平面上の矩形和の最大値問題の 並列プログラムの導出

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List Homomorphism は、理想的には分割統治アルゴリズムに適しており、並列プログラミングの分野で最近注目を集めているが、従来の議論はアドホックに行なわれる傾向があった。本稿では、「平面上の矩形和の最大値問題」を例題として、素朴で効率の悪いプログラムから効率のよい並列プログラムを系統的・形式的に導出する方法を報告し、List Homomorphism の再帰的構造に基づいて定義されるTupling と Fusion というふたつの変換技法を提案する。

Formal Derivation of Parallel Program for 2-Dimensional Maximum Segment Sum Problem

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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 are usually treated rather informally and ad-hoc in the development of efficient parallel programs. This paper reports a case study on systematic and formal development of a new parallel program for the 2-dimensional maximum segment problem. We show how a straightforward, and "obviously" correct, but quite inefficient solution to the problem can be successfully turned into a semantically equivalent "almost list homomorphism". Our derivation is based on two transformations, namely tupling and fusion, which are defined according to the specific recursive structures of list homomorphisms.

1 Introduction

List homomorphisms 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,

$$h(xs + + ys) = hxs \oplus hys$$

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 pieces of the list. The computations of hxs and hys 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 of 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. For example, we have

$$mss [3, -4, 2, -1, 6, -3] = 7$$

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 [Col93] proposed an approach showing how to embed these functions into list homomorphisms in an ad hoc manner. His method consists of constructing a homomorphism as a tuple of functions where the original function is one of its components. The main difficulty is 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 program developer.

This paper reports results of a case study on formal and systematic derivation of a new efficient and correct $O(\log^2 n)$ (n denotes the number of elements in a matrix) parallel program for the 2-dimensional maximum segment sum problem. This problem is of interest because there are efficient but non-obvious

algorithms to compute it in parallel. In [Smi87], the tuple consisting of eleven functions is used for the definition of $O(log^2n)$ parallel algorithm but the detailed derivation, which would be rather cumbersome with Smith's approach, was not given at all.

This paper is organized as follows. In Section 2, we review the notational conventions and basic concepts used in this paper. After giving a specification for the 2-dimensional maximum segment sum problem in Section 3, we focus ourselves on deriving an efficient (almost) list homomorphism from the specification with our two important theorems, namely the Tupling and the Almost Fusion Theorems, in Section 4.

2 Preliminary

In this section, we briefly review notational conventions and the basic concepts in [Bir87], known as Bird-Meertens Formalism, as well as the concept of almost list homomorphism, which will be used in the rest of this paper.

2.1 Functions

Functional application is denoted by a space and the argument which may be written without brackets. Thus fa means f(a). Functions are curried and application associates to the left. Thus fab means (fa)b. Functional application is regarded as more binding than any other operator, so $fa \oplus b$ means $(fa) \oplus b$ but not $f(a \oplus b)$. Functional composition is denoted by a centralized circle \circ . By definition, $(f \circ g)a = f(ga)$. Functional composition is an associative operator, and the identity function is denoted by id.

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 $(f \triangle g) a = (f a, g a)$ and $(f \times g) (a, b) = (f a, g b)$. The \triangle can be naturally extended to functions with two arguments. So, we have $a(\bigoplus \triangle \boxtimes) b = (a \oplus b, a \boxtimes b)$.

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: $(a\oplus) b = a \oplus b = (\oplus b) a$.

2.2 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 $[\cdot]$ is its unit. For example, the term [1] + + [2] + + [3] denotes a list with three elements, often abbreviated to [1, 2, 3].

2.3 List Homomorphisms

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

It soon follows from this definition that \oplus must be an associative binary operator with unit ι_{\oplus} . For example, the functions sum is a list homomorphisms,

$$\begin{array}{lll} \max \left[\right] & = & -\infty \\ \max \left[x \right] & = & x \\ \max \left(xs + + ys \right) & = & \max xs \uparrow \max ys \end{array}$$

$$\begin{array}{lll} \sup \left[\right] & = & 0 \\ \sup \left[x \right] & = & x \\ \sup \left(xs + + ys \right) & = & \sup xs + \sup ys \end{array}$$

where \uparrow denotes the binary maximum function and $-\infty$ denotes a smallest value w.r.t. \uparrow . For notational convenience, we write (f, \oplus) for the unique function h^1 , e.g., sum = (id, +).

2.4 Parallelism: Map and Reduction

Map is the operator which applies another function to every item in a list. It is written as an infix *. Informally, we have

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

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 with unit ι_{\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 [Bir87]: $(f, \oplus) = (\oplus/) \circ (f*)$. Every list homomorphism can be written as the composition of a reduction and a map.

2.5 Almost Homomorphisms

As stated in Introduction, quite a lot of interesting functions are not list homomorphisms. Fortunately, Cole argued informally that some of them can be converted into so-called almost (list) homomorphisms by tupling them with some extra functions [Col93]. 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.

3 Specification

It is strongly advocated by Bird [Bir87] that specifications should be direct solutions to problems. Therefore, our specification for a problem p is a simple, and "obviously" correct, but possibly inefficient solution with the form of

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

where each p_i is a (recursively defined) function. It reflects our way of solving problems; a (big) problem p is likely to be solved through multiple passes in which a simpler problem p_i is solved by a recursion.

3.1 1-Dimensional Maximum Segment Sum Problem

Before giving a specification for the 2-dimensional maximum segment sum problem, let's start with a simpler 1-dimensional maximum segment sum problem *mss*, an example given in the introduction. An obviously correct solution to the problem is:

$$mss = max \circ (sum*) \circ segs \tag{2}$$

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.

The only unknown function in (2) is segs, which is to compute all (contiguous) segments of a list. It

Strictly 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 .

would be quite natural to give the following definition.

$$segs (xs ++ ys) = segs xs ++ segs ys ++ (tails xs \mathcal{X}_{++} inits ys).$$

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, the functions, such as inits, tails, and \mathcal{X}_{++} , are standard functions in [Bir87], though our definitions are slightly different as will be seen later.

Unfortunately, this 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]). In fact, the two resulting lists indeed consist of the same elements, but these elements are listed in different order. To solve this problem, we may impose the order \prec to the resulting list. Let $[x_{i_1}, x_{i_1+1}, \dots, x_{j_1}]$ and $[x_{i_2}, x_{i_2+1}, \dots, x_{j_2}]$ be segments of the presumed list $[x_1, x_2, \dots, x_n]$.

$$[x_{i_1}, x_{i_1+1}, \cdots, x_{j_1}] \prec [x_{i_2}, x_{i_2+1}, \cdots, x_{j_2}] =_{def}$$

$$(i_1, \cdots, j_1) <_D (i_2, \cdots, j_2)$$

where $<_D$ stands for the dictionary order. To simplify our presentation while capturing the index information as above, for the rest of this paper, we shall assume that each atomic element x (not a list) in the presumed list (i.e., input list) is a record with two fields: value field x.v and index field x.d. Under this assumption, we can redefine \prec by

$$[x_1, \cdots, x_m] \prec [y_1, \cdots, y_n] =_{def}$$
$$(x_1, d, \cdots, x_m, d) <_D (y_1, d, \cdots, y_n, d)$$

Note that generally x.d should be a n-tuple in case of n-dimensional structures. Furthermore, for notational convenience, we write x for x.v when no ambiguity happens.

Now our definition for segs is defined by

$$\begin{array}{rcl} segs \; [] & = \; [] \\ segs \; [x] & = \; [[x]] \\ segs \; (xs + + ys) & = \; segs \; xs + + \ \ \, segs \; ys + + \ \ \, \\ & \; (tails \; xs \; \mathcal{X}_{++} \; inits \; ys) \end{array}$$

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

To make this paper self-contained, we give the definitions for other functions. The *inits* is a function

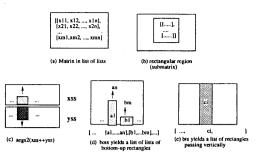


Figure 1: Explanation of specification mss2

returning all initial segments of a list, while the tails is a function returning all tail segments. They can be defined directly by:

$$\begin{array}{lll} inits \ [] & = & [] \\ inits \ [x] & = & [[x]] \\ inits \ (xs + + ys) & = & inits \ xs + + (xs + +) * (inits \ ys) \\ tails \ [] & = & [] \\ tails \ [x] & = & [[x]] \\ tails \ (xs + + ys) & = & (+ + ys) * (tails \ xs) + + tails \ ys. \\ \end{array}$$

The operator \mathcal{X}_{\oplus} is usually called *cross* operator, defined informally by $[x_1,\cdots,x_n]$ $\mathcal{X}_{\oplus}[y_1,\cdots,y_m]=[x_1\oplus y_1,\cdots,x_1\oplus y_m,\cdots,x_n\oplus y_1,\cdots,x_n\oplus y_m]$, which crosswisely combines elements in two lists with operator \oplus . The cross operator enjoys many algebraic identities, e.g., $(f*)\circ\mathcal{X}_{\oplus}=\mathcal{X}_{f\circ\oplus}$.

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

3.2 2-Dimensional Maximum Segment Sum 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 Figure 1 (a), and so does its rectangular subregion as in Figure 1 (b). Following the same thought we did for mss, we define mss2 straightforwardly as:

$$mss2 = max \circ (sum2*) \circ segs2$$

where 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.

Function sum2, computing the sum of a list of lists, is defined by $sum2 = sum \circ sum*$.

Function segs2, quite similar to that of segs, is defined below.

```
\begin{array}{lll} segs2 \ [] & = & [] \\ segs2 \ [xs] & = & [\cdot] * (segs \ xs) \\ segs2 \ (xss ++ yss) & = & segs2 \ xss ++_{\prec'} segs2 \ yss ++_{\prec'} \\ & & concat((bots \ xss)\Upsilon_{\mathcal{X}_{++}}(tops \ yss)) \end{array}
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The last equation reads that all rectangular subregions of xss + yss, a matrix connecting xss and yss vertically (see Figure 1 (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 shallow-grey rectangle) with every top-down rectangular subregion in yss (depicted by dark-grey rectangle) sharing the same edge. The order \prec' over rectangles is defined by

$$\begin{array}{ll} [[x_{11}, \cdots, x_{1m}], \cdots, [x_{n1}, \cdots, x_{nm}]] \prec' \\ [[y_{11}, \cdots, y_{1p}], \cdots, [y_{q1}, \cdots, y_{qp}]] &=_{def} \\ ((x_{11}.d, \cdots, x_{1m}.d), \cdots, (x_{n1}, d, \cdots, x_{nm}.d)) <_D \\ ((y_{11}.d, \cdots, y_{1p}.d), \cdots, (y_{q1}.d, \cdots, y_{qp}.d)) \end{array}$$

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 below, using another function bts which yields a list of rectangles passing through the matrix vertically (Figure 1 (e)).

$$\begin{array}{lll} bots \ [] & = & [] \\ bots \ [xs] & = & [\cdot] * ([\cdot] * (segs \ xs)) \\ bots \ (xss + + yss) & = & ((bots \ xss) \ \Upsilon_{\lambda(x,y).(+ \ y) * x} \\ & \ (bts \ yss)) \ \Upsilon_{+} \ (bots \ yss) \\ tops \ [] & = & [] \\ tops \ (xss + + yss) & = & ([\cdot] * ([\cdot] * (segs \ xs)) \\ & \ \Upsilon_{\lambda(x,y).((x+) * y)} \ (tops \ yss)) \\ bts \ [] & = & [] \\ bts \ [xs] & = & [\cdot] * (segs \ xs) \\ bts \ (xss + yss) & = & (bts \ xss) \ \Upsilon_{+} \ (bts \ yss) \\ \end{array}$$

concat, a function to flatten a list, and the zip-with operator Υ_{\oplus} , a function to apply \oplus pairwisely to two lists, are informally defined as follows.

$$concat [xs_1, \dots, xs_n] = xs_1 + + \dots + + xs_n$$
$$[x_1, \dots, x_n] \Upsilon_{\oplus} [y_1, \dots, y_n] = [x_1 \oplus y_1, \dots, x_n \oplus y_n]$$

4 Derivation

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

- Derive an almost homomorphism from the recursive definition of p₁ (Section 4.1);
- 2. Fuse p_2 with the derived almost homomorphism to obtain another almost homomorphism and repeat this fusion until p_n is fused (Section 4.2);

Let π₁ ∘ (f,⊕) be the result obtained in (2). If f or ⊕ are much complicated, repeat (1) and (2) to find an efficient parallel implementation for f and ⊕ (Section 4.3).

4.1 Deriving almost homomorphisms

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 mutually defined as follows.

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

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

$$h_{i}(xs ++ys) = ((\Delta_{1}^{n}h_{i})xs) \oplus_{i} ((\Delta_{1}^{n}h_{i})ys)$$
(3)

Then

$$\Delta_1^n h_i = (\![\Delta_1^n f_i, \, \Delta_1^n \oplus_i]\!)$$

and
$$(\iota_{\oplus_1}, \dots, \iota_{\oplus_n})$$
 is the unit of $\Delta_1^n \oplus_i$.

Proof: (omitted)

Theorem 1 says that if h_1 is mutually defined with other functions (i.e., $h_2, \cdots h_n$) traversing over the same lists in the specific form of (3), then tupling h_1, \cdots, h_n will give a list homomorphism. It follows that h_1 is an almost homomorphism: the projection function π_1 composed with the list homomorphism for the tuple-function. It is worth noting that this style of tupling can avoid multiple traversals of the same lists [Tak87] resulting in no repeatedly redundant computations of h_1, \cdots, h_n in the computation of the list homomorphism of $\Delta_1^n h_i$. That is, all previous computed results by h_1, \cdots, h_n can be fully reused, as expected in "true" almost homomorphisms.

Let's see how the tupling theorem is used in deriving a "true" almost homomorphism from the definition of segs2 given in Section 3.

First, we determine what functions are to be tupled, i.e., h_1, \dots, h_n . As the tupling theorem suggests, the functions to be tupled are those traversing over the same lists in the mutual definitions. So, from the definition of seas2:

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

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 (3), 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

$$\begin{array}{l} f_1 \ xs = [\cdot] * (segs \ xs) \\ (s_1,b_1,t_1,d_1) \oplus_1 (s_2,b_2,t_2,d_2) = \\ s_1 +_{\prec'} s_2 +_{+\prec'} concat \ (b_1 \Upsilon_{\mathcal{X}_{\#}} \ t_2) \\ f_2 \ xs = [\cdot] * ([\cdot] * (segs \ xs)) \\ (s_1,b_1,t_1,d_1) \oplus_2 (s_2,b_2,t_2,d_2) = \\ (b_1 \ \Upsilon_{\lambda(x,y),(+y)*x} \ d_2) \ \Upsilon_{\#} \ b_2 \\ f_3 \ xs = [\cdot] * ([\cdot] * (segs \ xs)) \\ (s_1,b_1,t_1,d_1) \oplus_3 (s_2,b_2,t_2,d_2) = \\ t_1 \ \Upsilon_{\#} \ (d_1 \ \Upsilon_{\lambda(x,y),((x\#)*y)} \ t_2) \\ f_4 \ xs = [\cdot] * (segs \ xs) \\ (s_1,b_1,t_1,d_1) \oplus_4 (s_2,b_2,t_2,d_2) = d_1 \ \Upsilon_{\#} \ d_2 \end{array}$$

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

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

And our almost homomorphism for segs2 is thus obtained:

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

4.2 Fusion with Almost Homomorphisms

In this section, we show how to fuse a function with an almost homomorphism. Our fusion theorem for this purpose is given below.

Theorem 2 (Almost Fusion) Let $(\Delta_1^n f_i, \Delta_1^n \oplus_i)$ and h be given. If there exist $\otimes_i (i = 1, \dots, n)$ and a map $\Lambda h = h_1 \times \dots \times h_n$ where $h_1 = h$ such that for all i,

$$\forall x, y. \ h_i(x \oplus_i y) = (\Lambda h) x \otimes_i (\Lambda h) y \tag{5}$$

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: (omitted)

We have two remarks on Theorem 2. First, this theorem suggests a rule 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 (5). Second, in order to simplify our presentation, without loss of generalization we restrict the projection function to be π_1 in the theorem.

Returning to our example, 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 can fuse sum2* with $\pi_1 \circ ([\Delta_1^4 f_i, \Delta_1^4 \oplus_i])$ by Theorem 2, and then repeat this fusion for max. And we can get the following result (see [HIT96] for detail).

$$mss2 = \pi_1 \circ \left(\left[\Delta_1^4 f_i^{\prime\prime}, \Delta_1^4 \otimes_i^{\prime} \right] \right) \tag{6}$$

where

$$\begin{array}{l} (s_1,b_1,t_1,d_1) \otimes_1' (s_2,b_2,t_2,d_2) = s1 \uparrow s_2 \uparrow (\uparrow/(b_1 \Upsilon_{\mathcal{X}_+} \ t_2)) \\ (s_1,b_1,t_1,d_1) \otimes_2' (s_2,b_2,t_2,d_2) = (b_1 \Upsilon_+ \ d_2) \Upsilon_\uparrow \ b_2 \\ (s_1,b_1,t_1,d_1) \otimes_3' (s_2,b_2,t_2,d_2) = t_1 \Upsilon_\uparrow \ (d_1 \Upsilon_+ \ t_2) \\ (s_1,b_1,t_1,d_1) \otimes_4' (s_2,b_2,t_2,d_2) = d_1 \Upsilon_+ \ d_2 \end{array}$$

and

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

4.3 Improving Operators in List Homomorphisms

Equation (6) has given a homomorphic solution to the 2-dimensional maximum segment sum problem. Let n be the number of elements of input matrix. By a simple divide-and-conquer implementation of list homomorphisms, the derived program can expect an $\max(O(\Delta_1^4f_i^{\prime\prime}), (O(\log n)*O(\Delta_1^4\otimes_i^{\prime})))$ parallel algorithm. With the assumption that Υ_{\oplus} can be implemented fully in parallel, i.e., $O(\Upsilon_{\oplus}) = O(\oplus)$, and that \mathcal{X}_{\oplus} with associative operator \oplus can be parallelly executed in $O(\log n)*O(\oplus)$, we can see that $O(\Delta_1^4\otimes_i^{\prime}) = O(\log n)$ and mss2 is an

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

parallel algorithm. It is, however, not so obvious about efficient parallel implementation of $f_i^{"}$. We can derive (almost) list homomorphisms for them using the above derivation strategy again (see [HIT96]).

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