#### PyPy Intro and JIT Frontend

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#### About this talk

- What is PyPy? What is RPython?
- Tracing JIT 101
- PyPy JIT frontend and optimizer
  - "how we manage to make things fast"

#### Part 1

**PyPy introduction** 

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## What is PyPy?

• For most people, the final product:

```
pypy
Python 2.7.10 (173add34cdd2, Mar 15 2016, 23:00:19)
[PyPy 5.1.0-alpha0 with GCC 4.8.4] on linux2
>>>> import test.pystone
>>> test.pystone.main()
Pystone(1.1) time for 50000 passes = 0.0473992
This machine benchmarks at 1.05487e+06 pystones/second
```

 More in general: a broader project, ecosystem and community

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## PyPy as a project

- rpython: a fancy compiler
  - source code: "statically typed Python with type inference and metaprogramming"
  - fancy features: C-like performance, GC, meta-JIT
  - "like GCC" (it statically produces a binary)
  - you can run RPython programs on top of CPython (veeery slow, for development only)
- pypy: a Python interpreter
  - ▶ "like CPython", but written in RPython
  - ► CPython : GCC = PyPy : RPythor

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  - CPython : GCC = PyPy : RPython

### Important fact

- We did not write a JIT compiler for Python
- The "meta JIT" works with all RPython programs
- The "Python JIT" is automatically generated from the interpreter
- Writing an interpreter is vastly easier than a compiler

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Other interpreters: smalltalk, prolog, ruby, php, ...

#### The final product

- rpython + pypy: the final binary you download and execute
  - a Python interpreter
  - with a GC
  - with a JIT
  - fast

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**Overview of tracing JITs** 

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#### **Assumptions**

- Pareto Principle (80-20 rule)
  - the 20% of the program accounts for the 80% of the runtime
  - hot-spots
- Fast Path principle
  - optimize only what is necessary
  - fall back for uncommon cases
- Most of runtime spent in loops
- Always the same code paths (likely)

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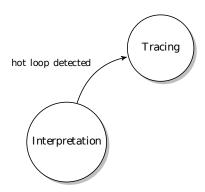
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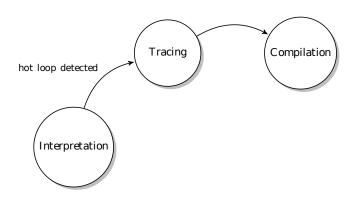
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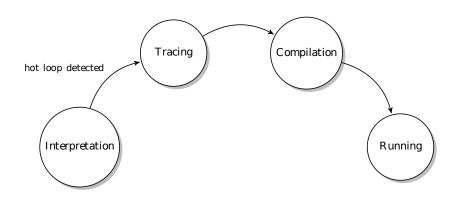
## Tracing JIT

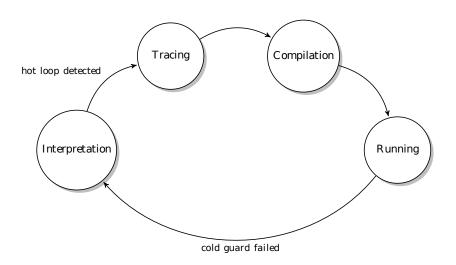
- Interpret the program as usual
- Detect hot loops
- Tracing phase
  - linear trace
- Compiling
- Execute
  - guards to ensure correctness
- Profit :-)

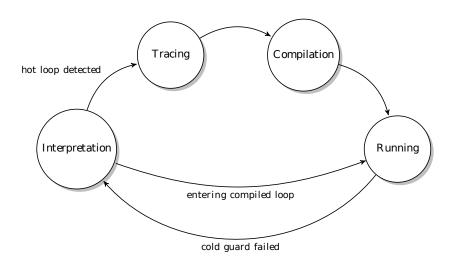


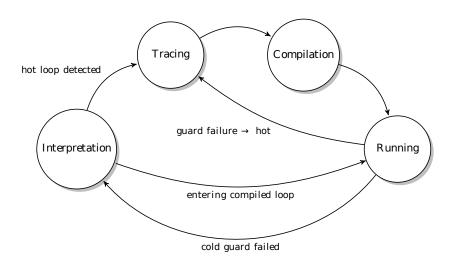


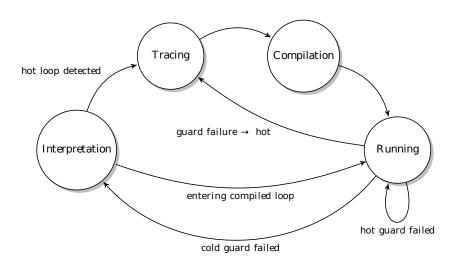












#### Trace trees

**WRITE ME** 

The PyPy JIT

## Terminology (1)

- translation time: when you run "rpython targetpypy.py" to get the pypy binary
- runtime: everything which happens after you start pypy
- interpretation, tracing, compiling
- assembler/machine code: the output of the JIT compiler
- execution time: when your Python program is being executed
  - by the interpreter
  - by the machine code

## Terminology (2)

- interp-level: things written in RPython
- [PyPy] interpreter: the RPython program which executes the final Python programs
- bytecode: "the output of dis.dis". It is executed by the PyPy interpreter.
- app-level: things written in Python, and executed by the PyPy Interpreter

## Terminology (3)

- (the following is not 100% accurate but it's enough to understand the general principle)
- low level op or ResOperation
  - low-level instructions like "add two integers", "read a field out of a struct", "call this function"
  - (more or less) the same level of C ("portable assembler")
  - knows about GC objects (e.g. you have getfield\_gc vs getfield\_raw)
- jitcodes: low-level representation of RPython functions
  - sequence of low level ops
  - generated at translation time
  - 1 RPython function --> 1 C function --> 1 jitcode

## Terminology (4)

- JIT traces or loops
  - a very specific sequence of llops as actually executed by your Python program
  - generated at runtime (more specifically, during tracing)
- JIT optimizer: takes JIT traces and emits JIT traces
- JIT backend: takes JIT traces and emits machine code

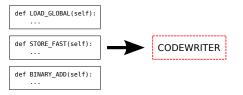
#### **RPYTHON**

```
def LOAD_GLOBAL(self):
...

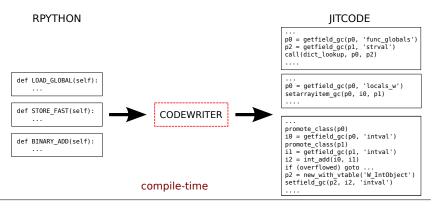
def STORE_FAST(self):
...

def BINARY_ADD(self):
...
```

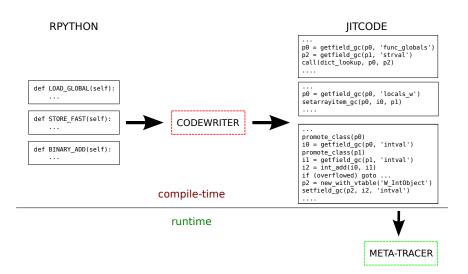
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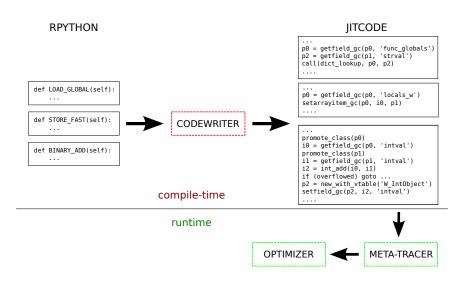


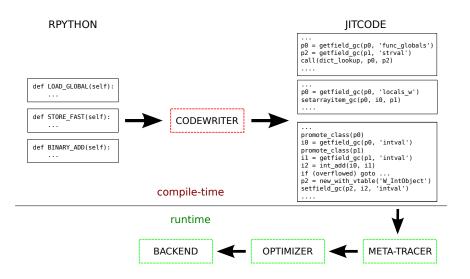
#### RPYTHON JITCODE p0 = getfield\_gc(p0, 'func\_globals') p2 = getfield\_gc(p1, 'strval') call(dict lookup, p0, p2) def LOAD GLOBAL(self): p0 = getfield\_gc(p0, 'locals\_w') setarravitem qc(p0, i0, p1) def STORE FAST(self): CODEWRITER promote class(p0) i0 = getfield gc(p0, 'intval') def BINARY ADD(self): promote class(p1) i1 = getfield gc(pl, 'intval') i2 = int add(i0, i1)if (overflowed) goto ... p2 = new with vtable('W IntObject') setfield qc(p2, i2, 'intval')

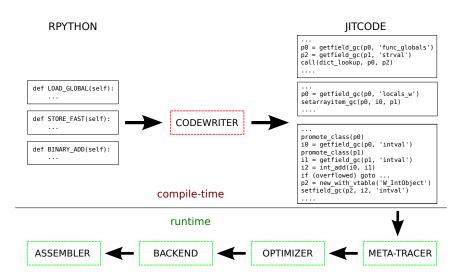


runtime









## PyPy trace example

```
def fn():
    c = a+b
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LOAD_GLOBAL A
LOAD_GLOBAL B
BINARY_ADD
STORE_FAST C
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p2 = getfield gc(p1, 'strval')
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guard class(p0, W IntObject)
i0 = getfield gc(p0, 'intval')
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. . .
p\theta = \text{getfield gc}(p\theta, 'locals w')
setarrayitem gc(p0, i0, p1)
. . . .
```

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#### PyPy optimizer

- intbounds
- constant folding / pure operations
- virtuals
- string optimizations
- heap (multiple get/setfield, etc)
- unroll

#### Intbound optimization (1)

```
intbound.py
def fn():
    i = 0
    while i < 5000:
        i += 2
    return i</pre>
```

#### Intbound optimization (2)

```
unoptimized
...
i17 = int_lt(i15, 5000)
guard_true(i17)
i19 = int_add_ovf(i15, 2)
guard_no_overflow()
...
```

```
optimized
...
i17 = int_lt(i15, 5000)
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- It works often
- array bound checking
- intbound info propagates all over the trace

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```

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## Virtuals (1)

```
virtuals.py
def fn():
    i = 0
    while i < 5000:
        i += 2
    return i</pre>
```

## Virtuals (2)

# unoptimized ... guard\_class(p0, W\_IntObject) i1 = getfield\_pure(p0, 'intval') i2 = int\_add(i1, 2) p3 = new(W\_IntObject) setfield\_gc(p3, i2, 'intval')

```
optimized
...
i2 = int_add(i1, 2)
...
```

- The most important optimization (TM)
- It works both inside the trace and across the loop
- It works for tons of cases
  - e.g. function frames

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#### Constant folding (1)

```
constfold.py
def fn():
    i = 0
    while i < 5000:
        i += 2
    return i</pre>
```

## Constant folding (2)

```
optimized
...
i1 = getfield_pure(p0, 'intval')
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```

- It "finishes the job"
- Works well together with other optimizations (e.g. virtuals)
- It also does "normal, boring, static" constant-folding

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## Out of line guards (1)

```
outoflineguards.py
N = 2
def fn():
    i = 0
    while i < 5000:
        i += N
    return i</pre>
```

#### Out of line guards (2)

```
unoptimized
...
quasiimmut_field(<Cell>, 'val')
guard_not_invalidated()
p0 = getfield_gc(<Cell>, 'val')
...
i2 = getfield_pure(p0, 'intval')
i3 = int_add(i1, i2)
```

```
optimized
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- Python is too dynamic, but we don't care :-)
- No overhead in assembler code
- Used a bit "everywhere"

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#### Guards

- guard\_true
- guard\_false
- guard\_class
- guard\_no\_overflow
- guard\_value

#### **Promotion**

- guard\_value
- specialize code
- make sure not to overspecialize
- example: type of objects
- example: function code objects, ...

#### Conclusion

- PyPy is cool :-)
- Any question?