# Divisible load theory

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# 1 The context

#### Context of the study

- Scientific computing : large needs in computation or storage resources.
- Need to use systems with "several processors":
  - Parallel computers with shared memory
  - Parallel computers with distributed memory
  - Clusters
  - Heterogeneous clusters
  - Clusters of clusters
  - Network of workstations
  - The Grid
- Problematic: to take into account the heterogeneity at the algorithmic level.

#### New platforms, new problems

Execution platforms: Distributed heterogeneous platforms (network of workstations, clusters, clusters of clusters, grids, etc.)

New sources of problems

- Heterogeneity of processors (computational power, memory, etc.)
- Heterogeneity of communications links.
- Irregularity of interconnection network.
- Non dedicated platforms.

We need to adapt our algorithmic approaches and our scheduling strategies : new objectives, new models, etc.

An example of application: seismic tomography of the Earth

- Model of the inner structure of the Earth

- The model is validated by comparing the propagation time of a seismic wave in the model to the actual propagation time.
- Set of all seismic events of the year 1999: 817101
- Original program written for a parallel computer:

```
if (rank = ROOT) raydata \leftarrow read n lines from data file; MPI_Scatter(raydata, n/P, ..., rbuff, ..., ROOT, MPI_COMM_WORLD); compute_work(rbuff);
```

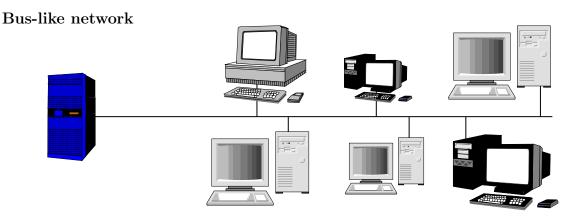
# Applications covered by the divisible loads model

Applications made of a very (very) large number of fine grain computations.

Computation time proportional to the size of the data to be processed.

Independent computations: neither synchronizations nor communications.

# 2 Bus-like network: classical resolution

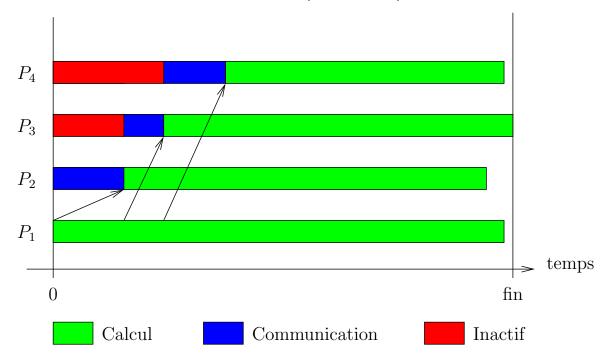


- The links between the master and the slaves all have the same characteristics.
- The slave have different computation power.

#### **Notations**

- A set  $P_1, ..., P_p$  of processors
- $-P_1$  is the master processor: initially, it holds all the data.
- The overall amount of work :  $W_{\text{total}}$ .
- Processor  $P_i$  receives an amount of work :  $n_i \in \mathbb{N}$  with  $\sum_i n_i = W_{\text{total}}$ . Length of a unit-size work on processor  $P_i$ :  $w_i$ . Computation time on  $P_i$ :  $n_i w_i$ .
- Time needed to send a unit-message from  $P_1$  to  $P_i$ : c. One-port bus:  $P_1$  sends a single message at a time over the bus, all processors communicate at the same speed with the master.

# Behavior of the master and of the slaves (illustration)



# Behavior of the master and of the slaves (hypotheses)

- The master sends its chunk of  $n_i$  data to processor  $P_i$  in a single sending.
- The master sends their data to the processors, serving one processor at a time, in the order  $P_2, ..., P_p$ .
- During this time the master processes its  $n_1$  data.
- A slave does not start the processing of its data before it has received all of them.

# **Equations**

 $-P_1:T_1=n_1.w_1$ 

 $- P_2: T_2 = n_2.c + n_2.w_2$ 

 $-P_3: T_3 = (n_2.c + n_3.c) + n_3.w_3$   $-P_i: T_i = \sum_{j=2}^{i} n_j.c + n_i.w_i \text{ for } i \ge 2$   $-P_i: T_i = \sum_{j=1}^{i} n_j.c_j + n_i.w_i \text{ for } i \ge 1 \text{ with } c_1 = 0 \text{ and } c_j = c \text{ otherwise.}$ 

# Execution time

$$T = \max_{1 \le i \le p} \left( \sum_{j=1}^{i} n_j \cdot c_j + n_i \cdot w_i \right)$$

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We look for a data distribution  $n_1, ..., n_p$  which minimizes T.

# Execution time: rewriting

$$T = \max \left( n_1.c_1 + n_1.w_1, \max_{2 \le i \le p} \left( \sum_{j=1}^{i} n_j.c_j + n_i.w_i \right) \right)$$

$$T = n_1.c_1 + \max\left(n_1.w_1, \max_{2 \le i \le p} \left(\sum_{j=2}^{i} n_j.c_j + n_i.w_i\right)\right)$$

An optimal solution for the distribution of  $W_{\text{total}}$  data over p processors is obtained by distributing  $n_1$  data to processor  $P_1$  and then optimally distributing  $W_{\text{total}} - n_1$  data over processors  $P_2$  to  $P_p$ .

# Algorithm

```
1: solution[0, p] \leftarrow cons(0, NIL); cost[0, p] \leftarrow 0
 2: for d \leftarrow 1 to W_{\text{total}} do
        solution[d,p] \leftarrow \operatorname{cons}(d,\mathit{NIL})
        cost[d, p] \leftarrow d \cdot c_p + d \cdot w_p
 5: for i \leftarrow p-1 downto 1 do
        solution[0, i] \leftarrow cons(0, solution[0, i + 1])
 7:
         cost[0,i] \leftarrow 0
        for d \leftarrow 1 to W_{\text{total}} do
 8:
 9:
            (sol, min) \leftarrow (0, cost[d, i+1])
10:
            for e \leftarrow 1 to d do
               m \leftarrow e \cdot c_i + \max(e \cdot w_i, cost[d - e, i + 1])
11:
12:
               if m < min then
13:
                   (sol, min) \leftarrow (e, m)
14:
            solution[d, i] \leftarrow cons(sol, solution[d - sol, i + 1])
15:
            cost[d, i] \leftarrow min
16: return (solution[W_{total}, 1], cost[W_{total}, 1])
```

#### Complexity

- Theoretical complexity

$$O(W_{\mathrm{total}}^2 \cdot p)$$

Complexity in practice

If  $W_{\text{total}} = 817101$  and p = 16, on a Pentium III running at 933 MHz : more than two days...

(Optimized version ran in 6 minutes.)

#### Disadvantages

- Cost
- Solution is not reusable
- Solution is only partial (processor order is fixed)

We do not need the solution to be so precise

# 3 Bus-like network : resolution under the divisible load model

#### **Notations**

- A set  $P_1, ..., P_p$  of processors
- $P_1$  is the master processor : initially, it holds all the data.
- The overall amount of work :  $W_{\text{total}}$ .
- Processor  $P_i$  receives an amount of work  $\alpha_i W_{\text{total}}$  with  $\alpha_i W_{\text{total}} \in \mathbb{Q}$  and  $\sum_i \alpha_i = 1$ .

Length of a unit-size work on processor  $P_i: w_i$ .

- Computation time on  $P_i : \alpha_i w_i$ .
- Time needed to send a unit-message from  $P_1$  to  $P_i$ : c. One-port model:  $P_1$  sends a *single* message at a time, all processors communicate at the same speed with the master.

# **Equations**

For processor  $P_i$  (with  $c_1 = 0$  and  $c_j = c$  otherwise):

$$T_i = \sum_{j=1}^{i} \alpha_j W_{\text{total}}.c_j + \alpha_i W_{\text{total}}.w_i$$

$$T = \max_{1 \le i \le p} \left( \sum_{j=1}^{i} \alpha_j W_{\text{total}}.c_j + \alpha_i W_{\text{total}}.w_i \right)$$

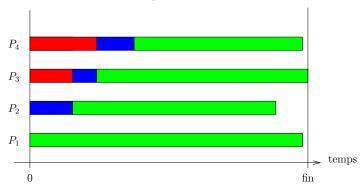
We look for a data distribution  $\alpha_1, ..., \alpha_p$  which minimizes T.

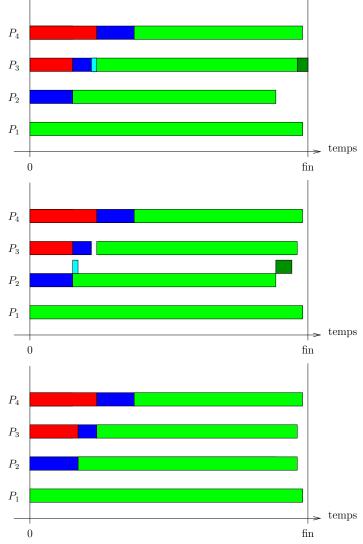
# Properties of load-balancing

**Lemma:** In an optimal solution, all processors end their processing at the same time.

#### Demonstration of lemma 1

Two slaves i and i + 1 with  $T_i < T_{i+1}$ . (the same results holds with i - 1 and i)





We decrease  $\alpha_{i+1}$  by  $\epsilon$ . We increase  $\alpha_i$  by  $\epsilon$ . The communication time for the following processors is unchanged. We end up with a better solution!

- Ideal: 
$$T'_{i} = T'_{i+1}$$
.

We choose  $\epsilon$  such that :

$$(\alpha_i + \epsilon)W_{\text{total}}(c + w_i) =$$

$$(\alpha_i + \epsilon)W_{\text{total}}(c + (\alpha_{i+1} - \epsilon)W_{\text{total}}(c + w_{i+1})$$

- The master stops before the slaves : absurd.
- The master stops after the slaves : we decrease  $P_1$  by  $\epsilon$ .

#### Property for the selection of resources

**Lemma:** In an optimal solution all processors work.

Demonstration: this is just a corollary of lemma 1...

# Resolution for a given ordering of the workers

$$T = \alpha_1 W_{\text{total}} w_1.$$

$$T = \alpha_2(c + w_2)W_{\text{total}}$$
. Therefore  $\alpha_2 = \frac{w_1}{c + w_2}\alpha_1$ .

$$T = (\alpha_2 c + \alpha_3 (c + w_3)) W_{\text{total}}$$
. Therefore  $\alpha_3 = \frac{w_2}{c + w_3} \alpha_2$ .

$$\alpha_i = \frac{w_{i-1}}{c+w_i} \alpha_{i-1} \text{ for } i \ge 2.$$

$$\sum_{i=1}^{n} \alpha_i = 1.$$

$$\alpha_1 \left( 1 + \frac{w_1}{c + w_2} + \dots + \prod_{k=2}^j \frac{w_{k-1}}{c + w_k} + \dots \right) = 1$$

# Impact of the order of communications

How important is the influence of the ordering of the processor on the solution?

Consider the volume processed by processors  $P_i$  and  $P_{i+1}$  during a time T.

**Processor** 
$$P_i$$
:  $\alpha_i(c+w_i)W_{\text{total}} = T$ . Therefore  $\alpha_i = \frac{1}{c+w_i}\frac{T}{W_{\text{total}}}$ .

Processor 
$$P_{i+1}$$
:  $\alpha_i c W_{\text{total}} + \alpha_{i+1} (c + w_{i+1}) W_{\text{total}} = T$ .  
Thus  $\alpha_{i+1} = \frac{1}{c + w_{i+1}} (\frac{T}{W_{\text{total}}} - \alpha_i c) = \frac{w_i}{(c + w_i)(c + w_{i+1})} \frac{T}{W_{\text{total}}}$ .

Processors  $P_i$  and  $P_{i+1}$ :

$$\alpha_i + \alpha_{i+1} = \frac{c + w_i + w_{i+1}}{(c + w_i)(c + w_{i+1})}$$

# No impact of the order of the communications

#### Choice of the master processor

We compare processors  $P_1$  and  $P_2$ .

**Processor** 
$$P_1: \alpha_1 w_1 W_{\text{total}} = T$$
. Then,  $\alpha_1 = \frac{1}{w_1} \frac{T}{W_{\text{total}}}$ .

**Processor** 
$$P_2: \alpha_2(c+w_2)W_{\text{total}} = T$$
. Thus,  $\alpha_2 = \frac{1}{c+w_2}\frac{T}{W_{\text{total}}}$ .

Total volume processed:

$$\alpha_1 + \alpha_2 = \frac{c + w_1 + w_2}{w_1(c + w_2)} = \frac{c + w_1 + w_2}{cw_1 + w_1w_2}$$

Minimal when  $w_1 < w_2$ .

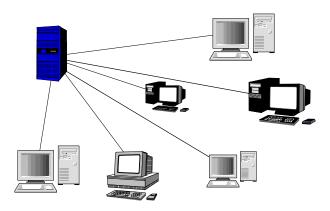
Master = the most powerful processor (for computations).

#### Conclusion

- Closed-form expressions for the execution time and the distribution of data.
- Choice of the master.
- The ordering of the processors has no impact.
- All processors take part in the work.

# 4 Star-like network

#### Star-like network



- The links between the master and the slaves have different characteristics.
- The slaves have different computational power.

#### **Notations**

- A set  $P_1, ..., P_p$  of processors
- $-P_1$  is the master processor : initially, it holds all the data.
- The overall amount of work :  $W_{\rm total}$ .
- Processor  $P_i$  receives an amount of work  $\alpha_i W_{\text{total}}$ , with  $\alpha_i W_{\text{total}} \in \mathbb{Q}$  and  $\sum_i \alpha_i = 1$ .
- Length of a unit-size work on processor  $P_i : w_i$ .
- Computation time on  $P_i : \alpha_i w_i$ .
- Time needed to send a unit-message from  $P_1$  to  $P_i : c_i$ .
- One-port model:  $P_1$  sends a *single* message at a time.

#### Resource selection

**Lemma:** In an optimal solution, all processors work.

We take an optimal solution. Let  $P_k$  be a processor which does not receive any work: we put it last in the processor ordering and we give it a fraction  $\alpha_k$  such that  $\alpha_k(c_k + w_k)W_{\text{total}}$  is equal to the processing time of the last processor which received some work.

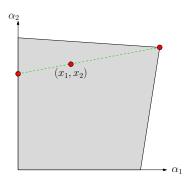
Why should we put this processor last?

## Load-balancing property

**Lemma:** In an optimal solution, all processors end at the same time.

- Most existing proofs are false.

$$\begin{aligned} & \text{Minimize } T, \\ & \text{Subject to} \\ & \left\{ \begin{array}{l} \sum_{i=1}^{n} \alpha_i = 1 \\ \forall i, \quad \alpha_i \geq 0 \\ \forall i, \quad \sum_{k=1}^{i} \alpha_k c_k + \alpha_i w_i \leq T \end{array} \right. \end{aligned}$$



- The constraints define a polyhedron
- One of the optimal solution is a vertex of the polyhedron, that is at least n among the 2n inequalities are equalities,
- It can not be a lower bound, because all processors participate, thus the only vertex with all  $\alpha_i > 0$  is an optimal solution
- Assume there is another optimal solution; it lies within the polyhedron
- We can build a linear combination of both optimal solution (with optimal objective), such that one variable is zero, which contradicts the previous result.

# Impact of the order of communications

Volume processed by processors  $P_i$  and  $P_{i+1}$  during a time T.

**Processor** 
$$P_i$$
:  $\alpha_i(c_i + w_i)W_{\text{total}} = T$ . Thus,  $\alpha_i = \frac{1}{c_i + w_i} \frac{T}{W_{\text{total}}}$ .

Processor 
$$P_{i+1}$$
:  $\alpha_i c_i W_{\text{total}} + \alpha_{i+1} (c_{i+1} + w_{i+1}) W_{\text{total}} = T$ .  
Thus,  $\alpha_{i+1} = \frac{1}{c_{i+1} + w_{i+1}} (1 - \frac{c_i}{c_i + w_i}) \frac{T}{W_{\text{total}}} = \frac{w_i}{(c_i + w_i)(c_{i+1} + w_{i+1})} \frac{T}{W_{\text{total}}}$ .

Volume processed : 
$$\alpha_i + \alpha_{i+1} = \frac{c_{i+1} + w_i + w_{i+1}}{(c_i + w_i)(c_{i+1} + w_{i+1})}$$

Communication time: 
$$\alpha_i c_i + \alpha_{i+1} c_{i+1} = \frac{c_i c_{i+1} + c_{i+1} w_i + c_i w_{i+1}}{(c_i + w_i)(c_{i+1} + w_{i+1})}$$

Processors must be served by decreasing bandwidths.

#### Conclusion

- The processors must be ordered by decreasing bandwidths
- All processors are working
- All processors end their work at the same time
- Formulas for the execution time and the distribution of data

# 5 Conclusion

What should we remember from that?

- A simple, approximate, model is sometimes better than a complete, but intractable one.
- In such a configuration (bus network, one-port model), communication times are more important than computation speed.
- Always start with a very simple model, (solve it) and then make it more complex if possible.