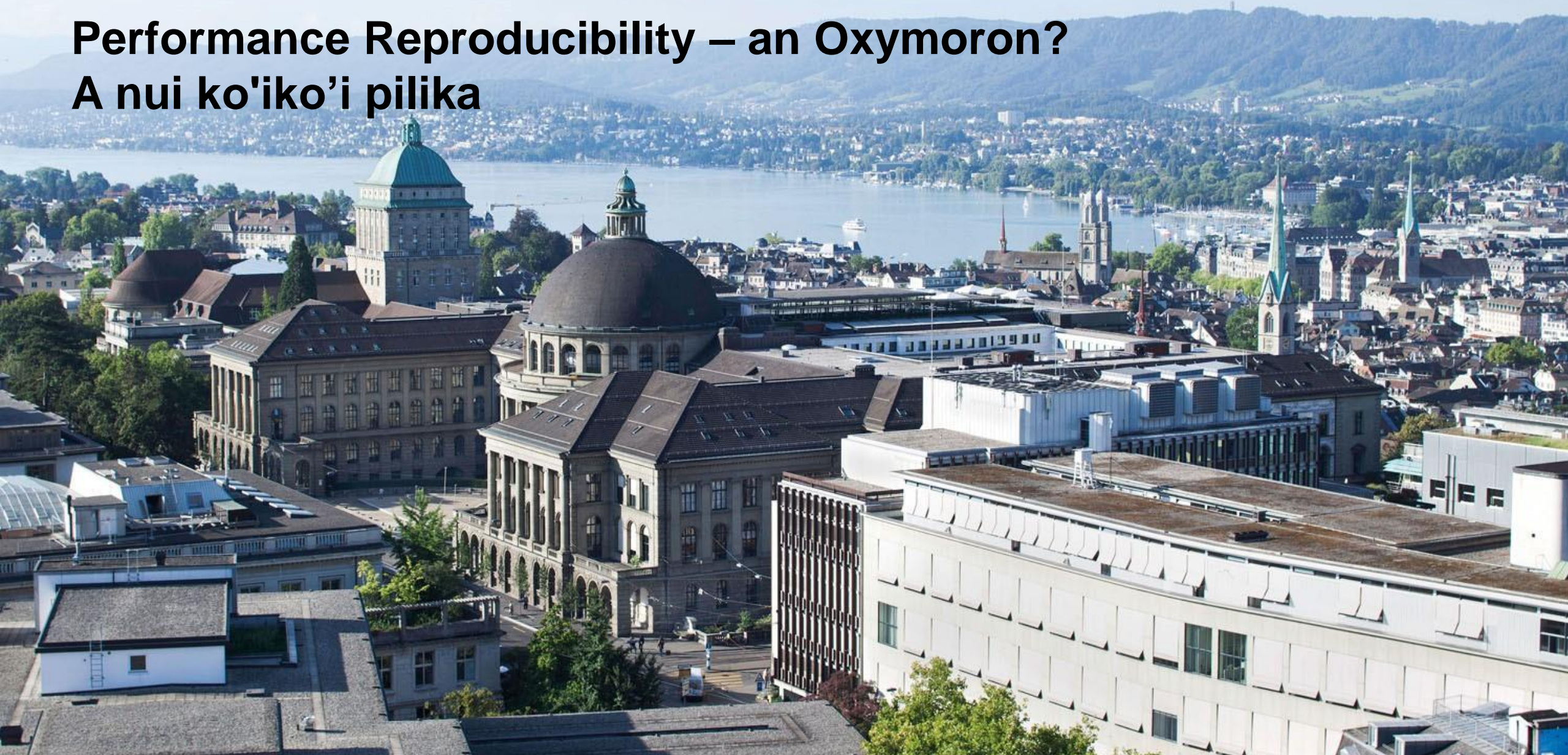


TORSTEN HOEFLER

Performance Reproducibility – an Oxymoron? A nui ko'iko'i pilika





OPINION

PNAS, Feb. 2015

Opinion: Reproducibility in Science

Jeffrey T. Leek
*Associate Professor of Biostatistics
 Johns Hopkins University*



“In the good old days physicists repeated each other’s experiments, just to be sure. Today they stick to FORTRAN, so that they can share each other’s programs, bugs included.” – Edsger Dijkstra (1930-2002), Dutch computer scientist, Turing Award 1972

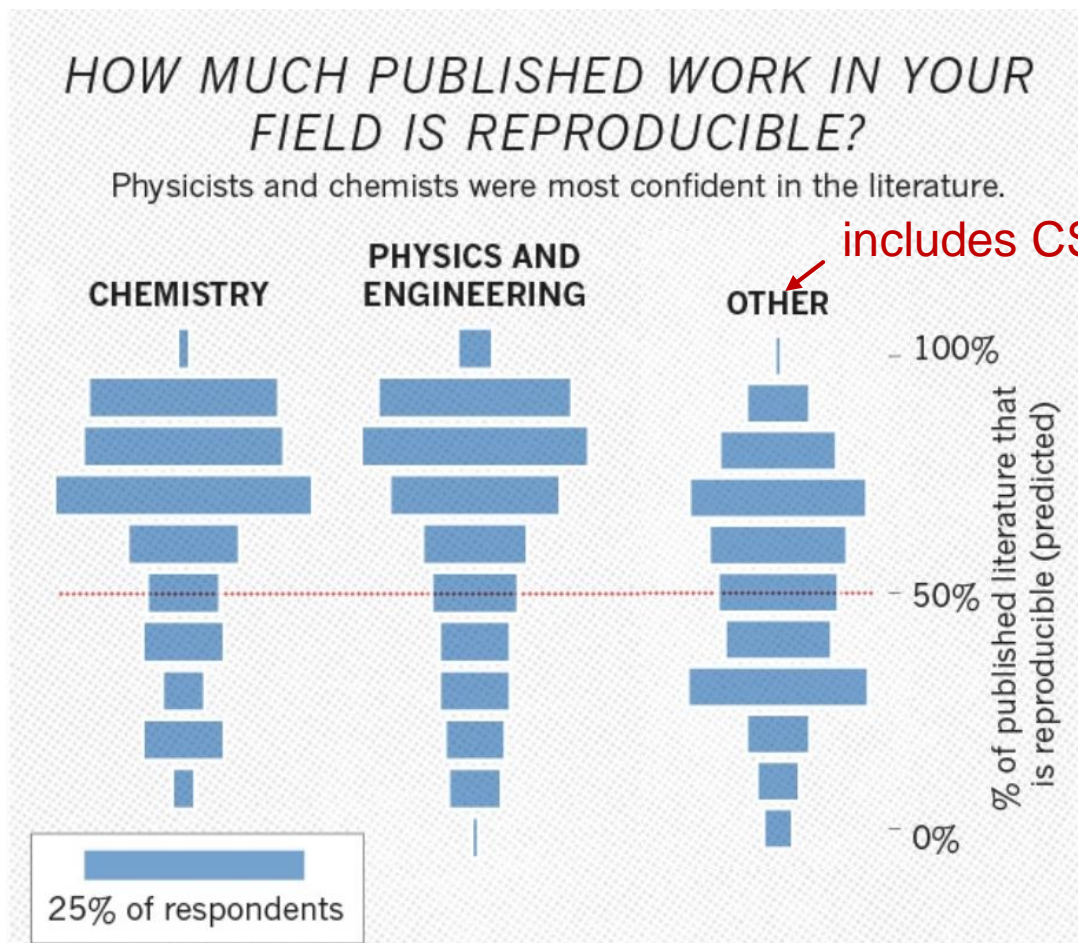
Reproducibility—repeating an experiment and getting the same result—is a cornerstone of scientific research. It is one of the primary means by which scientific evidence accumulates for or against a hypothesis. Yet, of late, there has been a crisis of confidence among researchers worried about the rate at which studies are either

been some very public failings of reproducibility across a range of disciplines from cancer genomics (3) to economics (4), and the data for many publications have not been made publicly available, raising doubts about the quality of data analyses. Popular press articles have raised questions about the reproducibility of all scientific research (5), and the US Congress has convened hearings focused on the transparency of scientific research (6). The result is that much of the

Unfortunately, the mere reproducibility of computational results is insufficient to address the replication crisis because even a reproducible analysis can suffer from many problems—confounding from omitted variables, poor study design, missing data—that threaten the validity and useful interpretation of the results. Although improving the reproducibility of research may increase the rate at which flawed analyses are uncovered, as recent high-profile examples have demonstrated (1), it does not change the fact that

Reproducibility and replicability?

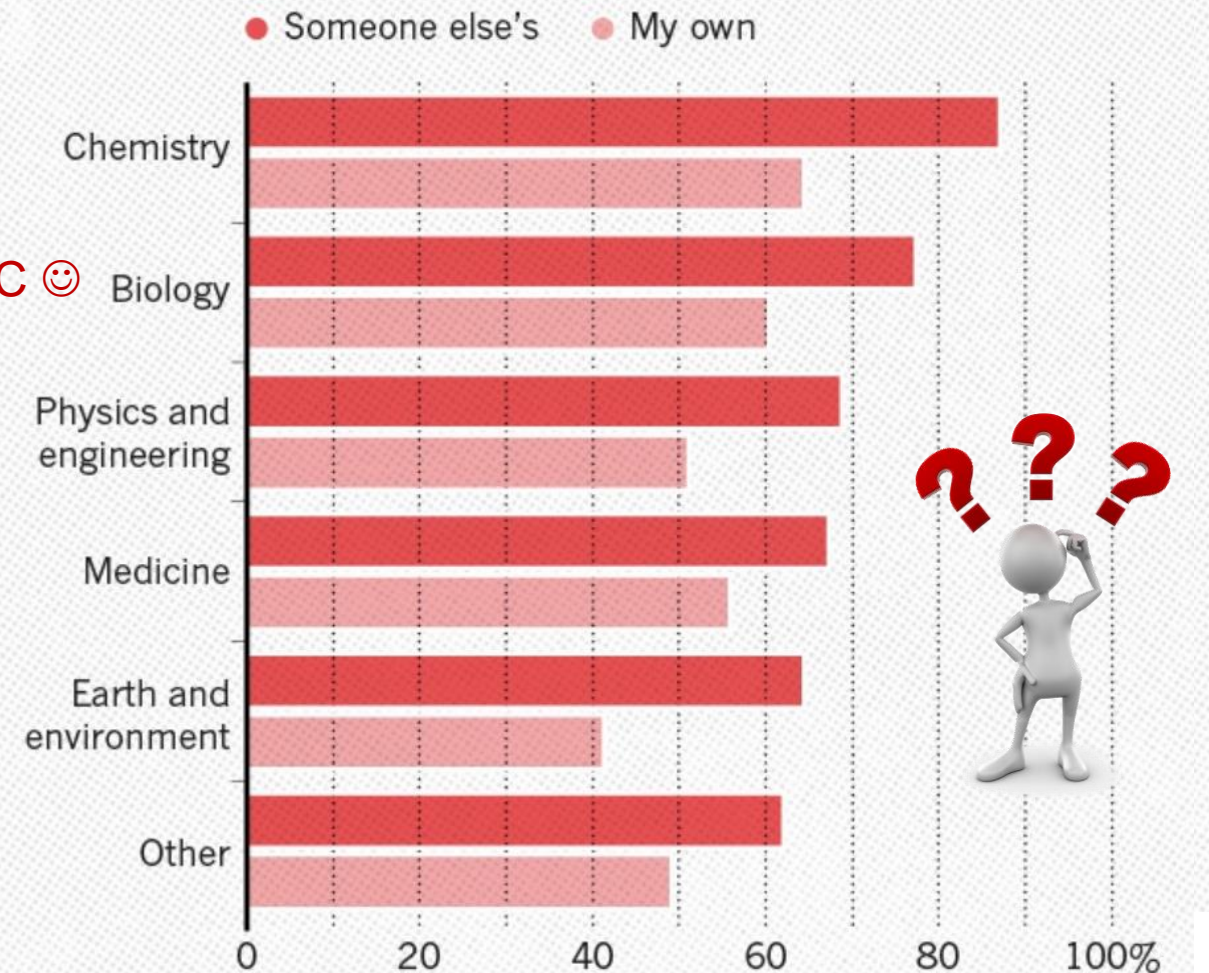
- **Reproducibility** – get the exact results
- **Replicability** – repeat the effect/insight



Nature, May 2016

HAVE YOU FAILED TO REPRODUCE AN EXPERIMENT?

Most scientists have experienced failure to reproduce results.



Functional reproducibility is relatively simple – release the code!



Notebook

Single-threaded, if you don't care much about performance

Gets a bit more complex when you share parallel codes (IEEE 754 is not associative)

IPDPS'14

Designing Bit-Reproducible Portable High-Performance Applications*

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Abstract—Bit-reproducibility has many advantages in the context of high-performance computing. Besides simplifying and making more accurate the process of debugging and testing the code, it can allow the deployment of applications on heterogeneous systems, maintaining the consistency of the computations. In this work we analyze the basic operations performed by scientific applications and identify the possible sources of non-reproducibility. In particular, we consider the tasks of evaluating transcendental functions and performing reductions using non-associative operators. We present a set of techniques to achieve reproducibility and we propose im-

runs is often of key importance in order to locate and isolate bugs. Especially, when refactoring an application in a way that the results should not change, reproducibility can significantly ease testing. However, debugging is only a secondary use-case for us. Many applications being run on large, parallel high performance computing facilities simulate the behavior of complex and highly non-linear systems. Prominent examples can be found in molecular dynamics or weather and climate simulation. For example, for weather

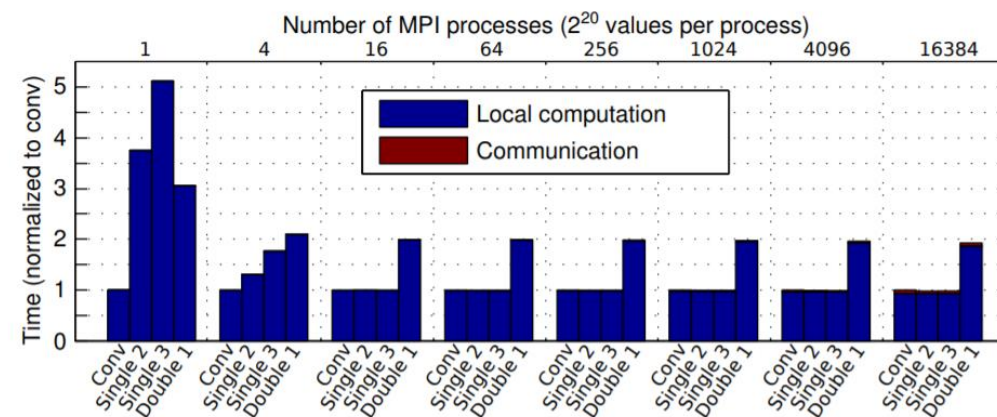
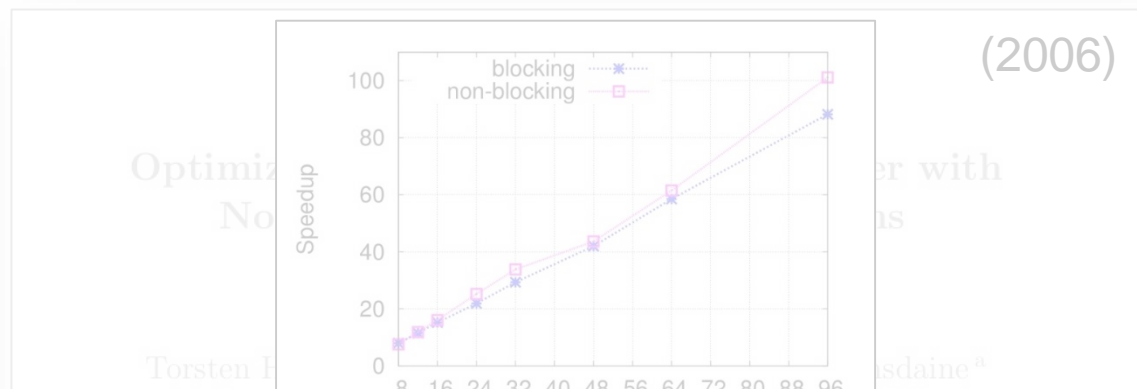


Figure 8. Performance comparison of conventional reduction performed with MKL (*Conv*), single-sweep reduction with two levels (*Single2*), with three levels (*Single3*) and double-sweep reduction with 1 level (*Double 1*) on varying number of processes, each owning 2^{20} double-precision values,

But what if performance is your science result?



Original findings:

- If carefully tuned, NBC speed up a 3D solver
Full code published
- 800³ domain – 4 GB (distributed) array

Deadline in a bit more than 24 hours, no extensions ☺

Reproducing performance results is hard! Is it even possible?



9 years later: attempt to reproduce ☺!

System A: 28 quad-core nodes, Xeon E5520

System B: 4 nodes, dual Opteron 6274

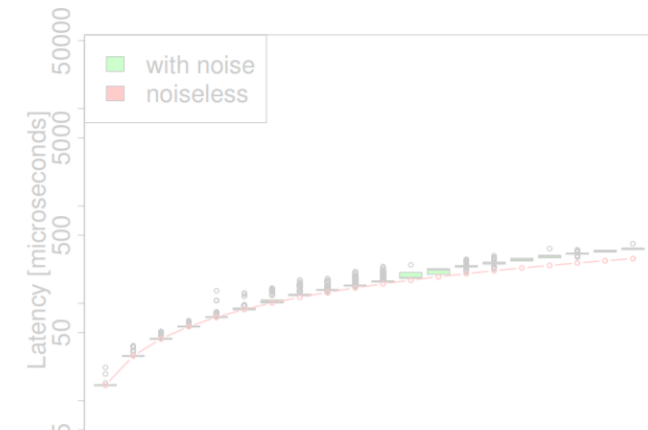
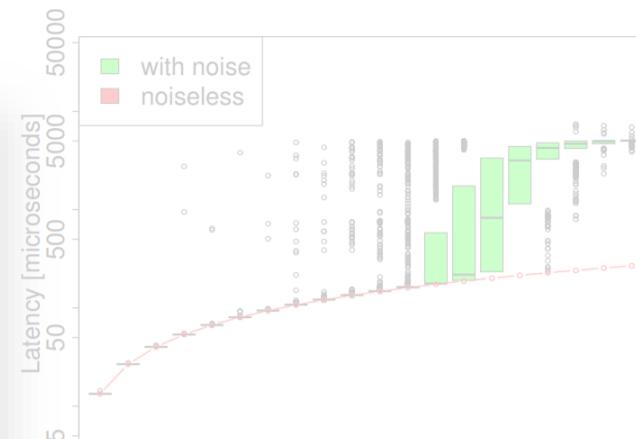
“Neither the experiment in A nor the one in B could reproduce the results presented in the original paper, where the usage of the NBC library resulted in a performance gain for practically all node counts, reaching a superlinear speedup for 96 cores (explained as being due to cache effects in the inner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

My own replication result

Characterizing the Influence of System Noise on Large-Scale Applications by Simulation

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 Indiana University
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Replicating performance results is possible but rare! Make it the default?

structure of the noise. Simulations with different network speeds show that a 10x faster network does not improve application scalability. We quantify noise and conclude that our tools can be utilized to tune the noise signatures of a specific system.

I. MOTIVATION AND BACKGROUND

The performance impact of operating system and architectural overheads (*system noise*) at massive scale is increasingly of concern. Even small local delays on compute nodes, which can be caused by interrupts, operating system daemons, or even cache or page misses, can affect global application performance significantly [1]. Such local delays often cause less than 1% overhead per process but severe performance losses can occur if noise is propagated (*amplified*) through communication or global synchronization. Previous analyses generally assume that the performance impact of system noise grows at scale and Tsafir et al. [2] even suggest that the

is supported directly by the BG/L hardware, allreduce used a pattern similar to the dissemination pattern. We use LogGP parameters from BlueGene/P running CNL because we do not have access to a BlueGene/L. Thus, we expect the impact to be slightly lower, but asymptotically similar. Like Beckman et al., we used unsynchronized noise with a fixed frequency of 1,000, 100, and 10 Hz causing detours of 16, 50, 100, and 1000 microseconds. ⁴<http://www.unixer.de/LogGOPSim> (2010)

"[...] a collective communication call may, or may not, have the effect of synchronizing all calling processes. This statement excludes, of course, the barrier function." This invalidates all simple models in use today. The synchronization properties of an application depend on the collective algorithm, point-to-point messaging, and the system's network parameters.

We chose a simulation approach similar to Sottile et al.'s [8] and improve it by using noise traces from existing systems combined with detailed simulation and extrapolation of collec-

results from Perera, Bridges, Brightwell as well as Beckman et al. both two years earlier on different machines



HPC Performance reproducibility – don't even try?

- ~~Reproducibility – get the exact results~~
- ~~Replicability – repeat the effect/insight~~

Small Quiz

Raise your hand if you believe one can reproduce any Gordon Bell finalist from before 2013!

HAVE YOU FAILED TO REPRODUCE
AN EXPERIMENT?



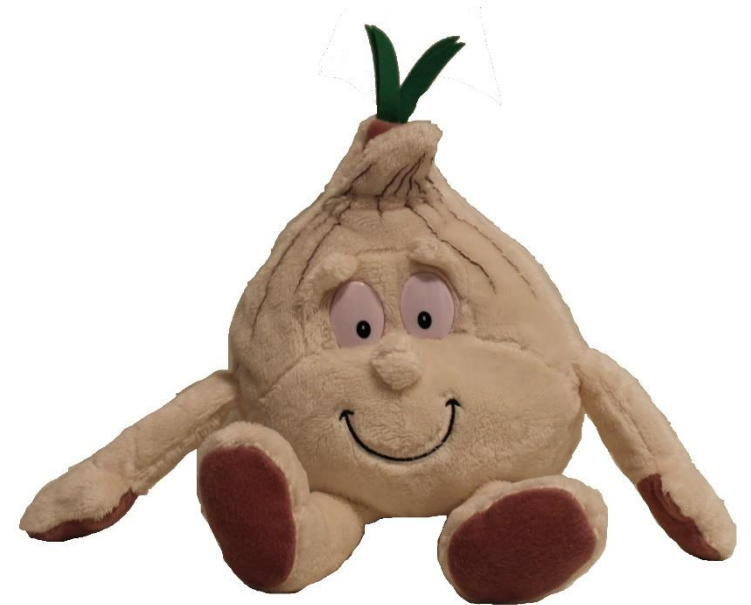
Interpretability: *We call an experiment interpretable if it provides enough information to allow scientists to understand the experiment, draw own conclusions, assess their certainty, and possibly generalize results.*

25% of respondents

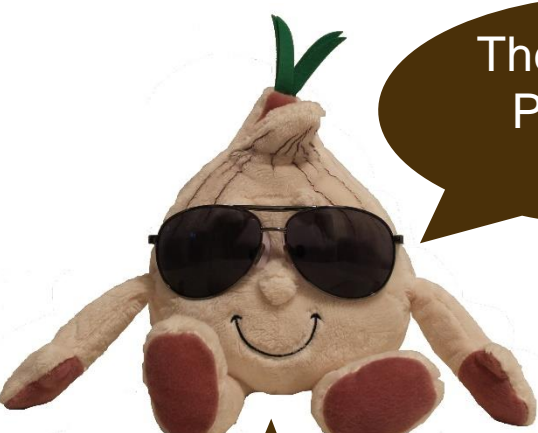
0 20 40 60 80 100%

How does Garth measure and report performance?

- We are 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?”
 - ...
- I asked: “How are they answered in the field today?”
 - “Experience”
 - “Gut feeling”
 - “Clever observation”
 - ...

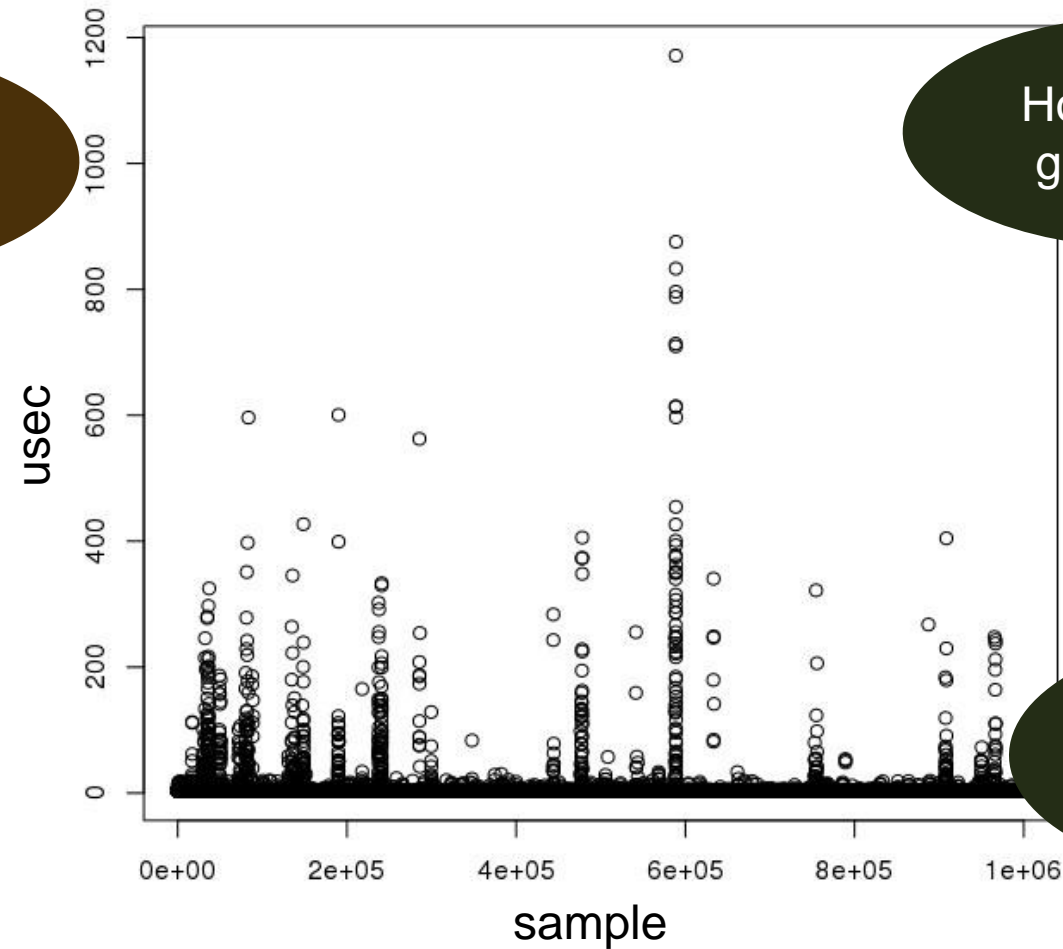



The simplest networking question: ping pong latency!



The latency of Piz Dora is 1.77us!

I averaged 10^6 tests, it must be right!

How did you get to this?

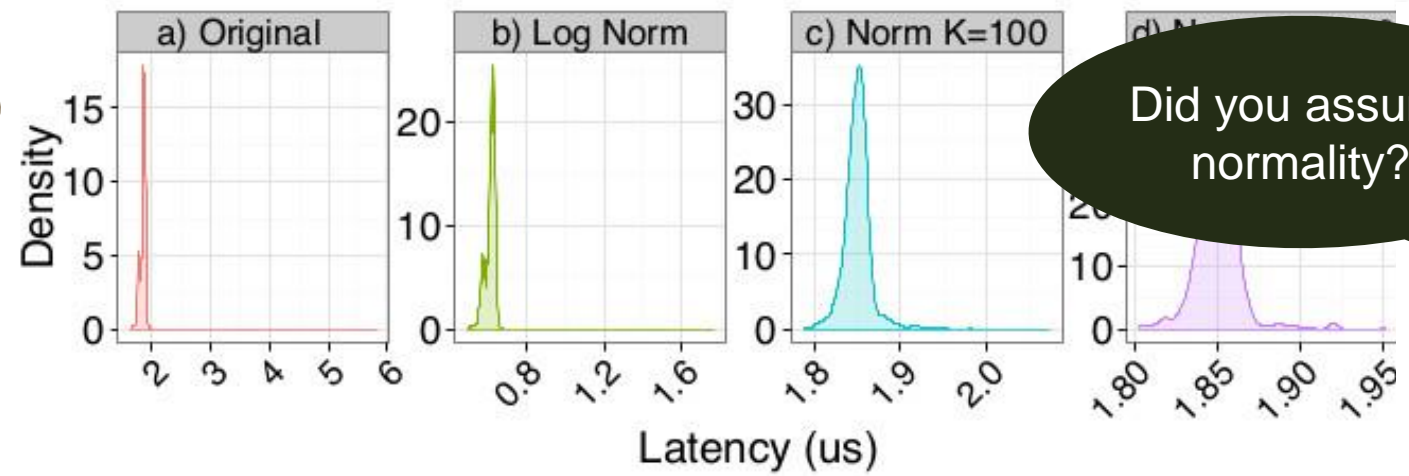
Why do you think so? Can I see the data?

Thou shalt not trust your average textbook!

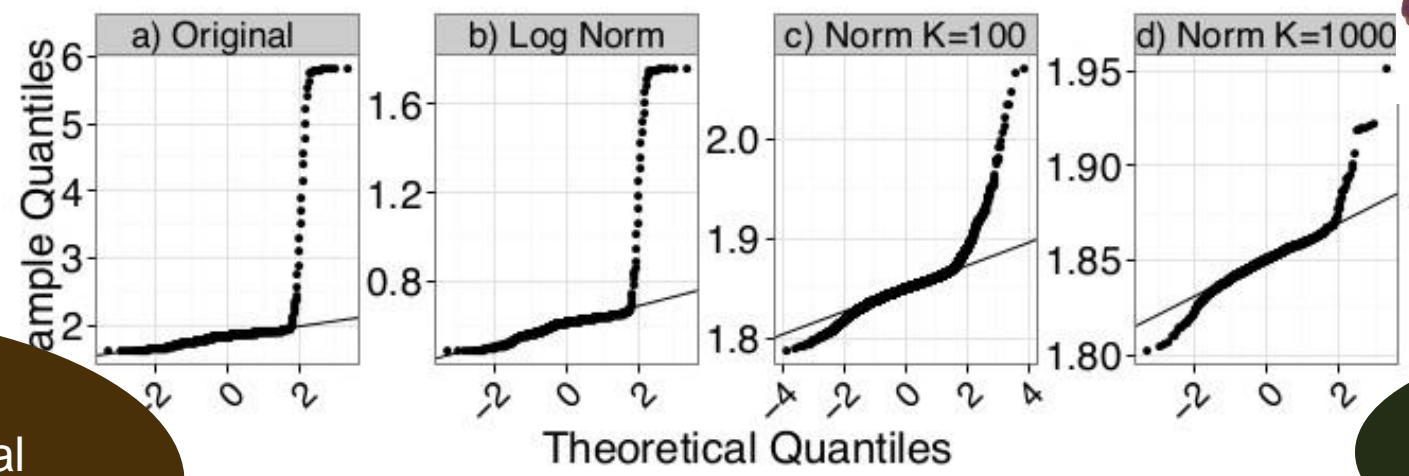
The confidence interval is 1.765us to 1.775us



Ugs, the data is not normal at all! The real CI is actually 1.6us to 1.9us!



Did you assume normality?



Can we test for normality?

Thou shalt not trust your system!

Look what data I got!

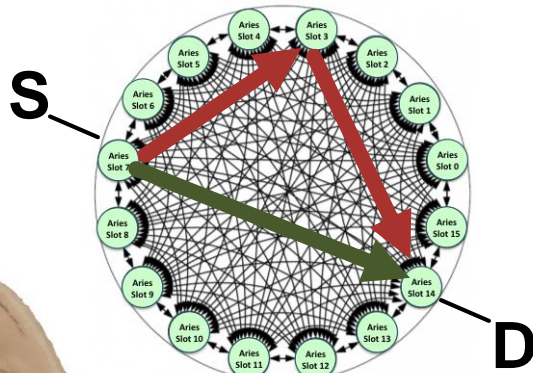
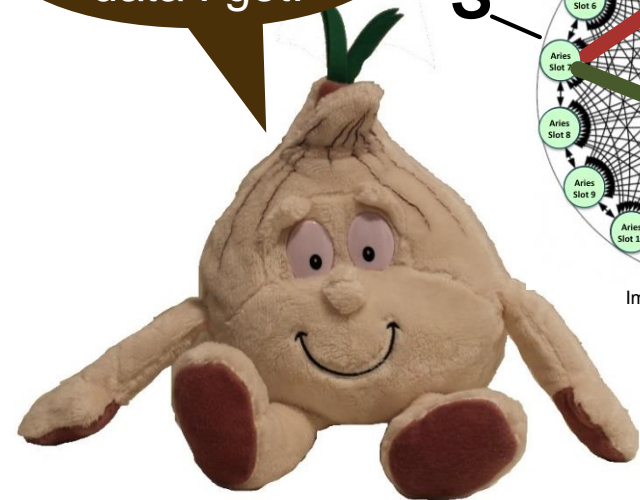
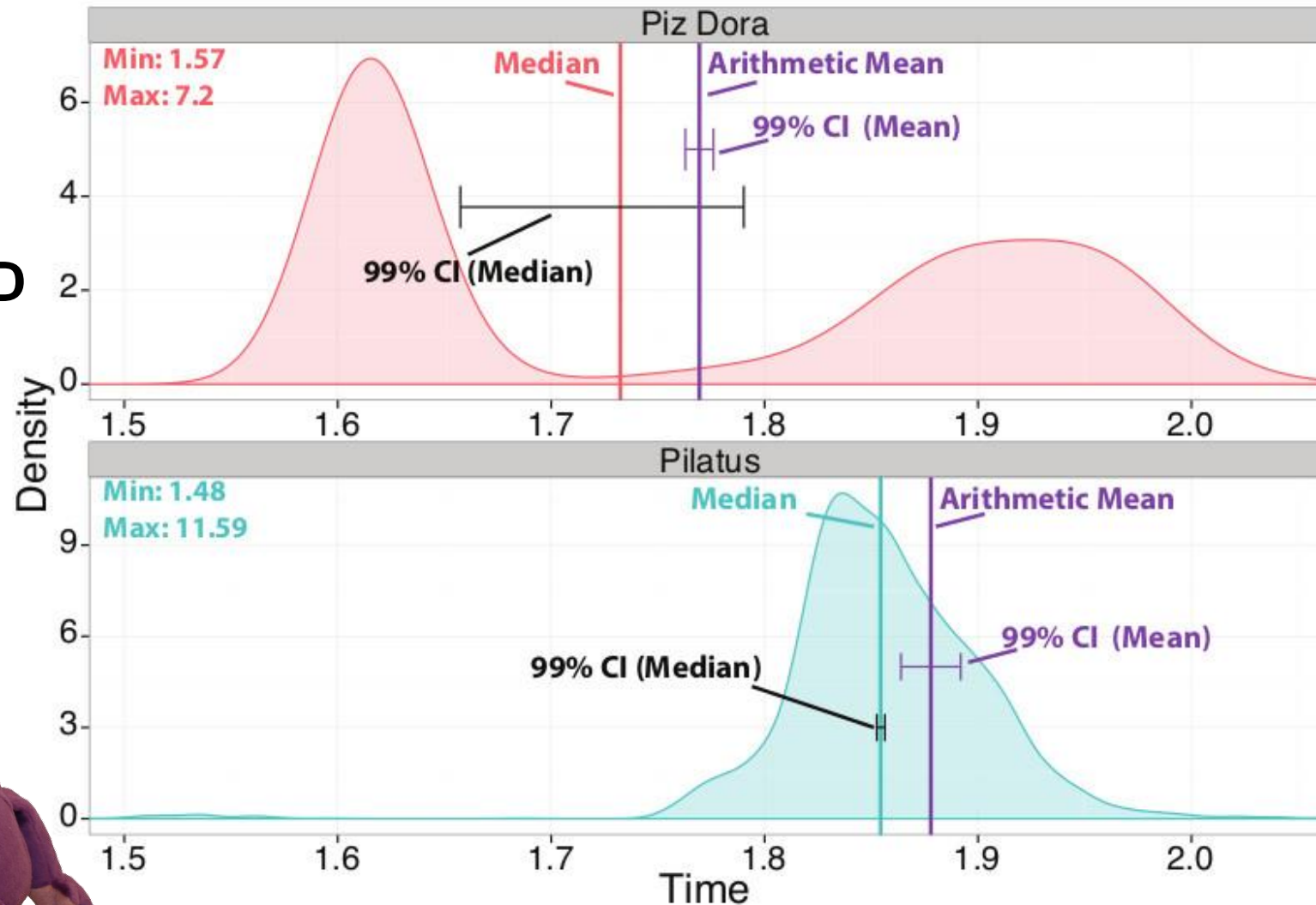


Image credit: nersc.gov

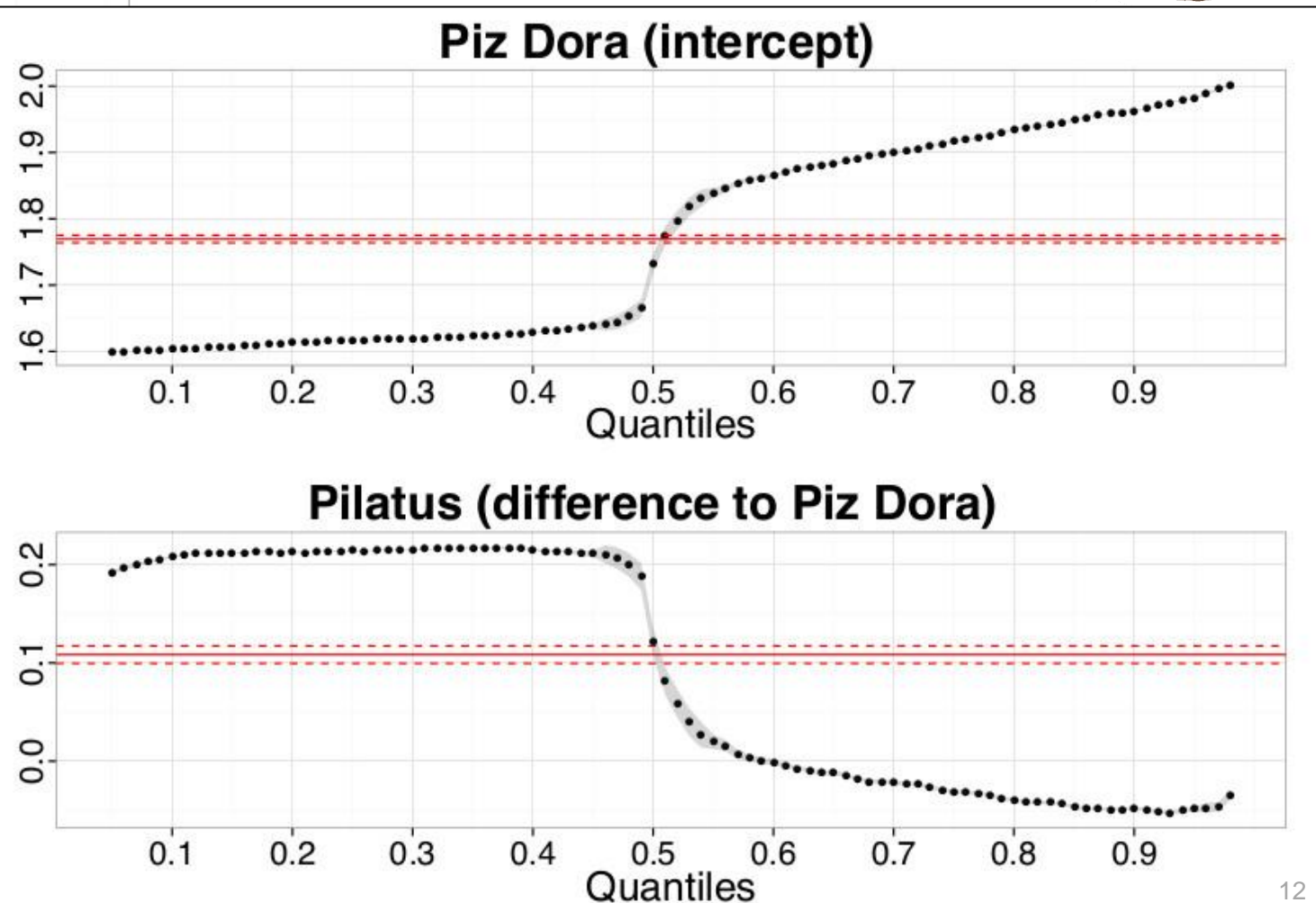
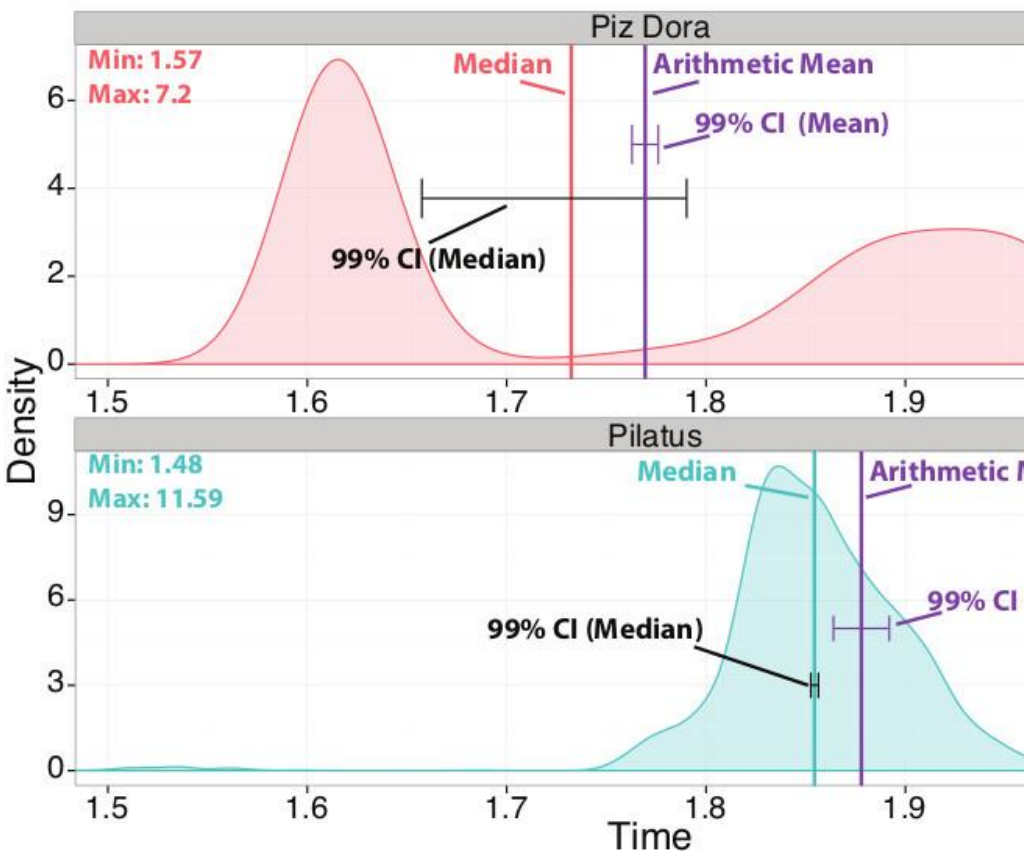
Clearly, the mean/median are not sufficient!

Try quantile regression!



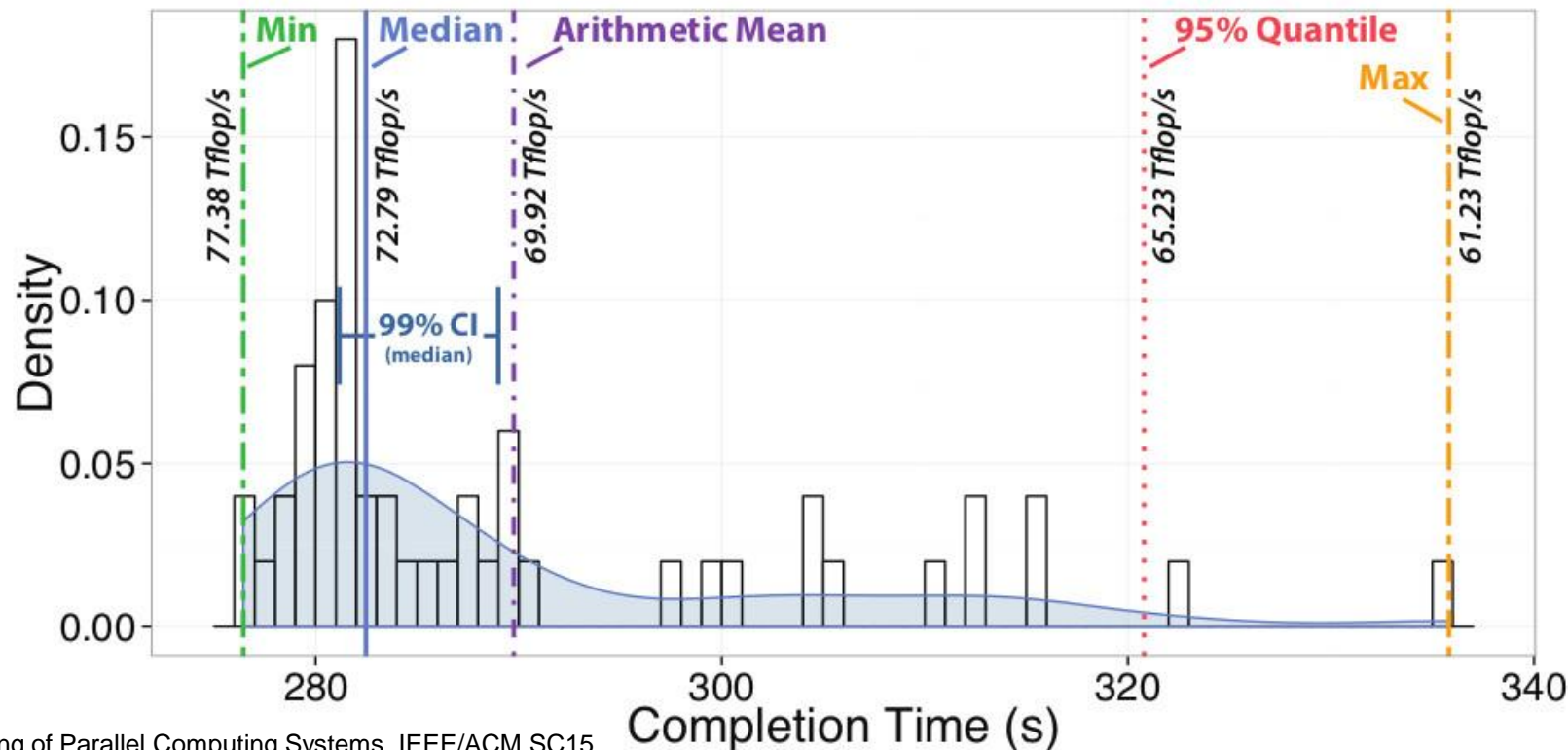
Quantile regression

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



A note on good scientific HPC practice ☺ - HPL

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



But it's ok, HPC people can laugh about ourselves!

1991 – the classic!



Twelve Ways to Fool the Masses When Giving Performance Results on Parallel Computers



2012 – the shocking

Abstract

Many of us quite difficult supercomputing scientific papers these results

How to Pitfall

2013 – the extension



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

Gerhard Wellein^(1,2), Georg Hager⁽²⁾

⁽¹⁾Department for Computer Science

⁽²⁾Erlangen Regional Computing Center

Friedrich-Alexander-Universität Erlangen-Nürnberg



Yes, this is a garlic press!



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

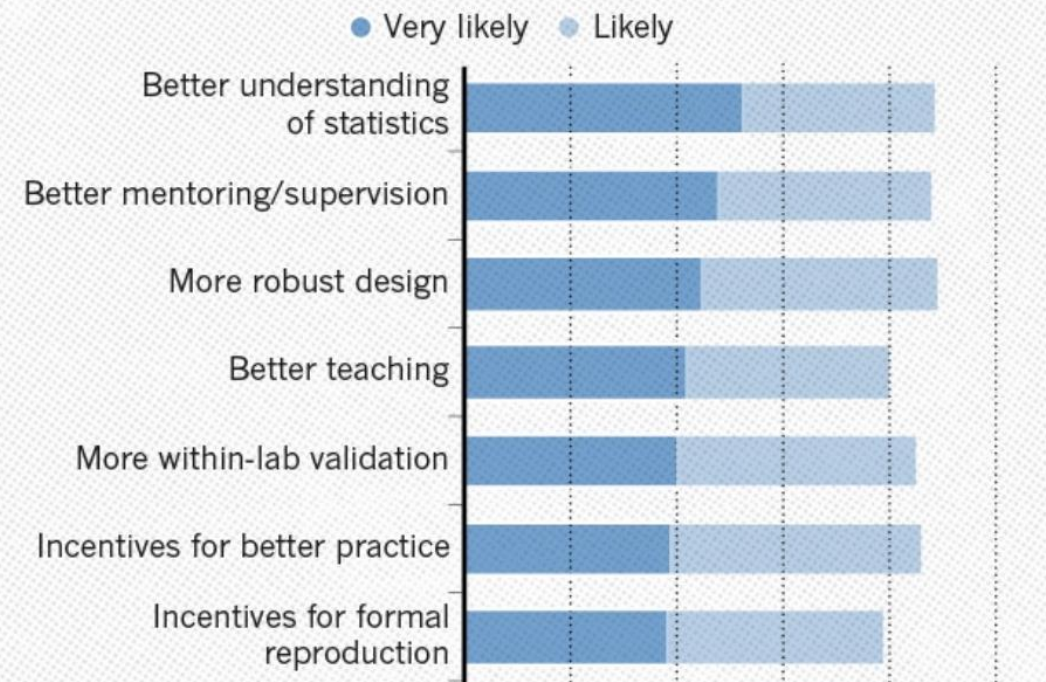
- **Attempt to emphasize interpretability of performance experiments**
 - Teach some basic statistics

- **The set of 12 rules is not complete**
 - And probably never will be
 - Intended to serve as a solid start
 - Call to the community to extend it

Nature, 2016

WHAT FACTORS COULD BOOST REPRODUCIBILITY?

Respondents were positive about most proposed improvements but emphasized training in particular.



Scientific Benchmarking of Parallel Computing Systems

Twelve ways to tell the masses when reporting performance results

Torsten Hoefler
Dept. of Computer Science
ETH Zurich
Zurich, Switzerland
htor@inf.ethz.ch

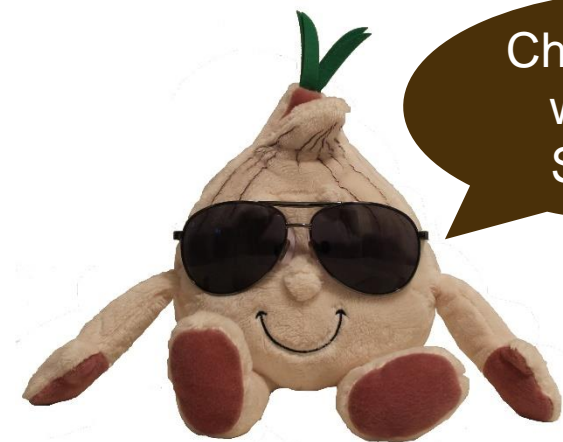
Roberto Belli
Dept. of Computer Science
ETH Zurich
Zurich, Switzerland
bellir@inf.ethz.ch

ABSTRACT

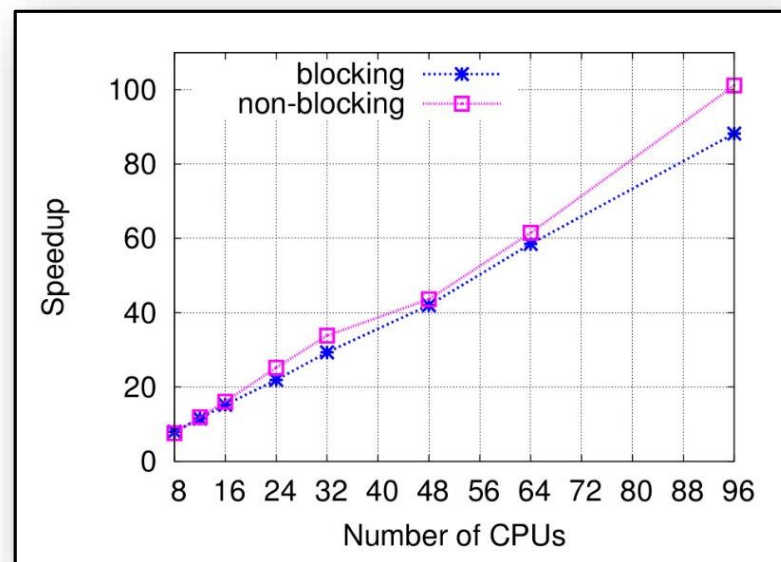
Measuring and reporting performance of parallel computers constitutes the basis for scientific advancement of high-performance

Reproducing experiments is one of the main principles of the scientific method. It is well known that the performance of a computer program depends on the application, the input, the compiler, the

The most common issue: speedup plots



Check out my wonderful Speedup!



- **Most common and oldest-known issue**

- First seen 1988 – also included in Bailey’s 12 ways
- 39/120 checked papers reported speedups
15 (38%) did not specify the base-performance 😞
- Recently rediscovered in the “big data” universe

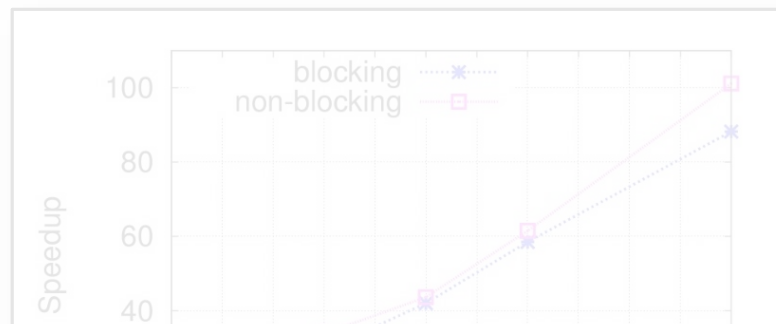
A. Rowstron et al.: Nobody ever got fired for using Hadoop on a cluster, HotCDP 2012

F. McSherry et al.: Scalability! but at what cost?, HotOS 2015



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.

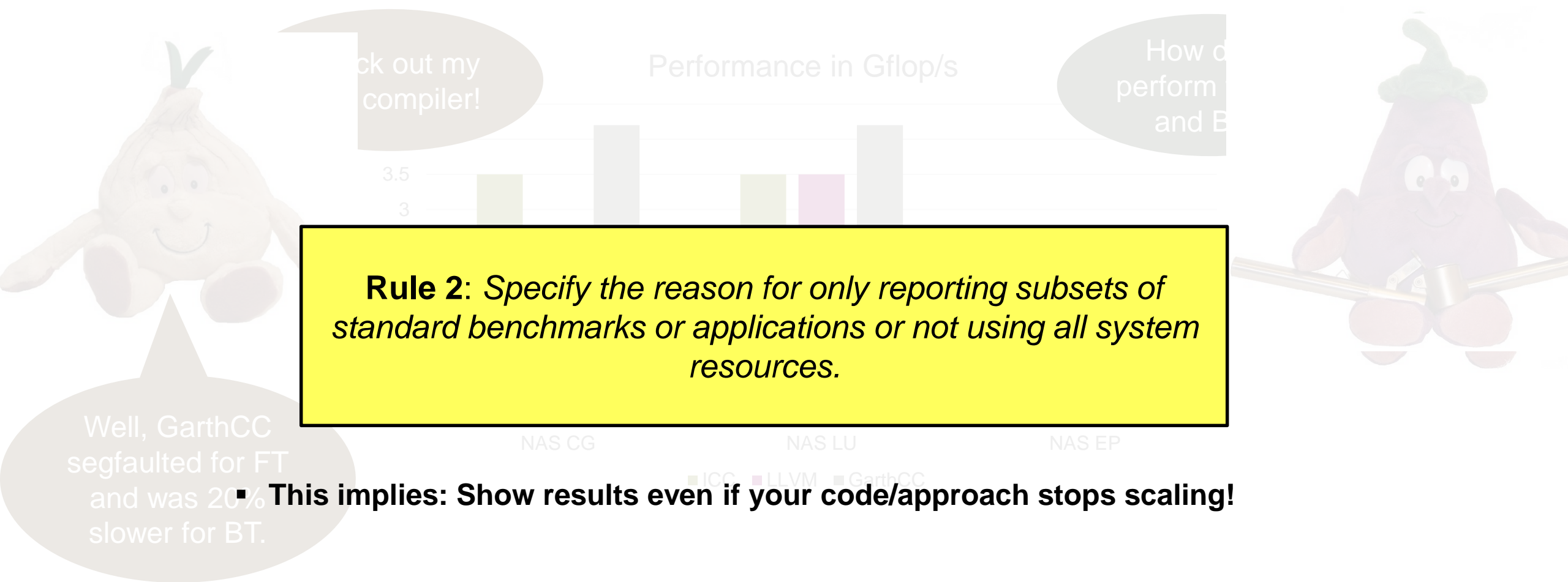
Most common and oldest known issue

- First seen 1988 – also included in Bailey’s 12 ways
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- 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

Garth's new compiler optimization



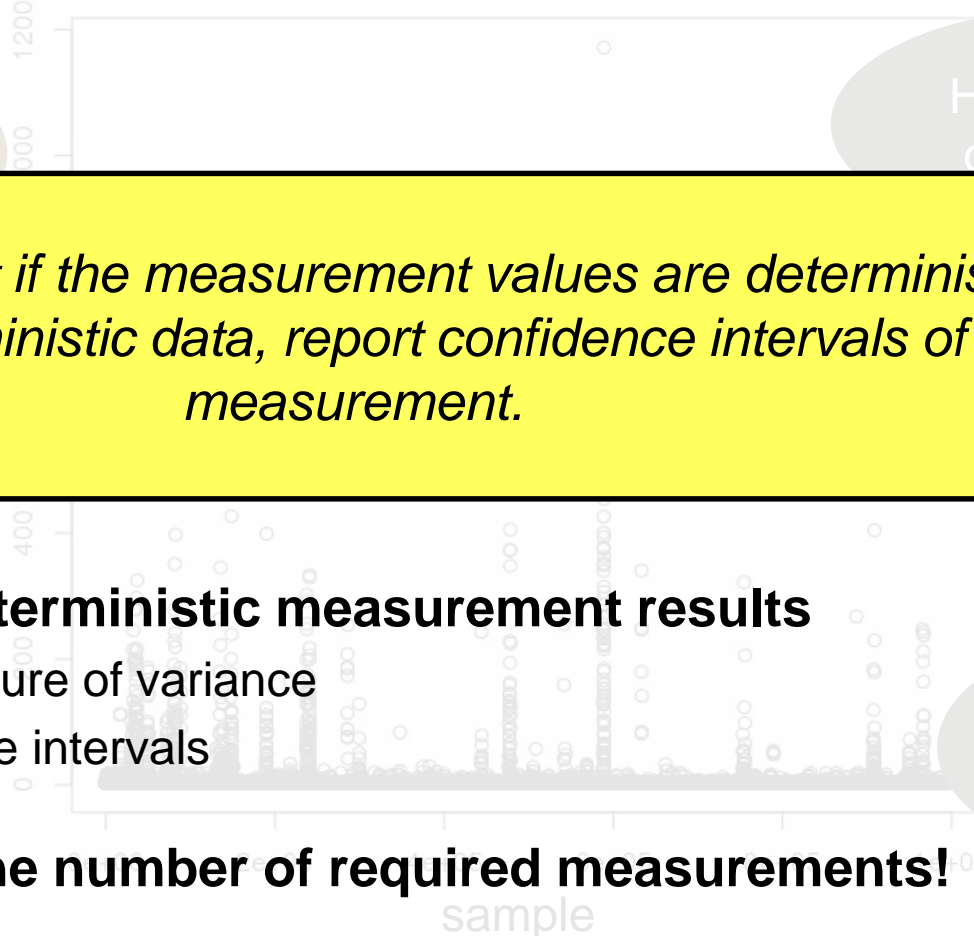
The mean parts of means – or how to summarize data

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.*

- **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**

Dealing with variation



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:**

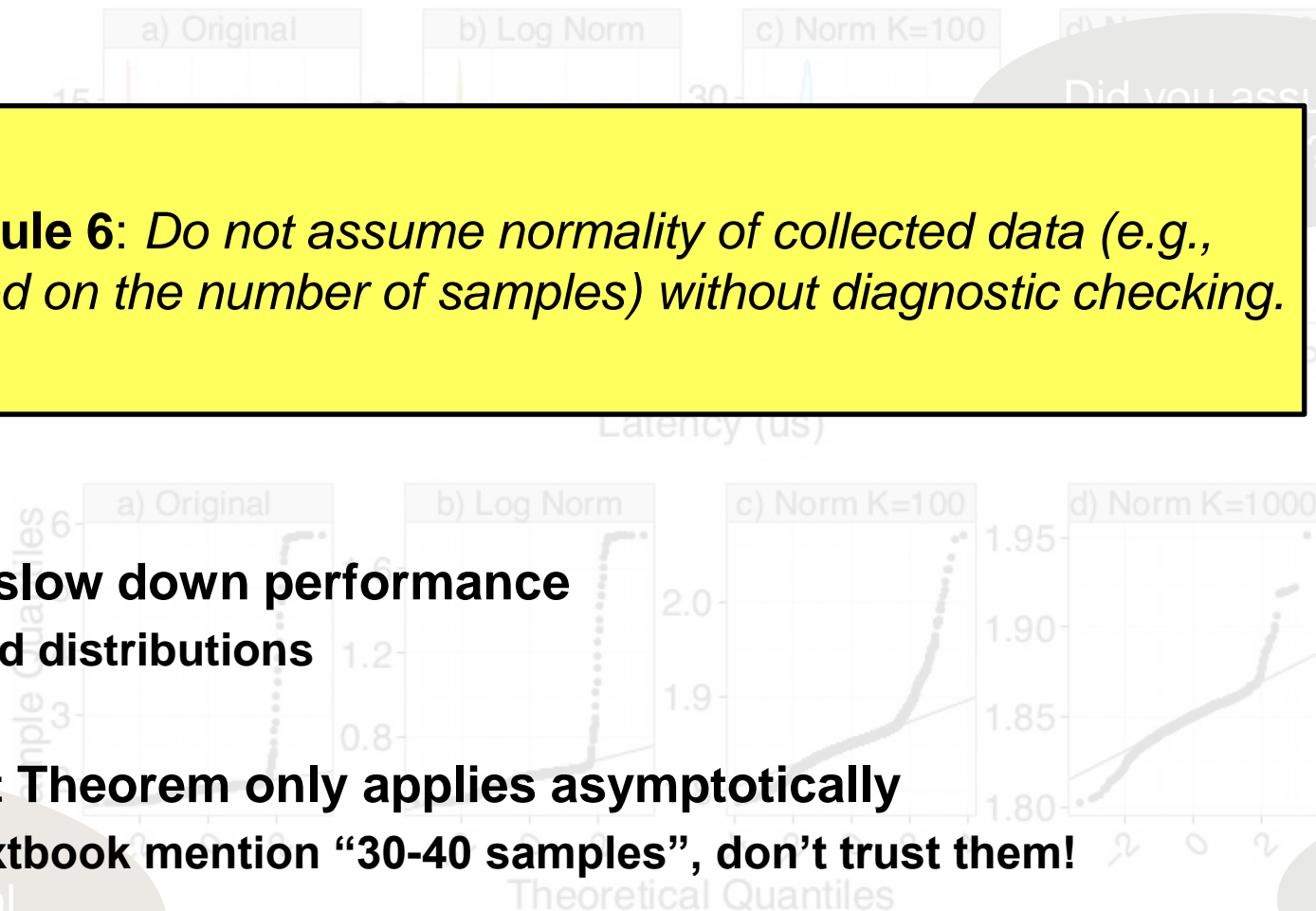
“We collected measurements until the 99% confidence interval was within 5% of our reported means.”

Dealing with variation

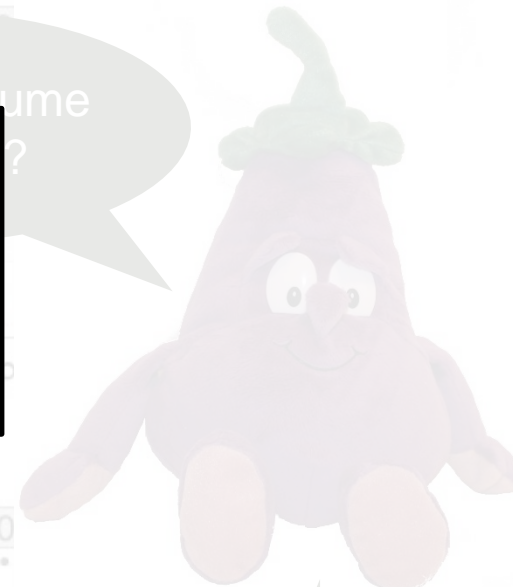
The confidence interval is 1.765us to 1.775us

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**



Did you assume ...?



Can we test for normality?

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*



Time in parallel systems

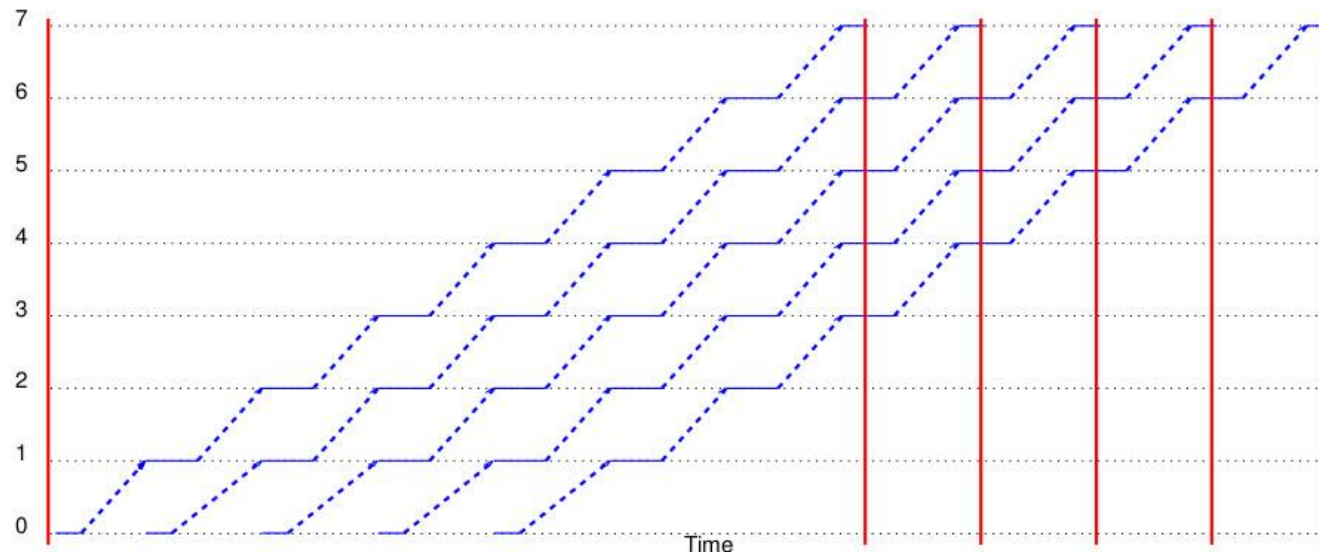


My simple broadcast takes only one latency!

But I measured it so it must be true!

```

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



That's nonsense!



...
Measure each operation separately!

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.*

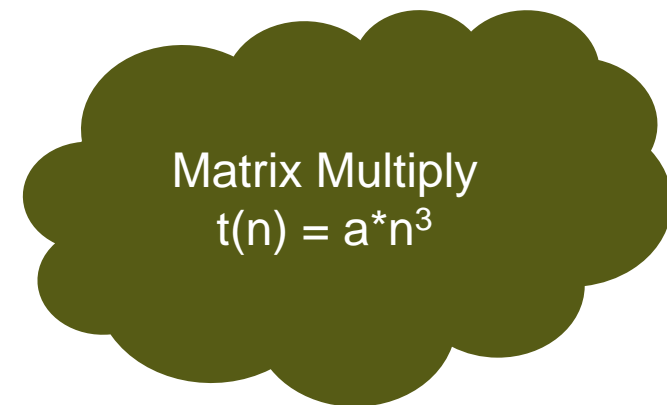
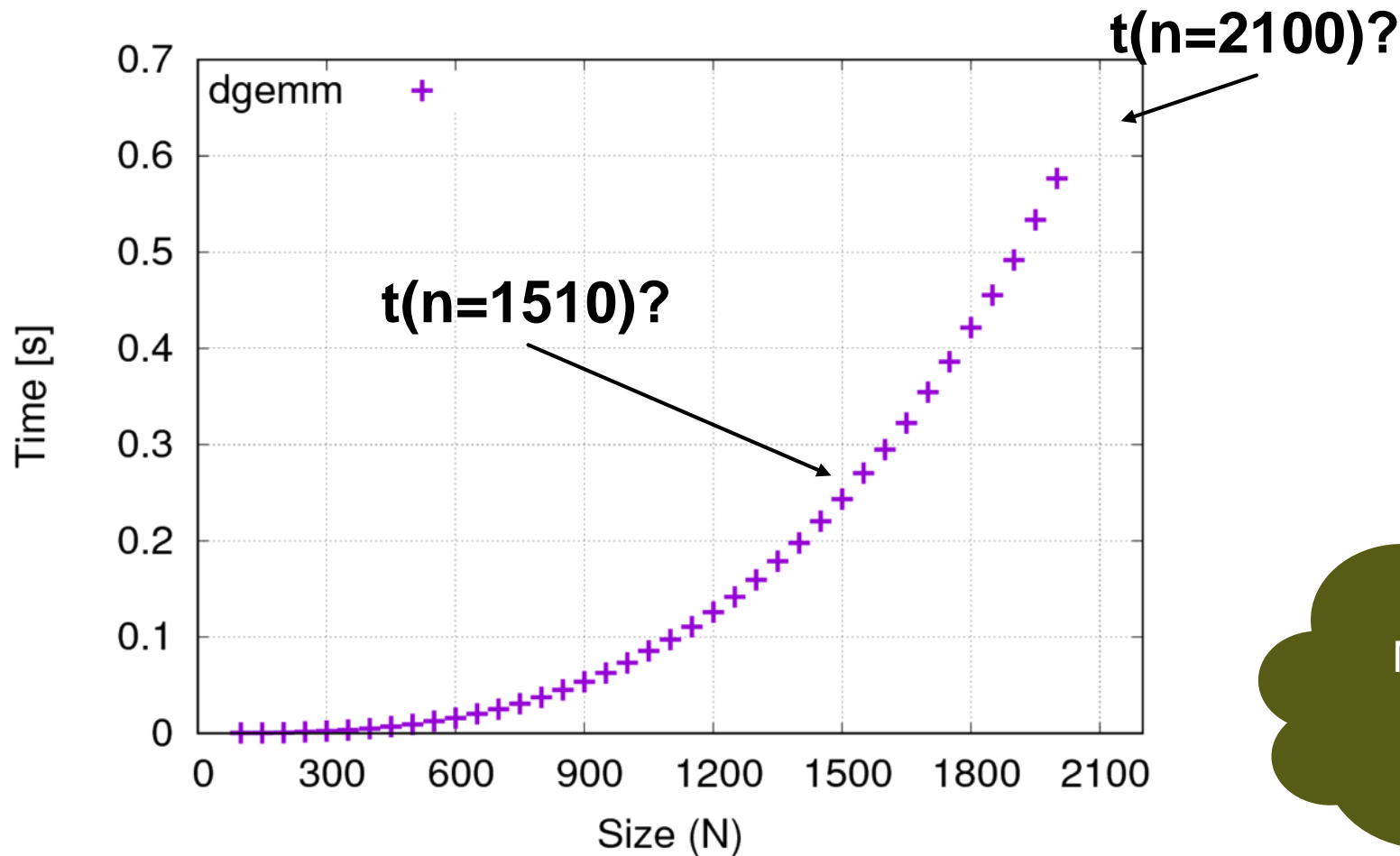
- **Measure events separately**
 - Use high-precision timers
 - Synchronize processes
- **Summarize across processes:**
 - Min/max (unstable), average, median – depends on use-case

Somebody Time 100-

whiskers depict the 1.5 IQR

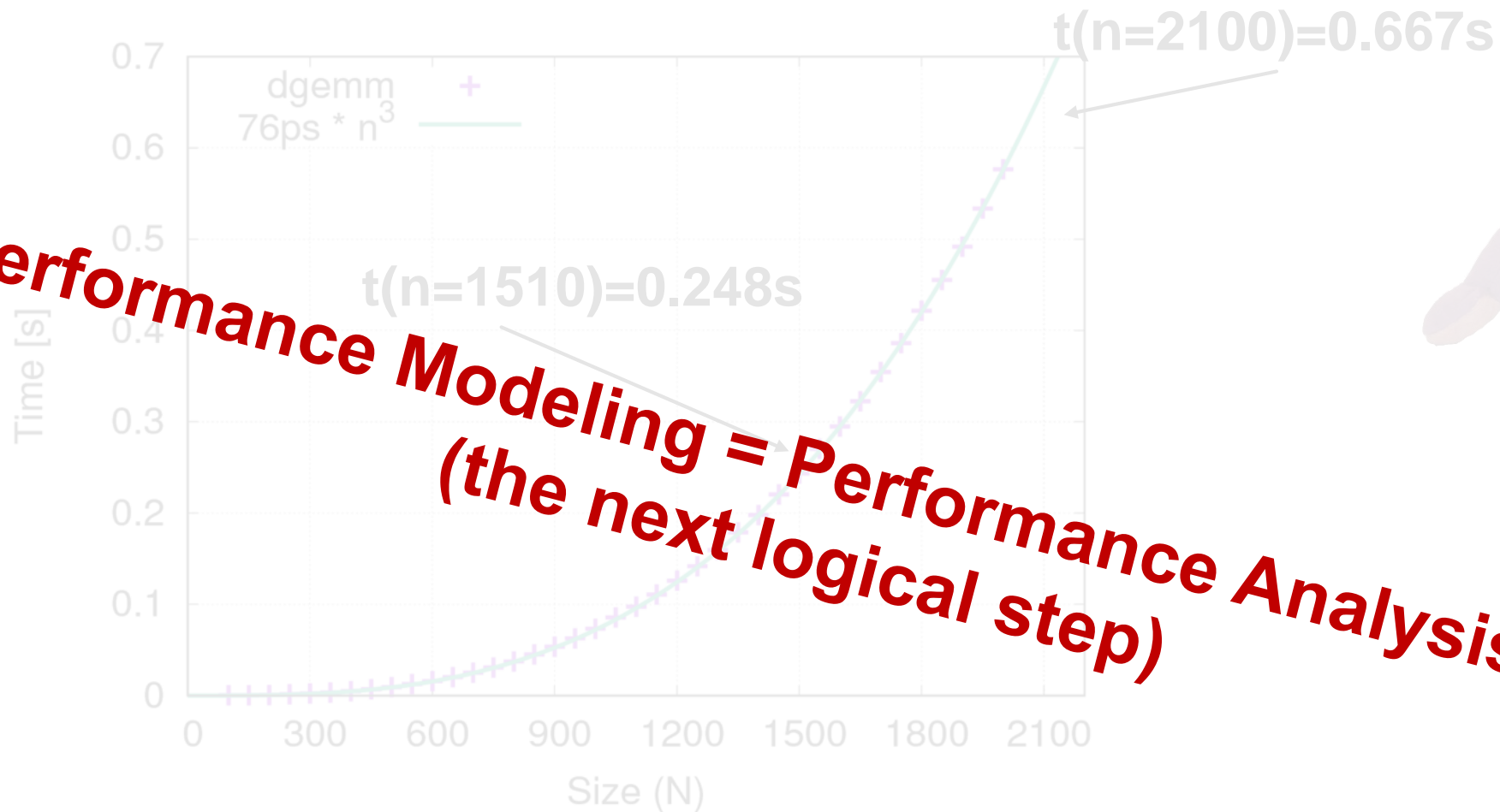
Processes

We have the (statistically sound) data, now what?



The 99% confidence interval is within 1% of the reported median.

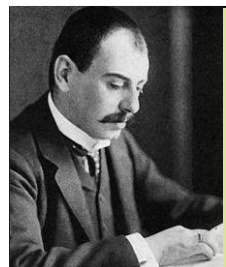
We have the (statistically sound) data, now what?



The 99% confidence interval is within 1% of the reported median.
The adjusted R² of the model fit is 0.99

Conclusions and call for action

- HPC performance is not reproducible
- **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]



My inner mathematician to the HPC crowd:

Landau really thought about this hard 😊

$$O(100) = O(10) = O(10^5) = O(0.5) = O(1)$$

Mahalo Keli'i!

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

and a shameless plug!

Demystifying Parallel and Distributed Deep Learning: An In-Depth Concurrency Analysis

TAL BEN-NUN* and TORSTEN HOEFLER, ETH Zurich

Deep Neural Networks (DNNs) are becoming an important tool in modern computing applications. Accelerating their training is a major challenge and techniques range from distributed algorithms to low-level circuit design. In this survey, we describe the problem from a theoretical perspective, followed by approaches for its parallelization. Specifically, we present trends in DNN architectures and the resulting implications on parallelization strategies. We discuss the different types of concurrency in DNNs; synchronous and asynchronous stochastic gradient descent; distributed system architectures; communication schemes; and performance modeling. Based on these approaches, we extrapolate potential directions for parallelism in deep learning.

CCS Concepts: • **General and reference** → *Surveys and overviews*; • **Computing methodologies** → **Neural networks**; **Distributed computing methodologies**; **Parallel computing methodologies**; *Machine learning*;

Additional Key Words and Phrases: Deep Learning, Distributed Computing, Parallel Algorithms

ACM Reference format:

Tal Ben-Nun and Torsten Hoefler. 2018. Demystifying Parallel and Distributed Deep Learning: An In-Depth Concurrency Analysis. 60 pages.

1 INTRODUCTION

Machine Learning, and in particular Deep Learning [LeCun et al. 2015], is a field that is rapidly taking over a variety of aspects in our daily lives. In the core of deep learning lies the Deep Neural Network (DNN), a construct inspired by the interconnected nature of the human brain. Trained properly, the expressiveness of DNNs provides accurate solutions for problems previously thought to be unsolvable, simply by observing large amounts of data. Deep learning has been successfully implemented for a plethora of subjects, ranging from image classification [Huang et al. 2017], through speech recognition [Amodei et al. 2016] and medical diagnosis [Cireşan et al. 2013], to autonomous driving [Bojarski et al. 2016] and defeating human players in complex games [Silver et al. 2017] (see Fig. 1 for more examples).

Give times a meaning!

I compute 10^{10} digits of Pi

I have no clue.

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

Can you provide?

- Ideal speedup
- Amdahl's speedup
- Parallel overheads

- **Model computer system as k-dimensional space**

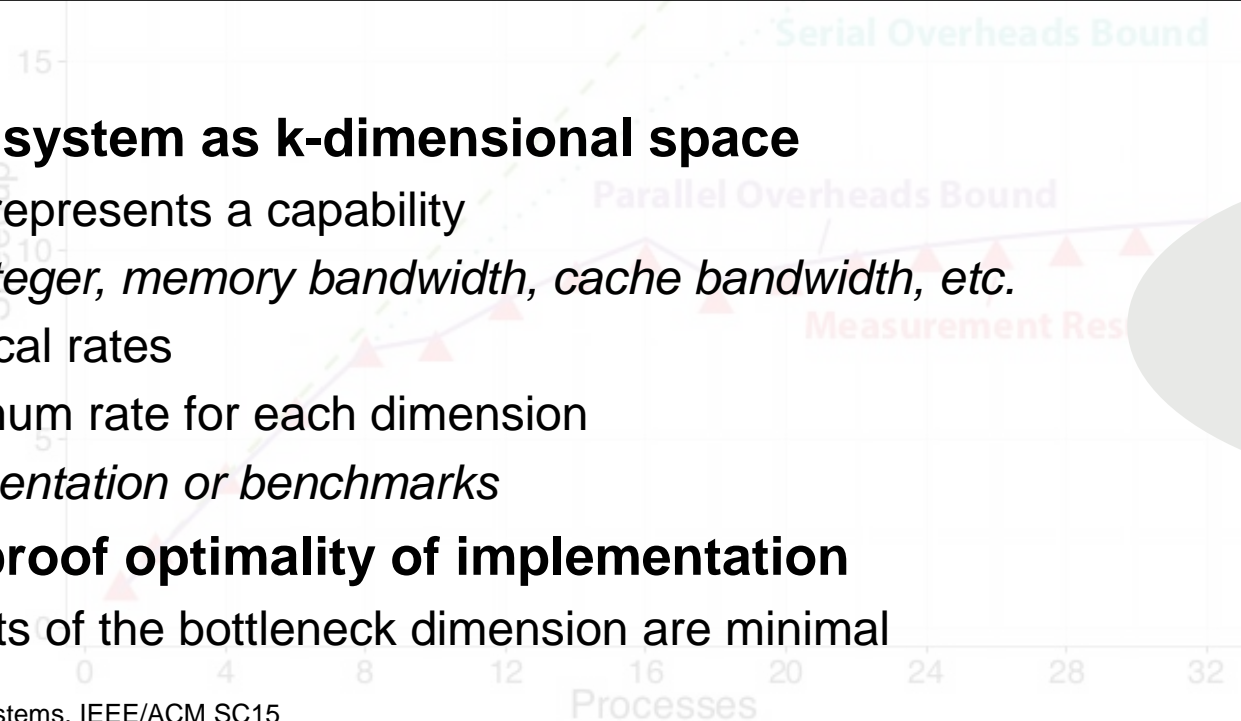
- Each dimension represents a capability
Floating point, Integer, memory bandwidth, cache bandwidth, etc.

Ok: The 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



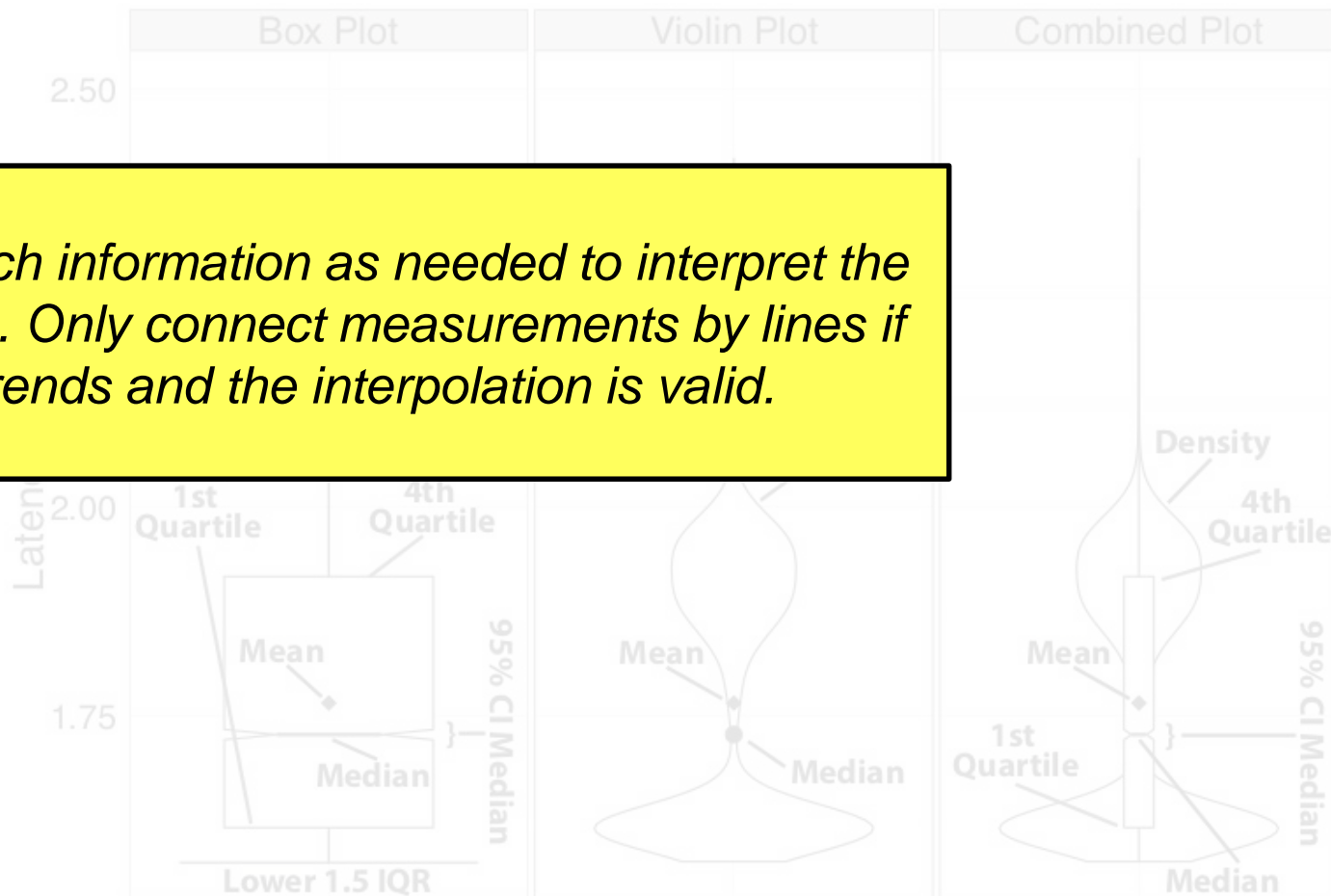
Plot as much information as possible!

My most common request was "show me the data"

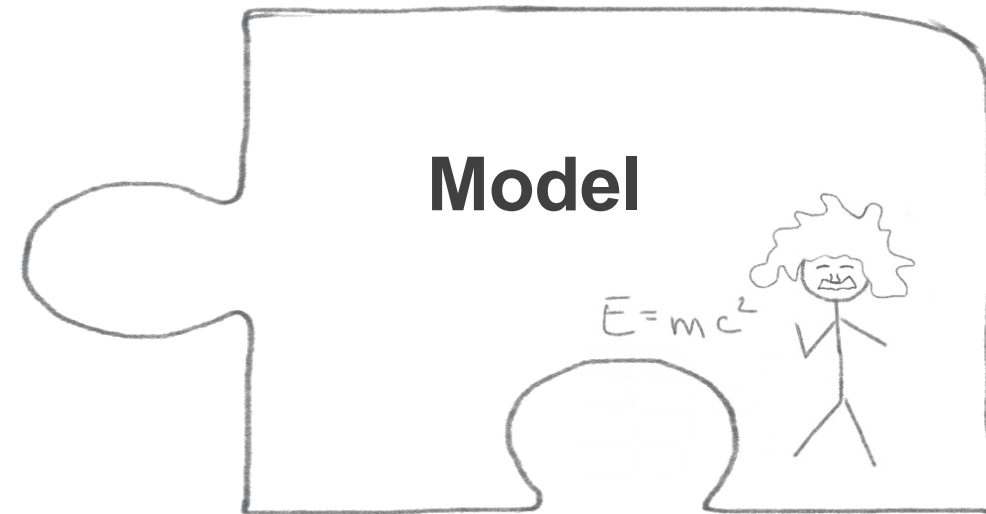
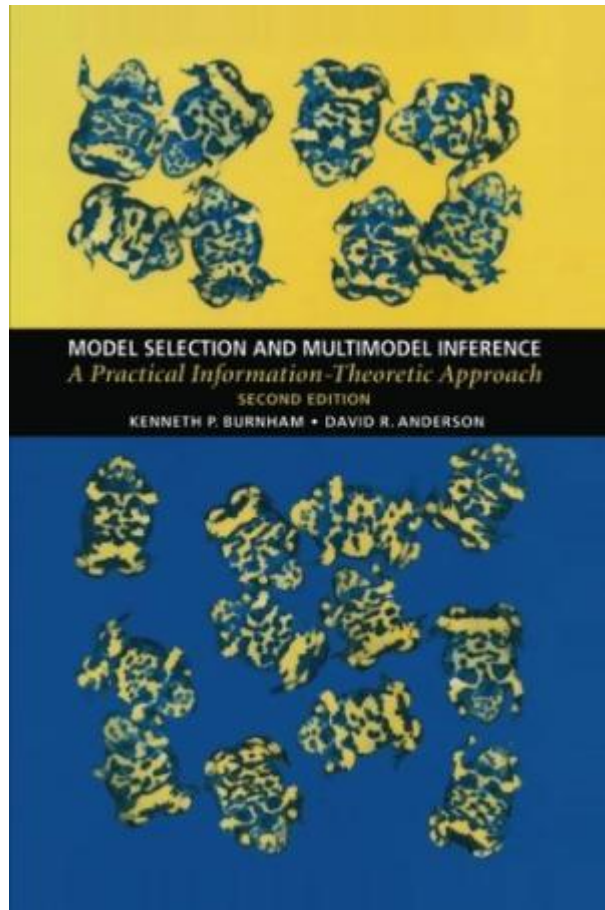


Rule 12: *Plot as much information as needed to interpret the experimental results. Only connect measurements by lines if they indicate trends and the interpolation is valid.*

This is how I should have presented the Dora results.



Part II: Model



Burnham, Anderson: *“A model is a simplification or approximation of reality and hence will not reflect all of reality. ... Box noted that “all models are wrong, but some are useful.” While a model can never be “truth,” a model might be ranked from very useful, to useful, to somewhat useful to, finally, essentially useless.”*

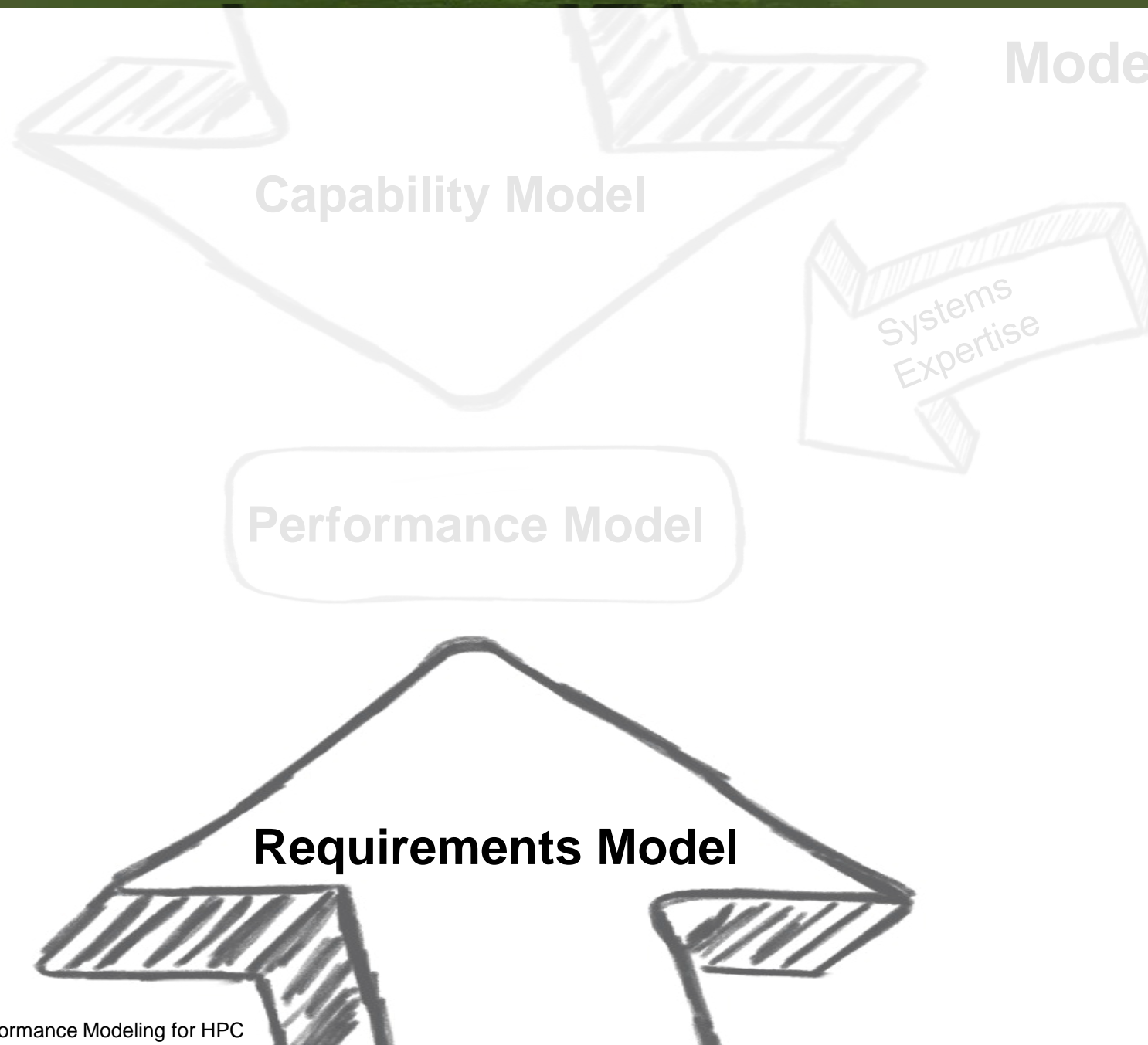
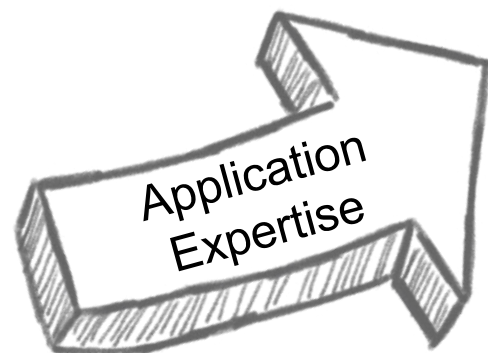
This is generally true for all kinds of modeling.
We focus on **performance modeling** in the following!

Cited by 33599

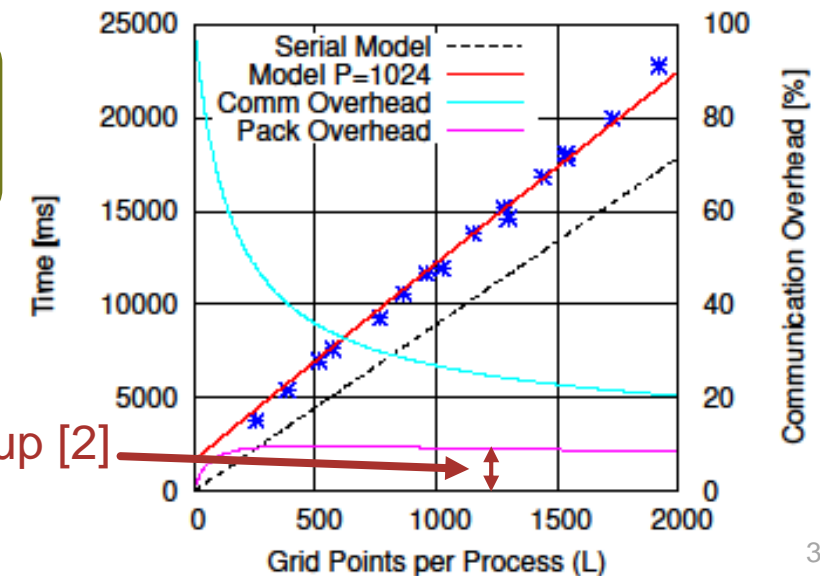
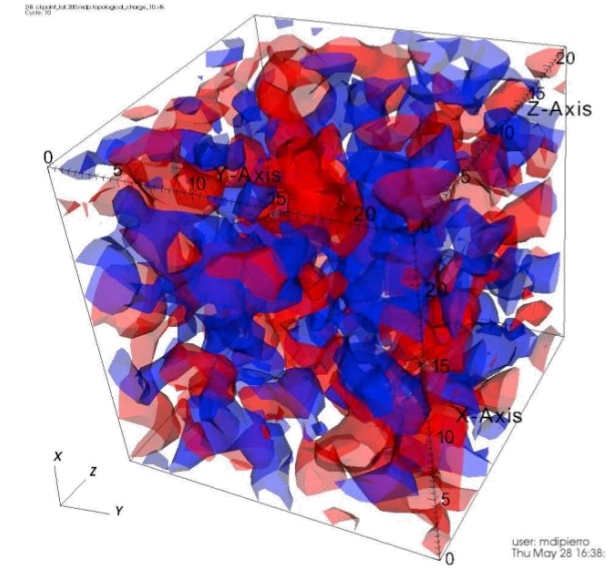
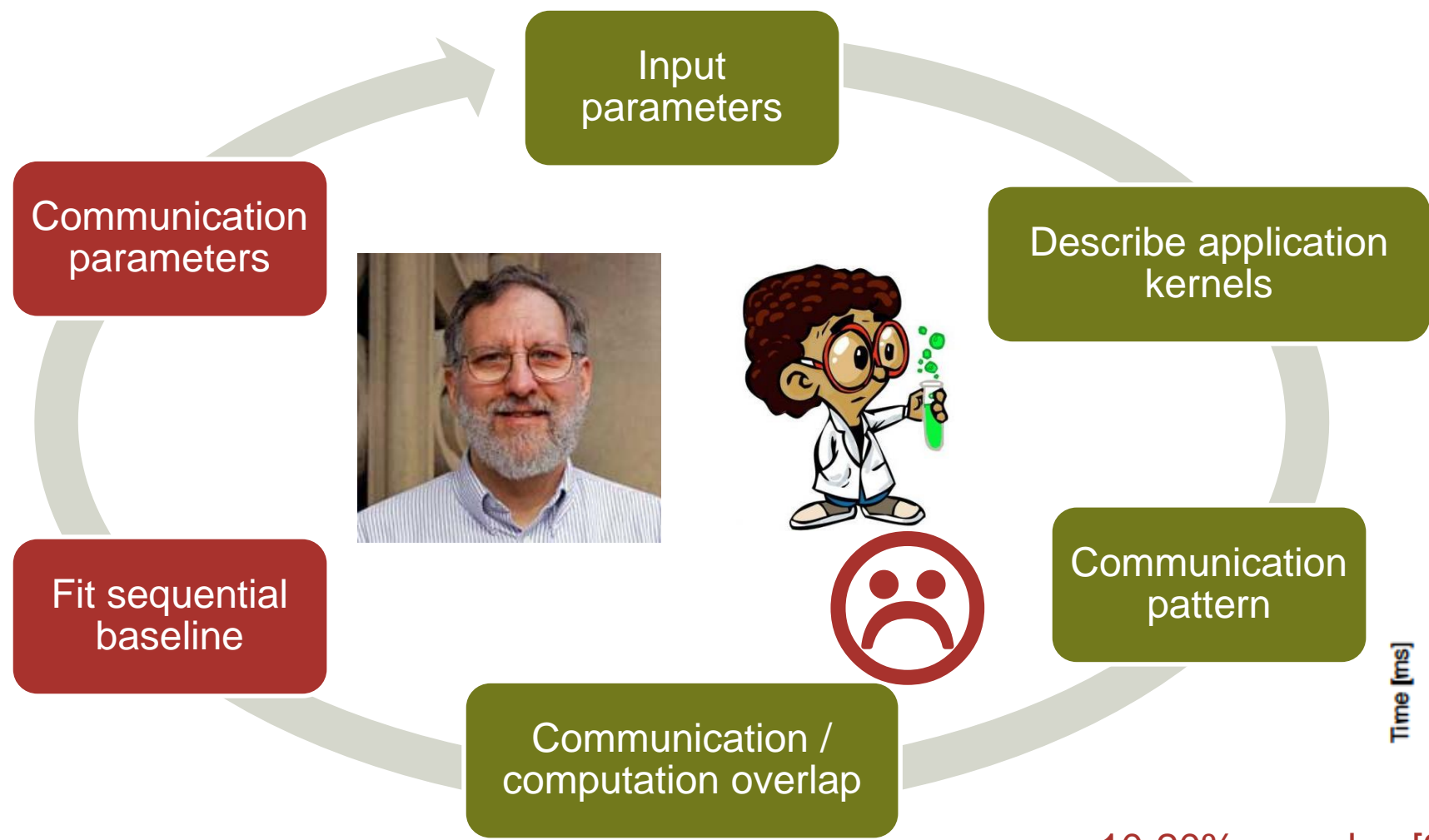


Performance

Modeling



Requirements modeling I: Six-step performance modeling



[1] TH, W. Gropp, M. Snir and W. Kramer: Performance Modeling for Systematic Performance Tuning, SC11
 [2] TH and S. Gottlieb: Parallel Zero-Copy Algorithms for Fast Fourier Transform and Conjugate Gradient using MPI Datatypes, EuroMPI'10

Requirements modeling II: Automated best-fit modeling

- Manual kernel selection and hypothesis generation is time consuming (boring and tricky)
- Idea: Automatically select best (scalability) model from predefined search space

Number of processes

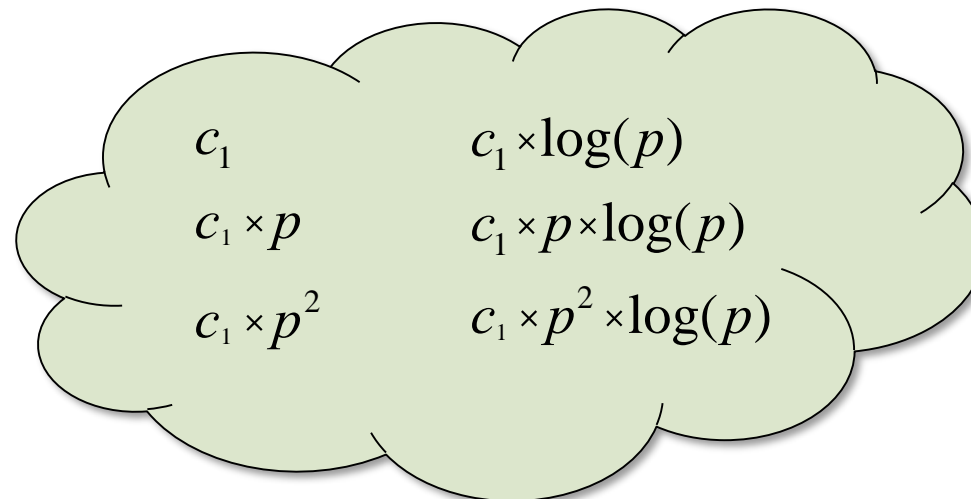
$$f(p) = \sum_{k=1}^n c_k \cdot p^{i_k} \cdot \log_2^{j_k}(p)$$

number of terms

(model) constant

$$\begin{aligned}
 n &\hat{=} \mathbb{N} \\
 i_k &\hat{=} I \\
 j_k &\hat{=} J \\
 I, J &\hat{=} \mathbb{Q}
 \end{aligned}$$

$$\begin{aligned}
 n &= 1 \\
 I &= \{0, 1, 2\} \\
 J &= \{0, 1\}
 \end{aligned}$$



Requirements modeling II: Automated best-fit modeling

- Manual kernel selection and hypothesis generation is time consuming (and boring)
- Idea: Automatically select best model from predefined space

$$f(p) = \prod_{k=1}^n c_k \times p^{i_k} \times \log_2^{j_k}(p)$$

$$n = 2$$

$$I = \{0, 1, 2\}$$

$$J = \{0, 1\}$$

$$c_1 + c_2 \times p$$

$$c_1 + c_2 \times p^2$$

$$c_1 + c_2 \times \log(p)$$

$$c_1 + c_2 \times p \times \log(p)$$

$$c_1 + c_2 \times p^2 \times \log(p)$$

$$c_1 \cdot \log(p) + c_2 \cdot p$$

$$c_1 \cdot \log(p) + c_2 \cdot p \cdot \log(p)$$

$$c_1 \cdot \log(p) + c_2 \cdot p^2$$

$$c_1 \cdot \log(p) + c_2 \cdot p^2 \cdot \log(p)$$

$$c_1 \cdot p + c_2 \cdot p \cdot \log(p)$$

$$c_1 \cdot p + c_2 \cdot p^2$$

$$c_1 \cdot p + c_2 \cdot p^2 \cdot \log(p)$$

$$c_1 \cdot p \cdot \log(p) + c_2 \cdot p^2$$

$$c_1 \cdot p \cdot \log(p) + c_2 \cdot p^2 \cdot \log(p)$$

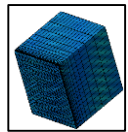
$$c_1 \cdot p^2 + c_2 \cdot p^2 \cdot \log(p)$$

$$\begin{aligned} n &\hat{=} \mathbb{N} \\ i_k &\hat{=} I \\ j_k &\hat{=} J \\ I, J &\hat{=} \mathbb{Q} \end{aligned}$$

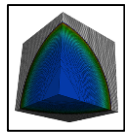
Tool support: Extra-P for automated best-fit modeling [1]



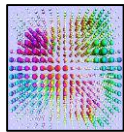
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DARMSTADT



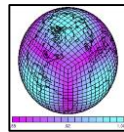
Sweep3d



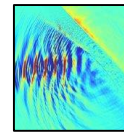
Lulesh



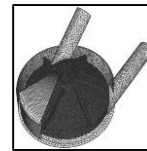
Milc



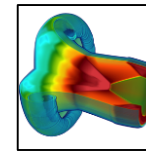
HOMME



JUSPIC



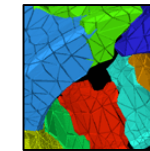
XNS



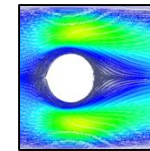
BLAST



NEST



UG4



MP2C

[1] Download Extra-P at: <http://www.scalasca.org/software/extra-p/download.html>

[2] A. Calotoiu, D. Beckingsale, C. W. Earl TH, I. Karlin, M. Schulz, F. Wolf: Fast Multi-Parameter Performance Modeling, IEEE Cluster 2016



Requirements modeling III: Source-code analysis [1]

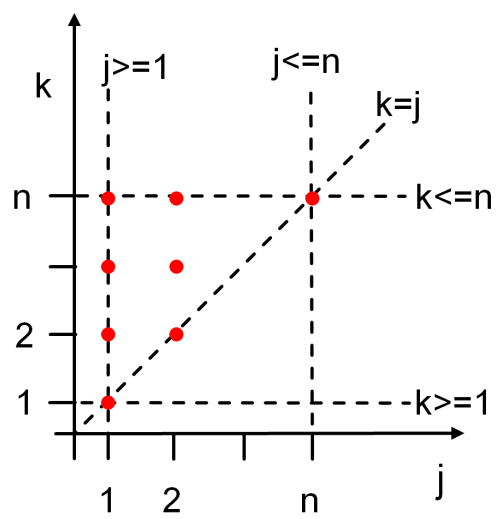
- **Extra-P** selects model based on best fit to the data
 - What if the data is not sufficient or too noisy?
- **Back to first principles**
 - The source code describes all possible executions
 - Describing all possibilities is too expensive, focus on counting loop iterations symbolically

```
for (j = 1; j <= n; j = j*2)
  for (k = j; k <= n; k = k++)
    OperationInBody(j,k);
```

Parallel program

```
do i = 1, procCols
  call mpi_irecv( buff, 2, dp_type, reduce_exch_proc(i),
    i, mpi_comm_world, request, ierr )
  call mpi_send( buff2, 2, dp_type, reduce_exch_proc(i),
    i, mpi_comm_world, ierr )
  call mpi_wait( request, status, ierr )
enddo

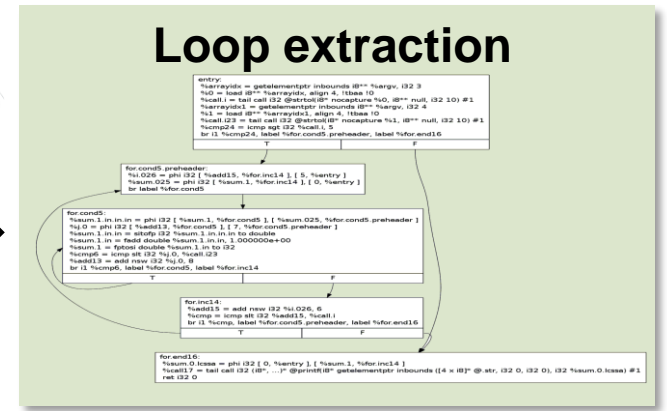
do i = id * n/p, ( id + 1 ) * n/p
  do j = 1, nSize
    call compute
```



$$N = (n + 1) \log_2 n - n + 2$$

Requirements Models

$$W = N \Big|_{p=1}$$

$$D = N \Big|_{p \rightarrow \infty}$$


Number of iterations

$$N = \sum_{i_1=0}^{n_1(x_0,1)} \sum_{i_2=0}^{n_2(x_0,2)} \dots \sum_{i_{r-1}=0}^{n_{r-1}(x_0,r-1)} n_r(x_0,r).$$

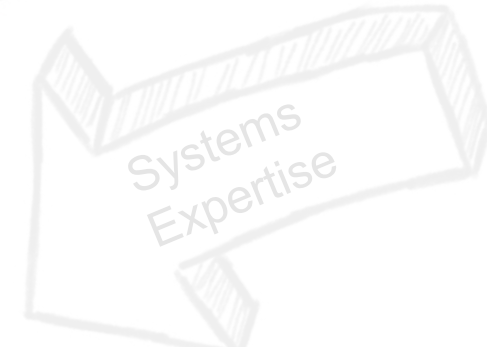

[1]: TH, G. Kwasniewski: Automatic Complexity Analysis of Explicitly Parallel Programs, ACM SPAA'14

Performance

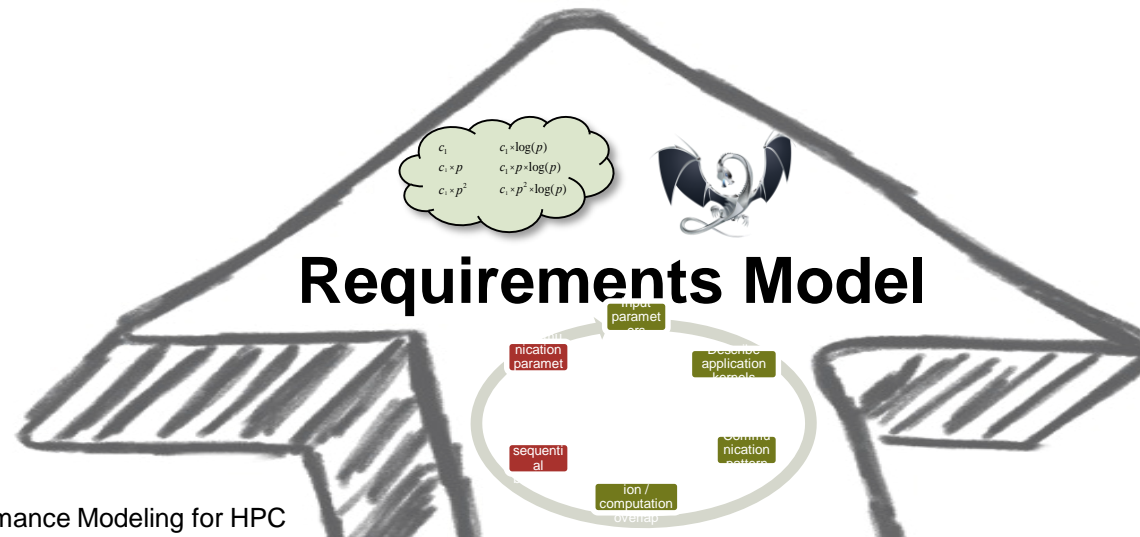
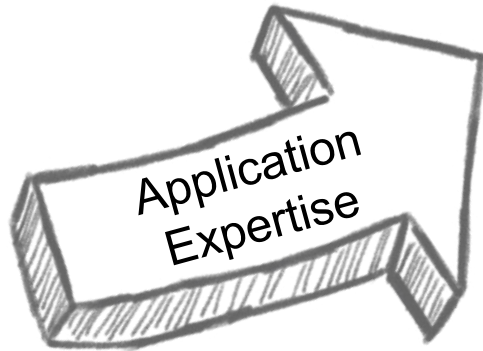
Modeling



Capability Model



Performance Model

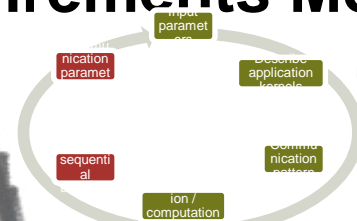


c_1
 $c_1 \cdot p$
 $c_1 \cdot p^2$

$c_1 \cdot \log(p)$
 $c_1 \cdot p \cdot \log(p)$
 $c_1 \cdot p^2 \cdot \log(p)$

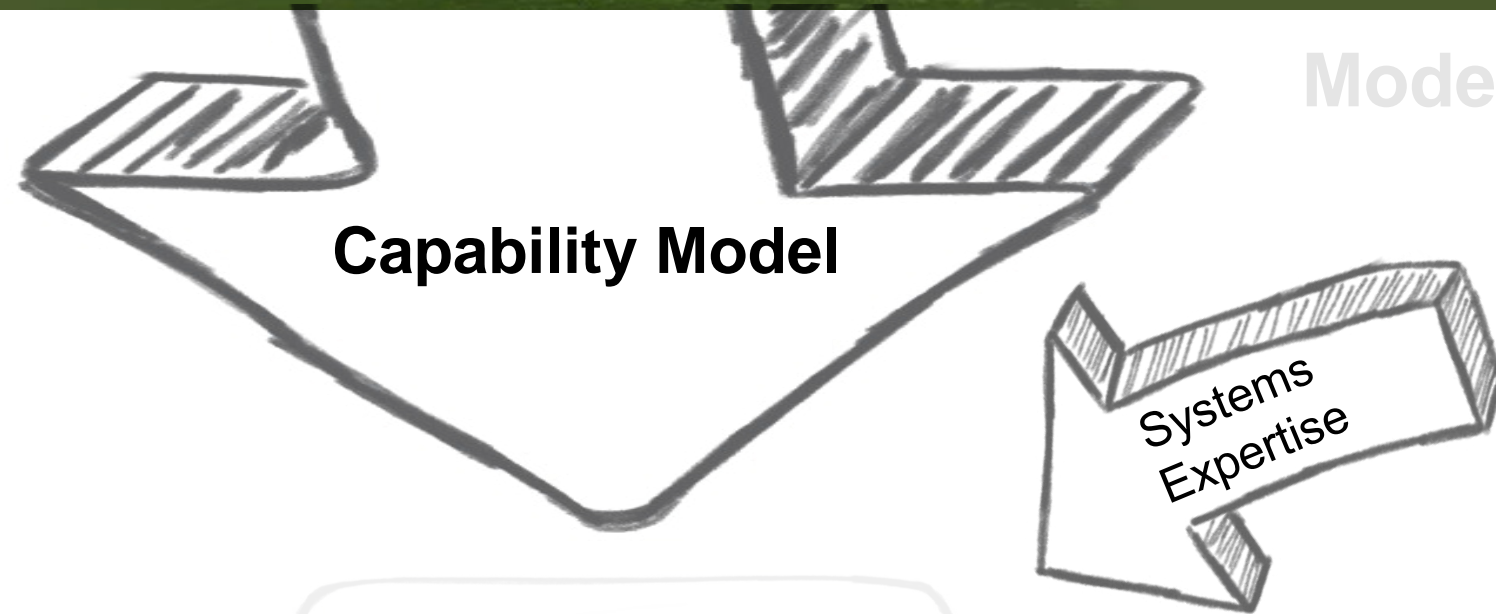


Requirements Model

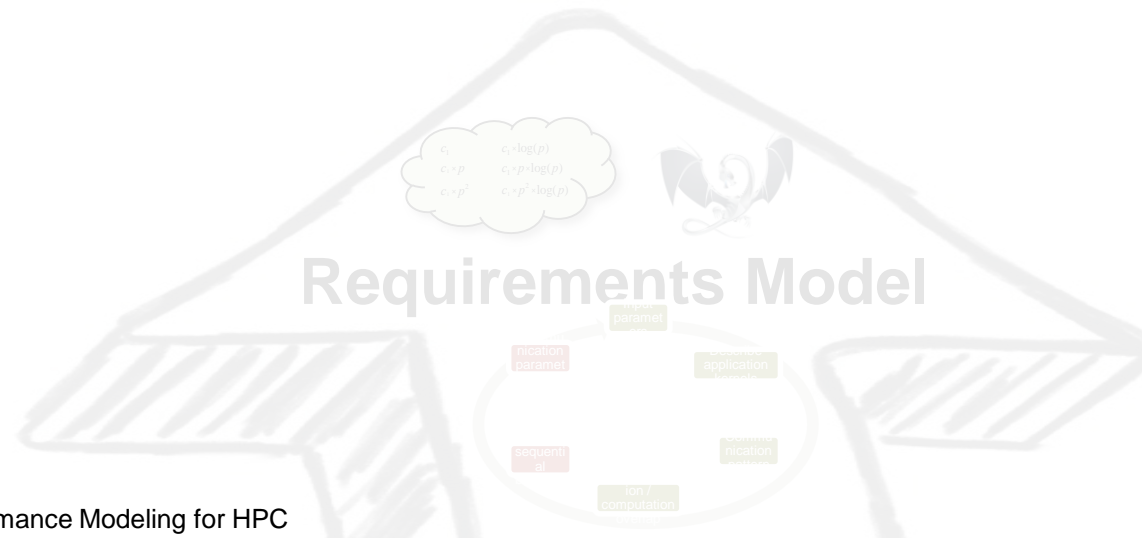


Performance

Modeling



Performance Model



Requirements Model

Capability models for network communication

The LogP model family and the LogGOPS model [1]

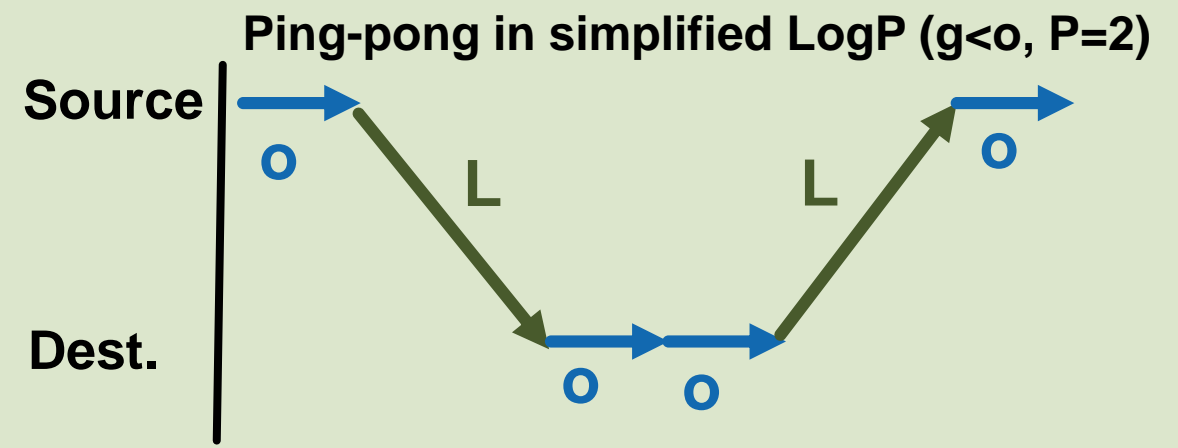
A new parallel machine model reflects the critical technology trends underlying parallel computers

A PRACTICAL MODEL of PARALLEL COMPUTATION

OUR GOAL IS TO DEVELOP A MODEL OF PARALLEL COMPUTATION THAT WILL serve as a basis for the design and analysis of fast, portable parallel algorithms, such as algorithms that can be implemented effectively on a wide variety of current and future parallel machines. If we look at the body of parallel algorithms developed under current parallel models, many are impractical because they exploit artificial factors not present in any real machine.

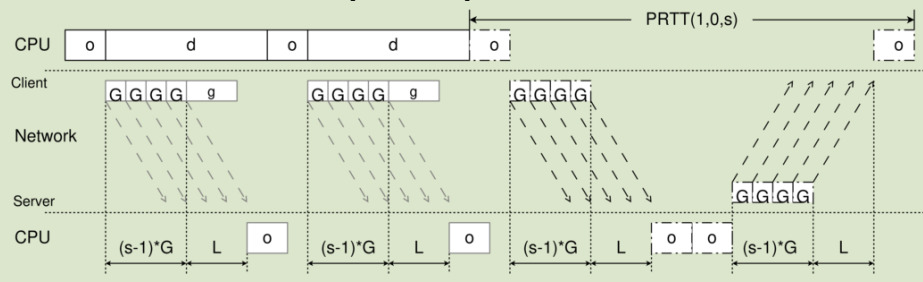
PRAM consists of a collection of processors which compute synchronously in parallel and communicate with a global random access memory.

David E. Culler, Richard M. Karp, David Patterson, Abhijit Sahay, Eunice E. Santos, Klaus Erik Schauer, Ramesh Subramonian, and Thorsten von Eicken



Finding LogGOPS parameters

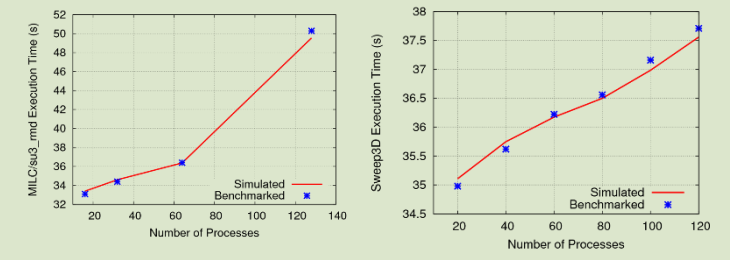
Netgauge [2], model from first principles, fit to data using special kernels



Large scale LogGOPS Simulation

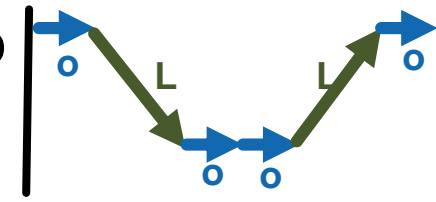
LogGOPSim [1], simulates LogGOPS with 10 million MPI ranks

<5% error

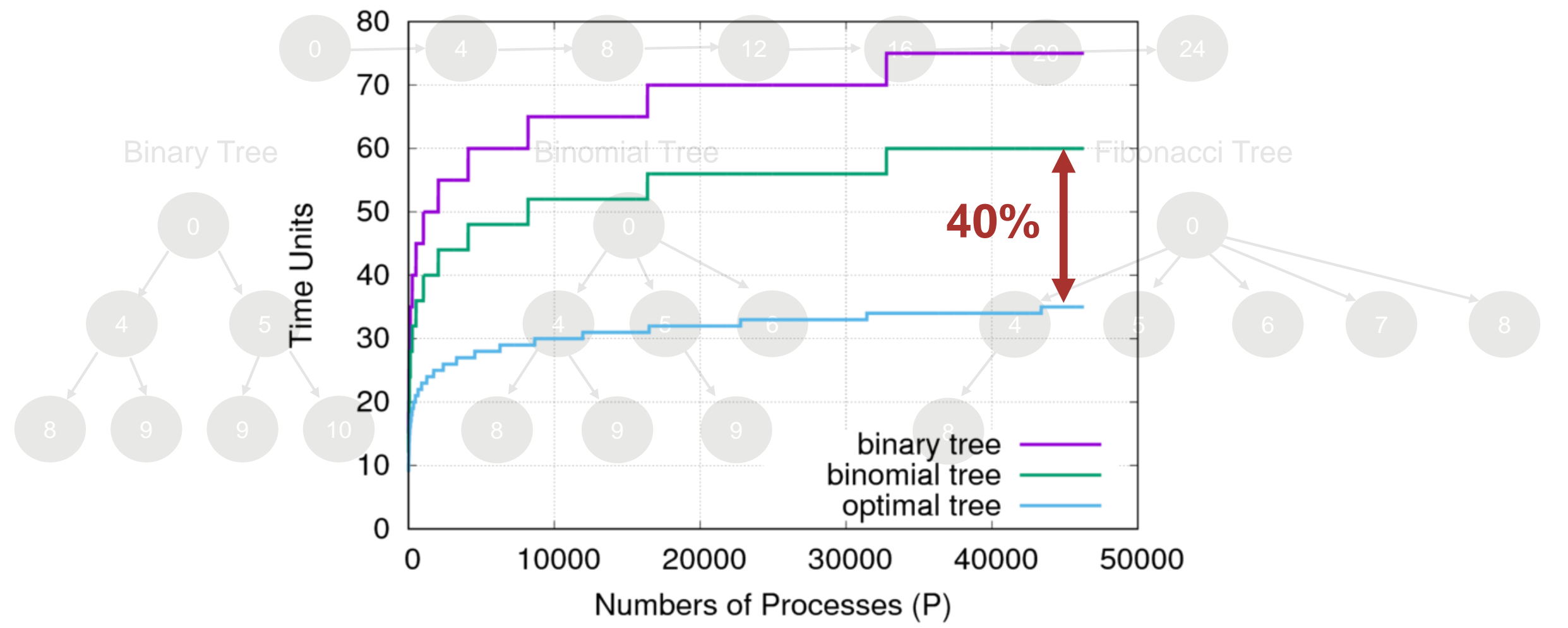


[1]: TH, T. Schneider and A. Lumsdaine: LogGOPSIm - Simulating Large-Scale Applications in the LogGOPS Model, LSAP 2010, <https://spcl.inf.ethz.ch/Research/Performance/LogGOPSIm/>
 [2]: TH, T. Mehlan, A. Lumsdaine and W. Rehm: Netgauge: A Network Performance Measurement Framework, HPC 2007, <https://spcl.inf.ethz.ch/Research/Performance/Netgauge/>

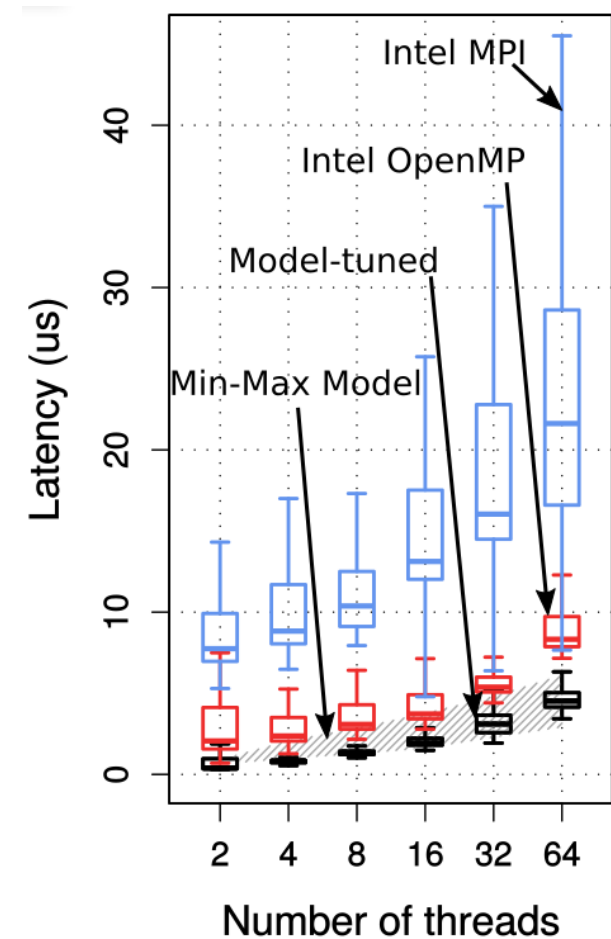
2) Design optimal algorithms – small broadcast in LogP



$L=2, o=1, P=7$

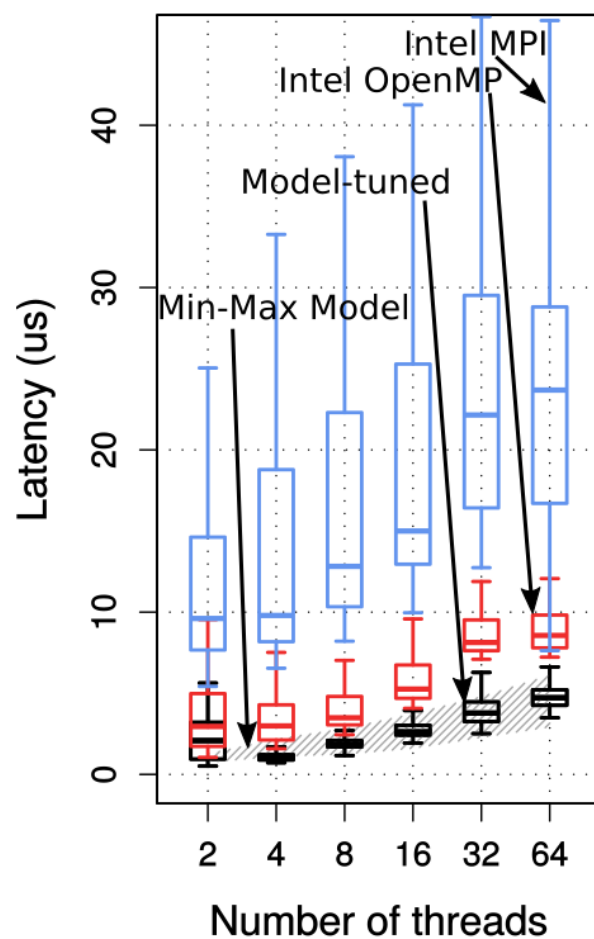


Model-tuned Barrier and Reduce vs. Intel's OpenMP and MPI

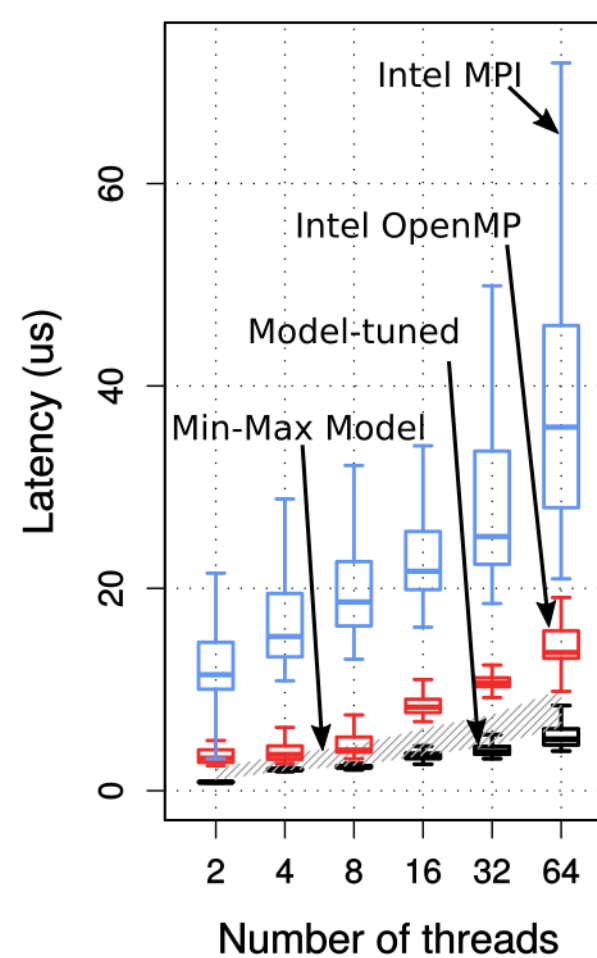


(a) Filling Tiles.

Barrier (7x faster than OpenMP)

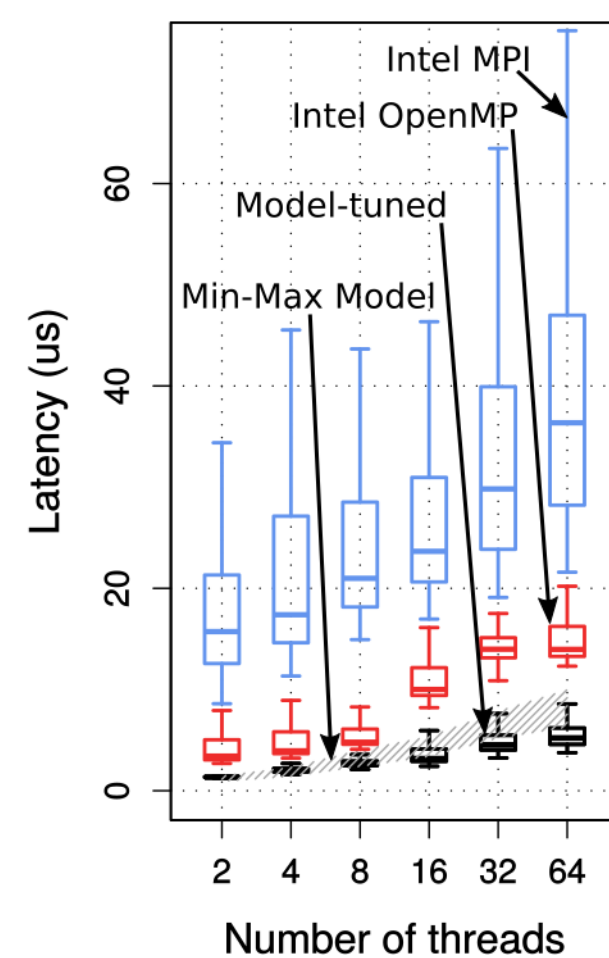


(b) Scatter.



(a) Filling Tiles.

Reduce (5x faster than OpenMP)



(b) Scatter.

Performance

Modeling



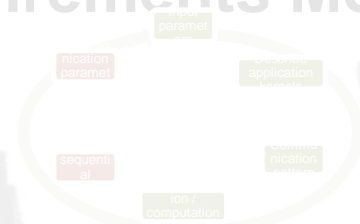
Capability Model



Performance Model

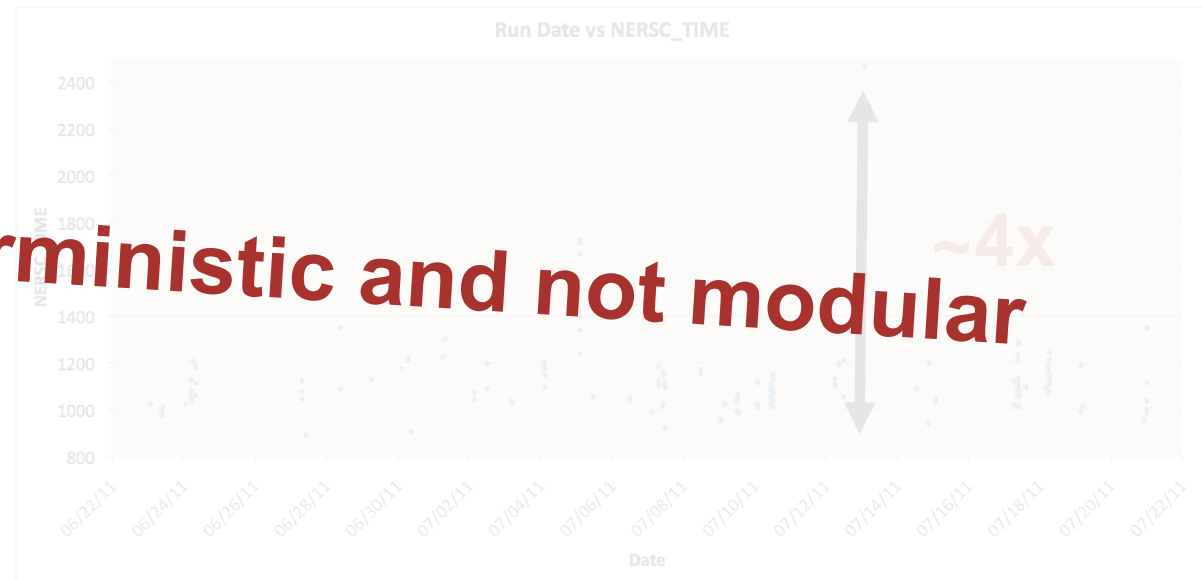


Requirements Model

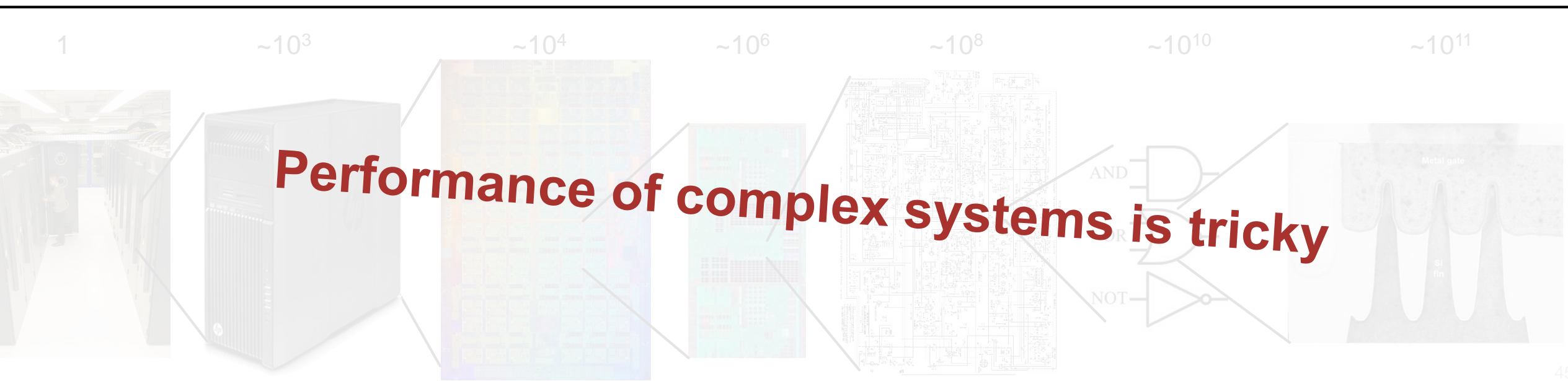


High Performance Computing

```
dgemm("N", "N", 50, 50, 50, 1.0, A, 50, B, 50, 1.0, C, 50);
```



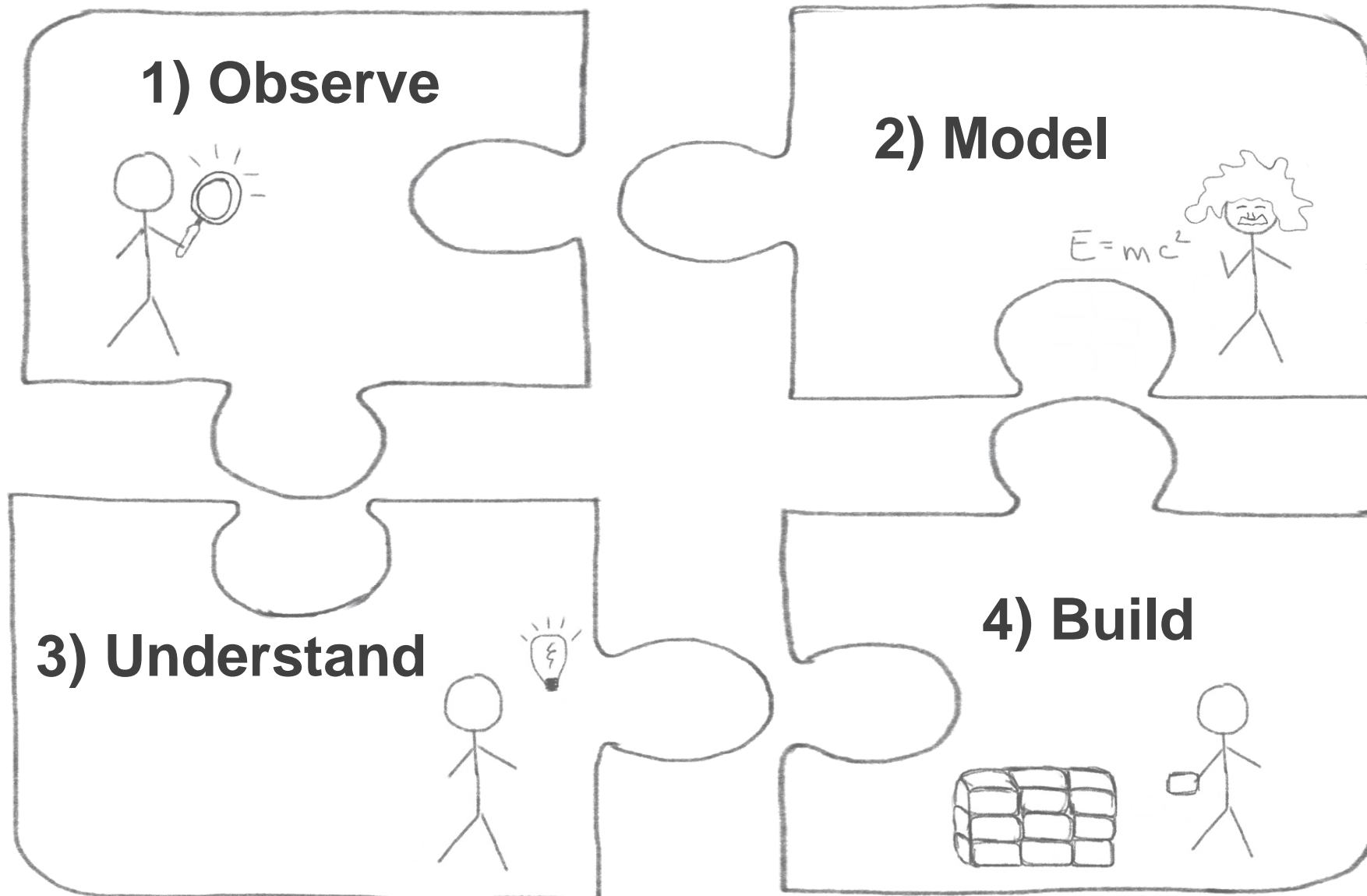
Performance is nondeterministic and not modular



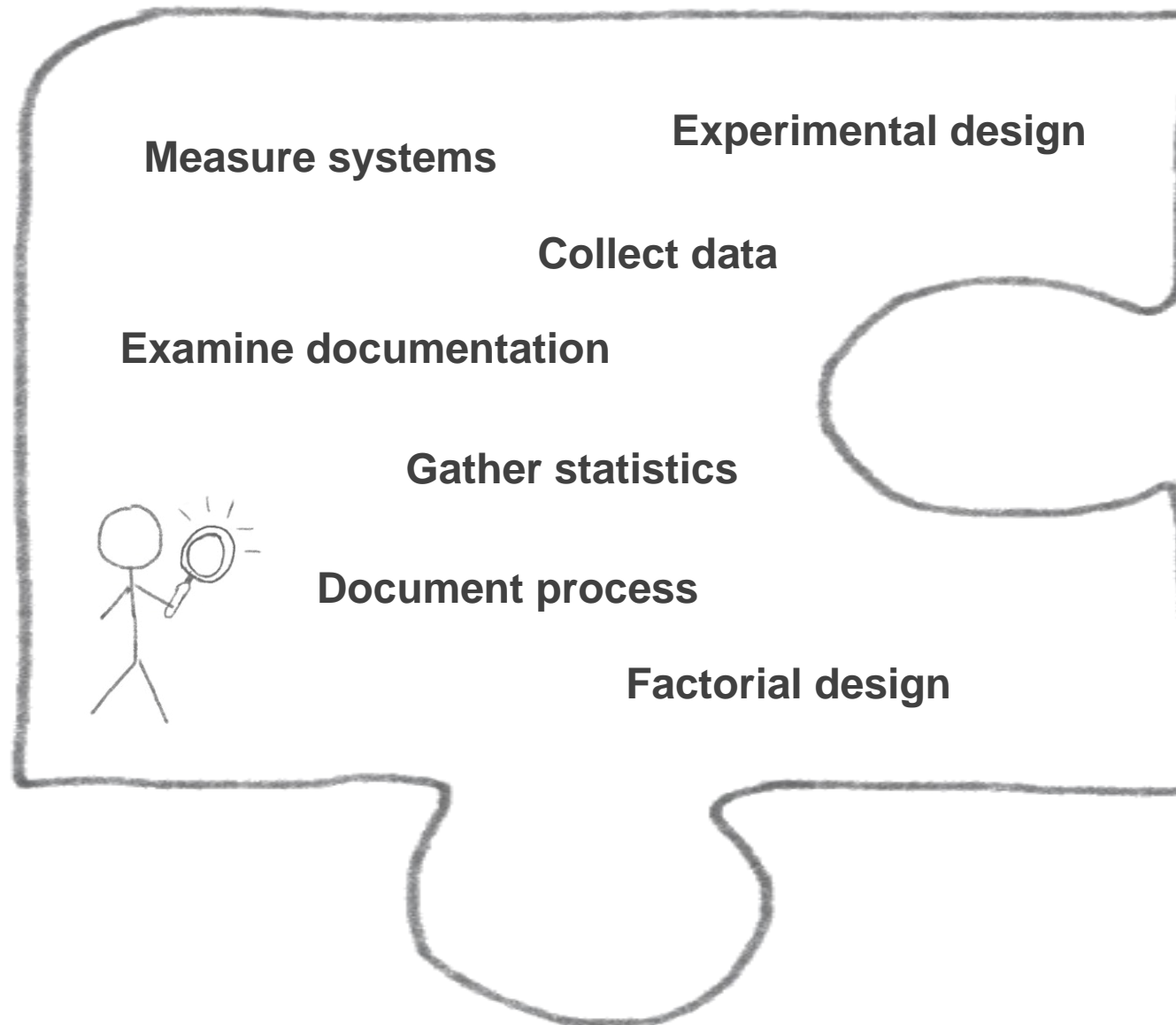
HPC is used to solve complex problems!

**Treat performance-centric programming
and system design like physical systems**

Scientific **Performance** Engineering



Part I: Observe



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!**



