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Scientific Benchmarking of Parallel Computing Systems

Twelve ways to tell the masses when reporting performance results

DPHPC special session on benchmarking for the final reports

Paper and video available at: <http://hpc.inf.ethz.ch/publications/index.php?pub=222>

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Disclaimer(s)

- This is an experience talk (published at SC 15 – State of the Practice!)
  - Explained in SC15 FAQ:  
"generalizable insights as gained from experiences with particular HPC machines/operations/applications/benchmarks, overall analysis of the status quo of a particular metric of the entire field or historical reviews of the progress of the field."
- Don't expect novel insights  
Given the papers I read, much of what I say may be new for many
- My musings shall not offend anybody
  - Everything is (now) anonymized
- Criticism may be rhetorically exaggerated
  - Watch for tropes!
- This talk should be entertaining!

CAUTION:

USE THESE WORDS WITH DISCRETION

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How does Garth measure and report performance?

- We are all interested in High Performance Computing
  - We (want to) see it as a science – reproducing experiments is a major pillar of the scientific method
- When measuring performance, important questions are
  - "How many iterations do I have to run per measurement?"
  - "How many measurements should I run?"
  - "Once I have all data, how do I summarize it into a single number?"
  - "How do I compare the performance of different systems?"
  - "How do I measure time in a parallel system?"
  - ...
- How are they answered in the field today?
  - Let me start with a little anecdote ... a reaction to this paper ☺

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(2006)

(2015)

- Original findings:
  - If carefully tuned, NBC speeds up a 3D solver  
Full code published
  - 800<sup>3</sup> domain – 4 GB (distributed) array  
1 process per node, 8-96 nodes  
Opteron 246 (old even in 2006, retired now)
  - Super-linear speedup for 96 nodes  
~5% better than linear
- 9 years later: attempt to reproduce ☹!
  - System A: 28 quad-core nodes, Xeon E5520
  - System B: 4 nodes, dual Opteron 6274

"Neither the experiment in A nor the one in B could reproduce the results presented in the original paper, where the usage of the NBC library resulted in a performance gain for practically all node counts, reaching a superlinear speedup for 96 cores (explained as being due to cache effects in the inner part of the matrix vector product)."

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State of the Practice in HPC

- Stratified random sample of three top-conferences over four years
  - HPDC, PPoPP, SC (years: 2011, 2012, 2013, 2014)
  - 10 random papers from each (10-50% of population)
  - 120 total papers, 20% (25) did not report performance (were excluded)
- Main results:
  - Most papers report details about the hardware but fail to describe the software environment.  
Important details for reproducibility missing
  - The average paper's results are hard to interpret and easy to question  
Measurements and data not well explained
  - No statistically significant evidence for improvement over the years ☺
- Our main thesis:  
Performance results are often nearly impossible to reproduce! Thus, we need to provide enough information to allow scientists to understand the experiment, draw own conclusions, assess their certainty, and possibly generalize results.

This is especially important for HPC conferences and activities such as the Gordon Bell award!

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Well, we all know this - but do we really know how to fix it?

1991 – the classic!  
Twelve Ways to Fool the Masses When Giving Performance Results in Parallel Computers

2012 – the shocking  
How to Fool the Masses When Giving Performance Results in Parallel Computers

2013 – the extension  
How to Fool the Masses When Giving Performance Results in Parallel Computers

Fooling the Masses with Performance Results: Old Classics & Some New Ideas

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Yes, this is a garlic press!

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This is not new – meet Eddie!

1991 – the classical

### Our constructive approach: provide a set of (12) rules

- Attempt to emphasize interpretability of performance experiments
- The set is not complete
  - And probably never will be
  - Intended to serve as a solid start
  - Call to the community to extend it
- I will illustrate the 12 rules now
  - Using real-world examples
  - All anonymized!
  - Garth and Eddie will represent the bad/good scientist

Yes, this is a generic press!

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### The most common issue: speedup plots

Check out my wonderful Speedup!

I can't tell if this is useful at all!

- Most common and oldest-known issue
  - First seen 1988 – also included in Bailey's 12 ways
  - 39 papers reported speedups
  - 15 (38%) did not specify the base-performance ©
  - Recently rediscovered in the "big data" universe
  - A. Rowstron et al.: Nobody ever got fired for using Hadoop on a cluster, HotCDP 2012
  - F. McSherry et al.: Scalability! but at what cost?, HotOS 2015

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### The most common issue: speedup plots

Check out my wonderful Speedup!

I can't tell if this is useful at all!

**Rule 1: When publishing parallel speedup, report if the base case is a single parallel process or best serial execution, as well as the absolute execution performance of the base case.**

- Most common
  - First seen 1988 – also included in Bailey's 12 ways
  - A simple generalization of this rule implies that one should never report ratios without absolute values.
  - Recently rediscovered in the "big data" universe
  - A. Rowstron et al.: Nobody ever got fired for using Hadoop on a cluster, HotCDP 2012
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### Garth's new compiler optimization

Check out my compiler!

Well, GarthCC segfaulted for FT and was 2x slower for BT

How did you perform and B?

**Rule 2: Specify the reason for only reporting subsets of standard benchmarks or applications or not using all system resources.**

- This implies: Show results even if your code/approach stops scaling!

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### The mean parts of means – or how to summarize data

**Rule 3: Use the arithmetic mean only for summarizing costs. Use the harmonic mean for summarizing rates.**

**Rule 4: Avoid summarizing ratios; summarize the costs or rates that the ratios base on instead. Only if these are not available use the geometric mean for summarizing ratios.**

Ah, true, the

- 51 papers use means to summarize data, only four (!) specify which mean was used
  - A single paper correctly specifies the use of the harmonic mean
  - Two use geometric means, without reason
  - Similar issues in other communities (PLDI, CGO, LCTES) – see N. Amaral's report
  - harmonic mean  $\leq$  geometric mean  $\leq$  arithmetic mean

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### Dealing with variation

The latency of PixiDora is

How did you get to this?

**Rule 5: Report if the measurement values are deterministic. For nondeterministic data, report confidence intervals of the measurement.**

- Most papers report nondeterministic measurement results
  - Only 15 mention some measure of variance
  - Only two (!) report confidence intervals
- CIs allow us to compute the number of required measurements!
- Can be very simple, e.g., single sentence in evaluation:
 

"We collected measurements until the 99% confidence interval was within 5% of our reported means."

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## Dealing with variation

The confidence interval is 1.765us to 1.775us

**Rule 6:** Do not assume normality of collected data (e.g., based on the number of samples) without diagnostic checking.

- Most events will slow down performance
  - Heavy right-tailed distributions
- The Central Limit Theorem only applies asymptotically
  - Some papers/textbook mention "30-40 samples", don't trust them!
- Two papers used CIs around the mean without testing for normality

Can we test for normality?

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## Dealing with non-normal data – nonparametric statistics

- Rank-based measures (no assumption about distribution)
  - Essentially always better than assuming normality
- Example: median (50<sup>th</sup> percentile) vs. mean for HPL
  - Rather stable statistic for expectation
  - Other percentiles (usually 25<sup>th</sup> and 75<sup>th</sup>) are also useful

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## Comparing nondeterministic measurements

I saw variance using GarthCC as well!

Show me the data!

Retract the paper! You have not shown anything!

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## What if the data looks weird!?

Look what data I got!

Clearly, the mean/median are not sufficient!

Try quantile regression!

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## Quantile Regression

Wow, so Pilatus is better for (worst-case) latency-critical workloads even though Dora is expected to be faster

**Rule 8:** Carefully investigate if measures of central tendency such as mean or median are useful to report. Some problems, such as worst-case latency, may require other percentiles.

Check Oliveira et al. "Why you should care about quantile regression". SIGARCH Computer Architecture News, 2013.

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## How many measurements are needed?

- Measurements can be expensive!
  - Yet necessary to reach certain confidence
- How to determine the minimal number of measurements?
  - Measure until the confidence interval has a certain acceptable width
  - For example, measure until the 95% CI is within 5% of the mean/median
  - Can be computed analytically assuming normal data
  - Compute iteratively for nonparametric statistics
- Often heard: "we cannot afford more than a single measurement"
  - E.g., Gordon Bell runs
  - Well, then one cannot say anything about the variance
  - Even 3-4 measurement can provide very tight CI (assuming normality)
  - Can also exploit repetitive nature of many applications

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Experimental design

MPI\_Reduce

I don't believe you, try other numbers of processes!

**Rule 9: Document all varying factors and their levels as well as the complete experimental setup (e.g., software, hardware, techniques) to facilitate reproducibility and provide interpretability.**

- We recommend factorial design
- Consider parameters such as node allocation, process-to-node mapping, network or node contention
  - If they cannot be controlled easily, use randomization and model them as random variable
- This is hard in practice and not easy to capture in rules

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Time in parallel systems

My simple broadcast takes only one latency!

That's nonsense!

But I measured it so it must be true!

Measure each operation separately!

```

t = -MPI_Wtime();
for(i=0; i<1000; i++) {
  MPI_Bcast(...);
}
t += MPI_Wtime();
t /= 1000;
  
```

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Summarizing times in parallel systems!

My new reduce

Come on, show me the data!

**Rule 10: For parallel time measurements, report all measurement, (optional) synchronization, and summarization techniques.**

- Measure events separately
  - Use high-precision timers
  - Synchronize processes
- Summarize across processes:
  - Min/max (unstable), average, median – depends on use-case

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Give times a meaning!

I compute 10%

I have no clue.

**Rule 11: If possible, show upper performance bounds to facilitate interpretability of the measured results.**

- Model computer system as k-dimensional space
  - Each dimension represents a capability: Floating point, Integer, memory bandwidth, cache bandwidth, etc.
- Features are typical rates
  - Determine maximum rate for each dimension: E.g., from documentation or benchmarks
- Can be used to proof optimality of implementation
  - If the requirements of the bottleneck dimension are minimal

Can you provide?

- Ideal speedup
- Amdahl's speedup
- Parallel overheads

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Plot as much information as possible!

My most common request was "show me the data"

**Rule 12: Plot as much information as needed to interpret the experimental results. Only connect measurements by lines if they indicate trends and the interpolation is valid.**

This is how I should have presented the Dora results.

Box Plot, Violin Plot, Combined Plot

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Conclusions and call for action

- Performance may not be reproducible
  - At least not for many (important) results
- Interpretability fosters scientific progress
  - Enables to build on results
  - Sounds statistics is the biggest gap today
- We need to foster interpretability
  - Do it ourselves (this is not easy)
  - Teach young students
  - Maybe even enforce in TPCs
- See the 12 rules as a start
  - Need to be extended (or concretized)
  - Much is implemented in LibSciBench [1]

No vegetables were harmed for creating these slides!

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[1]: <http://spcl.inf.ethz.ch/Research/Performance/LibLSB/>.

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## Backup slides