

Minsegsumtwodim

Just when I had posed "minsegsum" to my students, Jon Bentley's article [0] appeared in the CACM. The article mentioned the analogous problem in two dimensions and more or less suggested that it was a difficult one. As it mentioned no complexity results for the two-dimensional problem, both Jayadev Misra and I started to think about it and arrived independently at the same solution. Since the solution is complicated enough to present a problem of presentation, I decided to devote a note to it.

Remark about notation. By way of experiment I shall denote functional application by an infix period; it has the highest binding power of all operators and is (as usual) left-associative. By way of further experiment, I shall also use it for subscription. (End of Remark about notation.)

The functional specification for minsegsumtwodim is

```

[[ M, N: int { M ≥ 1 ∧ N ≥ 1 }
; C(m, n: 0 ≤ m < M ∧ 0 ≤ n < N) array of int
; [[ x: int
; minsegsumtwodim
{R: x = (MIN m0, m1, n0, n1: 0 ≤ m0 ≤ m1 ≤ M ∧ 0 ≤ n0 ≤ n1 ≤ N:
(S m, n: m0 ≤ m < m1 ∧ n0 ≤ n < n1: C.m.n))}
]]
]]

```

in which (as usual) the constants of the environment are declared in the outer block.

To begin with we observe that  $R \Rightarrow x \leq 0$  since the summation may be over an empty range ( $m_0 = m_1 \vee n_0 = n_1$ ). Furthermore we observe that  $R$  is not changed if we change the range for  $m_0$  and  $m_1$  into  $0 \leq m_0 < m_1 \leq M$ , since then the summation over the empty rectangle is still included. We adopt this change. (Condition  $M \geq 1$  has been chosen, rather than  $M \geq 0$ , to make this change permissible;  $N \geq 1$  has been chosen because, in general, our solution is most attractive for  $N \geq M$ .)

Our next step is to rewrite  $R$  in terms of nested MIN's. (At this stage the reader is invited to convince himself of the correctness of my re-writing and urged to shelve for the time being the question why  $R$  is re-written this way.)

$$R: x = (\underline{\text{MIN}} m_0: 0 \leq m_0 < M: \\ (\underline{\text{MIN}} m_1: m_0 < m_1 \leq M: \\ (\underline{\text{MIN}} n_1: 0 \leq n_1 \leq N: MS.m_0.m_1.n_1)))$$

$$MS.m_0.m_1.n_1 =$$

$$(\underline{\text{MIN}} n_0: 0 \leq n_0 \leq n_1: (\underline{\text{S}} n: n_0 \leq n < n_1: Q.m_0.m_1.n))$$

$$Q.m_0.m_1.n = (\underline{\text{S}} m: m_0 \leq m < m_1: C.m.n)$$

Lines 2 through 4 of the above describe the linear minsegsum "in the  $n$ -direction" with  $m_0$  and  $m_1$  as parameters: the  $m$ -direction has been

"pushed to the sides", i.e. to the outer minimizations and the inner summation respectively.

We next derive -analogously to EWD897 - two relations for MS.

$$MS.m0.m1.0 = 0 \quad (0)$$

and for  $0 \leq n1 < N$

$$\begin{aligned} & MS.m0.m1.(n1+1) \\ &= \{ \text{definition of MS} \} \\ & (\underline{\text{MIN}} n0: 0 \leq n0 \leq n1+1: (\underline{\text{S}} n: n0 \leq n < n1+1: Q.m0.m1.n)) \\ &= \{ \text{for } n0 = n1+1, \text{ the summation yields } 0 \} \\ & (\underline{\text{MIN}} n0: 0 \leq n0 \leq n1: (\underline{\text{S}} n: n0 \leq n < n1+1: Q.m0.m1.n)) \underline{\text{min}} 0 \\ &= \{ \text{isolation of last term of summation} \} \\ & (\underline{\text{MIN}} n0: 0 \leq n0 \leq n1: (\underline{\text{S}} n: n0 \leq n < n1: Q.m0.m1.n) \\ & \quad + Q.m0.m1.n1) \underline{\text{min}} 0 \\ &= \{ Q.m0.m1.n1 \text{ does not depend on } n0 \text{ and} \\ & \quad \text{addition then distributes over minimization} \} \\ & ((\underline{\text{MIN}} n0: 0 \leq n0 \leq n1: (\underline{\text{S}} n: n0 \leq n < n1: Q.m0.m1.n)) \\ & \quad + Q.m0.m1.n1) \underline{\text{min}} 0 \\ &= \{ \text{definition of MS} \} \\ & (MS.m0.m1.n1 + Q.m0.m1.n1) \underline{\text{min}} 0 \quad . \quad (1) \end{aligned}$$

For Q we derive (directly from its definition)

$$Q.m0.m0.n1 = 0 \quad (2)$$

$$Q.m0.(m1+1).n1 = Q.m0.m1.n1 + C.m1.n1 \quad . \quad (3)$$

Our rewritten  $R$  tells us that for any  $m_0, m_1$  combination we have to find the minimum of  $MS.m_0.m_1.n_1$  for all  $n_1$ , while (1) gives a recurrence relation for that sequence of values. The snag is that that recurrence relation contains the term  $Q.m_0.m_1.n_1$ , which -see (3)- satisfies a recurrence relation over  $m_1$ . In order to exploit these two "orthogonal" recurrence relations to the fullest the program evaluates for each value of  $m_0$  the  $M-m_0$  recurrences (1) in synchrony. To this end a local array  $MS_v(m: m_0 < m \leq M)$  is introduced, such that whenever  $MS_v.m_1$  is adjusted, its value changes from  $MS.m_0.m_1.n_1$  to  $MS.m_0.m_1.(n_1+1)$ . Furthermore a local scalar  $Q_v$  is introduced such that whenever  $Q_v$  is adjusted, its value changes from  $Q.m_0.m_1.n_1$  to  $Q.m_0.(m_1+1).n_1$ .

The annotated program is given below. In the initialization of  $x$ ,  $x \leq 0$  is used; in its adjustment we use that min is associative.

Because we wish to use each  $MS$ -value as soon as it is computed, i.e. equals  $MS.m_0.m_1.(n_1+1)$ , line 2 of the rewritten  $R$  is rewritten once more:

$$0 \text{ min ( MIN } n_1: 0 \leq n_1 < N: MS.m_0.m_1.(n_1+1) ) ,$$

where the initial 0 can be taken out and taken care of by the initialization.

```

[[ m0: int ; x:=0 ; m0 := 0
; do m0 ≠ M →
  [[ MSv(m: m0 < m ≤ M) array of int
  ; m1, n1: int
  ; m1 := m0 ; do m1 ≠ M → m1 := m1 + 1 ; MSv.m1 := 0 od
  ; n1 := 0 { (∃ m: m0 < m ≤ M: MSv.m = MS.m0.m.n1) }
  ; do n1 ≠ N →
    [[ Qv: int ; Qv, m1 := 0, m0 { Qv = Q.m0.m1.n1 }
    ; do m1 ≠ M →
      Qv, m1 := Qv + C.m1.n1, m1 + 1
      { Qv = Q.m0.m1.n1 }
      ; MSv.m1 := (MSv.m1 + Qv) min 0
      ; x := x min MSv.m1
      od
    ] ] ; n1 := n1 + 1
    od
  ] ] ; m0 := m0 + 1
  od
] ]

```

---

\* \* \*

Side remark Eventually, we came up with a program that works for  $M=0 \vee N=0$  as well. We could have made our analysis also applicable in that case by defining the minimum of an empty set as "big enough" - 0, say - . (End of Side remark.)