#### CMPP2000 – Ponte de Lima - July 2000

## The Meta Transformation Tool for Skeleton-Based Languages

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## Skeletons

- Structured programming models
- Skeletons
  - $\rightarrow$  known, reusable, parallelism exploitation patterns
  - $\rightarrow$  Think to the analogy with seq. prog. (while do, for ...)
- Programmers concentrate on qualitative aspects of parallelism
- Tools deal with implementation details and quantitative aspects
  - ightarrow load-balancing, parallelism degree, messages size, etc.
- Specification of the software architecture

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# Why skeletons?

- Parallel programming is difficult and error prone
  - → Skeletons have a clear, functional and parallel semantic
- Performance portability
  - → Only performance can justify HPC high costs
  - ightarrow Performance heavily depends on the "matching" of the program with the architecture
- Several MPI implementation of the same algorithm

	seq	farm	farm + pipe	pipe $+ 2*farm$	pipe + farm
$T_s$ (sec)	6.03	0.39	1.30	0.72	4.99
$T_c$ (sec)	1207.76	84.50	286.62	151.67	1004.69
#PE	1	20	20	20	20
$\mathcal{E}(\%)$		75.52	23.08	41.93	6.04

(Aldinucci & Danelutto. IASTED PDCS'99, MIT, Boston, USA)

## Common skeletons

- Task parallel skeletons
  - parallelism in the computation of stream tasks
  - -E.g. pipeline, task farm, etc.
- Data parallel skeletons
  - parallelism in the computation of a single task
  - E.g. map (independent forall), reduce,  $\mathcal{D}\&\mathcal{C},$  etc.
- Sequential skeletons
  - the degenerate case: no parallelism at all
  - sometime used to wrap functions written in a guest language

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# The methodology

- Ingredients
  - A skeleton-based language
  - A cost calculus to foresee program performance
  - A set of semantic-preserving rewriting rules
- Methodology
  - 1. Write an initial specification/program
  - 2. Evaluate its performance
  - 3. Transform it (until the performance is satisfactory)
- Wish list
  - A compiler for the language
  - A tool to transform (optimize) programs

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## Outline

- Framework & Motivation
- $\rightarrow$  Outline
- The Meta transformation tool
  - A short introduction
  - Dealing with languages and rules
  - How it work
  - The architecture and the implementation
  - Running the tool
- Conclusions

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## The Meta transformation tool: What does it do?

### Given:

1. A Target Language (TL)



2. A set of (sound) rewriting rules  $(L \to R)$  for TL



3. A program written in TL

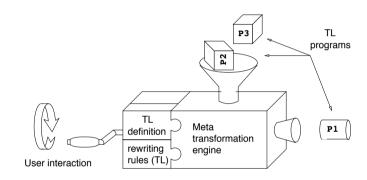
the Meta tool:

- locates applicable transformations
- provides performance estimates
- (possibly) transforms the program accordingly with the (user) chosen rule

Meta basically implements a (meta) term-rewriting system

# The Meta transformation tool: properties

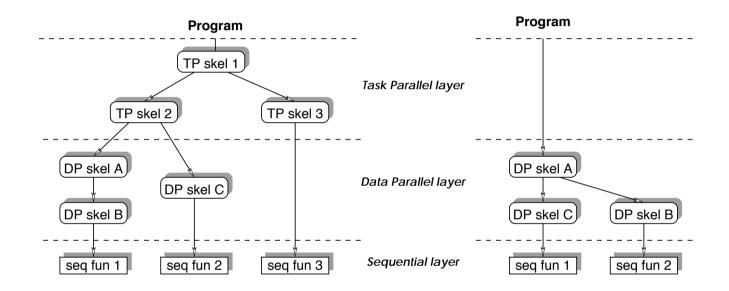
- Interactive
- Graphical
- Fast
- Language independent
- Rule independent
- Manages mixed data and task parallel languages



Meta can be instantiated with a broad class of TL (three-tier languages) and sets of rules for TL.

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# Skeleton-Based (up to) three-tier languages



Lower levels skeletons cannot call upper level ones

"The exploitation of task parallelism is (often) orthogonal with respect to the exploitation of data parallelism"

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## Three-tier languages: testbeds

### FAN: Functional Abstract Notation

(Aldinucci, Gorlatch, Lengauer, Pelagatti. Parallel Algorithms & Applications (to appear), Gordon & Breach.)

- Data parallel skeleton language
- FAN cost calculus

### Skel-BSP: Skeletons on top of the H-BSP

(Zavanella. Ph.D. Thesis, University of Pisa)

- Task and data parallel skeleton language
- Implemented on top of (hierarchical) BSP
- BSP-like cost calculus

Notice I don't present here neither new languages nor new rules Indeed, they have been presented and validated elsewhere

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# Three-tier languages: example Skel-BSP

```
TL\_prog ::= TP \mid DP
TP ::= \mathsf{farm} \; (TP) \mid \mathsf{pipe} \; \{TPlist\} \mid DP
DP ::= \mathsf{map} \; Seq \mid \mathsf{scanL} \; Seq \mid \mathsf{reduce} \; Seq \mid \mathsf{Seq} \mid \mathsf{comp} \; (\mathsf{out} \; \mathit{Var}, \; \mathsf{in} \; \mathit{Varlist}) \; \{\mathit{DPlist}\} \dots
Seq ::= \langle \; a \; \mathit{sequential} \; C \; \mathit{function} \; \rangle
\dots
\mathsf{comp}. name \; (\mathsf{out} \; \mathit{outvar}, \; \mathsf{in} \; \mathit{invars}) \{
\mathit{outvar}_1 = \mathsf{dp}. \; 1 \; Op_1 \; \mathit{invars}_1
\vdots
\mathit{outvar}_n = \mathsf{dp}. \; n \; Op_n \; \mathit{invars}_n \}
\mathsf{comp} \; \mathsf{definition} \; \mathsf{follows} \; \mathsf{the} \; \mathit{single-assignment} \; \mathsf{rule} \colon \; \mathsf{there} \; \mathsf{is} \; \mathsf{at}
```

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most one equation defining each variable. Skeletons into comp are

executed in sequence on a single set of PEs.

# Rewriting rules

A rule is a pair  $L \to R$  where

- L and R are fragments of TL programs with variables  $\nu_0, \nu_1, \ldots$
- $\bullet$   $\nu_0, \nu_1, \ldots$  act as placeholders for any piece of program
- ullet Every variable occurring in R must occur also in L
- L is not a variable

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# Rewriting rules: examples (1)

```
\mathsf{TSk} \quad \mathop{\leftarrow}\limits_{\leftarrow}^{\rightarrow} \quad \mathsf{farm} \; (\mathsf{TSk})
```

```
\begin{array}{ll} \textbf{pipe} \ \{ & \textbf{comp (out } z, \textbf{in } a) \ \{ \\ \textbf{DSk}_1 \ Op_1, & b = \textbf{DSk}_1 \ Op_1 \ a, \\ \textbf{DSk}_2 \ Op_2, & \rightarrow & c = \textbf{DSk}_2 \ Op_2 \ b, \\ < \cdots >_1 & < \cdots >_1 \\ \textbf{DSk}_n \ Op_n \} & z = \textbf{DSk}_n \ Op_n \ y \} \end{array}
```

A farm replicates TSk without changing the function it computes

Same functional (sequential) semantic. They differ in the execution model. pipe stages run on on different set of PEs, comp stages run in sequence on the same set of PEs

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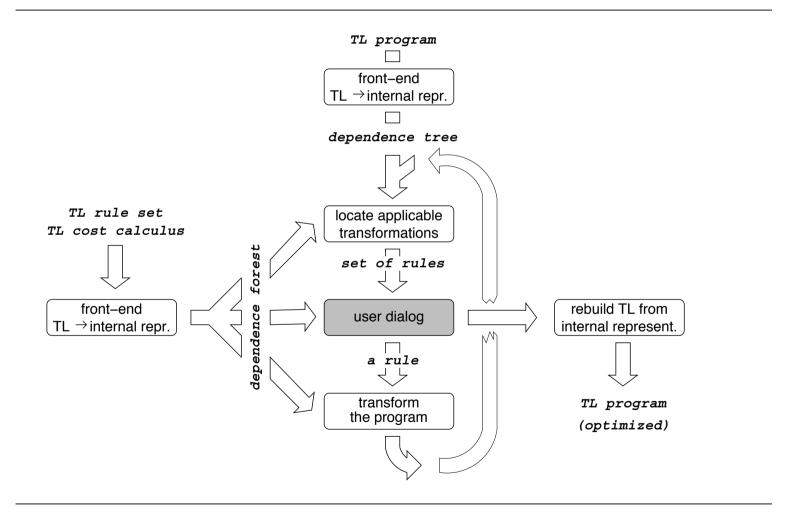
# Rewriting rules: examples (2)

```
comp (out outvar, in invars) { < \cdots >_1 q = \text{map } Op_1 \ p, < \cdots >_2 r = \text{map } Op_2 \ q, < \cdots >_3 } comp (out outvar, in invars) { < \cdots >_1 q = \text{map } Op_1 \ p, r = \text{map } Op_2 \ q, < \cdots >_2 < \cdots >_3 }
```

map (backwards) distribution through functional composition. We do not require the two maps to be adjacent in the program code. Meta provides the program with the additional assignment only if the intermediate result q is referenced into  $< \cdots >_2$  or  $< \cdots >_3$ 

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## Tool behaviour



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## Representing program and rules

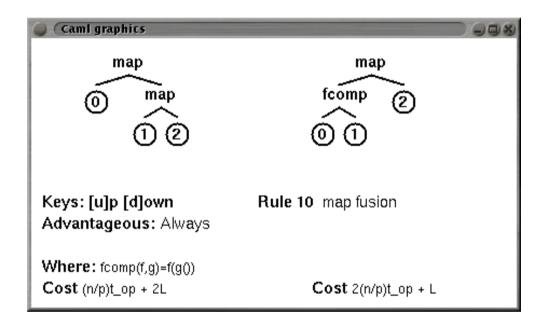
Both program and rules are represented by dependence trees

- A dependence tree (DT) is a labelled tree
- DT directly represents the data dependence among skeletons
- if Sk1 directly uses data produced by Sk2 then they are adjacent in DT
- DT is built starting from parse tree (PT) and data flow graph (DFG) of the program
- both PT and DFG can be build using standard tools
- Since program and patterns (L) are trees, the search for applicable rules reduces to subtree matching

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# Representing program and rules (cont'ed)

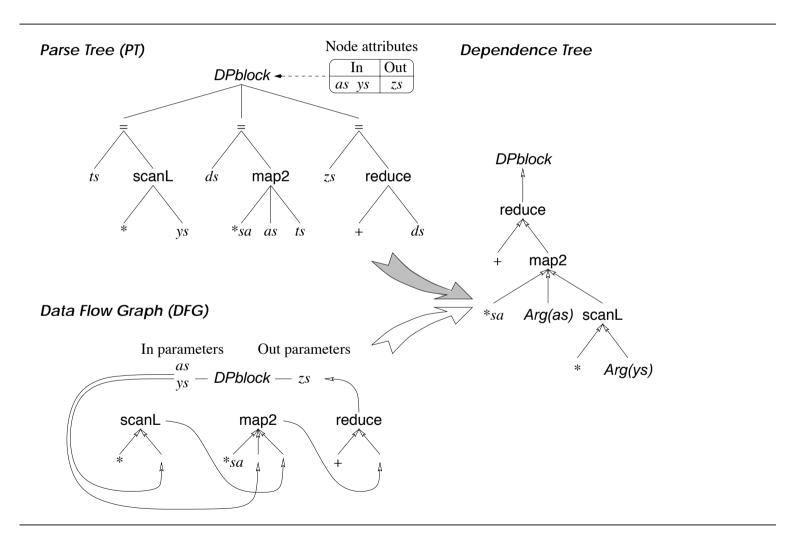
A rule is represented by a dependence tree pair



- Circled figures represents variables
- fcomp is special node representing functional composition

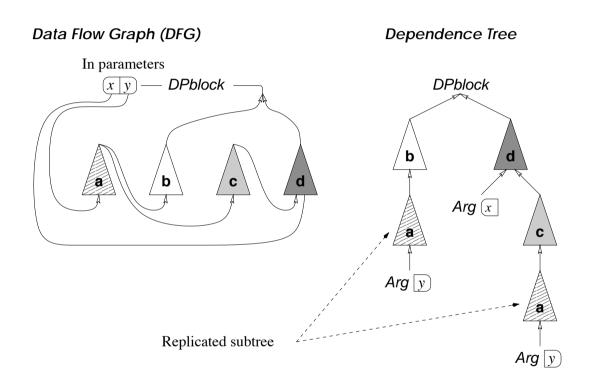
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# Building up the Dependence Tree



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# Building up the *Dependence Tree* (cont'ed)



shared subtrees: keep them shared or replicate them?

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# Locating applicable rules: subtree matching

Problem: Match a set of patterns against many subject trees

Solution: Hoffman-O'Donnell bottom-up algorithm:

Two phases: 1) preprocessing of the rules, 2) matching

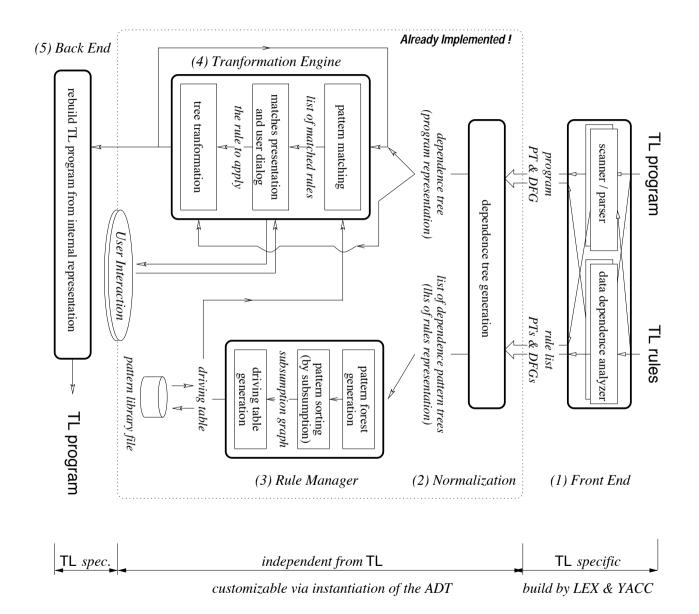
#### Good news:

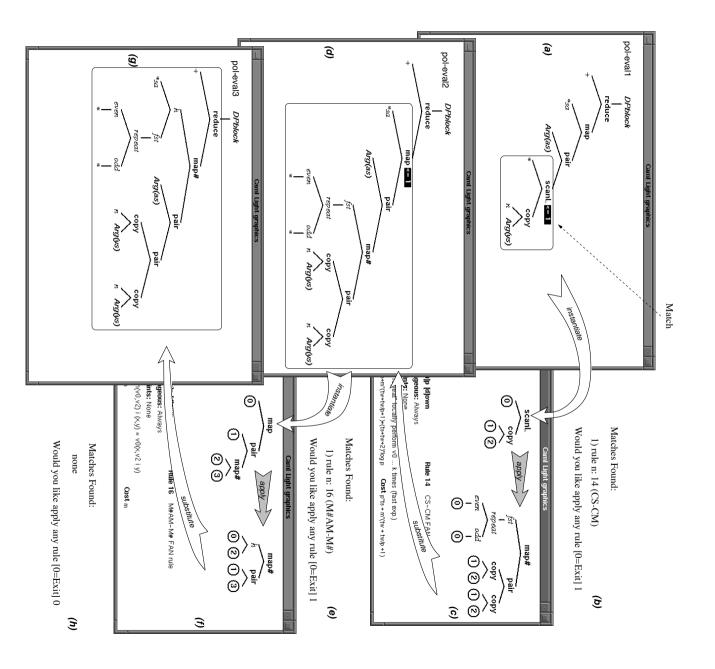
- The preprocessing phase have to be repeated only if either the rules or the language have been changed
- Matching really fast (even in practice): a single traversal of T

### Bad news:

- The preprocessing phase may be expensive
- but, it is fast for a broad class of pattern sets
- Details in the paper

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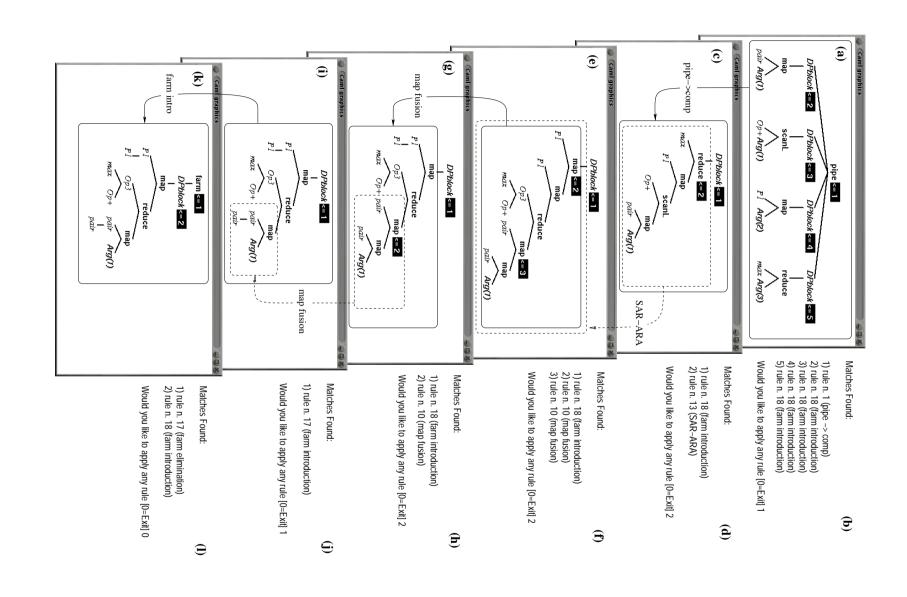




# MSS example in Skel-BSP (9 out of 20 formulations)

```
pipe.mss {
                                       pipe.mss {
                                                                                 pipe.mss {
                                        map pair,
 map pair,
                                                                                   map pair,
                           farm i/e
                                                                     farm i/e
                                        farm(scanL Op_+),
                                                                                   farm(scanL Op_{+}),
 scanL Op_{+},
 map P_1,
                                        map P_1,
                                                                                   farm(map P_1),
                                        reduce max }
 reduce max
                                                                                   reduce max }
      pipe→comp
                           \widehat{\mathsf{farm}}\ i/e
                                                                     farm i/e
comp.mss (out r, in x) {
                                       pipe.mss {
                                                                                  pipe.mss {
 y=map pair x,
                                        map pair,
                                                                                   map pair,
                                                                     farm i/e
 s=scanL Op_+ y,
                                        scanL Op_+,
                                                                                   scanL Op_+,
                                        farm(map P_1),
                                                                                   farm(map P_1),
 v = \text{map } P_1 s,
 r=reduce max v}
                                        reduce max }
                                                                                   farm(reduce max) \}
      SAR \frac{\uparrow}{\parallel} ARA
comp.mss (out r, in x) {
                                                                                  farm.mss (
                                       comp.mss (out r, in x) {
 a=map pair x,
                          map fusion
                                                                                   comp (out r, in x) {
 b=map pair a,
                                        a = \mathsf{map} \ (pair \circ pair) \ x
                                                                                    a=map (pair \circ pair) x,
 c=reduce
                                        b=reduce
                                                                   farm i/e
  Op_3(max, Op_+) b,
                                          Op_3(max, Op_+) a,
                                                                                    b=reduce
                                                                                      Op_3(max, Op_+) a
 d=map P_1 c,
                                        r=map (P_1 \circ P_1) b }
                          map fusion
                                                                                    r = map (P_1 \circ P_1) b \}
 r = map P_1 d
```

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# Implementation

- Prototype in Ocaml 2.02 (about 2000 lines of code)
- Tested under Windows'98 and Linux RH6.2
- Tested over 2 target languages and about 20 rules
- The implementation (except graphical interface) is based on a single ADT which describes the dependence tree and the function working on it
- The implementation can handle many Target Languages via instantiation of the ADT

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### Discussion

### 1. Why the tool is interactive?

→ Because the rewriting calculus of TL, in the general case, is not confluent in performance and the solution space (may) grow exponentially with the number rules

### 2. Does the tool make any decision about the rule to apply?

- → No. But, it can be extended with your own heuristics, if you have them
- → Currently Meta optimises Skel-BSP data-parallel-free programs with a standard sequence of rules. Such "Normal Form" is proved to be the fastest among the semantic-equivalent formulations that can be reached with these rules (Aldinucci, Danelutto. IASTED PDCS'99, Boston, USA)

### 3. Does the tool really optimise the program?

→ It really depend on the TL, the set of rule and the cost calculus, not on the tool. The tool make you happy playing with your new skeleton language, that's it.

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## Conclusions

Meta implements a (meta) rewriting system for skeleton-based languages

- It is independent from the target language and the rules
- It only requires a three-tier language schema
- Can be equipped with heuristics to make decisions on the rule to apply
- It is already implemented on a platform-independent language (Ocaml)
- It has a (simple) graphical interface and it is fast
- Is is easy to modify and to extend
- . . .
- It is free!

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# FAN rules (1)

#### Rule SR-ARA

```
b = reduce Op2 (scanL Op1 a)

b = proj1 (reduce Op3 (pair (a,a)))

If Op1 distributes forward over Op2

(a_1,b_1) Op3 (a_2,b_2) = (a_1 Op2 (b_1 Op1 a_2), b_1 Op2 b_2)
```

#### Rule AR-RA

```
b = \text{reduce } Op1 \text{ (proj1 } a)

b = \text{proj1 (reduce } Op2 \text{ } a)

(a_1, b_1) Op2 (a_2, b_2) = (a_1 Op1 a_2, b_1 Op1 b_2)
```

#### Rule SAR-ARA

```
c = \text{reduce } Op2 \text{ (proj1 (scanL } Op1 \text{ } a))
c = \text{proj1 (proj1 (reduce } Op3 \text{ (pair(} a,a))))}
If Op1 distributes forward over Op4
(a_1,b_1) Op3 (a_2,b_2) = (a_1 Op4 (b_1 Op1 a_2),b_1 Op4 b_2)
(a_1,b_1) Op4 (a_2,b_2) = (a_1 Op2 a_2,b_1 Op2 b_2)
```

#### Rule CS-CM

```
b = \operatorname{scanL} Op \ (\operatorname{copy} \ n \ a)
b = \operatorname{map}_{\#} f \ (\operatorname{copy} \ n \ a)
f \ i \ x = fst(repeat \ i \ (x, x))
repeat \ k \ x = \operatorname{if} \ k = 0 \ \operatorname{then} \ x \ \operatorname{else} \ repeat \ (k \ \operatorname{div} \ 2) \ (\operatorname{if} \ (k \ \operatorname{mod} \ 2 = 0) \ \operatorname{then} \ e \ x \ \operatorname{else} \ o \ x)
e(t, u) = (t, u \ Op \ u), \ o(t, u) = (t \ Op \ u, u \ Op \ u)
```

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# FAN rules (2)

#### Rule $M_{\#}M$ - $M_{\#}$

$b = map_\# \ f \ a$
$c = map\; g\; b$
$c = map_\# \ h \ a$
h i x = g (f i x)

#### Rule $M_{\#}AM$ - $M_{\#}$

$$b = \mathsf{map}_{\#} f \ a$$

$$c = \mathsf{map} \ g \ (\mathsf{pair} \ (d,b))$$

$$c = \mathsf{map}_{\#} \ h \ (\mathsf{pair} \ (d,a))$$

$$h \ i \ (x,y) = g \ (x, f \ i \ y)$$

FAN Operation	Time required
$map\ f\ x$	$m*t_f$
proj $1 x$	0
pair $(x,y)$	$2*m*t_{copy}$
copy $n x$	$p * t_s + \frac{m*(p-1)}{p} * t_w$
part $(r,s)$ $x$	$2*t_s + (r+s)*t_w$
reduce $(\oplus)$ $x$	$m * t_{\oplus} + \log p * (t_s + t_w + t_{\oplus})$
scanL $(\oplus)$ $x$	$2*m*t_{\oplus} + \log p*(t_s + t_w + 2*t_{\oplus})$

Rule	Time left hand side	Time right hand side	Improves if
SR-ARA	$3*m + \log p * (2*(t_s + t_w) + 3))$	$2*m + \log p*(t_s + 2*t_w + 2)$	always
AR-RA	$m + \log p * (t_s + t_w + 1)$	$2*m + \log p * (t_s + 2*(t_w + 1))$	never
SAR-ARA	$3*m + \log p * (2*(t_s + t_w) + 3)$	$5*m + \log p*(t_s + t_w + 1)$	$(t_s + t_w + 2) * \log p > 2m$
CS-CM	$p * t_s + m * (t_w + \frac{t_w}{p} + 1) + \log p * (t_s + t_w + 2)$	$p * t_s + m * (t_w + \frac{t_w}{n} + 1)$	always
	$+\log p*(t_s+t_w+2)$	r	
$M_{\#}M-M_{\#}$	2*m	m	always
$M_{\#}AM-M_{\#}$	4*m	m	always

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