# The speed of PyPy

Maciej Fijałkowski

merlinux GmbH

RuPy, November 7th 2009, Poznań

# Story about Python's speed



Python is slow

Python is slow



Is Python really slow?

- Is Python really slow?
- Sometimes, for some usecases

- Is Python really slow?
- Sometimes, for some usecases
- Let's have a look at some examples

#### Nomenclature

- Python a programming language
- CPython main implementation of Python
- JVM Java Virtual Machine VM used to run Java, among others
- JIT Just in time compiler
- Psyco JIT for Python

Float example, stolen from factor blog

	CPython	Java (hotspot client mode)
Average of 10 runs:	7.6s	0.77s

Float example, stolen from factor blog

	CPython	Java (hotspot client mode)
Average of 10 runs:	7.6s	0.77s

Python is 10x slower than Java

Float example, stolen from factor blog

	CPython	Java (hotspot client mode)
Average of 10 runs:	7.6s	0.77s

- Python is 10x slower than Java
- Python is 10x slower than Java on this particular benchmark

Float example, stolen from factor blog

	CPython	Java (hotspot client mode)
Average of 10 runs:	7.6s	0.77s

- Python is 10x slower than Java
- Python is 10x slower than Java on this particular benchmark
- CPython is 10x slower than Java on this particular benchmark

# More about this example

	CPython	JVM	Psyco	PyPy
Average of 10 runs	7.6s	0.77s	4.4s	1.3s

# More about this example

	CPython	JVM	Psyco	PyPy
Average of 10 runs	7.6s	0.77s	4.4s	1.3s

• So, it's CPython that is slow on this particular benchmark

# More about this example

	CPython	JVM	Psyco	PyPy
Average of 10 runs	7.6s	0.77s	4.4s	1.3s

- So, it's CPython that is slow on this particular benchmark
- Same example, using numpy and vectorization about 3x faster than JVM

### Python's speed

• Instead of: "Why is Python slow?"

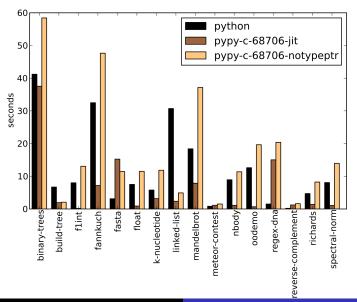
### Python's speed

- Instead of: "Why is Python slow?"
- Better: "Why is Python hard to optimize?"

### Python's speed

- Instead of: "Why is Python slow?"
- Better: "Why is Python hard to optimize?"
- Even better: "How are we going to fix it?"

#### Some evidence





# Why is Python hard to optimize?

- Duck typing (dynamic dispatch)
- Frames
- Object encapsulation
- Dictionaries of instances
- Changing globals
- Ability to dynamically change builtins

# **Duck typing**

- Dispatching over item type
- ullet Needs to check what the type of x and y is

#### **Frames**

- Python interpreters use frames on heap (instead of stack)
- Locals are stored on those frames
- Intermediate results are store on valuestack of frames
- In fact, you can access frames via sys.\_getframe() or traceback

#### Object encapsulation

- Also called boxing
- Each object, even int, has to be boxed
- Requires allocations and indirection

### Example - addition

- read value for x from frame, store on valuestack
- read value for x from frame, store on valuestack
- allocate new integer
- read two values from valuestack, add and store the result in freshly allocated integer
- move the result from valuestack to locals



### Example - addition

- read value for x from frame, store on valuestack
- read value for x from frame, store on valuestack
- allocate new integer
- read two values from valuestack, add and store the result in freshly allocated integer
- move the result from valuestack to locals
- in fact, should be one assembler instruction



#### Dictionaries of instances

- Need to perform a dictionary lookup for x.y
- There are three lookups per method call (descriptor, object, type)
- Two for attribute access (descriptor, object)
- Looks like list lookup should be enough

# Changing globals

- Global symbols can change at any moment in time
- Makes inlining hard
- Requires global dict lookups, even for constants

# Ability to dynamically change builtins

- You can say int = my\_function
- But you can't int.\_\_add\_\_ = my\_method
- Still messes up optimizations
- Global lookup is also a dictionary lookup, even if globals don't change

#### Interpreting vs compiling

- Apparently, processors are good at branch prediction
- We didn't measure much of a difference, less than 2x overall

### CPython specific problems

- In general, CPython is fairly well optimized
- refcounting is an inefficient garbage collection scheme
- GIL

#### Dynamic compilation to the rescue

- You don't pay for feature, until you actually use it
- In static compilation, compiler has to prove that bad things can't happen
- Impossible in Python
- With dynamic compilation, you just throw away compiled code in case things go wrong or start over

Allocate frame

- Allocate frame
- Use C stack, but remember where frame fields are living on the stack

- Allocate frame
- Use C stack, but remember where frame fields are living on the stack
- Be able to reconstruct frame on demand (for example sys.\_getframe() was called)

- Allocate frame
- Use C stack, but remember where frame fields are living on the stack
- Be able to reconstruct frame on demand (for example sys.\_getframe() was called)
- The effect is that you don't pay for frames, unless you really use them

### Dealing with dynamic dispatch

- The answer is to simply specialize over types
- Provides possibly multiple versions of compiled code for single Python code

# Dealing with object encapsulation

- "Virtual objects"
- Also known as escape analysis
- If object does not "escape", don't allocate it at all

# How does it work in practice?

## How does it work in practice?

Pretty well

# How does it work in practice?

- Pretty well
- A reasonable speedup over CPython

## Dealing with attribute access

- Fairly complex task
- Sharing dict, more or less the same effect as V8's hidden classes

# Dealing with attribute access

- Fairly complex task
- Sharing dict, more or less the same effect as V8's hidden classes
- Python is a very complex language
- Shadowing methods with attributes
- Descriptors before attributes

# Other important optimizations

- Caching globals
- Caching builtins
- A lot of smaller ones

### Bird's view of JIT

- Mixed mode interpreter & JIT for hot paths
- Tracing JIT (like TraceMonkey), not up-front

•

• I use CPython for everyday usage

- I use CPython for everyday usage
- But personally, I hope to change it in next months

- I use CPython for everyday usage
- But personally, I hope to change it in next months
- ... in places where performance matters, but that don't depend on third party C modules (like numpy)

- I use CPython for everyday usage
- But personally, I hope to change it in next months
- ... in places where performance matters, but that don't depend on third party C modules (like numpy)
- ... like building and developing PyPy

## Status of PyPy

- Very compliant Python interpreter
- Most of important stdlib modules
- Differencies are agreed to be implementation details

# Status of PyPy

- Very compliant Python interpreter
- Most of important stdlib modules
- Differencies are agreed to be implementation details
- or bugs

# Examples of working programs

- Django (sqlite only)
- Twisted
- PyPy's translation toolchain
- ctypes

#### Status of JIT

- Because the way it's constructed, handles all Python language features (unlike for example Psyco)
- Changes very quickly these days
- Ready for cautious tests
- Not ready as a drop-in replacement of CPython

#### Status of JIT

- Because the way it's constructed, handles all Python language features (unlike for example Psyco)
- Changes very quickly these days
- Ready for cautious tests
- Not ready as a drop-in replacement of CPython
- yet!

- A fact: People rely on deep obscure features of language
- Examples:

- A fact: People rely on deep obscure features of language
- Examples:

```
except ImportError, e:
 if str(e) != ...: raise
```

- A fact: People rely on deep obscure features of language
- Examples:
  - except ImportError, e:
    if str(e) != ...: raise
  - Exact naming of list comprehension variable

- A fact: People rely on deep obscure features of language
- Examples:

```
except ImportError, e:
 if str(e) != ...: raise
```

- Exact naming of list comprehension variable
- Reliance on reference counting

### Profit tomorrow!

 We plan to release JIT-ready version somewhere early 2010

### Profit tomorrow!

- We plan to release JIT-ready version somewhere early 2010
- It should be able to speed up real-world programs

## How you can help

- It's all open source after all ...
- Try running existing programs
- Profile, report bugs

## How you can help

- It's all open source after all ...
- Try running existing programs
- Profile, report bugs
- Talk to your boss

## Thank you!

- This talk is already online (with all examples):
   http://codespeak.net/svn/pypy/dist/extradoc/talk
- http://morepypy.blogspot.com
- #pypy on freenode
- If you want to know more about PyPy, feel free to bug me around (like, how does the JIT work?)
- Any questions?