Amortised memory analysis using the *depth* of data structures

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## Outline

Hofmann-Jost Heap Memory Analysis

Example for stack space

The Stack Analysis Type System

The Stack Space Inference

Further Work

The andtrees function computes the pointwise 'and' of two boolean trees (up to the smaller tree):



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- The typings (and bounds) are not unique. |t2| is also sufficient.

The andtrees function computes the pointwise 'and' of two boolean trees (up to the smaller tree):



let x = andtrees y z in ...

- Signatures also 'translate' requirements:
- If ... requires |x| + 4 units, then 2 × |y| + 4 is sufficient for both allocation and |x| + 4 later.

### Hofmann-Jost rules

 $n \ge \text{size(bool tree node)} + \mathbf{k} + n'$ 

*l*:bool tree[*k*], *r*:bool tree[*k*], *v*:bool,  $n \vdash \text{node}(l, v, r)$ : bool tree[*k*], n' (NODE)

means that if we have

 $|I| \times \mathbf{k} + |\mathbf{r}| \times \mathbf{k} + \mathbf{n}$ 

units of free memory then we can allocate the node and end up with

 $|\text{node}(l, v, r)| \times k + n' = (1 + |l| + |r|) \times k + n'.$ 

# Hofmann-Jost inference

 $n \ge \text{size(bool tree node)} + \mathbf{k} + n'$ 

*l*:bool tree[*k*], *r*:bool tree[*k*], *v*:bool,  $n \vdash \text{node}(l, v, r)$ : bool tree[*k*], n' (NODE)

- Collect constraints from typing rules;
- Solve linear program, minimising the bound.

## Hofmann-Jost as an amortized analysis

 $n \ge \text{size(bool tree node)} + \mathbf{k} + n'$ 

*I*:bool tree[*k*], *r*:bool tree[*k*], *v*:bool,  $n \vdash \text{node}(I, v, r)$ : bool tree[*k*], *n'* (NODE)

The type annotations define potential functions

$$\Upsilon_{\Gamma}(l,r) = |l| \times \frac{k}{k} + |r| \times \frac{k}{k} + \frac{n}{k}$$

for the context, and for the result:

$$\Upsilon_R(R)=|R|\times \mathbf{k}+\mathbf{n}'.$$

Constraint ensures that the allocation is accounted for by a drop in potential. (See *Physicist's view* in Tarjan 1985)

Hofmann-Jost Heap Memory Analysis



- means andtrees t1 t2 uses at most |t1| (or |t2|) units of stack space.
- Stack space is reusable.



- 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|<sub>d</sub>).

Developing an analysis with maximums

Previously we just added all the 'potential' from the context:

```
l:bool tree [k], r:bool tree [k], v:bool, n \vdash \dots
|l| \times k + |r| \times k + 0 + n
```

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

(*I*:bool tree [k]; r:bool tree [k];v:bool),  $n \vdash \dots$ max{ $|I|_d \times k$ ,  $|r|_d \times k$ , 0}+n

- Treat tree types as 'folded up' version of above context.
- So t: bool tree [k] has potential of  $|t|_d \times k$ .
- Note that contexts are now trees.

Inspired by O'Hearn's Bunched Typing.

### **Bunched Contexts**

 $\Gamma := \cdot \mid x \colon T \mid \Gamma, \Gamma \mid \Gamma; \Gamma \mid k$ 

$$\begin{split} \Upsilon_{\Gamma}(S,\cdot) &= 0\\ \Upsilon_{\Gamma}(S,x:T) &= \Upsilon_{T}(S(x),T)\\ \Upsilon_{\Gamma}(S,(\Gamma_{1},\Gamma_{2})) &= \Upsilon_{\Gamma}(S,\Gamma_{1}) + \Upsilon_{\Gamma}(S,\Gamma_{2})\\ \Upsilon_{\Gamma}(S,(\Gamma_{1};\Gamma_{2})) &= \max\{\Upsilon_{\Gamma}(S,\Gamma_{1}),\Upsilon_{\Gamma}(S,\Gamma_{2})\}\\ \Upsilon_{\Gamma}(S,k) &= k \end{split}$$

So  $(x:T_1, k)$ ;  $y:T_2$  has potential

$$\max\{|x|_{\mathrm{d}}+k,|y|_{\mathrm{d}}\}.$$

THE STACK ANALYSIS TYPE SYSTEM

# Unfolding trees in the context

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

$$\begin{array}{c} \Gamma(\cdot) \vdash e_1 \colon T, k' \\ \hline \Gamma((l: \text{bool tree}[k]; r: \text{bool tree}[k]; v: \text{bool}), k) \vdash e_2 \colon T, k' \\ \hline \Gamma(t: \text{bool tree}[k]) \vdash \text{match } t \text{ with } \begin{array}{c} \text{leaf} \to e_1 \\ | \text{ node}(l, v, r) \to e_2 \colon T, k' \\ \end{array} \\ \end{array}$$



t:bool tree[k]

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(I:bool tree[k]; r:bool tree[k]; v:bool), k

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(I:bool tree[k]; ((rl:bool tree[k]; rr:bool tree[k]; rv:bool), k); v:bool), k

#### Folding trees in the context

 $k' \ge k + k''$ 

(*I*:bool tree[*k*]; *r*:bool tree[*k*]; *v*:bool),  $k' \vdash \text{node}(I, v, r)$ :bool tree[*k*], k''(NODE)



(I:bool tree[k]; ((rl:bool tree[k]; rr:bool tree[k]; rv:bool), k); v:bool), k

The Stack Analysis Type System

#### New rules

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

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

$$\begin{array}{ll} \Gamma, (\Delta; \Delta') \cong (\Gamma, \Delta); (\Gamma, \Delta') & (\text{distribution}) \\ \Gamma \cong \Gamma; \Gamma & (\text{max-contraction}) \\ \Gamma \cong q\Gamma, (1-q)\Gamma & q \in [0, 1] & (\text{plus-contraction}) \end{array}$$

All preserve the 'potential' (i.e. give the same potential function).

Function signatures are also 'structured'.



means andtrees t1 t2 uses at most |t1|<sub>d</sub> or |t2|<sub>d</sub> units of stack space.

Function signatures are also 'structured'.



- means andtreesmax t1 t2 uses at most max{|t1|<sub>d</sub>, |t2|<sub>d</sub>} units of stack space.
- We can now also type a version of andtrees which keeps all the nodes which are only in one of the arguments.

Function signatures are also 'structured'.



# Extra benefit from maxima

```
let maybetail(1,b) =
  match l with cons(h,t)' ->
    if b then t else l
```

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  $|I| \Rightarrow \max\{|I|, |I|\} \Rightarrow \max\{|t|, |I|\}$ .
- Context goes  $I : \text{list} \Rightarrow I : \text{list}; I : \text{list} \Rightarrow t : \text{list}; I : \text{list}$

#### What about let?

$$\frac{\Gamma_1 \vdash e_1 \colon T_1 \qquad \Gamma_2, x \colon T_1 \vdash e_2 \colon T_2}{\Gamma_1, \Gamma_2 \vdash \text{let } x = e_1 \text{ in } e_2 \colon T_2} \text{ (BORING LET)}$$

We don't necessary want to sum requirements, and we do want to preserve tree structure.

- Really want to use a small subcontext Γ<sub>1</sub> inside Γ<sub>2</sub>(Γ<sub>1</sub>), replace with x : T<sub>1</sub>.
- This kind of thing appears in Bunched Typing too.
- Prepared to forget about bounding heap space.

#### Local context replacement

For stack space we always get back the memory that we put in. So we have the same amount of 'free memory' at  $e_2$  as the start of the let expression.

Thus we can change the form of potential function so long the values it produces cannot increase.

$$\frac{\Gamma_1 \vdash e_1 : T_1, n_1 \qquad \Gamma(x : T_1, n_1) \vdash e_2 : T, n'}{\Gamma(\Gamma_1) \vdash \text{let } x = e_1 \text{ in } e_2 : T, n'} \qquad \text{(LET)}$$

Relies on stack discipline.

Heap allocation is harder: If we have Γ<sub>1</sub>; Γ<sub>2</sub> and allocate in e<sub>1</sub>, then we might not have enough left for Γ<sub>2</sub>.

# Local replacement example

let x = node(l, r, v) in  $e_2$ 

Say  $e_2$  requires max{ $|x|_d + |y|_d, 5$ }.

(x:bool tree[1]; y:bool tree[1]),  $5 \vdash e_2$ : T, k

The typing of  $e_1$  can 'translate' the x part of the bound:

(*I*:bool tree[1]; *r*:bool tree[1]; *v*:bool),  $1 \vdash \text{node}(I, r, v)$ :bool tree[1], 0

So replace x in  $e_2$ 's context:

((*I*:bool tree[1]; *r*:bool tree[1]; *v*:bool), 1, *y*:bool tree[1]),  $5 \vdash e_2 : T, k$ 

 $\max\{\max\{|I|_{\rm d},|r|_{\rm d}\}+1+|y|_{\rm d},5\}$ 

But for heap part of the 5 units may be used up in the allocation.

### 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)
↓
additional 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.

# Expanding contexts

We can simplify the problems by distributing over ';' as far as possible to get a maximum-of-sums context:

$$(((((a, b); c), d); e), f \cong (((a, b); c), d, f); (e, f) \\\cong (a, b, d, f); (c, d, f); (e, f)$$

 $\max\{\max\{a+b,c\}+d,e\}+f = \max\{a+b+d+f,c+d+f,e+f\}$ 

(Potentially exponential, but contexts are small. May be possible to reduce amount of expansion.)

# Binding

We can pick out the plus-bunches of the expanded context involving the bound variable and factor them out:

$$\begin{split} & \Gamma_1; \ldots; \Gamma_n; (x:T, \Gamma_{n+1}); \ldots; (x:T, \Gamma_m) \vdash e:T, n' \\ & \downarrow \\ & \Gamma_1; \ldots; \Gamma_n; (x:T, (\Gamma_{n+1}; \ldots; \Gamma_m)) \vdash \mathsf{distribute}(n+1, e):T, n' \end{split}$$

- Add contraction rules as necessary.
- The multivariable case is similar, but may require more approximation.

- Expand both ends;
- pick out bunch(es) in the expanded signature containing all the variables of each generated bunch;
- weaken away any extras;

duplicating the signature's bunches as necessary.

Generated context (a; b), (c; d)

function signature 
$$a, b, ((c, e); (d, e))$$

- Expand both ends;
- pick out bunch(es) in the expanded signature containing all the variables of each generated bunch;
- weaken away any extras;

duplicating the signature's bunches as necessary.

$$\begin{array}{rcl} \text{Generated context} & (a; b), (c; d) \\ & (\text{expanded}) & \cong & (a, c); (a, d); (b, c); (b, d) \end{array}$$

 $\begin{array}{ll} (\text{expanded}) & (a, b, c, e); (a, b, d, e) \\ \text{function signature} & \cong & a, b, ((c, e); (d, e)) \end{array}$ 

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$$\begin{array}{rcl} \text{Generated context} & (a; b), (c; d) \\ & (expanded) &\cong & (a, c); (a, d); (b, c); (b, d) \\ & (rearrange) &\cong & (a, b, c, e); (a, b, d, e); (a, b, c, e); (a, b, d, e) \\ & (max-contract) &\cong & (a, b, c, e); (a, b, c, e); (a, b, d, e); (a, b, d, e) \\ & (expanded) &\cong & (a, b, c, e); (a, b, d, e) \\ & function signature &\cong & a, b, ((c, e); (d, e)) \end{array}$$

Weakening bridges the gap between lines 2 and 3.

The Stack Space Inference

- Expand both ends;
- pick out bunch(es) in the expanded signature containing all the variables of each generated bunch;
- weaken away any extras;

duplicating the signature's bunches as necessary.

Nasty case:

Generated *a*, *b* 

Signature a; b

Use rule corresponding to  $\max\{x, y\} \ge \frac{1}{2}(x + y)$ . Sound, but sometimes imprecise.

The Stack Space Inference

# Fixed amounts of potential in contexts

Fixed amounts may appear anywhere in the context, but are not explicitly introduced by binding.

When expanding contexts to the maximum-of-plus form, we can ensure that every plus-bunch has exactly one fixed amount

$$(x_{11}: T_{11}, \ldots, k_1); \ldots; (x_{n1}: T_{n1}, \ldots, k_n)$$

When partitioning the expanded context for a binding, add context manipulation terms so that the resulting fixed amounts can come from the binding *or* the subcontexts:

$$\Gamma_{1}; \ldots; \Gamma_{n}; (x: T, \Gamma_{n+1}, k_{n+1}); \ldots; (x: T, \Gamma_{m}, k_{m})$$

$$\downarrow$$

$$\Gamma_{1}; \ldots; \Gamma_{n}; (x: T, k, ((\Gamma_{n+1}, k'_{n+1}); \ldots; (\Gamma_{m}, k'_{m})))$$

The typing rules require that  $k + k'_i \ge k_i$ .

# Finishing inference

#### source program (plain types) ↓ additional terms for context manipulation – variables ↓ additional terms for context manipulation – fixed potential ↓ bound (using linear programming)

- Now collect constraints and solve LP as before.
- (Rough) SML implementation.

# The full analysis

- Have algebraic data types, not just trees.
- Can specify the calculation of potential: depth, total size, mixture

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

 Resource polymorphism (different function signatures at different points).

#### Further work

- Nested types don't behave that well. Have done some work on separating contents and structure
- Inferring the structure of function signatures.
- Reduce complexity of inference.
- Try heap space version.