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

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

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 : RPython

PyPy as a project

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- pypy: a Python interpreter
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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
- 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

Overview of tracing JITs

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)

Assumptions

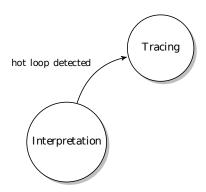
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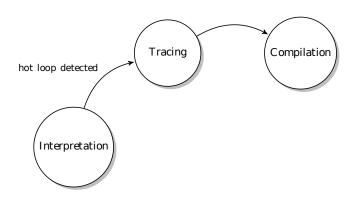
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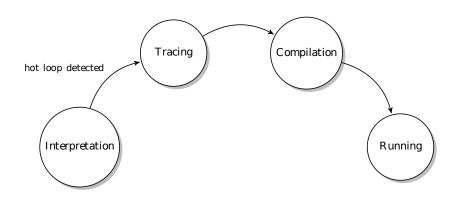
Tracing JIT

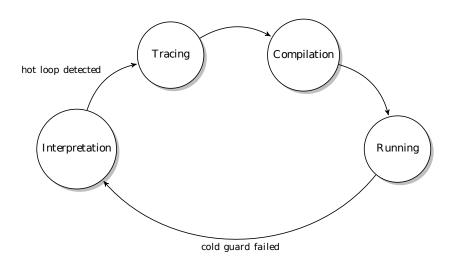
- Interpret the program as usual
- Detect hot loops
- Tracing phase
 - linear trace
- Compiling
- Execute
 - guards to ensure correctness
- Profit :-)

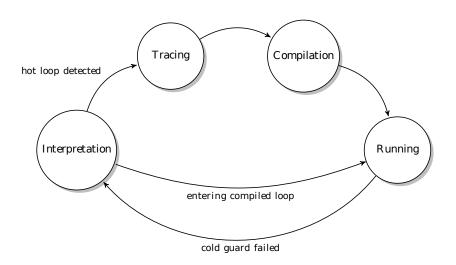


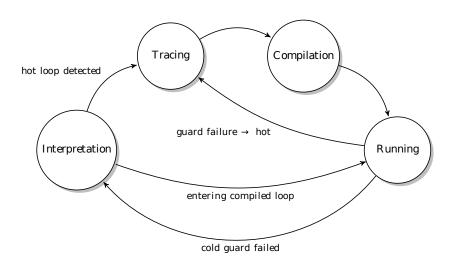


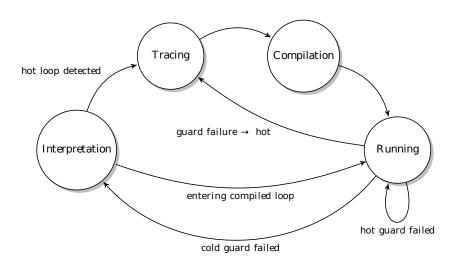












Trace trees (1)

tracetree.py def foo(): a = 0 i = 0 N = 100 while i < N: if i%2 == 0: a += 1 else: a *= 2; i += 1 return a</pre>

Trace trees (2)

```
label(start, i0, a0)

v0 = int_lt(i0, 2000)

guard_true(v0)

v1 = int_mod(i0, 2)

v2 = int_eq(v1, 0)

guard_true(v1)

a1 = int_add(a0, 10)

i1 = int_add(i0, 1)

jump(start, i1, a1)
```

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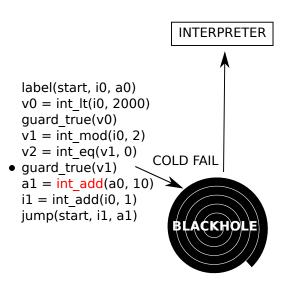
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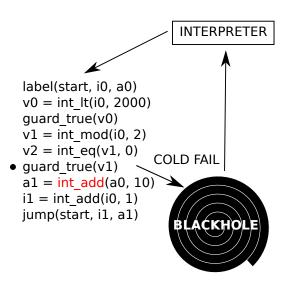
jump(start, i1, a1)
```

Trace trees (2)

```
label(start, i0, a0)
  v0 = int lt(i0, 2000)
  guard true(v0)
  v1 = int mod(i0, 2)
  v2 = int eq(v1, 0)
                       COLD FAIL
• guard true(v1)
  a1 = int add(a0, 10)
  i1 = int add(i0, 1)
 jump(start, i1, a1)
```



Trace trees (2)



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● guard_true(v1)
a1 = int_add(a0, 10)
i1 = int_add(i0, 1)
jump(start, i1, a1)

HOT FAIL
a1 = int_mul(a0, 2)
i1 = int_add(i0, 1)
jump(start, i1, a1)
```

```
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jump(start, i1, a1)
int_add(i0, 1)
jump(start, i1, a1)
```

Part 3

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

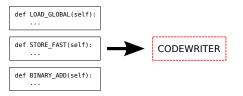
RPYTHON

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

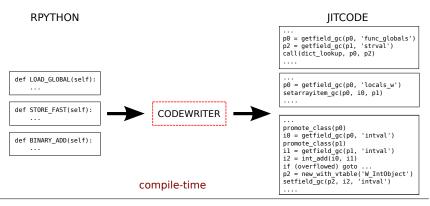
def STORE_FAST(self):
...

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

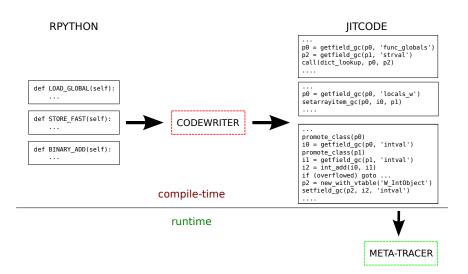
RPYTHON

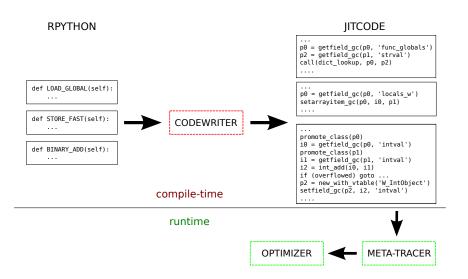


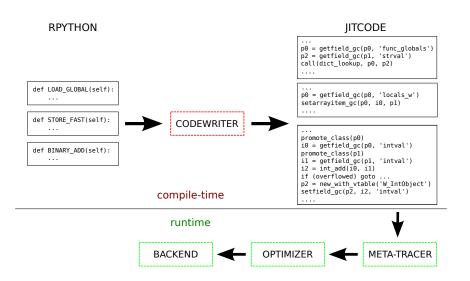
RPYTHON JITCODE p0 = getfield_gc(p0, 'func_globals') p2 = getfield_gc(p1, 'strval') call(dict lookup, p0, p2) def LOAD GLOBAL(self): p0 = getfield_gc(p0, 'locals_w') setarravitem qc(p0, i0, p1) def STORE FAST(self): CODEWRITER promote class(p0) i0 = getfield gc(p0, 'intval') def BINARY ADD(self): promote 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')

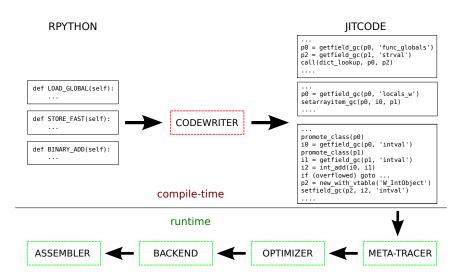


runtime









```
def fn():
    c = a+b
```

```
def fn():
    c = a+b
    ...

LOAD_GLOBAL A
LOAD_GLOBAL B
BINARY_ADD
STORE_FAST C
```

```
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```
...
p0 = getfield_gc(p0, 'func_globals')
p2 = getfield_gc(p1, 'strval')
call(dict_lookup, p0, p2)
...
```

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```

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```
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p2 = getfield gc(p1, 'strval')
call(dict lookup, p0, p2)
p0 = qetfield qc(p0, 'func qlobals')
p2 = getfield gc(p1, 'strval')
call(dict lookup, p0, p2)
guard class(p0, W IntObject)
i0 = getfield gc(p0, 'intval')
guard class(p1, W IntObject)
i1 = getfield_gc(p1, 'intval')
i2 = int add(00, i1)
quard not overflow()
p2 = new with vtable('W IntObject')
setfield gc(p2, i2, 'intval')
. . .
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setfield gc(p2, i2, 'intval')
. . .
p\theta = \text{getfield gc}(p\theta, 'locals w')
setarrayitem gc(p0, i0, p1)
. . . .
```

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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 - e.g. function frames

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')
i3 = int_add(i1, 2)
...
```

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

Constant folding (2)

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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
...
guard_not_invalidated()
...
i3 = int_add(i1, 2)
...
```

- 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

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Promotion

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

Conclusion

- PyPy is cool :-)
- Any question?