

Performance Measures: Part I

The Single Processor Case

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Time Measurement and Operation Counts



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That means the time we have to wait for a response of the program includes the waiting times besides the CPU time.



Instructions: Timings and Counts

clock rate and cycle time

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Example

A CPU with a clock rate of 3.5 GHz = $3.5 \times 10^9 \text{ 1/s}$ executes $3.5 \times 10^9 \text{ clock}$ ticks per second. The length of a clock cycle thus is

$$1/(3.5\,10^9)\,\mathrm{s} = 1/3.5\,10^{-9}\,\mathrm{s} \approx 0.29\,\mathrm{ns}$$



Instructions: Timings and Counts

Different instructions require different times to get executed. This is represented by the so called cycles per instruction (CPI) of the corresponding instruction. An average CPI is connected to a process A via CPI(A).

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This number determines the total user CPU time together with the number of instructions and cycle time via

$$T_{U_CPU}(A) = n_{instr}(A) \cdot CPI(A) \cdot t_{cycle}$$

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Clever choices of the instructions can influence the values of $n_{instr}(A)$ and CPI(A). \leadsto compiler optimization.

Time Measurement and Operation Counts



A common performance measure of CPU manufacturers is the Million instructions per second (MIPS) rate.

It can be expressed as

$$MIPS(A) = \frac{n_{instr}(A)}{T_{U_CPU}(A) \cdot 10^6} = \frac{r_{cycle}}{CPI(A) \cdot 10^6},$$

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This measure can be misleading in high performance computing, since higher instruction throughput does not necessarily mean shorter execution time.

Time Measurement and Operation Counts



More common for the comparison in Scientific computing is the rate of floating point operations (FLOPS) executed. The MFLOPS rate of a program A can be expressed as

$$MFLOPS(A) = \frac{n_{FLOPS}(A)}{T_{U_CPU}(A) \cdot 10^6} [1/s],$$

with $n_{FLOPS}(A)$ the total number of FLOPS issued by the program A.

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Note that not all FLOPS (see also Chapter 3 summer term) take the same time to execute. Usually divisions and square roots are much slower. The MFLOPS rate, however, does not take this into account.



Example

```
Input:
```

```
ct0=0;
A=randn(1500);
tic
ct0=cputime;
pause(2)
toc
cputime-ct0
tic
ct0=cputime;
[Q,R]=qr(A);
toc
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CPU_Time versus Execution Time



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```

Output:

```
Elapsed time is 2.000208 seconds.

ans =

0.0300

Elapsed time is 0.733860 seconds.

ans =

21.6800
```

Executed on a 4x8core Xeon® system.

cputime-ct0

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In any case it can be much smaller, as well.



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The first result is easily explained by the splitting of the execution time into user/system CPU time and waiting time. The process is mainly waiting for the sleep system call to return whilst basically accumulating no active CPU time.



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The second result is due to the fact that the activity is distributed to several cores. Each activity accumulates its own CPU time and these are summed up to the total CPU time of the process.

Parallel Cost and Optimality



Definition (Parallel cost and cost-optimality)

The cost of a parallel program with data size n is defined as

$$C_p(n) = p * T_p(n).$$

Here $T_p(n)$ is the parallel runtime of the process its execution time on p processors.

The parallel program is called cost-optimal if

$$C_p = T^*(n).$$

 $T^*(n)$ represents the execution time of the fastest sequential program solving the same problem.

In practice $T^*(n)$ is often approximated by $T_1(n)$.

Speedup



The speedup of parallel program

$$S_p(n) = \frac{T^*(n)}{T_p(n)},$$

is a measure for the acceleration in terms of execution time we can expect from a parallel program.

The speedup is strictly limited from above by p Since otherwise the parallel program would motivate a faster sequential algorithm. See [Rauber/Rünger '10] for details.

In practice often the speedup is computed with respect to the sequential version of the code, i.e.,

$$S_p(n) pprox rac{T_1(n)}{T_p(n)}.$$



Usually the parallel execution of the work a program has to perform comes at the cost of certain management of subtasks. Their distribution, organization and interdependence leads to a fraction of the total execution, that has to be done extra.

Definition

The fraction of work that has to be performed by a sequential algorithm as well is described by the parallel efficiency of a program. It is described by

$$E_p(n) = \frac{T^*(n)}{C_p(n)} = \frac{S_p(n)}{p} = \frac{T^*}{p \cdot T_p(n)}.$$

The parallel efficiency obviously is limited from above by $E_p(n) = 1$ representing the perfect speedup of p.



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The total parallel runtime $T_p(n)$ then consists of

- $f \cdot T^*(n)$ the time for the sequential fraction and
- $(1-f)/p \cdot T^*(n)$ the time for the fully parallel part.

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The best attainable speedup can thus be expressed as

$$S_p(n) = \frac{T^*(n)}{f \cdot T^*(n) + \frac{1-f}{p}T^*(n)} = \frac{1}{f + \frac{1-f}{p}} \leq \frac{1}{f}.$$

Scalability of Parallel Programs



Question

Is the parallel efficiency of a parallel program independent of the number of processors p used?

The question is answered by the concept of parallel scalability. Scientific computing and HPC distinguish two forms of scalability:

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strong scalability captures the dependence of the parallel runtime on the number of

processors for a fixed total problem size.

weak scalability

captures the dependence of the parallel runtime on the number of processors for a fixed problem size per processor.