

CPython, Grumpy, PyPy

When, How, Why?





About Me

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CPython # Python

CPython = A Language Engine

Python = Programming Language

Python and CPython

CPython:

- Is used by the majority of developers and users.
- Compiles your python code into bytecode (transparently) and interprets that bytecode in an evaluation loop.
- Performance is mainly limited by the GIL

GIL (Global Interpreter Lock)

- A binary semaphore that prevents multiple native threads from executing Python bytecodes at once.
- Must be held by the current thread before it can safely access Python objects.

GIL # Mutex

GIL (Global Interpreter Lock)

 In order to emulate concurrency of execution, the interpreter regularly tries to switch threads (sys.setswitchinterval).

Python 2.x

GIL (Global Interpreter Lock)

The waiting thread waits until a timeout (By default TIMEOUT is 5 milliseconds, but it can be changed).

Python 3.2<

Simple Solution



<u>MultiProcessing</u>

<u>PROS</u>

- Code is usually straightforward
- Takes advantage of multiple CPUs & cores
- Avoids GIL limitations for CPython (basically each process has it's own GIL)
- Child processes are interruptible/killable

CONS

- Separate memory space
- Costly to run a huge number of processes
- Costly to create and destroy processes frequently
- IPC a little more complicated with more overhead
- Larger memory footprint

MultiThreading

PROS

- Shared memory
- Low memory footprint

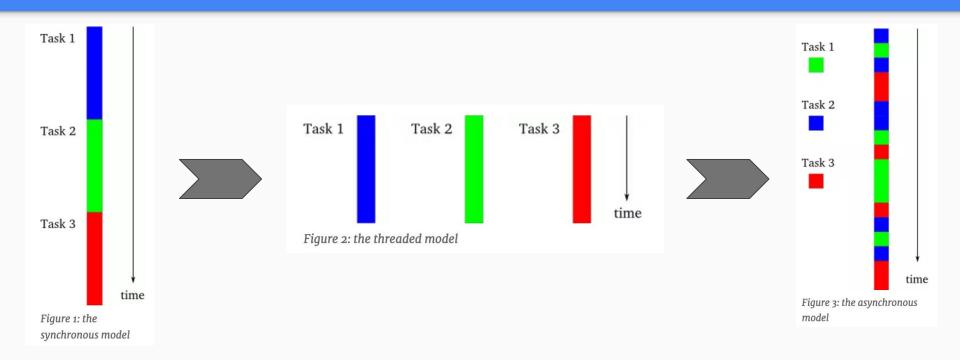
<u>CONS</u>

- CPython subject to the GIL
- Code is usually harder to understand and to get right - the potential for race conditions increases dramatically

Not So Simple Solution



Twisted/async.io (Asynchronous Programming/Event driven programing)



When Should I Use CPython

- GUI applications
- Network servers

Demo

GIL



PyPy

- Supports all of the core language (passing Python test suite)
- Supports most of the commonly used Python standard library modules (just check if it imports. If it imports, it should work):
 - __builtin___, __pypy___, _ast, _cffi_backend, _codecs, _collections, _continuation, _csv, _file, _hashlib, _io, _locale, _lsprof, _md5, _minimal_curses, _multibytecodec, _multiprocessing, _numpypy, _pickle_support, _pypyjson, _random, _rawffi, _sha, _socket, _sre, _ssl, _struct, _testing, _warnings, _weakref, array, binascii, bz2, cStringIO, cmath, cppyy, cpyext, crypt, errno, exceptions, fcntl, gc, imp, itertools, marshal, math, mmap, operator, parser, posix, pwd, pyexpat, pypyjit, select, signal, symbol, sys, termios, thread, time, token, unicodedata, zipimport, zlib
- Built using the RPython language that was co-developed with it:
 - A framework for producing implementations of dynamic languages (like python)
 - Able to automatically generate a Just-in-Time compiler for any dynamic language
- Implements a JIT compiler in Python.
- Support for Stackless and Lightweight concurrent programming (Greenlets).

Why did the STM effort not work out?

STM was a research project that proved that the idea is possible. However, the amount of user effort that is required to make programs run in a parallelizable way is significant, and we never managed to develop tools that would help in doing so. At the moment we're not sure if more work spent on tooling would improve the situation or if the whole idea is really doomed. The approach also ended up adding significant overhead on single threaded programs, so in the end it is very easy to make your programs slower. (We have some money left in the donation pot for STM which we are not using; according to the rules, we could declare the STM attempt failed and channel that money towards the present GIL removal proposal.)

Software Transactional Memory

```
def f(list1, list2):
 while True:
     t = transaction()
     x = list1.pop(t)
     list2.append(t, x)
     if t.commit():
         break
```

JIT(Just In Time)

- Done during execution (runtime).
- Continuously analyses the code being executed and identifies parts of the code where the speedup gained from compilation would outweigh the overhead of compiling that code
- In theory JIT compilation can yield faster execution than static compilation.

Demo

Performance + How to use PyPy



When Should I Use PyPy

- Long running processes
- When running mostly Python code.

Grumpy

- Google has YouTube which is:
 - Written In CPython (2.7)
 - Serves millions of requests per second
- Google decides to Implement an alternative runtime optimized for real-time serving.
- Google decides to use Go.
- Grumpy was born!



Grumpy

- Python to Go source code transcompiler and runtime.
- Grumpy has no VM.
- Has no GIL.
- Uses Go's garbage collection instead of counting references.
- Uses Go's goroutines
- No support for C extension modules.



Grumpy Features

Missing features Language features ☑ decorators pow operator relative imports ☑ importing module members ■ wildcard import **Builtins Types** buffer ■ BufferError bytearray ■ FloatingPointError ■ GeneratorExit ■ KeyboardInterrupt memoryview ■ UnicodeTranslateError

[™] Functions apply() coerce() filter() format() globals() input() intern() locals() map() pow() reduce() reload() reversed() round() ✓ sorted() vars()

Demo

Performance + How to use Grumpy



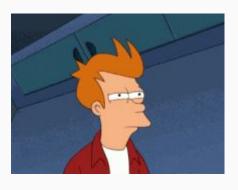
So which one has the best performance???



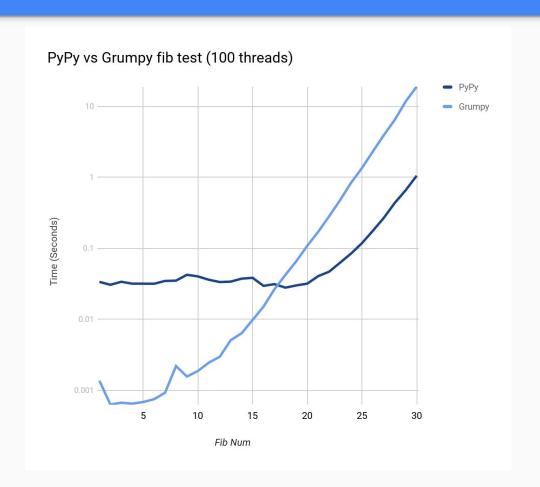
Dict Operations

	руру 2.7	руру 3.6	python 2.7	python 3.6	grumpy	Max	Min
BenchmarkDictCreate	1450018991	1146059661	9150119.157	9046784.877	1695006.364	руру 2.7	grumpy
BenchmarkDictCreateFunc	1503687375	4329709.169	3932658.568	3374312.332	1020829.856	руру 2.7	grumpy
BenchmarkDictGetItem	140226856.3	124192851.8	31829670.34	21461999.59	10997887.58	руру 2.7	grumpy
BenchmarkDictSetItem	95185928.29	99022084.33	24672325.76	19936729.46	5331109.432	руру 3.6	grumpy
Benchmark Dict String Only Get Item	126787806.7	129258907.5	33041546.44	24173192.88	11138902.38	руру 3.6	grumpy
BenchmarkDictStringOnlySetItem	90706527.51	91076428.76	26744788.29	20688567.85	5621035.744	руру 3.6	grumpy
BenchmarkHashStrCached	1146345922	1119756541	30558516.86	14645639.98	11977268.42	руру 2.7	grumpy

Do Some Thinking



It All Comes Down To This



When Should I Use Grumpy

- When you're using a lot of threads, each executing a very small task.
- When you need to use any of the go libraries.

Basic Comparison

CPython

- The reference implementation of Python
- Written in C
- compiles Python
 code to
 intermediate
 bytecode which is
 then interpreted by
 a virtual machine

PyPy

- Supports all of the core language (passing Python test suite)
- supports most of the commonly used Python standard library modules
- Implements a JIT compiler in Python

Grumpy

- Python to Go source code transcompiler.
- Grumpy has no VM.
- Uses Go's garbage collection instead of counting references.

Performance Comparison

GIL

- A binary semaphore that prevents multiple native threads from executing Python bytecodes at once.
- Must be held by the current thread before it can safely access Python objects

JIT

- Done during execution (runtime).
- In theory JIT compilation can yield faster execution than static compilation.

Go

- Has no GIL.
- Statically compiled.
- Uses Go's garbage collection instead of counting references.
- Uses Go's goroutines

Thanks

- David Beazley (https://www.youtube.com/watch?v=ph374fJqFPE
- Dave Peticolas http://krondo.com/an-introduction-to-asynchronous-programming-and-twisted/
- Emmanuel Klinger http://emmanuel-klinger.net/python-100-times-faster-than-grumpy.html
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- Stackoverflow
 - https://stackoverflow.com/questions/18946662/why-shouldnt-i-use-pypy-over-cpython-if-pypy-is-6-3-time
 s-faster
 - https://stackoverflow.com/questions/3044580/multiprocessing-vs-threading-python
 - https://stackoverflow.com/questions/62814/difference-between-binary-semaphore-and-mutex

More Questions

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