Why Aren't More Users More Happy With Our VMs?

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Tratt

Warmup work in collaboration with:
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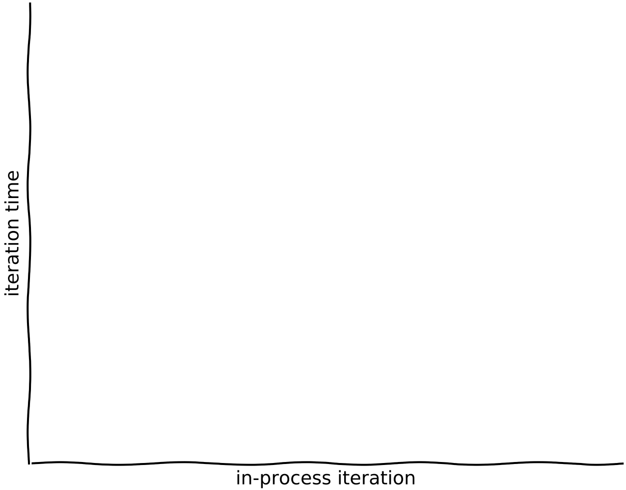
JVMs bring "gcc -O2" to the masses

–*Cliff Click*: A JVM does that?

What do VM claims pertain to?



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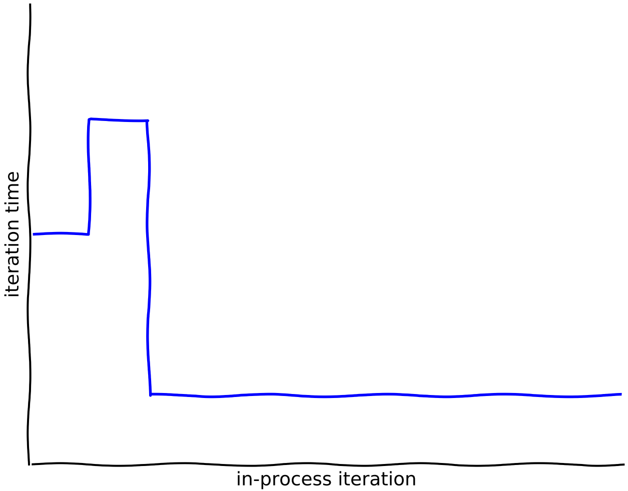

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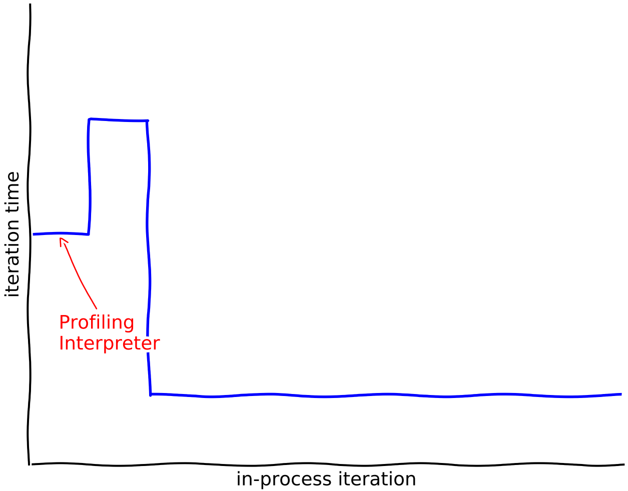

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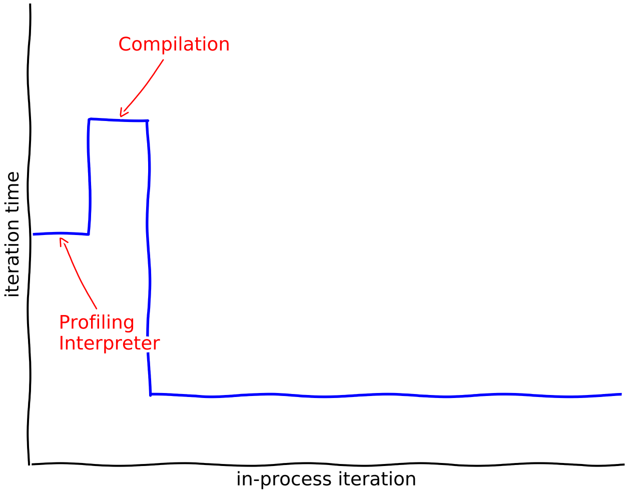

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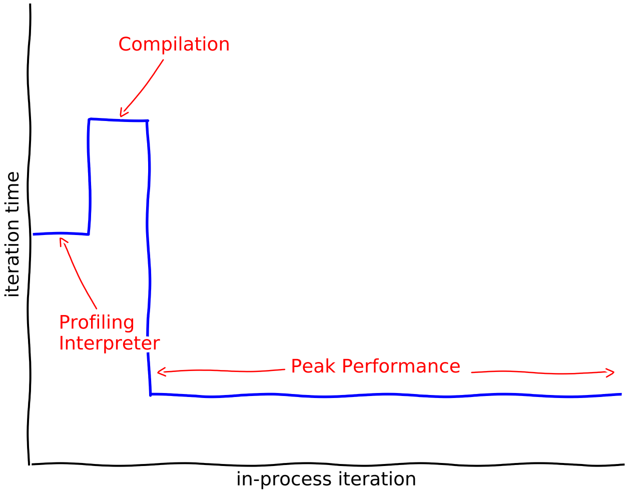

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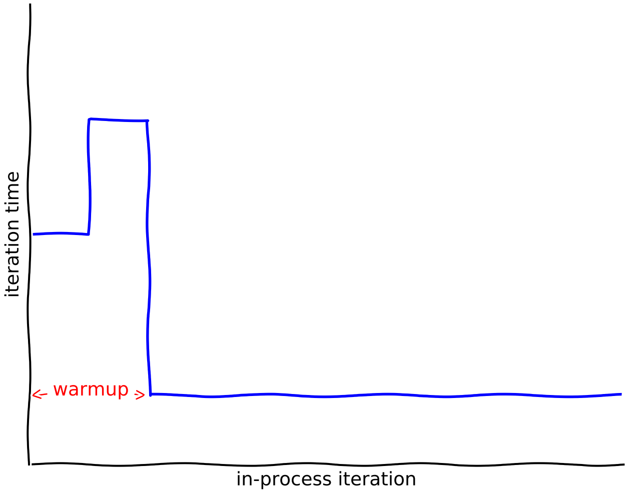

Users *always* perceive warmup

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Maybe we should know how long it is?

Measure warmup of modern language implementations

The Warmup Experiment

Measure warmup of modern language implementations

Hypothesis: Small, deterministic programs reach a steady state of peak performance.

Method 1: Which benchmarks?

The language benchmark games are perfect for us (unusually)

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The language benchmark games are perfect for us (unusually)

We removed any CFG non-determinism

We added checksums to all benchmarks

Method 2: How long to run?

2000 *in-process iterations*

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2000 *in-process iterations*

30 *process executions*

Method 3: VMs

- Graal-0.22
- HHVM-3.19.1
- TruffleRuby 20170502
- Hotspot-8u121b13
- LuaJit-2.0.4
- PyPy-5.7.1
- V8-5.8.283.32
- GCC-4.9.4

Note: same GCC (4.9.4) used for all compilation

Method 4: Machines

- Linux_{4790}, Debian 8, 24GiB RAM
- $\text{Linux}_{\text{E3-1240v5}}$, Debian 8, 32GiB RAM
- OpenBSD_{4790}, OpenBSD 6.0, 32GiB RAM

Method 4: Machines

- Linux_{4790}, Debian 8, 24GiB RAM
- $\text{Linux}_{\text{E3-1240v5}}$, Debian 8, 32GiB RAM
- OpenBSD_{4790}, OpenBSD 6.0, 32GiB RAM

- Turbo boost and hyper-threading disabled
- Network card turned off.
- Daemons disabled (cron, smtpd)

Method 5: Krun

Benchmark runner: tries to control as many confounding variables as possible

Benchmark runner: tries to control as many confounding variables as possible e.g.:

- Minimises I/O
- Sets fixed heap and stack ulimits
- Drops privileges to a 'clean' user account
- Automatically reboots the system prior to each proc. exec
- Reruns any proc. exec where the CPU was throttled
- Checks `dmesg` for changes after each proc. exec
- Checks system at (roughly) same temperature for proc. execs
- Enforces kernel settings (tickless mode, CPU governors, ...)

And now for data...

The experiment has gone through many versions.

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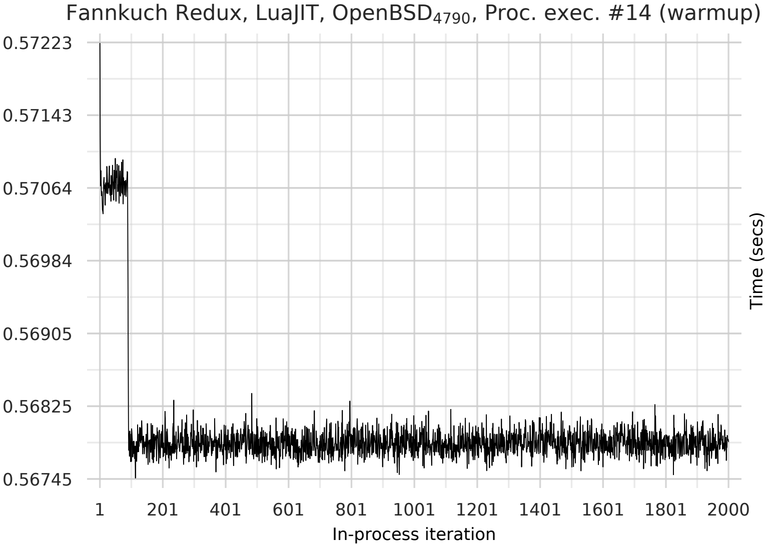
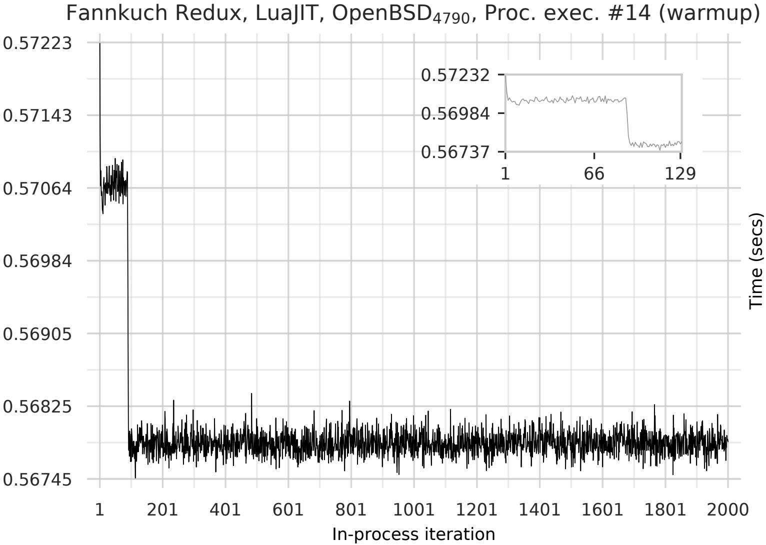
The following data is from the 1.5 run.

Warmup & flat (1)



Warmup & flat (1)



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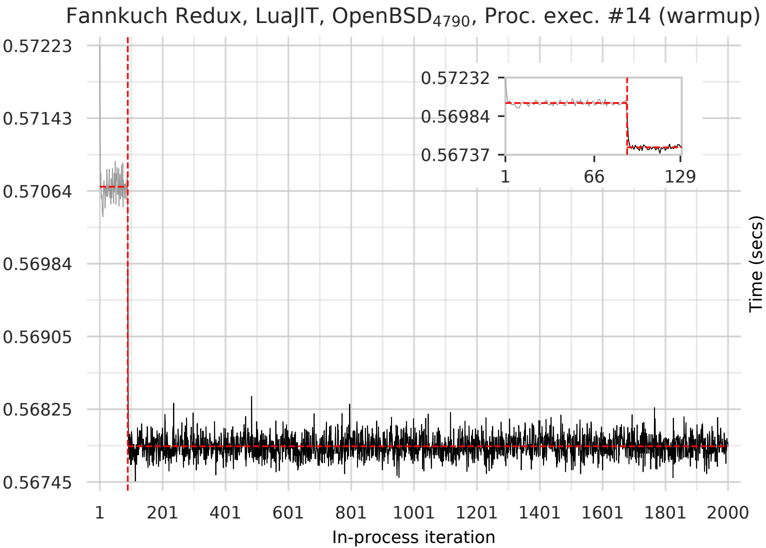
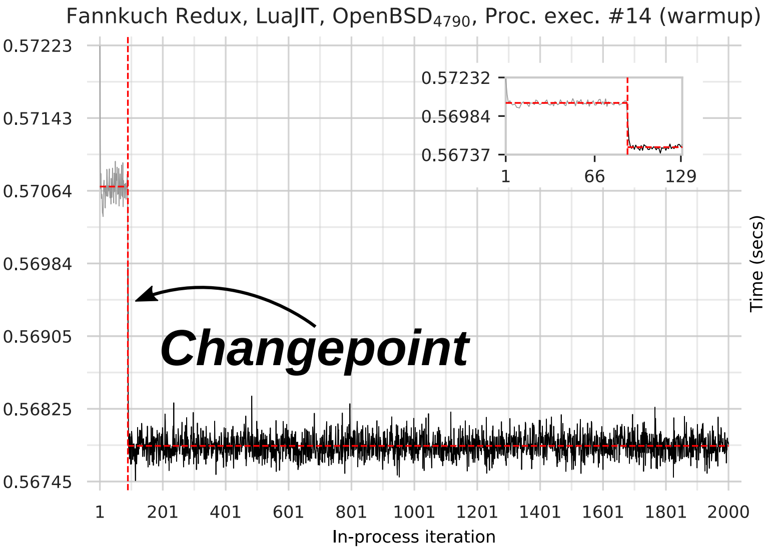
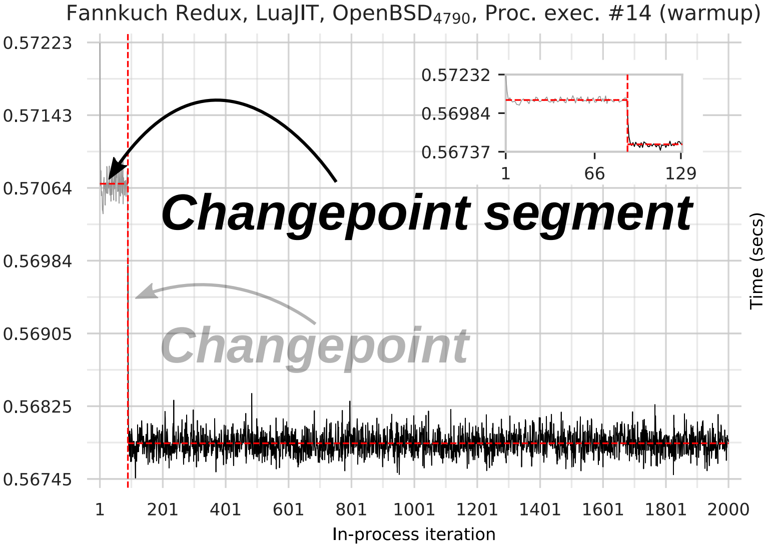



Fannkuch Redux, LuaJIT, OpenBSD4790, Proc. exec. #14 (warmup)
0.57223
0.57143
0.57064
0.56984
0.56905
0.56825
0.56745
1
201
401
601
801
1001
1201
1401
1601
1801
2000
In-process iteration
Time (secs)
0.57232
0.56984
0.56737
1
66
129
Changepoint

Warmup & flat (1)



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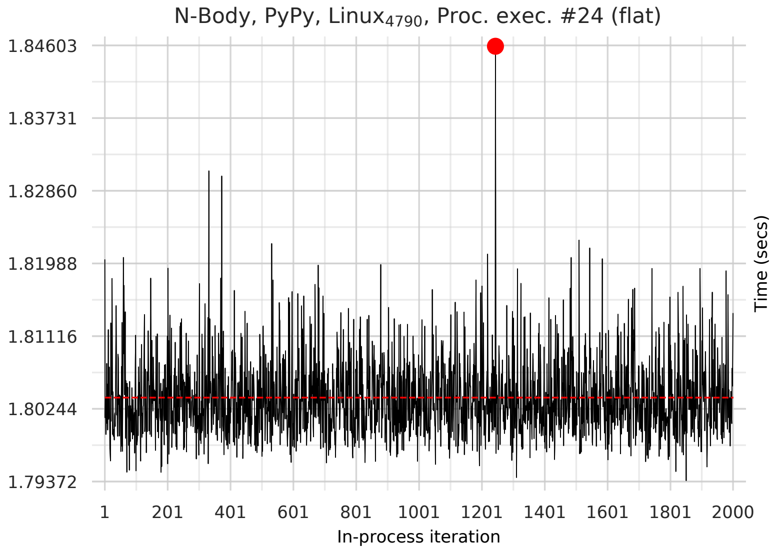


Classification algorithm (steps in order):

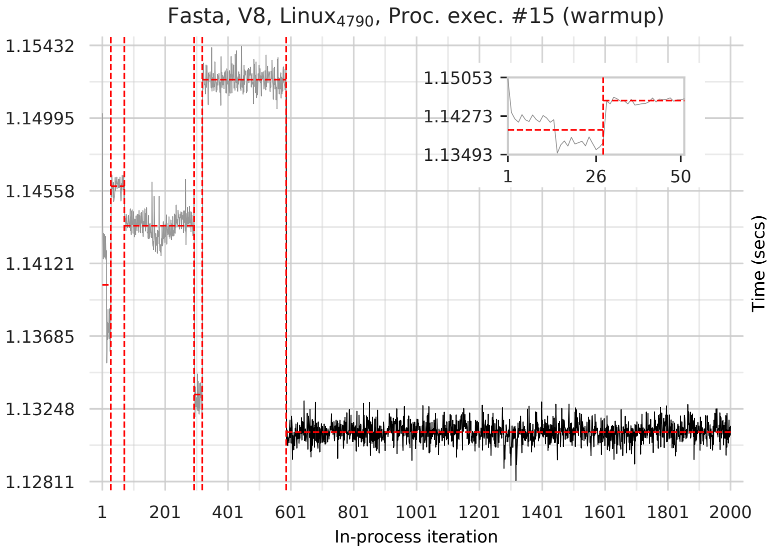
All segs are equivalent: *flat*

Method 7: Classification

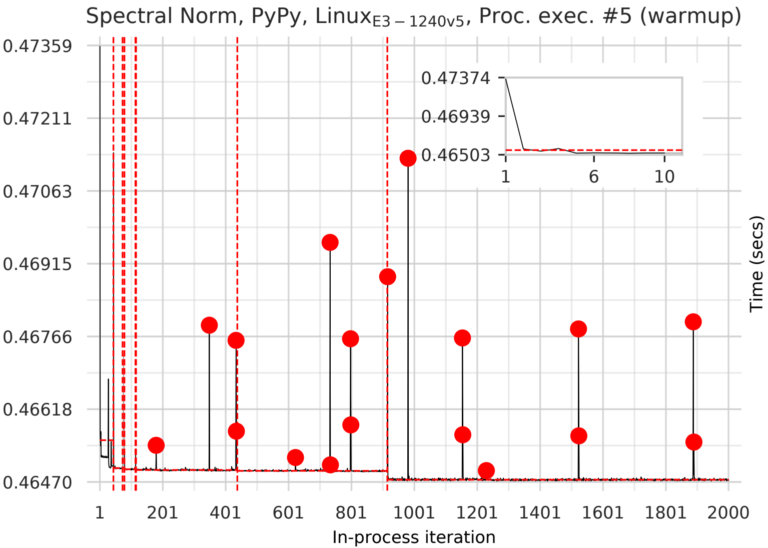
Classification algorithm (steps in order):

- All segs are equivalent: *flat*
- Final seg is in fastest set: *warmup*

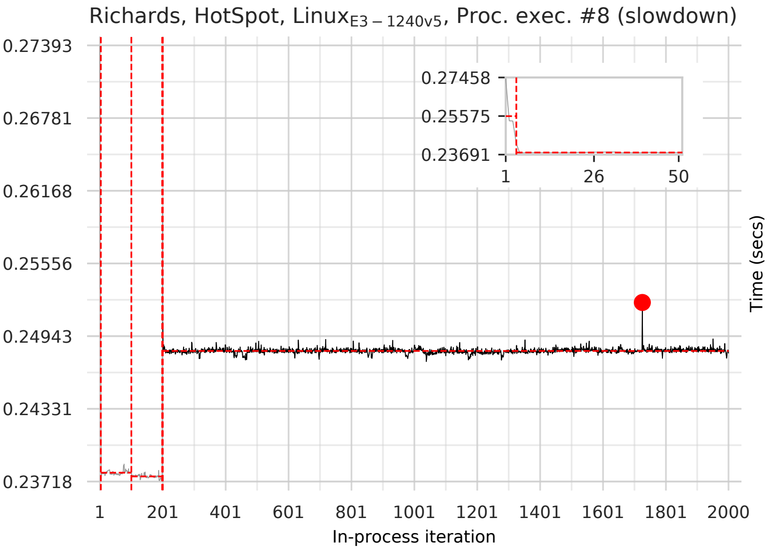
Warmup & flat (2)



Warmup & flat (2)



Slowdown (1)



Richards, HotSpot, Linux$_{E3-1240v5}$, Proc. exec. #8 (slowdown)
0.27393
0.26781
0.26168
0.25556
0.24943
0.24331
0.23718
0.27458
0.25575
0.23691
1
26
50
Time (secs)
1
201
401
601
801
1001
1201
1401
1601
1801
2000
In-process iteration

Method 7: Classification

Classification algorithm (steps in order):

- All segs are equivalent: *flat*
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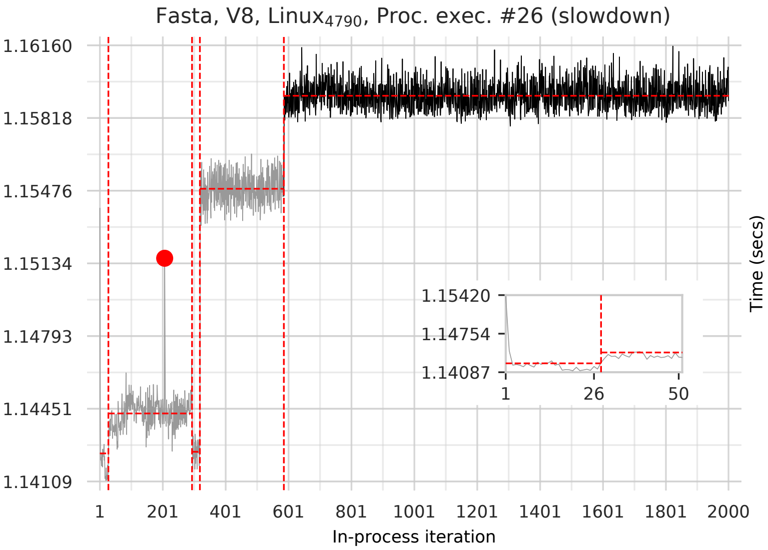
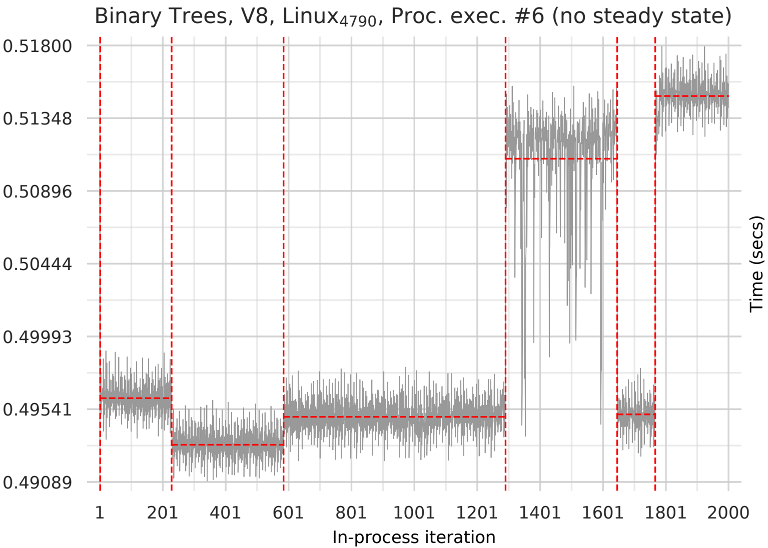
Slowdown (2)



Fasta, V8, Linux$_{4790}$, Proc. exec. #26 (slowdown)
1.16160
1.15818
1.15476
1.15134
1.14793
1.14451
1.14109
1
201
401
601
801
1001
1201
1401
1601
1801
2000
In-process iteration
Time (secs)
1.15420
1.14754
1.14087
1
26
50

No steady state (1)



Classification (3)

Classification algorithm (steps in order):

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Classification algorithm, in order:

- All segs are equivalent: *flat*
- Final seg is in fastest set: *warmup*
- Final seg is not in fastest set: *slowdown*
- Else: *no steady state*

Good

Classification algorithm, in order:

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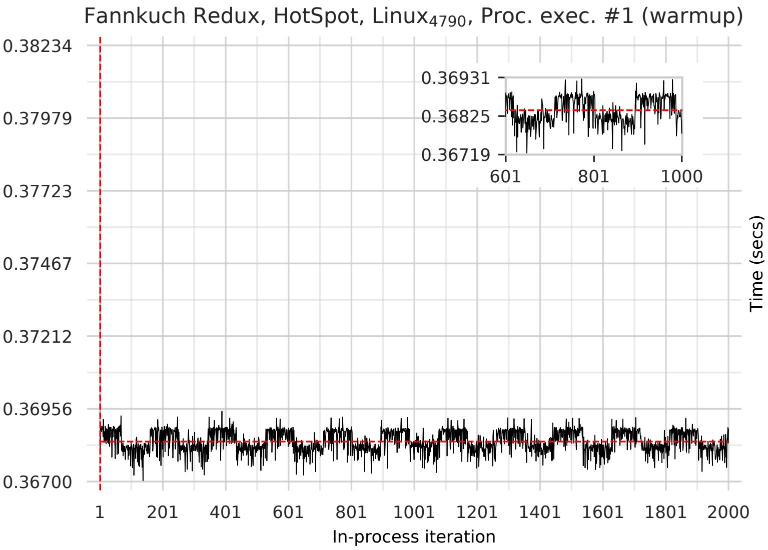
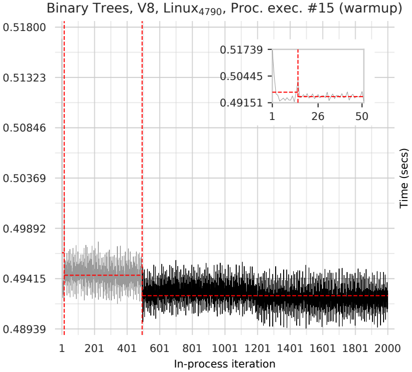
- Final seg is in fastest set: *warmup*
- Final seg is not in fastest set: *slowdown*
- Else: *no steady state*

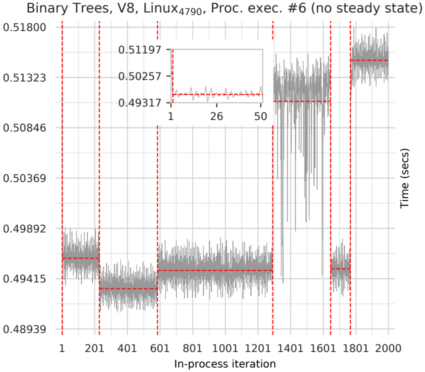
Bad

Warmup or no steady state?



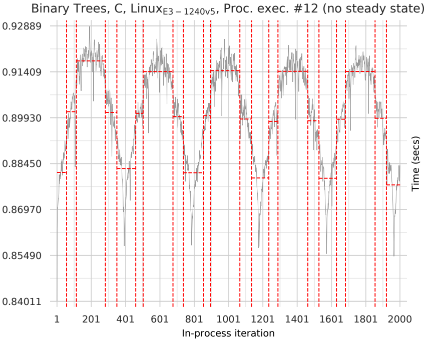
Inconsistent Process-executions

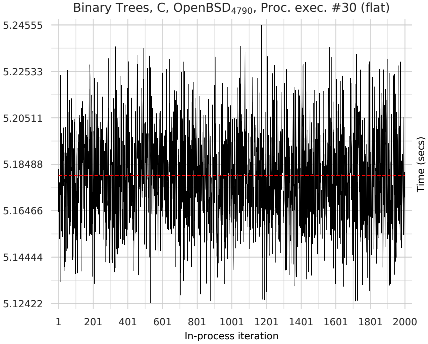




(Same machine)

Inconsistent Process-executions





(Different machines. Bouncing ball Linux-specific)

Individual benchmark stats

		Class.	Steady iter (#)	Steady iter (s)	Steady perf (s)		Class.	Steady iter (#)	Steady iter (s)	Steady perf (s)
binary trees	C	∿				*n-body*	—			0.40557 ±0.006155
	Graal	⨯ (27⅂, 2⅃, 1∿)					⅂	6.0 (5.0,7.0)	0.86 (0.704,1.090)	0.13677 ±0.000035
	HHVM	⨯ (19⅂, 7⅃, 4∿)								
	HotSpot	⨯ (15⅂, 13⅃, 2∿)					⅂	2.0 (2.0,2.0)	0.14 (0.142,0.151)	0.13699 ±0.000029
	LuaJIT	⨯ (25⅂, 3∿, 2⅃)					= (29—, 1⅂)	1.0 (1.0,1.0)	0.00 (0.000,0.000)	0.25401 ±0.002040
	PyPy	⨯ (27∿, 3⅃)					—			1.83314 ±0.024563
	TruffleRuby	⨯ (19⅃, 11⅂)	1003.0 (283.8,1009.2)	2621.96 (749.093,2639.554)	2.61136 ±0.036511		⨯ (28⅂, 2⅃)	68.0 (68.0,1003.0)	40.45 (40.056,481.916)	0.46998 ±0.001516
	V8	⨯ (13—, 9⅃, 8⅂)	228.0 (1.0,588.0)	106.34 (0.000,275.806)	0.46927 ±0.003135		= (20—, 10⅂)	1.0 (1.0,363.0)	0.00 (0.000,87.733)	0.24148 ±0.000304
fannkuch redux	C	⨯ (25—, 5⅃)	1.0 (1.0,538.9)	0.00 (0.000,224.243)	0.41742 ±0.011343	*Richards*	⨯ (18—, 8⅃, 4∿)			
	Graal	⨯ (26⅂, 4∿)					⅂	2.0 (2.0,35.3)	0.95 (0.879,9.745)	0.26465 ±0.007761
	HHVM	⨯ (28⅂, 2∿)								
	HotSpot	⨯ (17⅃, 10⅂, 3∿)					⨯ (28⅃, 1∿, 1⅂)			
	LuaJIT	—			0.56287 ±0.000113		⨯ (18∿, 7⅃, 5⅂)			
	PyPy	⨯ (19—, 9⅂, 2⅃)	1.0 (1.0,1120.3)	0.00 (0.000,1802.797)	1.60962 ±0.022821		⅂	2.0 (2.0,2.0)	1.04 (1.023,1.102)	0.88940 ±0.035495
	TruffleRuby	⨯ (26⅃, 3∿, 1⅂)					⅃	1021.0 (334.6,1068.9)	1521.33 (499.847,1585.296)	1.48585 ±0.011040
	V8	= (28⅂, 2—)	2.0 (1.5,25.0)	0.32 (0.142,7.583)	0.30634 ±0.000160		⅂	3.0 (3.0,189.8)	1.04 (0.961,97.303)	0.47681 ±0.020468
fasta	C	—			0.06848 ±0.000167	*spectralnorm*	⨯ (27⅂, 2—, 1⅃)	775.0 (1.5,780.0)	425.16 (0.246,426.809)	0.54581 ±0.033116
	Graal	⅂	4.0 (3.0,34.4)	0.75 (0.535,5.873)	0.16188 ±0.000738		⅃	14.0 (2.0,94.6)	13.60 (0.830,98.737)	1.05685 ±0.000126
	HHVM	⨯ (13⅃, 10⅂, 7∿)					⨯ (29⅂, 1∿)			
	HotSpot	⨯ (27⅂, 3⅃)	6.0 (5.0,595.0)	0.62 (0.500,70.062)	0.11691 ±0.001949		⅂	7.0 (7.0,7.5)	1.91 (1.902,3.645)	0.31472 ±0.169143
	LuaJIT	⨯ (19⅃, 9—, 1∿, 1⅂)					—			0.22181 ±0.000039
	PyPy	∿					= (27—, 3⅂)	1.0 (1.0,45.2)	0.00 (0.000,20.597)	0.46480 ±0.000085
	TruffleRuby						⨯ (25⅃, 5∿)			
	V8	⨯ (27⅃, 3⅂)	586.0 (318.0,587.6)	612.12 (330.270,622.328)	1.04904 ±0.020905		⅂	3.0 (3.0,3.0)	0.52 (0.523,0.526)	0.25362 ±0.000034

Individual benchmark stats

		Class.	Steady iter (#)	Steady iter (s)	Steady perf (s)
C	*spectralnorm*	⤫ (27⅂, 2—, 1⅃)	775.0 (1.5,780.0)	425.16 (0.246,426.809)	0.54581 ±0.033116
Graal		⅃	14.0 (2.0,94.6)	13.60 (0.830,98.737)	1.05685 ±0.000126
HHVM		⤫ (29⅂, 1∿)			
HotSpot		⅂	7.0 (7.0,7.5)	1.91 (1.902,3.645)	0.31472 ±0.169143
LuaJIT		—			0.22181 ±0.000039
PyPy		= (27—, 3⅂)	1.0 (1.0,45.2)	0.00 (0.000,20.597)	0.46480 ±0.000085
TruffleRuby		⤫ (25⅃, 5∿)			
V8		⅂	3.0 (3.0,3.0)	0.52 (0.523,0.526)	0.25362 ±0.000034

Overall benchmark stats

Class.	Linux_{4790}	Linux_{1240v5}	OpenBSD_{4790} †
		⟨VM, benchmark⟩ pairs	
—	8.9%	11.1%	13.3%
‾|_	20.0%	17.8%	20.0%
_|‾	4.4%	4.4%	3.3%
⋀⋁	4.4%	4.4%	0.0%
=	11.1%	8.9%	13.3%
⋇	51.1%	53.3%	50.0%
		Process executions	
—	22.0%	23.3%	37.7%
‾|_	48.3%	43.9%	35.2%
_|‾	20.1%	22.1%	12.1%
⋀⋁	9.6%	10.8%	15.0%

Overall benchmark stats

Class.	Linux_{4790}	Linux_{1240v5}	OpenBSD_{4790} †
		⟨VM, benchmark⟩ pairs	
—	8.9%	11.1%	13.3%
⎿	20.0%	17.8%	20.0%
⎾	4.4%	4.4%	3.3%
ʌʌ	4.4%	4.4%	0.0%
=	11.1%	8.9%	13.3%
≭	51.1%	53.3%	50.0%
		Process executions	
—	22.0%	23.3%	37.7%
⎿	48.3%	43.9%	35.2%
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ʌʌ	9.6%	10.8%	15.0%

Classical warmup occurs for only:

Summary

Classical warmup occurs for only:

67.2%–70.3% of process executions

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67.2%–70.3% of process executions

37.8%–40.0% of (VM, benchmark) pairs

Summary

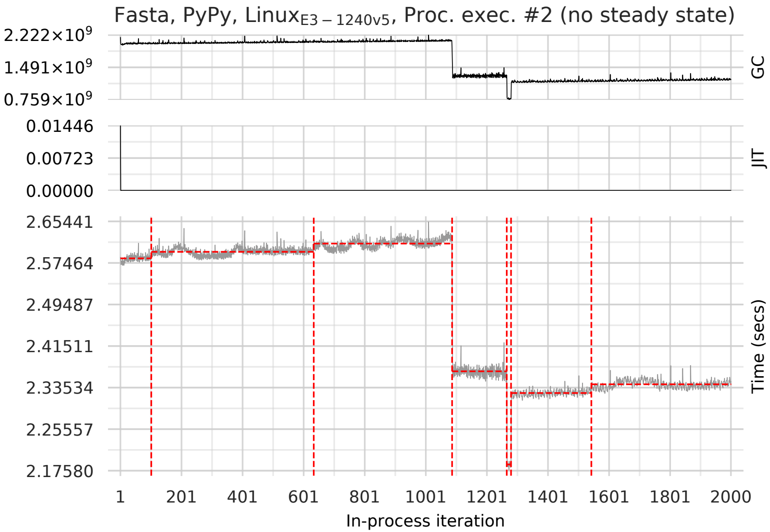
Classical warmup occurs for only:

67.2%–70.3% of process executions

37.8%–40.0% of (VM, benchmark) pairs

12.5% of benchmarks for (VM, benchmark, machine) triples

Are odd effects correlated with compilation and GC?



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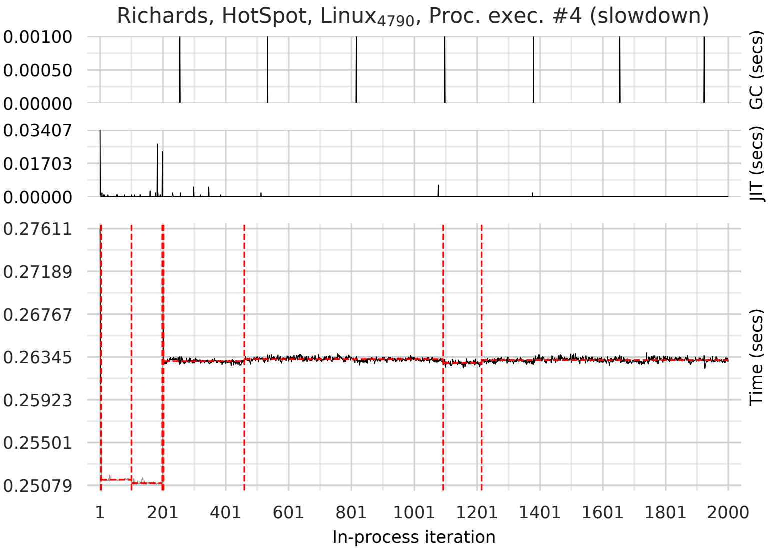

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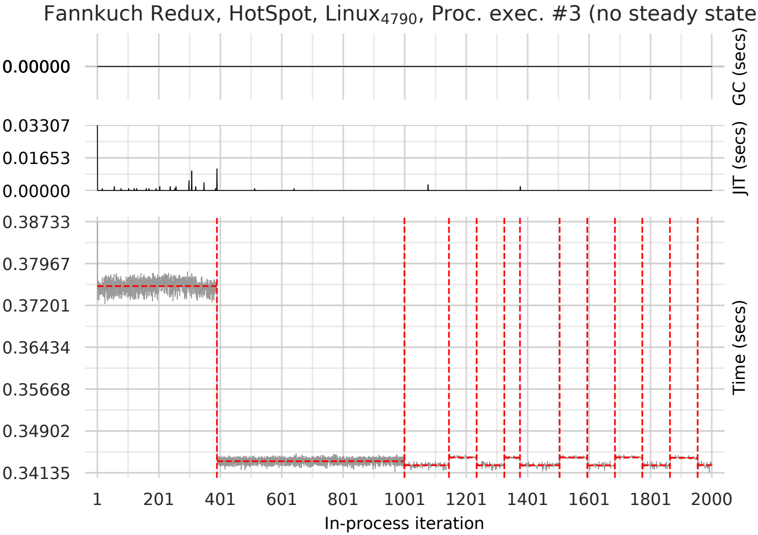

Benchmark suites

Benchmarks guide our optimisations

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Are they complete guides?

A war story

Symptom: poor performance of a Pyston benchmark on PyPy

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Cause: RPython traces recursion

A war story

Symptom: poor performance of a Pyston benchmark on PyPy

Cause: RPython traces recursion

Fix: Check for recursion before tracing

A war story: the basis of a fix

```
diff --git a/rpython/jit/metainterp/pyjitpl.py b/rpython/jit/metainterp/pyjitpl.py
--- a/rpython/jit/metainterp/pyjitpl.py
+++ b/rpython/jit/metainterp/pyjitpl.py
@@ -951,9 +951,31 @@
         if warmrunnerstate.inlining:
             if warmrunnerstate.can_inline_callable(greenboxes):
+                # We've found a potentially inlinable function; now we need to
+                # see if it's already on the stack. In other words: are we about
+                # to enter recursion? If so, we don't want to inline the
+                # recursion, which would be equivalent to unrolling a while
+                # loop.
                 portal_code = targetjitdriver_sd.mainjitcode
-                return self.metainterp.perform_call(portal_code, allboxes,
-                                                    greenkey=greenboxes)
+                inline = True
+                if self.metainterp.is_main_jitcode(portal_code):
+                    for gk, _ in self.metainterp.portal_trace_positions:
+                        if gk is None:
+                            continue
+                        assert len(gk) == len(greenboxes)
+                        i = 0
+                        for i in range(len(gk)):
+                            if not gk[i].same_constant(greenboxes[i]):
+                                break
+                        else:
+                            # The greenkey of a trace position on the stack
+                            # matches what we have, which means we're definitely
+                            # about to recurse.
+                            inline = False
+                            break
+                if inline:
+                    return self.metainterp.perform_call(portal_code, allboxes,
+                                greenkey=greenboxes)
```

A war story: mixed fortunes

Success: slow benchmark now 13.5x faster

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Failure: some PyPy benchmarks slow down

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Success: slow benchmark now 13.5x faster

Failure: some PyPy benchmarks slow down

Solution: allow *some* tracing into recursion

A war story: data

#unrollings	1	2	3	5	7	10
hexiom2	1.3	1.4	1.1	1.0	1.0	1.0
raytrace-simple	3.3	3.1	2.8	1.4	1.0	1.0
spectral-norm	3.3	1.0	1.0	1.0	1.0	1.0
sympy_str	1.5	1.0	1.0	1.0	1.0	1.0
telco	4	2.5	2.0	1.0	1.0	1.0
polymorphism	0.07	0.07	0.07	0.07	0.08	0.09

http://marc.info/?l=pypy-dev&m=141587744128967&w=2

A war story: conclusion

The benchmark suite said 7 levels, so that’s what I suggested

A war story: conclusion

The benchmark suite said 7 levels, so that's what I suggested

Even though I doubted it was the right global value

Benchmark suites (2)

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Benchmarks guide our optimisations

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Are they correct guides?

Octane

17 JavaScript benchmarks from V8

Octane

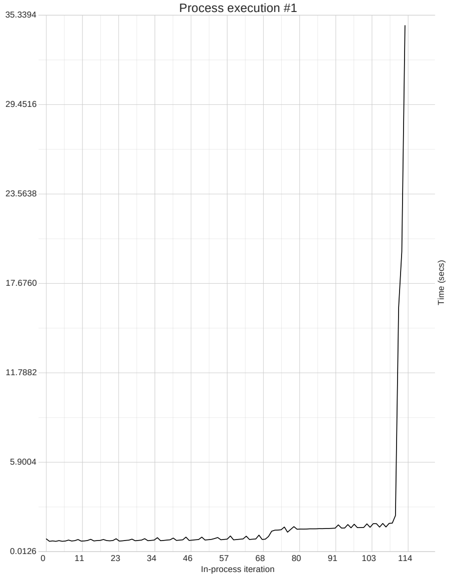
17 JavaScript benchmarks from V8

Let’s make each benchmark run for 2000 iterations

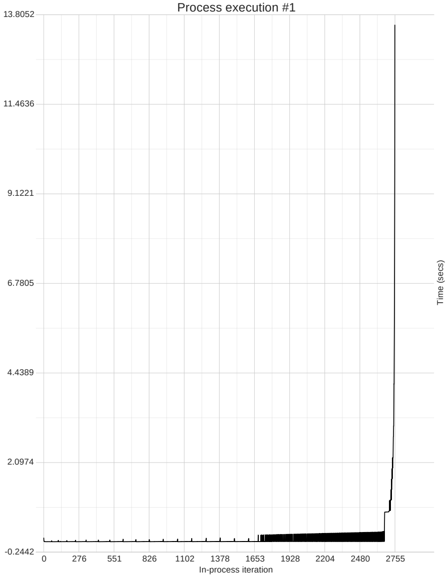
Octane: pdf.js explodes

```
$ d8 run.js
Richards
DeltaBlue
Encrypt
Decrypt
RayTrace
Earley
Boyer
RegExp
Splay
NavierStokes
PdfJS

<--- Last few GCs --->

14907865 ms: Mark-sweep 1093.9 (1434.4) -> 1093.4 (1434.4) MB, 274.8 / 0.0 ms [allocation failure] [GC in old space
14908140 ms: Mark-sweep 1093.4 (1434.4) -> 1093.3 (1434.4) MB, 274.4 / 0.0 ms [allocation failure] [GC in old space
14908421 ms: Mark-sweep 1093.3 (1434.4) -> 1100.5 (1418.4) MB, 280.9 / 0.0 ms [last resort gc].
14908703 ms: Mark-sweep 1100.5 (1418.4) -> 1107.8 (1418.4) MB, 282.1 / 0.0 ms [last resort gc].


<--- JS stacktrace --->

==== JS stack trace =========================================

Security context: 0x20d333ad3ba9 <JS Object>
    2: extractFontProgram(aka Type1Parser_extractFontProgram) [pdfjs.js:17004] [pc=0x3a13b275421b] (this=0x3de358283
    3: new Type1Font [pdfjs.js:17216] [pc=0x3a13b2752078] (this=0x4603fbdaea9 <a Type1Font with map 0x1f822134f7e1>,


#
# Fatal error in CALL_AND_RETRY_LAST
# Allocation failed - process out of memory
#

zsh: illegal hardware instruction  d8 run.js
```

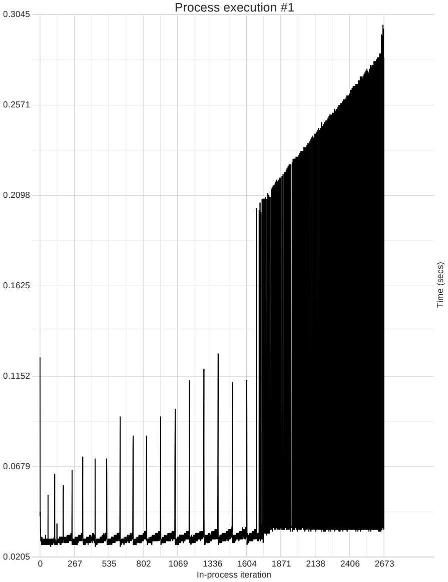
Octane: analysing pdf.js



Octane: analysing pdf.js



Octane: analysing pdf.js

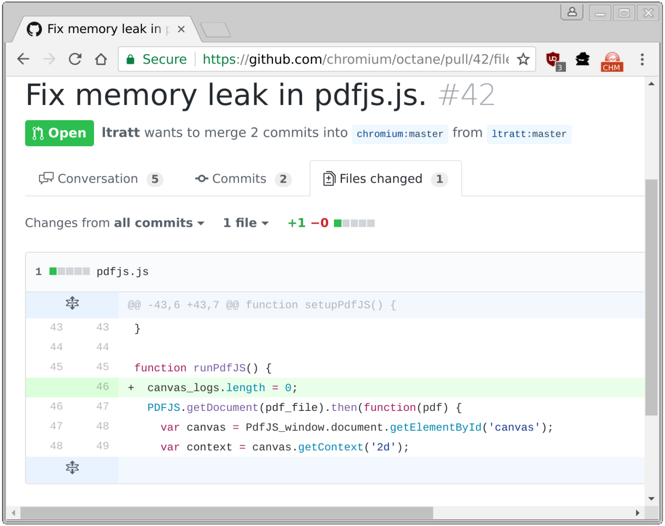


Octane: debugging

pdfjs.js (/tmp/octane) - GVIM2

```
pdfjs.js                                                                    buffers

//////// Benchmark set-up. (c) by Google. ////////

var pdf_file = "test.pdf";
var canvas_logs = [];

var PdfJS = new BenchmarkSuite("PdfJS", [10124921], [
  new Benchmark("PdfJS", false, false, 24,
    runPdfJS, setupPdfJS, tearDownPdfJS, null, 4)
]);
NORMAL  master  pdfjs.js                jav…  utf-8[unix]    0% ¶  28/33053 :  1
function runPdfJS() {
  PDFJS.getDocument(pdf_file).then(function(pdf) {
    var canvas = PdfJS_window.document.getElementById('canvas');
    var context = canvas.getContext('2d');
    var renderContext = {canvasContext: context};
    canvas_logs.push(context.__log__);

    // Cycle through all pages.
    function renderPages(i, j) {
      if (i > j) return;
      context.clearRect(0, 0, canvas.width, canvas.height);
      pdf.getPage(i).then(function(page) {
        renderContext.viewport = page.getViewport(1);
        canvas.height = renderContext.viewport.height;
        canvas.width = renderContext.viewport.width;
        page.render(renderContext).then(renderPages.bind(null, i + 1, j));
      });
    }
    renderPages(1, pdf.numPages);
  });

  // Wait for everything to complete.
  PdfJS_window.flushTimeouts();
}
pdfjs.js                                  jav…  utf-8[unix]    0% ¶  50/33053 : 38
```

Octane: fixing



Octane: other issues

pdfjs isn't the only problem

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zlib complains that `Cannot enlarge memory arrays in asm.js` (a memory leak? I don't know)

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Timings are made with a non-monotonic microsecond timer

Summary

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Why aren’t more users more happy with our VMs?

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My thesis: benchmarking *and* benchmarks are performance destiny.

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My thesis: benchmarking *and* benchmarks are performance destiny.

Ours have misled us.

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1. Run benchmarks for longer to uncover issues.
2. Accept that neither peak performance or a steady state may occur.
3. Always report warmup time.
4. Stop over-training on small benchmark suites.
5. Collect more benchmarks.
6. Focus on predictable performance.

The big question

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At least a bit... but a lot? Unclear.

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In case we can't, I have an idea...

FL Interpreter

```
program_counter = 0; stack = []
vars = {...}
while True:
  jit_merge_point(program_counter)
  instr = load_instruction(program_counter)
  if instr == INSTR_VAR_GET:
    stack.push(
     vars[read_var_name_from_instruction()])
    program_counter += 1
  elif instr == INSTR_VAR_SET:
    vars[read_var_name_from_instruction()]
     = stack.pop()
    program_counter += 1
  elif instr == INSTR_INT:
    stack.push(read_int_from_instruction())
    program_counter += 1
  elif instr == INSTR_LESS_THAN:
    rhs = stack.pop()
    lhs = stack.pop()
    if isinstance(lhs, int) and isinstance(rhs, int):
      if lhs < rhs:
        stack.push(True)
      else:
        stack.push(False)
    else: ...
    program_counter += 1
```

```
  elif instr == INSTR_IF:
    result = stack.pop()
    if result == True:
      program_counter += 1
    else:
      program_counter +=
        read_jump_if_instruction()
  elif instr == INSTR_ADD:
    lhs = stack.pop()
    rhs = stack.pop()
    if isinstance(lhs, int)
     and isinstance(rhs, int):
      stack.push(lhs + rhs)
    else: ...
    program_counter += 1
```

FL Interpreter

```
program_counter = 0; stack = []
vars = {...}
while True:
  jit_merge_point(program_counter)
  instr = load_instruction(program_counter)
  if instr == INSTR_VAR_GET:
    stack.push(
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    program_counter += 1
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    vars[read_var_name_from_instruction()]
     = stack.pop()
    program_counter += 1
  elif instr == INSTR_INT:
    stack.push(read_int_from_instruction())
    program_counter += 1
  elif instr == INSTR_LESS_THAN:
    rhs = stack.pop()
    lhs = stack.pop()
    if isinstance(lhs, int) and isinstance(rhs, int):
      if lhs < rhs:
        stack.push(True)
      else:
        stack.push(False)
    else: ...
    program_counter += 1
```

Meta-tracing JITs

FL Interpreter	User program (lang *FL*)

```
program_counter = 0; stack = []
vars = {...}
while True:
  jit_merge_point(program_counter)
  instr = load_instruction(program_counter)
  if instr == INSTR_VAR_GET:
    stack.push(
     vars[read_var_name_from_instruction()])
    program_counter += 1
  elif instr == INSTR_VAR_SET:
    vars[read_var_name_from_instruction()]
     = stack.pop()
    program_counter += 1
  elif instr == INSTR_INT:
    stack.push(read_int_from_instruction())
    program_counter += 1
  elif instr == INSTR_LESS_THAN:
    rhs = stack.pop()
    lhs = stack.pop()
    if isinstance(lhs, int) and isinstance(rhs, int):
      if lhs < rhs:
        stack.push(True)
      else:
        stack.push(False)
    else: ...
    program_counter += 1
```

```
if x < 0:
  x = x + 1
else:
  x = x + 2
x = x + 3
```

Meta-tracing JITs

FL Interpreter	Initial trace

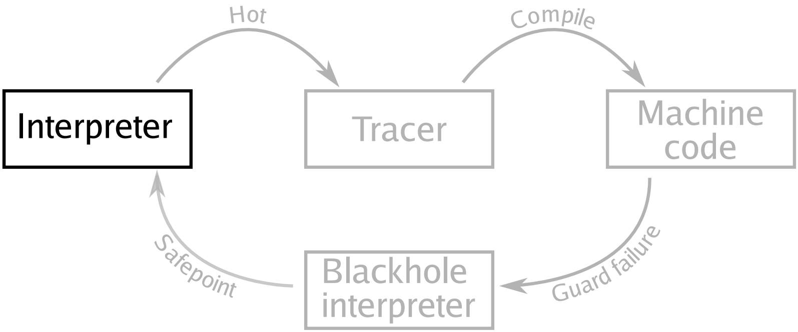
FL Interpreter

```
program_counter = 0; stack = []
vars = {...}
while True:
  jit_merge_point(program_counter)
  instr = load_instruction(program_counter)
  if instr == INSTR_VAR_GET:
    stack.push(
     vars[read_var_name_from_instruction()])
    program_counter += 1
  elif instr == INSTR_VAR_SET:
    vars[read_var_name_from_instruction()]
     = stack.pop()
    program_counter += 1
  elif instr == INSTR_INT:
    stack.push(read_int_from_instruction())
    program_counter += 1
  elif instr == INSTR_LESS_THAN:
    rhs = stack.pop()
    lhs = stack.pop()
    if isinstance(lhs, int) and isinstance(rhs, int):
      if lhs < rhs:
        stack.push(True)
      else:
        stack.push(False)
    else: ...
    program_counter += 1
```

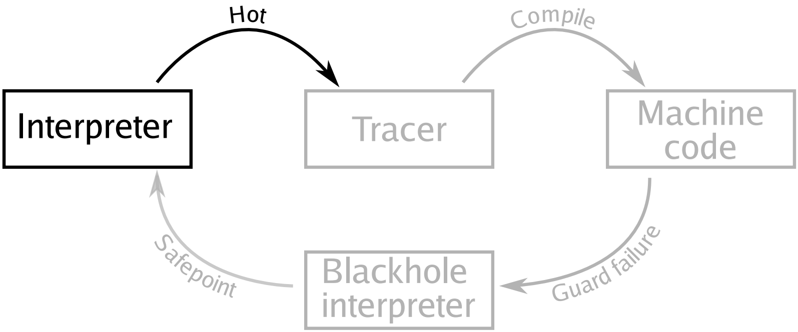
Initial trace

```
v0 = <program_counter>
v1 = <stack>
v2 = <vars>
v3 = load_instruction(v0)
guard_eq(v3, INSTR_VAR_GET)
v4 = dict_get(v2, "x")
list_append(v1, v4)
v5 = add(v0, 1)
v6 = load_instruction(v5)
guard_eq(v6, INSTR_INT)
list_append(v1, 0)
v7 = add(v5, 1)
v8 = load_instruction(v7)
guard_eq(v8, INSTR_LESS_THAN)
v9 = list_pop(v1)
v10 = list_pop(v1)
guard_type(v9, int)
guard_type(v10, int)
guard_not_less_than(v9, v10)
list_append(v1, False)
v11 = add(v7, 1)
v12 = load_instruction(v11)
guard_eq(v12, INSTR_IF)
v13 = list_pop(v1)
guard_false(v13)
...
```

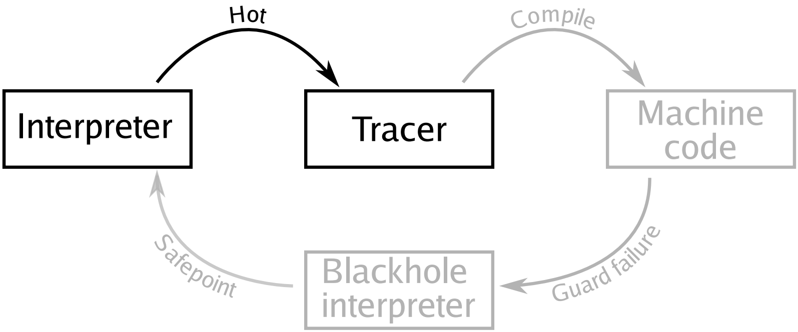
Meta-tracer states

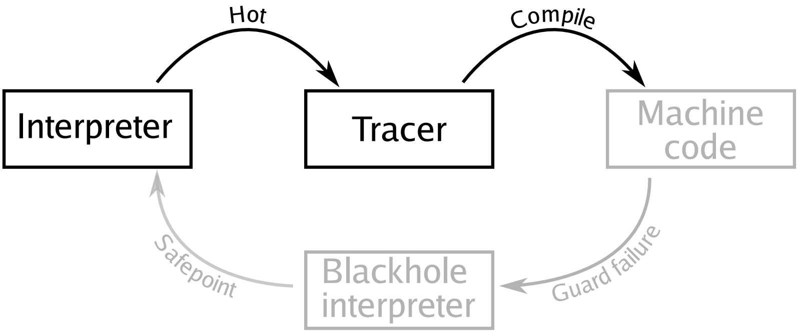


Meta-tracer states

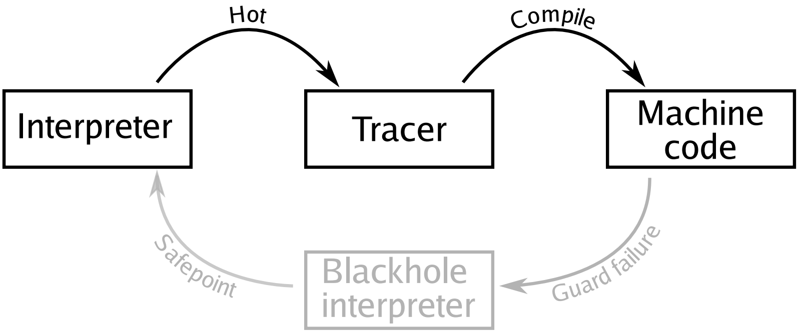


Meta-tracer states




Hot
Compile
Interpreter
Tracer
Machine code
Safepoint
Blackhole interpreter
Guard failure

Meta-tracer states



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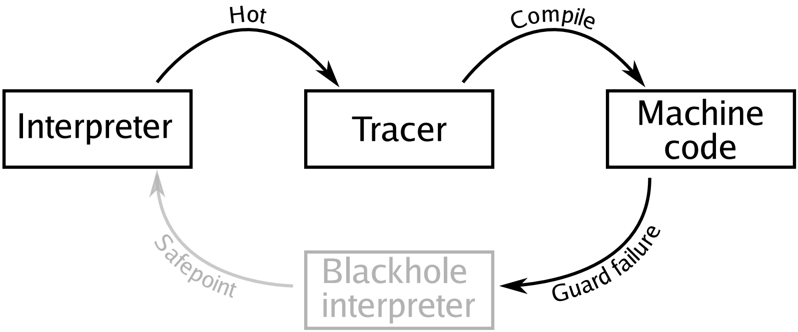

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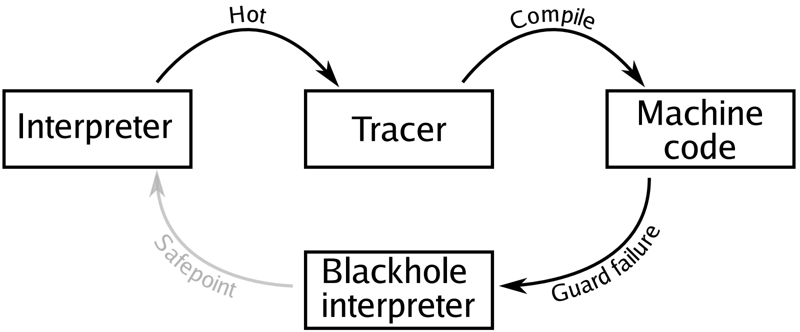

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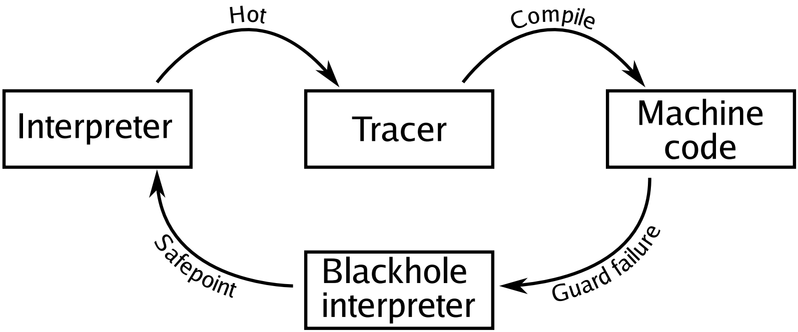
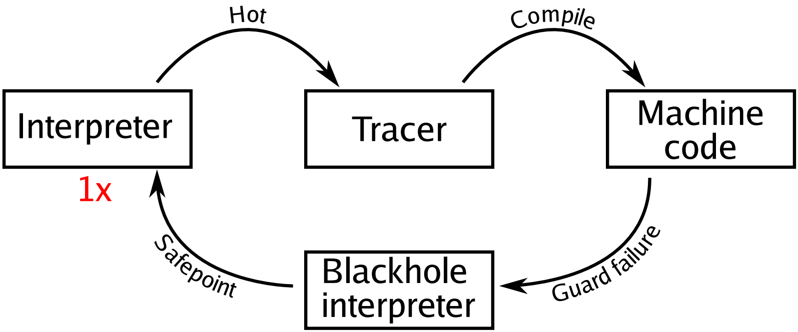


Meta-tracer performance (now)



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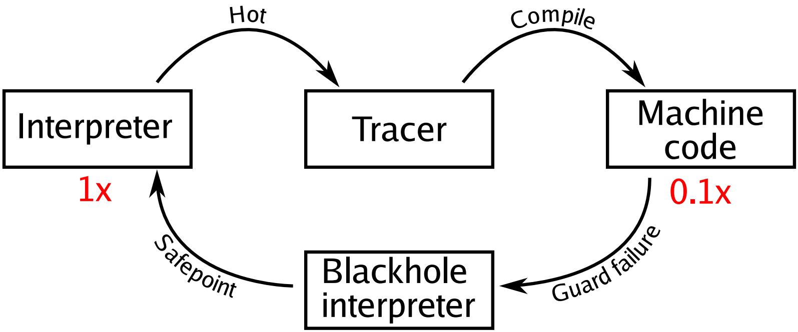

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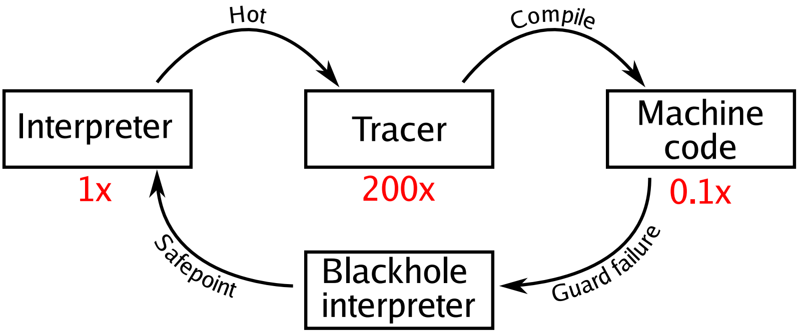
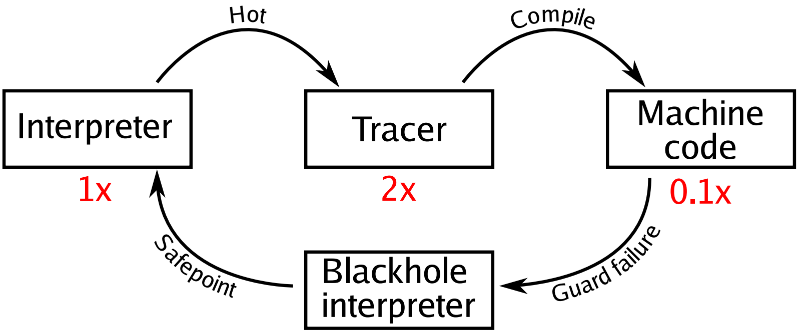


Meta-tracer performance (our aim)



References

VM Warmup Blows Hot and Cold
E. Barrett, C. F. Bolz, R. Killick, V. Knight, S. Mount and L. Tratt.

Rigorous Benchmarking in Reasonable Time
T. Kalibera and R. Jones

Specialising Dynamic Techniques for Implementing the Ruby Programming Language
C. Seaton (Chapter 4)

Quantifying performance changes with effect size confidence intervals
T. Kalibera and R. Jones

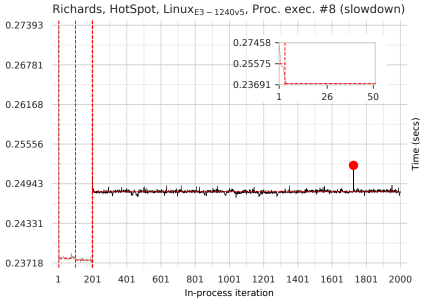
Software

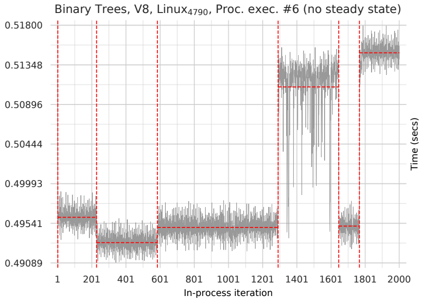
warmup_stats	Use our statistical method on your VMs
	http://soft-dev.org/src/warmup_stats/
Krun	Run experiments in a controlled manner
	http://soft-dev.org/src/krun/

Thanks

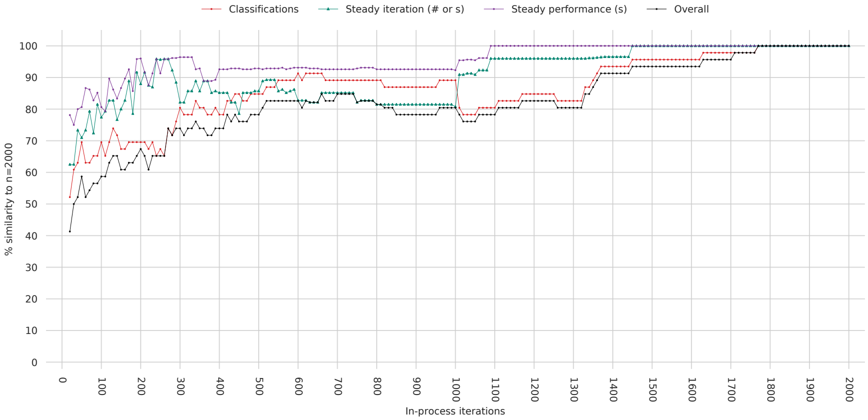
- EPSRC: *COOLER* and *Lecture*.
- Oracle.
- Cloudflare.

Thanks for listening





How long to run things for (0.8)



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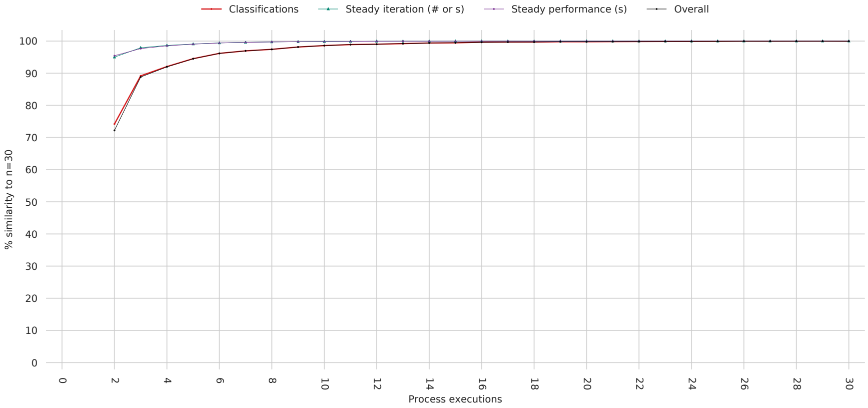

		Class.	Steady iter (#)	Steady iter (s)	Steady perf (s)
binarytrees		≭ (27⎺⎽, 2⎽⎺, 1⁓)			
fannkuch_redux		≭ (26⎺⎽, 4⁓)			
fasta		⎺⎽	4.0 (3.0,34.4)	0.75 (0.535,5.873)	0.16188 ±0.000738
nbody	Graal	⎺⎽	6.0 δ=−2.0 (5.0,7.0)	0.86 δ=−0.352 (0.704,1.090)	0.13677 δ=+0.00343 ±0.000035
richards		⎺⎽	2.0 (2.0,35.3)	0.95 (0.879,9.745)	0.26465 ±0.007761
spectralnorm		⎽⎺	14.0 (2.0,94.6)	13.60 (0.830,98.737)	1.05685 δ=+0.16392 ±0.000126