Python and PyPy performance (not) for dummies

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EuroPython 2015

July 21, 2015

About us

- PyPy core devs
- vmprof, cffi, pdb++, fancycompleter, ...
- Consultants
- http://baroquesoftware.com/

About you

- You are proficient in Python
- Your Python program is slow
- You want to make it fast(er)

Optimization for dummies

- Obligatory citation
 - premature optimization is the root of all evil (D. Knuth)
- Pareto principle, or 80-20 rule
 - 80% of the time will be spent in 20% of the program
 - 20% of 1 mln is 200 000
- Two golden rules:
 - 1. Identify the slow spots
 - 2. Optimize them

This talk

- Two parts
 - 1. How to identify the slow spots
 - 2. How to address the problems

Part 1

- profiling
- tools

What is performance?

- you need something quantifiable by numbers
- usually, time spent doing task X
- sometimes number of requests, latency, etc.
- some statistical properties about that metric (average, minimum, maximum)

Do you have a performance problem?

- define what you're trying to measure
- measure it (production, benchmarks, etc.)
- see if Python is the cause here (if it's not, we can't help you, but I'm sure someone can)
- make sure you can change and test stuff quickly (e.g. benchmarks are better than changing stuff in production)
- same as for debugging

We have a python problem

- tools, timers etc.
- systems are too complicated to guess which will be faster
- find your bottlenecks
- 20/80 (but 20% of million lines is 200 000 lines, remember that)

Profilers landscape

- cProfile, runSnakeRun (high overhead) event based profiler
- plop, vmprof statistical profiler
- cProfile & vmprof work on pypy

vmprof

- inspired by gperftools
- statistical profiler run by an interrupt (~300Hz on modern linux)
- sampling the C stack
- CPython, PyPy, possibly more virtual machines

why not just use gperftools?

- C stack does not contain python-level frames
- 90% PyEval_EvalFrame and other internals
- we want python-level functions
- picture is even more confusing in the presence of the JIT

using vmprof

- demo
- http://vmprof.readthedocs.org

using vmprof in production

- low overhead (5-10%), possibly lower in the future
- possibility of realtime monitoring (coming)

vmprof future

- profiler as a service
- realtime advanced visualization

Part 2

Make it fast

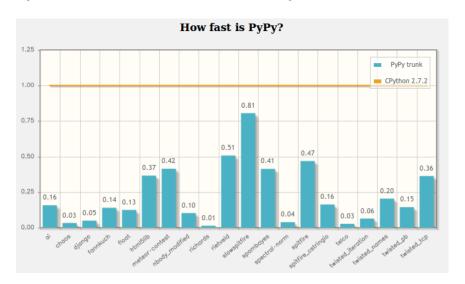
Tools

- Endless list of tools/techniques to increment speed
- C extension
- Cython
- numba
- "performance tricks"
- PyPy
 - We'll concentrate on it
 - WARNING: we wrote it, we are biased:)
 - gives you most wins for free (*)

What is PyPy

- Alternative, fast Python implementation
- Performance: JIT compiler, advanced GC
- PyPy 2.6.0 (Python version 2.7.9)
- Py3k as usual in progress (3.2.5 out, 3.3 in development)
- http://pypy.org
- EP Talks:
 - The GIL is dead: PyPy-STM (July 23, 16:45 by Armin Rigo)
 - PyPy ecosystem: CFFI, numpy, scipy, etc (July 24, 15:15 by Romain Guillebert)

Speed: 7x faster than CPython

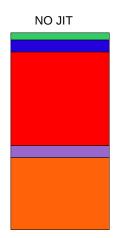


The JIT

```
def main():
  init()
  some_quick_code()
  for x in large_list:
      do_something(x)
  some_other_code()
  while condition():
      expensive_computation()
```

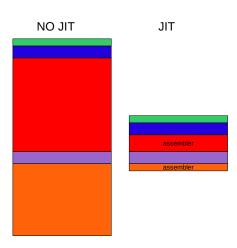
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JIT overview

- Tracing JIT
 - detect and compile "hot" code
- Specialization
- Precompute as much as possible
- Constant propagation
- Aggressive inlining

Specialization (1)

- obj.foo()
- which code is executed? (SIMPLIFIED)
 - lookup foo in obj.__dict__
 - lookup foo in obj.__class___
 - lookup foo in obj.__bases__[0], etc.
 - finally, execute foo
- without JIT, you need to do these steps again and again
- Precompute the lookup?

Specialization (2)

- pretend and assume that obj.__class__ IS constant
 - "promotion"
- guard
 - check our assumption: if it's false, bail out
- now we can directly jump to foo code
 - ...unless foo is in obj.___dict___: GUARD!
 - ...unless foo.__class__._dict__changed:
 GUARD!
- Too many guard failures?
 - Compile some more assembler!
- guards are cheap
 - out-of-line guards even more

Specialization (3)

- who decides what to promote/specialize for?
 - we, the PyPy devs :)
 - heuristics
- instance attributes are never promoted
- class attributes are promoted by default (with some exceptions)
- module attributes (i.e., globals) as well
- bytecode constants

Specialization trade-offs

- Too much specialization
 - guards fails often
 - explosion of assembler
- Not enough specialization
 - inefficient code

Guidos points



Guido van Rossum Shared publicly - Sep 10, 2012

Some patterns for fast Python. Know any others?

- Avoid overengineering datastructures. Tuples are better than objects (try namedtuple too though). Prefer simple fields over getter/setter functions.
- Built-in datatypes are your friends. Use more numbers, strings, tuples, lists, sets, dicts. Also check out the collections library, esp. deque.
- Be suspicious of function/method calls; creating a stack frame is expensive.
- Don't write Java (or C++, or Javascript, ...) in Python.
- Are you sure it's too slow? Profile before optimizing!
- The universal speed-up is rewriting small bits of code in C. Do this only when all else fails.

Don't do it on PyPy (or at all)

- simple is better than complicated
- avoid string concatenation in the loop
- avoid replacing simple loop with itertools monsters
- "move stuff to C" is (almost) never a good idea
- use cffi when calling C
- avoid C extensions using CPython C API
- avoid creating classes at runtime

More about PyPy

- we are going to run a PyPy open space
- come ask more questions

Q&A?

- Thank you!
- http://baroquesoftware.com
- http://pypy.org
- http://vmprof.readthedocs.org