



PyPy Crash Course/Sprint Intro

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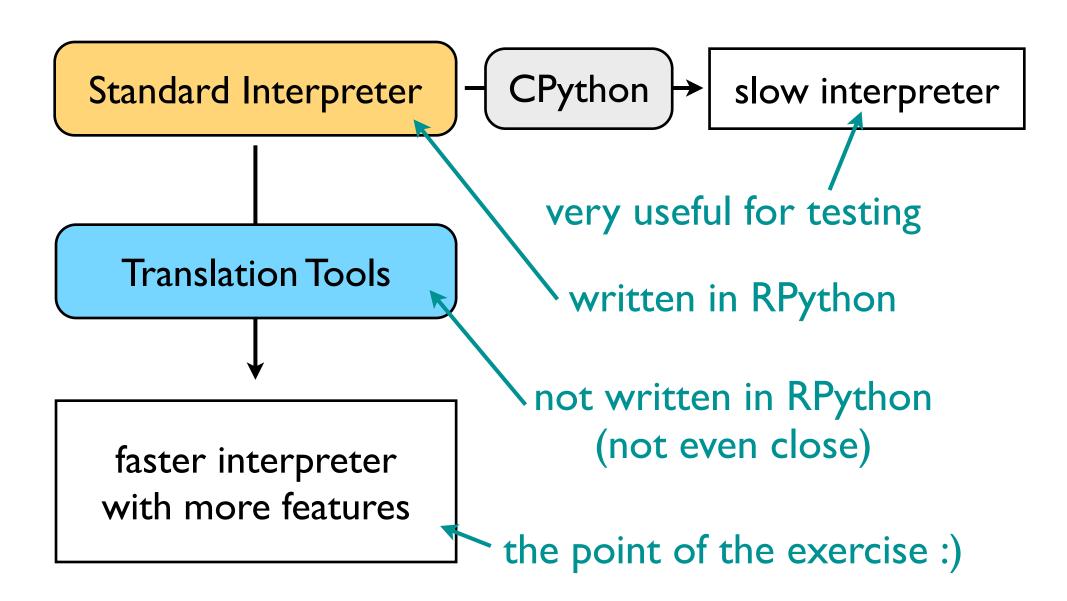
Something to look at if I'm too boring

 "getting started" has a lot of good stuff, including where to get the source, links to subversion clients and entry points:

http://codespeak.net/pypy/dist/pypy/doc/getting-started.html



Big Picture





Standard Interpreter

Standard Interpreter

Standard Object Space

Parser/Compiler

Bytecode Evaluator

CPython can be divided into the same parts with sufficient imagination — which is hardly a coincidence

independent of object space implementation, which is important

all written in RPython



The "what is RPython?" question

- RPython is first and foremost Python code
- Basically defined by being static enough for our toolchain to be able to cope with it
- Some of the restrictions are documented in the coding guide



Some Jargon

- There are many levels in PyPy
- Two of the more important, defined in terms of running PyPy on top of CPython:
 - "interp-level": code that will be executed by CPython (and get translated to C)
 - "app-level": code that will be executed by PyPy's bytecode evaluator, not CPython's



Interp/App-level

- The standard interpreter is written in a mixture of app-level and interp-level code
- Can call from one to the other
- Advantages of app-level: can use full power of Python, less code
- Advantages of interp-level: faster, closer to the metal



Status

- Standard Interpreter very complete, passes a large majority (>90%) of CPython's core regression tests
- Work being done on making the parser/ compiler configurable at runtime
- Standard Object Space and bytecode evaluator now very stable
- Potential sprint topic: adding 2.5 features



Translation Tools

Flow Object Space

Annotator

RTyper

Backend



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Analyzes a single code object to deduce control flow

We have a funky pygame flow graph viewer that we use to view these flow graphs (demo)



Translation Tools

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Analyzes an entire program to deduce type and other information

Uses abstract interpretation, rescheduling and other funky stuff



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Uses the information found by the annotator to decide how to lay out the types used by the input program in memory, and translates high level operations to lower level more pointer-ish operations



Translation Tools

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Translates low level operations and types from the RTyper to (currently)
C, JavaScript or LLVM code

Sounds like it should be easy, in fact a bit painful



Flow Analysis

- Control flow graphs are built by abstractly interpreting the code object using the standard interpreter's bytecode evaluator in an abstract domain – the Flow Object Space
- Uses state-saving tricks to consider both parts of a branch
- That's enough on how it works what it produces is much more relevant today



The Flow Model

- All defined in pypy.objspace.flow.model
- Values are either Variables or Constants
- A function's control flow graph is described by a FunctionGraph
- This contains Blocks and Links
- Blocks contain a list of SpaceOperations



The Flow Model

- SpaceOperations have an opname, a result variable and a list of args.
- The graph is naturally produced in Static
 Single Information form, which is very handy
 for later analysis
- This means each Variable is used in exactly one block and is renamed if it needs to be passed across a link

The Flow Model

Some examples:



The Annotator

- Type annotation is a fairly widely known concept – it associates variables with information about which values they might take at run time
- An unusual feature of PyPy's approach is that the annotator works on live objects
- This means it never sees initialization code,
 so that can use exec and other insane tricks



The Annotator

- Works by abstractly interpreting (a popular phrase:) the control flow graphs produced by the flow analysis
- Annotation starts at an entry point and discovers as it proceeds which functions are needed
- Read "Compiling dynamic language implementations" on the web site for more than is on these slides



The Annotator

- Works a block at a time, maintaining a pile of blocks that need analysis
- As analysis proceeds, the information about a block may get invalidated – in other words, the annotation reschedules as needed
- A fix-point approach:

```
while work_to_do: do_work()
```



The Annotation Model

- Does not modify the graphs; end result is essentially a big dictionary mapping Variables to instances of a subclass of pypy.annotation.model.SomeObject.
- Important subclasses are SomeInteger,
 SomeList, SomeInstance, SomePBC ("some prebuilt constant", includes classes and functions)



The RTyper

- An apology: "RTyping" is a pretty bad name.
 Just treat it as a random atomic identifier
 ("Frobnostication" is too hard to spell)
- Performs "representation selection" and converts high-level operations to low-level
- Potentially can target a C-ish language or an OO-language like Java or Smalltalk (OO backend somewhat theoretical at this point)



The RTyper

- Originally we tried to do the job the RTyper does at the same time as source generation
- Failed
- Miserably
- It does a job that's not part of the standard "Introduction to Compilers 101" course



Representation Selection

- The fact that the annotator performs a global analysis gives us a novel opportunity
- For example, in:

```
l = range(10)
for x in l: print l
```

can represent the return value of range as just start/stop/step, but if we know the return value of range() is going to be mutated we just return a normal list



lltypes

- pypy.rpython.lltypesystem.lltype contains a collection of Python classes that describe (and implement!) a C-like memory model with Structs, Arrays, Pointers, Signeds (integers), GcStructs, Floats... all subclasses of LowLevelType
- Convention is that variables holding instances of Iltypes are in ALLCAPS: TYPE = GcStruct("T", ("x", Signed))



Representation Selection

- The RTyper attaches an attribute "concretetype" containing an Iltype to all Constants and Variables
- During the process of RTyping, however, an instance of pypy.rpython.rmodel.Repr is created and associated with each Variable's annotation, which knows how to translate operations involving the Variable



Translating High Level to Low Level

- The high level operations such as "add" apply to different types; you can add strings, floats or integers and continually having to distinguish is annoying
- Better to have monomorphic operations int_add, float_add, str_add (well...)
- Some operations are more complex, e.g. instantiation of a class



Translating High Level to Low Level

- For each operation:
 - an instance of a subclass of Repr is created/found for each argument's annotation
 - and these are asked what low-level operation(s) the high level operation should be replaced with



Translating High Level to Low Level, example

- Start with, say,
 SpaceOperation("add", [v_x, v_y], v_z)
 with v_x and v_y (and v_z) all annotated as
 SomeInteger()
- rtype.getrepr(v_x) returns an instance of
 IntegerRepr (same for v_y)
- We end up calling a method rtype_add on a "pairtype" which makes a "int_add"
 operation



Source Generation

- Maintained backends: C, JavaScript(!) and LLVM
- All proceed in two phases:
 - Traverse the forest of rtyped graphs, computing names for everything
 - Spit out the code



Things I haven't talked about

- Specialization of functions/annotation policies
- External functions
- Low level helpers
- Garbage collection policies
- The JIT
- Constraint solving
- Geninterp
- Pairtypes
- Backend optimizations
- Exceptions in the flow model
- Multimethods in the Standard Object Space



Coding Issues

- See http://codespeak.net/pypy/dist/pypy/ doc/coding-guide.html
- But in summary: PEP 8, test driven development (using py.test)
- Away from sprints, much talking in #pypy
- Also weekly #pypy-sync meetings (horrendously timed for the US, though)