

The BSP Model

In the BSP model [1], a computer is a set of p uniform processor-memory pairs and a communication network. A BSP program is executed as a sequence of *super-steps* (Fig. 1), each one divided into three successive disjoint phases:

- 1) Each processor only uses its local data to perform sequential computations and to request data transfers to other nodes;
- 2) The network delivers the requested data;
- 3) A global synchronisation barrier occurs, making the transferred data available for the next super-step.

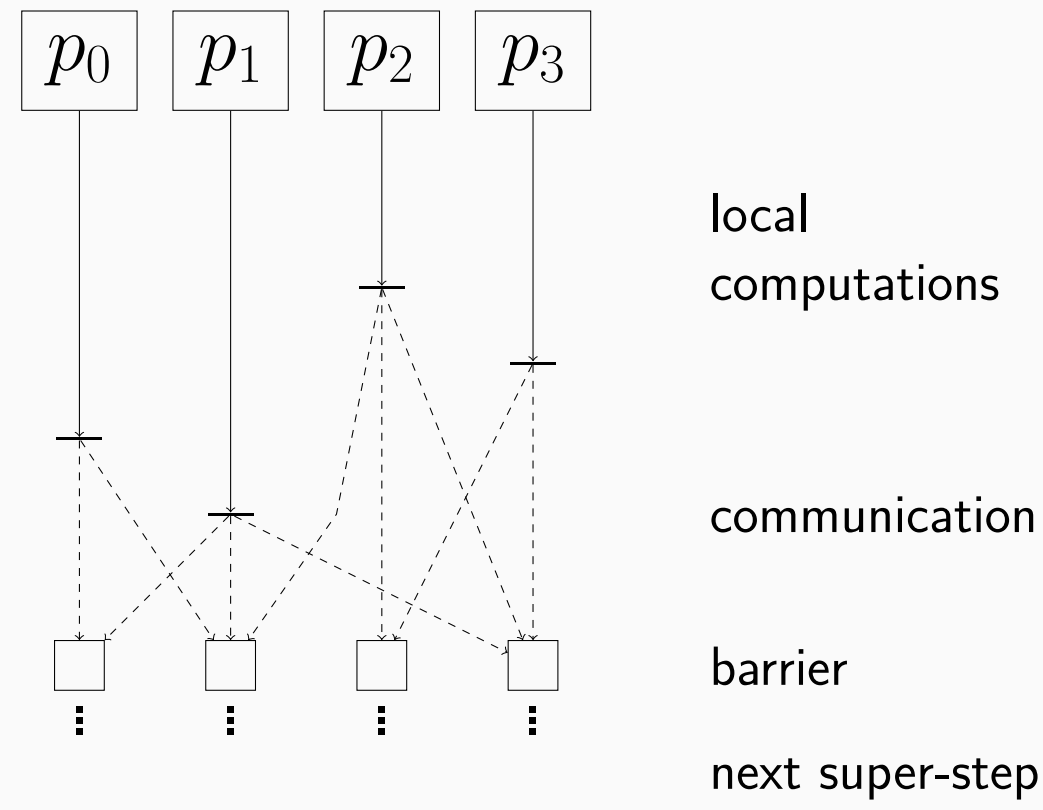


Figure 1: A BSP super-step

BSP Programming in ML : BSML

BSML [3] uses a *small set of primitives* and is currently implemented as a library for the ML programming language OCAML. A BSML program is built as a ML one but using a specific data structure called *parallel vector*. Its ML type is `'a par`. A vector expresses that each of the p processors *embeds* a value of any type `'a`. The BSML primitives are summarized in Fig. 6 :

Primitive	Level	Type	Informal semantics
<code><<e>></code>	g	<code>'a par</code> (if <code>e:'a</code>)	$\langle e, \dots, e \rangle$
<code>pid</code>	g	<code>int par</code>	A predefined vector: i on processor i
<code>\$v\$</code>	l	<code>'a</code> (if <code>v: 'a par</code>)	v_i on processor i , assumes $v \equiv \langle v_0, \dots, v_{p-1} \rangle$
<code>proj</code>	g	<code>'a par</code> \rightarrow <code>(int \rightarrow 'a)</code>	$\langle x_0, \dots, x_{p-1} \rangle \mapsto (\text{fun } i \rightarrow x_i)$
<code>put</code>	g	<code>(int \rightarrow 'a) par</code> \rightarrow <code>(int \rightarrow 'a) par</code>	$\langle f_0, \dots, f_{p-1} \rangle \mapsto \langle (\text{fun } i \rightarrow f_i 0), \dots, (\text{fun } i \rightarrow f_i (p-1)) \rangle$

Figure 6: The BSML primitives

An example of a parallel vector construction using the BSML toplevel :

```
#let vec = << "GDR" >> in << $vec$ ^ ",_proc_" ^ (string_of_int $pid$) >> ;;
val vec : string par = <"GDR,_proc_0", "GDR,_proc_1", "GDR,_proc_2">
```

The Multi-BSP Model

The MULTI-BSP model [2] is another *bridging model* as the original BSP, but adapted to *clusters of multicores*. The MULTI-BSP model introduces a vision where a *hierarchical architecture* is a *tree* structure of *nested components* (*sub-machines*) where the lowest stage (*leaf*) are processors and every other stage (*node*) contains memory. A node executes some codes on its nested components (*aka "children"*), then waits for results, do the communication and synchronised the sub-machine.

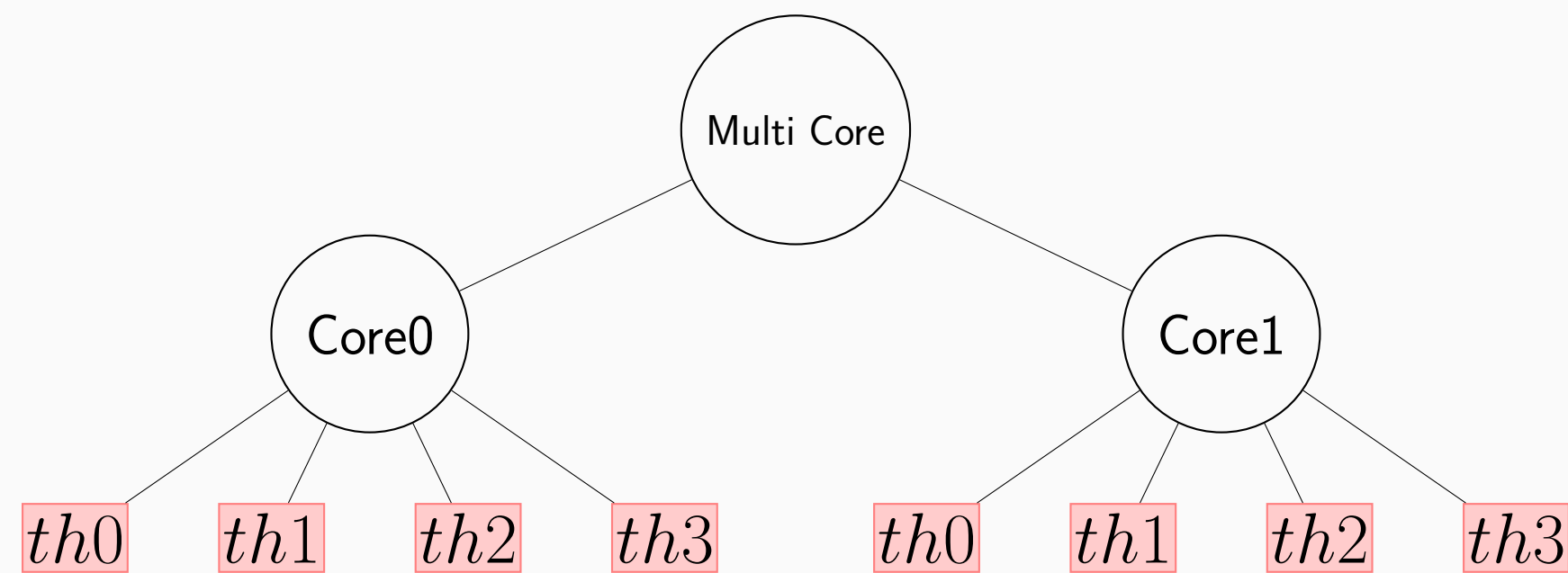


Figure 2: A MULTI-BSP view of a multi-core architecture

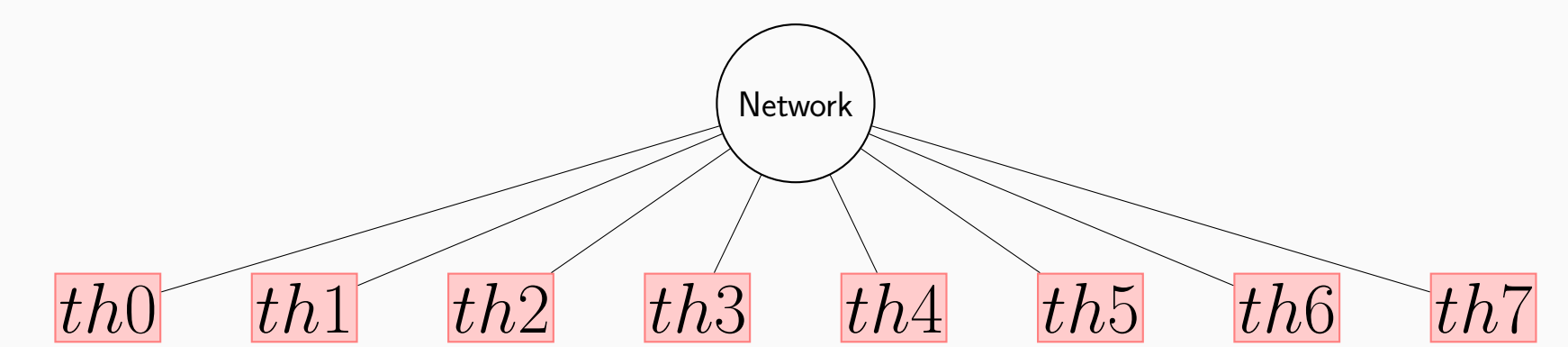


Figure 3: A BSP view of a multi-core architecture

For a multicore architecture it is possible to distinguish all the level thanks to MULTI-BSP (Fig. 2). On the contrary, the BSP model (Fig. 3) flattens the architecture.

Benchmarks

Fig. 4 shows the results of our experimentations. We can see that the efficiency on small list is poor but as the list grows, MULTI-ML exceeds BSML. This difference is due to the fact that BSML communicates through the network at every super steps; while MULTI-ML focusing on communications through local memories and finally communicates through the distributed level.

	100_000		1_000_000		3_000_000	
	MULTI-ML	BSML	MULTI-ML	BSML	MULTI-ML	BSML
8	0.7	1.8	125.3	430.7
16	0.5	0.8	68.1	331.5	1200.0	...
32	0.3	0.5	11.3	122.2	173.2	...
48	0.5	0.4	5.5	88.4	69.3	...
64	0.3	0.3	4.1	56.1	51.1	749.9
96	0.3	0.38	3.9	30.8	38.1	576.1
128	0.5	0.45	4.7	24.3	30.6	443.7

Figure 4: Execution time of Eratosthenes (naive) using MULTI-ML and BSML.

Fig. 5 gives the computation time of the simple scan using a summing operator. We can see that MULTI-ML introduce a small overhead due to the level management; however it is as efficient as BSML and concord to the estimated execution times.

	5_000_000			
	MULTI-ML	BSML	Pred MULTI-ML	Pred BSML
8	2.91	2.8	3.44	1.83
16	1.42	1.4	1.72	0.92
32	0.92	0.73	0.43	0.46
48	0.84	0.75	0.28	0.31
64	0.83	0.74	0.21	0.23

Figure 5: Execution time and predictions of scan (sum of integers)

The Multi-ML language

MULTI-ML is based on the idea of executing a BSML-like code on every stage of the MULTI-BSP architecture, that is on every sub-machine. For this, we add a *specific syntax* to ML in order to code special functions, called *multi-functions*, that recursively go through the MULTI-BSP tree. At each stage, a multi-function allows the execution of any BSML code. The main idea of MULTI-ML is to structure parallel codes to control all the stage of a tree: we generate the parallelism by allowing a node to call recursively a code on each of its sub-machines (children). When leaves are reached, they will execute their own codes and produce values, accessible by the top node using a vector. The data are distributed on the stages (toward leaves) and results are gathered on nodes toward the root node as shown in Fig. 7. Let us consider a code where, on a node, `<< e >>` is executed. As shown in Fig. 8, the node creates a vector containing, for each sub-machine i , the expression e . As the code is run asynchronously, the execution of the node code will continue until reaching a barrier.

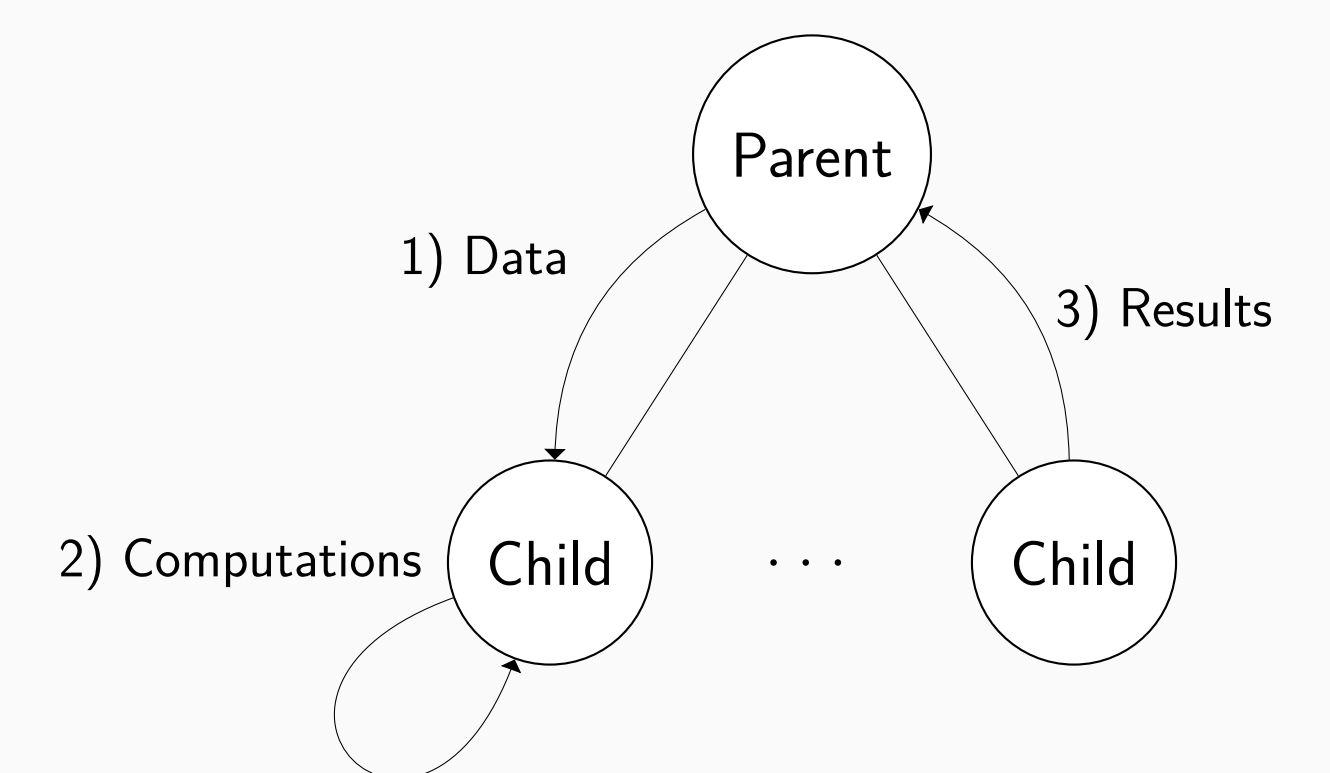


Figure 7: Code propagation

Fig. 9 shows the MULTI-ML primitives (without recall the BSML ones); their authorised level of execution and their informal semantics.

Primitive	Level	Type	Informal semantics
<code>\$e\$</code>	m	<code>'a tree</code>	Build $\lambda e \lambda$, a tree of e
<code>\$t\$</code>	s	<code>'a</code>	In a <code>\$e\$</code> code, t_n on node/leaf n of the tree λt
v (if $v: 'a tree$)	b	<code>'a</code>	v_n on node n of tree λv ,
<code>\$v\$</code>	l	<code>'a</code>	In the i th component of a vector, $v_{n,i}$ on node/leaf n of the tree λv
<code>gid</code>	m	<code>id</code>	The predefined tree of nodes and leaves ids
<code><<...f...>></code>	l	<code>'a</code>	In a component of a vector, recursive call of the multi-function
<code>#x#</code>	l	<code>'a</code>	In a component of a vector, reading the value x at upper stage (id)
<code>mkpar f</code>	b	<code>'a par</code>	$\langle v_0, \dots, v_p \rangle$, where $\forall i, f i = v_i$, at id n of the tree
<code>finally v1 v2</code>	b,s	<code>'a</code>	Return value v_1 to upper stage (id) and keep v_2 in the tree
<code>this</code>	b,l,s	<code>'a option</code>	Current value of the tree if exists, <code>None</code> otherwise

Figure 9: The MULTI-ML primitives

An example of a tree construction using the MULTI-ML toplevel :

```
#let multi f n =
  where node =
    let _ = <<f ($pid$ + #n# + 1)>> in
    finally ~up:() ~keep:(gid ^ "=" ^ n)
  where leaf = finally ~up:() ~keep:(gid ^ "=" ^ n);;
val f : int  $\rightarrow$  string tree = <multi-fun>
```

```
#f 0
o "0  $\rightarrow$  0"
o "0.0  $\rightarrow$  1"
  |  $\rightarrow$  "0.0.0  $\rightarrow$  2"
  |  $\rightarrow$  "0.0.1  $\rightarrow$  3"
o "0.1  $\rightarrow$  2"
  |  $\rightarrow$  "0.1.0  $\rightarrow$  3"
  |  $\rightarrow$  "0.1.1  $\rightarrow$  4"
```

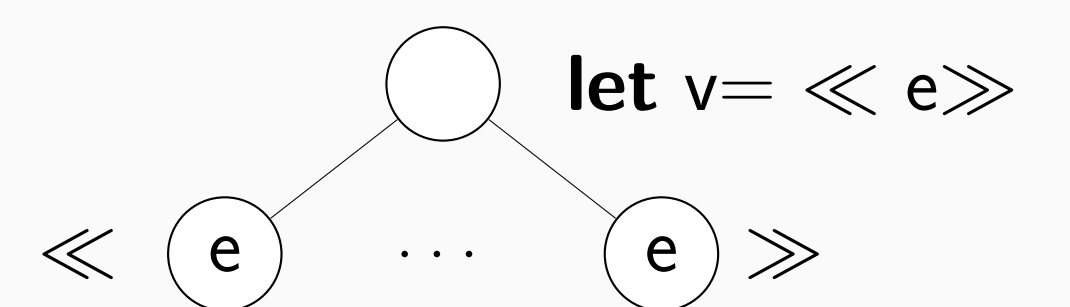


Figure 8: Data distribution

References

- [1] L. G. Valiant. "A Bridging Model for Parallel Computation". In: *Comm. of the ACM* 33.8 (1990), pp. 103–111.
- [2] L. G. Valiant. "A bridging model for multi-core computing". In: *J. Comput. Syst. Sci.* 77.1 (2011), pp. 154–166.
- [3] Louis Gesbert et al. "Bulk Synchronous Parallel ML with Exceptions". In: *Future Generation Computer Systems* 26 (2010), pp. 486–490.