# Higher Order Escape Analysis: Optimizing Stack Allocation in Functional Program Implementations<sup>1</sup>

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#### Abstract.

In this paper, we present a method for optimizing the allocation of closures in memory. This method is based on *escape analysis*, an application of abstraction interpretation to higher order functional languages. Escape analysis determines, at compile time, if any arguments to a function have a greater lifetime than the function call itself. Such arguments, especially if they are closures, must be allocated in the heap rather than in the stack. In most implementations, however, stack allocation of closures is preferable due to the lower cost of allocation and reclamation. Therefore, we use escape analysis to determine when arguments can be stack allocated safely.

In the past, first order escape analysis has been used in optimizing LISP compilers, and has been described in various data-flow analysis frameworks for a language with complex types. The analysis described here, being higher order, provides more accurate escape information, although for a very simple higher order functional language.

#### 1. Introduction

Higher order functions are an important part of functional languages. They have generally been seen, however, as having a high implementation overhead. Two reasons for this view are that 1) they force an implementation to use a heap to store closures, and 2) programs using higher order functions are particularly difficult to analyze for optimization purposes.

The need for heap allocation arises when parameters and locally defined objects within a function outlive a call to that function. For example, in the following program fragment

```
let f x y z = x + y + z

g a b = f b a

in g 1 2
```

a closure representing the partial application of f during the execution of g will outlive the call to g. Thus, the closure containing the parameters a and b will have to be heap allocated. In this case, the partial application of f, along with g's parameters a and b, are said to escape from the call to g.

Notice that in the following program fragment

```
let f x y z = x + y + z

h x = x 3

g a b = h (f b a)

in g 1 2
```

the closure representing the partial application of f does not escape from g. In this case, the closure and pa-

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```
c \in Con
                   constants
x \in Id
                   identifiers
e \in Exp
                   expressions, defined by
                            c | x | e_1 e_2 | e_1 \rightarrow e_2, e_3 | e_1 + e_2 | e_1 = e_2 | \lambda x. e |
                             let x_1 = e_1; \ldots; x_n = e_n in e
```

Figure 1. The syntax of nml

rameters can be allocated in g's activation record on the stack. Escape analysis is a compile time analysis that determines whether an object, such as a closure or a parameter, needs to be heap allocated.

In this paper we assume that stack allocation is less expensive than heap allocation, although we recognize that there are safety issues (such as stack overflow, etc.) that are important (see [Chase88]). There are also persuasive arguments in favor of heap allocation. In the Standard ML compiler [AM87], all closures are allocated in the heap. It appears that with a large amount of memory and a sophisticated garbage collection strategy, the overhead of garbage collection is quite small. However, on most current systems (and especially in distributed systems where garbage collection is more expensive), stack allocation of closures is preferable.

A simple escape analysis was used in the Orbit compiler for Scheme [Kranz88]. It is a first order escape analysis in which the following program could not be analyzed accurately.

```
let fab = a + b
   ghab=hba
in gf12
```

because h is an unknown function (i.e. a function bound to a formal parameter) in the body of g. Orbit's escape analysis assumes that any argument to an unknown function will escape from that function. Thus, it will assume that both a and b escape from the call to h inside of g, and therefore escape from g. In fact, neither a nor b escape from g.

Another analysis, called lifetime analysis [RM88], was developed to compute, if possible, the relative lifetimes of dynamically allocated objects in a first order language with structures and recursive types (such as trees). Escape analysis is a particular instance of lifetime analysis in which the lifetime of a function's activation record is compared to the objects defined inside the function. Other analyses for optimizing storage allocation were proposed in [JM76], [MJ81], [Schwartz75], [Barth77], and [Chase87].

In the following sections, we present an analysis that, using abstract interpretation ([CC77],[Mycroft81]), gives escape information in the presence of higher order functions (although we do not deal with structures and recursive types). Higher order abstract interpretation has been mainly used for strictness analysis ([BHA85],[HY86]), although other higher order analyses have been developed (such as sharing analysis [Goldberg87]). Our use of abstraction interpretation differs from that of Mycroft, and is similar to that of Hudak and Young, because we form an abstraction of a nonstandard semantics, rather than the standard semantics, of our programming language.

# 2. A Simple Higher Order Functional Language

For this discussion a very simple higher order monomorphically typed strict functional language, nml (for not much of a language), will suffice. The syntax of the language is given in figure 1, although we omit the type declarations. The standard semantic domains of nml are as follows:

```
D, the standard domain of values,
Env: Id \rightarrow D, the domain of environments,
```

$$E[[c]] \quad Env = c, \text{ for each constant } c \in Con$$

$$E[[x]] \quad Env = Env[[x]], \text{ for each identifier } x \in Id$$

$$E[[e_I + e_2]] \quad Env = (E[[e_I]] \quad Env) + (E[[e_2]] \quad Env)$$

$$E[[e_I = e_2]] \quad Env = (E[[e_I]] \quad Env = E[[e_2]] \quad Env)$$

$$E[[e_1 e_2]] \quad Env = (E[[e_I]] \quad Env) \quad (E[[e_2]] \quad Env)$$

$$E[[e_1 \to e_2, e_3]] \quad Env = E[[e_I]] \quad Env \to \quad E[[e_2]] \quad Env, \quad E[[e_3]] \quad Env$$

$$E[[\lambda x.e]] \quad Env = \lambda y.E[[e]] \quad Env[y/x]$$

$$E[[et x_1 = e_I; \dots; x_n = e_n \text{ in } e]] \quad Env = E[[e]] \quad Env')/x_1, \dots, (E[[e_n]] \quad Env')/x_n]$$

$$\text{where } Env' = [(E[[e_I]] \quad Env')/x_1, \dots, (E[[e_n]] \quad Env')/x_n]$$

Figure 2. The standard semantics of nml

 $E: Exp \rightarrow Env \rightarrow D$ , the semantic function for expressions.

The standard semantic function E is defined in figure 2. For notational convenience, all syntactic objects are printed in boldface type, including all nml identifiers. Variables referring to syntactic objects are printed in boldface italic type. Semantic variables are printed in non-bold italic type.

# 3. An Exact Escape Semantics

In this section, we describe a nonstandard semantics for nml such that the result of a function call indicates whether a particular parameter escapes. The escape semantic domains are defined as follows:

$$D_e^{int} = D_e^{bool} = \dots = 2 \times \{err\}$$
, where 2 is the two element domain ordered by  $0 \le I$ ,  $D_e^{T_I \to T_2} = 2 \times (D_e^{T_I} \to D_e^{T_2})$ , for any types  $T_I$  and  $T_2$ ,  $D_e = \sum_T D_e^{T}$ ,

 $Env_e = Id \rightarrow D_e$ , the domain of escape environments.

The semantic function is

$$E_e: Exp \rightarrow Env_e \rightarrow D_e$$

and is defined in figure 3.

Under these semantics, the value of an expression is a pair whose first element is a boolean (0 or 1) that

$$E_e[\![\,c\,]\!] \ Env = <0, err> \ , \text{ for any constant } c$$

$$E_e[\![\,x\,]\!] \ Env = Env[\![\,x\,]\!]$$

$$E_e[\![\,e_1 + e_2]\!] \ Env = <0, err> \ , \text{ likewise for the other arithmetic operators}$$

$$E_e[\![\,e_1 + e_2]\!] \ Env = (E_e[\![\,e_1]\!] \ Env)_{(2)} \ (E_e[\![\,e_2]\!] \ Env)$$

$$E_e[\![\,e_1 - e_2, e_3]\!] \ Env = Oracle [\![\,e_1]\!] \ \rightarrow E_e [\![\,e_2]\!] \ Env \ , E_e [\![\,e_3]\!] \ Env$$

$$E_e[\![\,e_1 - e_2, e_3]\!] \ Env =$$

$$\text{where } v = \bigvee_{z \in F} (Env[\![\,z]\!])_{(I)}, \text{ and } F \text{ is the set of free variables in } (\lambda x.e)$$

$$E_e[\![\,e_1 - e_1; \ldots; x_n = e_n \text{ in } e]\!] \ Env = E_e[\![\,e_1]\!] \ Env')$$

$$\text{where } Env' = [(E_e[\![\,e_1]\!] \ Env')/x_1, \ldots, (E_e[\![\,e_n]\!] \ Env')/x_n]$$

$$\text{Figure 3. The exact escape semantics of nml}$$

indicates whether a particular parameter escapes, and whose second element is a function that captures the higher order behavior of the expression. *err* denotes a non-function value.

Given an  $x \in D_e$  we use the notation  $x_{(1)}$  and  $x_{(2)}$  to refer to the first and second elements of x, respectively. The domain  $D_e$  is partially ordered in the standard way:

$$\forall x,y \in D_e, x \le y \text{ iff } x_{(1)} \le y_{(1)} \text{ and } x_{(2)} \le y_{(2)}$$

For each type  $T = T_1 \rightarrow T_2$ , there is a bottom element  $\perp_T$  of  $D_e$  defined as follows:

$$\perp_T = <0, \lambda x. \perp_{T_2}>$$

For each primitive type T,  $\perp_T = < 0$ , err>.

In order to return the actual escape value of each expression, we must be able to determine which branch of a conditional would be evaluated at run-time. The only way to do this would be to embed the standard semantics within the escape semantics. For convenience, we instead resort to an oracle to choose the appropriate branch of the conditional.

Given an nml function f, its meaning under the escape semantics will be a pair  $\langle f_{(1)}f_{(2)}\rangle$ . We then use  $f_{(2)}$  to determine if a particular argument in a call to f escapes. Suppose, for example, we want to know, given the function application (f x), if x escapes. To do so, we let  $\langle x_{(1)}, x_{(2)}\rangle$  be the value of x under that escape semantics and let  $y = f_{(2)} \langle 1, x_{(2)}\rangle$ . If  $y_{(1)} = 1$  then x escapes in the standard semantics, otherwise x does not. Section 6 gives a detailed description of how the escape semantics is used.

# 4. The Abstract Escape Semantics

We now present an abstraction of the exact escape semantics that allows an approximation of the exact escape behavior to be found at compile time. The semantic domains are essentially identical to those of the exact escape semantics:

$$\begin{aligned} &D_{ae}^{int} = D_{ae}^{bool} = \dots = 2 \times \{err\}, \\ &D_{ae}^{T_I \to T_2} = 2 \times (D_{ae}^{T_I} \to D_{ae}^{T_2}), \\ &D_{ae} = \sum_{\tau} D_{ae}^{T} \end{aligned}$$

$$Env_{ae} = Id \rightarrow D_{ae}$$
.

The semantic function

$$E_{ae}: Exp \rightarrow Env_{ae} \rightarrow D_{ae}$$

is defined in figure 4. The difference between the exact and the abstract semantics lies in the handling of the conditional. Rather than referring to the standard semantics (as denoted by the oracle) the conditional is handled by taking the least upper bound of the escape values of the two branches.

#### 5. Termination

In our abstract escape semantics, a function may be expressed recursively and is thus defined as the least fixpoint of the corresponding functional. That is, for the function

$$f = F(f)$$

where f is of type T and F is a functional (corresponding to the body of f), the meaning of f is defined to be the least function satisfying the above equation. Domain theory tells us that the least fixpoint f can be found as follows:

$$f = \lim_{i \to \infty} F^{i} (\perp_{T})$$

where  $F^{0}(x) = x$  and  $F^{i}(x) = F(F^{i-1}(x))$ .

$$E_{ae}[\![\![\,x]\!]\!] Env = <0, err>$$

$$E_{ae}[\![\![\,x]\!]\!] Env = Env[\![\,x]\!]$$

$$E_{ae}[\![\![\,e_1\!]\!] + e_2\!]\!] Env = <0, err>, \text{ likewise for the other arithmetic operators}$$

$$E_{ae}[\![\![\,e_1\!]\!] e_1 = 0 = (E_{ae}[\![\![\,e_1\!]\!]\!] Env)_{(2)} \quad (E_{ae}[\![\![\,e_2\!]\!]\!] Env)$$

$$E_{ae}[\![\![\,e_1\!]\!] \rightarrow e_2, e_3\!]\!] Env = (E_{ae}[\![\![\,e_2\!]\!]\!] Env) \sqcup (E_{ae}[\![\![\,e_3\!]\!]\!] Env)$$

$$E_{ae}[\![\![\,\lambda x.e]\!]\!] Env = < v, \lambda y. E_{ae}[\![\![\,e_1\!]\!]\!] Env[y/x] >$$

$$\text{where } v = \bigvee_{z \in F} (Env[\![\![\,z]\!]\!])_{(1)}, \text{ and } F \text{ is the set of free variables in } (\lambda x.e)$$

$$E_{ae}[\![\![\![\,e_1\!]\!]\!] Env = E_{ae}[\![\![\,e_1\!]\!]\!] Env'$$

$$\text{where } Env' = [(E_{ae}[\![\![\,e_1\!]\!]\!] Env')/x_1, \ldots, (E_{e}[\![\![\,e_n\!]\!]\!] Env')/x_n]$$

To ensure that our analysis terminates, we must show that a fixpoint is reached in finite time. That is, for all functions defined according to the above equation, there must exist some j such that

Figure 4. The abstract escape semantics of nml

$$F^k(\perp_T) = F^j(\perp_T)$$

for all k > j.

A simple way of showing that there exists such a j is to show that every functional F must be monotonic and that the fixpoint iteration is performed over a finite domain (using the technique described in [BHA85]).

Every functional is composed of the monotonic operation  $\vee$  (logical or) and the least upper bound operator (as defined in figure 4) and is thus monotonic. Furthermore, each subdomain  $D_{ae}^{T}$  is finite since  $D_{ae}^{T_0}$  is finite for each primitive type  $T_0$  (int, bool, etc.) and  $D_{ae}^{T_1 \to T_2}$  is finite whenever  $D_{ae}^{T_1}$  and  $D_{ae}^{T_2}$  are finite. When finding the least fixpoint of a function of type T we need only search over the subdomain  $D_{ae}^{T}$ . Thus, the least fixpoint can be computed in a finite number of iterations. Figure 5 contains an example of fixpoint finding.

# 6. Using the Abstract Functions

In this section, we describe how the abstract functions are used to detect the escape properties of the corresponding functions in an nml program.

# 6.1. Global Escape Analysis

Using a global escape analysis, we find escape information about each nml function f that holds true for every possible application of f. To do so, we apply the corresponding abstract function to arguments that cause the greatest escapement possible. For each type T, we define the abstract function  $R^T$  that corresponds to an nml function from which every argument escapes.

$$R^T = \lambda z_1 . \langle z_{1(1)}, \lambda z_2 . \langle z_{1(1)} \vee z_{2(1)}, \dots, \lambda z_m . \langle \bigvee_{p=1}^m z_{p(1)}, err \rangle \dots > \rangle$$

where m is the number of arguments that a function of type T can take (before returning a primitive value).

Given an identifier f bound to a function of n arguments in some environment Env,  $G_i(f, Env)$  returns I if the ith parameter could escape and 0 otherwise.  $G_i$  is defined as follows:

$$G_i(f, Env) = (E_{ae} \llbracket (f x_1 \dots x_n) \rrbracket Env[y_1/x_1, \dots, y_n/x_n])_{(1)}$$
 where for all  $j \le n, j \ne i$ ,

Consider the following nml function definition (of type  $int \rightarrow int$ , for example):

$$f = \lambda x. \lambda y. (x=0) \rightarrow y, f (x-1) y$$

Using the definitions in figure 4, the corresponding function in  $D_{ae}$  is described by:

$$\begin{split} f &= <0, \lambda x. < x_{(1)}, \lambda y. \ E_{ae} \llbracket \ (\mathbf{x=0}) \to \ \mathbf{y}, \ \mathbf{f} \ (\mathbf{x-1}) \ \mathbf{y} \rrbracket \ \llbracket \ (\mathbf{x}/\mathbf{x}, \mathbf{y}/\mathbf{y}, \mathbf{f}/\mathbf{f} \rrbracket >> \\ &= <0, \lambda x. < x_{(1)}, \lambda y. \ \mathbf{y} \ \sqcup \ (f_{(2)} < 0, err >)_{(2)} \ \mathbf{y}) >> \end{split}$$

Therefore f is the least fixpoint of the functional F defined by

$$F = \lambda f. <0, \ \lambda x. < x_{(1)}, \ \lambda y. \ y \sqcup (f_{(2)} < 0, err >)_{(2)} \ y)>>$$

The least fixpoint is found by the following fixpoint iteration:

$$\begin{split} f^0 &= F(\bot_{int \to int \to int}) = <0, \, \lambda x. \, < x_{(1)}, \, \lambda y. \, y \sqcup (\bot_{int \to int \to int(2)} < 0, err >)_{(2)} \, y) >> \\ &= <0, \, \lambda x. \, < x_{(1)}, \, \lambda y. \, y \sqcup \bot_{int \to int(2)} \, y >> \\ &= <0, \, \lambda x. \, < x_{(1)}, \, \lambda y. \, y \sqcup \bot_{int}) >> \\ &= <0, \, \lambda x. \, < x_{(1)}, \, \lambda y. \, y \sqcup \bot_{int}) >> \\ f^l &= F(f^0) &= <0, \, \lambda x. \, < x_{(1)}, \, \lambda y. \, y \sqcup ((\lambda x. \, < x_{(1)}, \, \lambda y. y >) < 0, err >)_{(2)} \, y)) >> \\ &= <0, \, \lambda x. \, < x_{(1)}, \, \lambda y. \, y \sqcup (\lambda y. y) \, y)) >> \\ &= <0, \, \lambda x. \, < x_{(1)}, \, \lambda y. \, y) >> \end{split}$$

Since  $f^0 = f^1 = F(f^0)$ , a fixpoint has been found, thus  $f = \langle 0, \lambda x. \langle x_{(1)}, \lambda y. y \rangle >$ 

This means that when f is applied to two arguments, only the second argument may escape.

Figure 5. An example of fixpoint finding

$$y_j = \langle 0, R^{T_j} \rangle$$

where  $T_i$  is the type of  $x_j$  and

$$y_i = \langle I, R^{T_i} \rangle.$$

Since each  $y_{i(2)}$  is a function from which every argument escapes, and since the abstract function for f is monotonic,  $G_i(f, Env)$  provides the worst case behavior with respect to the escapement of f's ith argument.

### 6.2. Local Escape Analysis

Generally, we would like to know if an argument escapes from a particular call to a function f. This depends on the values of the arguments of that call. We define the function  $L_i$  such that  $L_i(f, e_1, ..., e_n, Env)$ returns l if the *i*th argument of  $(fe_1 \dots e_n)$  might escape,  $\theta$  otherwise. The environment Env must be an environment mapping the free identifiers within  $e_1$  through  $e_n$  to elements of  $D_{ae}$ . The function  $L_i$  is defined as follows

$$L_{i}(f, e_{1},..., e_{n}, Env) = (E_{ae} [ (f x_{1} ... x_{n}) ] Env[y/x_{j}])_{(1)}$$
where, for all  $j \le n, j \ne i$ ,
$$y = <0, (E_{ae} [ e : ] Env)_{(2)} >$$

$$y_j = \langle 0, (E_{ae} [[e_j]] Env)_{(2)} \rangle$$

and

$$y_i = \langle l, (E_{ae} [[e_i]] Env)_{(2)} \rangle.$$

```
Env = [f/f, h/h, p/p, q/q, g/g], \text{ where } \\ f = <0, \lambda x. < x_{(1)}, \ \lambda y. \ E_{ae} [[x+y]] \ Env[x/x,y/y]>> \\ = <0, \lambda x. < x_{(1)}, \ \lambda y. <0, err>>> \\ p = q = <0, \ \lambda b. <0, err>> \\ h = <0, \lambda a. \ E_{ae} [[(a=0) \rightarrow p, q]] \ Env[a/a]> \\ = <0, \lambda a. \ p \sqcup q> \\ = <0, \lambda a. <0, \ \lambda b. <0, err>>> \\ g = <0, \lambda m. < m_{(1)}, \ \lambda n. \ E_{ae} [[mn]] \ Env[m/m,n/n]>> \\ = <0, \lambda m. < m_{(1)}, \ \lambda n. \ m_{(2)} \ n>>
```

Figure 6. Abstract Escape Functions

# 7. Examples

Consider the following nml program:

```
let f = \lambda x. \lambda y. x+y;

h = \lambda a. (a=0) \rightarrow p, q;

p = \lambda b. b+1;

q = \lambda b. b-1;

g = \lambda m. \lambda n. m n;

in ... (g f 4) ... (g h 4) ...
```

Figure 6 shows the corresponding abstract escape functions. Our aim to is analyze the escape properties of g, both globally and locally.

# 7.1. A Global Escape Analysis Example

To find the global (i.e. worst case) escape property of g, we apply the global analysis function  $G_i$  described in section 6.1 to the abstract function g and the environment Env shown in figure 6. We assume g is of type  $(int \rightarrow int \rightarrow int \rightarrow int \rightarrow int \rightarrow int)$ .

$$G_{I}(\mathbf{g}, Env) = (E_{ae} [ (\mathbf{g} \ x_{I} \ x_{2})] Env[y_{I}/x_{I}, y_{2}/x_{2}])_{(I)}$$
where  $y_{I} = \langle I, \lambda z_{I}, \langle z_{I(1)}, \lambda z_{2}, \langle z_{I(1)} \vee z_{2(I)}, \text{ err} \rangle \rangle$  and  $y_{2} = \langle 0, err \rangle$ .
Thus
$$G_{I}(\mathbf{g}, Env) = ((g_{(2)} \ y_{I})_{(2)} \ y_{2})_{(I)}$$

$$= (((\lambda m. \langle m_{(I)}, \lambda n. m_{(2)} \ n \rangle) \ y_{I})_{(2)} \ y_{2})_{(I)}$$

$$= ((\lambda z_{I}, \langle z_{I(1)}, \lambda z_{2}, \langle z_{I(1)} \vee z_{2(I)}, \text{ err} \rangle) \langle 0, err \rangle)_{(I)}$$

$$= 0$$

indicating that g's first parameter can never escape.

$$G_2(\mathbf{g}, Env) = (E_{ae} [ (\mathbf{g} x_1 x_2) ] Env[y_1/x_1, y_2/x_2])_{(1)}$$
 where  $y_1 = \langle 0, \lambda z_1, \langle z_{1(1)}, \lambda z_2, \langle z_{1(1)} \vee z_{2(1)}, \text{err} \rangle \rangle$  and  $y_2 = \langle l, err \rangle$ . Thus 
$$G_2(\mathbf{g}, Env) = ((g_{(2)} y_1)_{(2)} y_2)_{(1)}$$
 
$$= (((\lambda m. \langle m_{(1)}, \lambda n. m_{(2)} n \rangle) y_1)_{(2)} y_2)_{(1)}$$
 
$$= ((\lambda z_1, \langle z_{1(1)}, \lambda z_2, \langle z_{1(1)} \vee z_{2(1)}, \text{err} \rangle) \langle l, err \rangle)_{(1)}$$

indicating that the first parameter to g might escape in some situations.

# 7.2. A Local Escape Analysis Example

To find what arguments escape from the each application of g in the program, we use the local analysis function  $L_i$  described in section 6.2. Since we know from the global analysis of g that its first parameter can never escape, we only need to test if the second parameter escapes (using  $L_2$ ). For the expression (g f 4),

$$\begin{split} L_{2}(\mathbf{g},\mathbf{f},\mathbf{4},Env) &= (E_{ae} \, \llbracket \, (\mathbf{g} \, \mathbf{x}_{1} \, \mathbf{x}_{2}) \rrbracket \, \, Env[<0,f_{(2)}>/x_{1},<1,\,err>/x_{2}])_{(1)} \\ &= ((g_{(2)}<0,f_{(2)}>)_{(2)}<1,\,err>)_{(1)} \\ &= (((\lambda m.\,< m_{(1)},\,\,\lambda n.\,m_{(2)}\,n>)<0,\,\lambda x.\,< x_{(1)},\,\,\lambda y.\,<0,\,err>>>)_{(2)}<1,\,err>)_{(1)} \\ &= ((\lambda x.\,< x_{(1)},\,\,\lambda y.\,<0,\,err>>>)_{(2)}<1,\,err>)_{(1)} \\ &= 1. \end{split}$$

This indicates that the second argument to g escapes. For the expression (g h 4),

$$\begin{split} L_2(\mathbf{g},\,\mathbf{h},\,\mathbf{4},\mathit{Env}) &= (E_{ae}\, [\![\, (\mathbf{g}\,x_1\,x_2)]\!]\, \mathit{Env}[\,<0,\,h_{(2)}\!>\!/x_1,\,< l\,,\,err\!>\!/x_2])_{(1)} \\ &= ((g_{(2)}<0,\,h_{(2)}\!>)_{(2)}< l\,,\,err\!>)_{(1)} \\ &= (((\lambda m.\,< m_{(1)},\,\,\lambda n.\,m_{(2)}\,n\!>)<0,\,\lambda a.\,<0,\,\,\lambda b.\,<0,\,err\!>>)_{(2)}< l\,,\,err\!>)_{(1)} \\ &= ((\,\lambda a.\,<0,\,\,\lambda b.\,<0,\,err\!>>)_{(2)}< l\,,\,err\!>)_{\,(1)} \\ &= 0. \end{split}$$

This indicates that no argument to g escapes (even though the result is a partial application).

# 8. Escape Analysis on Lists

We have not yet discussed escape analysis in the presence of the list operators cons, car, and cdr. We extend the abstract semantic function  $E_{de}$  as follows:

$$E_{ae}[[cons]] Env = <0, \lambda a. < a_{(1)}, \lambda b. < a_{(1)} \lor b_{(1)}, err>>>$$
 $E_{ae}[[cor]] Env = <0, \lambda x. < x_{(1)}, err>>$ 
 $E_{ae}[[cor]] Env = <0, R^T>$ 

where T is the type of the elements of the list to which car is being applied and  $R^T$  is defined in section 6.1. In other words, once an object has been placed on a list, we are unable to determine when it is removed. This means that if some head or tail of a list escapes from a function then all of the elements of the list are seen as escaping. In addition, the function value of the car of a list is seen as the maximally escaping function of that type. Admittedly, this is an unsatisfactory analysis on lists. We are working on an escape analysis that could be applied to lists in a manner similar to the way that strictness analysis was extended to lists [Wadler87].

#### 9. Conclusions

We have taken an existing, useful optimization and used denotational semantics and abstract interpretation to apply it to higher order programming languages. We have yet to implement the analysis in a real compiler and thus it remains to be seen if the benefit of the analysis outweighs its cost (mainly fixpoint finding).

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