for stack space.
 Especially for functional programming, tree structured data.
- ▶ Want maxima and bounds in terms of depth.

Keep close to efficient Linear Programming inference.

Principles of Hofmann-Jost analyses

- ▶ The type system *certifies* bounds;
- annotations describe bounding function in terms of input sizes:

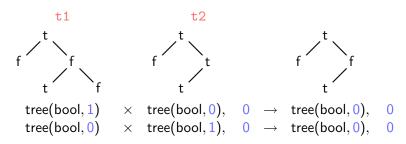
```
x : \text{tree(bool, 2)}, y : \text{tree(bool, 3)}, 5 \vdash e : \text{tree(bool, 2)}, 1
     2 \times |x| + 3 \times |y| +5
                                                   \rightarrow 2 × |result| + 1
```

Side conditions in rules guarantee the soundness of the bounding functions.

- Inference: construct 'plain' typing
 - collect side conditions together
 - solve the resulting LP.

Heap memory example

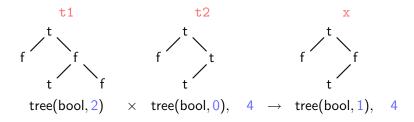
The andtrees function computes the pointwise 'and' of two boolean trees (up to the largest common subtree):



- means andtrees t1 t2 uses no more than |t1| units of space.
- ► The typings (and bounds) are not unique. |t2| is also sufficient.

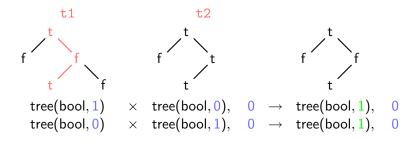
Heap memory example

The andtrees function computes the pointwise 'and' of two boolean trees (up to the largest common subtree):



let x =andtrees t1 t2in e_2

- Signatures also 'translate' requirements:
- ▶ If e_2 requires |x| + 4 units, then $2 \times |y| + 4$ is sufficent for both allocation and |x| + 4 later.



- means andtrees t1 t2 uses at most |t1| (or |t2|) units of stack space.
- Stack space is reusable.
- ▶ But now we want to use the depth to get a better bound (i.e., |t1|_d).

Example for stack space 6/20

Developing an analysis with maximums

Previously we just added all the contributions from the context:

I:tree(bool, k), r:tree(bool, k), v:bool,
$$n \vdash e : ...$$

 $|I| \times k + |r| \times k + 0 + n$

Now we introduce a second context former to denote 'max' (;):

```
(/:tree(bool, k); r:tree(bool, k);v:bool), n \vdash e:...
\max\{ |I|_d \times k , |r|_d \times k , 0 \} + n
```

- Note that contexts are now trees.
- Treat tree types as 'folded up' version of above context.
- ▶ So t:tree(bool, k) denotes $|t|_d \times k$.

Inspired by O'Hearn's Bunched Typing.

Unfolding trees in the context

 $\Gamma()$ is a context with a 'hole'.

$$\frac{\Gamma(\cdot) \vdash e_1 : T, k'}{\Gamma((l : \mathsf{tree}(T, k); r : \mathsf{tree}(T, k); v : T), k) \vdash e_2 : T, k'}{\Gamma(t : \mathsf{tree}(T, k)) \vdash \mathsf{match} \ t \ \mathsf{with} \ \ | \mathsf{leaf} \to e_1 \\ | \ \mathsf{node}(l, v, r) \to e_2 : T, k' \\ (\mathsf{MATCH})$$

$$\to t \\ t : \mathsf{tree}(\mathsf{bool}, k)$$

Unfolding trees in the context

 $\Gamma()$ is a context with a 'hole'.

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$$\Gamma((I : \mathsf{tree}(T, k); r : \mathsf{tree}(T, k); v : T), k) \vdash e_2 : T, k'$$

$$\Gamma(t : \mathsf{tree}(T, k)) \vdash \mathsf{match} \ t \ \mathsf{with} \quad \mathsf{leaf} \to e_1 \\ \mid \mathsf{node}(I, v, r) \to e_2 : T, k'$$

$$(\mathsf{MATCH})$$

$$t : \mathsf{tree}(\mathsf{bool}, k)$$

$$(I : \mathsf{tree}(\mathsf{bool}, k); r : \mathsf{tree}(\mathsf{bool}, k); v : \mathsf{bool}), k$$

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$$(\mathsf{MATCH})$$

$$t : \mathsf{tree}(\mathsf{bool}, k)$$

$$(I : \mathsf{tree}(\mathsf{bool}, k); r : \mathsf{tree}(\mathsf{bool}, k); v : \mathsf{bool}), k$$

(!:tree(bool, k); ((r!:tree(bool, k); rr:tree(bool, k); rv:bool), k); v:bool), k

let expressions

let
$$x = e_1$$
 in e_2

- ▶ Overall bound is $\max\{\text{bound for } e_1, \text{bound for } e_2\}$.
- ▶ But we also need to translate bound for e_2 .

Instead replace subcontext for x with that needed to produce it, Δ :

$$\frac{\Delta \vdash e_1 : T_0, n_0}{\Gamma(\Delta) \vdash \text{let } x = e_1 \text{ in } e_2 : T, k'}$$
(Let)

(Only sound for stack discipline.)

New rules

We need to be able to manipulate contexts to get the right shape. Hence new rules such as:

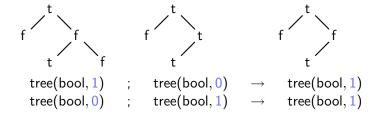
$$\frac{\Gamma(\Delta') \vdash e : \mathcal{T}, n' \qquad \Delta \cong \Delta'}{\Gamma(\Delta) \vdash e : \mathcal{T}, n'} \qquad (\equiv)$$

$$\Gamma, (\Delta; \Delta') \cong (\Gamma, \Delta); (\Gamma, \Delta')$$
 (distribution)
 $\Gamma \cong \Gamma; \Gamma$ (max-contraction)
 $\Gamma \cong q\Gamma, (1-q)\Gamma$ $q \in [0,1]$ (plus-contraction)

All preserve the bounding functions derived from the context.

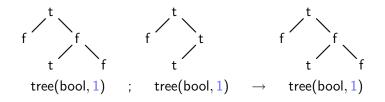
Also: weakening and a max-to-plus approximate conversion.

Function signatures are also 'structured'.



▶ means and trees t1 t2 uses at most $|t1|_d$ or $|t2|_d$ units of stack space.

Function signatures are also 'structured'.



- ▶ means and trees max t1 t2 uses at most $\max\{|\mathtt{t1}|_d, |\mathtt{t2}|_d\}$ units of stack space.
- We can now also type a version of andtrees which returns false for all the nodes which are only in one of the arguments.

Function signatures are also 'structured'.

Extra benefit from maxima

```
let maybeleft(t,b) =
  match t with leaf -> leaf | node(l,r,v) ->
  if b then l else t
```

In heap analysis need to sum requirements because of use of contraction at match. Doubles the bound unnecessarily.

- ▶ In depth type system we can use max-contraction.
- ▶ So requirement goes $|t| \Rightarrow \max\{|t|, |t|\} \Rightarrow \max\{|t|, |t|\}$.
- ▶ Context goes t : tree $\Rightarrow t$: tree; t : tree $\Rightarrow t$: tree; t : tree

Stack space inference

- Would like to take advantage of linear programming again
- ▶ But new context manipulation rules are not syntax-directed

We add an extra stage to the inference process:

source program (plain types)

ditional terms for context manipulation

bound (using linear programming)

Assume context structure given for function signatures to make problem more tractable.

Basic ideas for inference

- ▶ Work from the leaves of the expression outwards.
- At every stage, keep track of a generated context derived from subexpressions and the typing rule.

$$\frac{\Gamma \vdash e_1 \mapsto \Gamma_1 \qquad \Gamma \vdash e_2 \mapsto \Gamma_2}{\Gamma \vdash \text{if } x \text{ then } e_1 \text{ else } e_2 \mapsto \Gamma_1; \Gamma_2; x \text{:bool}}$$

Need to add context manipulation at two points:

- 1. where binding occurs, to deal with contraction, etc;
- 2. to make the generated context match the function signature.

Expand context to max-of-sums form, factor out bound variables.

The full analysis

- Have algebraic data types, not just trees.
- Can specify the form of bounds:

in terms of depth, total size, or a mixture.

Bounds w.r.t. total size useful when depth analysis fails.

Resource polymorphism (different function signatures at different points).

Implementation in Standard ML.

Conclusion and Further work

Type-based amortized analysis can provide good bounds on stack space for programs using tree-structured data.

- ▶ Nested containers don't behave that well due to the definition of depth.
 - May be possible to separate structure of containers from their contents.
- ▶ Inferring the structure of function signatures.
- ▶ Deal with construction of log-depth trees (e.g, heap sort).

Potentially the most interesting:

Find sound Let rule for heap space, get max-plus bounds for heap.