

# How to make a virtual machine less virtual

*Or: an “integrated” approach  
to dynamic language implementation*

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# Programming languages...

Programming languages are great, but...

- requirements are diverse (“none is perfect”)
- even *within one program!*

However,

- incorporating foreign code is costly
  - ◆ (think JNI, Python C API, Swig, ...)
- *per-language* debugging tools are a poor solution
  - ◆ programmer burden; lack whole-program view
- performance suffers
  - ◆ reimplementation  $\rightarrow$  re-optimisation

# One-slide summary of this talk

For the rest of this talk, I'll

- describe an approach for tackling these problems
- by changing how we implement higher-level languages
- focusing on the case of *dynamic languages*
- based on aggressive re-use of existing infrastructure
- ... esp. of *debugging*
- “the process is the VM”
- zoom in on the memory management bit
- relate it to my mainline work

This work is ongoing, unfinished, background, hangover, ...

# Unifying infrastructures help

“Isn’t this already solved?”

- JVM, CLR et al. unify many languages...
- “unify”ing FFI and debugging issues

But we could do better:

- what about *native* code? C, C++, ...
- not all languages available on all VMs
- ... FFI coding is still a big issue

What’s the “most unifying” infrastructure?

# What's in a virtual machine?

A virtual machine comprises...

- support for language implementors
  - ◆ GCing allocator; interpreter/JIT of some kind
  - ◆ object model: “typed”, flat...
  - ◆ ... on heap only
- support for end programmers, coding
  - ◆ *core* runtime library (e.g. reflection, loader, ...)
  - ◆ “native interface” / FFI
- support for end programmers, debugging / “reasoning”
  - ◆ interfaces for debuggers, ...
- support for users / admins (security, res. man't, ...)

# What's in a ~~virtual machine~~? an OS process + minimal libc?

A **The “null”** virtual machine comprises...

- support for language implementors
  - ◆ ~~GCing~~ allocator; ~~interpreter/JIT~~ of some kind
  - ◆ object model: “typed”, flat **opaque**...
  - ◆ ... on heap ~~only~~ **or stack or bss/rodata**
- support for end programmers, coding
  - ◆ *core* runtime library (e.g. **reflection**, loader, ...)
  - ◆ “~~native~~ interface” / FFI
- support for end programmers, debugging / “reasoning”
  - ◆ interfaces for debuggers, ... **at whole process scale**
- support for users / admins (security, res. man't, ...)

# Astonishing claim

For most omissions, we can plug in libraries:

- JIT/interpreter...
- choose a GC (Boehm; can do better?)

Wha about reflection?

- ... more generally, “dynamic” features

Debugging infrastructure supports all kinds of dynamism:

- name resolution, dynamic dispatch, ...
- object schema updates (with some work)

... on *compiled* code, in *any* (compiled) language!

# Well, almost...

Building “null VM” Python means plugging a few holes:

- ... that are *already* problems for debuggers!
- that fit neatly into runtime and/or debugger facilities

I’m going to focus on a “hole”.

- For the rest, ask me (or trust me...)

# Some equivalences

debugging-speak

---

backtrace

state inspection

memory leak detection

altered execution

edit-and-continue

breakpoint

bounds checking

runtime-speak

stack unwinding

reflection

garbage collection

eval function

dynamic software update

dynamic weaving

(spatial) memory safety

For each pair, implement using the same infrastructure...

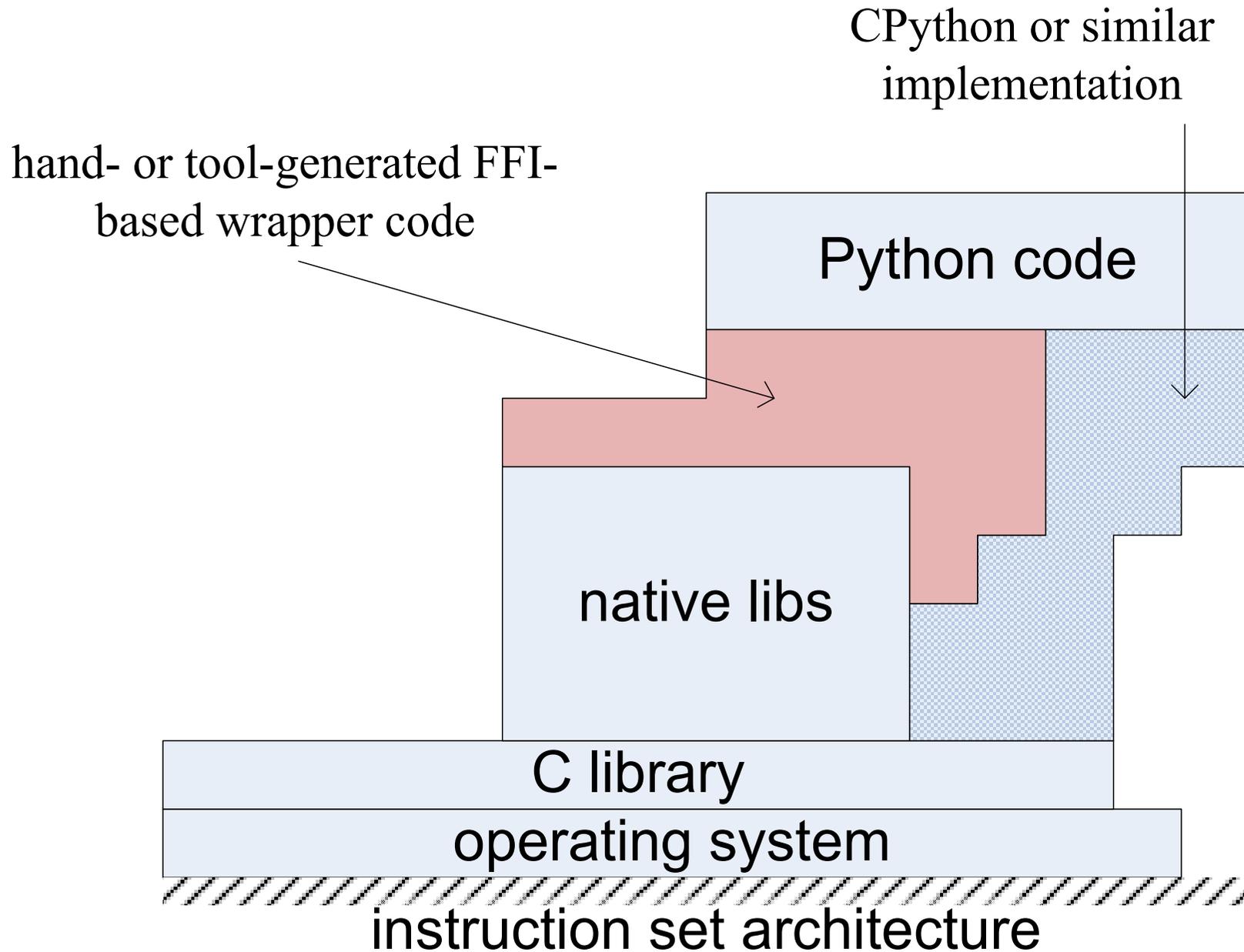
# DwarfPython in one slide

DwarfPython is an implementation of Python which

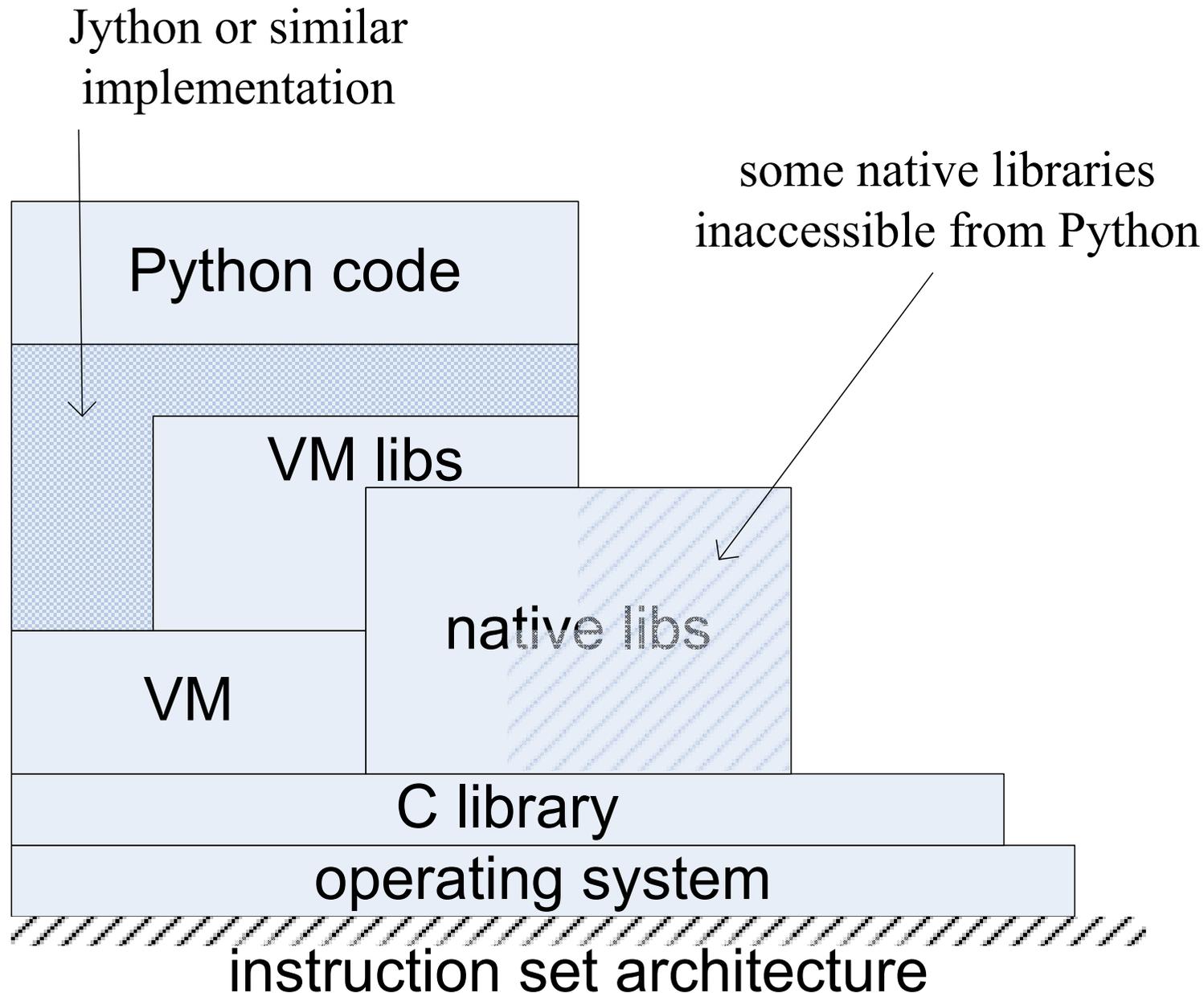
- uses DWARF debug info to understand native code...
  - ◆ ... and itself!
- unifies Python object model with native (general) model
  - ◆ this is key!
- small, uniform changes allow `gdb`, `valgrind`, ...
  - ◆ as a consequence of above two points
- deals with other subtleties...
  - ◆ I count 19 “somewhat interesting” design points

Not (yet): parallel / high-perf., Python *libraries*, ...

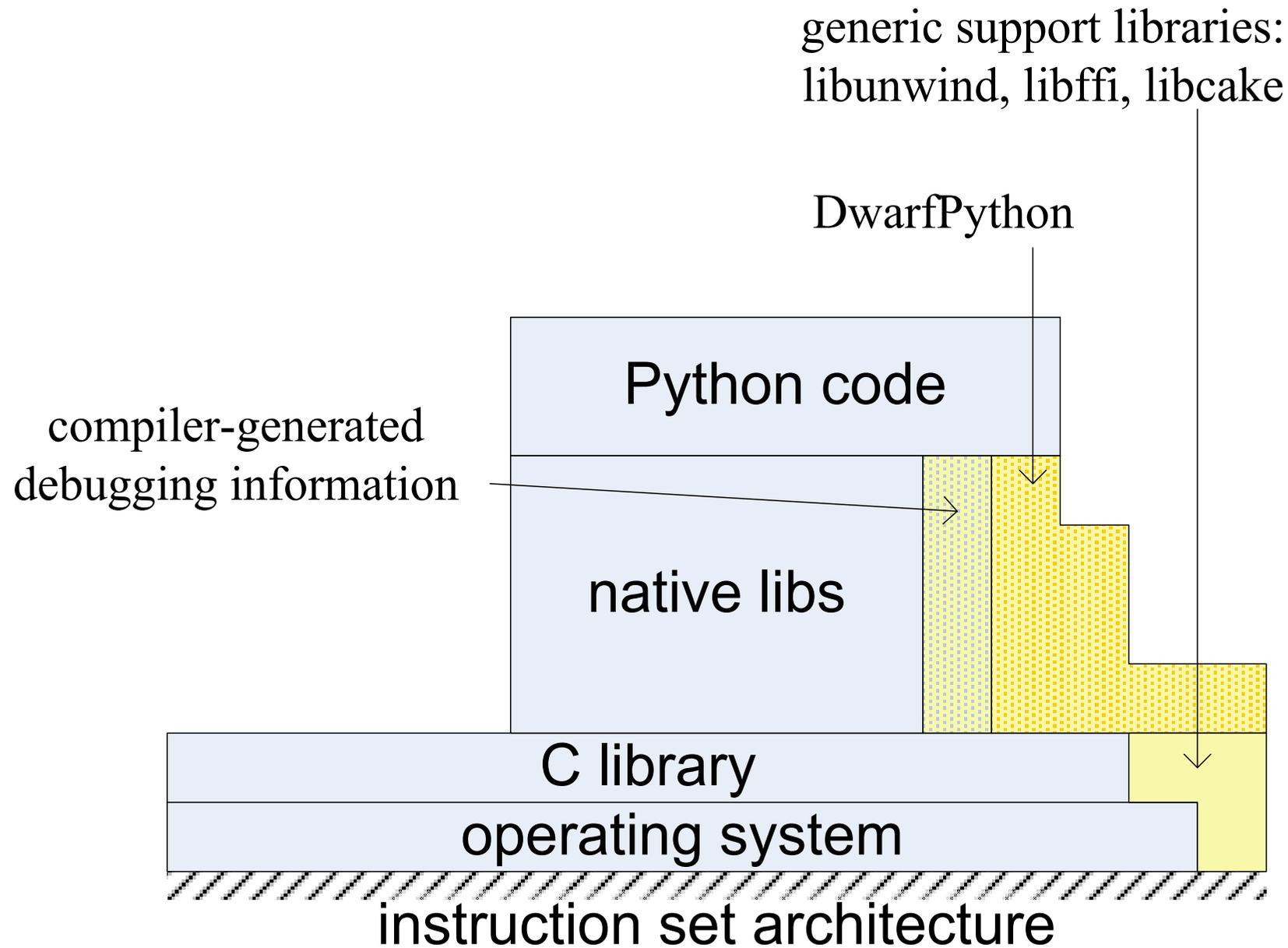
# Implementation tetris (1)



# Implementation tetris (2)



# Implementation tetris (3)



# Objects are not really opaque...

```
>>> import ellipse # dlopen()s libellipse.so
>>> my_ellipse = native_new_ellipse()
>>> print my_ellipse
```

## Invariant 1: all objects have DWARF layout descriptions...

my\_ellipse



```
struct ellipse {
    double maj;
    double min;
    struct point {
        double x, y;
    } ctr;
}
```

2d: DW\_TAG\_structure\_type

DW\_AT\_name : point

39: DW\_TAG\_member

DW\_AT\_name : x

DW\_AT\_type : <0x52>

DW\_AT\_location: (DW\_OP\_plus\_uconst: 0

45: DW\_TAG\_member

DW\_AT\_name : y

DW\_AT\_type : <0x52>

DW\_AT\_location: (DW\_OP\_plus\_uconst: 8

52: DW\_TAG\_base\_type

DW\_AT\_byte\_size : 8

DW\_AT\_encoding : 4 ( float )

DW\_AT\_name : double

How to make a VM... - p.13

59: DW\_TAG\_structure\_type

# Calling functions

```
>>> import c # libc.so already loaded
>>> def bye(): print "Goodbye, world!"
...
>>> atexit(bye)
```

Invariant 2: *all* functions have  $\geq 1$  “native” entry point

- for Python code these are generated at run time

DwarfPython uses `libffi` to implement *all* calls

# Object models

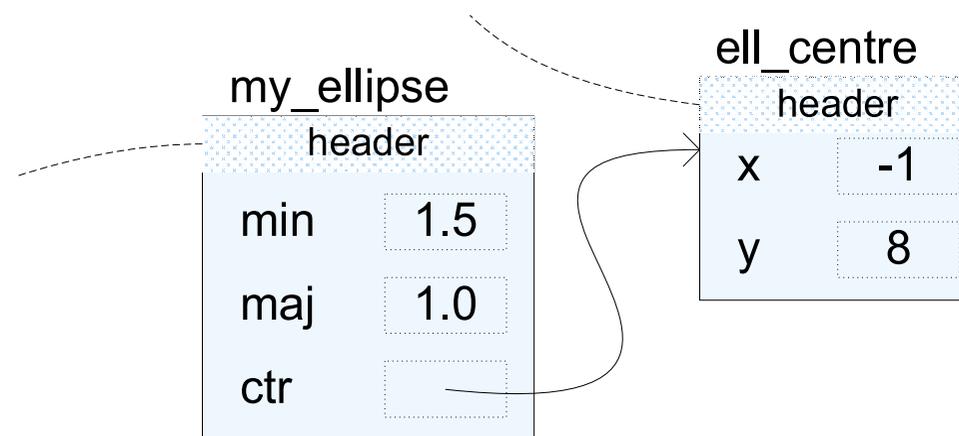
Dynamic dispatch means finding object metadata. Problem!

my\_ellipse



```
struct ellipse {  
    double maj;  
    double min;  
    struct point {  
        double x, y;  
    } ctr;  
}
```

Native objects are trees; no descriptive headers, whereas...



VM-style objects: “no interior pointers” + custom headers

# Wanted: fast metadata lookup

How can we locate an object's DWARF info

- ... without object headers?
- ... given possibly an *interior* pointer?

Solution:

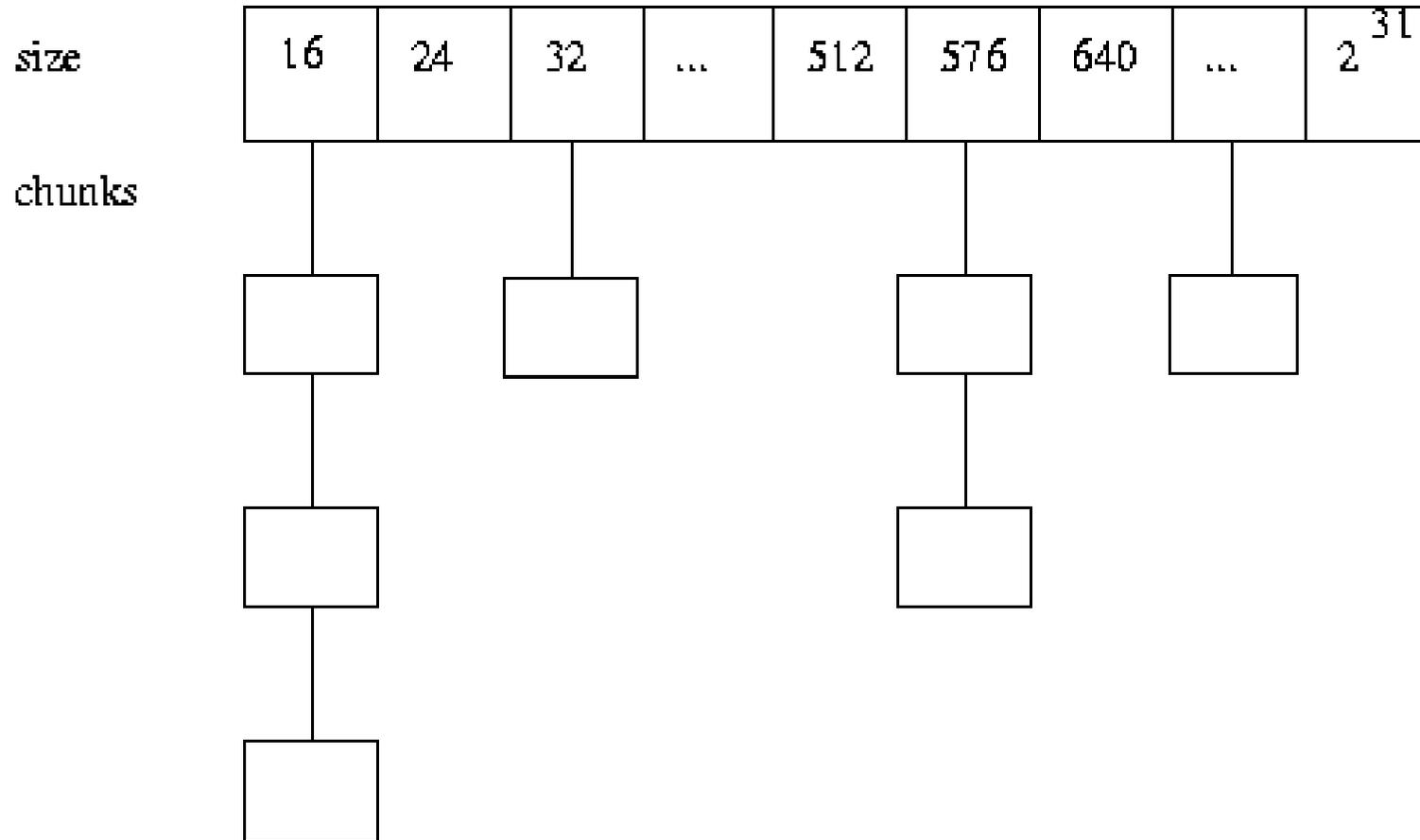
- is object on stack, heap or bss/rodata? ask memory map
- if static or stack, just use debug info (+ stack walker)

In the heap (difficult) case:

- we'll need some malloc() hooks...
- ... and a *memtable*.
  - ◆ read: efficient *address-keyed* associative structure

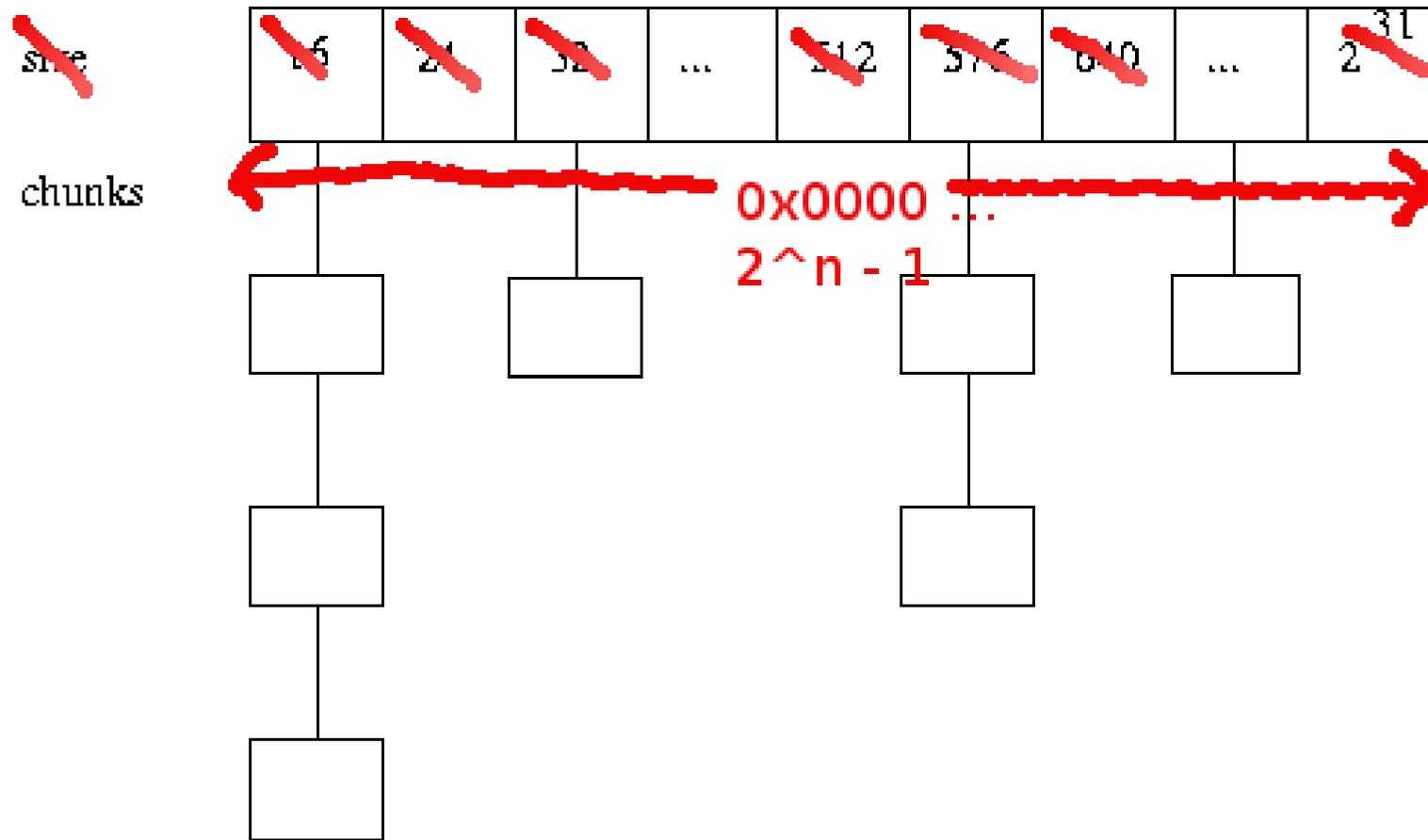
# Indexing chunks

Inspired by free chunk binning in Doug Lea's (old) malloc.



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As well as indexing *free* chunks binned by *size*,  
... index *allocated* chunks binned by *address*

# How many bins?

Each bin is a linked list of chunks

- thread next/prev pointers through allocated chunks...
  - ◆ hook can add space, if no spare bits
- also store allocation site (key to DWARF info)
- can compress all this quite small (48 bits)

Q: How big should we make the bin index?

A: As big as we can!

- given an interior pointer, finding chunk is  $O(\text{binsize})$

Q: How big *can* we make the bin index?

A: Really really huge!

# Really, how big?

## Exploit

- sparseness of address space usage
- lazy memory commit on “modern OSes” (Linux)

Bin index resembles a linear page table.



After some tuning...

- 32-bit AS requires  $2^{22}$  bytes of VAS for bin index
- covering  $n$ -bit AS requires  $2^{n-10}$ -byte bin index...
- use bigger index for smaller expected bin size

# What's the benefit?

Faster and more space-efficient than a hash table

- also better cache and demand-paging behaviour?

Some preliminary figures (timed gcc, 3 runs):

- gcc uninstrumented: 1.70, 1.76, 1.72
- gcc + no-op hooks: 1.73, 1.76, 1.72
- gcc + vgHash index: 1.83, 1.82, 1.85
- gcc + memtable index: 1.77, 1.78, 1.77

Memtables are not limited to this application!

- e.g. Cake “corresponding objects” look-up
- ... your idea here

# Status of DwarfPython

Done: first-pass simplified implementation

- DWARF-based foreign function access
- no dynamic lang. features, debugger support, ...

Full implementation in progress...

- including proof-of-concept extension of LLDB
- + feedback into DWARF standards!

What's the big picture behind DwarfPython?

- habilitation of new / dynamic / unusual languages
- ... into a mainstream toolchain
- language-independent notion of “API”
- orthogonalise language from tool support

What other neat tools might now be applicable to Python?

- tracers (e.g. ltrace)
- race detectors (helgrind or similar)
- heap profilers (massif, ...)

What about verification / bug-finding tools?

# Very quick summary

Wanted: a tool that can answer questions of the form:

- “how does my program exercise this API?” (general)
- e.g. “how does my program use the filesystem API?”
  - ◆ what data will it write? delete/overwrite?
  - ◆ what data will it *not* write? lose on crash?

How? Using Klee, a “dynamic symbolic execution” engine.

- works on binaries (LLVM bitcode as it happens)
- is it a static or a dynamic analysis? Hmm!

Ask me for more about this...

# Conclusions & work in progress

Language implementors can do more to

- make using foreign code easier;
- orthogonalise language from tool support.

Questions for the audience:

- pessimal cases / bad GC interactions?
- can we do better?
- other uses of memtables? (“less conservative” GC?)

Still to do: implementation, benchmarks...

Thanks for listening. Any questions?

## Taster: wrapper-free FFI (2)

Calling native functions:

- instantiate the data types the function expects
- call using `libffi`

In Parathon, an earlier effort, we had:

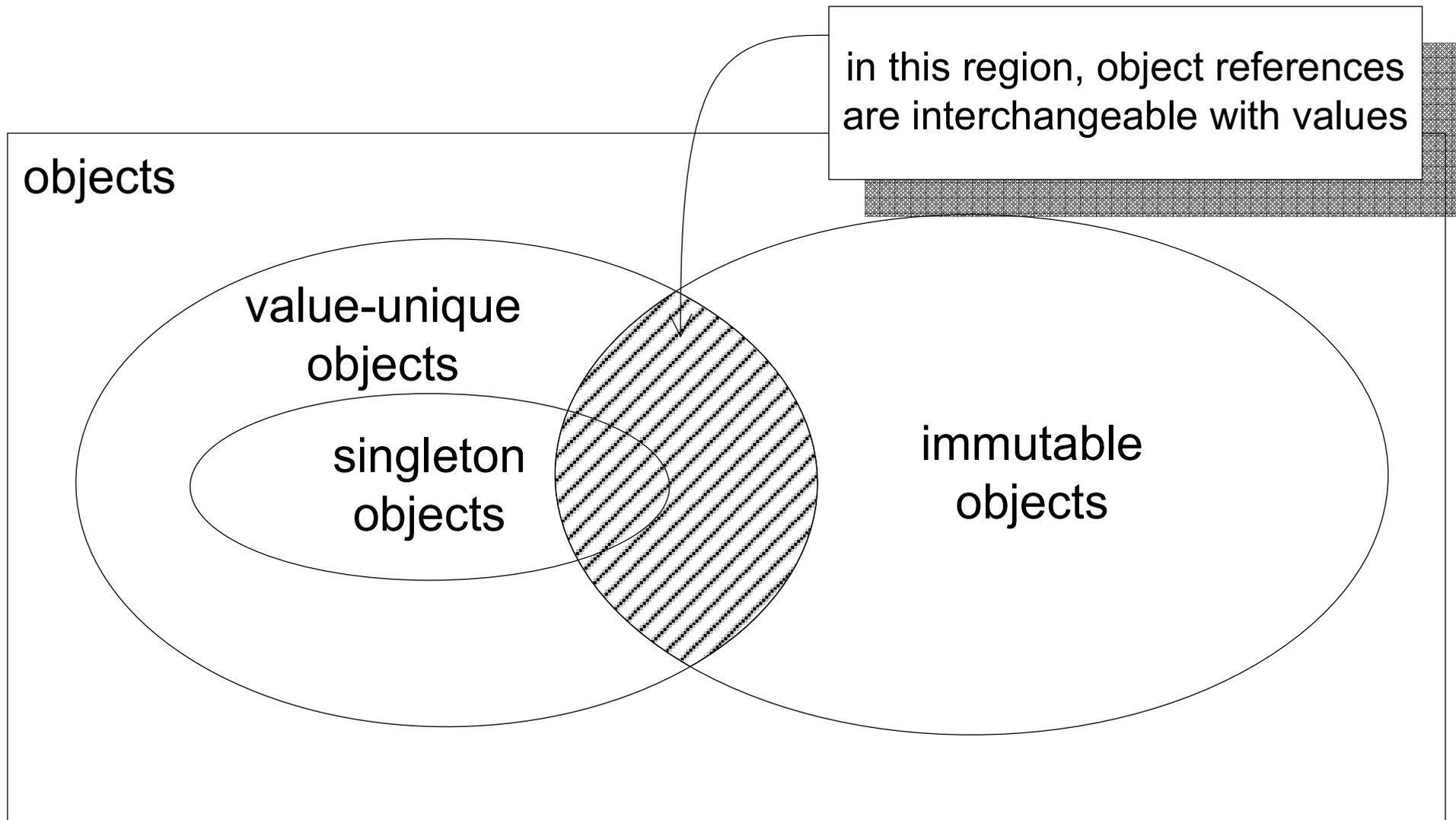
```
ParathonValue* FunctionCall::evaluate(ParathonContext& c)
{   return call_function (this→base_phrase→evaluate(c),
    /* invokes libffi ^ */ this→parameter_list→asArgs(c)); }
```

Now we have:

```
val FunctionCall::evaluate() // ← only context is the *process* i.e. stack
{   return call_function (this→base_phrase→evaluate(),
                        this→parameter_list→asArgs()); }
```

The interpreter context *is* the process context!

# Primitive values



# Out-of-band metadata

