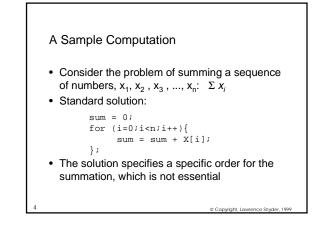
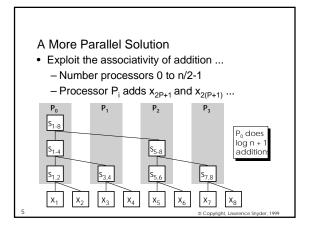
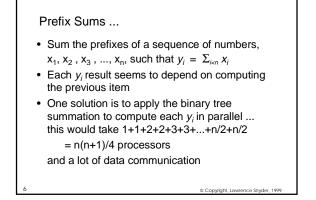


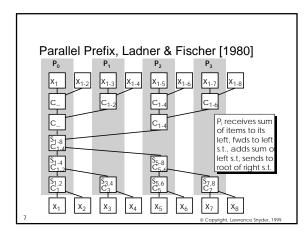
 Thinking In Parallel

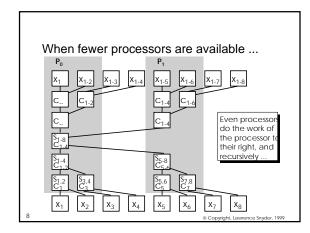
 Usually when we formulate a computation, we think of a sequential solution. Good parallel computations rarely result from transforming a sequential solution. A paradigm shift is required. So, it is essential to acquire a "parallel point of view" to produce good parallel computations from the start











### Essential Features of the Example

- Arbitrary ordering constraints removed by exploiting associativity -- focus on problem characteristics
- · Chose direct solution rather than "reducing to an earlier solution" that "over-parallelizes" -too parallel is no more useful than sequential
- · Ladner & Fischer solution can use any number of processors in the range 1 - n/2 -- scalable parallelism is essential in practice

These Guidelines Will Be Elaborated Furthe

### Consider Another Example ...

}

- Matrix multiplication is a common operation in scientific computing
- The C code for multiplying an mxn matrix A times an nxp matrix B and to produce an mxp matrix C is ...

# }

Properties of the Computation ...

- · Addition and multiplication are associative
- Each position c<sub>ii</sub> in the result is the sum of the ith row times the jth column ... all of them could be computed simultaneously
- · Each position admits plenty of parallelsim ...
  - All multiplys in row ixj column are independent · Sum of products could use binary addition tree

Notice that ideas from sequential complexity theory such a Strassen's Algorithm, that reduces the number of multiplicat ons from O(n<sup>3</sup>) to O(n<sup>2.81</sup>), do not apply. Concurrency counts h

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## A Very Parallel Solution ... · Each c<sub>ii</sub> is computed in parallel such that One processor dedicated to each a[i][k]\*b[k][j] Addition tree computes sum of those products · How many steps? · How many processors running concurrently? · Is this solution even remotely practical? • Data access -- conflicts/transit time/resources Computation time vs communication time • Processor demands -- n<sup>3</sup> procs for n<sup>2</sup> results Hello? © Copyright, Lawrence Snyder, 199

### Realities of Parallel Computers ...

### Dissiderata

- · Every computer has a fixed number of processors
- Present large computers have a few hundred processors up to a few thousand
- Using all available processors (usually) gives the best performance
- Processors can be very simple, but as first approximation, assume Pentium, PowerPC, MIPS
- The transmission of data from processor to processor is a significant (often the most signficant) cost

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### What's Important?

- · Maximizing number of processors used
- Minimizing execution time
- · Minimizing the amount of work performed
- Reducing size of memory footprint
- Maximizing (minimizing) degree of data sharing
- Reducing data motion (interprocessor comm.)
- Maximizing synchroneity or maybe asynchroneity
- Guaranteeing portability among platforms
- · Balancing work load across processors
- Maximizing programming convenience
- · Avoiding races, guaranteeing determinacy
- Improve SoftEng... robust, maintain, debug, etc

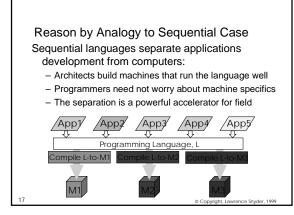
### My Answers ...

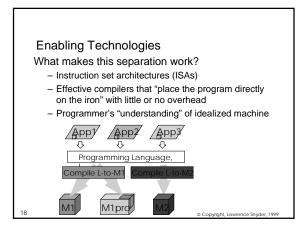
- NA Maximizing number of processors used
- 1 Minimizing execution time
- NA Minimizing the amount of work performed
- -- Reducing size of memory footprint
- -- Maximizing (minimizing) degree of data sharing
- 1 Reducing data motion (interprocessor comm.)
- -- Maximizing synchroneity or maybe asynchroneity
- <sup>1</sup> Guaranteeing portability among platforms
- 7 Balancing work load across processors
- 4 Maximizing programming convenience
- <sup>4</sup> Avoiding races, guaranteeing determinacy
- 4 Improve SoftEng... robust, maintain, debug, etc

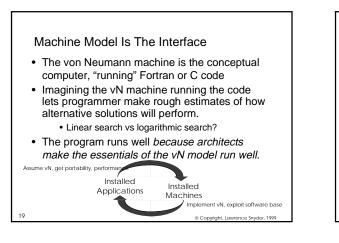
These answers are in conflict ...

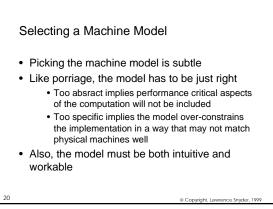
- No. 1 Goals Conflict --
  - Minimizing execution time ==> code close to the hardware
  - Portability ==> keep distance from hardware because machines differ
- No. 1 Goal Conflicts with No. 4 Goal
  - Convenience ==> ignore data motion
  - Minimizing data motion ==> attend to data motion

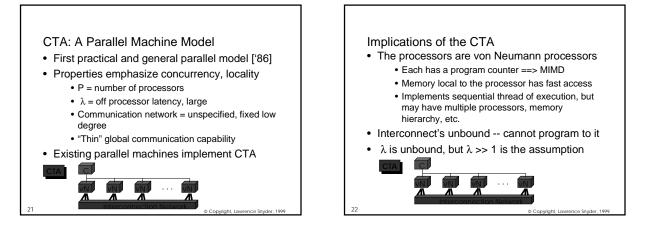
How are these conflicts solved in the sequential word?





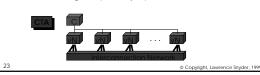




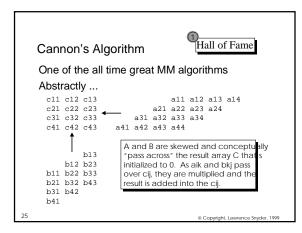


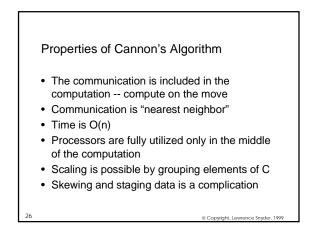
### Further Implications of CTA

- The memory is physically distributed (it must be), but there is no mention of for shared address space or shared memory
- Since  $\lambda$  is large, programs exploit locality run faster, i.e. try to compute on data in the local memory
- Fixed degree (usually 1) limits burst rate



# Reconsider the Matrix Multiplication If every processor had a copy of the A,B matrices, each could compute a rectangular subarray Memory footprint would be huge, P(mn+np) + C<sub>r</sub> Transfer time of arrays to each memory would be λ(mn+np), also huge Optimization -- C[i..i+x,j..j+y] requires rows i..i+x and columns j..j+y Total numeric operations would be O(mpn) which should benefit from a P-way speedup Alternatives?





### **Further Reading**

- L. F. Cannon [1969] A Cellular Computer to Implement the (Kalman) Filter Algorithm, Ph.D. Thesis, Montana State University
- R. E. Ladner & M. J. Fischer [1980] Parallel Prefix Computation, *Journal of the ACM* 27(4):831-838
- L. Snyder [1995] Experimental Validation of Models of Parallel Computation, A. Hofmann & J. van Leeuwen (eds), *Lecture Notes in Computer Science, Special Volume 1000*, Springer, pp. 78-100
- L. Snyder [1986] Type Architecture, Shared Memory and the Corollary of Modest Potential, Annual Review of Computer Science 1, pp. 289-318

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