

Profiling & Optimization

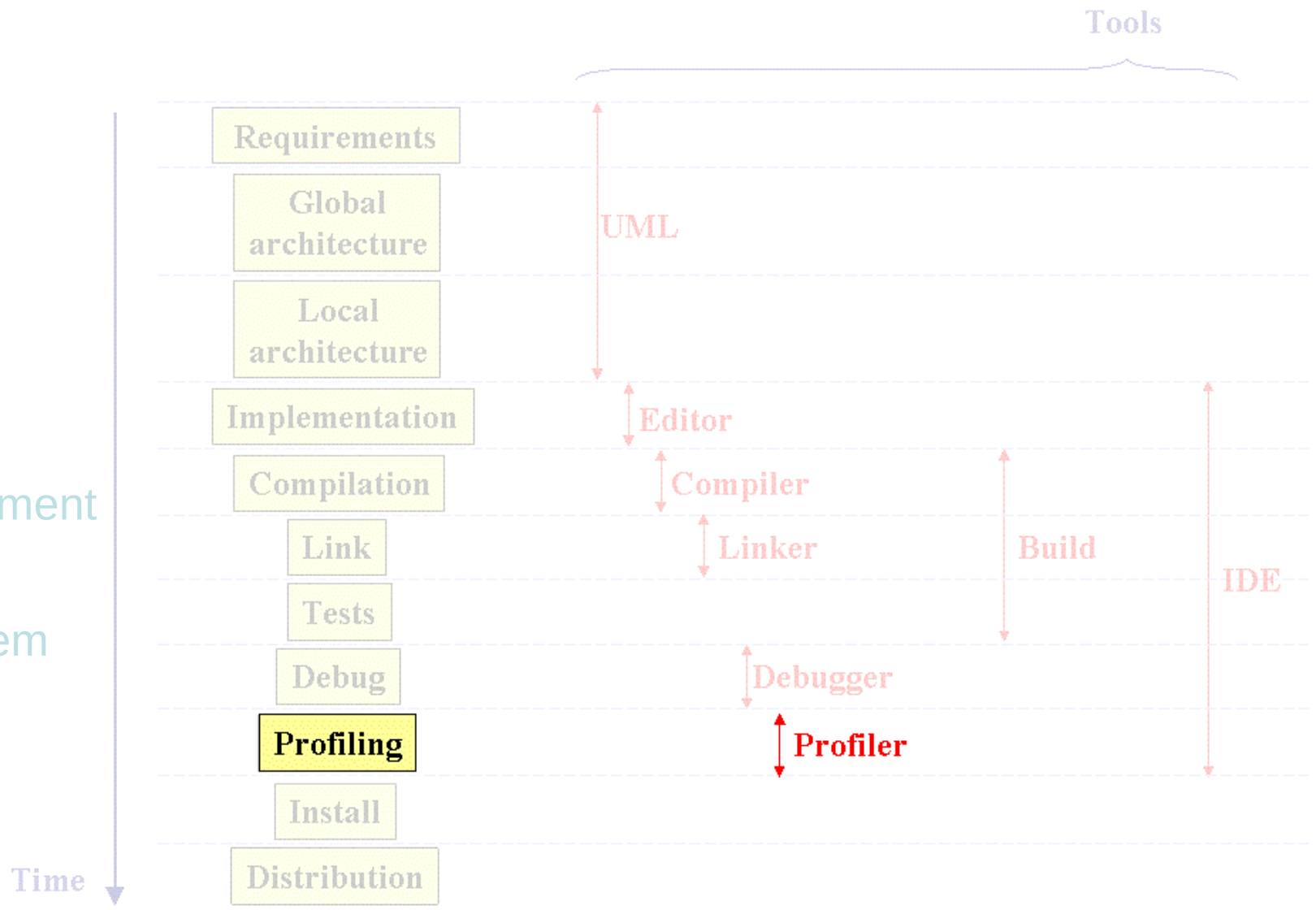
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DE RECHERCHE
EN INFORMATIQUE
ET EN AUTOMATIQUE



Profiling & Optimization

- Introduction
- Analysis/Design
- Build
- Tests
- Debug
- **Profiling**
- Project management
- Documentation
- Versioning system
- IDE
- GForge
- Conclusion



Outline

- Profiling
- Tools
- Optimization



Profiling

No optimization without profiling

Not everyone has a P4 @ 3Ghz

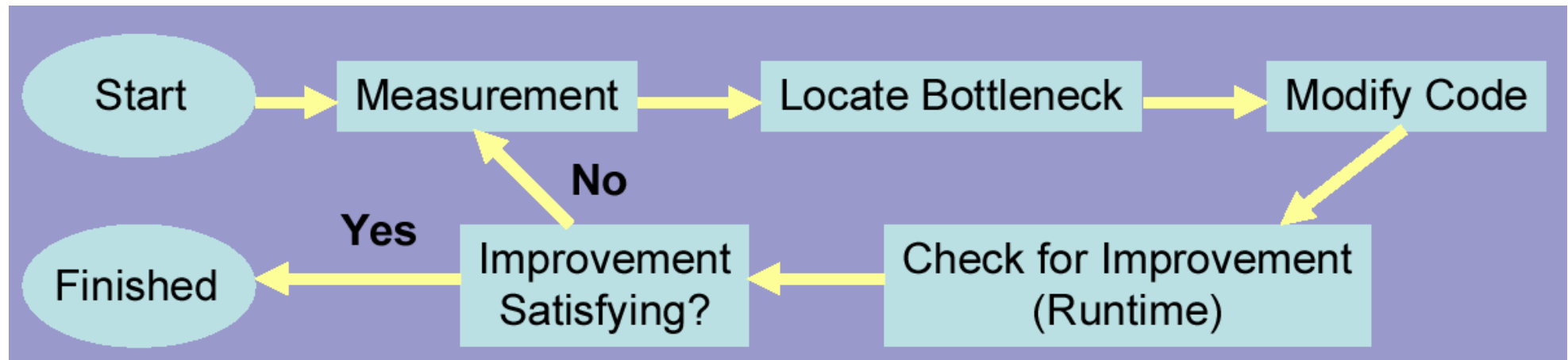
“We should forget about small efficiencies, say about 97% of the time : premature optimization is the root of all evil.” Donald Knuth

Profiling : When ?

- To choose among several algorithms for a given problem
- To check the awaited behavior at runtime
- To find the parts of the code to be optimized

Profiling : How ?

- On fully implemented (and also tested) code
- On a “release” version (optimized by the compiler, ...)
- On representative data
- The optimization cycle :



What do we measure ?

- Understand what's going on :
 - OS: scheduling, memory management, hard drives, network
 - Compiler : optimization
 - CPU architecture, chipset, memory
 - Libraries used
- If an application is limited by its I/O, useless to improve the calculation part.
- Here we'll limit ourselves to CPU & memory performance.

Measurement methods

- Manual
- Source instrumentation
- Statistical measure (sampling)
- Simulation
- Hardware counters

Outline

- Profiling
- Tools
- Optimization

Tools

	Manual	Instr.	Sampling	Simulation	Hardware Counter
▪ system timers	X				
▪ gprof / gcc -pg		X	X		
▪ valgrind(callgrind)/kcachegrind				X	
IBM Rational quantify		X	X		
Oprofile			X		
Intel Vtune			X		X
PAPI (Performance API)					X
JVMPI (Java Virtual Machine Profiler) ▪ runhprof, ▪ Eclipse TPTP (Test&Performance) ▪ OptimizeIT, JProbe, JMP (Memory Profiler)			X		
Shark (MacOS X)			X		X

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Using the tools : system timers

- You have to **know the timer's resolution**
- Windows :
 - `QueryPerformanceCounter() / QueryPerformanceFrequency()`
- Linux/Unix :
 - `gettimeofday()`
 - `clock()`
- Java :
 - `System.currentTimeMillis()`
 - `System.nanoTime()`
- Intel CPU counter : RDTSC (Read Time Stamp Counter)

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Using the tools : gprof

gprof (compiler generated instrumentation) :

- instrumentation : count the function calls
- temporal sampling
- compile with gcc -pg
- create gmon.out file at runtime

Drawbacks

- line information not precise
- needs a complete recompilation
- results not always easy to analyze for large software

Using the tools : gprof

```
dgeld@enesco: ~/src/versions/dgeld/cours/optimisation
enesco$ make timapg
g++ -O3 -g -march=pentium4 -pg -o timapg tima.cpp image.cpp image2.cpp timer.cpp
enesco$ ./timapg
invert  : 304.63          911650545 ticks
invert2 : 107.179 ms     320743665 ticks
invert3 : 80.948 ms      242239312 ticks
invert4 : 63.167 ms      189028665 ticks
invert4b : 63.115 ms     188874075 ticks
invert5 : 63.17 ms       189038527 ticks
invert6 : 14.781 ms      44223030 ticks
enesco$
```

Using the tools : gprof

```
dgeld@enesco:~/src/versions/dgeld/cours/optimisation
enesco$ gprof ./timapg | head
Flat profile:

Each sample counts as 0.01 seconds.
 %   cumulative    self           self      total      name
time  seconds  seconds   calls  ms/call  ms/call  name
45.00      0.27      0.27         1    270.00    270.00  image::invert()
15.00      0.36      0.09         1     90.00     90.00  image::invert2()
11.67      0.43      0.07         1     70.00     70.00  image::invert3()
10.00      0.49      0.06         1     60.00     60.00  image::invert4b()
 8.33      0.54      0.05         1     50.00     50.00  image::invert4()
enesco$ gprof -l ./timapg | head
Flat profile:

Each sample counts as 0.01 seconds.
 %   cumulative    self           self      total      name
time  seconds  seconds   calls  Ts/call  Ts/call  name
43.33      0.26      0.26         1    image::invert() (image.cpp:77 @ 8048e3d)
 9.17      0.32      0.06         1    image::invert2() (image.cpp:88 @ 8048e7c)
 8.33      0.36      0.05         1    image::invert5() (image.cpp:132 @ 8048f35)
 6.67      0.41      0.04         1    image::invert4() (image.cpp:108 @ 8048ee1)
 5.83      0.44      0.04         1    image::invert3() (image.cpp:98 @ 8048eb2)
enesco$
```


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Using the tools : callgrind

callgrind/kcachegrind : <http://kcachegrind.sf.net/cgi-bin/show.cgi>

- cache simulator on top of valgrind
- CPU simulation : estimate CPU cycles for each line of code
- analyze the data more easily with kcachegrind

Drawbacks

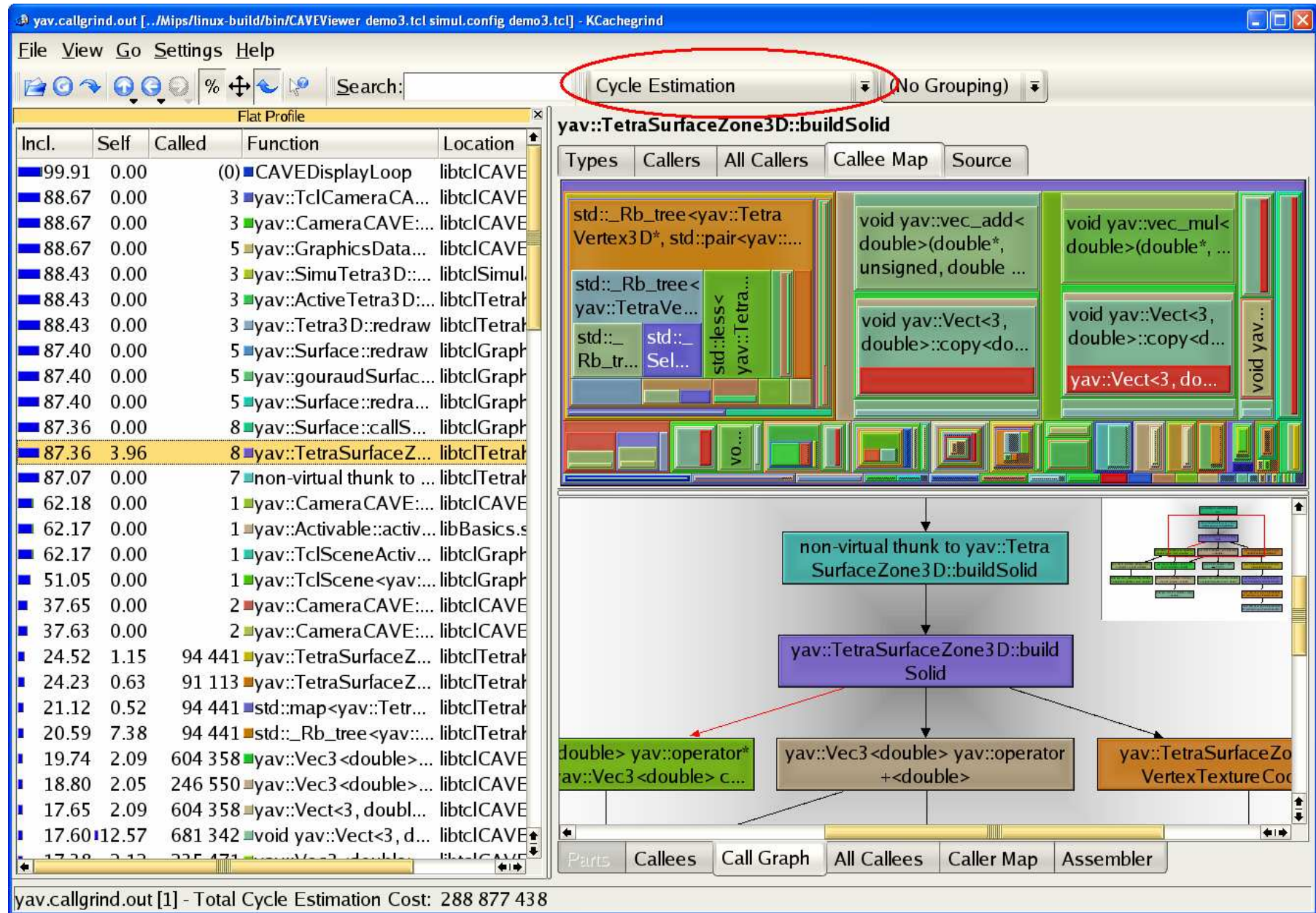
- time estimates can be inaccurate
- measure only the user part of the code
- analyzed software is 20-100 times slower, uses huge amount of memory.

Using the tools : callgrind

callgrind/kcachegrind usage :

- on already compiled software :
`valgrind --tool=callgrind prog`
- generates `callgrind.out.xxx`
- analyzed with `callgrind_annotate` or `kcachegrind`
- to be usable in spite of its slowness :
do not simulate cache usage : `--simulate-cache=no`
start instrumentation only when needed :
`--instr-atstart=no / callgrind_control -i on`

Using the tools : callgrind



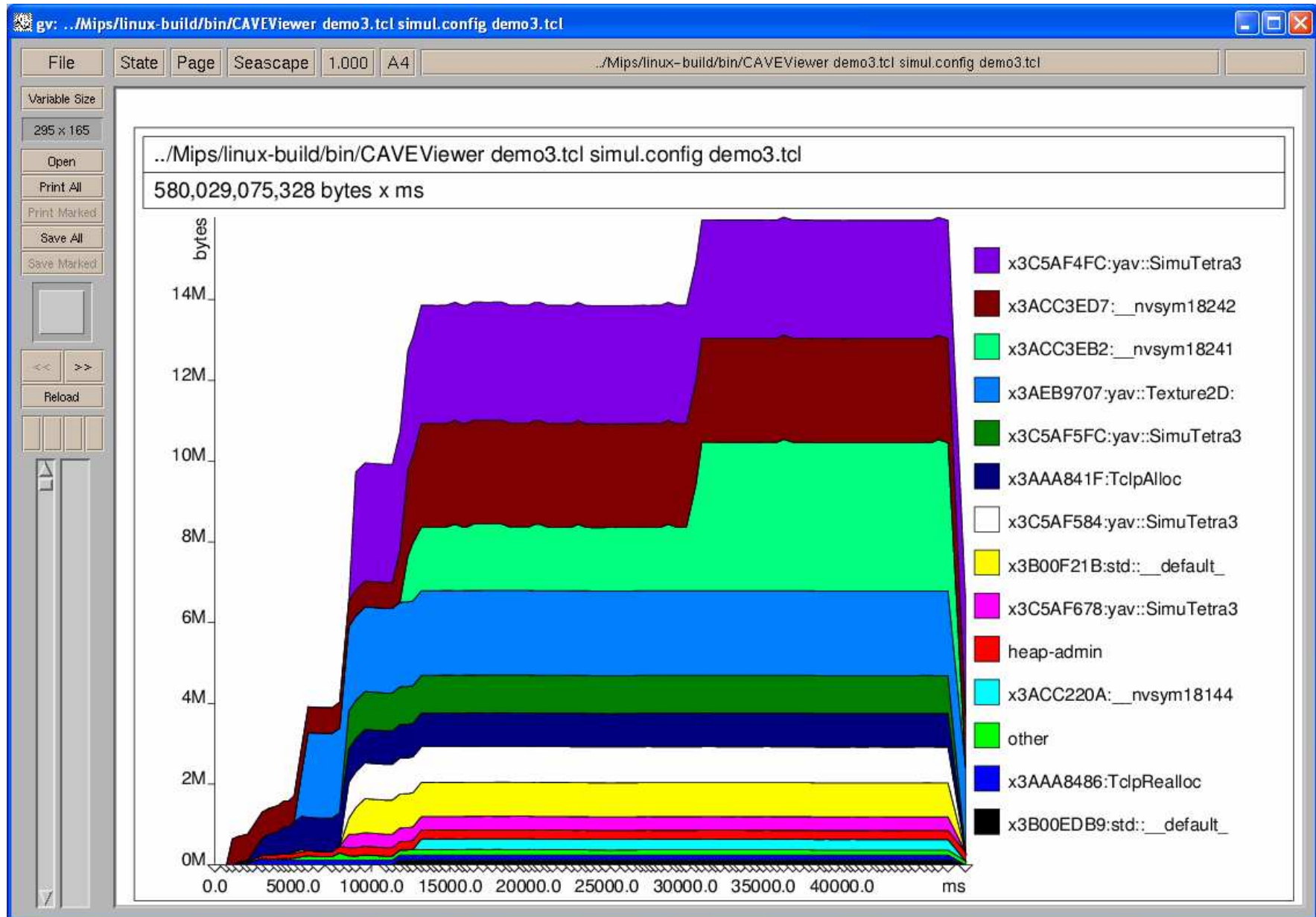
Using the tools : massif

massif (heap profiler) : <http://valgrind.org/info/tools.html#massif>

- another valgrind tool to be used on compiled software :
`valgrind --tool=massif prog`
- generates `massif.xxx.ps` : memory usage vs. time
- `massif.xxx.txt` : which part of code uses what



Using the tools : massif



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Using the tools : runhprof

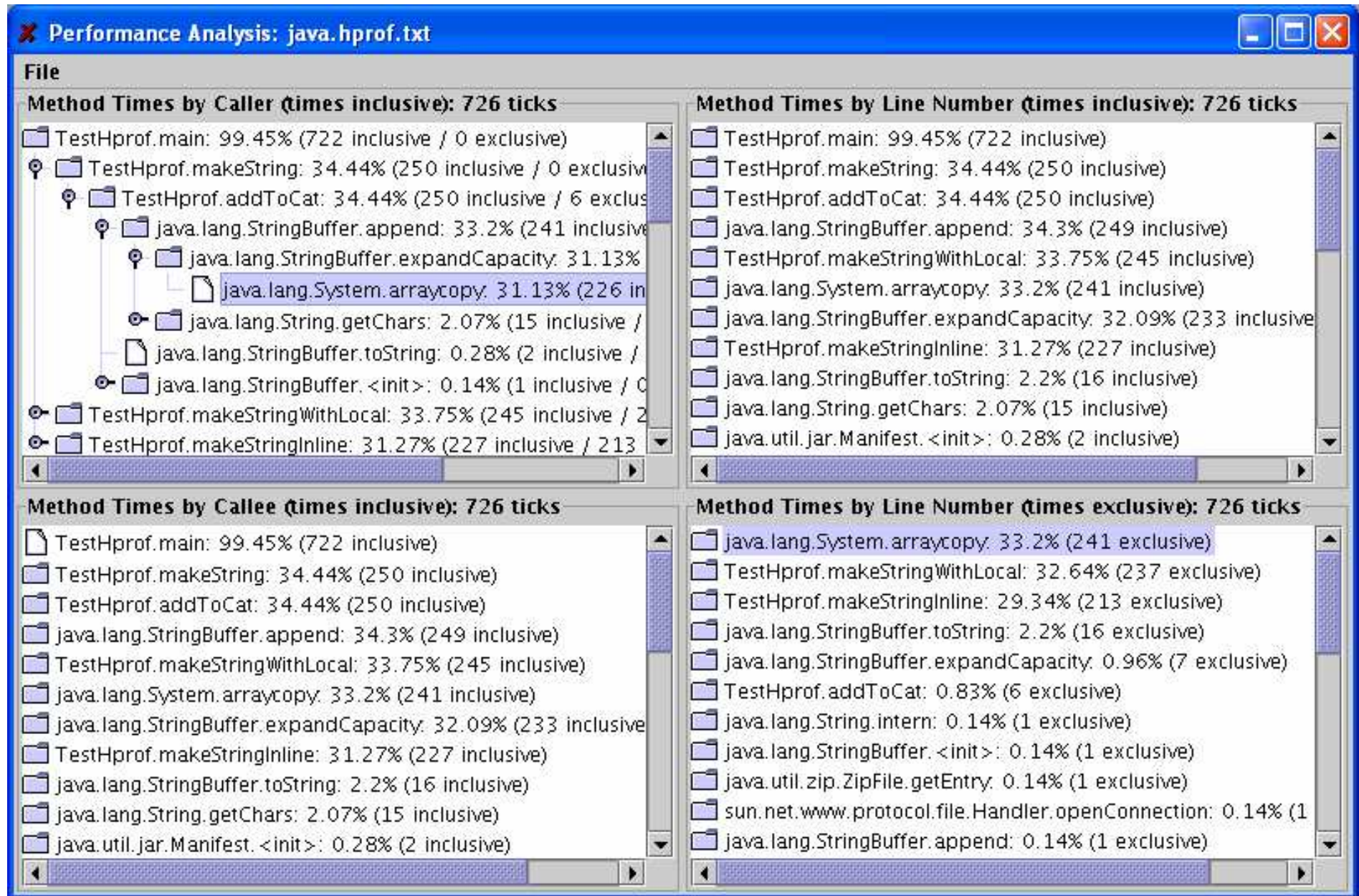
java/runhprof

- SUN's JVM extension
`java -Xrunhprof:cpu=samples,depth=6,thread=y prog`
- generates `java.hprof.txt` file
- analyzed with perfanal :
`java -jar PerfAnal.jar java.hprof.txt`
- memory :
`java -Xrunhprof:heap=all prog`

Drawbacks

- coarse sampling
- does not use all the possibilities of JVMPi

Using the tools : runhprof



Using the tools :

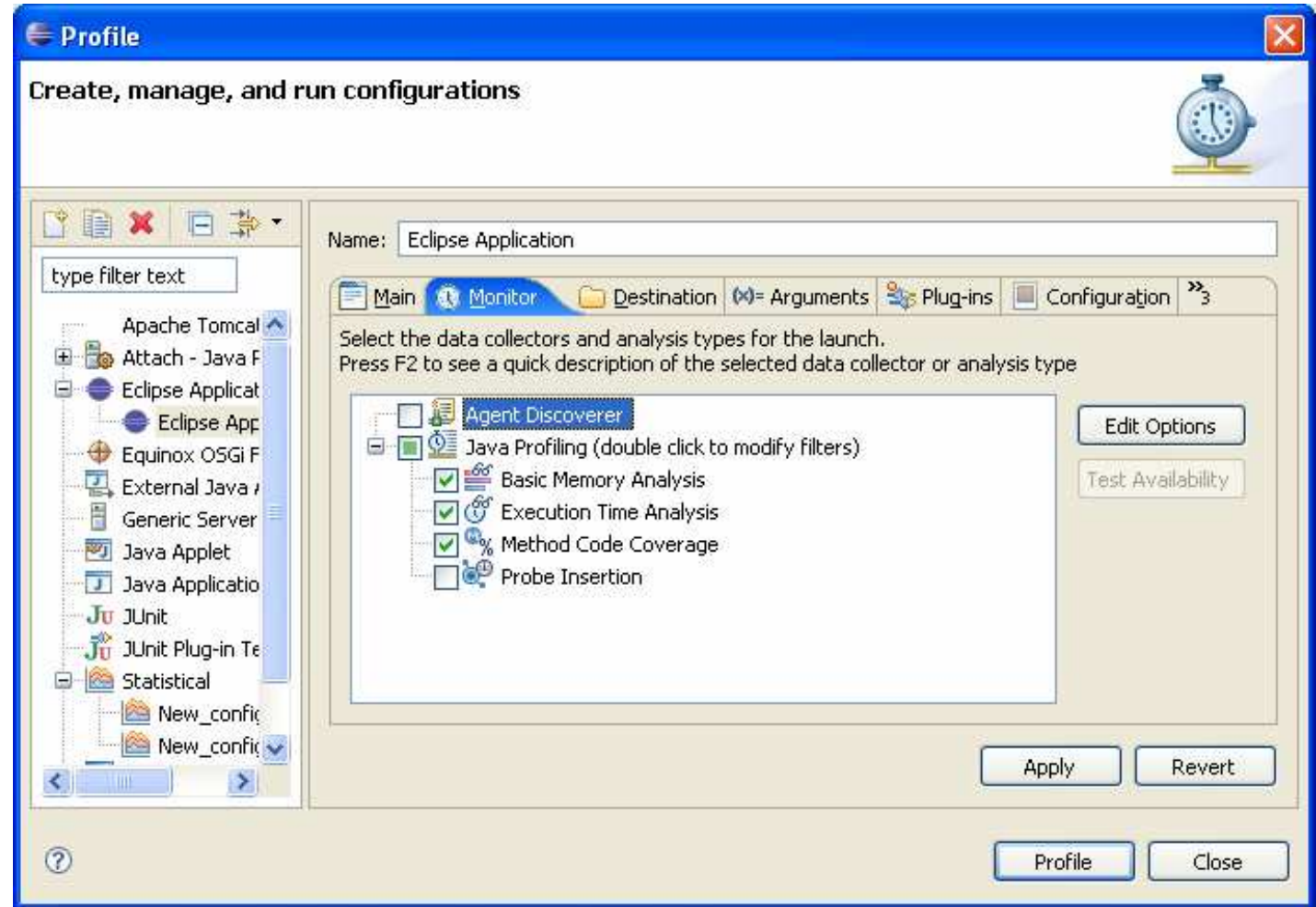
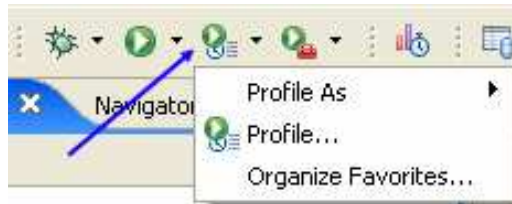
Eclipse Test & Performance Tools Platform

TPTP : tools integrated into Eclipse

Supports only Java (for the moment)

Profiling of local or distributed software

Using the tools : Eclipse Test & Performance Tools Platform



Using the tools : Eclipse Test & Performance Tools Platform

The screenshot shows the Eclipse IDE interface. The **Profiling Monitor** window is open, displaying a tree view of the application's execution. The **Log Navigator** shows the current execution state. A context menu is open over the **Method Code Coverage** item, listing various visualization types. A blue arrow points to the **Coverage Statistics** option in this menu. The **Coverage Statistics** window is also open, showing a table of coverage data for the package **com.sysdeo.tprcp**.

Liste des types de visualisation

- Coverage Statistics
- Execution Flow
- Execution Statistics
- Memory Statistics
- Object References
- UML2 Class Interactions
- UML2 Object Interactions
- UML2 Thread Interactions

>Package	Average Base ...	Cumulative Time...	Calls
com.sysdeo.tprcp.editors	0,005750	0,104097	18
com.sysdeo.tprcp.extensions	0,000178	0,000533	3
com.sysdeo.tprcp.modele	0,003338	0,854440	256
Appellation	0,000008	0,000612	76
Bouteille	0,000005	0,000683	136
Cave	0,000175	0,001398	8
CaveModelePlugin	0,053214	0,852767	16
CaveModelePlugin()	0,000025	0,000025	1
chargerCave() void	0,841824	0,843467	1
enregistrerCave() void	0,000000	0,000000	0
getCave() com.sysdeo.tp	0,000005	0,852240	5
getChargeurEnregistreur	0,000000	0,000000	0
getDefault() com.sysdeo	0,000001	0,000005	4
netMmmFichierCave() iav	0,004453	0,008909	2

Outline

- Profiling
- Tools
- Optimization

Optimization

- No premature optimization
- Keep the code maintainable
- Do not over-optimize

Optimization (continued)

- Find a better algorithm
 - the constant factor of the complexity can be significant
- Memory access : first cause of slowness
- Use already optimized libraries
- Limit the number of calls to expensive functions
- Write performance benchmarks/tests
 - allows one to check that the performance has not degraded
- The bottleneck moves at each optimization step
 - example : I/O can become blocking

Optimization example

- Optimization example : image inversion (5000x5000)

```
for (int x = 0; x < w; x++)  
    for (int y = 0; y < h; y++)  
        data[y][x] = 255 - data[y][x];
```

- Without compiler optimization : 435 ms
- Compiler optimization (-O3) : 316 ms
- Improve memory access locality : 107 ms
- Suppress double dereference on data : 94 ms
- Constant code outside of the loop : 63 ms (~7x)
- OpenMP parallelization : 38 ms
- Using MMX assembly : 26 ms

Conclusion

- No premature optimization
- Know your profiling tool
- Keep the code maintainable
- Do not over-optimize

Questions ?

