

Parallel Computations for Large-Scale Problems

Lecture 1. Introduction

Def: T_s^* → execution time of the fastest serial algorithm.

T_p → total execution time of P processors

Parallel speedup $S_p = T_s^* / T_p$.

Efficiency in distributed memory machines

- * Data Parallelism: similar processing
- * Data Partitioning: partition into adjacent regions
- * Relaxed Algorithm: Embarassingly parallel: No sync
- * Synchronous Iterations: Do synchronous.
- * Replicated Workers: Central pool.
- * Pipelined Computation: arranged in a structure

Lecture 2. Performance Evaluation

Amdahl's Law: Assume serial part $t_s = f T_1 \Rightarrow$

$$T_p = f T_1 + \frac{(1-f) T_1}{P} \Rightarrow S_p = \frac{T_1}{T_p} = \frac{P}{1 + (P-1)f}$$

$$\Rightarrow \lim_{P \rightarrow \infty} S_p = \frac{1}{f}$$

Gustafson's Law: Assume T_p constant. \Rightarrow

$$T_1 = f' T_p + (1-f') P T_p \Rightarrow S_p' = f' + (1-f') P$$

Def: Parallel efficiency $\eta_p = S_p / P$.

Cost $C = T_p P$.

Parallel Execution Time:

$$T_p = t_{\text{comp}} + t_{\text{comm}} ; t_{\text{comp},i} = f_i(n, P) ;$$

$$t_{\text{comp}} = t_{\text{comp},1} + t_{\text{comp},2} + \dots ; \text{Similarly}$$

$$t_{\text{comm}} = t_{\text{comm},1} + t_{\text{comm},2} + \dots ; t_{\text{comm},i} = t_{\text{startup}} + w t_{\text{data}}$$

Lecture C: C-programming

* GDB

- gcc -g -Wall -o prog prog.c -lm

- gdb prog

. In gdb: type run to start your program

. bt: prints current call stack (list of nested functions)

. p x: prints value of variable x

. break file.c:123: set break point

. continue: continues execution

. clear 1: remove break point 1.

. l: list program code

* Valgrind

* Gprof.

* IDE

Lecture 3: Introduction to MPI.

Blocking send: `int MPI_Send`

Blocking Receive: `int MPI_Recv`

Non-blocking: `MPI_Isend` ; `MPI_Wait`.

Data Movement Routines

* Broadcast * Scattering \Leftrightarrow Gathering

Global Computation Routines

* Reduce

MPI Calls:

* `MPI_INIT`: establishes an environment.

* `MPI_Comm_size`: return # of processes

* `MPI_Comm_rank`: return rank of the process.

* `MPI_SEND` * `MPI_RECV`

* `MPI_FINALIZE`: exit MPI cleanly.

Lecture 4. Image Reconstruction and Poisson's equation

The common denominator: use a uniform mesh.

* Load-balanced linear data distribution

* Load-balanced scatter distribution

Fill the Ghost Cells: Communication

\Rightarrow Safe solution: red-black coloring.

Red-Black Communication.

* Associate each process with a color (red or black) in the p and q directions such that no neighbor has the same color.

Optimal process Topology: Minimize $\Phi(P) = \left(\frac{P}{M} + \frac{Q}{N}\right)$.

$$\Rightarrow P = \sqrt{\frac{M}{N} R}, \quad R = P \times Q$$

Surface to Volume Ratio:

* T_{comp} is proportional to $I_p * J_q$.

* T_{comm} is proportional to $2(I_p + J_q)$

Gauss-Seidal Iteration: Change value immediately

Lexicographic Order: for $n=1..N$; for $m=1..M$: $u(m,n)$

Pipelined Computations: Gauss-Seidal. Red-black ordering.

Lecture 5. Matrix Multiplication and Collective Communication

Addition of Matrices:

Embarrassingly parallel, all data access purely local.

Recursive Doubling for taking sums.

Matrix-Vector Multiplication

* Use a load-balanced data distribution on $R = P \times Q$ processors in array topology to store A .

* Once data distributed, all individual components can be computed in parallel using the Recursive Doubling Algorithm

Matrix-Matrix Multiplication.

Assume $M=N=K$ and $R=P \times P$.

Cannon's Algorithm:

- * In a first step, assume $P=M$.
- * Data distribution: processor $(m-1, n-1)$ holds $a_{m,n}$
 $b_{m,n}$, $c_{m,n}$
- * On the "diagonal" processors $(m-1, n-1)$, parts of the sum $c_{m,m} = a_{m,1} \cdot b_{1,m} + \dots + a_{m,m} \cdot b_{m,m} + \dots + a_{m,M} \cdot b_{M,m}$ are available.
- * Shifting the rows of A cyclic to the right and those of B cyclic downwards provides the next term.
- * Repeat this cyclic exchange once again completes the calculation of the diagonal elements.

Algorithm:

- Initially, processor (p, q) has elements $a_{p+1, q+1}$, $b_{p+1, q+1}$
- Shift the rows of A and the columns of B into the structure described above.
- For $k=1 \dots N$
 - Multiply the local values on each processor
 - Shift the rows of A and the columns of B cyclically by one processor
- If necessary: undo step 1.

Definition of PageRank

- * Let u be a webpage
- * Let F_m be the set of pages m points to
- * Let B_m be the set of pages that point to m .
- * Let $N_m = |F_m|$ be the number of forward links, and c be a normalization factor
- * Then we define PageRank $R_m = c \sum_{n \in B_m} \frac{R_n}{N_n}$
- * Define a matrix A by $a_{mn} = \begin{cases} 1/N_n, & \text{if there is link from } n \text{ to } m \\ 0, & \text{otherwise} \end{cases}$
- * Let $R = (R_1, \dots, R_M)^T$ be the pagerank vector. Then it holds $R = cAR$.

The Power Method for Eigenvalue Problems.

- * Given a square matrix A and an initial guess $x^{[0]} \neq 0$
 - * Form the sequence $x^{[k+1]} = Ax^{[k]}$, $k=0, 1, \dots$
 - * For many A , $\{x^{[k]} / \|x^{[k]}\|\}$ converges towards an eigenvector to the largest eigenvalue c of A .
 - * An estimation C_k of c is given by $C_k = \frac{\langle x^{[k+1]}, x^{[k]} \rangle}{\langle x^{[k]}, x^{[k]} \rangle}$
- Vector Transposition
- * If $P=Q$, MPI-Alltoall can be used to transpose y .
 - * The general case is equivalent to one recursive doubling step.

Lecture 6. Linear Systems of Equations by Gaussian Elimination

Gaussian Elimination

Idea: For $n=1, \dots, N-1$: Use equation n for the elimination of x_n from the equations $n+1, \dots, N$

For $n=N, \dots, 1$: Compute x_n from the n -th equation

Forward elimination + Backsubstitution

Pivoting + LU decomposition

Use data partition to distribute

Forward substitution: Sequential algorithm

Use pipelined computation: Lots of communication

Use Data Distribution μ :

- * Assume a row data distribution μ

- * The last process ($P-1$) must wait for x_1 to arrive.

- * x_1 is ready to send if the first processor (0) has done its work.

- * In an ideal world, further communication is completely overlapped by computation.

- * $P \times Q$ processor mesh gives rather poor performance.

Lecture 7. Parallel Sorting

Symmetric Distributed Compare And Exchange

Linear data distribution place N/P consecutive elements on each processor. First step is local sorting.

Then do the merging. Complexity $(2N/P-1)t_a$.

Parallel Bubble Sort: Odd-Even Sort

2 phases, all processes are comparing. Guaranteed to terminate after $N/2$ odd-even and even-odd steps.

Bitonic Mergesort: Bitonic Sort implemented on

$P=N=2^D$ processor has a time complexity of $O(\log^2 N)$.

Bucket Sort: Uniformly distributed keys.

Lecture 8. Algorithms on Graph.

Assume (connected) weighted undirected graph.

Algorithms For Dense Graphs

Prim's algorithm: Inner loop can be parallelized.

Distribute V load balanced over P processes.

Distribute the weighted adjacency matrix A correspondingly over a $1 \times P$ process mesh.

Single Source Shortest Paths - Dijkstra's Algorithm
Similar to Prim's algorithm

Floyd's Algorithm: Parallel innermost loop

Algorithms For Sparse Graphs