Parallel computing models and their performances

A high level exploration of the HPC world

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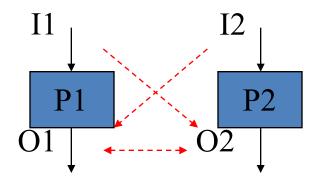
Overview

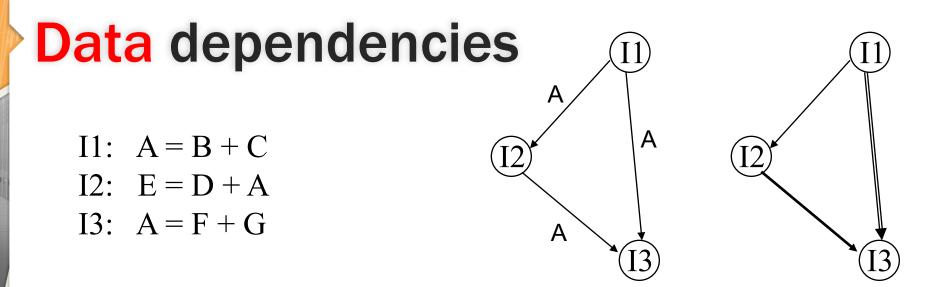
- Definition of parallel application
- Architectures taxonomy
- What is quantifiable ? Laws managing the parallel applications field
- Modeling performance of parallel applications

Formal definition of parallelism

The Bernstein Conditions Let's define:

- I(P) all variables, registers and memory locations used by P
- O(P) all variables, registers and memory locations written by P Then P1; P2 is equivalent to P1 || P2 if and only if {I(P1) ∩ O(P2) = Ø & I(P2) ∩ O(P1) = Ø & O(P1) ∩ O(P2) = Ø} General case: P1... Pn are parallel if and only if each for each pair Pi, Pj we have Pi || Pj.
- 3 limit to the parallel applications:
 - 1. Data dependencies
 - 2. Flow dependencies
 - 3. Resources dependencies

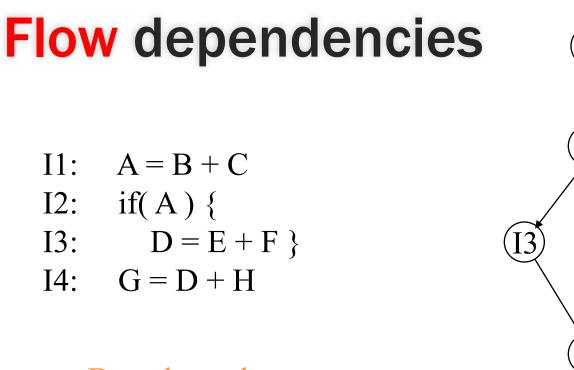




- Flow dependency (RAW): a variable assigned in a statement is used in a later statement
- Anti-dependency (WAR): a variable used in a statement is assigned in a subsequent statement
- Output dependency (WAW): a variable assigned in a statement is subsequently re-assigned

How to avoid them?

Which type of data dependency can be avoided ?



- Data dependency
- Control dependency

How to avoid ?

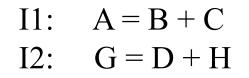


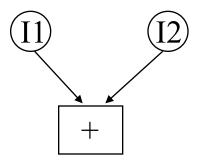
I3

I1

[4

Resources dependencies





How to avoid ?



A more complicated example (loop)

A[9] = A[8]

for i = 0 to 9 A[i] = B[i]

All statements are independent, as they relate to different data. They are concurrent. for i = 1 to 9 A[i] = A[i-1]

 $A[1] = A[0] \quad All \text{ statements} \\ A[2] = A[1] \quad are \text{ dependent}, \\ A[3] = A[2] \quad as every 2 \\ A[3] = A[2] \quad as every 2 \\ A[4] = A[3] \quad statements \text{ are} \\ A[4] = A[3] \quad statements \text{ are} \\ strictly \\ sequential. \\ A[6] = A[5] \\ A[7] = A[6] \\ A[8] = A[7] \\ A[8] = A[7] \\ All statements \\ A$

Flynn Taxonomy (1966)

- Computers classified by instruction delivery mechanism and data stream(s)
 - I for instruction, P for program. Conceptually similar, technically at a different granularity.
- 4 characters code: 2 for instruction stream and 2 for data stream

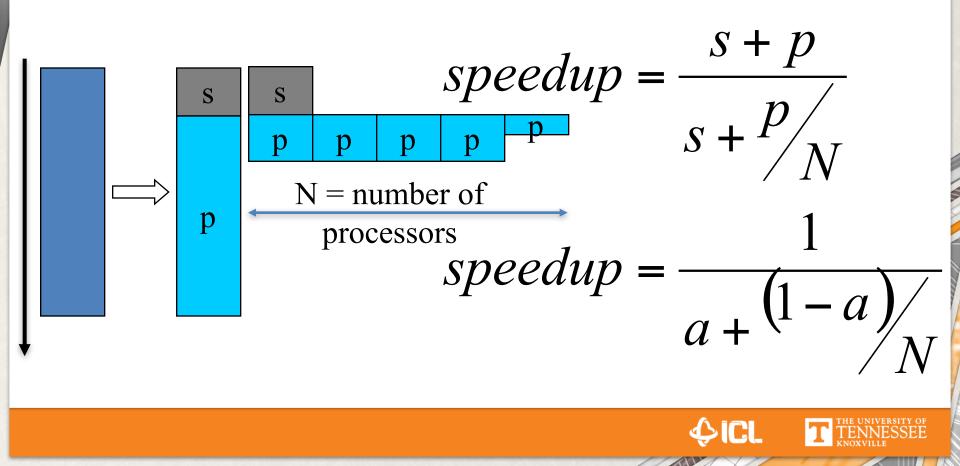
	1 Instruction flow	> 1 Instruction flow
1 data stream	SISD Von Neumann	MISD pipeline
> 1 data stream	SIMD	MIMD

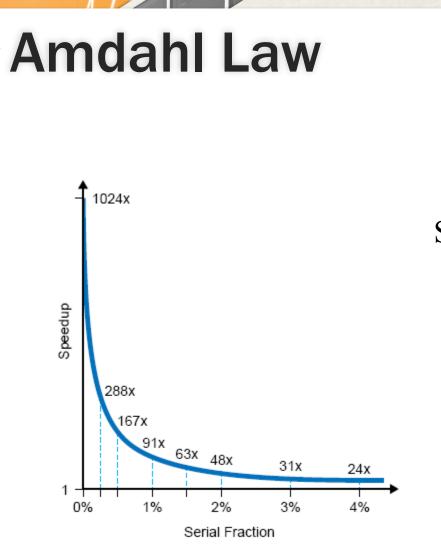
Flynn Taxonomy: Analogy

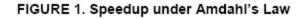
- SISD: assembly line work (no parallelism)
- SIMD: systolic, GPU computing (vector computing MMX, SSE, AVX)
- MISD: more unusual type. Safety requirements, replication capabilities, think space shuttle.
- MIMD: airport facility, several desks working at their own pace, synchronizing via a central entity (database). Most distributed algorithms, as well as multi-core applications.

Amdahl Law

- First law of parallel applications (1967)
- Limit the speedup for all parallel applications







Speedup is bound by 1/a.

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Amdahl Law

- Bad news for parallel applications
- 2 interesting facts:
 - We should limit the sequential part
 - A parallel computer should be a fast sequential computer to be able to resolve the sequential part quickly
- What about increasing the size of the initial problem ?

Gustafson's Law

- Less constraints than the Amdahl law.
- In a parallel program the quantity of data to be processed increase, so the sequential part decrease.

$$\begin{cases} t = s + P/n \\ P = a^*n \end{cases}$$
 speedup = $\frac{s + a^*n}{s + a}$

$$a \rightarrow \infty \Rightarrow speedup \rightarrow n$$



Gustafson's Law

 The limit of Amdahl Law can be transgressed if the quantity of data to be processed increase.

 $speedup \leq n + (1-n)s$

Rule stating that if the size of most problems is scaled up sufficiently, then any required efficiency can be achieved on any number of processors.

Speedup

Superlinear speedup ?

Sometimes superlinear speedups can be observed!

- Memory/cache effects
- More processors typically also provide more memory/cache.

Superlinear

Sub-linear

- Total computation time decreases due to more page/cache hits.
- Search anomalies
 - Parallel search algorithms.
 - Decomposition of search range and/or multiple search strategies.
 - One task may be "lucky" to find result early.

Parallel execution models

- Amdahl and Gustafson laws define the limits without taking in account the properties of the computer architecture
- They can only loosely be used to predict (in fact mainly to cap) the real performance of any parallel application
- We should integrate in the same model the architecture of the computer and the architecture of the application

What are models good for ?

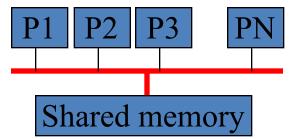
- Abstracting the computer properties
 - Making programming simple
 - Making programs portable ?
- Reflecting essential properties
 - Functionality
 - Costs
- What is the von-Neumann model for parallel architectures ?

Parallel Random Access Machine

- World described as a collection of synchronous processors which communicate with a global shared memory unit.
 - A collection of numbered RAM processors (P_i)
 - A collection of indexed memory cells (M[i])

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- Each processor P_i has it's own unbounded local memory (registers) and knows it's index (rank)
- Each processor can access any shared memory cell in unit time
- Input and output of a PRAM algorithm consist in N distinct items
- A PRAM instruction consist in 3 synchronous steps: read (acquire the input data), computation, write (save the data back to a shared memory cell).
 - Exchanging data is realized through the writing and reading of memory cells



Parallel Random Access Machine

• Algorithmic Complexity:

- Time = the time elapsed for P_0 computations
- Space = the number of memory cells accessed

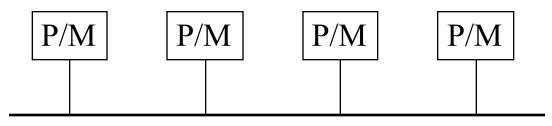
Specialized in parallel algorithms

- Natural: the number of operations per cycle on N processors is at most N
- Strong: all accesses are realized in a single time unit
- Simple: keep the complexity and correctness overheads low y abstracting all communication or synchronization overheads

The PRAM corresponds intuitively to the programmers' view of a parallel computer: it ignores lower level architectural constraints, and details, such as memory access contention and overhead, synchronization overhead, interconnection network throughput, connectivity, speed limits and link bandwidths, etc.

Bulk Synchronous Parallel – BSP Valiant 1990

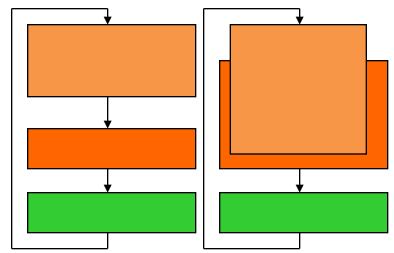
- Differs from PRAM by taking in account communications and synchronizations and by distributing the memory across participants
 - Compute: Components capable of computing or executing local memory transactions
 - Communication: A network routing messages between components
 - Synchronization: A support for synchronization on all or a sub-group of components



 Each processor can access his own memory without overhead and have a uniform slow access to remote memory

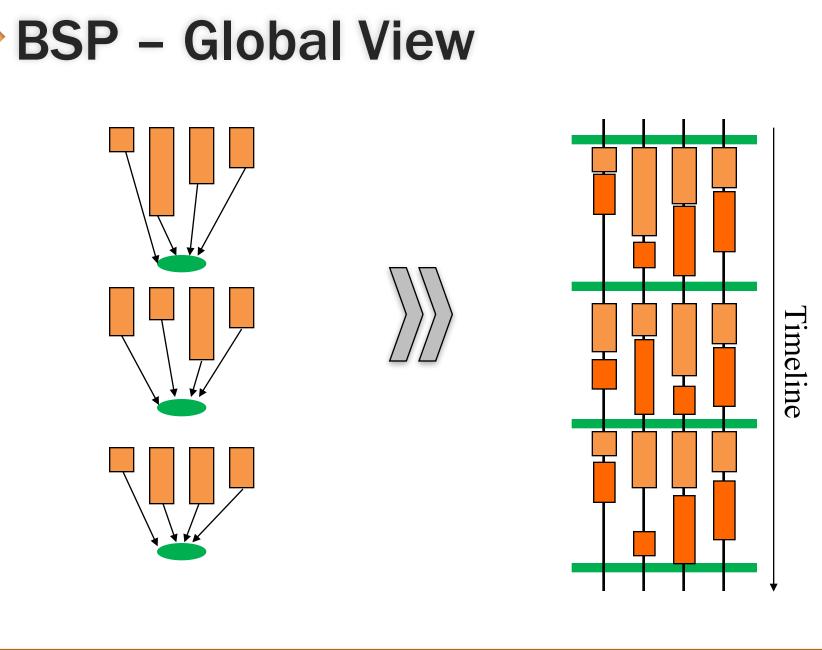
BSP - Superstep

- Applications composed by Supersteps separated by global synchronizations.
- A superstep contains:
 - A computation step
 - A communication step
 - A synchronization step



Synchronization used to insure that all processors complete the computation + communication steps in the same amount of time.

As communications are remote memory accesses (one sided) there are no synchronizations during the computation + communication step



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BSP – The communication step

- BSP consider communication not at the level of individual actions, but as a whole (per step)
- The goal being to define an upper bound on the time necessary to complete all data movements
- *h* = the maximum number of messages (incoming or outgoing) per superstep
- g = the network capability to deliver messages
 - It takes hg time for a processor to deliver h messages of size 1
 - A message of size m takes the same time to send as m messages of size 1

BSP – The synchronization cost

- The cost of synchronization is noted by *I* and is generally determined empirically
- With the increase in scale of the computing resources, the synchronizations are becoming the main bottleneck
 - Removing them might introduce deadlock or livelock
 - Decrease the simplicity of the model

BSP – Compute the cost

 $T_{superstep}$

$$= \max_{i=1}^{p} (w_i) + g * \max_{i=1}^{p} (h_i) + l \\ = w + g * h + l$$

Where:

w = max of computation time

g = 1/(network bandwidth)

h = max of number of messages

I = time for the synchronization

$$T_{total} = \sum_{s=1}^{S} T_{superstep}$$



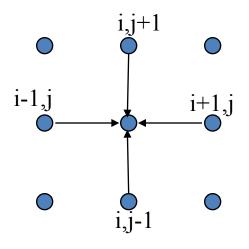
BSP

- An algorithm can be described using only w, h and the problem size.
- Collections of algorithms are available depending on the computer characteristics.
 - Small L
 - Small g
- The best algorithm can be selected depending on the computer properties.

Numerical solution to Laplace's equation

end for

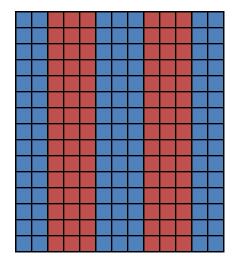
$$U_{i,j}^{n+1} = \frac{1}{4} \left(U_{i-1,j}^{n} + U_{i+1}^{n} + U_{i,j-1}^{n} + U_{i,j+1}^{n} \right)$$



for j = 1 to jmax for i = 1 to imax Unew(i,j) = 0.25 * (U(i-1,j) + U(i+1,j) + U(i,j-1) + U(i,j+1))end for

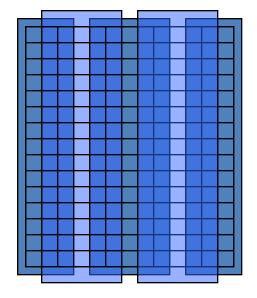


• The approach to make it parallel is by partitioning the data



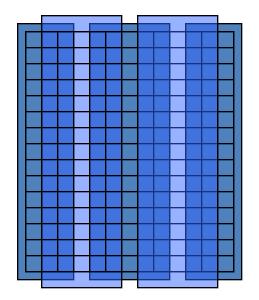


• The approach to make it parallel is by partitioning the data



Overlapping the data boundaries allow computation without communication for each superstep

On the communication step each processor update the corresponding columns on the remote processors.



for j = 1 to jmax for i = 1 to imax unew(i,j) = 0.25 * (U(i-1,j) + U(i+1,j))+ U(i,j-1) + U(i,j+1))end for end for if me not 0 then bsp put(to the left) endif if me not NPROCS -1 then bsp put(to the right) Endif bsp sync()

$$T_{superstep} = w + g * h + l$$

h = max number of messages = N values to the left + N values to the right = 2 * N (ignoring the inverse communication!) w = 4 * N * N / p

$$T_{superstep} = 4 * \frac{N^2}{p} + 2 * g * N + l$$



• BSP parameters for a wide variety of architectures has been published.

Machine	S	р	I	g
Origin 2000	101	4	1789	10.24
		32	39057	66.7
Cray T3E	46.7	4	357	1.77
		16	751	1.66
Pentium 10Mbit	61	4	139981	1128.5
		8	826054	2436.3
Pentium II	88	4	27583	39.6
100Mbit		8	38788	38.7