Formal Derivation of Efficient Parallel Programs by Construction of List Homomorphisms

ZHENJIANG HU

Department of Information Engineering, University of Tokyo

and

HIDEYA IWASAKI

Department of Computer Science, Tokyo University of Agriculture and Technology and

MASATO TAKECHI

Department of Information Engineering, University of Tokyo

It has been attracting much attention to make use of list homomorphisms in parallel programming because they ideally suit the divide-and-conquer parallel paradigm. However, they have been usually treated rather informally and ad-hoc in the development of efficient parallel programs. What is worse is that some interesting functions, e.g., the maximum segment sum problem, are basically not list homomorphisms. In this paper, we propose a systematic and formal way for the construction of a list homomorphism for a given problem so that an efficient parallel program is derived. We show, with several well-known but non-trivial problems, how a straightforward, and "obviously" correct, but quite inefficient solution to the problem can be successfully turned into a semantically equivalent "almost list homomorphism". The derivation is based on two transformations, namely tupling and fusion, which are defined according to the specific recursive structures of list homomorphisms.

Categories and Subject Descriptors: I.2.2 [Artificial Intelligence]: Automatic Programming— Program transformation; D.1.1 [Programming Techniques]: Applicative (Functional) Programming; D.1.3 [Programming Techniques]: Concurrent Programming—Parallel Programming; F.3.3 [Logics and Means of Programs]: Studies of Program Constructs—Program and recursion schemes

 $General\ terms:\ Program\ Transformation,\ Parallel\ Programming$

Additional Key Words and Phrases: List homomorphism, Program transformation and derivation, Parallel functional programming

1. INTRODUCTION

It has been attracting wide attention to make use of list homomorphisms in parallel programming [Bird 1987; Cole 1993b; Gorlatch 1995; Gorlatch 1996a; Hu et al. 1996a; Hu et al. 1996c; Chin 1996]. List homomorphisms [Bird 1987] are those functions on finite lists that promote through list concatenation — that is, function h for which there exists an associative binary operator \oplus such that, for all finite lists xs and ys, we have $h(xs + ys) = h xs \oplus h ys$, where # denotes list concatenation. Intuitively, the definition of list homomorphisms means that the value of h on the larger list depends in a particular way (using binary operation \oplus) on the values of h applied to the two pieces of the list. The computations of h xs and h ys

Author's address: Zhenjiang HU, Department of Information Engineering, University of Tokyo. 7-3-1 Hongo, Bunkyo-ku, Tokyo 113, Japan. Email: hu@ipl.t.u-tokyo.ac.jp

are independent each other and can thus be carried out in parallel. This simple equation can be viewed as expressing the well-known divide-and-conquer paradigm in parallel programming.

Therefore, the implications for parallel program development become clear; if the problem is a list homomorphism, then it only remains to define a cheap \oplus in order to produce a highly parallel solution. However, there are a lot of useful and interesting list functions that are not list homomorphisms and thus have no corresponding \oplus . One example is the function mss known as (1-dimensional) maximum segment $sum\ problem$, which finds the maximum of the sums of contiguous segments within a list of integers. For example, $mss\ [3, -4, 2, -1, 6, -3] = 7$, where the result is contributed by the segment [2, -1, 6]. The mss is not a list homomorphism, since knowing $mss\ xs$ and $mss\ ys$ is not enough to allow computation of $mss\ (xs + + ys)$.

To solve this problem, Cole [Cole 1993b] proposed an informal approach showing how to embed these functions into list homomorphisms. His method consists of constructing a homomorphism as a tuple of functions where the original function is one of the components. The main difficulties are to guess which functions must be included in a tuple in addition to the original function and to prove that the constructed tuple is indeed a list homomorphism. The examples given by Cole show that this usually requires a lot of ingenuity from the programmer.

The purpose of this paper is to give a systematic and formal derivation of such list homomorphisms containing the original non-homomorphic function as its component. It is mainly based on our previous works reported in [Hu et al. 1996a; Hu et al. 1996c]. Our main contributions are as follows.

- —Unlike Cole's informal study, we propose a *systematic* way of discovering extra functions which are to be tupled with the original function to form a list homomorphism. We base our method on two main theorems, the Tupling Theorem and the Almost Fusion Theorem, showing how to derive a *true* list homomorphism from recursively defined functions by means of tupling and how to calculate a new homomorphism *incrementally* from the old by means of fusion. It would be interesting to see that our systematic construction of list homomorphisms is of much help in discovering new efficient parallel programs (Section 5).
- —Our main theorems for tupling and fusion are given in a calculational style [Meijer et al. 1991; Takano and Meijer 1995; Hu et al. 1996b] rather than being based on the fold/unfold transformation [Chin 1992; Chin 1993]. Therefore, infinite unfoldings, once inherited in the fold/unfold transformation, can be definitely avoided by the theorems themselves. Furthermore, although we restrict ourselves to list homomorphisms, our theorems could be extended naturally for homomorphisms of arbitrary data structures (e.g., trees) with the theory of constructive algorithmics [Fokkinga 1992].
- —Our derivation of parallel program proceeds in a formal way, leading to a correct solution with respect to the initial specification. We start with a simple, and "obviously" correct, but possibly inefficient solution to the problem, and then transform it based on our rules and algebraic identities into a semantically equivalent list homomorphism. Furthermore, as will be seen later, most of our derivation are mechanical and thus could be made automatically and embedded in a parallel compiler.

We shall illustrate our idea using the maximum segment sum problem mss as our

running example. This problem is of interest because there are efficient but non-obvious algorithms to compute it, both in sequential [Bird 1987] and in parallel [Cai and Skillicorn 1992; Cole 1993b].

This paper is organized as follows. In Section 2, we review the notational conventions and some basic concepts used in this paper. After showing how to specify problems in Section 3, we focus ourselves on deriving an efficient (almost) list homomorphism from the specification by using our two important theorems, namely the Tupling and the Almost Fusion Theorems in Section 4. In Section 5, we illustrate how our systematic way is also very useful in discovering new efficient parallel programs. Concluding remarks are given in Section 6.

2. PRELIMINARY

In this section, we briefly review the notational conventions known as Bird-Meertens Formalisms [Bird 1987] and some basic concepts which will be used in the rest of this paper.

Functions

Functional application is denoted by a space and the argument which may be written without brackets. Thus f a means f (a). Functions are curried and application associates to the left. Thus f a b means (f a) b. Functional application is regarded as more binding than any other operator, so f $a \oplus b$ means (f $a) \oplus b$ but not f $(a \oplus b)$. Functional composition is denoted by a centralized circle \circ . By definition, $(f \circ g)$ a = f (g a). Functional composition is an associative operator, and the identity function is denoted by id. Infix binary operators will often be denoted by \oplus , \otimes and can be sectioned; an infix binary operators like \oplus can be turned into unary functions by: $(a \oplus)$ $b = a \oplus b = (\oplus b)$ a.

The followings are some important operators (functions) used in the paper.

—The projection function π_i will be used to select the *i*-th component of tuples, e.g., $\pi_1(a,b) = a$. The \triangle and \times are two important operators related to tuples, defined by

$$(f \triangle g) a = (f a, g a), \quad (f \times g) (a, b) = (f a, g b).$$

The \triangle can be naturally extended to functions with two arguments. So, we have $a (\oplus \triangle \otimes) b = (a \oplus b, a \otimes b)$.

—The *cross* operator \mathcal{X}_{\oplus} , which crosswisely combines elements in two lists with operator \oplus , is defined informally by:

$$[x_1, \dots, x_n] \ \mathcal{X}_{\oplus} [y_1, \dots, y_m] = [x_1 \oplus y_1, \dots, x_1 \oplus y_m, \dots, x_n \oplus y_1, \dots, x_n \oplus y_m].$$

The cross operator enjoys many algebraic identities, e.g., $(f*) \circ \mathcal{X}_{\oplus} = \mathcal{X}_{f \circ \oplus}$.

—The concat, a function to flatten a list, is defined by:

$$concat [xs_1, \dots, xs_n] = xs_1 + + \dots + xs_n.$$

—The zip-with operator Υ_{\oplus} , a function to apply \oplus pairwisely to two lists, is informally defined by

$$[x_1, \dots, x_n] \Upsilon_{\oplus} [y_1, \dots, y_n] = [x_1 \oplus y_1, \dots, x_n \oplus y_n].$$

Lists

Lists are finite sequences of values of the same type. A list is either empty, a singleton, or the concatenation of two other lists. We write [] for the empty list, [a] for the singleton list with element a (and $[\cdot]$ for the function taking a to [a]), and xs + + ys for the concatenation of xs and ys. Concatenation is associative, and [] is its unit. For example, the term [1] ++ [2] ++ [3] denotes a list with three elements, often abbreviated to [1, 2, 3].

List Homomorphisms

A function h satisfying the following three equations will be called a *list homomorphism*.

$$\begin{array}{ll} h \; [\;] & = \; \iota_{\oplus} \\ h \; [x] & = \; f \; x \\ h \; (xs \; +\!\!\!+\; ys) \; = \; h \; xs \oplus h \; ys \end{array}$$

It soon follows from this definition that \oplus must be an associative binary operator with unit ι_{\oplus} . For notational convenience, we write $([f, \oplus])^1$ for the unique function h, e.g., sum = ([id, +]) and $max = ([id, \uparrow])$, where \uparrow denotes the binary maximum function whose unit is $-\infty$. Note when it is clear from the context, we usually abbreviate "list homomorphisms" to "homomorphism."

Two important list homomorphisms are *map* and *reduction*. Map is the operator which applies a function to every item in a list. It is written as an infix *. Informally, we have

$$f * [x_1, x_2, \dots, x_n] = [f x_1, f x_2, \dots, f x_n].$$

Reduction is the operator which collapses a list into a single value by repeated application of some binary operator. It is written as an infix /. Informally, for an associative binary operator \oplus , we have

$$\oplus$$
 / $[x_1, x_2, \cdots, x_n] = x_1 \oplus x_2 \cdots \oplus x_n$.

It is not difficult to see that * and / have simple massively parallel implementations on many architectures. For example, \oplus / can be computed in parallel on a tree-like structure with the combining operator \oplus applied in the nodes, whereas f* is totally parallel. The relevance of list homomorphisms to parallel programming can be seen clearly from the Homomorphism Lemma [Bird 1987]: $(f, \oplus) = (\oplus/) \circ (f*)$, saying that every list homomorphism can be written as the composition of a reduction and a map. This implies that a list homomorphism (f, \oplus) can be simply implemented using $O(\log n) \times C(\oplus) + C(f)$ parallel time where n stands for the size of input list, $C(\oplus)$ for the cost of \oplus , and C(f) for the cost of f.

Almost Homomorphisms

Simple as they are, list homomorphisms cannot specify a lot of interesting functions as explained in Introduction. To solve this problem, Cole [Cole 1993b] argued informally that some of them can be converted into so-called *almost* (*list*) homomorphisms by tupling them with some extra functions so that the tupled function

¹Strictly speaking, we should write $([\iota_{\oplus}, f, \oplus])$ to denote the unique function h. We can omit the ι_{\oplus} because it is the unit of \oplus .

can be specified by a list homomorphism. In other words, an almost homomorphism is a composition of a projection function and a list homomorphism. Since projection functions are simple, almost homomorphisms are also suitable for parallel implementation as list homomorphisms do.

In fact, it may be surprising to see that every function can be represented in terms of an almost homomorphism [Gorlatch 1995]. Let k be a non-homomorphic function. Consider a new function g such that g x = (x, k x). The tupled function g is homomorphic, i.e., $g(xs + ys) = (xs + ys, k(xs + ys)) = gxs \oplus gys$, where $(xs_1, k_1) \oplus (xs_2, k_2) = (xs_1 + xs_2, k(xs_1 + xs_2))$, and we have the following almost homomorphism for k:

$$k = \pi_2 \circ g = \pi_2 \circ ([g \circ [\cdot], \oplus]).$$

However, a closer look at the definition of operation \oplus reveals the drawback: it is quite expensive and meaningless in that it does not make use of the previously computed values k_1 (= k xs_1) and k_2 (= k xs_2) and computes k from scratch! In this sense, we say it is not an expected "true" almost homomorphism.

In order to derive a "true" almost homomorphism, a suitable tupled function should be carefully defined, making full use of previously computed values. Cole reported several case studies of such derivation with parallel algorithms as a result, and stressed that in each case the derivation requires a lot of intuition [Cole 1993b; Cole 1993a]. In this paper, we shall propose a systematic approach to this derivation.

3. SPECIFICATION

Given problems, we aim at a formal derivation of efficient parallel programs by constructing list homomorphisms including the original as its component (i.e., almost homomorphisms)². To talk about parallel program derivation, we should be clear about specifications. It is advocated by transformational programming [Bird 1984; Feather 1987; Pettorossi and Proietti 1993] that specifications should be given as naive solutions to problems where we only focus ourselves on a simple but correct solutions without concerning with efficiency or parallelism. More precisely, our specification for a problem p will be a simple, and "obviously" correct, but possibly inefficient solution with the form in a compositional style:

$$p = p_n \circ \dots \circ p_2 \circ p_1 \tag{1}$$

where each p_i is a (recursively defined) function. This reflects our way of solving problems; a (big) problem p may be solved through multiple passes while in each pass a simpler problem p_i is solved by a recursion.

Consider our running example of maximum segment sum problem. An obviously correct solution to the problem is $mss: [Int] \rightarrow Int$ defined by:

$$mss = max \circ (sum*) \circ seqs$$

which is implemented by three passes; (1) computing all contiguous segments of a sequence by segs, (2) summing up each contiguous segment by sum, (3) selecting the largest value by max.

 $^{^{2}}$ Note that list homomorphisms can be considered as a special case of almost list homomorphisms where the projection part is *identity* function.

```
mss: [(Index, Int)] \rightarrow Int
mss = max' \circ (sum'*) \circ segs
where
                    =([\pi_2,\uparrow'])
max'
                        where (is, x) \uparrow' (js, y)). x \uparrow y
                    = ([\lambda(i,x).([i],x),+'])
sum
                        where (is, x) +' (js, y) = (is ++ js, x + y)
segs
segs [x]
                    = [[x]]
segs (xs ++ ys) = segs xs ++ \prec segs ys ++ \prec (tails xs \mathcal{X}_{++} inits ys)
                    = []
inits
                    = [[x]]
inits [x]
inits (xs ++ ys) = inits xs ++ (xs ++) * (inits ys)
tails []
                    = []
tails [x]
                    = [[x]]
tails (xs ++ ys) = (++ ys) * (tails xs) ++ tails ys
```

Fig. 1. Specification for mss Problem

The only unknown function in the specification is $segs:[Int] \to [[Int]]$, computing all segments of a list. It would be likely to define it simply as

```
segs (xs ++ ys) = segs xs ++ segs ys ++ (tails xs X_{++} inits ys).
```

The equation reads that all segments in the sequence $xs +\!\!\!\!+ ys$ are made up of three parts: all segments in xs, all segments in ys, and all segments produced by crosswisely concatenating every tail segment of xs (i.e., the segment in xs ending with xs's last element) with every initial segment of ys (i.e., the segment in ys starting with ys's first element). Here, inits and tails are standard functions in [Bird 1987], though our definitions are slightly different as will be seen later. Being simple, it is a wrong definition for segs, as you may have noticed that, for example, segs ([1,2] ++[3]) $\neq segs$ ([1] ++[2,3]) while they are expected to be equal (to segs [1,2,3]). A closer look reveals that the two resulting lists indeed consist of all segments of [1,2,3] but in different order. One way to remedy this situation is to force segs to give result of a sorted list of segments under a total order, say \prec , and thus we can define segs correctly as

```
segs\ (xs ++ ys) = segs\ xs ++ \prec segs\ ys ++ \prec (tails\ xs\ \mathcal{X}_{++}\ inits\ ys)
```

where ++ merges two sorted lists into one with respect to the order of \prec .

Let's see how we can define such \prec in a simple way. Let $[x_{i_1}, x_{i_1+1}, \cdots, x_{j_1}]$ and $[x_{i_2}, x_{i_2+1}, \cdots, x_{j_2}]$ be two segments of the presumed list $[x_1, \cdots, x_n]$. Then, \prec is defined by $[x_{i_1}, x_{i_1+1}, \cdots, x_{j_1}] \prec [x_{i_2}, x_{i_2+1}, \cdots, x_{j_2}] =_{def} [i_1, \cdots, j_1] <_D [i_2, \cdots, j_2]$, where $<_D$ stands for the lexicographic order on indices. To capture the index information in our specification, we extend the input type of mss and segs from lists of integers, [Int], to lists of pairs of index and integer, [(Index, Int)]. Also, we change max to max' and sum' to sum taking account of this additional index

information.

So much for the specification of the *mss* problem, which is summarized in Fig. 1. It is a naive solution of the problem without concerning efficiency and parallelism at all, but its correctness is obvious.

4. DERIVATION

Our derivation of a "true" almost homomorphism from the specification (1) in a compositional style is carried out *incrementally* by the following procedure.

Step 1. Derive an almost homomorphism from the recursive definition of p_1 (Section 4.1);

Step 2. Fuse p_2 into the derived almost homomorphism to obtain a new almost homomorphism for $p_2 \circ p_1$ and repeat this derivation until p_n is fused (Section 4.2);

Step 3. Let $\pi_1 \circ ([f, \oplus])$ be the resulting almost list homomorphism for $p_n \circ \cdots \circ p_2 \circ p_1$ obtained at Step 2. For the functions inside the homomorphism, namely f and \oplus , try to repeat Step 1 and 2 to find efficient parallel implementations for them.

We are confronted with two problems here: (a) How an almost homomorphism can be derived from a recursive definition? (b) How a new almost homomorphism can be calculated out of a composition of a function and an old one?

4.1 Deriving almost homomorphisms

Although some functions cannot be described directly by list homomorphisms, they may be easily described by (mutual) recursive definitions while some other functions might be used (see segs in Section 3 for an example) [Fokkinga 1992]. In this section, we propose a way of deriving almost homomorphisms from such (mutual) recursive definitions, systematically discovering extra functions that should be tupled with the original function to turn it into a "true" list homomorphism. The "true" list homomorphism must fully reuse the previously computed values in the sense that there are no redundant recursive calls to the original function or to any newly-discovered extra function, as discussed in Section 2. Our approach is based on the following theorem. For notational convenience, we define $\Delta_1^n f_i = f_1 \triangle f_2 \triangle \cdots \triangle f_n$.

THEOREM 1. (TUPLING) Let h_1, \dots, h_n be mutual-recursively defined by:

$$h_{i}[] = \iota_{\bigoplus_{i}}$$

$$h_{i}[x] = f_{i}x$$

$$h_{i}(xs + + ys) = ((\Delta_{1}^{n}h_{i})xs) \oplus_{i} ((\Delta_{1}^{n}h_{i})ys)$$

$$(2)$$

Then $\Delta_1^n h_i$ is a list homomorphism $([\Delta_1^n f_i, \Delta_1^n \oplus_i])$ and $(\iota_{\oplus_1}, \dots, \iota_{\oplus_n})$ is the unit of $\Delta_1^n \oplus_i$.

PROOF. According to the definition of list homomorphisms, it is sufficient to prove that

$$\begin{array}{lll} (\Delta_1^n h_i) \ [] & = \ (\iota_{\oplus_1}, \cdots, \iota_{\oplus_n}) \\ (\Delta_1^n h_i) \ [x] & = \ (\Delta_1^n f_i) \ x \\ (\Delta_1^n h_i) \ (xs ++ ys) & = \ ((\Delta_1^n h_i) \ xs) \ (\Delta_1^n \oplus_i) \ ((\Delta_1^n h_i) \ ys). \end{array}$$

The first two equations are trivial. The last can be proved by the following calculation.

```
 \begin{array}{ll} LHS \\ =& \{ \text{ Def. of } \Delta \text{ and } \triangle \ \} \\ & (h_1(xs +\!\!\!\!+ ys), \cdots, h_n(xs +\!\!\!\!+ ys)) \\ =& \{ \text{ Def. of } h_i \ \} \\ & (((\Delta_1^n h_i) \, xs) \oplus_1 \left( (\Delta_1^n h_i) \, ys), \, \cdots, \, \left( (\Delta_1^n h_i) \, xs \right) \oplus_n \left( (\Delta_1^n h_i) \, ys \right)) \\ =& \{ \text{ Def. of } \triangle \text{ and } \Delta \ \} \\ & RHS \end{array}
```

Theorem 1 says that if h_1 is mutually defined with other functions (i.e., $h_2, \dots h_n$) which traverse over the same lists in the specific form of (2), then tupling h_1, \dots, h_n will definitely give a list homomorphism. It follows that every h_i is an almost homomorphism. Particularly, h_1 can be represented in the way of the projection function π_1 composed with the list homomorphism for the tupled function. It is worth noting that this style of tupling can avoid repeatedly redundant computations of h_1, \dots, h_n in the computation of the list homomorphism of $\Delta_1^n h_i$ [Takeichi 1987]. That is, all previous computed results by h_1, \dots, h_n can be fully reused, as expected in "true" almost homomorphisms.

Practically, not all recursive definitions are in the form of (2). They, however, can be turned into such form by a simple transformation. Let's demonstrate how the tupling theorem works in deriving a "true" almost homomorphism from the definition of segs given in Section 3.

First, we determine what functions are to be tupled, i.e., finding h_1, \dots, h_n . As explained above, the functions to be tupled are those which traverse over the same lists in the definitions. So, from the definition of segs:

```
segs\ (xs ++ ys) = \underline{segs\ xs} ++ \underline{\langle\ segs\ ys} ++ \underline{\langle\ (tails\ xs\ \mathcal{X}_{++}\ \underline{inits\ ys})}
```

we know that segs needs to be tupled with tails and inits, because segs and inits traverse the same list xs whereas segs and tails traverse the same list ys as underlined. Going to the definition of inits,

$$inits (xs ++ ys) = \underline{inits xs} ++ (\underline{xs} ++) * (inits ys)$$

we find that the *inits* needs to be tupled with id, the identity function, since xs = id xs. Similarly, The tails needs to be tupled with id. Note that id is the identity function over lists defined by

$$\begin{array}{lll} id & [&] & = & [&] \\ id & [x] & = & [x] \\ id & (xs ++ ys) & = & id & xs ++ id & ys \\ \end{array}$$

To summarize the above, the functions to be tupled are segs, inits, tails, and id, i.e., our tuple function will be $segs \triangle inits \triangle tails \triangle id$.

Next, we rewrite the definitions of the functions in the above tuple to the form of (2), i.e., deriving f_1, \oplus_1 for segs, f_2, \oplus_2 for inits, f_3, \oplus_3 for tails, and f_4, \oplus_4 for id. In fact, this is straightforward: just selecting the corresponding recursive calls

from the tuples. From the definition of segs, we have

It would be helpful for understanding the above derivation if we notice the following correspondences: s_1 to $segs\ xs$, i_1 to $inits\ xs$, t_1 to $tails\ xs$, d_1 to $id\ xs$, s_2 to $segs\ ys$, i_2 to $inits\ ys$, t_2 to $tails\ ys$, d_2 to $id\ ys$. Similarly, for inits, tails and id, we have

Now we are ready to apply Theorem 1 and get the following list homomorphism.

$$segs \triangle inits \triangle tails \triangle id = ([\Delta_1^4 f_i, \Delta_1^4 \oplus_i])$$

And our almost homomorphism for segs is thus obtained:

$$segs = \pi_1 \circ ([\Delta_1^4 f_i, \Delta_1^4 \oplus_i]). \tag{3}$$

It would be intersting to see that the above derivation is practically mechanical. Note that the derivation of the unit of the new binary operator (e.g., $\Delta_1^4 \oplus_i$) is omitted because this is trivial; the new tuple function applying to empty list will give exactly this unit (e.g., (segs \triangle inits \triangle tails \triangle id)[]). The derivation of units will be omitted in the rest of the paper as well.

4.2 Fusion with Almost Homomorphisms

In this section, we show how to fuse a function with an almost homomorphism, the second problem (b) as listed at the beginning of Section 4.

It is well known that list homomorphisms are suitable for program transformation in that there is a general rule called *Fusion Theorem* [Bird 1987], showing how to fuse a function with a list homomorphism to get another new list homomorphism.

THEOREM 2. (FUSION) Let h and $([f, \oplus])$ be given. If there exists \otimes such that $\forall x, y. \ h \ (x \oplus y) = h \ x \otimes h \ y$, then $h \circ ([f, \oplus]) = ([h \circ f, \otimes])$.

This fusion theorem, however, cannot be used directly for our purpose. As seen in Equation (3), we usually derive an almost homomorphism and we hope to know how to fuse functions with almost homomorphisms, namely we want to deal with the following case:

$$h \circ (\pi_1 \circ ([\Delta_1^n f_i, \Delta_1^n \oplus_i])).$$

We'd like to shift π_1 left and promote h into the list homomorphism. Our fusion theorem for this purpose is given below.

THEOREM 3. (ALMOST FUSION) Let h and $([\Delta_1^n f_i, \Delta_1^n \oplus_i])$ be given. If there exist $\otimes_i (i = 1, \dots, n)$ and a map $H = h_1 \times \dots \times h_n$ where $h_1 = h$ such that for all j,

$$\forall x, y. \ h_i(x \oplus_i y) = H \ x \otimes_i H \ y \tag{4}$$

then

$$h \circ (\pi_1 \circ ([\Delta_1^n f_i, \Delta_1^n \oplus_i])) = \pi_1 \circ ([\Delta_1^n (h_i \circ f_i), \Delta_1^n \otimes_i])$$

PROOF. We prove it by the following calculation.

$$\begin{array}{ll} h \circ \left(\pi_1 \circ \left(\left[\Delta_1^n f_i, \Delta_1^n \oplus_i\right]\right)\right) \\ = & \left\{ \begin{array}{ll} \operatorname{By} \ \pi_1 \ \operatorname{and} \ H \end{array} \right\} \\ \pi_1 \circ H \circ \left(\left[\Delta_1^n f_i, \Delta_1^n \oplus_i\right]\right) \\ = & \left\{ \begin{array}{ll} \operatorname{Theorem} \ 2, \ \operatorname{and} \ \operatorname{the proves} \ \operatorname{below} \end{array} \right\} \\ \pi_1 \circ \left(\left[\Delta_1^n (h_i \circ f_i), \Delta_1^n \otimes_i\right]\right) \end{array}$$

To complete the above proof, we need to show that for any x and y,

$$\begin{array}{ll} H\left(x\left(\Delta_{1}^{n} \oplus_{i}\right) y\right) \ = \ \left(H \ x\right)\left(\Delta_{1}^{n} \otimes_{i}\right)\left(H \ y\right) \\ H \circ \left(\Delta_{1}^{n} f_{i}\right) & = \ \Delta_{1}^{n}\left(h_{i} \circ f_{i}\right) \end{array}$$

The second equation is easy to prove. For the first, we argue that

$$\begin{array}{ll} LHS \\ = & \{ \text{ Expanding } \Delta, \text{ Def. of } \triangleq \} \\ H\left(x \oplus_1 y, \cdots, x \oplus_n y\right) \\ = & \{ \text{ Expanding } H, \text{ Def. of } \times \} \\ \left(h_1(x \oplus_1 y), \cdots, h_n\left(x \oplus_n y\right)\right) \\ = & \{ \text{ Assumption } \} \\ \left(Hx \otimes_1 Hy, \cdots, Hx \otimes_n Hy\right) \\ = & \{ \text{ Def. of } \triangleq, \Delta \ \} \\ RHS \end{array}$$

Theorem 3 suggests a way of fusing a function h with the almost homomorphism $\pi_1 \circ ([\Delta_1^n f_i, \Delta_1^n \oplus_i])$ in order to get another almost homomorphism; trying to find h_2, \dots, h_n together with $\oplus_1, \dots, \oplus_n$ that meet the equation (4). Note that, without loss of generality we restrict the projection function of our almost homomorphisms to π_1 in the theorem.

Returning to our running example, recall that we have reached the point:

$$mss = max' \circ (sum'*) \circ (\pi_1 \circ ([\Delta_1^4 f_i, \Delta_1^4 \oplus_i]).$$

We demonstrate how to fuse sum'* with $\pi_1 \circ ([\Delta_1^4 f_i, \Delta_1^4 \oplus_i])$ by Theorem 3. Let $H = h_1 \times h_2 \times h_3 \times h_4$ where $h_1 = (sum'*)$, and h_2, h_3, h_4 await to be determined. In addition, we need to derive $\otimes_1, \otimes_2, \otimes_3$, and \otimes_4 based on the following equations according to Theorem 3:

```
sum'*((s_1,i_1,t_1,d_1) \oplus_i (s_2,i_2,t_2,d_2)) = (sum'*s_1,h_2 i_1,h_3 t_1,h_4 d_1) \otimes_i (sum'*s_2,h_2 i_2,h_3 t_2,h_4 d_2) \quad (i = 1,\dots,4).
```

Now the derivation procedure becomes clear; calculating each LHS of the above equations to promote sum'* into s_1 and s_2 , and determining the unknown functions $(h_i \text{ and } \otimes_i)$ by matching with its RHS. As an example, consider the following

calculation of the LHS of the the equation for i = 1.

Matching the last expression with its corresponding RHS:

$$(sum'*s_1, h_2 i_1, h_3 t_1, h_4 d_1) \otimes_1 (sum'*s_2, h_2 i_2, h_3 t_2, h_4 d_2)$$

will give

$$h_2 = h_3 = sum'*$$

 $(s_1, i_1, t_1, d_1) \otimes_1 (s_2, i_2, t_2, d_2) = s_1 + + \downarrow_1 s_2 + + \downarrow_1 (t_1 \mathcal{X}_{+'} i_2).$

The others can be similarly derived.

To use Theorem 3, we also need to consider the f part whose results are as follows.

$$\begin{array}{lll} f_1' \; (i,x) \; = \; \left((sum'*) \circ f_1 \right) x \; = \left[([i],x) \right] \\ f_2' \; (i,x) \; = \; \left((sum'*) \circ f_2 \right) x \; = \left[([i],x) \right] \\ f_3' \; (i,x) \; = \; \left((sum'*) \circ f_3 \right) x \; = \left[([i],x) \right] \\ f_4' \; (i,x) \; = \; \left(sum' \circ f_1 \right) x \; & = \; ([i],x) \end{array}$$

According to Theorem 3, we soon have

$$(sum'*) \circ segs = \pi_1 \circ ([\Delta_1^4 f_i', \Delta_1^4 \otimes_i]). \tag{5}$$

Again, we can fuse max' with the above almost homomorphism (in this case, $H = max' \times max' \times max' \times id$) and get the following almost homomorphism, the final result for mss.

$$mss = \pi_1 \circ ([\Delta_1^4 \pi_2, \Delta_1^4 \otimes_i']) \tag{6}$$

where

$$\begin{array}{l} (s_1,i_1,t_1,d_1) \otimes_1' (s_2,i_2,t_2,d_2) = s1 \uparrow s_2 \uparrow (t_1+i_2) \\ (s_1,i_1,t_1,d_1) \otimes_2' (s_2,i_2,t_2,d_2) = i_1 \uparrow (d_1+i_2) \\ (s_1,i_1,t_1,d_1) \otimes_3' (s_2,i_2,t_2,d_2) = (t_1+d_2) \uparrow t_2 \\ (s_1,i_1,t_1,d_1) \otimes_4' (s_2,i_2,t_2,d_2) = d_1+d_2. \end{array}$$

Since the operators of $\Delta_1^4 \pi_2$ and $\Delta_1^4 \otimes_i'$ inside the obtained almost homomorphism are simple and efficient enough, we need not repeat Step 1 and 2 to make them efficient according to our derivation procedure given at the beginning of this Section. Thus we got our result, the same as informally given by Cole [Cole 1993b]. In practical terms, the algorithm looks so promising that on many architectures.

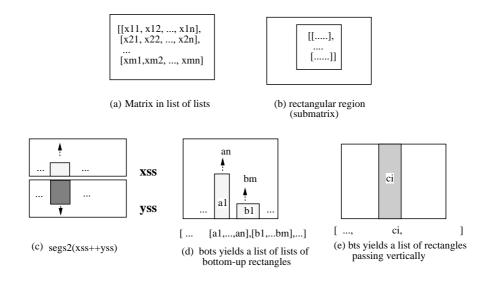


Fig. 2. The mss2 Problem

We can expect an $O(\log n)$ parallel algorithm according to the simple parallel implementation of list homomorphisms (Section 2), observing that $C(\Delta_1^4 \pi_2) = 1$ and $C(\Delta_1^4 \otimes_i') = 1$.

5. 2-DIMENSIONAL MAXIMUM SEGMENT SUM PROBLEM

In this section, we consider a more complicated problem, namely 2-dimensional maximum segment sum problem. In [Smith 1987], the tuple consisting of eleven functions is used for the definition of a $O(\log^2 n)$ parallel algorithm, but the detailed derivation, which would be rather cumbersome with Smith's approach, was not given at all. In the following, we'd like to show that although this problem looks very difficult it can be solved in a quite similar way as we did for the (1-dimensional) maximum segment sum problem resulting in a new efficient parallel program. It would be very intersting to see that our systematic construction of list homomorphisms is of much help in discovering new efficient parallel programs.

5.1 Specification of the Problem

Let's turn to the specification for the 2-dimensional maximum segment sum problem, mss2, a generalization of mss, which finds the maximum over the sum of all rectangular subregions of a matrix. The matrix can be naturally represented by a list of lists with the same length as shown in Fig. 2 (a), and so does its rectangular subregion as in Fig. 2 (b). Following the same thought we did for mss, we define mss2 straightforwardly as in Fig. 3. Here, segs2 computes all rectangular subregions of a matrix, then sum2 is applied to every rectangular subregion and sums up all elements, and finally max returns the largest value as the result.

Function segs2 is defined in a quite similar way to segs. The last equation reads that all rectangular subregions of xss ++ yss, a matrix connecting xss and yss vertically (Fig. 2 (c)), are made up from those in both xss and yss, and those produced by combining every bottom-up rectangular subregion in xss (depicted by

```
mss2: [[(Index', Int)]] \rightarrow Int
mss2 = max' \circ (sum2*) \circ segs2
where
sum 2
                         = sum' \circ sum' *
segs2 []
                         = []
segs2 [xs]
                         = [\cdot] * (segs xs)
segs2 (xss ++ yss) = segs2 xss ++_{\prec '} segs2 yss ++_{\prec '}
                                concat ((bots \ xss)\Upsilon_{\mathcal{X}_{++}} (tops \ yss))
bots []
                         = [\cdot] * ([\cdot] * (segs xs))
bots [xs]
bots~(xss~++~yss)~=~((bots~xss)~\Upsilon_{\lambda(x,y).(++y)*x}~(bts~yss))~\Upsilon_{++}~(bots~yss)
tops []
                         = []
tops [xs]
                         = [\cdot] * ([\cdot] * (segs xs))
tops (xss ++ yss) = (tops xss) \Upsilon_{++} ((bts xss) \Upsilon_{\lambda(x,y).((x++)*y)} (tops yss))
b ts []
bts[xs]
                         = [\cdot] * (segs xs)
bts (xss ++ yss)
                         = (bts \ xss) \Upsilon_{++} (bts \ yss)
```

Fig. 3. Specification for mss2 Problem

shallow-grey rectangle) with every top-down rectangular subregion in yss (depicted by dark-grey rectangle) sharing the same edge.

Let's see the definition of the total order \prec' among rectangular subregions. Note that the index type Index' in this case should be a pair denoting the row and column of elements. So we define \prec' by $[[((r_1,c_1),x_1),\cdots],\cdots,[\cdots,((r_2,c_2),x_2)]] \prec' [[((r'_1,c'_1),y_1),\cdots],\cdots,[\cdots,((r'_2,c'_2),y_2)]] =_{def} [(r_1,c_1),(r_2,c_2)) <_D ((r'_1,c'_1),(r'_2,c'_2)].$

For other functions in Fig. 3, bots is used to calculate a list of lists each of which comprises all rectangles with the same bottom edge. Symmetrically, tops calculates a list of lists each of which comprises all rectangles with the same top edge. They are defined by using another function bts which yields a list of rectangles passing through the matrix vertically (Fig. 2 (e)).

It should be noted that segs, sum' and max' are in fact polymorphic functions over any index type. This is why we can use them in the definition of segs2 even though the index type is Index' instead of Index as in Fig. 3.

5.2 Derivation of an List Homomorphism for mss2

Our derivation of an almost homomorphism for mss2 from the specification in Fig. 3 is carried out according to the procedure in Section 4. First, we derive an almost homomorphism from the recursive definition of segs2. Then, we fuse (sum2*) with the derived almost homomorphism to obtain another almost homomorphism and again repeat this fusion for max'. Finally, assuming that we've got the almost list homomorphism $\pi_1 \circ ([f, \oplus])$ for mss2, we repeat the above procedure to find an efficient parallel implementation for f and \oplus .

Step 1: Deriving an almost homomorphism for segs2. We'd like to apply the tupling theorem for this derivation. First, we determine the functions that should

be tupled, similar as we did for segs in Section 4. From the definition of segs2:

$$segs2\ (xss\ ++\ yss) = \underline{segs2\ xss}\ ++_{\prec'} \underline{segs2\ yss}\ ++_{\prec'} concat((\underline{bots\ xss})\Upsilon_{\mathcal{X}_{\#}}\ (\underline{tops\ yss}))$$

we know that segs2 should be tupled with bots and tops, because segs2 and bots traverse over the same list xss whereas segs2 and tops traverse over the same list yss as underlined. Similarly, the definitions of bots and tops requires that bts be tupled with bots and tops. In summary, the functions to be tupled are segs2, bots, tops and bts, i.e., our tuple function will be:

$$segs2 \triangle bots \triangle tops \triangle bts.$$

Next, we rewrite the definition of each function in the above tuple to be in the form of (2), i.e., deriving f_1, \oplus_1 for segs2, f_2, \oplus_2 for bots, f_3, \oplus_3 for tops, and f_4, \oplus_4 for bts. This is straightforward. The results are as follows. For example, from the definition of segs2, we can easily derive that

Finally, we apply Theorem 1 and get the following list homomorphism.

$$segs2 \triangle bots \triangle tops \triangle bts = ([\Delta_1^4 f_i, \Delta_1^4 \oplus_i])$$

It follows that we have our almost homomorphism for segs2:

$$segs2 = \pi_1 \circ ([\Delta_1^4 f_i, \Delta_1^4 \oplus_i]).$$

Step 2: Fusion with almost homomorphisms. Recall that we have reached the point:

$$mss2 = max' \circ (sum2*) \circ (\pi_1 \circ ([\Delta_1^4 f_i, \Delta_1^4 \oplus_i]).$$

We proceed to fuse sum2* with $\pi_1 \circ ([\Delta_1^4 f_i, \Delta_1^4 \oplus_i])$ by Theorem 3, and then repeat this fusion for max', giving the following result.

$$mss2 = \pi_1 \circ ([\Delta_1^4 f_i', \Delta_1^4 \oplus_i']) \tag{7}$$

where

$$\begin{array}{l} (s_1,b_1,t_1,d_1) \oplus_1' (s_2,b_2,t_2,d_2) = s1 \uparrow s_2 \uparrow (\uparrow / (b_1 \Upsilon_{\mathcal{X}_+} \ t_2)) \\ (s_1,b_1,t_1,d_1) \oplus_2' (s_2,b_2,t_2,d_2) = (b_1 \Upsilon_+ \ d_2) \Upsilon_\uparrow \ b_2 \\ (s_1,b_1,t_1,d_1) \oplus_3' (s_2,b_2,t_2,d_2) = t_1 \Upsilon_\uparrow (d_1 \Upsilon_+ \ t_2) \\ (s_1,b_1,t_1,d_1) \oplus_4' (s_2,b_2,t_2,d_2) = d_1 \Upsilon_+ \ d_2 \end{array}$$

and

$$\begin{array}{ll} f_1' &= max' \circ (sum'*) \circ segs \\ f_2' &= (sum'*) \circ segs \\ f_3' &= (sum'*) \circ segs \\ f_4' &= (sum'*) \circ segs \end{array}$$

Step 3: Improving operators in list homomorphisms. Equation (7) has given a homomorphic solution to the 2-dimensional maximum segment sum problem. It is, however, not so obvious about efficient parallel implementation for f'_i . We need to repeat the Step 1 and 2 to derive true (almost) list homomorphisms for them. In fact, this has been done in Section 4 as given in Equation (5) and (6). It is not difficult to check that they $(f'_i s)$ can be parallelly implemented in $O(\log n)$ parallel time

Let n be the size of the input matrix. By a simple divide-and-conquer implementation of list homomorphisms, the derived program can expect a

$$max(C(\Delta_1^4 f_i'), (O(\log n) * C(\Delta_1^4 \oplus_i')))$$

parallel algorithm. With assumptions that Υ_{\otimes} and \mathcal{X}_{\otimes} can be implemented fully in parallel, i.e., $C(\Upsilon_{\otimes}) = C(\otimes)$ and $C(\mathcal{X}_{\otimes}) = C(\otimes)$, we can see that $C(\Delta_1^4 \oplus_i') = O(\log n)$ due to the inherited parallelism in the reduction $(\uparrow /)$. It follows that mss2 is a

$$max(C(\Delta_1^4 f_i'), O(\log^2 n))$$

parallel algorithm. We, therefore, obtain a $O(\log^2 n)$ parallel program for the 2-dimensional maximum segment sum problem.

6. CONCLUDING REMARKS

In this paper, we propose a formal and systematic approach to the derivation of efficient parallel programs from specifications of problems via *manipulation* of almost homomorphisms, namely the construction of almost list homomorphisms from recursive definitions (Theorem 1) and the fusion of a function with almost homomorphisms (Theorem 3). It is different from Cole's informal way[Cole 1993b].

We demonstrate our idea through the derivation of efficient parallel algorithms for several non-trivial problems. After the initial naive solution, all the derivation are proceeded in a formal setting based on our theorems and algebraic identities of list functions. Therefore, the resulting parallel algorithm is guaranteed to be semantically equivalent to the initial naive but inefficient solution. Furthermore, most of our derivation is mechanical, which would be expected to be used in a parallel compiler. As in Section 4.1, the derivation of almost homomorphisms from mutually-recursive defined functions is fully mechanical. What is difficult for being fully automatic is the fusion with almost homomorphism as shown in Section 4.2 where new functions have to be derived based on the equation (4) in the Almost Fusion Theorem. But some attempts have been made to make the fusion process automatic with some suitable restrictions as in [Gill, Launchbury, and Jones 1993; Takano and Meijer 1995; Hu, Iwasaki, and Takeichi 1996b].

Tupling and fusion are two well-known techniques for improving programs. Chin [Chin 1992; Chin 1993] gave an intensive study on it. His method tries to fuse and/or tuple arbitrary functions by fold-unfold transformations while keeping track of function calls and using clever control to avoid infinite unfolding. In contrast to his costly and complicated algorithm to keep out of non-termination, our approach makes use of structural knowledge of list homomorphisms and constructs our tupling and fusion rules in a calculational style where infinite unfoldings can be definitely avoided.

Our approach to the tupling of mutual recursive definitions is basically similar to the generalization algorithm [Takeichi 1987]. Takeichi showed how to define a higher order function common to all functions mutually defined so that multiple traversals of the same data structures in the mutual recursive definition can be eliminated. Because higher order functions are suitable for partial evaluation but not good for program derivation, we employ tupled functions and develop the corresponding fusion theorem. A similar idea to tupling can also be found in [Fokkinga 1992].

Construction of list homomorphisms has gained great interest because of its importance in parallel programming. Barnard et.al. [Barnard, Schmeiser, and Skillicorn 1991 applied the Third Homomorphism Theorem [Gibbons 1994] for the language recognition problem. The Third Homomorphism Theorem says that an algorithm h which can be formally described by two specific sequential algorithms (leftwards and rightwards reduction algorithms) is a list homomorphism. Although the existence of an associative binary operator is guaranteed, the theorem does not address the question of the existence - let alone the construction - of a direct and efficient way of calculating it. To solve this problem, Gorlatch [Gorlatch 1995 imposed additional restrictions, left associativity and right associativity, on the leftwards and rightwards reduction functions so that an associative binary operator \oplus could be derived in a systematic way. However, finding left-associative binary operators is usually not easier than finding associative operators. Recently Gorlatch [Gorlatch 1996a; Gorlatch 1996b] extended his previous work and proposed an idea of synthesizing list homomorphisms by generalizing both leftwards and rightwards reduction functions. Since his idea is studied in an informal way and the generalization algorithm is not given, it is not so clear how to do it in general. In comparison, rather than relying on the Third Homomorphism Theorem we construct list homomorphisms based on tupling and fusion transformation. Our derivation is more constructive: we derive list homomorphism directly from mutual recursive representations and then fuse it with other functions.

Smith [Smith 1987] applied a strategy of divide-and-conquer approach to both 1-dimensional and 2-dimensional mss problems as applications. He constructs the composing operator (analog to our associative operator \oplus) by employing the suitable mathematical properties of the problem. Although our initial specification is less abstract than his, our derivation is more systematic and less prone to errors. As seen in the paper, by our approach one could concisely derive a $O(\log^2 n)$ parallel program for the 2-dimensional mss problem. In Comparison, in [Smith 1987] the tuple consisting of eleven functions is given for the 2-dimensional mss problem but the corresponding manipulation with Smith's approach is not presented at all which must be cumbersome.

ACKNOWLEDGMENTS

Many thanks are to Akihiko Takano, Fer-Jan de Vries, Wei-Ngan Chin, Sergei Gorlatch for many enjoyable discussions. Thanks are also to the referees for their useful advice.

REFERENCES

Barnard, D., Schmeiser, J., and Skillicorn, D. 1991. Seriving associative operators for language recognition. In *Bulletin of EATCS* (43), pp. 131–139.

BIRD, R. 1984. The promotion and accumulation strategies in transformational programming.

- ACM Transactions on Programming Languages and Systems 6, 4, 487-504.
- BIRD, R. 1987. An introduction to the theory of lists. In M. BROY (Ed.), Logic of Programming and Calculi of Discrete Design, pp. 5-42. Springer-Verlag.
- Cai, W. and Skillicorn, D. 1992. Calculating recurrences using the Bird-Meertens Formalism.

 Technical report, Department of Computing and Information Science, Queen's University.
- CHIN, W. 1992. Safe fusion of functional expressions. In Proc. Conference on Lisp and Functional Programming, San Francisco, California.
- Chin, W. 1993. Towards an automated tupling strategy. In *Proc. Conference on Partial Eval*uation and *Program Manipulation*, Copenhagen, pp. 119–132. ACM Press.
- CHIN, W. 1996. Parallelizing conditional recurrences. In Annual European Conference on Parallel Processing, LNCS 1123, LIP, ENS Lyon, France, pp. 579-586. Springer-Verlag.
- Cole, M. 1993a. List homomorphic parallel algorithms for bracket matching. Technical report CSR-29-93 (Aug.), Department of Computing Science, The University of Edinburgh.
- Cole, M. 1993b. Parallel programming, list homomorphisms and the maximum segment sum problems. Report CSR-25-93 (May), Department of Computing Science, The University of Edinburgh.
- FEATHER, M. 1987. A survey and classification of some program transformation techniques. In *TC2 IFIP Working Conference on Program Specification and Transformation*, Bad Tolz (Germany), pp. 165–195. North Holland.
- FOKKINGA, M. 1992. A gentle introduction to category theory the calculational approach —.

 Technical Report Lecture Notes (Sept.), Dept. INF, University of Twente, The Netherlands.
- GIBBONS, J. 1994. The third homomorphism theorem. Technical report, University of Auckland.
- GILL, A., LAUNCHBURY, J., AND JONES, S. P. 1993. A short cut to deforestation. In Proc. Conference on Functional Programming Languages and Computer Architecture, Copenhagen, pp. 223-232.
- GORLATCH, S. 1995. Constructing list homomorphisms. Technical Report MIP-9512 (Aug.), Fakultät für Mathematik und Informatik, Universität Passau.
- GORLATCH, S. 1996a. Systematic efficient parallelization of scan and other list homomorphisms.
 In Annual European Conference on Parallel Processing, LNCS 1124, LIP, ENS Lyon, France, pp. 401–408. Springer-Verlag.
- GORLATCH, S. 1996b. Systematic extraction and implementation of divide-and-conquer parallelism. *Microprocessing and Microprogramming* 41, 571-578. (Also appears in PLILP'96).
- HU, Z., IWASAKI, H., AND TAKEICHI, M. 1996a. Construction of list homomorphisms via tupling and fusion. In 21st International Symposium on Mathematical Foundation of Computer Science, LNCS 1113, Cracow, pp. 407-418. Springer-Verlag.
- HU, Z., IWASAKI, H., AND TAKEICHI, M. 1996b. Deriving structural hylomorphisms from recursive definitions. In ACM SIGPLAN International Conference on Functional Programming, Philadelphia, PA, pp. 73-82. ACM Press.
- Hu, Z., IWASAKI, H., AND TAKEICHI, M. 1996c. Formal derivation of parallel program for 2dimensional maximum segment sum problem. In Annual European Conference on Parallel Processing, LNCS 1123, LIP, ENS Lyon, France, pp. 553-562. Springer-Verlag.
- Meijer, E., Fokkinga, M., and Paterson, R. 1991. Functional programming with bananas, lenses, envelopes and barbed wire. In *Proc. Conference on Functional Programming Languages and Computer Architecture* (LNCS 523), Cambridge, Massachuetts, pp. 124-144.
- Pettorossi, A. and Proietti, M. 1993. Rules and strategies for program transformation. In IFIP TC2/WG2.1 State-of-the-Art Report, pp. 263-303. LNCS 755.
- SMITH, D. 1987. Applications of a strategy for designing divide-and-conquer algorithms. Science of Computer Programming 9, 213-229.
- Takano, A. and Meijer, E. 1995. Shortcut deforestation in calculational form. In *Proc. Conference on Functional Programming Languages and Computer Architecture*, La Jolla, California, pp. 306–313.
- Takeichi, M. 1987. Partial parametrization eliminates multiple traversals of data structures. Acta Informatica 24, 57-77.