Performance evaluation

Disclaimer

## Performance Evaluation in Database Research: Principles and Experiences

#### Stefan Manegold

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• This is more a collection of anecdotes or fairy tales — not

always to be taken literally, only, but all provide some general

Planning & conducting experiments

• There is no single way how to do it right. • There are many ways how to do it wrong.

• This is not a "mandatory" script.

rules or guidelines what (not) to do.

Planning & conducting experiments Presentation

Repeatability

Summary

Data sets & workloads

### Planning & conducting experiments

- From micro-benchmarks to real-life applications
- Choosing the hardware
- Choosing the software
- What and how to measure
- How to run
- Comparison with others
- CSI
- Presentation
- Repeatability
- Summary

## What do you plan to do / analyze / test / prove / show?

- Which data / data sets should be used?
- Which workload / queries should be run?
- Which hardware & software should be used?
- Metrics:
  - What to measure?
  - How to measure?
- How to compare?
- CSI: How to find out what is going on?

- Micro-benchmarks
- Standard benchmarks
- Real-life applications
- No general simple rules, which to use when
- But some guidelines for the choice...

Micro-benchmarks

## Micro-benchmarks

## Micro-benchmarks

## Pros

- Focused on problem at hand
- Controllable workload and data characteristics
  - Data sets (synthetic & real)
  - Data size / volume (scalability)
  - Value ranges and distribution
  - Correlation
  - Queries
  - Workload size (scalability)
- Allow broad parameter range(s)
- Useful for detailed, in-depth analysis
- Low setup threshold; easy to run

### Cons

- Neglect larger picture
- Neglect contribution of local costs to global/total costs
- Neglect impact of micro-benchmark on real-life applications
- Neglect embedding in context/system at large
- Generalization of result difficult
- Application of insights in full systems / real-life applications not obvious
- Metrics not standardized
- Comparison?

### Definition

- Isolating one particular piece of a larger system
- E.g., single DB operator (select, join, aggregation, etc.)

## • Specialized, stand-alone piece of software

tation Repeatability Summary Benchmarks

Standard benchmarks Standard benchmarks Standard benchmarks Examples Cons • RDBMS, OODBMS, ORDMBS: Pros TPC-{A,B,C,H,R,DS}, 007, ... Mimic real-life scenarios • Often "outdated" (standardization takes (too?) long) • XML, XPath, XQuery, XUF, SQL/XML: Publicly available Often compromises MBench, XBench, XMach-1, XMark, X007, TPoX, ... • Well defined (in theory ...) • Often very large and complicated to run • Stream Processing: • Scalable data sets and workloads (if well designed ...) Limited dataset variation Linear Road, ... • Metrics well defined (if well designed ...) • Limited workload variation General Computing: • Easily comparable (?) • Systems are often optimized for the benchmark(s), only! SPEC, ... Real-life applications Real-life applications Two types of experiments Analysis: "CSI" • Investigate (all?) details Analyze and understand behavior and characteristics Pros Cons • Find out where the time goes and why! • There are so many of them • There are so many of them • Existing problems and challenges • Proprietary datasets and workloads Publication • "Sell your story" • Describe picture at large • Highlight (some) important / interesting details Compare to others Choosing the hardware Choosing the software Choosing the software Which DBMS to use? Commercial Require license Other choices depend on your problem, knowledge, background, Choice mainly depends on your problem, knowledge, background, taste, etc. • "Free" versions with limited functionality and/or optimization taste, etc. capabilities? Operating system • Limitations on publishing results Programming language What ever is required by / adequate for your problem No access to code Compiler Optimizers Scripting languages A laptop might not be the most suitable / representative database server... Analysis & Tuning Tools System tools Visualization tools Open source Freely available • No limitations on publishing results Access to source code

Metrics: What to measure? Metrics: What to measure? Metrics: What to measure?

• Laptop: 1.5 GHz Pentium M (Dothan), 2 MB L2 cache, 2 GB RAM,

Basic

• Throughput: queries per time

Evaluation time

- wall-clock time ("real")
- user CPU time ("user")
- system CPU time ("system")
- Server-side vs. client-side
- Memory and/or storage usage / requirements

Comparison

- Scale-up
- Speed-up
- Analysis
  - System events & interrupts
  - Hardware events

	ser	ver	client		
	3rd		3rd	4th	run
	user	real	real	real	time (milliseconds)
Q					
1	2830	3533	3534	3575	
16	550	618	707	1468	

• Laptop: 1.5 GHz Pentium M (Dothan), 2 MB L2 cache, 2 GB RAM, 5400 RPM disk

- TPC-H (sf = 1)
- MonetDB/SQL v5.5.0/2.23.0
- measured 3rd (& 4th) of four consecutive runs

	ser	ver	client		
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S. Manegold (CWI) Performance Evaluation: Principles & Experiences 19/125	S. Manegold (CWI) Performance Evaluation: Principles & Experiences 20/125	S. Manegold (CWI) Performance Evaluation: Principles & Experiences 21/12
Planning Presentation Repeatability Summary Benchmarks HW SW Metrics How to run Compare CSI	Planning Presentation Repeatability Summary Benchmarks HW SW Metrics How to run Compare CSI	Planning Presentation Repeatability Summary Benchmarks HW SW Metrics How to run Compare
Metrics: What to measure?	Metrics: What to measure?	Metrics: How to measure?

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Metrics: How to measure?

• measured 3rd (& 4th) of four consecutive runs

	ser	ver	client			
	3rd		3rd 4th			run
	user	real	real	real	result	time (milliseconds)
Q	file	file	file	terminal	size	output went to
1	2830	3533	3534	3575	1.3 KB	
16	550	618	707	1468	1.2 MB	

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Be aware what you measure!

Tools, functions and/or system calls to measure time: Unix

- /usr/bin/time, shell built-in time
  - Command line tool ⇒ works with any executable
  - Reports "real", "user" & "sys" time (milliseconds)
  - Measures entire process incl. start-up
  - Note: output format varies!
- gettimeofday()

Metrics: How to measure?

- System function ⇒ requires source code
- Reports timestamp (microseconds)

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S. Manegold (CWI)	Performance Evaluation:	Principles & Exp	eriences			22/12	5
Planning Presentation		Benchmarks		Metrics			

Tools, functions and/or system calls to measure time: Windows

- TimeGetTime(), GetTickCount()
  - System function ⇒ requires source code
  - Reports timestamp (milliseconds)
  - Resolution can be as coarse as 10 milliseconds
- QueryPerformanceCounter() / QueryPerformanceFrequency()
  - System function ⇒ requires source code
  - Reports timestamp (ticks per seconds)
  - Resolution can be as fine as 1 microsecond
- cf., http://support.microsoft.com/kb/172338

Use timings provided by the tested software (DBMS)

- IBM DB2
  - db2batch

Metrics: How to measure?

- Microsoft SQLserver
  - GUI and system variables
- PostgreSQL

### postgresql.conf

log\_statement\_stats = on

log\_min\_duration\_statement = 0

- log\_duration = on
- MonetDB

  - TRACE select ...

echo 'TRACE select 1;' | mclient +----+ | single\_value | +=======+ 1 tuple (5.977ms) | ticks | stmt 16 | sql.exportValue(1, ". ", "single\_value":str, "tinyint 9 | end s0\_1; 50 | function user.s0\_1(A0=1:bte); 318 | X\_5:void := user.s0\_1(1:bte); 4 tuples (6.164ms)

How to run experiments

"We run all experiments in warm memory."

"We run all experiments in warm memory."





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S. Manegold (CWI) Performance Evaluation:	Principles & Experiences	28/125	S. Manegold (CWI)	Performance Evaluation:	Principles & Experiences
Planning Presentation Repeatability Summary		low to run Compare CSI	Planning Presentation		
"hot" vs. "cold"			"hot" vs. "cold	"	

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How to run experiments

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	cc	old	hot		
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	C	old	ho	ot	
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- Depends on what you want to show / measure / analyze
- No formal definition, but "common sense"

### Cold run

A cold run is a run of the query right after a DBMS is started and no (benchmark-relevant) data is preloaded into the system's main memory, neither by the DBMS, nor in filesystem caches. Such a clean state can be achieved via a system reboot or by running an application that accesses sufficient (benchmark-irrelevant) data to flush filesystem caches, main memory, and CPU caches.

A hot run is a run of a query such that as much (query-relevant) data is available as close to the CPU as possible when the measured run starts. This can (e.g.) be achieved by running the query (at least) once before the actual measured run starts.

• Be aware and document what you do / choose

S. Manego	old (CW	/I)	Performanc	e Evaluation:	Principles & Exp			 	
P	lanning				Benchmarks		How to run		
"hot"	VS.	"cold"	' &	user v	vs. real	time			

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	cold		hot		
Q	user	real	user	real	time (milliseconds)
1	2930	13243	2830	3534	

## "hot" vs. "cold" & user vs. real time Of apples and oranges Of apples and oranges

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Be aware what you measure!

#### Once upon a time at CWI ..

- Two colleagues A & B each implemented one version of an algorithm, A the "old" version and B the improved "new"
- They ran identical experiments on identical machines, each for
- Though both agreed that B's new code should be significantly better, results were consistently worse.

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- Though both agreed that B's new code should be significantly better, results were consistently worse.
- They tested, profiled, analyzed, argued, wondered, fought for several days ...
- ... and eventually found out that A had compiled with optimization enabled, while B had not ...

## Of apples and oranges

- Compiler optimization ⇒ up to factor 2 performance difference
- DBMS configuration and tuning  $\Rightarrow$  factor x performance difference (2 < x < 10?)
  - "Self-\*" still research
  - Default settings often too "conservative"
  - Do you know all systems you use/compare equally well?

#### DeBuG

#### Of apples and oranges: MonetDB

configure --enable-debug --disable-optimize --enable-assert

### $CFLAGS = "-g [-00] \dots$ "

#### OPTimized

configure --disable-debug --enable-optimize --disable-assert CFLAGS = "-03 -fomit-frame-pointer -pipe ..."

#### in case of doubt, check

mserver5 --version

Compilation: gcc -03 -fomit-frame-pointer -pipe ...

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Our problem-specific, hand-tuned, prototype X outperforms an out-of-the-box installation of a full-fledged off-the-shelf system Y;

## DBG/OPT time: execution 1.4 relative 1 4 7 10 13 16 19 22 **TPC-H** queries

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## Of apples and oranges Of apples and oranges Do you know what happens?

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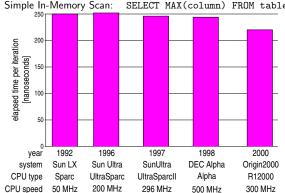
- "Absolutely fair" comparisons virtually impossible
- But:

Be at least aware of the the crucial factors and their impact, and document accurately and completely what you do.

Simple In-Memory Scan: SELECT MAX(column) FROM table

Performance Evaluation: Principles & Experience

Do you know what happens? Do you know what happens? Do you know what happens? Simple In-Memory Scan: SELECT MAX(column) FROM table Simple In-Memory Scan: SELECT MAX(column) FROM table Simple In-Memory Scan: SELECT MAX(column) FROM table



- No disk-I/O involved
- Up to 10x improvement in CPU clock-speed
- ⇒ Yet hardly any performance improvement!??

No disk-I/O involved

Find out what happens!

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- → Yet hardly any performance improvement!??
- Research: Always question what you see!

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## Do you know what happens?

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- → Yet hardly any performance improvement!??
- Research: Always question what you see!
- Standard profiling (e.g., 'gcc -gp' + 'gprof') does not reveal more (in this case)
- Need to dissect CPU & memory access costs
- Use hardware performance counters to analyze cache-hits, -misses & memory accesses
- VTune, oprofile, perfctr, perfmon2, PAPI, PCL, etc.

Simple In	-Memory	/ Scan:	SELECT MAX	(column)	FROM table
					Memory CPU
elapsed time per iteration [nanoseconds]					
Φ 50· year system	1992 Sun LX	1996 Sun Ultra	1997 SunUltra	1998 DEC Alpha	2000 Origin2000
CPU type CPU speed	Sparc 50 MHz	UltraSparo 200 MHz	UltraSparcII	Alpha 500 MHz	R12000 300 MHz

## Find out what happens!

Use info provided by the tested software (DBMS)

- IBM DB2
  - db2expln
- Microsoft SQLserver
  - GUI and system variables
- MySQL, PostgreSQL
  - EXPLAIN select ...
- MonetDB/SQL
  - (PLAN|EXPLAIN|TRACE) select ...

### Use profiling and monitoring tools

'gcc -gp' + 'gprof'

Find out what happens!

- Reports call tree, time per function and time per line
- Requires re-compilation and static linking
- 'valgrind --tool=callgrind' + 'kcachegrind'
  - Reports call tree, times, instructions executed and cache misses
  - Thread-aware
  - Does not require (re-)compilation
  - Simulation-based ⇒ slows down execution up to a factor 100
- Hardware performance counters
  - to analyze cache-hits, -misses & memory accesses
  - VTune, oprofile, perfctr, perfmon2, PAPI, PCL, etc.
- System monitors
  - ps, top, iostat, ...

### Planning & conducting experiments

- Presentation
  - Guidelines
  - Mistakes
- Repeatability
- Summary

Planning Presentation Repeatability Summary Graphical presentation of results Graphical presentation of results Guidelines for preparing good graphic charts

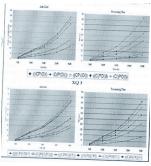
### We all know

A picture is worth a thousand words

#### We all know

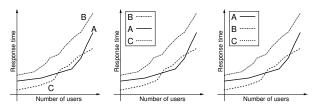
A picture is worth a thousand words

Er, maybe not all pictures...



### Require minimum effort from the reader

- Not the minimum effort from you
- Try to be honest: how would you like to see it?



Guidelines for preparing good graphic charts Guidelines for preparing good graphic charts

## Guidelines for preparing good graphic charts

Maximize information: try to make the graph self-sufficient

- Use keywords in place of symbols to avoid a join in the reader's brain
- Use informative axis labels: prefer "Average I/Os per query" to "Average I/Os" to "I/Os"
- Include units in the labels: prefer "CPU time (ms)" to "CPU time"

## Maximize information: try to make the graph self-sufficient

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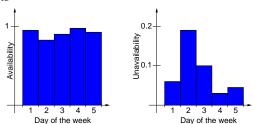
- reader's brain
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- Include units in the labels: prefer "CPU time (ms)" to "CPU time"

Use commonly accepted practice: present what people expect

- Usually axes begin at 0, the factor is plotted on x, the result
- Usually scales are linear, increase from left to right, divisions are equal
- Use exceptions as necessary

Minimize ink: present as much information as possible with as little ink as possible

Prefer the chart that gives the most information out of the same data





Reading material

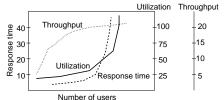
Edward Tufte: "The Visual Display of Quantitative Information"

http://www.edwardtufte.com/tufte/books\_vdqi

Presenting too many alternatives on a single chart Rules of thumb, to override with good reason:

- A line chart should be limited to 6 curves
- A column chart or bar should be limited to 10 bars
- A pie chart should be limited to 8 components
- Each cell in a histogram should have at least five data points

Presenting many result variables on a single chart Commonly done to fit into available page count :-(



Planning Presentation Repeatability Summary Planning Presentation Repeatability Summary Planning Presentation Repeatability Summary Common presentation mistakes Common presentation mistakes Common presentation mistakes Changing the graphical layout of a given curve from one figure to Presenting many result variables on a single chart Using symbols in place of text another Commonly done to fit into available page count :-( 1 job/sec/ //tn time (different machines) Utilization Throughput - 20 2 iobs/sec 10 3 jobs/sec Arrival rate 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 Number of users Human brain is a poor join processor Humans get frustrated by computing joins Huh? What do you mean "my graphs are not legible"? Pictorial games Pictorial games Pictorial games Plot random quantities without confidence intervals MINE is better than YOURS! MINE is better than YOURS! MINE 2610 5200 2610 5200 YOURS MINE YOURS: 2600-YOURS YOURS Overlapping confidence intervals sometimes mean the two A-ha quantities are statistically indifferent Pictorial games: gnuplot & LATEX Pictorial games: gnuplot & LATEX Pictorial games Manipulating cell size in histograms Response time Rule of thumb: each cell should have at least five points Not sufficient to uniquely determine what one should do.

Planning Presentation Repeatability Summary Guidelines Mistakes Planning Presentation Repeatability Summary Guidelines Mistakes Planning Presentation Repeatability Summary Guidelines Mistakes Pictorial games: gnuplot & LATEX Pictorial games: gnuplot & LATEX Specifying hardware environments default: default: better: better: set size ratio 0 1,1 set size ratio 0 0.5,0.5 set size ratio 0 1,1 set size ratio 0 0.5,0.5 "We use a machine with 3.4 GHz." TPC-H queries Rule of thumb for papers: width of plot =  $x \setminus \text{textwidth}$  $\Rightarrow$  set size ratio 0 x\*1.5, yPerformance Evaluation: Principles & Experiences Performance Evaluation: Principles & Experiences Specifying hardware environments Specifying hardware environments Specifying hardware environments "We use a machine with 3.4 GHz." "We use a machine with 3.4 GHz." cat /proc/cpuinfo vendor\_id : GenuineIntel cpu family model : 13 : Intel(R) Pentium(R) M processor 1.50GHz model name stepping : 600.000 cpu MHz cache size : 2048 KB fdiv\_bug hlt\_bug f00f\_bug coma\_bug fpu fpu\_exception cpuid level flags : fpu vme de pse tsc msr mce cx8 mtrr pge mca cmov pat clflush dts acpi mmx fxsr sse sse2 ss tm pbe up bts est tm2 : 1196.56 ⇒ Under-specified! clflush size Performance Evaluation: Principles & Experiences Specifying hardware environments Specifying hardware environments Specifying hardware environments /sbin/lspci -v /sbin/lspci -v 00:00.0 Host bridge: Intel Corporation 82852/82855 GM/GME/PM/GMV Processor to I/O Controller (rev 02) 00:00.0 Host bridge: Intel Corporation 82852/82855 GM/GME/PM/GMV Processor to I/O Controller (rev 02) cat /proc/cpuinfo Flags: bus master, fast devsel, latency 0 Flags: bus master, fast devsel, latency 0 Memory at <unassigned> (32-bit, prefetchable) Memory at <unassigned> (32-bit, prefetchable) Capabilities: <access denied>
Kernel driver in use: agpgart-intel Capabilities: <access denied>
Kernel driver in use: agpgart-intel vendor id : GenuineIntel cpu family model : 13 model name : Intel(R) Pentium(R) M processor  $1.50 \text{GHz} \Leftarrow !$ stepping 01:08.0 Ethernet controller: Intel Corporation 82801DB PRO/100 VE (MOB) Ethernet Controller (rev 83) 01:08.0 Ethernet controller: Intel Corporation 82801DB PRO/100 VE (MOB) Ethernet Controller (rev 83) : 600.000 == throttled down by speed stepping! Subsystem: Benq Corporation Unknown device 5002 Subsystem: Benq Corporation Unknown device 5002 cache size : 2048 KB Flags: bus master, medium devsel, latency 64, IRQ 10 Memory at e0000000 (32-bit, non-prefetchable) [size=4K] Flags: bus master, medium devsel, latency 64, IRQ 10 Memory at e0000000 (32-bit, non-prefetchable) [size=4K] fdiv\_bug hlt bug : no I/O ports at c000 [size=64] Capabilities: <access denied> I/O ports at c000 [size=64] Capabilities: <access denied> f00f\_bug coma\_bug no Kernel driver in use: e100 Kernel driver in use: e100 Kernel modules: e100 yes Kernel modules: e100 cpuid level wp flags : fpu vme de pse tsc msr mce cx8 mtrr pge mca cmov pat clflush /sbin/lspci -v | wc /sbin/lspci -v | wc dts acpi mmx fxsr sse sse2 ss tm pbe up bts est tm2 : 1196.56 151 lines 151 lines 861 words 861 words 6663 characters 6663 characters ⇒ Over-specified!

S. Manegold (CWI) Performance Evaluation: Principles & Experiences

S. Manegold (CWI) Performance Evaluation: Principles & Experiences

Specifying software environments Specifying hardware environments Planning & conducting experiments CPU: Vendor, model, generation, clockspeed, cache size(s) 2 Presentation • 1.5 GHz Pentium M (Dothan), 32 KB L1 cache, 2 MB L2 cache • Main memory: size Repeatability 2 GB RAM • Product names, exact version numbers, and/or sources where • Portable parameterizable experiments • Disk (system): size & speed obtained from Test suite • Documenting your experiment suite • 120 GB Laptop ATA disk @ 5400 RPM • 1 TB striped RAID-0 system (5x 200 GB S-ATA disk @ Summary Network (interconnection): type, speed & topology • 1 GB shared Ethernet Making experiments repeatable Making experiments repeatable Making experiments repeatable Purpose: another human equipped with the appropriate software Purpose: another human equipped with the appropriate software Purpose: another human equipped with the appropriate software and hardware can repeat your experiments. and hardware can repeat your experiments. and hardware can repeat your experiments. Your supervisor / your students Your supervisor / your students Your colleagues Your colleagues • Yourself, 3 months later when you have a new idea • Yourself, 3 months later when you have a new idea • Yourself, 3 years later when writing the thesis or answering • Yourself, 3 years later when writing the thesis or answering requests for that journal version of your conference paper requests for that journal version of your conference paper • Future researchers (you get cited!) Future researchers (you get cited!) Making experiments repeatable means: Making experiments portable and parameterizable Building a test suite and scripts Writing instructions Performance Evaluation: Principles & Experier
ion Repeatability Summary Portability Test s ntation Repeatability Summary Portability Making experiments portable Making experiments portable Making experiments portable Try to use not-so-exotic hardware Try to use not-so-exotic hardware Try to use not-so-exotic hardware Try to use free or commonly available tools (databases, compilers, Try to use free or commonly available tools (databases, compilers, Try to use free or commonly available tools (databases, compilers, plotters...) plotters...) plotters...) Clearly, scientific needs go first (joins on graphic cards; smart card Clearly, scientific needs go first (joins on graphic cards; smart card research; energy consumption study...) research; energy consumption study...) You may omit using Matlab as the driving platform for the experiments

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Matlab as the driving platform for the experiments

20-years old software that only works on an old SUN and is now unavailable

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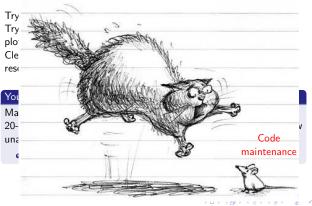
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• If you really love your code, you may even maintain it



Which abstract do you prefer?

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/usr/bin/time to time execution, parse the output with perl, divide by zero

## Abstract (Take 1)

We provide a new algorithm that consistently outperforms the state of the art.

Making experiments parameterizable

## Which abstract do you prefer?

### Abstract (Take 1)

We provide a new algorithm that consistently outperforms the state of the art.

### Abstract (Take 2)

We provide a new algorithm that on a Debian Linux machine with 4 GHz CPU, 60 GB disk, DMA, 2 GB main memory and our own brand of system libraries consistently outperforms the state of the art.

#### Abstract (Take 1)

Which abstract do you prefer?

We provide a new algorithm that consistently outperforms the state of the art.

## Abstract (Take 2)

We provide a new algorithm that on a Debian Linux machine with 4 GHz CPU, 60 GB disk, DMA, 2 GB main memory and our own brand of system libraries consistently outperforms the state of the

There are obvious, undisputed exceptions

This is huge

Planning Presentation Repeatability Summary Planning Presentation Repeatability Summary Making experiments parameterizable Making experiments parameterizable Making experiments parameterizable

## This is huge

Parameters your code may depend on:

- credentials (OS, database, other)
- values of important environment variables (usually one or two)
- various paths and directories (see: environment variables)
- where the input comes from
- switches (pre-process, optimize, prune, materialize, plot ...)
- where the output goes

Purpose: have a very simple mean to obtain a test for the values

$$f_1 = v_1, f_2 = v_2, \dots, f_k = v_k$$

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Many tricks. Very simple ones:

- argc / argv: specific to each class' main
- Configuration files
- Java Properties pattern
- + command-line arguments

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S. Manegold (CWI) Performance Evaluation: Principles & Experiences 100/:	S. Manegold (CWI) Performance Evaluation: Principles & Experiences	101/125 S. Manegold (CWI) Performance Evaluation: Principles & Experiences 102/125
Planning Presentation Repeatability Summary Portability Test suite Documenting	Planning Presentation Repeatability Summary Portability Test suite Documenting	Planning Presentation Repeatability Summary Portability Test suite Documenting
Making experiments parameterizable	Making your code parameterizable	Making your code parameterizable

### Configuration files

Omnipresent in large-scale software

- Crucial if you hope for serious installations: see gnu software install procedure
- Decide on a specific relative directory, fix the syntax
- Report meaningful error if the configuration file is not found

Pro: human-readable even without running code

Con: the values are read when the process is created

The bottom line: you will want to run it in different settings

- With your or the competitor's algorithm or special optimization
- On your desktop or your laptop
- With a local or remote MySQL server
- Make it easy to produce a point
- If it is very difficult to produce a new point, ask questions

The bottom line: you will want to run it in different settings

- With your or the competitor's algorithm or special optimization
- On your desktop or your laptop
- With a local or remote MvSQL server
- Make it easy to produce a point
- If it is very difficult to produce a new point, ask questions

#### You may omit coding like this:

The input data set files should be specified in source file util.GlobalProperty.java.

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Building a test suite	Building a test suite		Automatically generated gr	aphs

You already have:

- Designs
- Easy way to get any measure point

- Suited directory structure (e.g.: source, bin, data, res, graphs)
- Control loops to generate the points needed for each graph, under res/, and possibly to produce graphs under graphs
  - Even Java can be used for the control loops, but...
  - It does pay off to know how to write a loop in shell/perl etc.

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### You may omit coding like this:

Change the value of the 'delta' variable distribution.DistFreeNode.java into 1,5,15,20 and so

#### You have:

- files containing numbers characterizing the parameter values and the results
- basic shell skills

Planning Presentation Repeatability Summary Portability Test suite Planning Presentation Repeatability Summary Automatically generating graphs with Gnuplot

#### Automatically generated graphs Automatically generating graphs with Gnuplot

Data file results-m1-n5.csv:

1	1234
2	2467
3	4623

	1234	
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## You have:

1	1234
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#### Data file results-m1-n5.csv:

1	1234
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@ Gnuplot command file plot-m1-n5.gnu to plot this graph:

```
set data style linespoints
set terminal postscript eps color
set output "results-m1-n5.eps"
set title "Execution time for various scale factors"
set xlabel "Scale factor"
set ylabel "Execution time (ms)"
plot "results-m1-n5.csv"
```

• files containing numbers characterizing the parameter values

## Automatically generating graphs with Gnuplot

Data file results-m1-n5.csv:

and the results basic shell skills

Most frequently used solutions:

Based on Excel or OpenOffice clone

Other solutions: R; Matlab (remember portability)

• Based on Gnuplot

You need: graphs

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```

3 Call gnuplot plot-m1-n5.gnu

## Automatically producing graphs with Excel

Oreate an Excel file results-m1-n5.xls with the column labels:

Α	В	С
1	Scale factor	Execution time
2		
3		

Automatically producing graphs with Excel

Oreate an Excel file results-m1-n5.xls with the column

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Insert in the area B2-C3 a link to the file results-m1-n5.csv

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Automatically producing graphs with Excel

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- Oreate in the .xls file a graph out of the cells A1:B3, chose the layout, colors etc.
- When the .csv file will be created, the graph is automatically filled in.

## You may omit working like this:

Graph generation

In avgs.out, the first 15 lines correspond to xyzT, the next 15 lines correspond to XYZT, and the next 15 lines correspond to XyZT. In each of these sets of 15, the numbers correspond to queries 1.1,1.2,1.3,1.4,2.1,2.2,2.3,2.4,3.1,3.2,3.3,3.4,4.1,4.2,and 4.3.

Planning Presentation Repeatability Summary Portability Test suite Why you should take care to generate your own graphs Graph generation Why you should take care to generate your own graphs File avgs.out contains average times over three runs: File avgs.out contains average times over three runs: a b a b You may omit working like this: 1 13.666 1 13.666 In avgs.out, the first 15 lines correspond to xyzT, the next 15 lines 2 15 2 15 correspond to xYZT, the next 15 lines correspond to Xyzt, the next 3 12.3333 3 12.3333 15 lines correspond to xyZT, the next 15 lines correspond to XyzT, 4 13 4 13 the next 15 lines correspond to XYZT, and the next 15 lines corre-Copy-paste into OpenOffice 2.3.0-6.11-fc8: spond to XyZT. In each of these sets of 15, the numbers correspond to gueries 1.1,1.2,1.3,1.4,2.1,2.2,2.3,2.4,3.1,3.2,3.3,3.4,4.1,4.2,and 4.3. 13666 2 15 3 123333 ... either because you want to do clean work, or because you don't 13 want this to happen: Why you should take care to generate your own graphs Documenting your experiment suite File avgs.out contains average times over three runs: File avgs.out contains average times over three runs: ('.' decimals) 1 13.666 1 13.666 Very easy if experiments are already portable, parameterizable, and 2 15 2 15 if graphs are automatically generated. 3 12.3333 3 12.3333 Specify: 4 13 4 13 What the installation requires; how to install Copy-paste into OpenOffice 2.3.0-6.11-fc8: (expecting ',' decimals) Copy-paste into OpenOffice 2.3.0-6.11-fc8 For each experiment Extra installation if any 13666 13666 Script to run Where to look for the graph 15 15 3 123333 3 123333 13 13 The graph doesn't look good :-( The graph doesn't look good :-( Hard to figure out when you have to produce by hand 20 such graphs and most of them look OK Summary & conclusions • Good and repeatable performance evaluation and Very easy if experiments are already portable, parameterizable, and experimental assessment require no fancy magic but rather if graphs are automatically generated. solid craftsmanship Specify: • Proper planning helps to keep you from "getting lost" and What the installation requires; how to install ensure repeatability For each experiment • Repeatable experiments simplify your own work (and help others to understand it better) • Extra installation if any

## Why you should take care to generate your own graphs

## Documenting your experiment suite

- Script to run
- Where to look for the graph
- How long it takes

- There is no single way how to do it right.
- There are many ways how to do it wrong.
- We provided some simple rules and guidelines what (not) to do.