Python performance Past, Present, Future Victor Stinner





Python Implementations

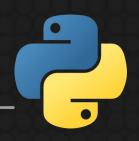


- 1989: CPython (Guido van Rossum)
- 1997: Jython (Jim Hugunin)
- 1998: Stackless Python (Christian Tismer)
- 2006: IronPython (Jim Hugunin)
- 2014: MicroPython (Damien George)





Faster Pythons



- 2002-2012: psyco (Armin Rigo)
- 2007: PyPy
- 2009-2010: Unladen Swallow (Google)
- 2014-2017: Pyston (Dropbox)
- 2016-2017: Pyjion (Microsoft)
 Notice end date for most of these projects...





Two approaches



- Fork CPython
- Implementation from scratch





Fork CPython

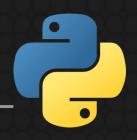


- Unladen Swallow, Pyston, Pyjion, ...
- Performance limited by old CPython design (1989)
- Specific memory allocators, C structures, reference counting, specific GC, ...
- CPython is limited to 1 thread because of the GIL (more later)





From scratch



- PyPy, Jython, IronPython, ...
- Jython and IronPython have no GIL!
- PyPy uses an efficient tracing garbage collector
- C extensions: no support, or slow
- cpyext creates CPython PyObject on demand, sync with PyPy objects





Competition with CPython

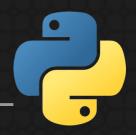


- CPython has around 30 active core developers to maintain it
- New features first land into CPython
- Why would users prefer an outdated and incompatible implementation?
- Who will sponsor the development?





Unladen Swallow (2011)



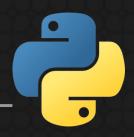
- "Most Python code at Google isn't performance critical"
- "deployment was too difficult: being a replacement was not enough."
- "Our potential customers eventually found other ways of solving their performance problems."

http://qinsb.blogspot.com/2011/03/unladen-swallow-retrospective.html





Pyston & Dropbox (2017)



- "Dropbox has increasingly been writing its performance-sensitive code in other languages, such as Go"
- "We spent much more time than we expected on compatibility"

https://blog.pyston.org/2017/01/31/ pyston-0-6-1-released-and-future-plans/





Summary



- CPython remains the reference implementation but shows its age
- Multiple optimization projects failed
- PyPy: drop-in replacement, 4.4x faster, but not widely adopted yet: why?







Optimize your code



- Let's say that you identified your code causing the performance bottleneck
- How to make your code faster?





PyPy just works!



- Drop-in replacement for CPython!
- 4.4x faster than CPython in average (exact speedup depends on your workload)
- Fully compatible with CPython



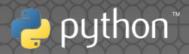


PyPy issues



- Support C extension with cpyext: heavily optimized in 2018, but still slower than CPython (more about the C API later...)
- Larger memory footprint (JIT)
- Longer startup time (JIT)





C and Rust extensions



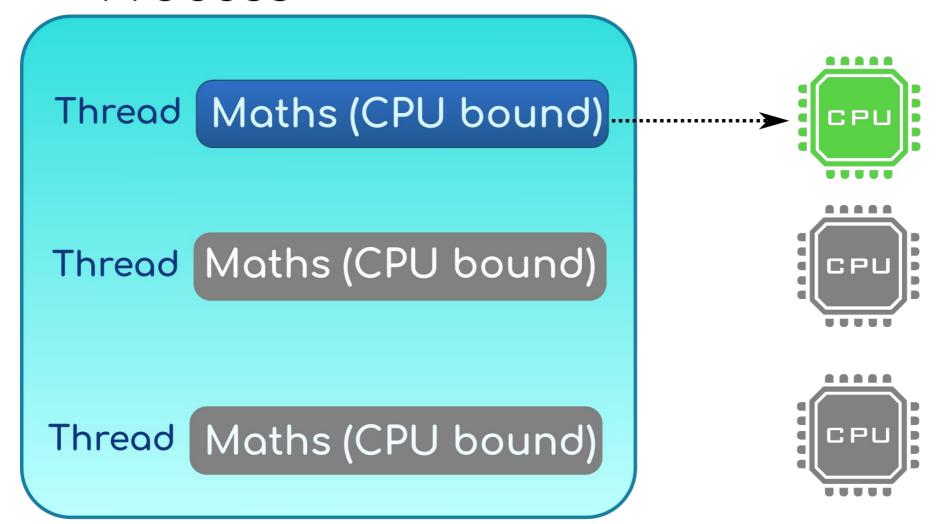
- Identify the performance bottleneck
- Rewrite a small part of your application as a C or Rust extension
- github.com/dgrunwald/rust-cpython
- github.com/Py03/pyo3





CPython GIL

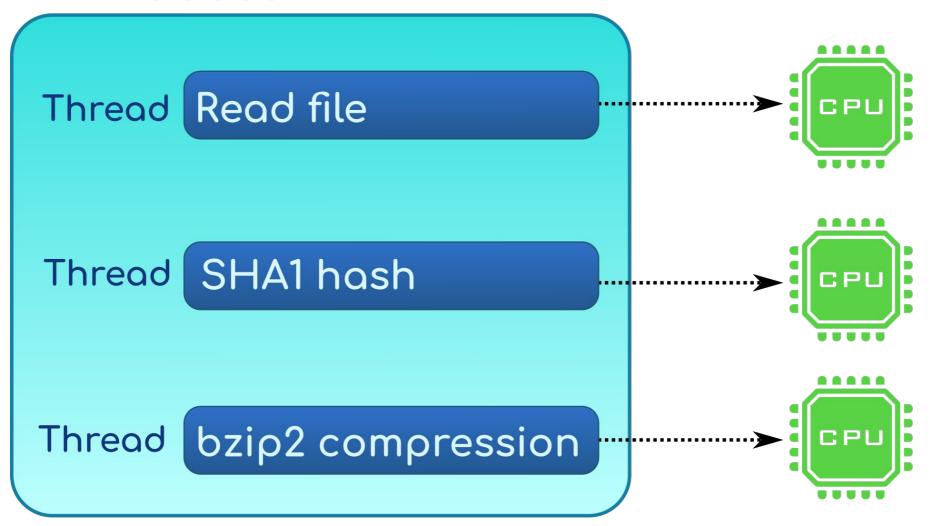
Process



Efficiency: 1/3

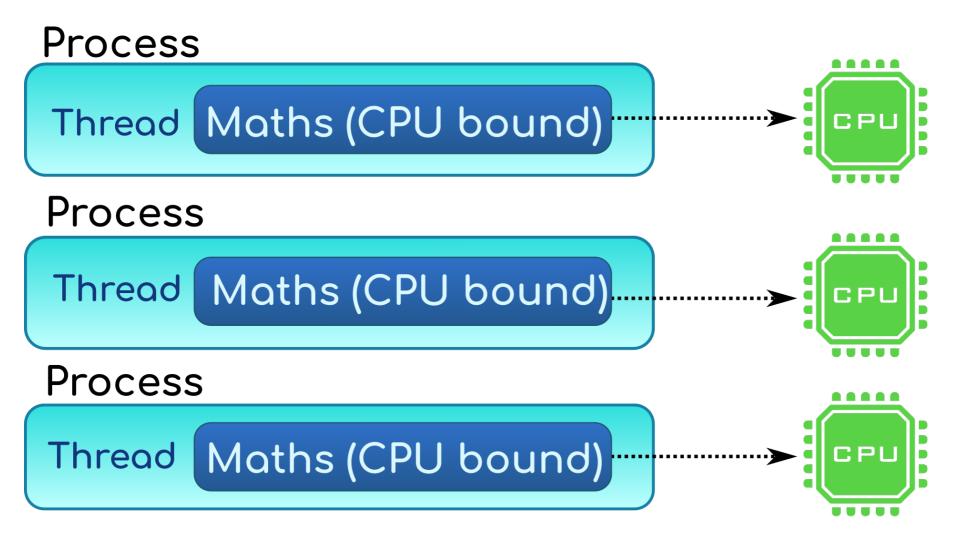
CPython: release the GIL

Process



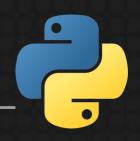
Efficiency: 100%

CPython: multiprocessing



Efficiency: 100%

Multiprocessing



- Work around the GIL limitation
- Shared memory (Python 3.8) avoids memory copies between workers
- New pickle version 5 (Python 3.8) avoids copying large objects: PEP 574





Cython

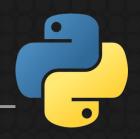


- Easy way to write C extension
- Syntax similar to Python
- Support multiple Python versions
- Handle reference counting for you
- The optimizer emits efficient code using CPython internals for you





Numba

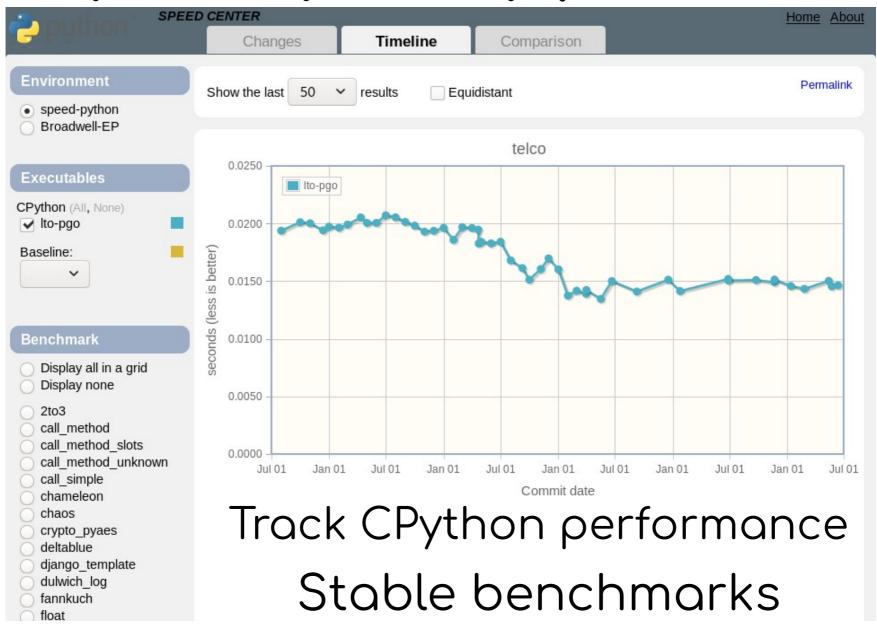


- JIT compiler translating subset of Python and NumPy into fast code
- Simplified Threading: release GIL
- SIMD Vectorization: SSE, AVX, AVX-512
- GPU Acceleration: NVIDIA's CUDA and AMD's ROCm





https://speed.python.org



Summary



- PyPy doesn't require any code change
- C and Rust extensions
- Multiprocessing scales
- Use Cython, don't use the C API directly
- Numba makes numpy faster







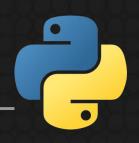
Python C API







Python C API



- Evolved organically: internal functions are exported by mistake
- First written to be consumed by CPython itself
- No design: expose everything
- It exposes too many implementation details





Python 3.8 changes



Python 3.8 now has 3 levels of C API:

- Include/: public "stable" C API
- Include/cpython/: C API specific to CPython
- Include/internal/: internal C API
 Many private functions ("_Py...") and PyInterpreterState structure moved to the internal API.





Stable ABI



- Support multiple Python versions: Python 3.8, 3.9, ...
- CPython 3.8 debug build is ABI compatible with the release build
- It can use C extensions compiled in release mode
- It has much more sanity checks at runtime to detect bugs





Specialized lists



- CPython list: array of pointers Py0bject*
- PyPy specialized list: list of integers int64_t array[n]
- Can it be implemented in CPython?
- Can we modify PyListObject?





Accessing structs



- Problem 1: PyList_GET_ITEM() macro access directly
 PyListObject.ob_item[index] (PyObject*)
- C extensions must not access
 PyListObject fields directly
- PyList_GET_ITEM() macro could be modified to convert int64_t to
 PyObject*, but...





Borrowed reference



- Problem 2: PyList_GET_ITEM() returns a borrowed reference, Py_DECREF() must not be called
- If PyList_GET_ITEM() would create a temporary object, when should it be destroyed? We don't know...
- Many C functions return borrowed references





Better C API



- Make structures opaque. Such code must fail with a compiler error:
 - PyObject *obj = PyLong_FromLong(1); PyTypeObject *type = obj->ob_type;
- Remove functions using borrowed references or "stealing" references
- Replace macros with function calls





Break compatibility?



- Any C API change can break an unknown number of projects
- Maybe not all C API design issues can be fixed
- Updating all C extensions on PyPI will take a lot of time
 - ... there should be another way...





New HPy C API

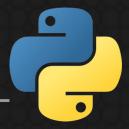


- New C API: correct since day 1
- PyHandle: opaque integer
- Similar to an Unix file descriptor or Windows HANDLE: open(), dup(), close()





New PyHandle C API



- CPython: Implemented on top of the current C API
- PyPy: More efficient than the current C API
- Cython: no need to change your code, Cython will generate code using PyHandle for you





Reference counting







Gilectomy

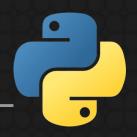


- "Remove the GIL": replace unique GIL with one lock per mutable object
- Atomic increment/decrement
- Log INCREF/DECREF
- Reference counting doesn't scale with the number of threads





Tracing GC for CPython



- Many modern language implementations use tracing GC
- PyPy has a tracing GC
- Existing C API would continue to use reference counting





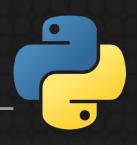
Subinterpreters







Subinterpreters



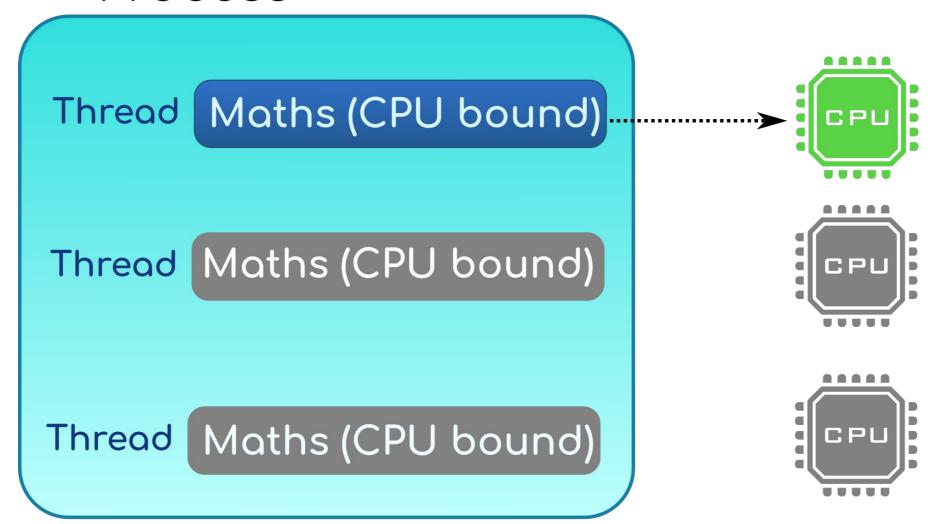
- Eric Snow's PEP 554 "Multiple Interpreters in the Stdlib"
- Replace the unique Global Interpreter Lock (GIL) with one lock per interpreter
- Work-in-progress refactoring of CPython complex internals





CPython GIL

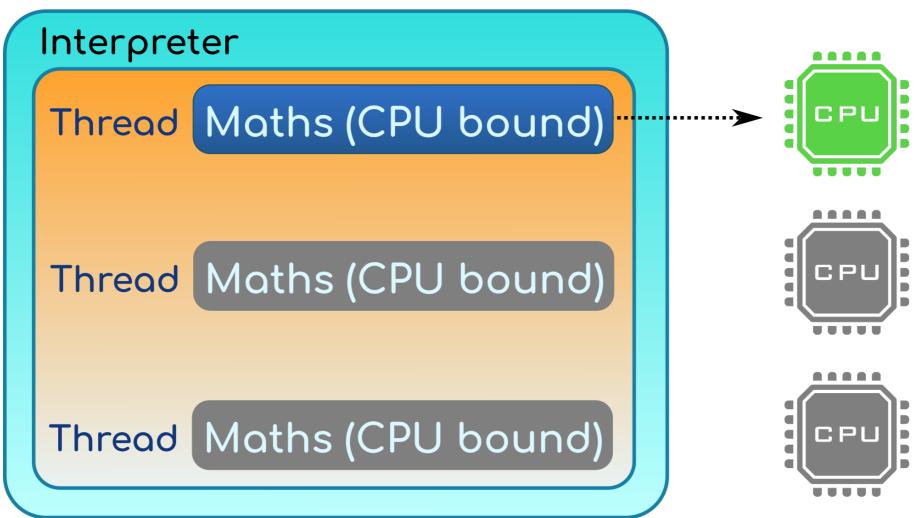
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Efficiency: 1/3

CPython GIL

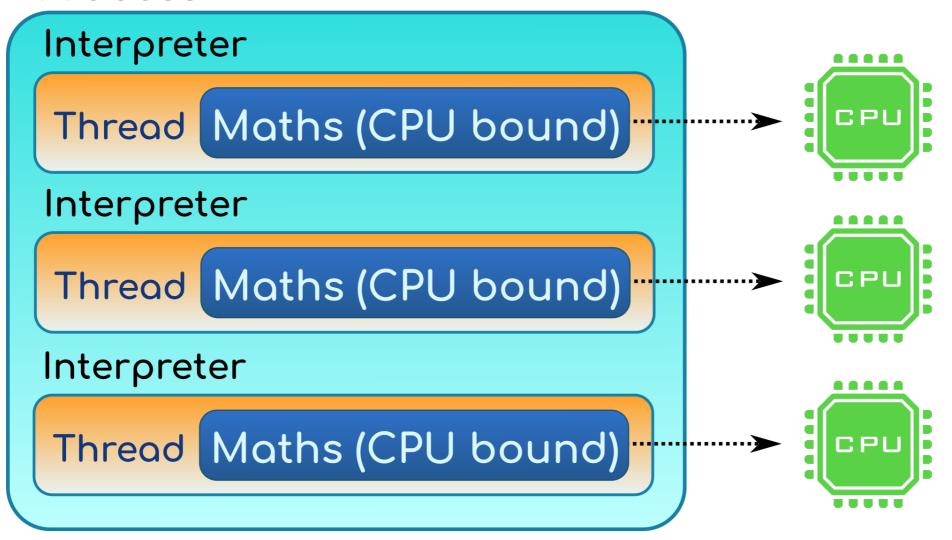
Process



Efficiency: 1/3

CPython subinterpreters

Process



Efficiency: 100%

Subinterpreters



Expectation (should be verified with a working implementation):

- Lower memory footprint: share more memory
- Faster locks (no syscall?)
- Limitation: Python objects cannot be shared between interpreters





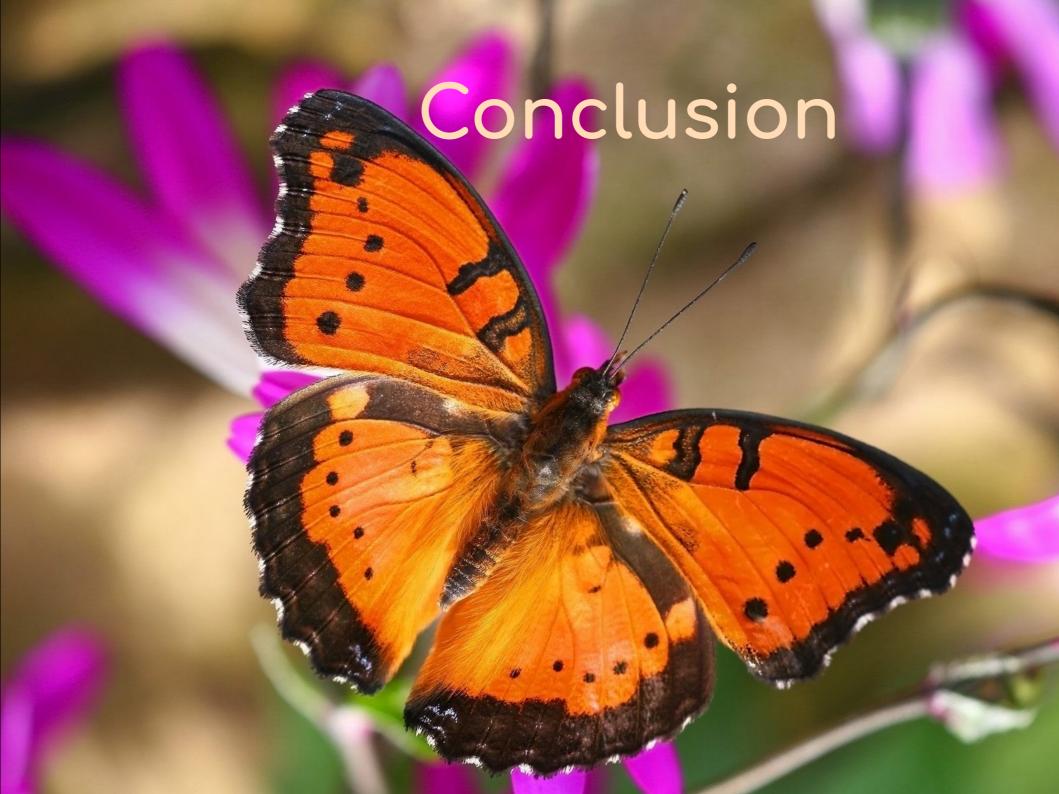
Summary



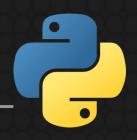
- Current C API has design issues
- New "PyHandle" C API
- Tracing garbage collector for CPython
- CPython subinterpreters







Conclusion



- Many previous optimizations projects failed
- Cython, multiprocessing and Numba are working well to make your code faster
- HPy, tracing GC and subinterpreters are very promising!





Questions?



- http://pypy.org/
- https://faster-cpython.readthedocs.io/
- https://pythoncapi.readthedocs.io/
- https://speed.python.org/
- https://mail.python.org/mailman3/lists/ speed.python.org/
- https://github.com/hpyproject/hpy



