Parallel Computing

Lecture 6: Performance Analysis

Mohamed Zahran (aka Z)
mzahran@cs.nyu.edu
http://www.mzahran.com
Defining Performance

- Which airplane has the best performance?
Standard Definition of Performance

- For some program running on machine X,
  \[
  \text{Performance}_X = \frac{1}{\text{Execution time}_X}
  \]

- "X is n times faster than Y"
  \[
  \frac{\text{Performance}_X}{\text{Performance}_Y} = n
  \]

  Example: time taken to run a program
  - 10s on A, 15s on B
  - Execution Time_B / Execution Time_A
    = 15s / 10s = 1.5
  - So A is 1.5 times faster than B
Speedup

- Number of cores = \( p \)
- Serial run-time = \( T_{\text{serial}} \)
- Parallel run-time = \( T_{\text{parallel}} \)

\[
S = \frac{T_{\text{serial}}}{T_{\text{parallel}}}
\]
Example

Perfect parallelization!
Does it ever occur?

\[ S_p = \frac{100}{25} = 4.0, \]

\[ S_p = \frac{100}{35} = 2.85, \]

perfect load balancing
Example (cont.)

\[ S_p = \frac{100}{40} = 2.5, \]
load imbalance

\[ S_p = \frac{100}{50} = 2.0, \]
load imbalance and sync cost

closest to real life parallel programs
## A Glimpse at the Top 500

<table>
<thead>
<tr>
<th>Rank</th>
<th>Site</th>
<th>System</th>
<th>Cores</th>
<th>$R_{max}$ (TFlop/s)</th>
<th>$R_{peak}$ (TFlop/s)</th>
<th>Power (kW)</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>National Super Computer Center in Guangzhou, China</td>
<td><strong>Tianhe-2 (MilkyWay-2)</strong> - TH-IVB-FEP Cluster, Intel Xeon E5-2692 12C 2.200GHz, TH Express-2, Intel Xeon Phi 31S1P NUDT</td>
<td>3120000</td>
<td>33862.7</td>
<td>54902.4</td>
<td>17808</td>
</tr>
<tr>
<td>2</td>
<td>DOE/SC/Oak Ridge National Laboratory, United States</td>
<td><strong>Titan</strong> - Cray XK7, Opteron 6274 16C 2.200GHz, Cray Gemini interconnect, NVIDIA K20x Cray Inc.</td>
<td>560640</td>
<td>17590.0</td>
<td>27112.5</td>
<td>8209</td>
</tr>
<tr>
<td>3</td>
<td>DOE/NNSA/LLNL, United States</td>
<td><strong>Sequoia</strong> - BlueGene/Q, Power BQC 16C 1.60 GHz, Custom IBM</td>
<td>1572864</td>
<td>17173.2</td>
<td>20132.7</td>
<td>7890</td>
</tr>
<tr>
<td>4</td>
<td>RIKEN Advanced Institute for Computational Science (AICS), Japan</td>
<td><strong>K computer</strong>, SPARC64 VIIIfx 2.0GHz, Tofu interconnect Fujitsu</td>
<td>705024</td>
<td>10510.0</td>
<td>11280.4</td>
<td>12660</td>
</tr>
</tbody>
</table>
Sources of Parallel Overheads

- Overhead of creating threads/processes
- Synchronization
- Load imbalance
- Communication
- Extra computation
- Memory access (for both sequential and parallel!)
Efficiency of a parallel program

\[ E = \frac{S}{p} = \frac{T_{\text{serial}}}{T_{\text{parallel}}} p \]
Be Careful about $T$

- Both $T_{seq}$ and $T_{par}$ are wall-clock times, and as such they are not objective. They can be influenced by:
  - The skill of the programmer who wrote the implementations
  - The choice of compiler (e.g. GNU C++ versus Intel C++)
  - The compiler switches (e.g. turning optimization on/off)
  - The operating system
  - The type of filesystem holding the input data (e.g. EXT4, NTFS, etc.)
  - The time of day... (different workloads, network traffic, etc.)
Scalability

- **Scalability** is the ability of a (software or hardware) system to handle a growing amount of work efficiently.

- If we keep the efficiency fixed by increasing the number of processes/threads and without increasing problem size, the problem is **strongly scalable**.

- If we keep the efficiency fixed by increasing the problem size at the same rate as we increase the number of processes/threads, the problem is **weakly scalable**.
Scalability

With scalability we want two things:

- Make use of more computing resources

\[
\text{strongScalingEfficiency}(N) = \frac{t_{seq}}{N \cdot t_{par}}
\]

- Solving bigger problems

\[
\text{weakScalingEfficiency}(N) = \frac{t_{seq}}{t'_{par}}
\]

t'_{par} is the time to solve a N-times bigger problem than t_{seq}
Let’s take a closer look at timing.
Taking Timings

• What is time?
• Start to finish?
• A program segment of interest?
• CPU time?
• Wall clock time?
Execution Time

• **Elapsed Time**
  - counts everything (disk and memory accesses, I/O, etc.)
  - a useful number, but often not good for comparison purposes

• **CPU time**
  - doesn't count I/O or time spent running other programs
  - can be broken up into system time, and user time

• **Our focus: user CPU time**
  - time spent executing the lines of code that are "in" our program
Execution Time (Elapsed Time)

- I/O Time
- CPU Time
- Disk and Memory time

  - User CPU Time
  - System CPU Time
**Taking Timings**

Inside your C program:

```c
#include <time.h>
#include <stdio.h>

int main() {
    clock_t start, end, total;
    int i;

    start = clock();

    for(i=0; i< 10000000; i++) { }

    end = clock();
    total= (double)(end – start) / CLOCKS_PER_SEC;

    printf("Total time taken by CPU: %f\n", total);
}
```
Let’s Look at Two Simple Metrics

• **Response time (aka Execution Time)**
  – The time between the start and completion of a task

• **Throughput**
  – Total amount of work done in a given time

What is the relationship between execution time and throughput?
Timing for sequential programs

- IPC
- MIPS
- CPI
- Execution Time
- # insts
Execution time for sequential program:

\[
\text{seconds per program} = \frac{\text{cycles per program}}{\text{cycle}} \times \frac{\text{seconds}}{\text{cycle}}
\]

\[
ET = IC \times CPI \times CT
\]

ET = Execution Time
CPI = Cycles Per Instruction
IC = Instruction Count
Example

A program runs in 10 seconds on computer A, which has a 4 GHz clock. We are trying to help a computer designer build a new machine B, that will run this program in 6 seconds. The designer can use new (or perhaps more expensive) technology to substantially increase the clock rate, but has informed us that this increase will affect the rest of the CPU design, causing machine B to require 1.2 times as many clock cycles as machine A for the same program. What clock rate should we tell the designer to target?“
CPI Example

• Suppose we have two implementations of the same instruction set architecture (ISA).

For some program, Machine A has a clock cycle time of 250 ps and a CPI of 2.0
Machine B has a clock cycle time of 500 ps and a CPI of 1.2
What machine is faster for this program, and by how much?

[ $10^{-3} = \text{milli}, \ 10^{-6} = \text{micro}, \ 10^{-9} = \text{nano}, \ 10^{-12} = \text{pico}, \ 10^{-15} = \text{femto}$ ]
#Instructions Example

- A compiler designer is trying to decide between two code sequences for a particular machine. Based on the hardware implementation, there are three different classes of instructions: Class A, Class B, and Class C, and they require one, two, and three cycles (respectively).

  The first code sequence has 5 instructions:
  2 of A, 1 of B, and 2 of C

  The second sequence has 6 instructions:
  4 of A, 1 of B, and 1 of C.

Which sequence will be faster? How much? What is the CPI for each sequence?
MIPS Example

- Two different compilers are being tested for a 4 GHz. machine with three different classes of instructions: Class A, Class B, and Class C, which require one, two, and three cycles (respectively). Both compilers are used to produce code for a large piece of software.

  The first compiler's code uses 5 million Class A instructions, 1 million Class B instructions, and 1 million Class C instructions.

  The second compiler's code uses 10 million Class A instructions, 1 million Class B instructions, and 1 million Class C instructions.

- Which sequence will be faster according to MIPS?
- Which sequence will be faster according to execution time?
Pitfalls in timing in Parallel Machines
For Multithreaded Programs

• Shall we use execution time or throughput? or both?

• IPC is not accurate here
  – small timing variations may lead to different execution
  – Order at which threads enter critical section may vary
  – Different interrupt timing may lead to different scheduling decisions

The total number of instructions executed may be different across different runs!
For Multithreaded Programs

The total number of instructions executed may be different across different runs!

This effect increases with the number of cores

System-level code account for a significant fraction of the total execution time
Your Program Does Not Run in A Vacuum

- OS at least is there.
- Multi-programming and/or multithreading setting is very common in multicore settings
- Independent programs affect each other performance (why?)
How to check the performance of a parallel machine?
Benchmarks

• Performance best determined by running a real application
  - Use programs typical of expected workload
  - Or, typical of expected class of applications
  - e.g., compilers/editors, scientific applications, graphics, etc.

• Small benchmarks
  - nice for architects and designers
  - easy to standardize

• Parallel Benchmarks: PARSEC, Rodinia, SPLASH-2

• SPEC (System Performance Evaluation Cooperative)
  - companies have agreed on a set of real program and inputs
  - valuable indicator of performance (and compiler technology)
Role of Benchmarks

• help designer explore architectural designs
• identify bottlenecks
• compare different systems
• conduct performance prediction
Example: PARSEC

- Princeton Application Repository for Shared-Memory Computers
- Benchmark Suite for Chip-Multiprocessors
- Freely available at: http://parsec.cs.princeton.edu/

Objectives:
- Multithreaded Applications: Future programs must run on multiprocessors
- Emerging Workloads: Increasing CPU performance enables new applications
- Diverse: Multiprocessors are being used for more and more tasks
- State-of-Art Techniques: Algorithms and programming techniques evolve rapidly
### Example: PARSEC

<table>
<thead>
<tr>
<th>Program</th>
<th>Application Domain</th>
<th>Parallelization</th>
</tr>
</thead>
<tbody>
<tr>
<td>Blackscholes</td>
<td>Financial Analysis</td>
<td>Data-parallel</td>
</tr>
<tr>
<td>Bodytrack</td>
<td>Computer Vision</td>
<td>Data-parallel</td>
</tr>
<tr>
<td>Canneal</td>
<td>Engineering</td>
<td>Unstructured</td>
</tr>
<tr>
<td>Dedup</td>
<td>Enterprise Storage</td>
<td>Pipeline</td>
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<tr>
<td>Facesim</td>
<td>Animation</td>
<td>Data-parallel</td>
</tr>
<tr>
<td>Ferret</td>
<td>Similarity Search</td>
<td>Pipeline</td>
</tr>
<tr>
<td>Fluidanimate</td>
<td>Animation</td>
<td>Data-parallel</td>
</tr>
<tr>
<td>Freqmine</td>
<td>Data Mining</td>
<td>Data-parallel</td>
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<tr>
<td>Streamcluster</td>
<td>Data Mining</td>
<td>Data-parallel</td>
</tr>
<tr>
<td>Swaptions</td>
<td>Financial Analysis</td>
<td>Data-parallel</td>
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<tr>
<td>Vips</td>
<td>Media Processing</td>
<td>Data-parallel</td>
</tr>
<tr>
<td>X264</td>
<td>Media Processing</td>
<td>Pipeline</td>
</tr>
</tbody>
</table>
Example: Rodinia

- A Benchmark Suite for Heterogeneous Computing: multicore CPU and GPU
- University of Virginia

<table>
<thead>
<tr>
<th>Application / Kernel</th>
<th>Dwarf</th>
<th>Domain</th>
</tr>
</thead>
<tbody>
<tr>
<td>K-means</td>
<td>Dense Linear Algebra</td>
<td>Data Mining</td>
</tr>
<tr>
<td>Needleman-Wunsch</td>
<td>Dynamic Programming</td>
<td>Bioinformatics</td>
</tr>
<tr>
<td>HotSpot*</td>
<td>Structured Grid</td>
<td>Physics Simulation</td>
</tr>
<tr>
<td>Back Propagation*</td>
<td>Unstructured Grid</td>
<td>Pattern Recognition</td>
</tr>
<tr>
<td>SRAD</td>
<td>Structured Grid</td>
<td>Image Processing</td>
</tr>
<tr>
<td>Leukocyte Tracking</td>
<td>Structured Grid</td>
<td>Medical Imaging</td>
</tr>
<tr>
<td>Breadth-First Search*</td>
<td>Graph Traversal</td>
<td>Graph Algorithms</td>
</tr>
<tr>
<td>Stream Cluster*</td>
<td>Dense Linear Algebra</td>
<td>Data Mining</td>
</tr>
<tr>
<td>Similarity Scores*</td>
<td>MapReduce</td>
<td>Web Mining</td>
</tr>
</tbody>
</table>
Conclusions

• Performance evaluation is very important to assess programming quality as well as the underlying architecture and how they interact.

• The following capture some aspects of the system but do not represent overall performance: MIPS, #instructions, #cycles, frequency

• **Execution time is what matters**: system time, CPU time, I/O and memory time

• Scalability and efficiency measure the quality of your code.