



Cornell University  
Computer Systems Laboratory

























# Cross-Layer Workload Characterization of Meta-Tracing JIT VMs

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# Dynamic languages are popular

Language Rank	Types	Spectrum Ranking
1. Python	 	100.0
2. C	  	99.7
3. Java	  	99.4
4. C++	  	97.2
5. C#	  	88.6
6. R		88.1
7. JavaScript	 	85.5
8. PHP		81.4
9. Go	 	76.1
10. Swift	 	75.3

S. Cass. "The 2017 Top Programming Languages." IEEE Spectrum.



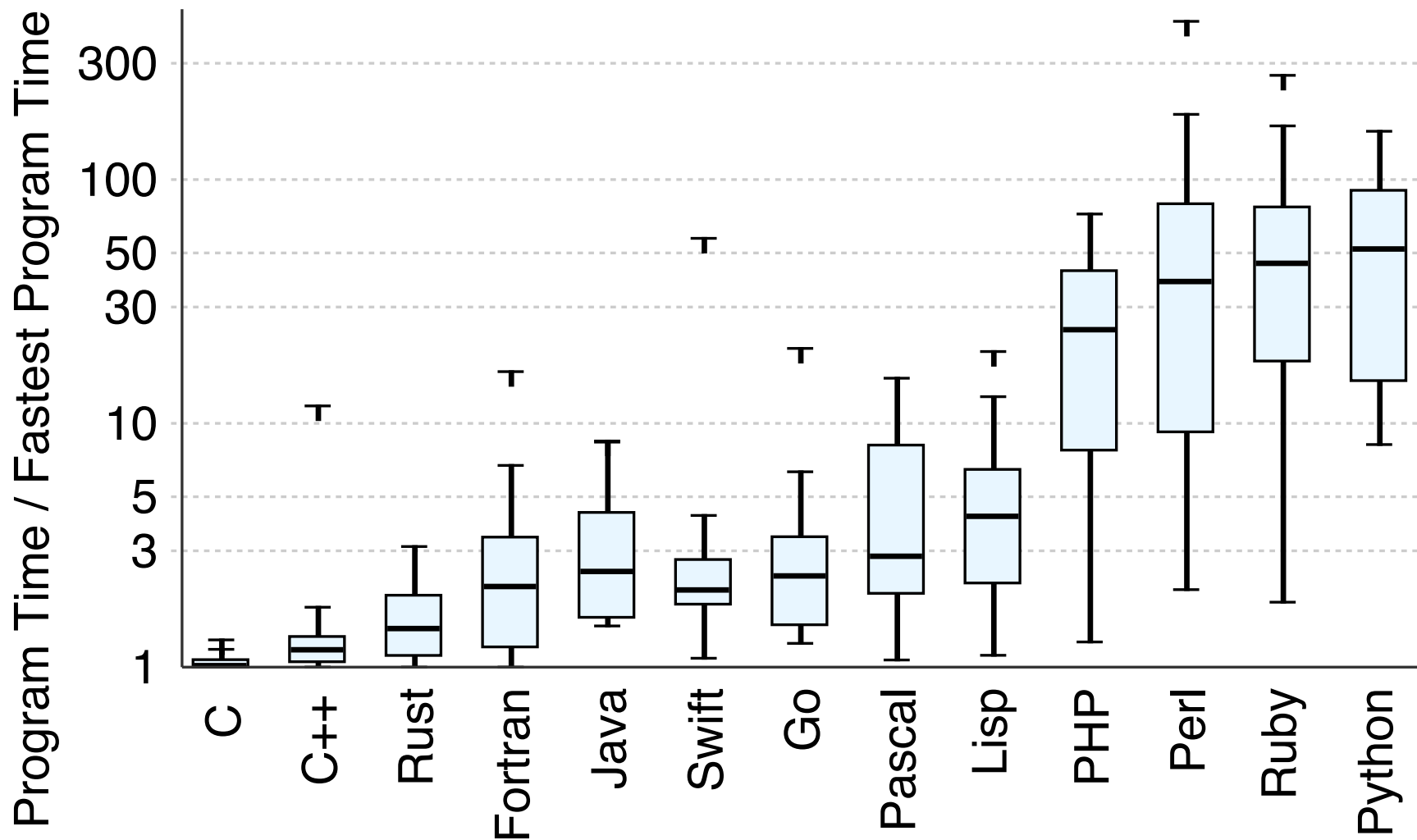
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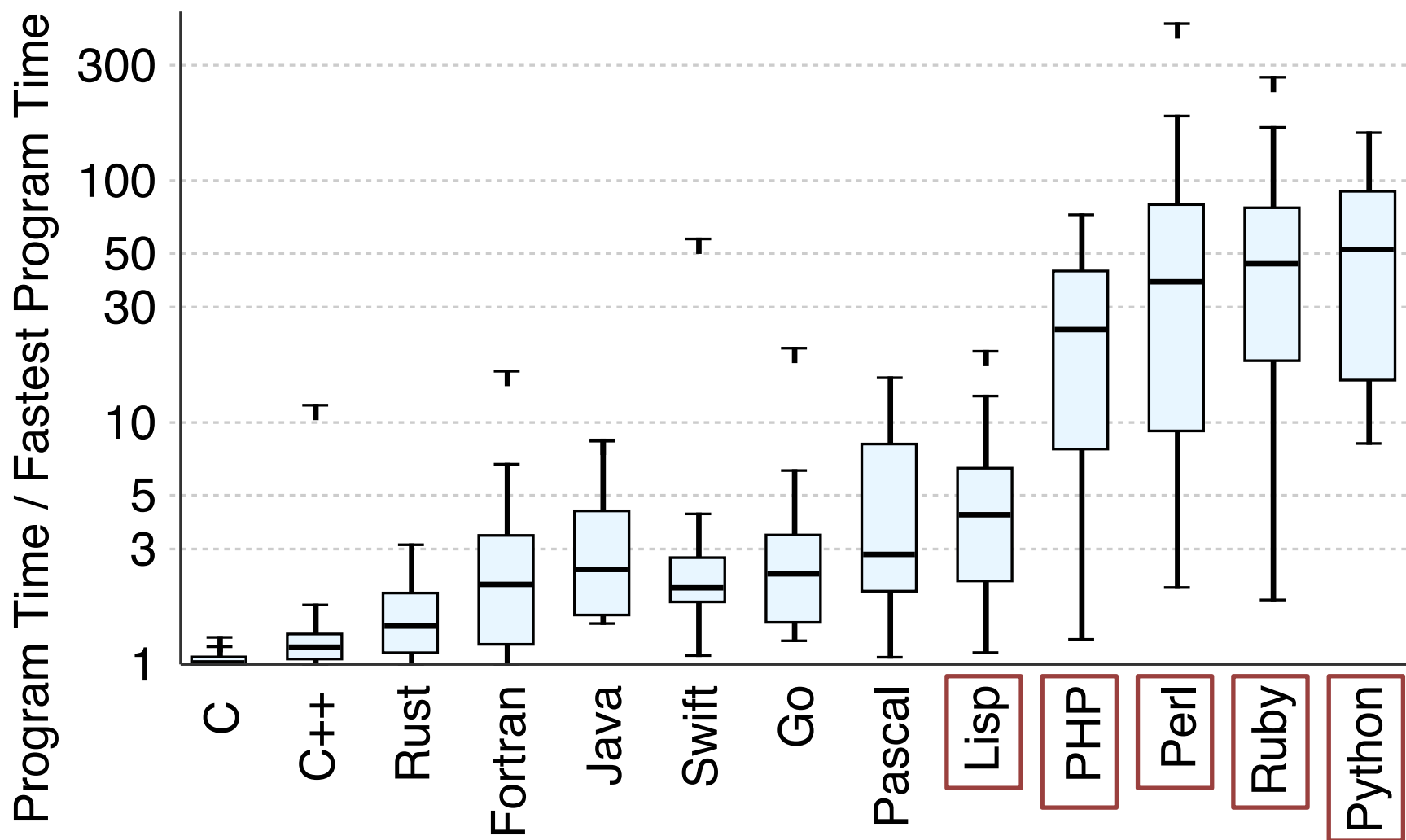
# Dynamic languages are slow



I. Guoy. "The Computer Languages Benchmarks Game."



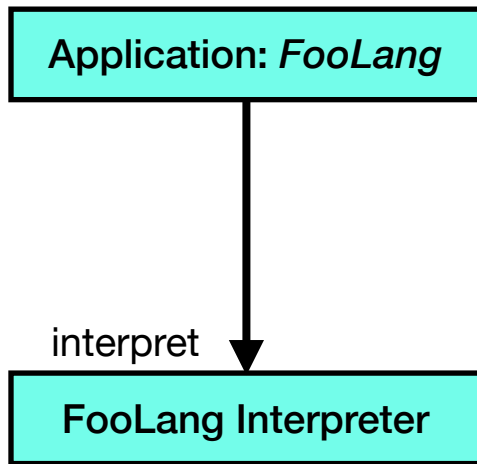
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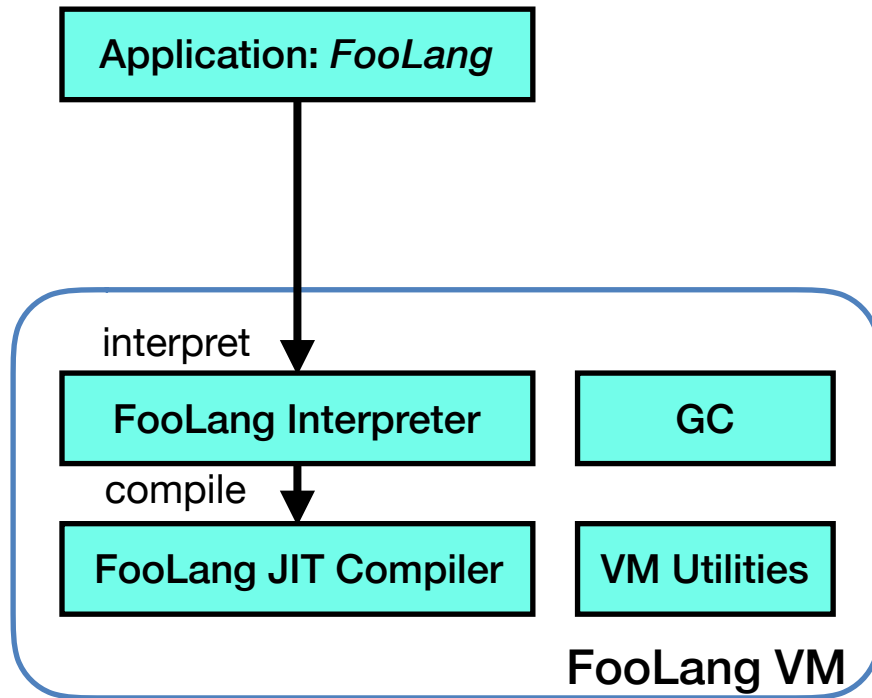
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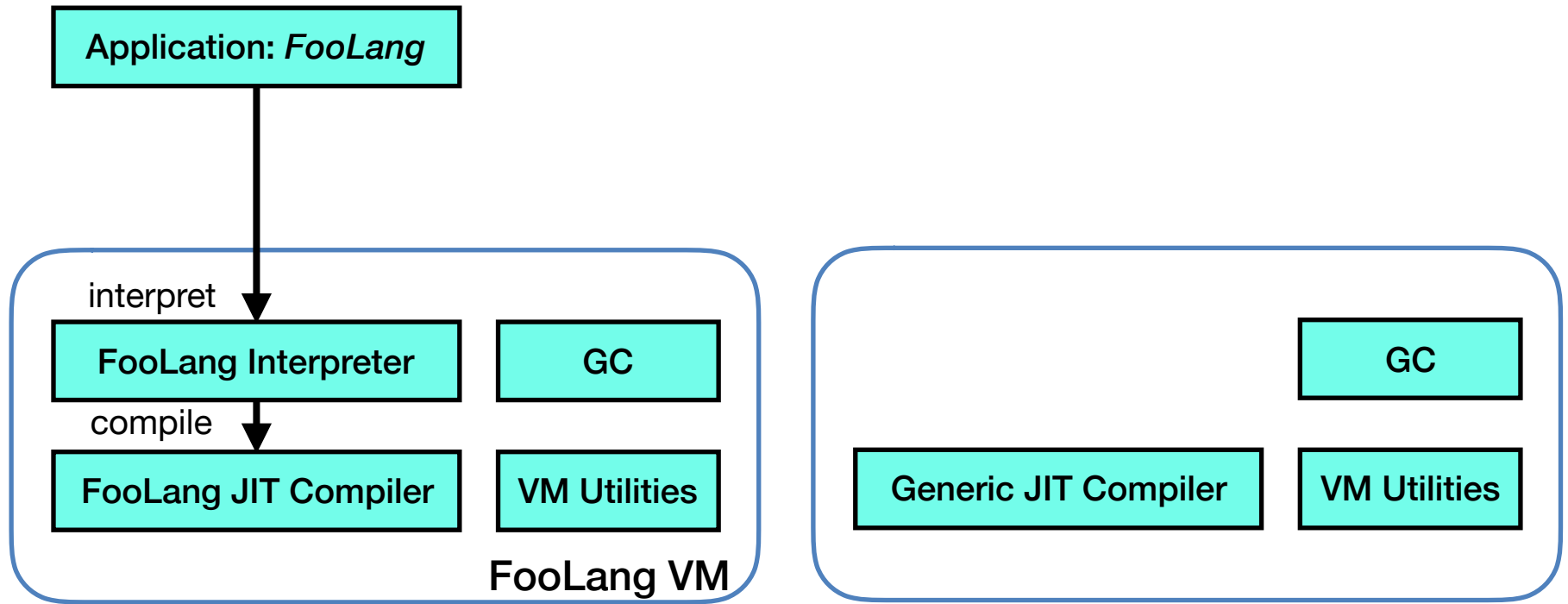
# Just-in-time-compiling virtual machines



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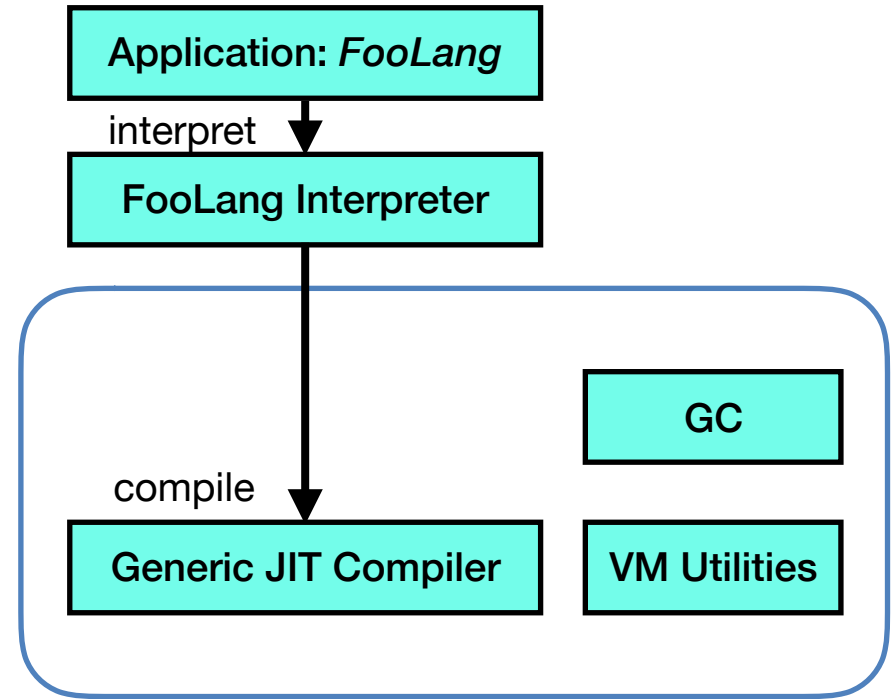
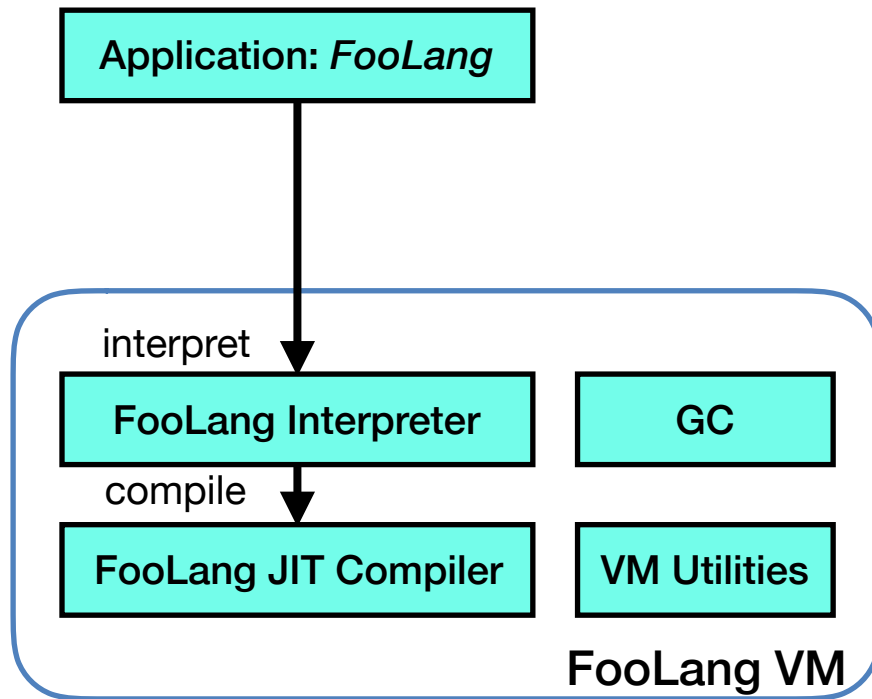


# Just-in-time-compiling virtual machines

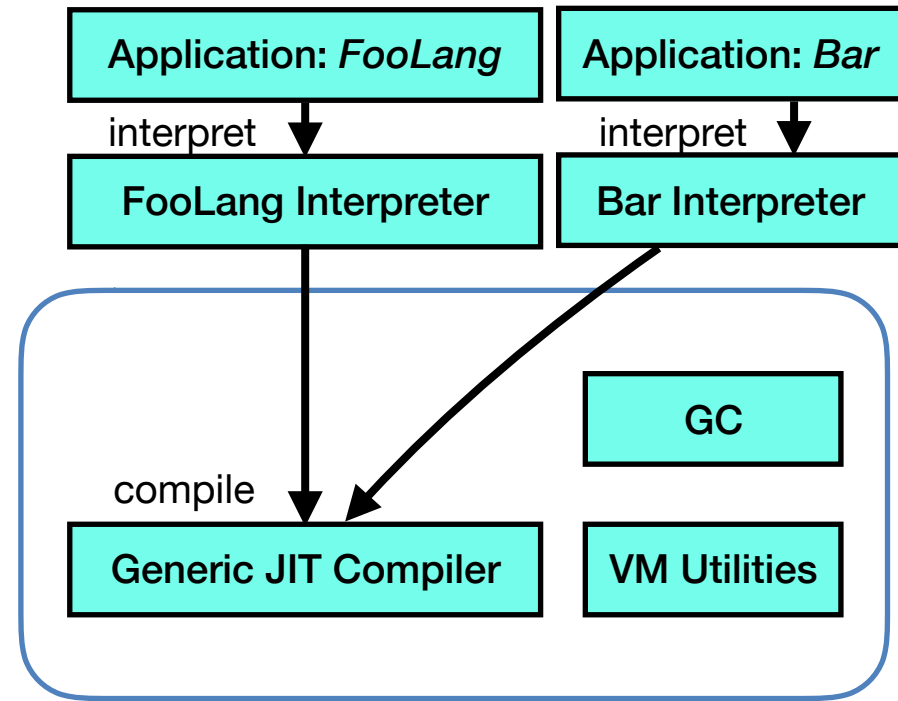
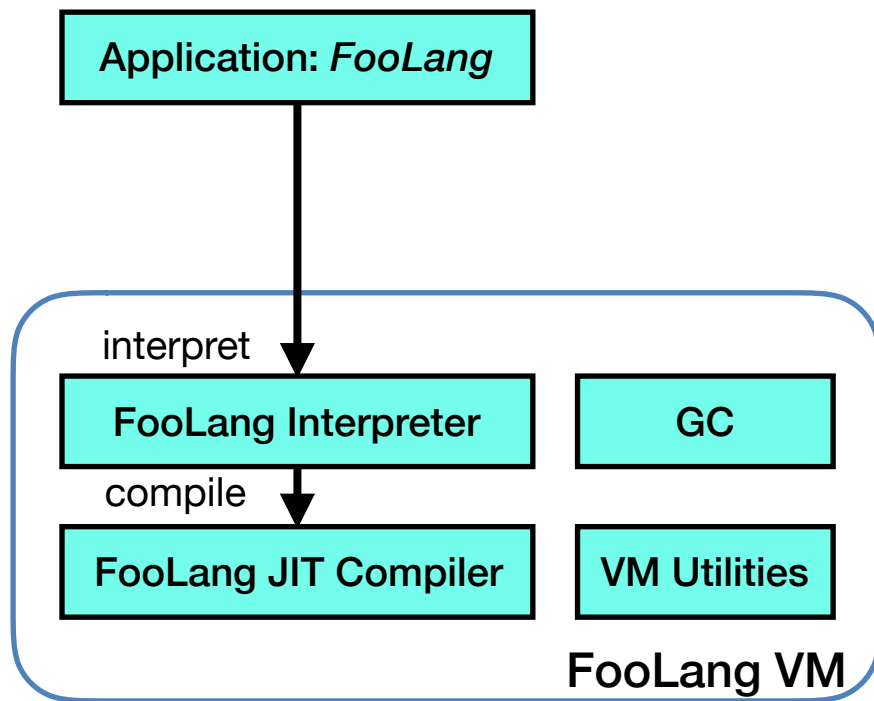




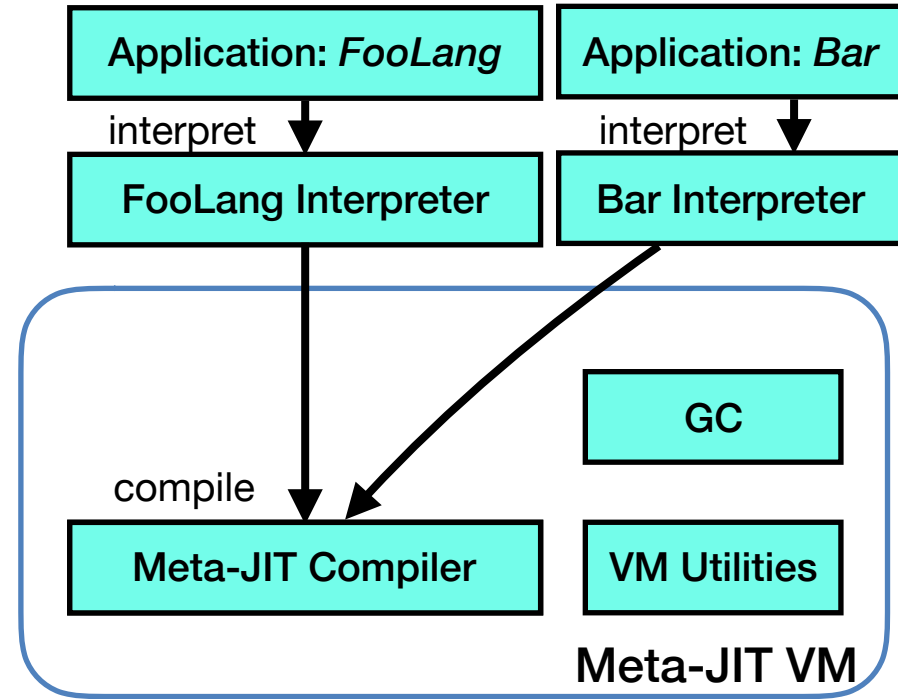
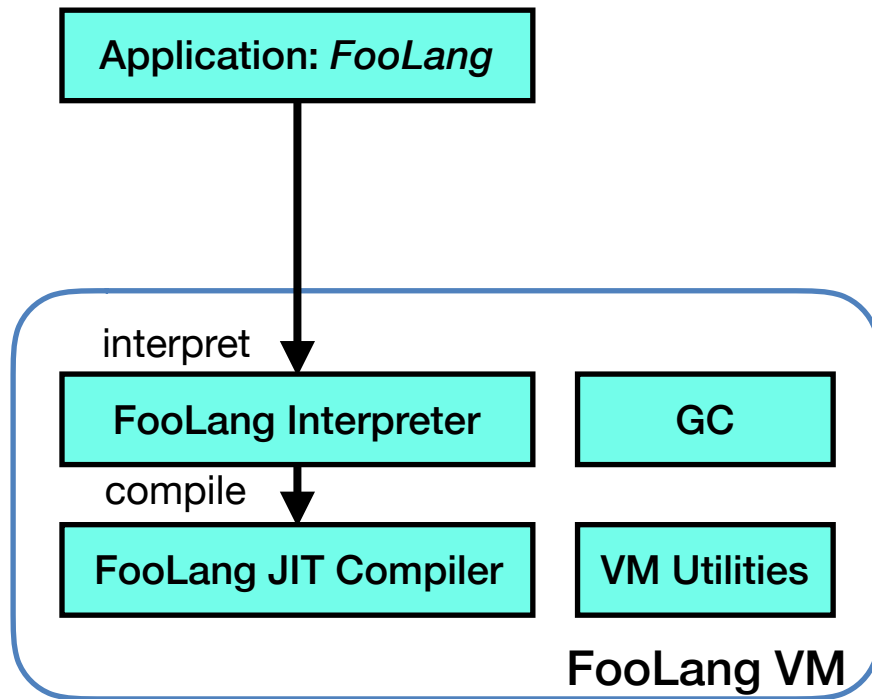
# Just-in-time-compiling virtual machines



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# Just-in-time-compiling virtual machines



# Meta-JIT approaches: meta-tracing and partial evaluation

## RPython Framework

Meta-tracing: meta-interpreter  
and tracing JIT

## Truffle/Graal Framework

Partial evaluation: partial evaluator  
and method JIT

```
def max(a, b):  
    if a > b:  
        return a  
    else:  
        return b
```



# Meta-JIT approaches: meta-tracing and partial evaluation

## RPython Framework

Meta-tracing: meta-interpreter  
and tracing JIT

### Linear JIT IR

```
guard_type(a, int)
guard_type(b, int)
c = int_gt(a, b)
guard_true(c)
return(a)
```

```
def max(a, b):
    if a > b:
        return a
    else:
        return b
```

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def max(a, b):
    if a > b:
        return a
    else:
        return b
```

## Truffle/Graal Framework

Partial evaluation: partial evaluator  
and method JIT

### JIT IR

```
guard_type(a, int)
guard_type(b, int)
c = int_gt(a, b)
jump_if_false(c, L1)
return(a)
L1: return(b)
```



# Meta-JIT approaches: meta-tracing and partial evaluation

## RPython Framework

Meta-tracing: meta-interpreter  
and tracing JIT

### Linear JIT IR

```
guard_type(a, int)
guard_type(b, int)
c = int_gt(a, b)
guard_true(c)
return(a)
```

**Bridge (a <= b)**

```
return(b)
```

```
def max(a, b):
    if a > b:
        return a
    else:
        return b
```

## Truffle/Graal Framework

Partial evaluation: partial evaluator  
and method JIT

### JIT IR

```
guard_type(a, int)
guard_type(b, int)
c = int_gt(a, b)
jump_if_false(c, L1)
return(a)
L1: return(b)
```



# Meta-JIT approaches: meta-tracing and partial evaluation

## RPython Framework

Meta-tracing: meta-interpreter  
and tracing JIT

### Linear JIT IR

```
guard_type(a, int)
guard_type(b, int)
c = int_gt(a, b)
guard_true(c)
return(a)
```

### Bridge (a <= b)

```
return(b)
```

### Bridge(float)

```
guard_type(a, float)
guard_type(b, float)
c = float_gt(a, b)
guard_true(c)
return(a)
```

```
def max(a, b):
    if a > b:
        return a
    else:
        return b
```

## Truffle/Graal Framework

Partial evaluation: partial evaluator  
and method JIT

### JIT IR

```
guard_type(a, int)
guard_type(b, int)
c = int_gt(a, b)
jump_if_false(c, L1)
return(a)
L1: return(b)
```





# Meta-JIT approaches: meta-tracing and partial evaluation

## RPython Framework

Meta-tracing: meta-interpreter  
and tracing JIT

### Linear JIT IR

```
guard_type(a, int)
guard_type(b, int)
c = int_gt(a, b)
guard_true(c)
return(a)
```

### Bridge (a <= b)

```
return(b)
```

### Bridge (float)

```
guard_type(a, float)
guard_type(b, float)
c = float_gt(a, b)
guard_true(c)
return(a)
```

```
def max(a, b):
    if a > b:
        return a
    else:
        return b
```

## Truffle/Graal Framework

Partial evaluation: partial evaluator  
and method JIT

### JIT IR

### Re-optimized JIT IR

```
i = is_type(a, int)
jump_if_false(i, L2)
guard_type(b, int)
c = int_gt(a, b)
jump_if_false(c, L1)
return(a)
L1: return(b)
L2: guard_type(a, float)
guard_type(b, float)
c = float_gt(a, b)
jump_if_false(c, L3)
return(a)
L3: return(b)
```



# Cross-layer workload characterization of meta-tracing JIT VMs

PyPy >> CPython



# Cross-layer workload characterization of meta-tracing JIT VMs

## PyPy >> CPython

- How can meta-tracing JITs significantly improve the performance of multiple dynamic languages?



# Cross-layer workload characterization of meta-tracing JIT VMs

**PyPy >> CPython**

- How can meta-tracing JITs significantly improve the performance of multiple dynamic languages?

**PyPy << C**



# Cross-layer workload characterization of meta-tracing JIT VMs

## PyPy >> CPython

- How can meta-tracing JITs significantly improve the performance of multiple dynamic languages?

## PyPy << C

- Why are meta-tracing JITs for dynamic programming still slower than C?



# Python-based interpreter

Application: *FooLang*



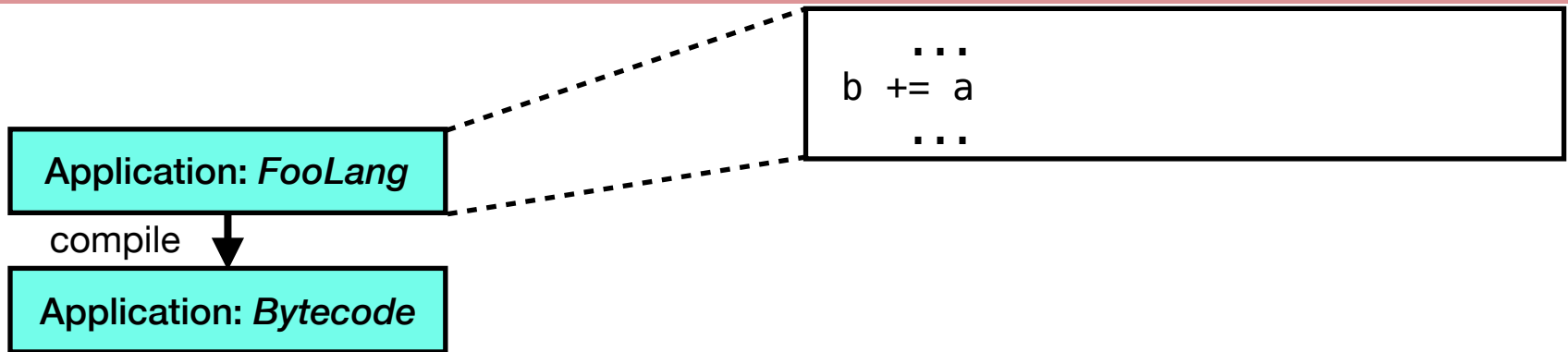
# Python-based interpreter

Application: *FooLang*

```
...  
b += a  
...
```

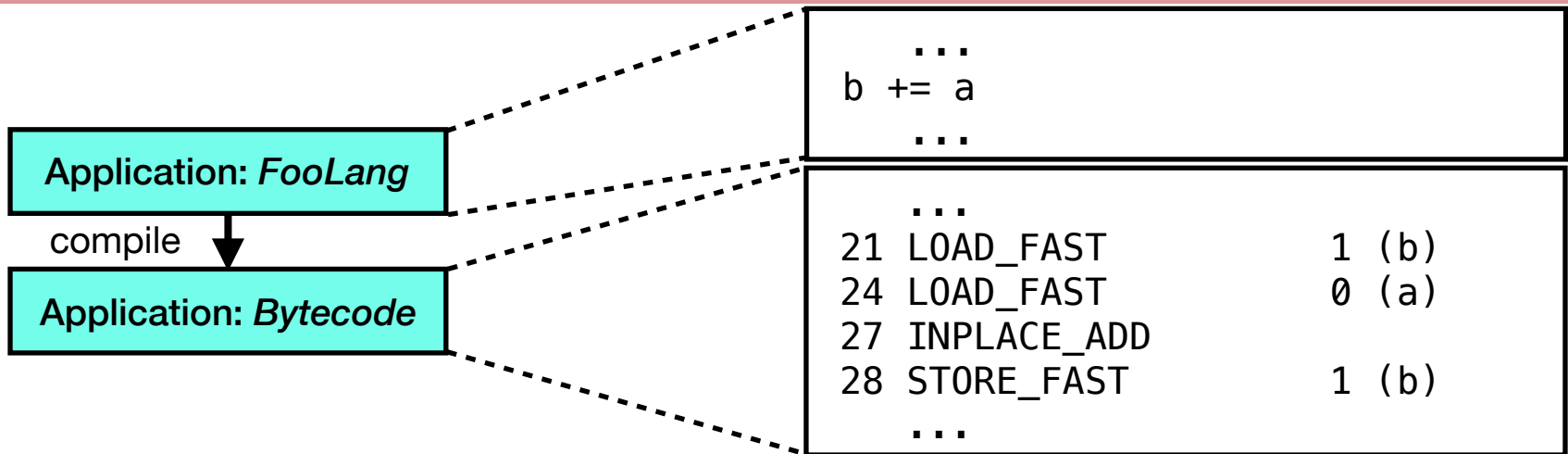


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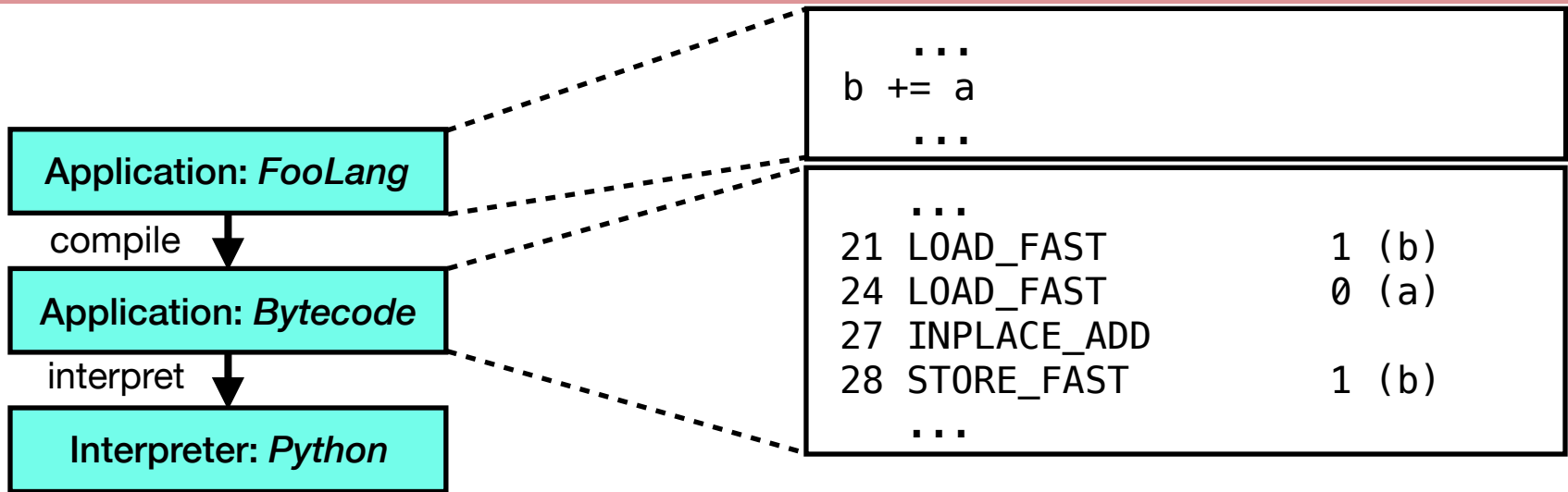




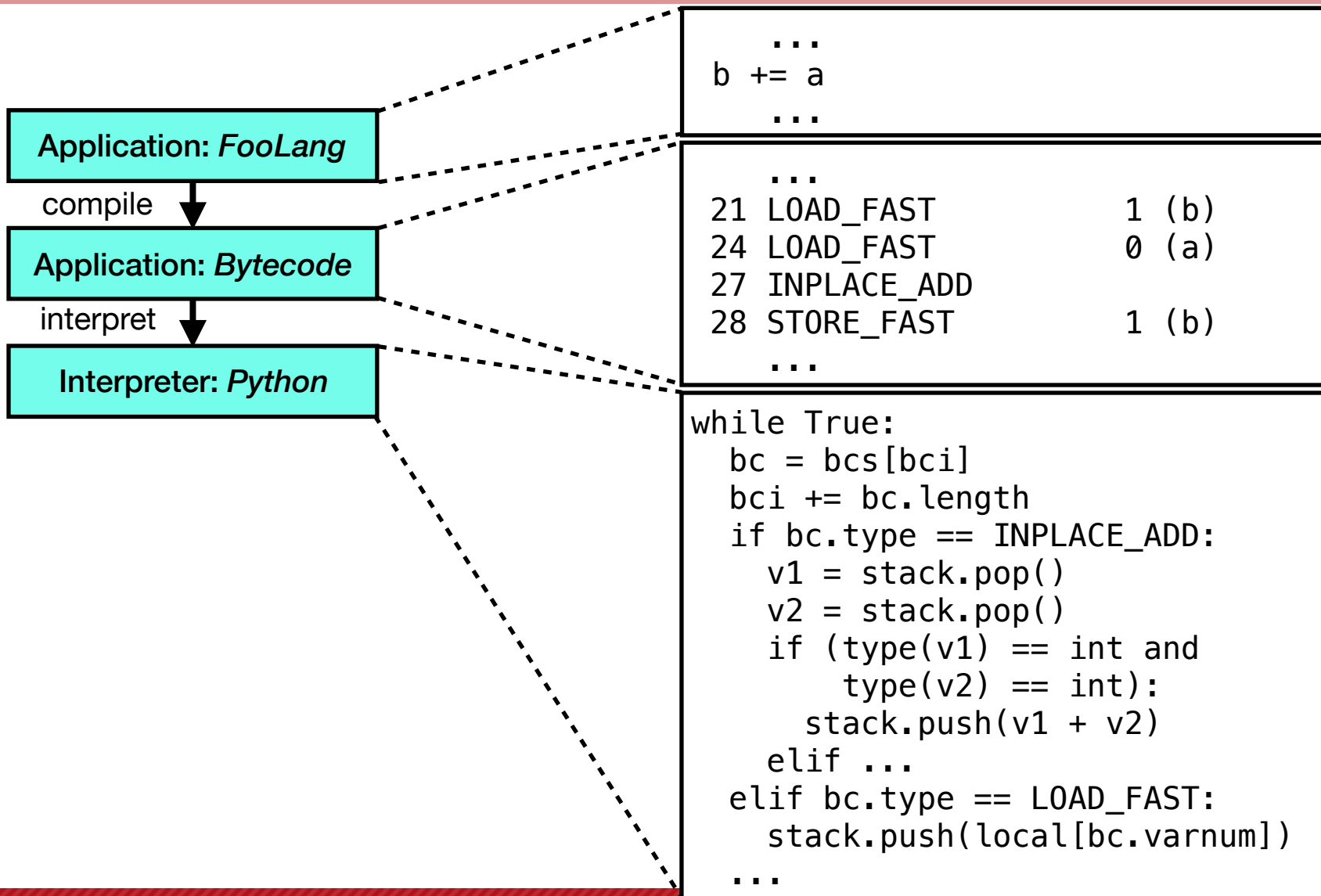
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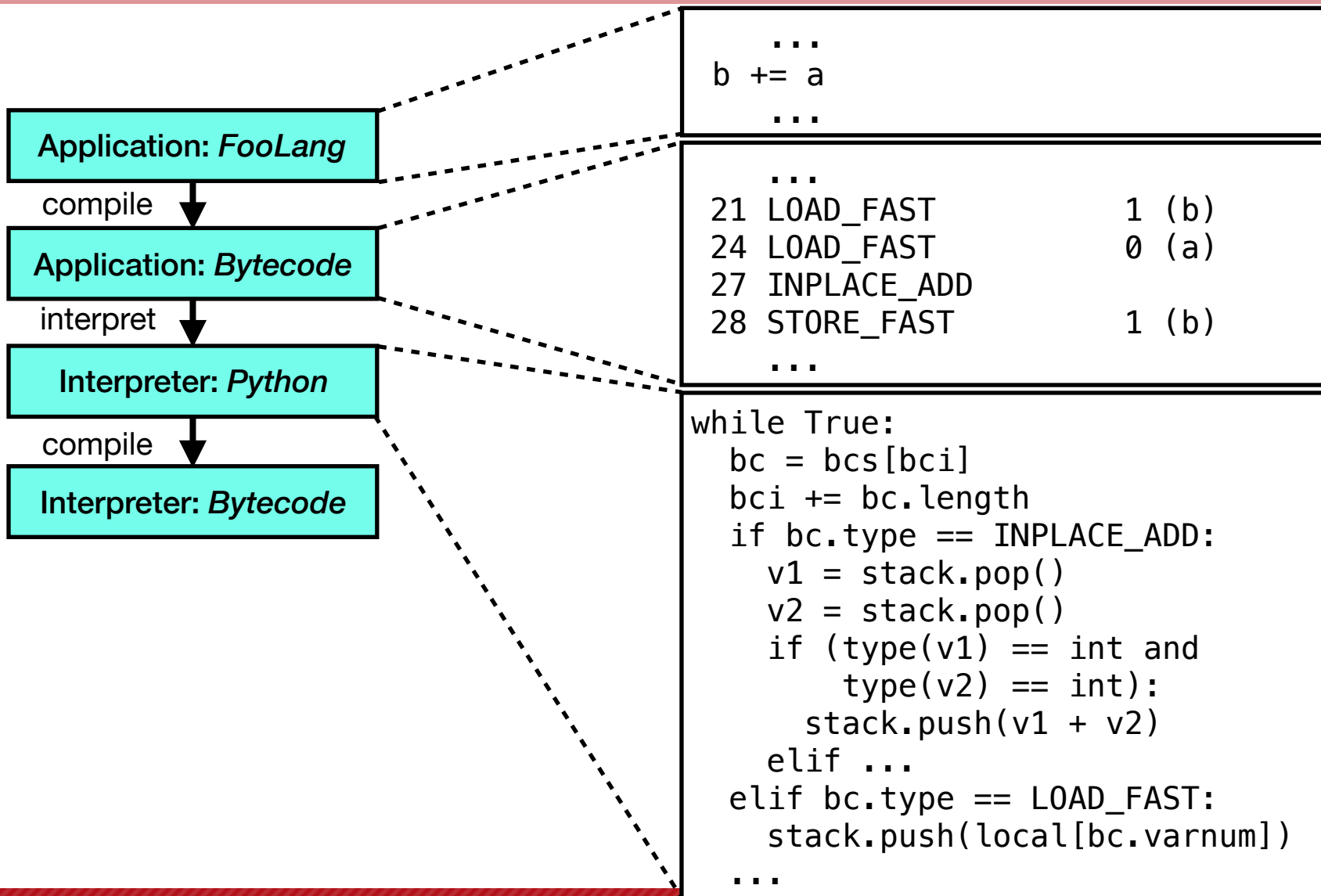
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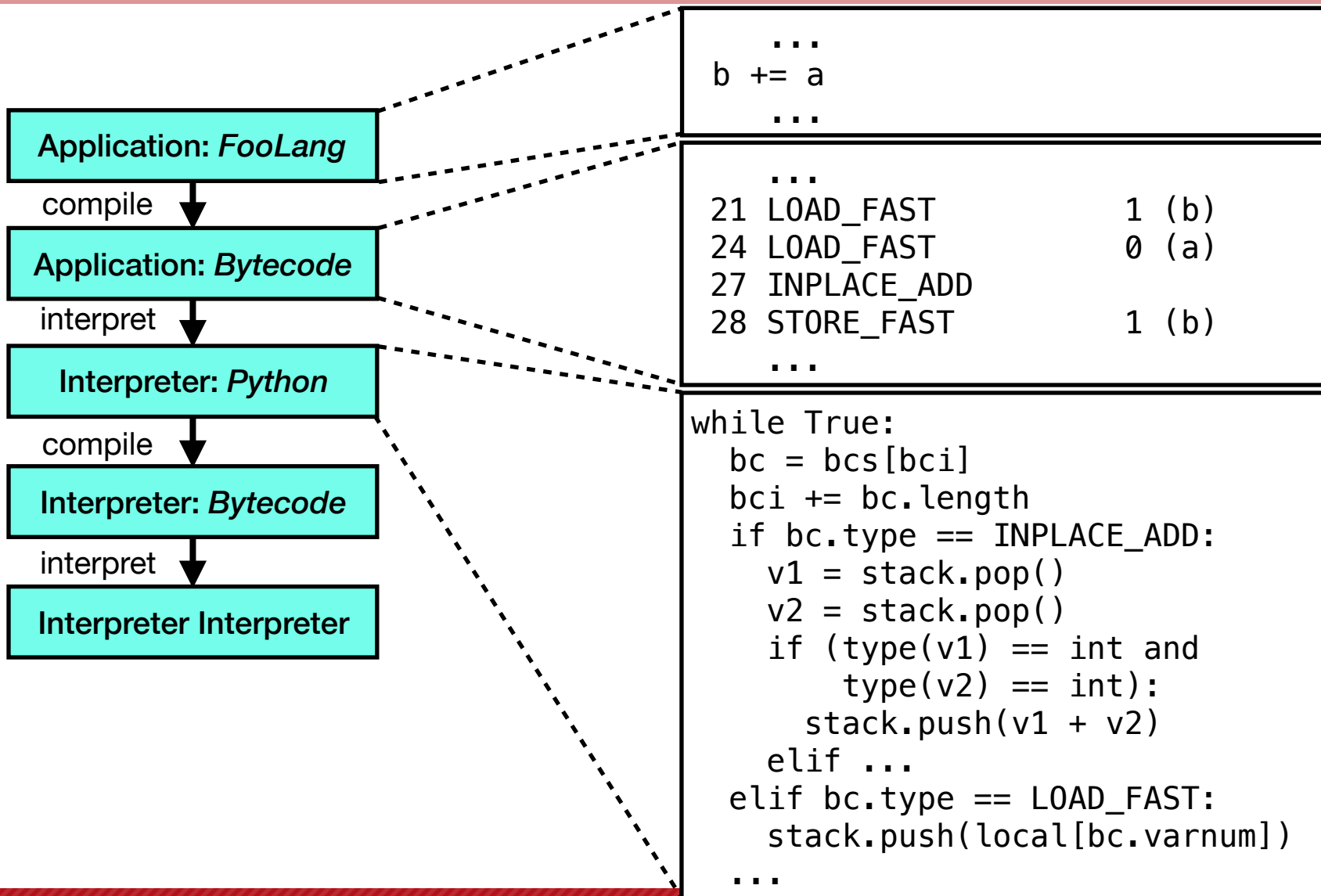
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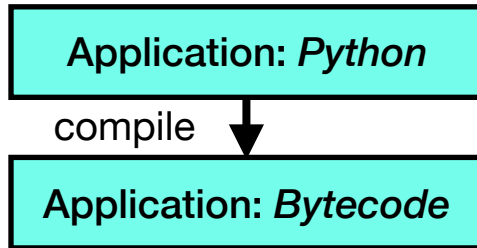
# Python-based interpreter



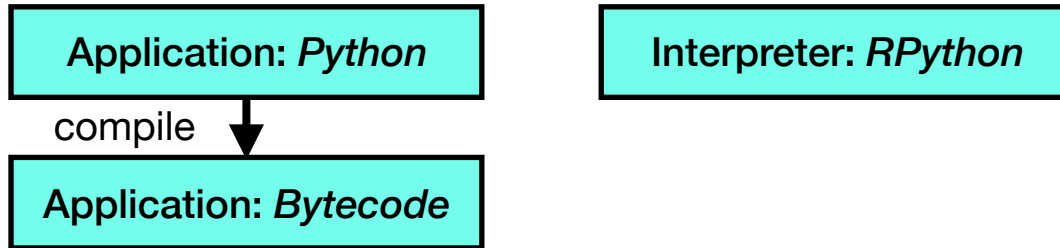
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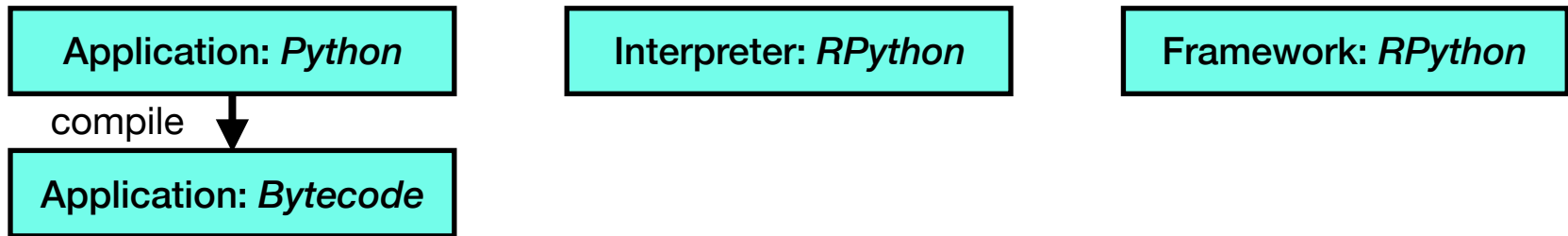
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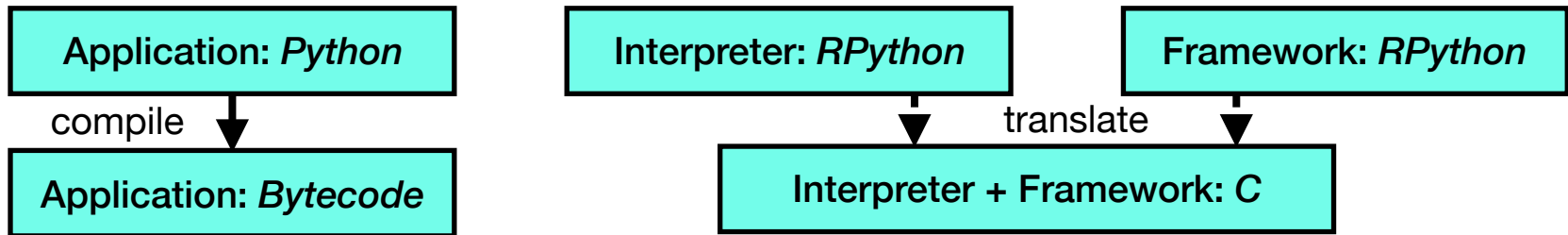


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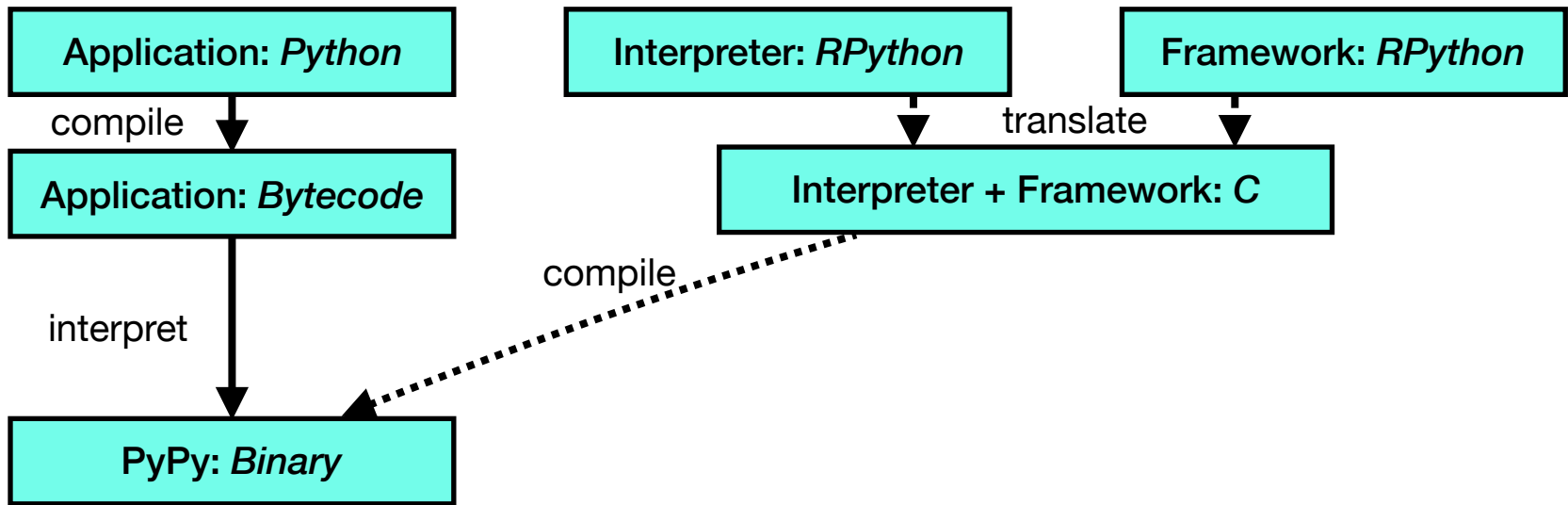




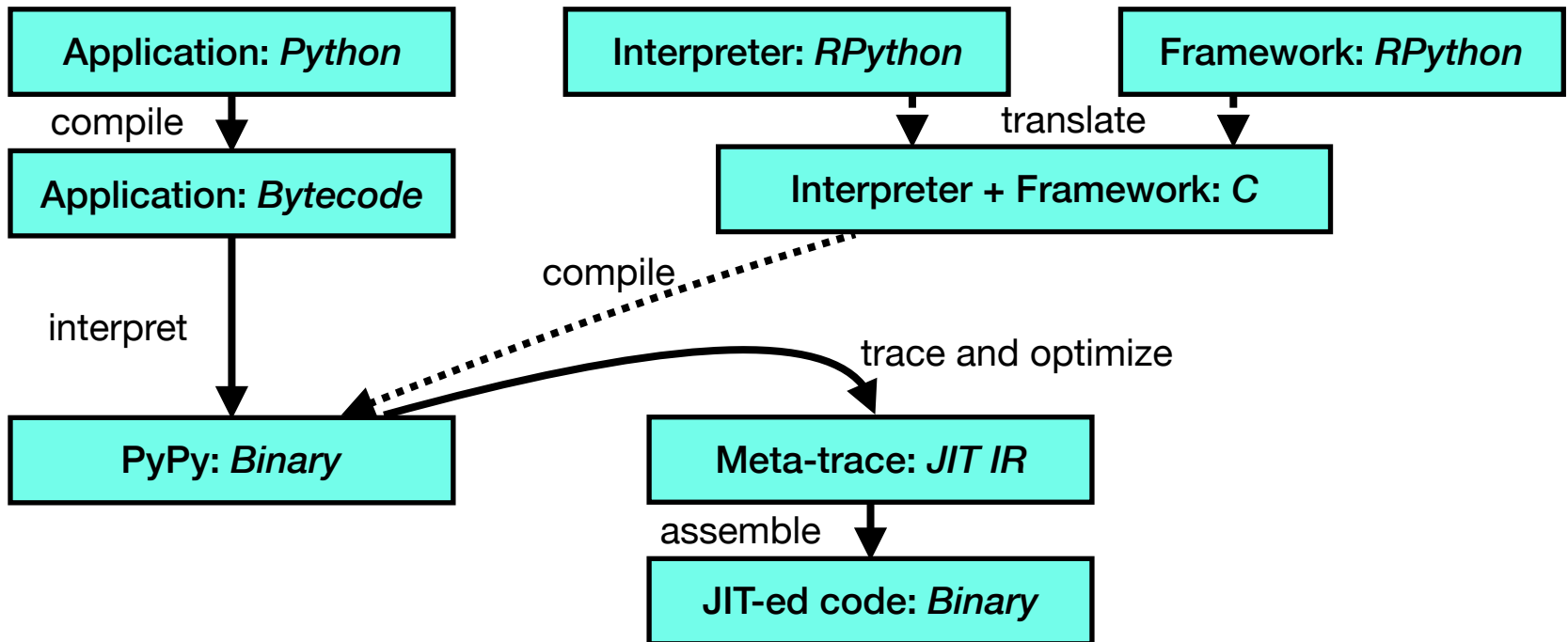
# RPython Framework



# RPython Framework



# RPython Framework



# Meta-trace

## Application bytecode

```
...  
21 LOAD_FAST          1 (b)  
24 LOAD_FAST          0 (a)  
27 INPLACE_ADD  
28 STORE_FAST        1 (b)  
...
```

## Interpreter

```
while True:  
    bc = bcs[bci]  
    bci += bc.length  
    if bc.type == INPLACE_ADD:  
        v1 = stack.pop()  
        v2 = stack.pop()  
        if (type(v1) == int and  
            type(v2) == int):  
            stack.push(v1 + v2)  
        elif ...  
    elif bc.type == LOAD_FAST:  
        stack.push(local[bc.varnum])  
    ...
```



# Meta-trace

## Application bytecode

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...  
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        stack.push(local[bc.varnum])  
    ...
```

## Meta-interpreter

## Meta-trace

```
...
```



# Meta-trace

## Application bytecode

```
...  
21 LOAD_FAST      1 (b)  
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            type(v2) == int):  
            stack.push(v1 + v2)  
        elif ...  
    elif bc.type == LOAD_FAST:  
        stack.push(local[bc.varnum])  
    ...
```

## Meta-interpreter

## Meta-trace

```
...
```



# Meta-trace

## Application bytecode

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            type(v2) == int):  
            stack.push(v1 + v2)  
        elif ...  
    elif bc.type == LOAD_FAST:  
        stack.push(local[bc.varnum])  
    ...
```

## Meta-interpreter

## Meta-trace

```
...
```



# Meta-trace

## Application bytecode

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...  
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    ...
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## Meta-interpreter

## Meta-trace

```
...  
p1 = getarrayitem(p0, 1)
```





# Meta-trace

## Application bytecode

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...  
21 LOAD_FAST          1 (b)  
24 LOAD_FAST          0 (a)  
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## Meta-trace

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```

## Meta-interpreter

## Meta-trace

```
...  
p1 = getarrayitem(p0, 1)  
p2 = getarrayitem(p0, 0)
```



# Meta-trace

## Application bytecode

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    elif bc.type == LOAD_FAST:  
        stack.push(local[bc.varnum])  
    ...
```

## Meta-interpreter

## Meta-trace

```
...  
p1 = getarrayitem(p0, 1)  
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# Meta-trace

## Application bytecode

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            stack.push(v1 + v2)  
        elif ...  
    elif bc.type == LOAD_FAST:  
        stack.push(local[bc.varnum])  
    ...
```

## Meta-interpreter

## Meta-trace

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...  
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p2 = getarrayitem(p0, 0)
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# Meta-trace

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        if (type(v1) == int and  
            type(v2) == int):  
            stack.push(v1 + v2)  
        elif ...  
    elif bc.type == LOAD_FAST:  
        stack.push(local[bc.varnum])  
    ...
```

## Meta-interpreter

## Meta-trace

```
...  
p1 = getarrayitem(p0, 1)  
p2 = getarrayitem(p0, 0)  
guard_class(p1, int)  
guard_class(p2, int)
```



# Meta-trace

## Application bytecode

```
...  
21 LOAD_FAST          1 (b)  
24 LOAD_FAST          0 (a)  
27 INPLACE_ADD  
28 STORE_FAST         1 (b)  
...
```

## Interpreter

```
while True:  
    bc = bcs[bci]  
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    if bc.type == INPLACE_ADD:  
        v1 = stack.pop()  
        v2 = stack.pop()  
        if (type(v1) == int and  
            type(v2) == int):  
            stack.push(v1 + v2)  
        elif ...  
    elif bc.type == LOAD_FAST:  
        stack.push(local[bc.varnum])  
    ...
```

## Meta-interpreter

## Meta-trace

```
...  
p1 = getarrayitem(p0, 1)  
p2 = getarrayitem(p0, 0)  
guard_class(p1, int)  
guard_class(p2, int)  
i3 = getfield(p1, intval)  
i4 = getfield(p2, intval)
```



# Meta-trace

## Application bytecode

```
...  
21 LOAD_FAST          1 (b)  
24 LOAD_FAST          0 (a)  
27 INPLACE_ADD  
28 STORE_FAST        1 (b)  
...
```

## Interpreter

```
while True:  
    bc = bcs[bci]  
    bci += bc.length  
    if bc.type == INPLACE_ADD:  
        v1 = stack.pop()  
        v2 = stack.pop()  
        if (type(v1) == int and  
            type(v2) == int):  
            stack.push(v1 + v2)  
        elif ...  
    elif bc.type == LOAD_FAST:  
        stack.push(local[bc.varnum])  
    ...
```

## Meta-interpreter

## Meta-trace

```
...  
p1 = getarrayitem(p0, 1)  
p2 = getarrayitem(p0, 0)  
guard_class(p1, int)  
guard_class(p2, int)  
i3 = getfield(p1, intval)  
i4 = getfield(p2, intval)  
i5 = int_add_ovf(i3, i4)  
guard_no_overflow()  
...
```



# Meta-trace

## Application bytecode

```
...  
21 LOAD_FAST          1 (b)  
24 LOAD_FAST          0 (a)  
27 INPLACE_ADD  
28 STORE_FAST         1 (b)  
...
```

## Interpreter

```
while True:  
    bc = bcs[bci]  
    bci += bc.length  
    if bc.type == INPLACE_ADD:  
        v1 = stack.pop()  
        v2 = stack.pop()  
        if (type(v1) ==  
            type(v2) ==  
                stack.push(v1 + v2)  
        elif ...  
    elif bc.type == LOAD_FAST:  
        stack.push(local[bc.varnum])  
    ...
```

## Meta-interpreter

## Meta-trace

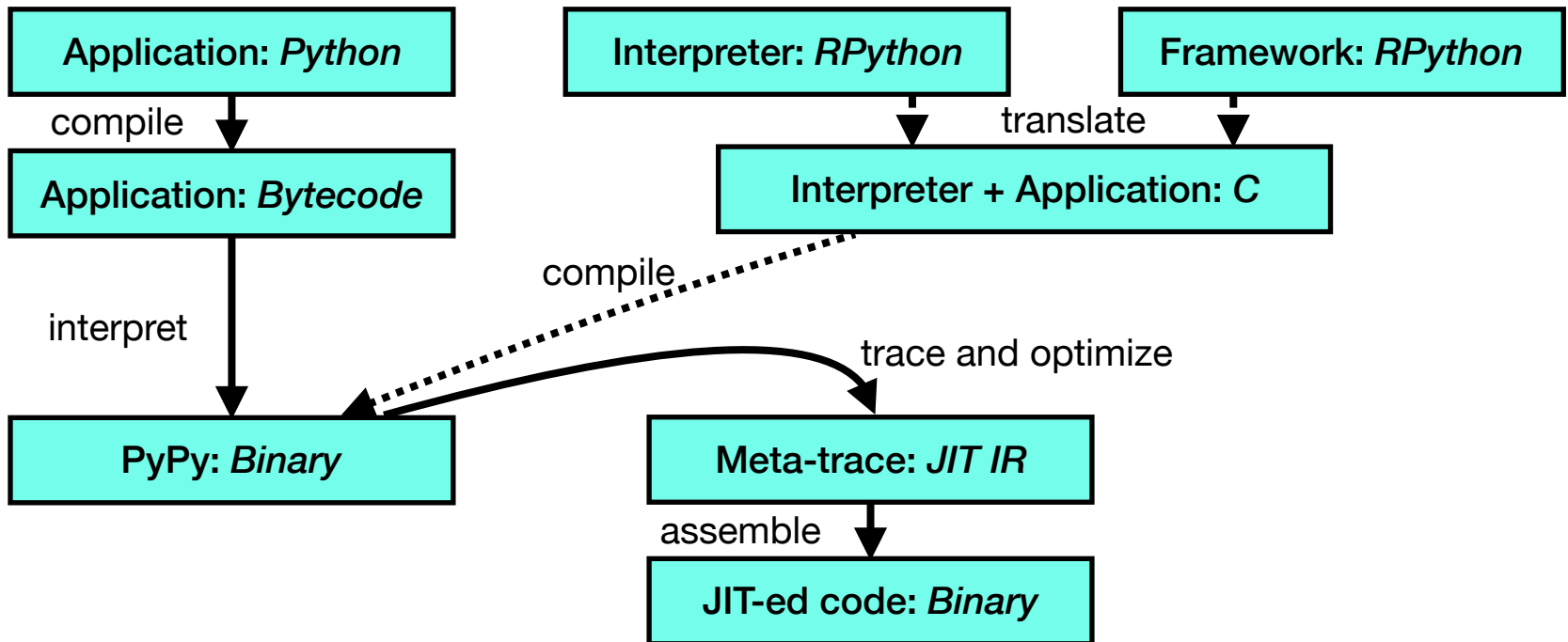
```
...  
p1 = getarrayitem(p0, 1)  
p2 = getarrayitem(p0, 0)  
guard_class(p1, int)  
  
i4 = getfield(p2, intval)  
i5 = int_add_ovf(i3, i4)  
guard_no_overflow()  
...
```

**Deoptimization back to interpreter on guard failure**



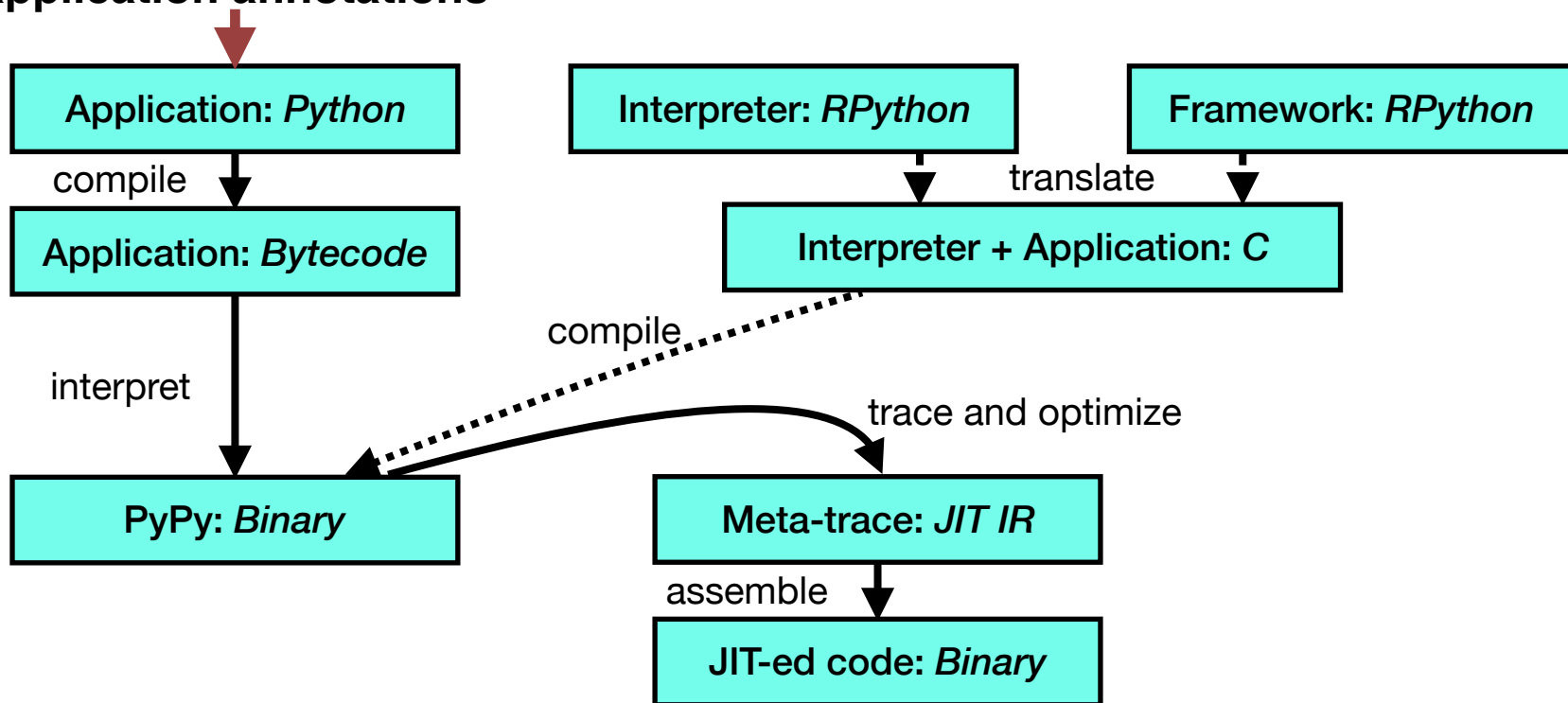


# Cross-layer annotations

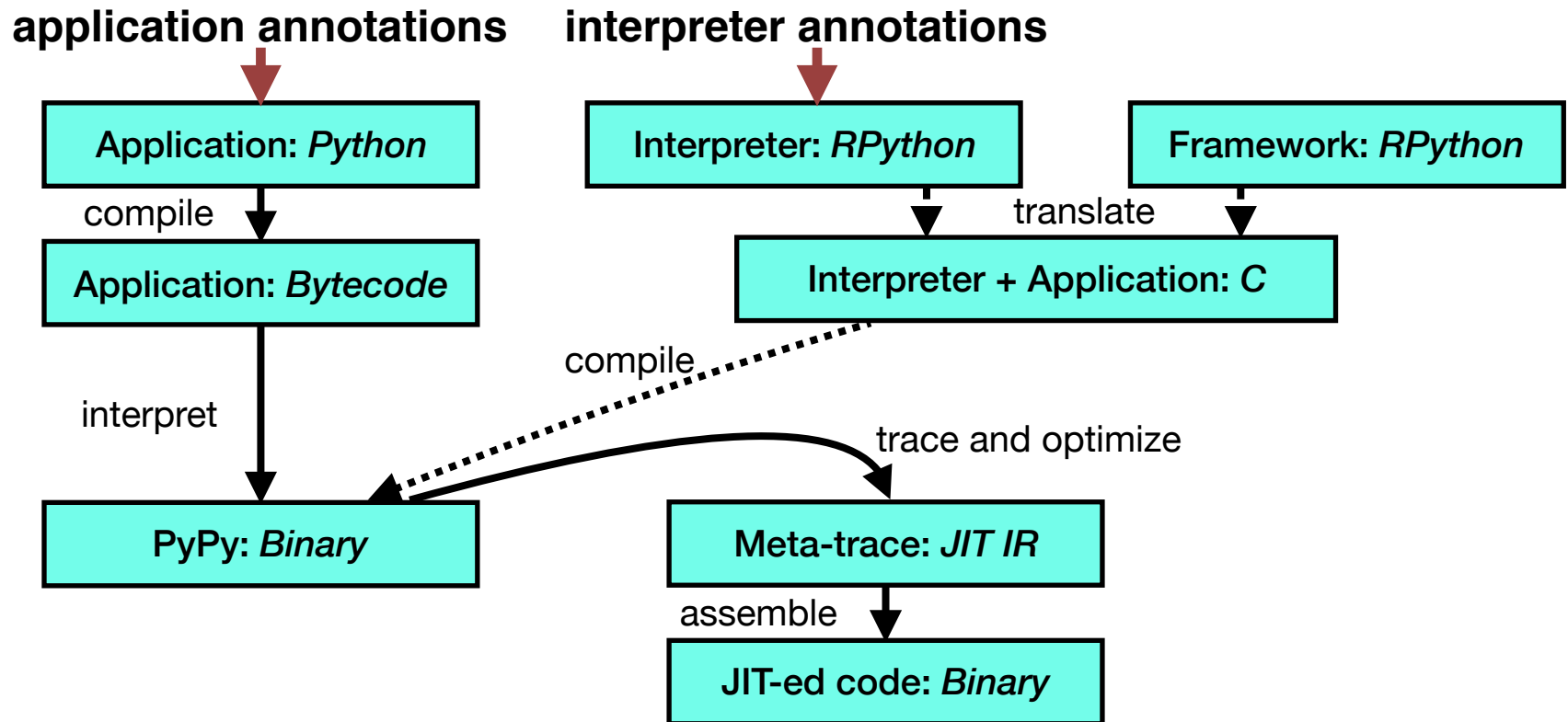


# Cross-layer annotations

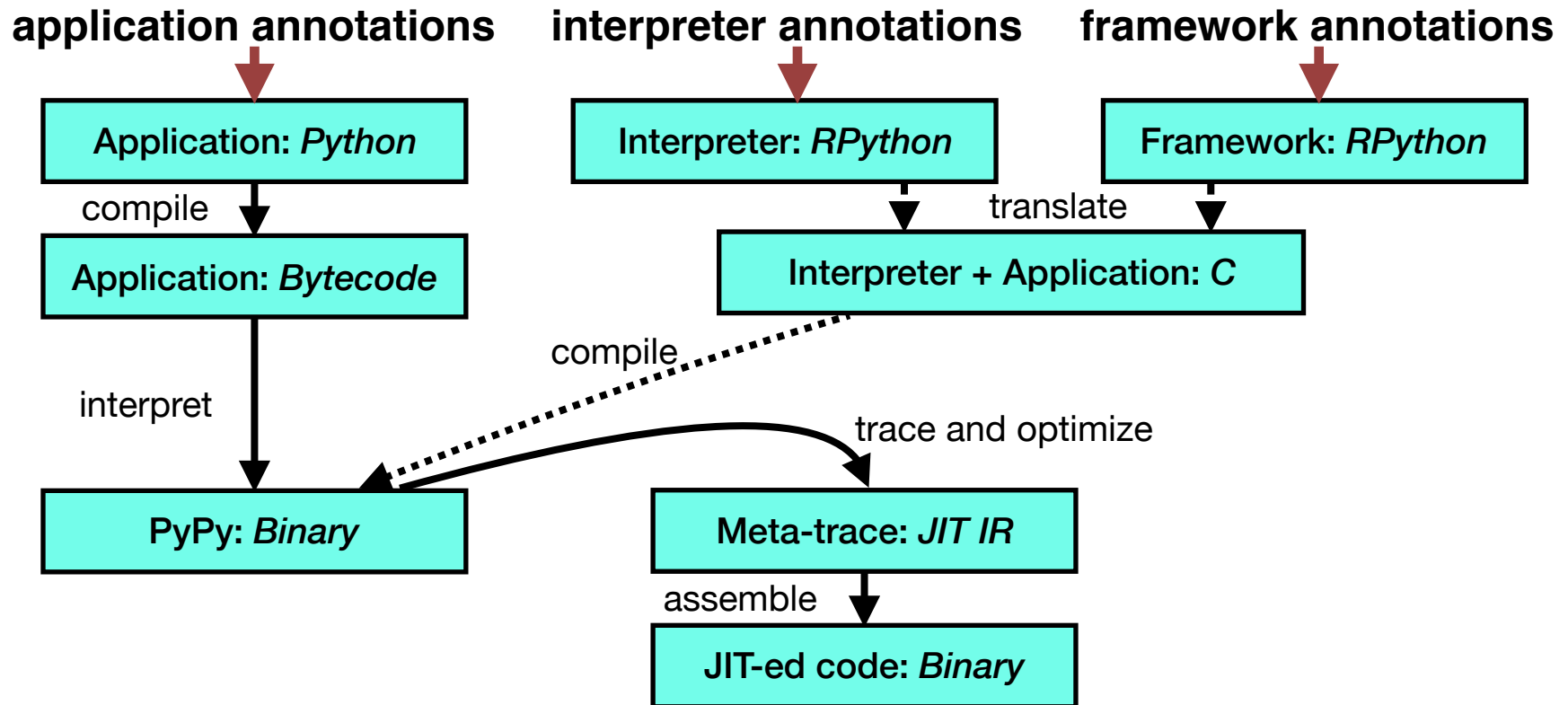
application annotations



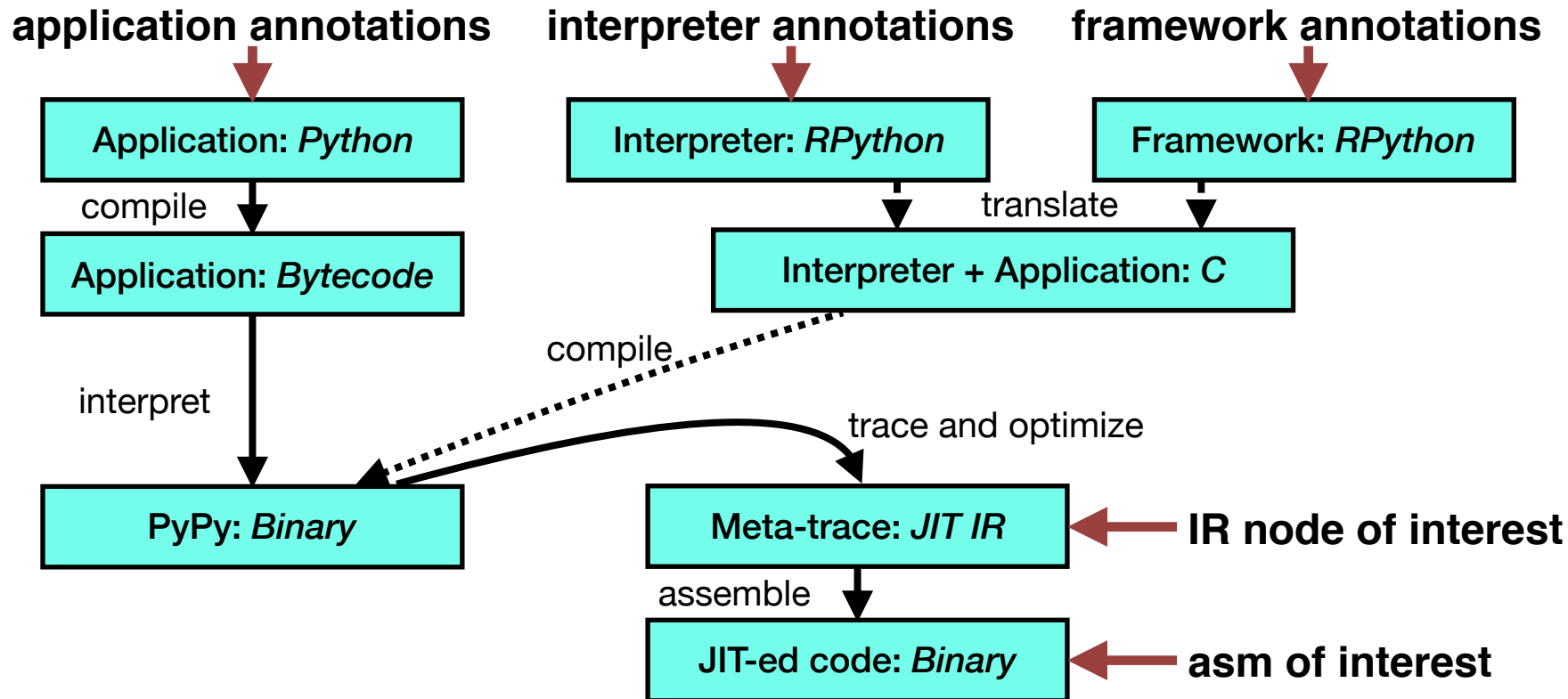
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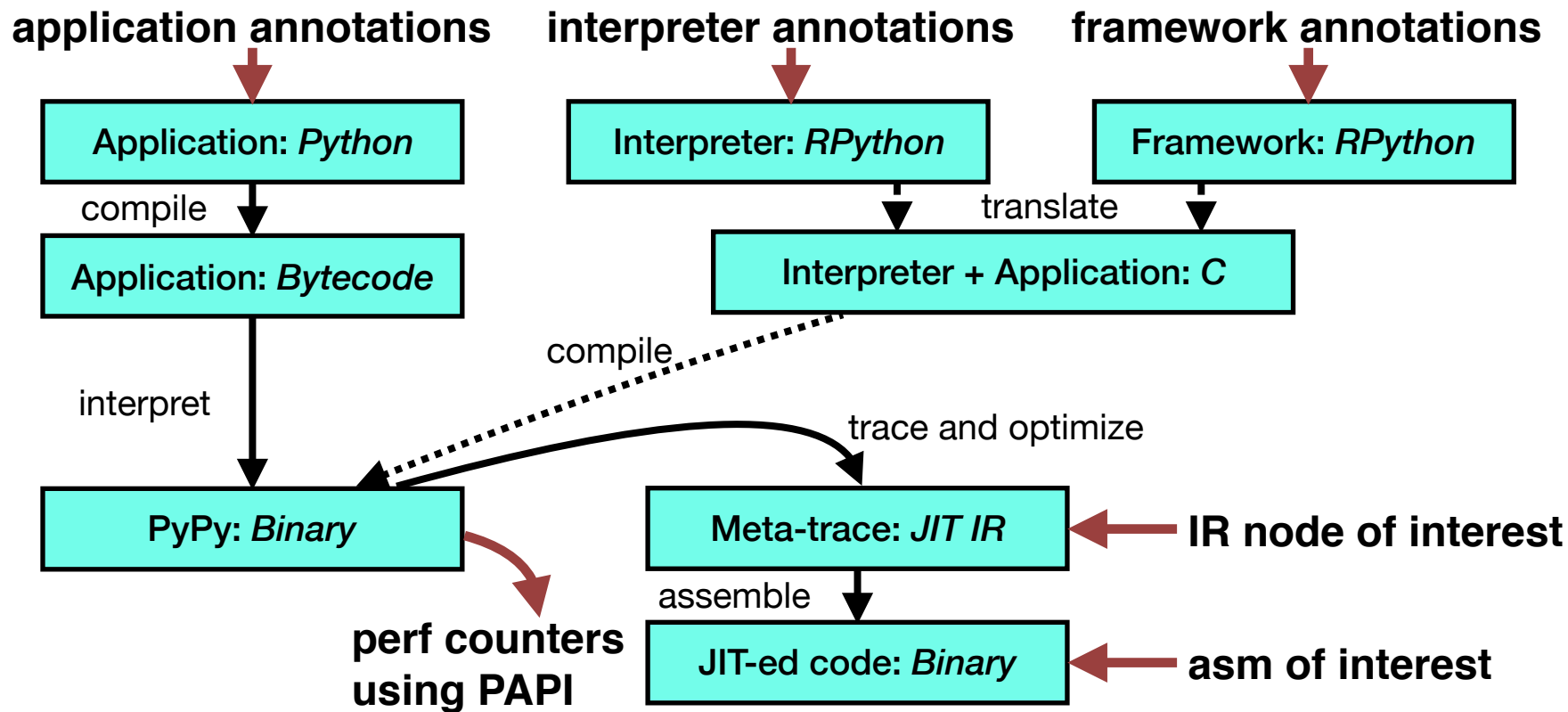
# Cross-layer annotations



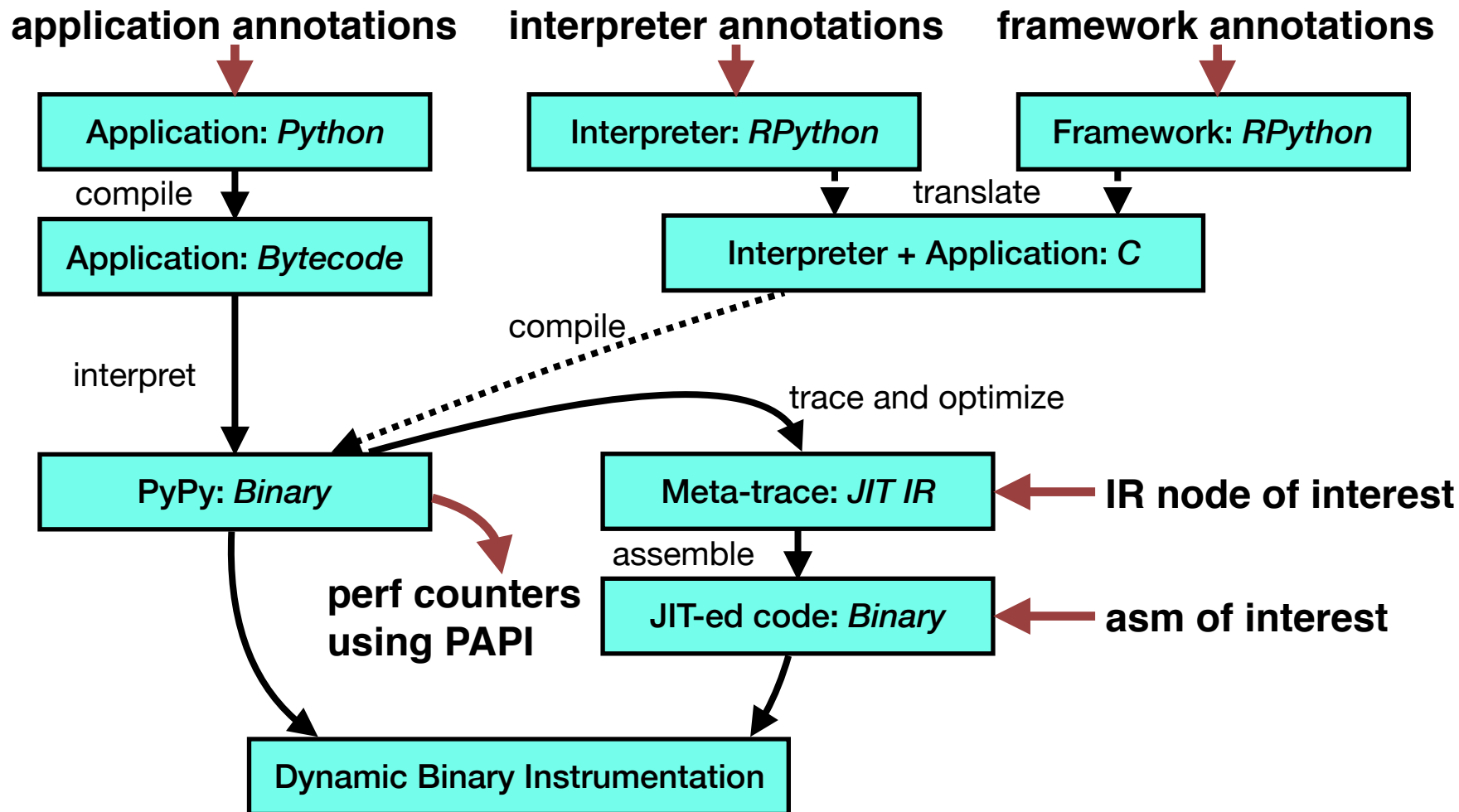
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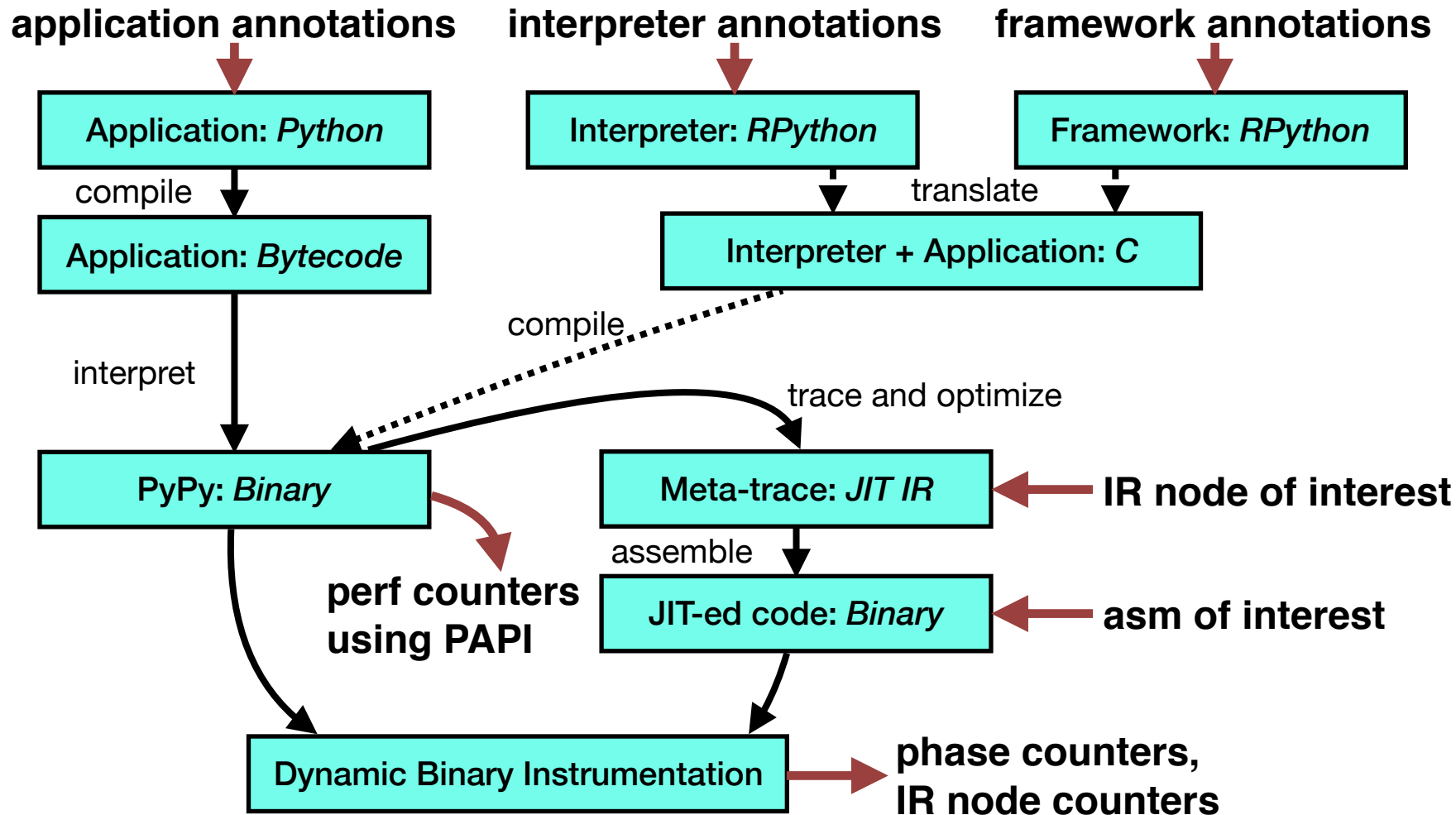
# Cross-layer annotations



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# Cross-layer annotations





# Cross-layer workload characterization of meta-tracing JIT VMs

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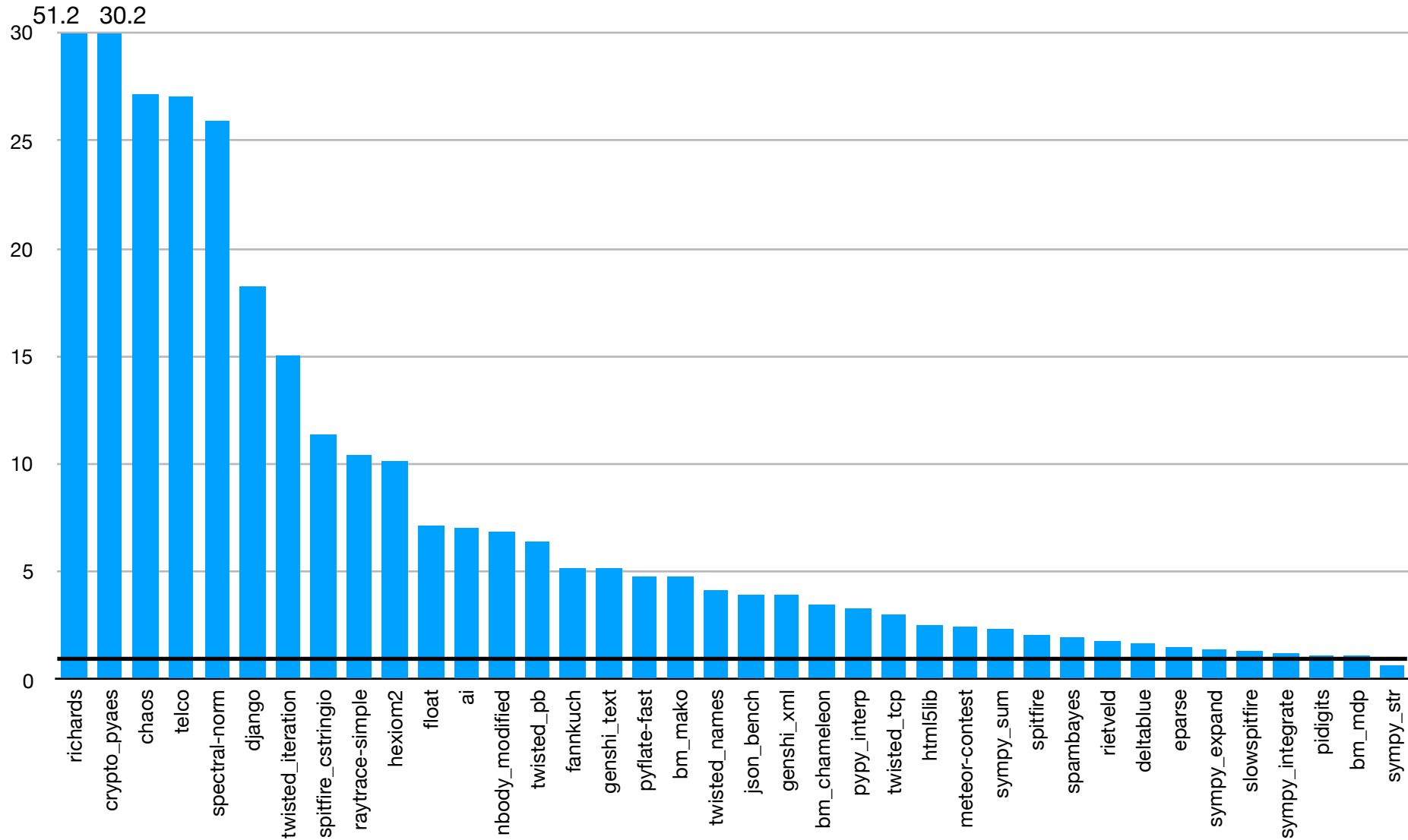
- How can meta-tracing JITs significantly improve the performance of multiple dynamic languages?

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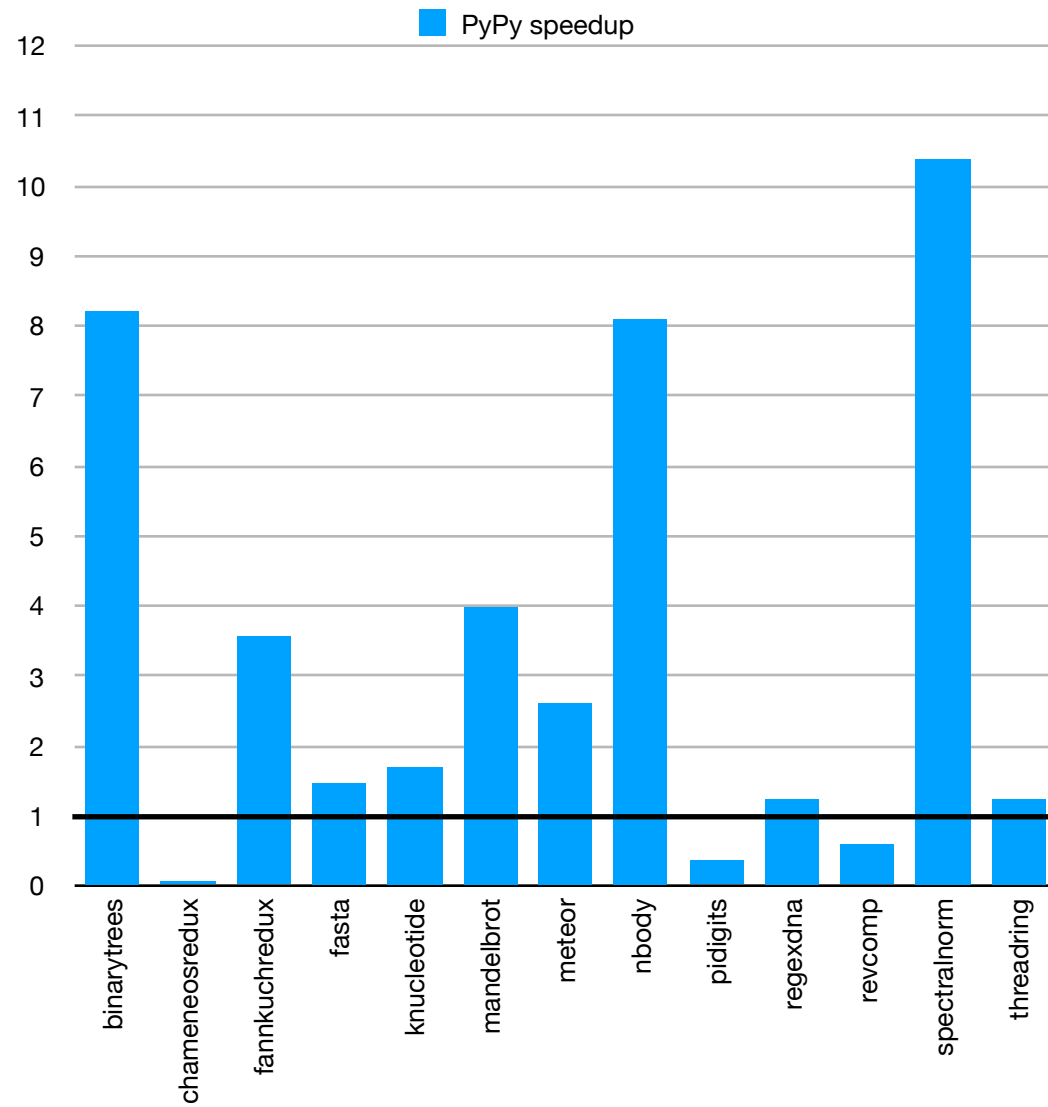
- Why are meta-tracing JITs for dynamic programming still slower than C?



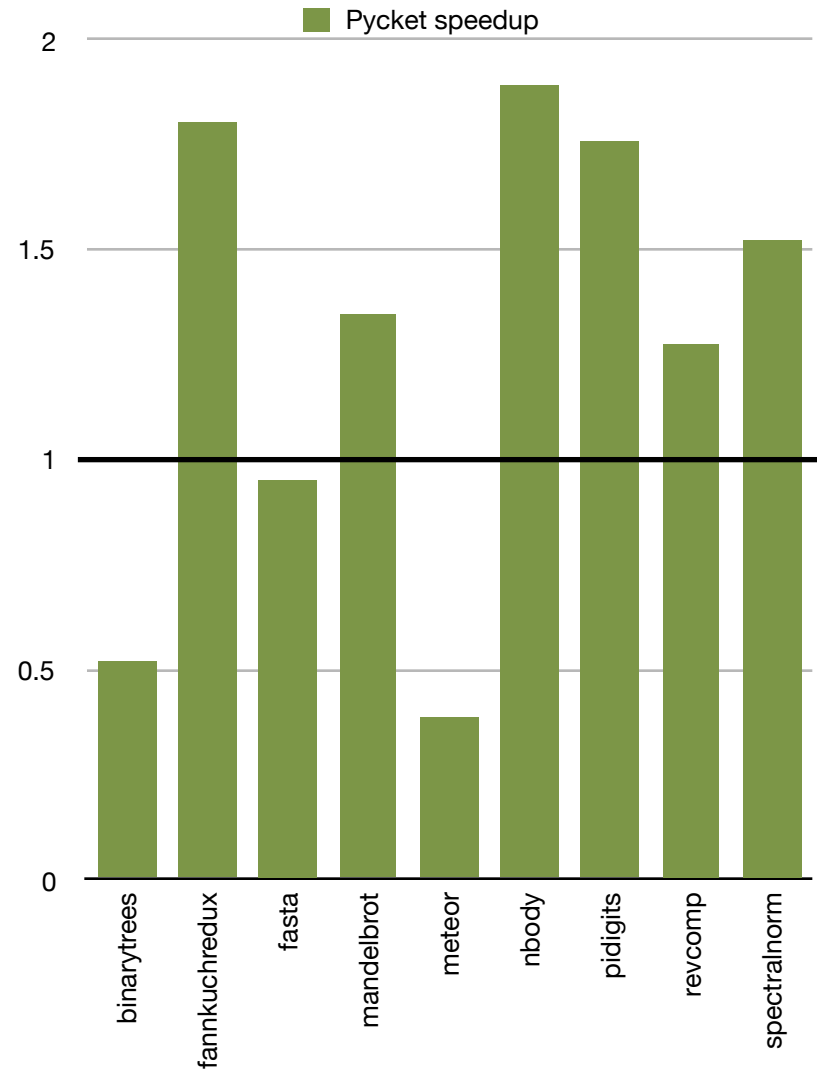
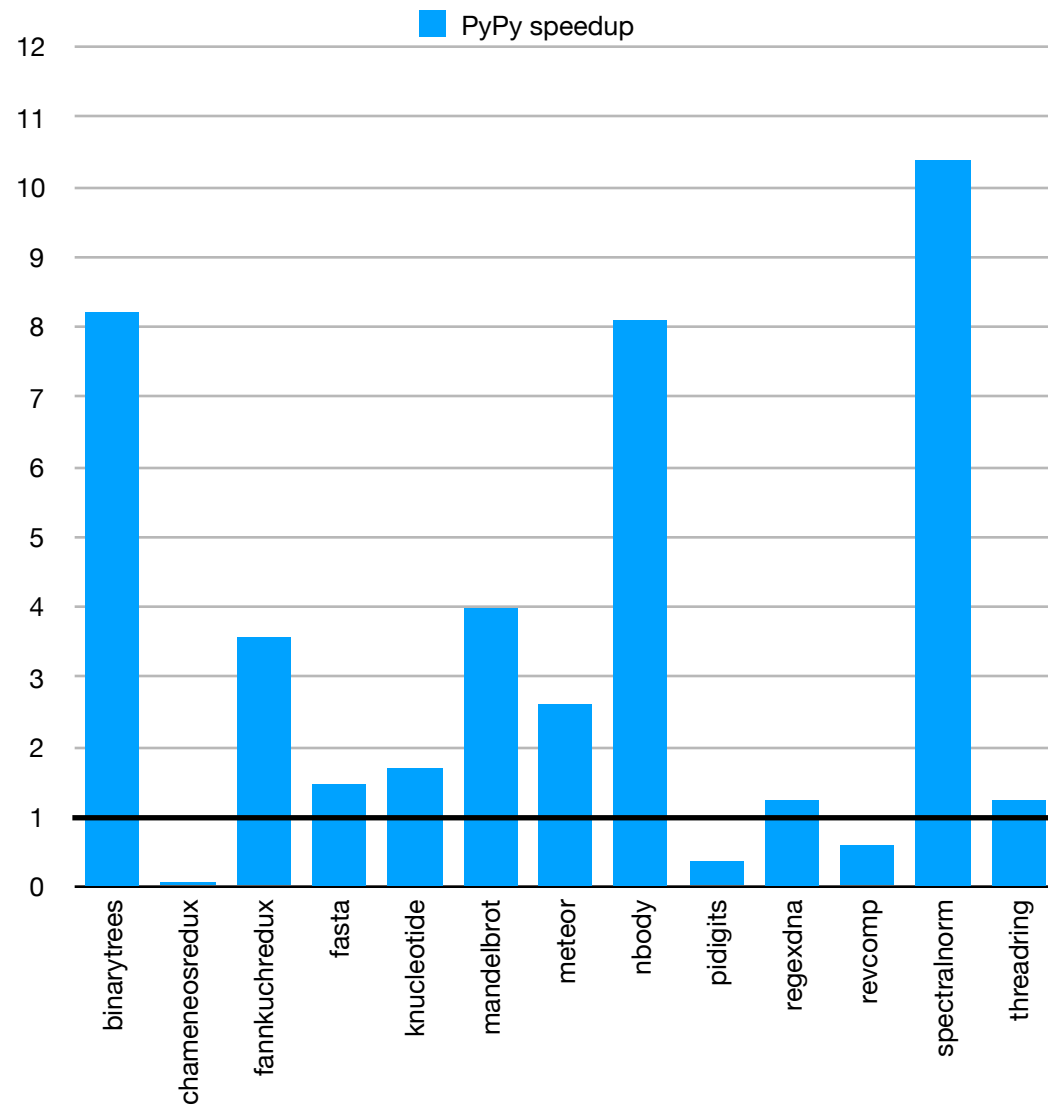
# PyPy with meta-tracing JIT speedup over CPython: Meta-tracing JIT improves the performance significantly



# PyPy speedup over CPython and Pycket speedup over Racket: Meta-tracing JIT improves performance significantly across multiple languages

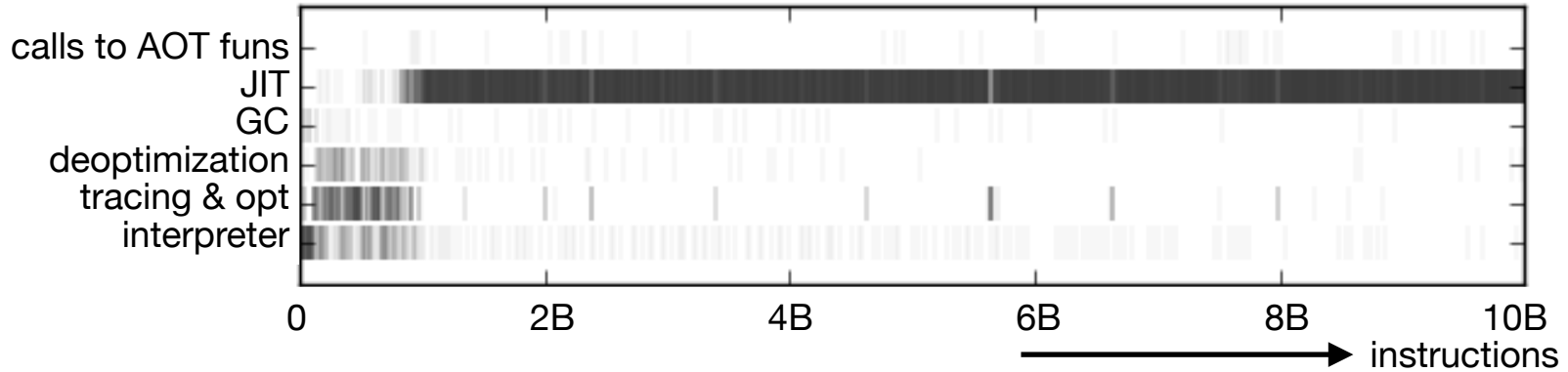


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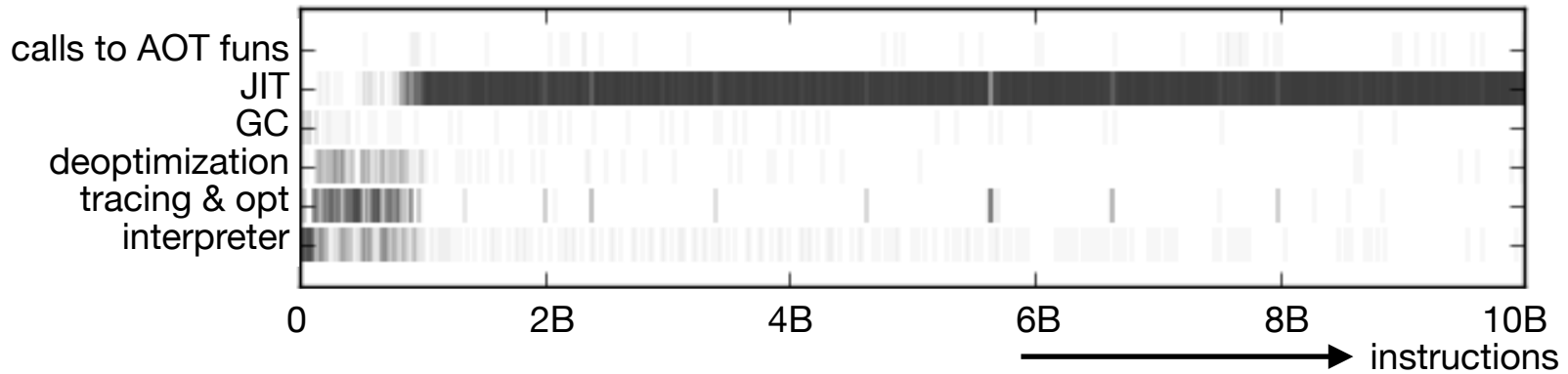
# Meta-tracing JIT VM phases

richards

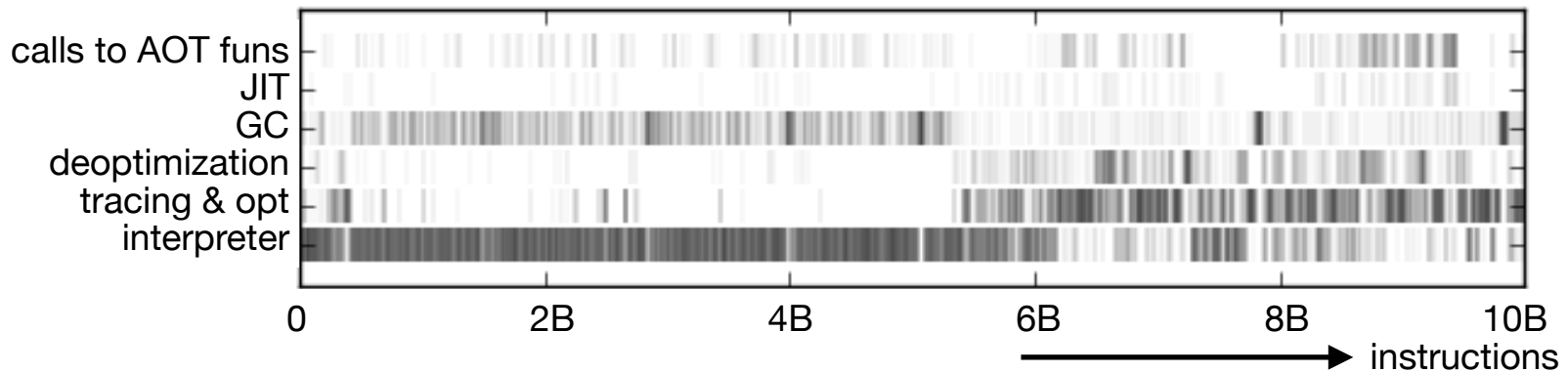


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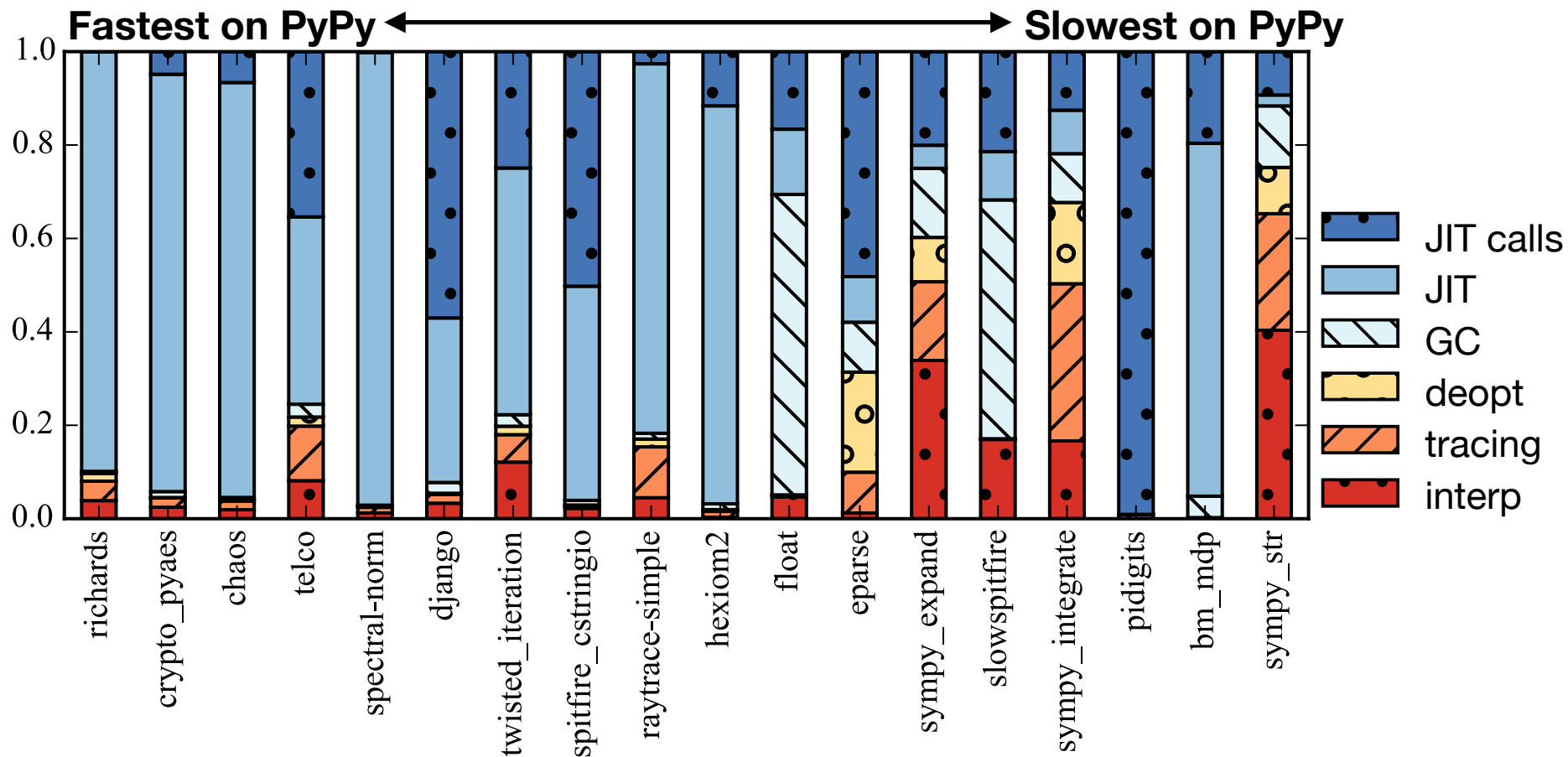
richards



sympy\_str



# Meta-tracing JIT VM phases







# Meta-tracing inlines all loops and can hurt performance

## Interpreter

```
while True:
    ...
    memcpy(d, s, n)
    ...

def memcpy(dest, src, n):
    i = 0
    while i < n:
        dest[i] = src[i]
        i += 1
```



# Meta-tracing inlines all loops and can hurt performance

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Meta-**interpreter**

Meta-**trace**

...



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## Meta-interpreter

## Meta-trace

```
...
guard_gt(i0, 0)
i3 = getarrayitem(p1, 0)
setarrayitem(p2, 0, i3)
```



# Meta-tracing inlines all loops and can hurt performance

## Interpreter

```
while True:
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## Meta-interpreter



## Meta-trace

```
...
guard_gt(i0, 0)
i3 = getarrayitem(p1, 0)
setarrayitem(p2, 0, i3)
guard_gt(i0, 1)
i4 = getarrayitem(p1, 1)
setarrayitem(p2, 1, i4)
```



# Meta-tracing inlines all loops and can hurt performance

## Interpreter

```
while True:
    ...
    memcpy(d, s, n)
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i4 = getarrayitem(p1, 1)
setarrayitem(p2, 1, i4)
guard_gt(i0, 2)
i5 = getarrayitem(p1, 2)
setarrayitem(p2, 2, i5)
...
```

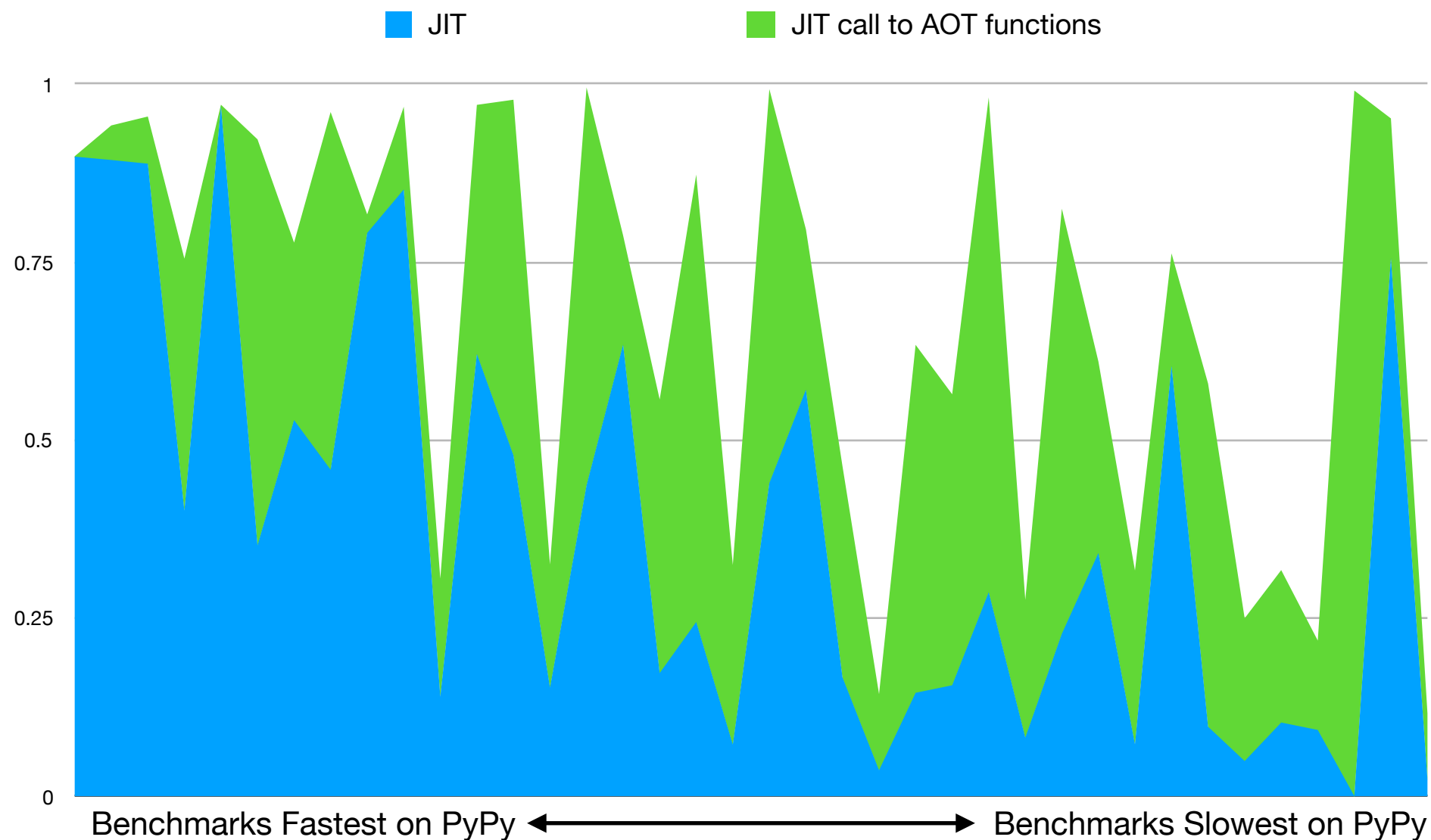


# Examples of significant AOT-compiled functions

Benchmark	%	Source	Function
ai	19.4	interpreter	<code>setobject.get_storage_from_list</code>
bm_chameleon	17.9	RPython types	<code>rorderdict.ll_call_lookup_function</code>
bm_mako	26.1	RPython lib	<code>runicode.unicode_encode_ucs1_helper</code>
json_bench	18.5	PyPy module	<code>_pypyjson.raw_encode_basestring_ascii</code>
nbody_modified	44.6	external lib	<code>pow</code>



# JIT calls to AOT-compiled functions: AOT-compiled functions can improve performance by avoiding long traces



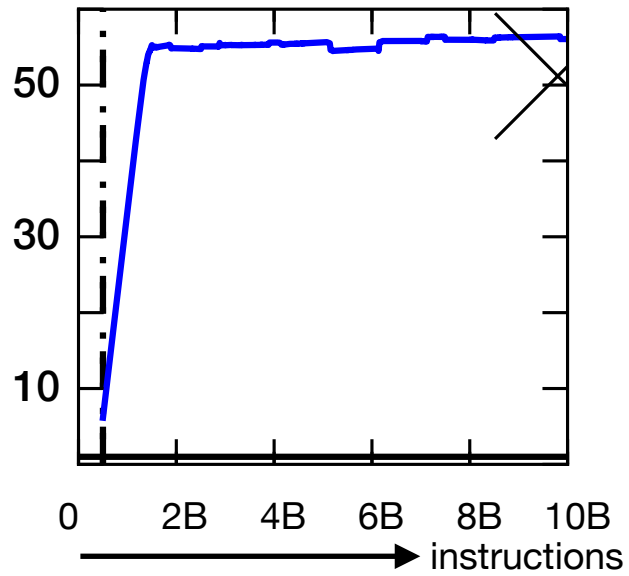
# PyPy bytecode execution rate compared to CPython: Benchmarks that perform the best also warm up the fastest





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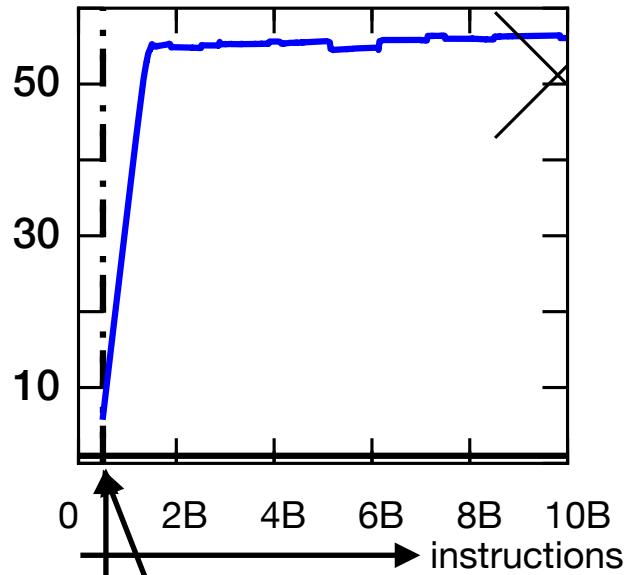
richards



# PyPy bytecode execution rate compared to CPython: Benchmarks that perform the best also warm up the fastest

**Breakeven point: the performance of the two VMs at this point is equal**

**richards**



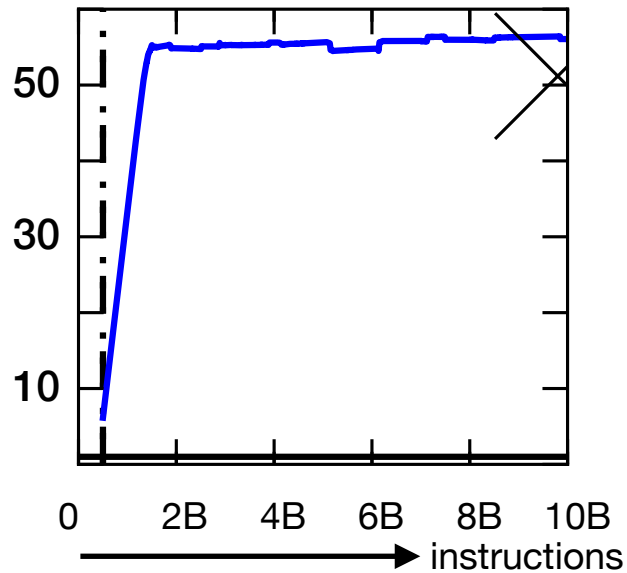
**PyPy w/o JIT breakeven point**  
**CPython breakeven point**



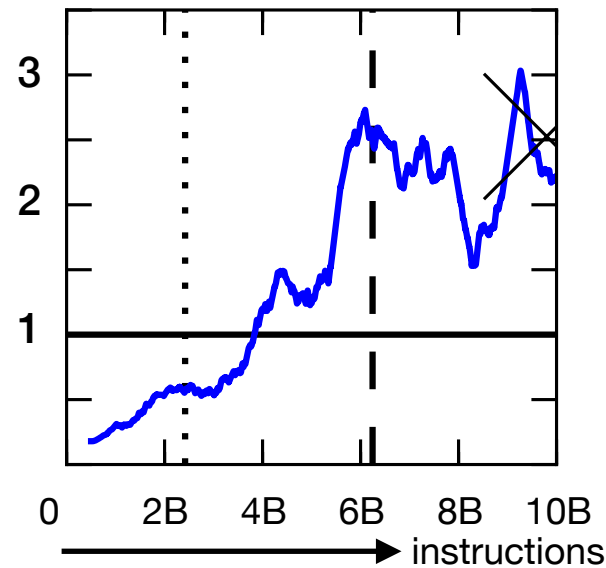
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richards



html5lib

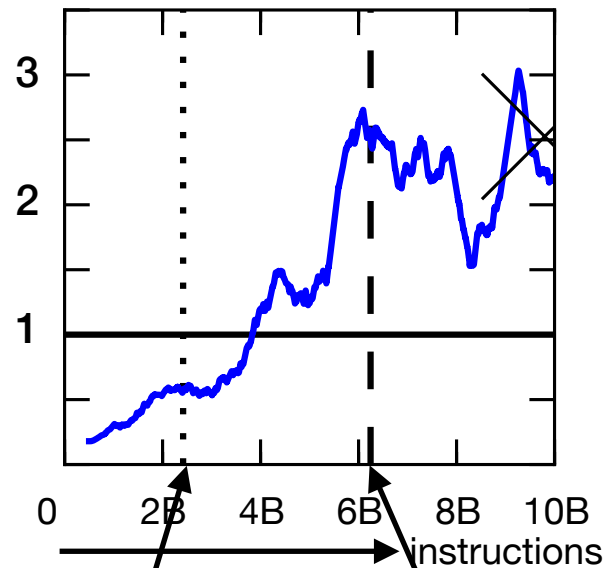
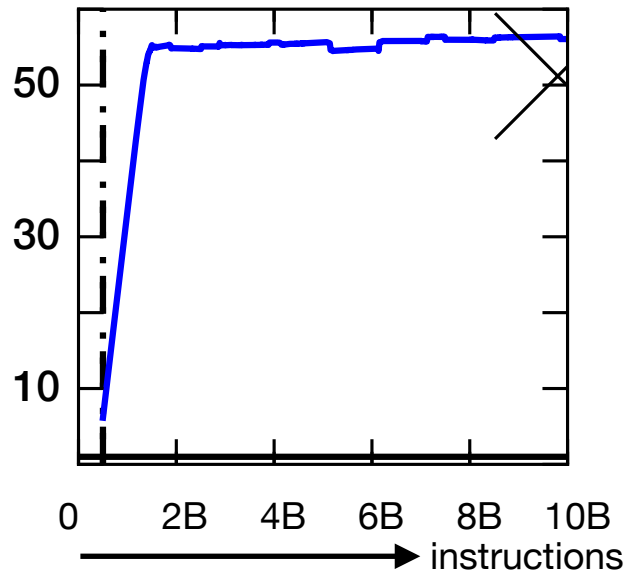


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richards

html5lib



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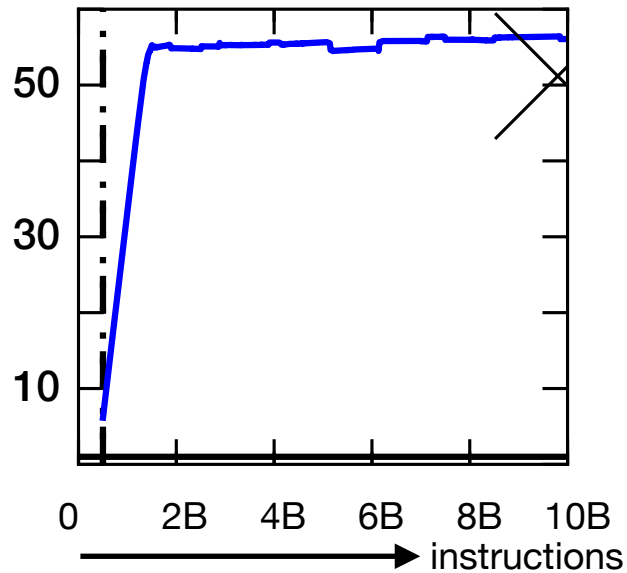
CPython breakeven point



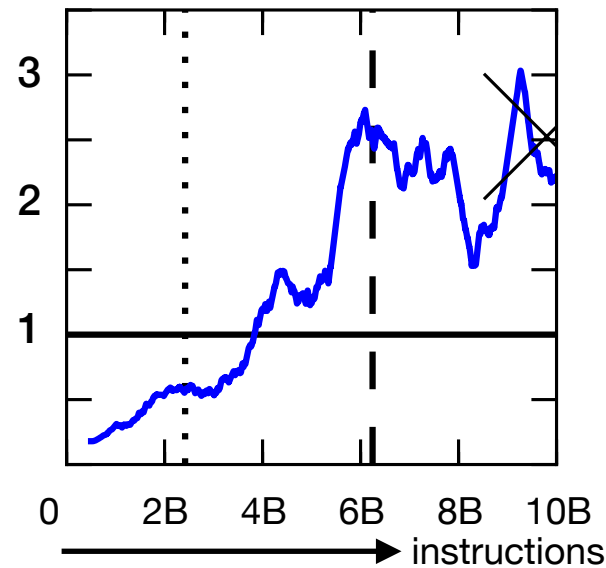
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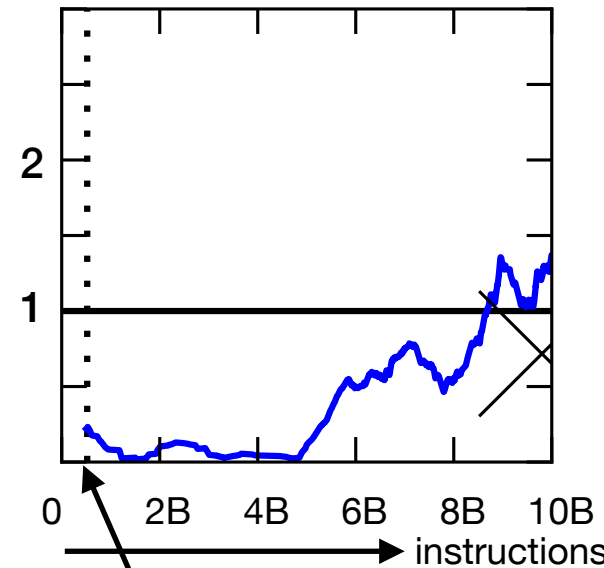
richards



html5lib



sympy\_str



PyPy w/o JIT breakeven point



# Cross-layer workload characterization of meta-tracing JIT VMs

## PyPy >> CPython

- How can meta-tracing JITs significantly improve the performance of multiple dynamic languages?

## PyPy << C

- Why are meta-tracing JITs for dynamic programming still slower than C?



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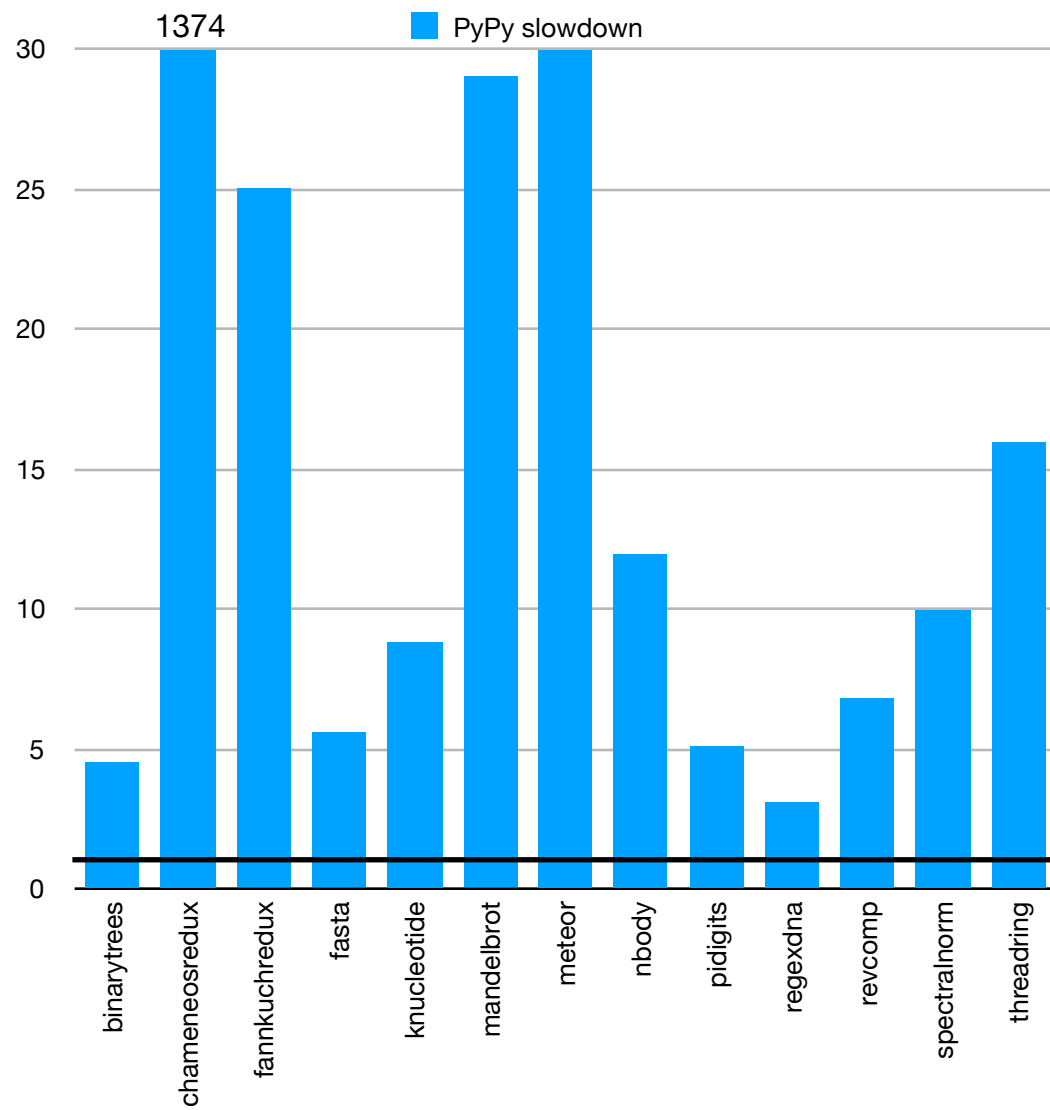
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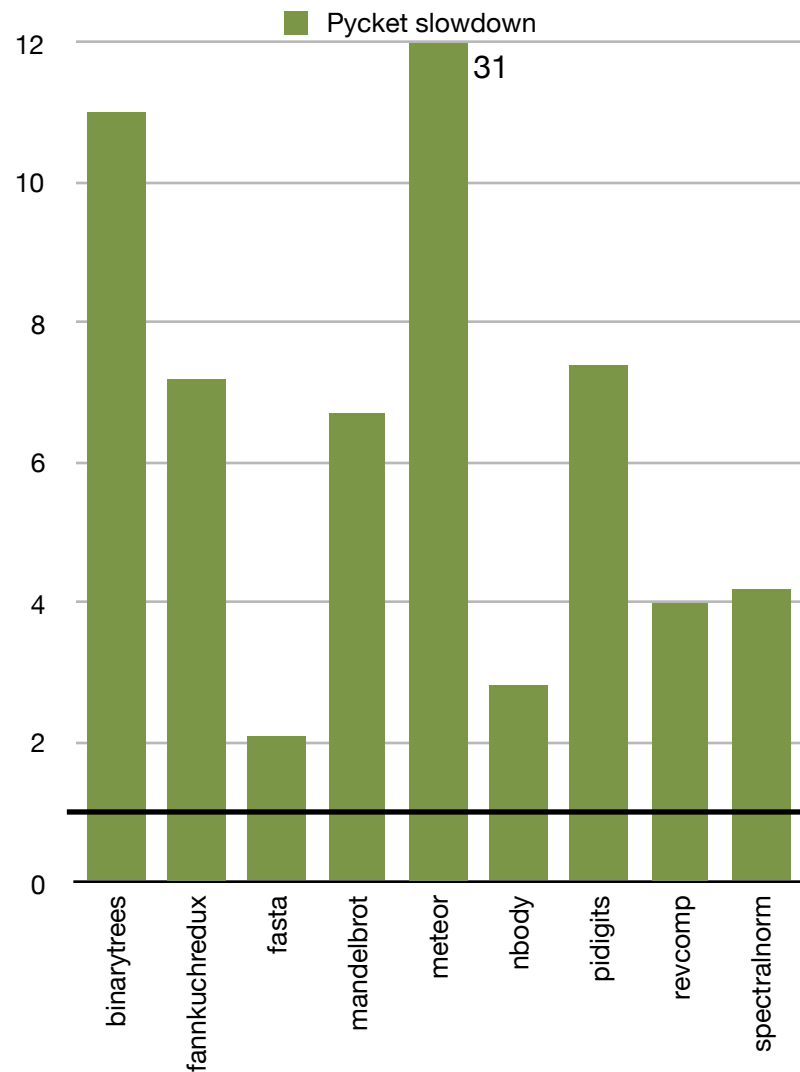
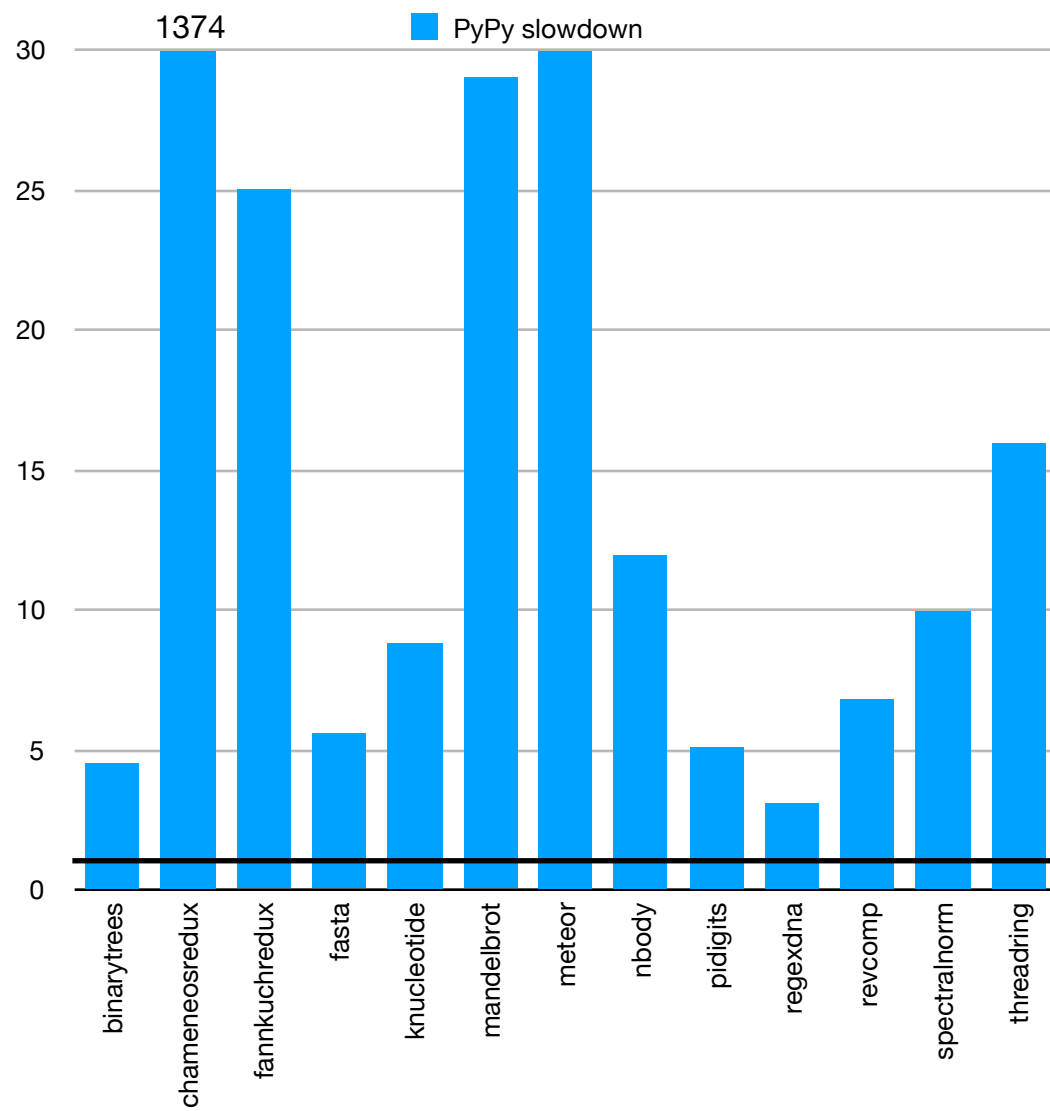
# PyPy and Pycket slowdown over C/C++: Meta-tracing JIT has a big performance gap between static languages



31

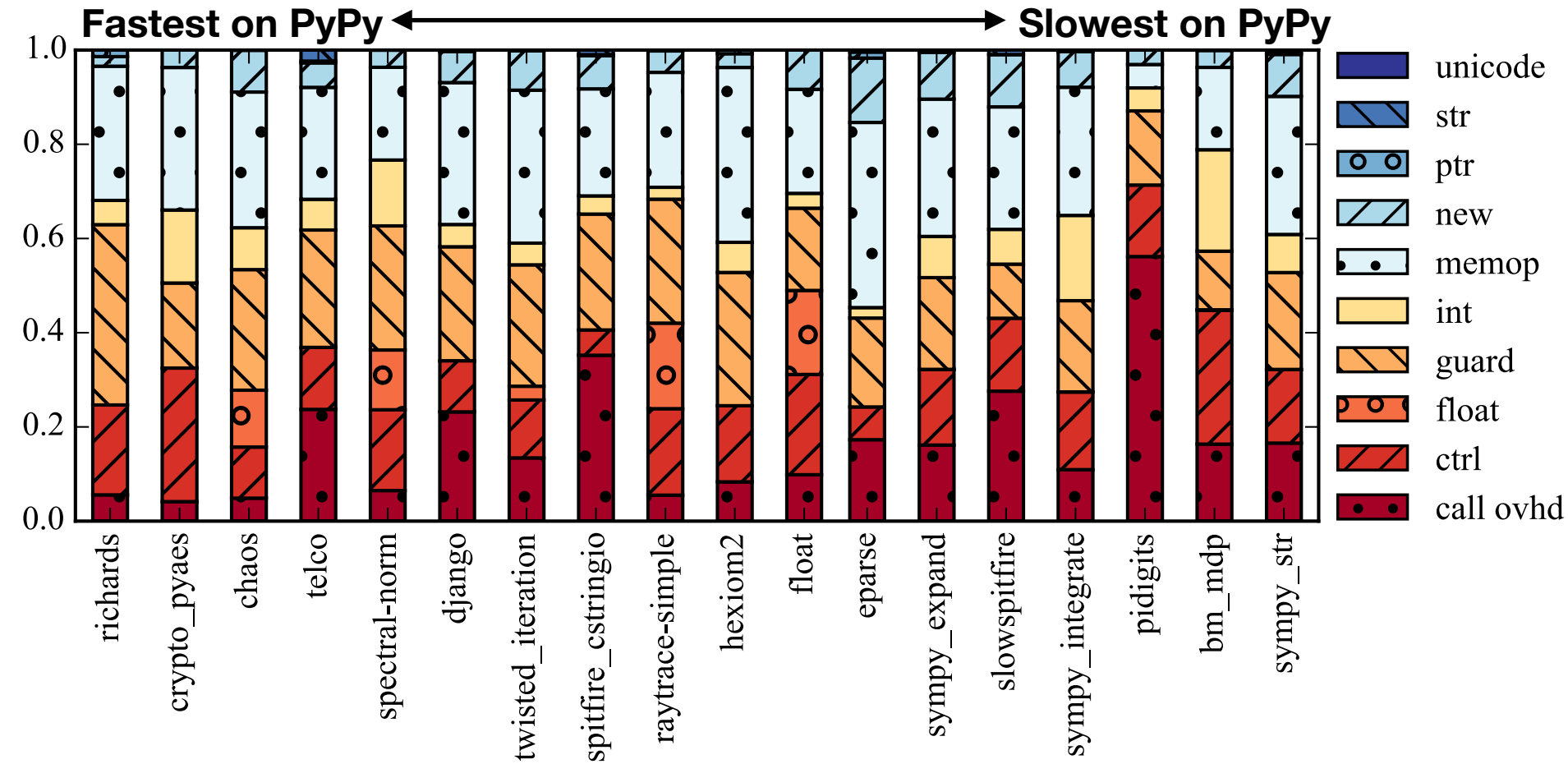


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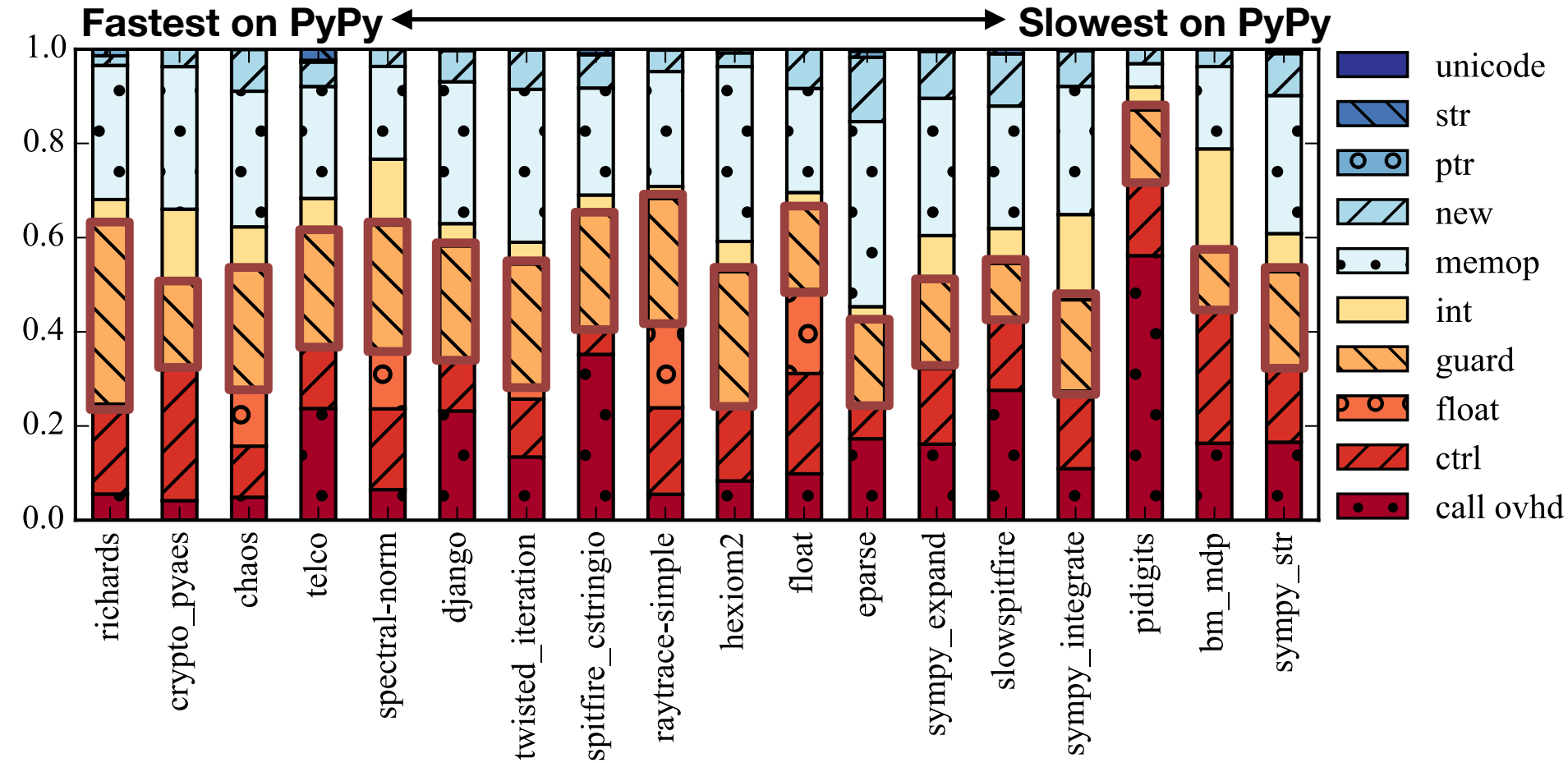




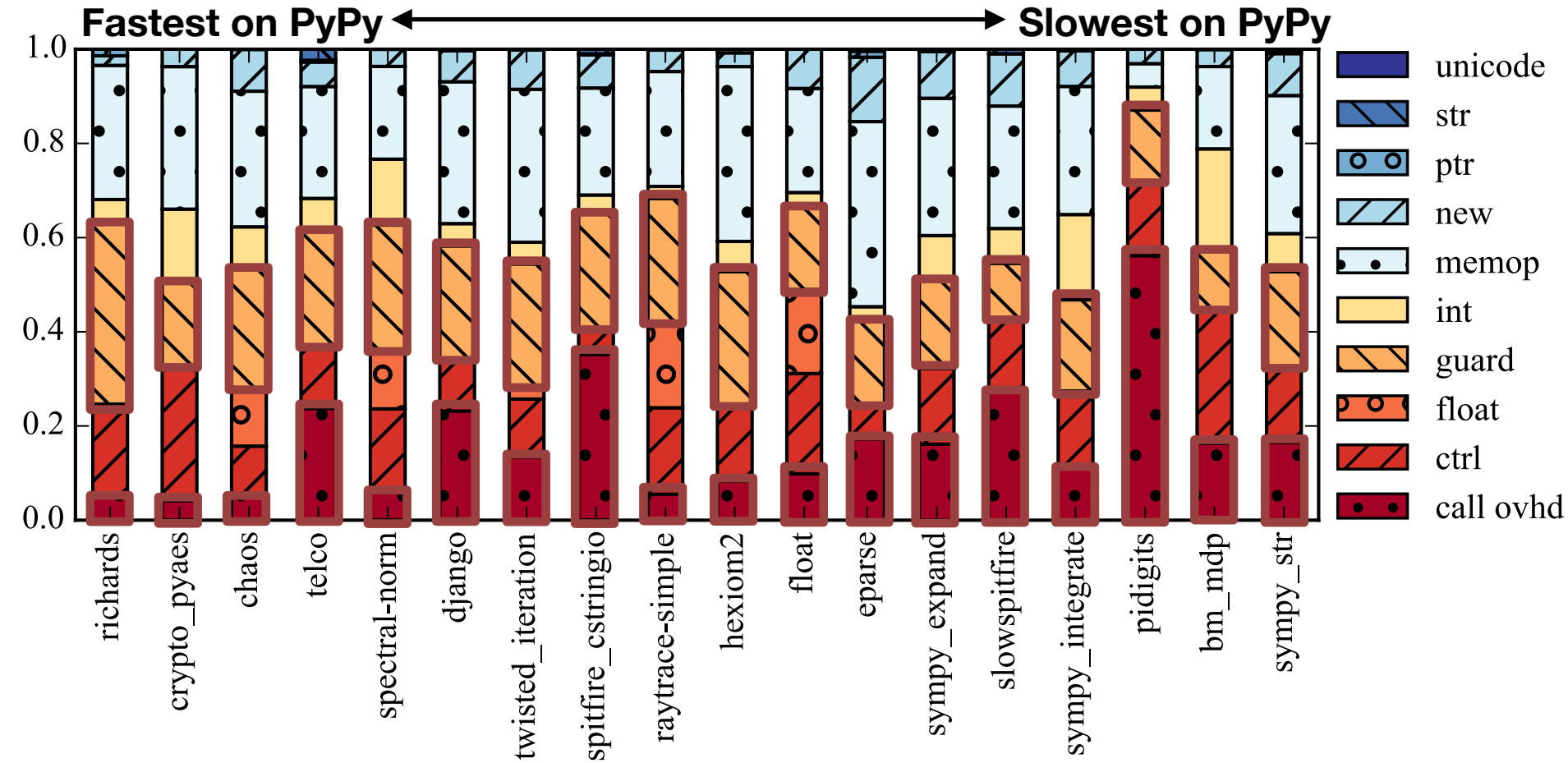
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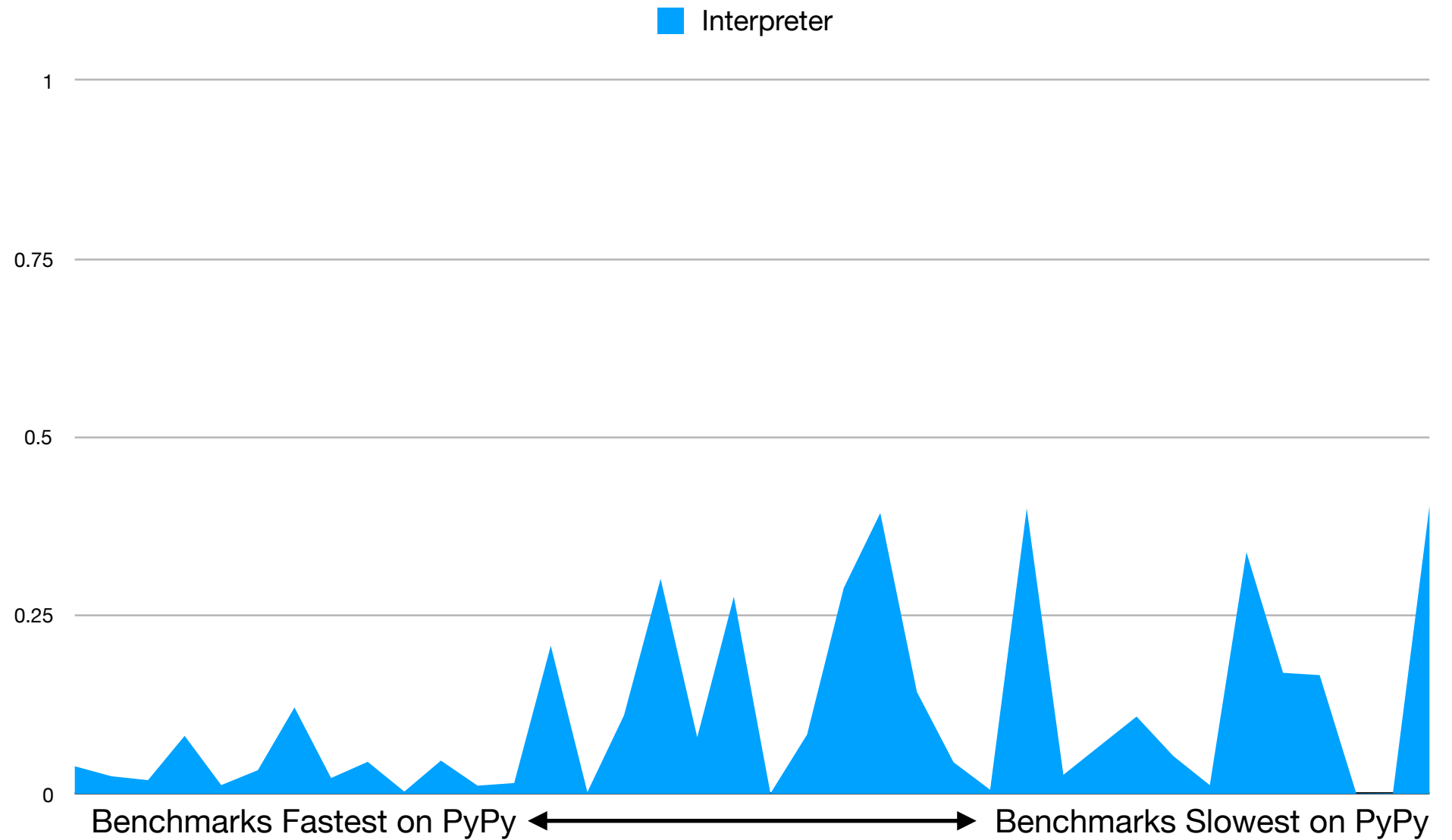
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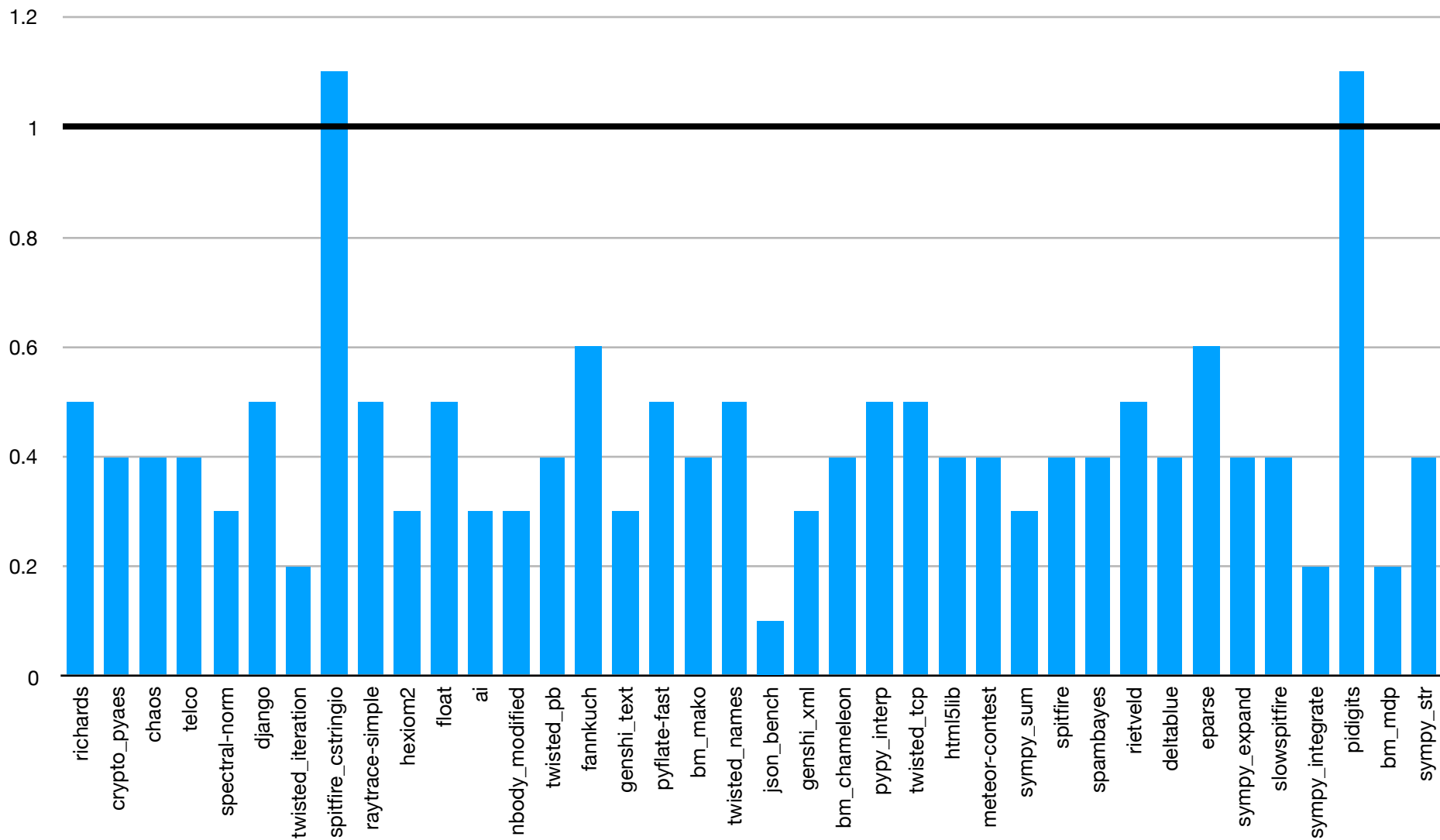




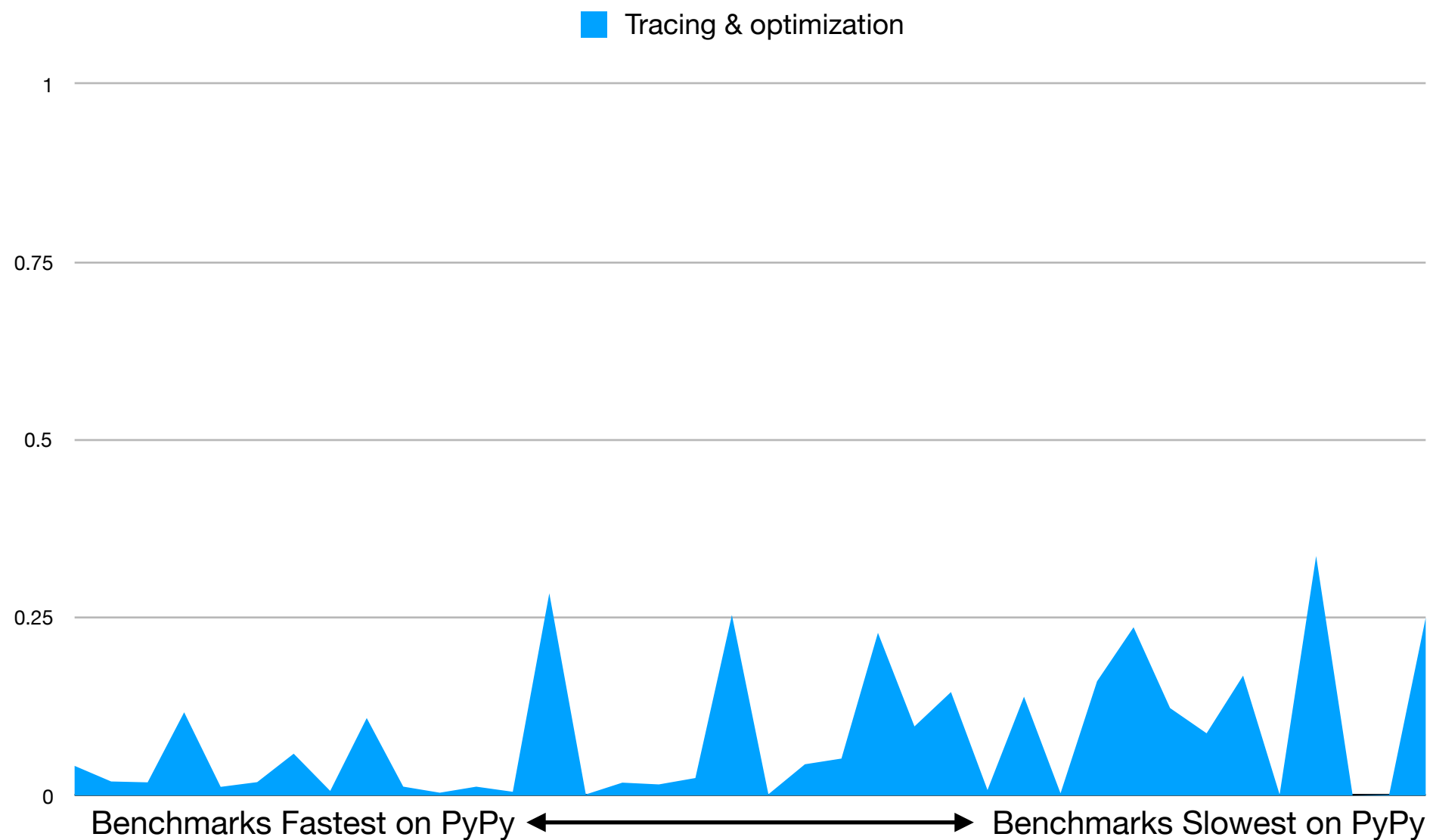
# Interpreter phase



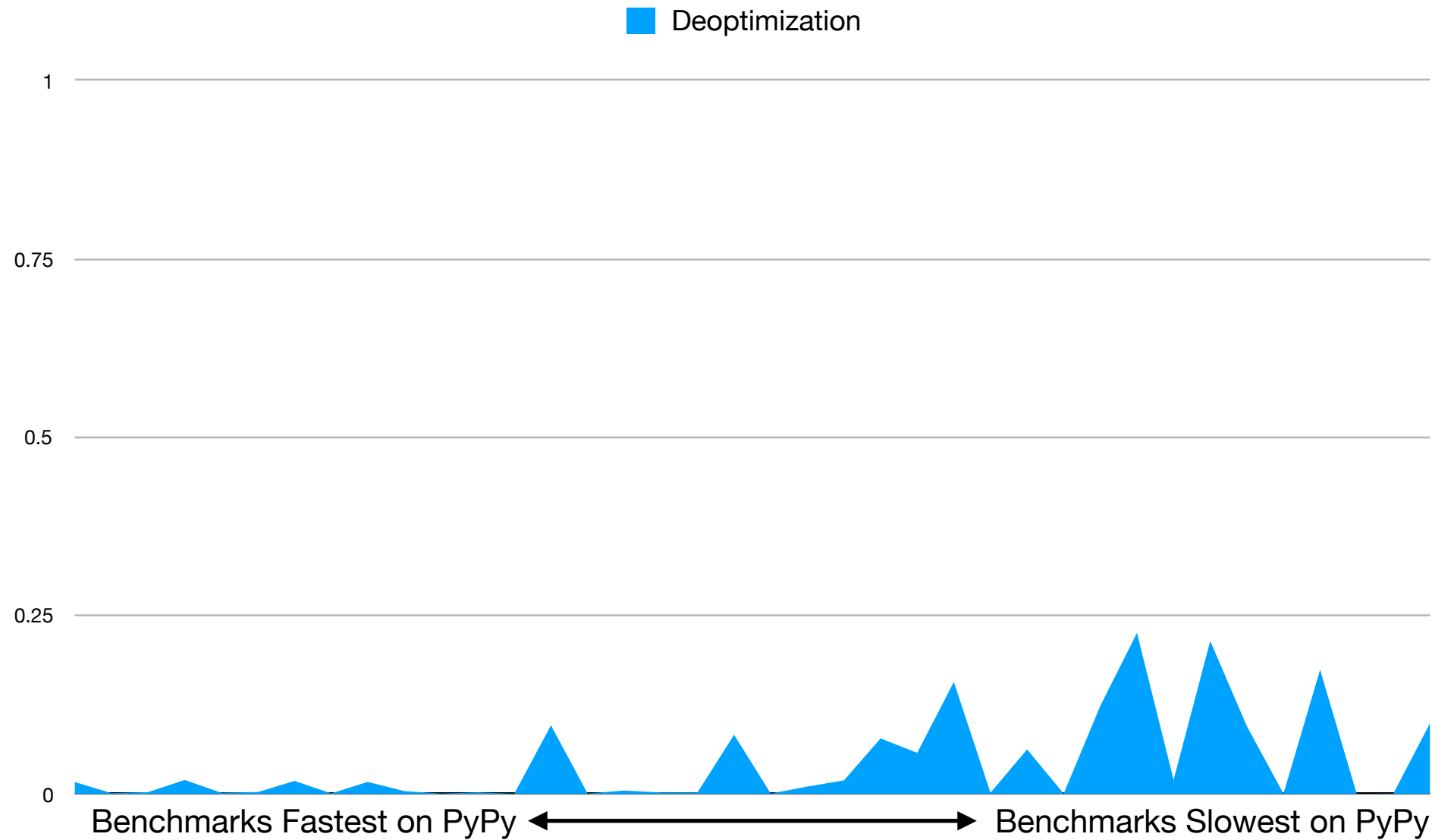
# PyPy *without* meta-tracing JIT speedup over CPython: RPython-to-C translation has overheads



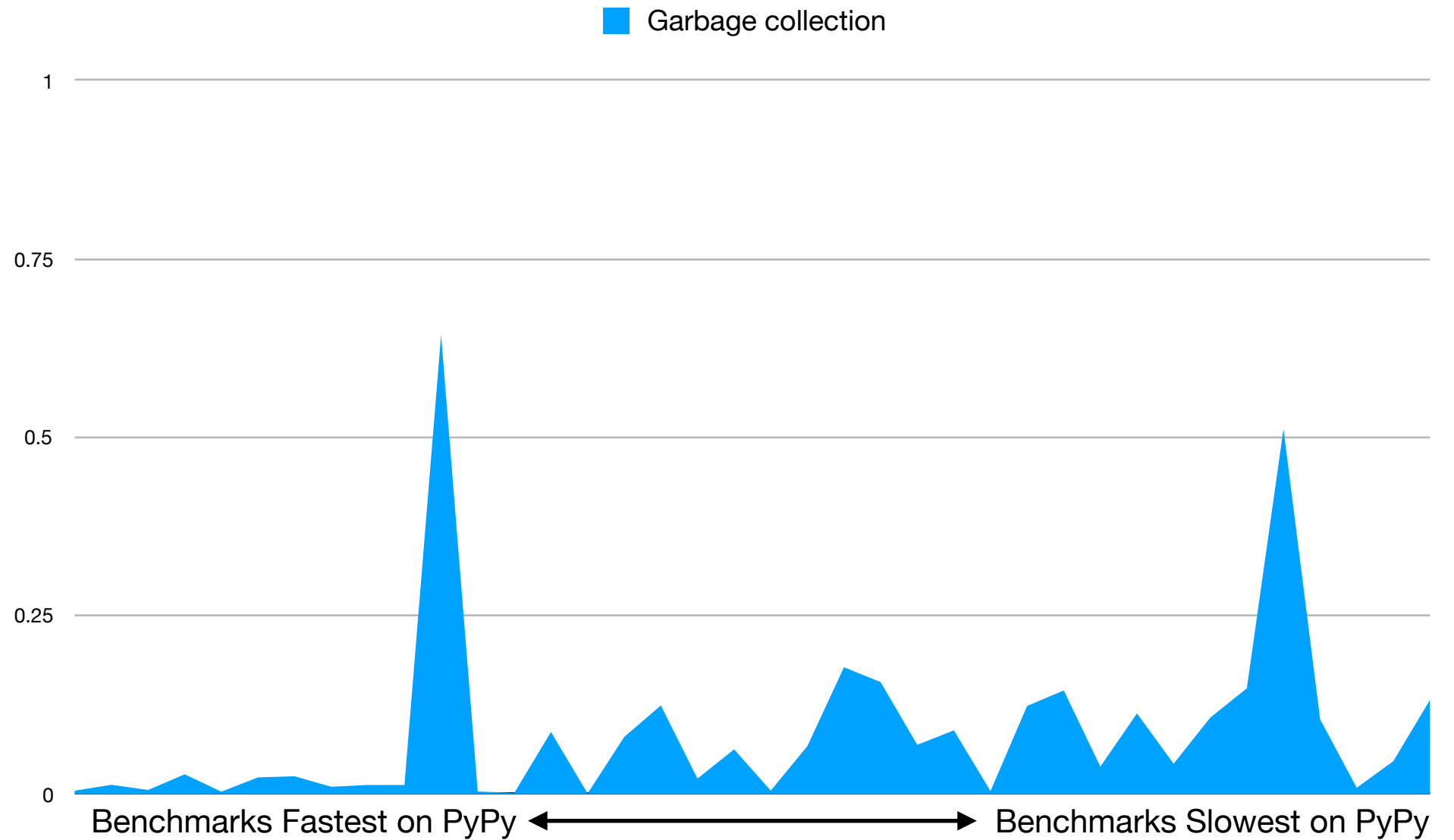
# Tracing and optimization phase



# Deoptimization phase

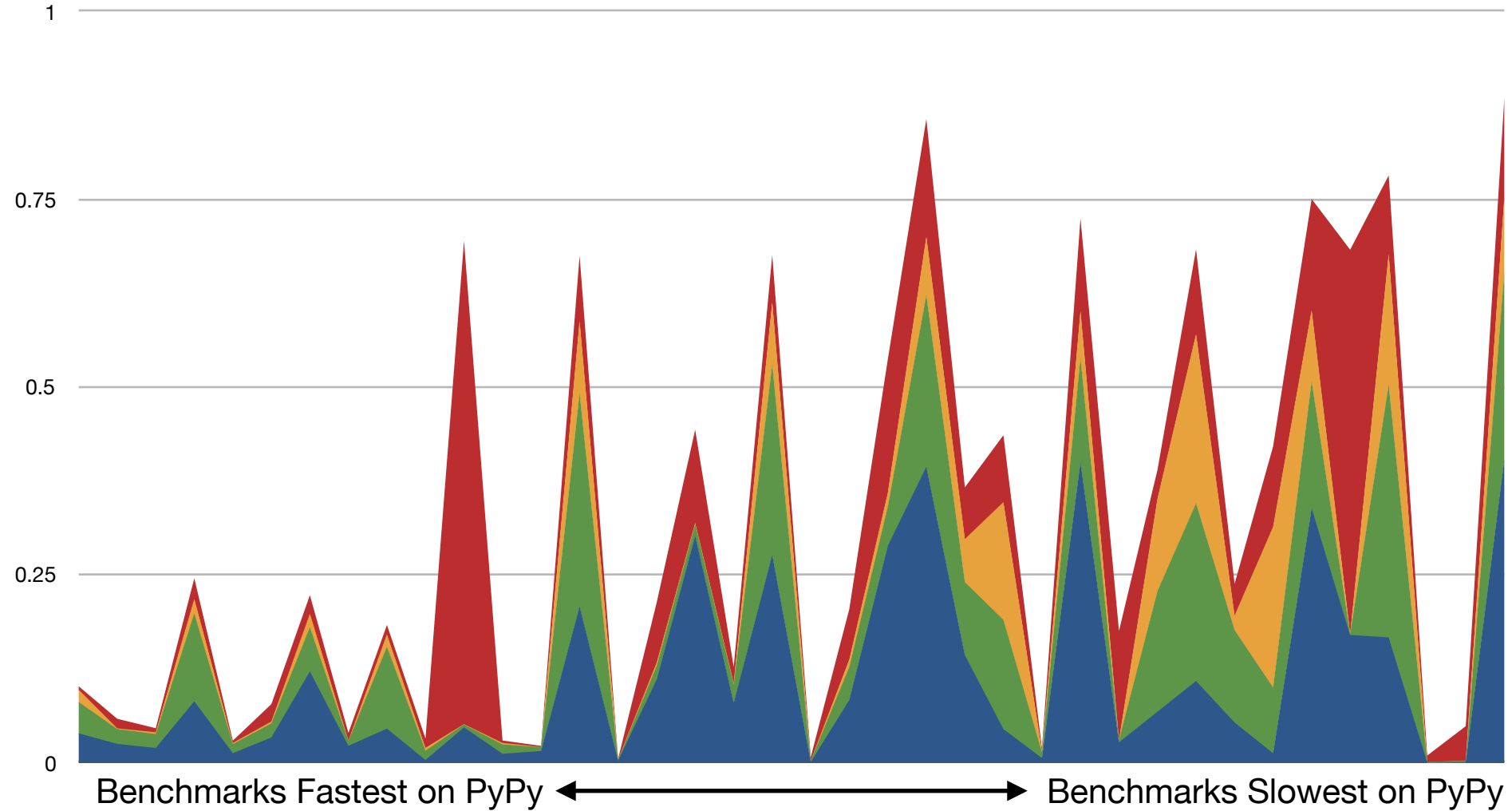


# Garbage collection phase



# Meta-tracing JIT VM overheads: Overheads are diverse and can add up to significant portion of execution

■ Interpreter   ■ Tracing & optimization   ■ Deoptimization   ■ Garbage collection



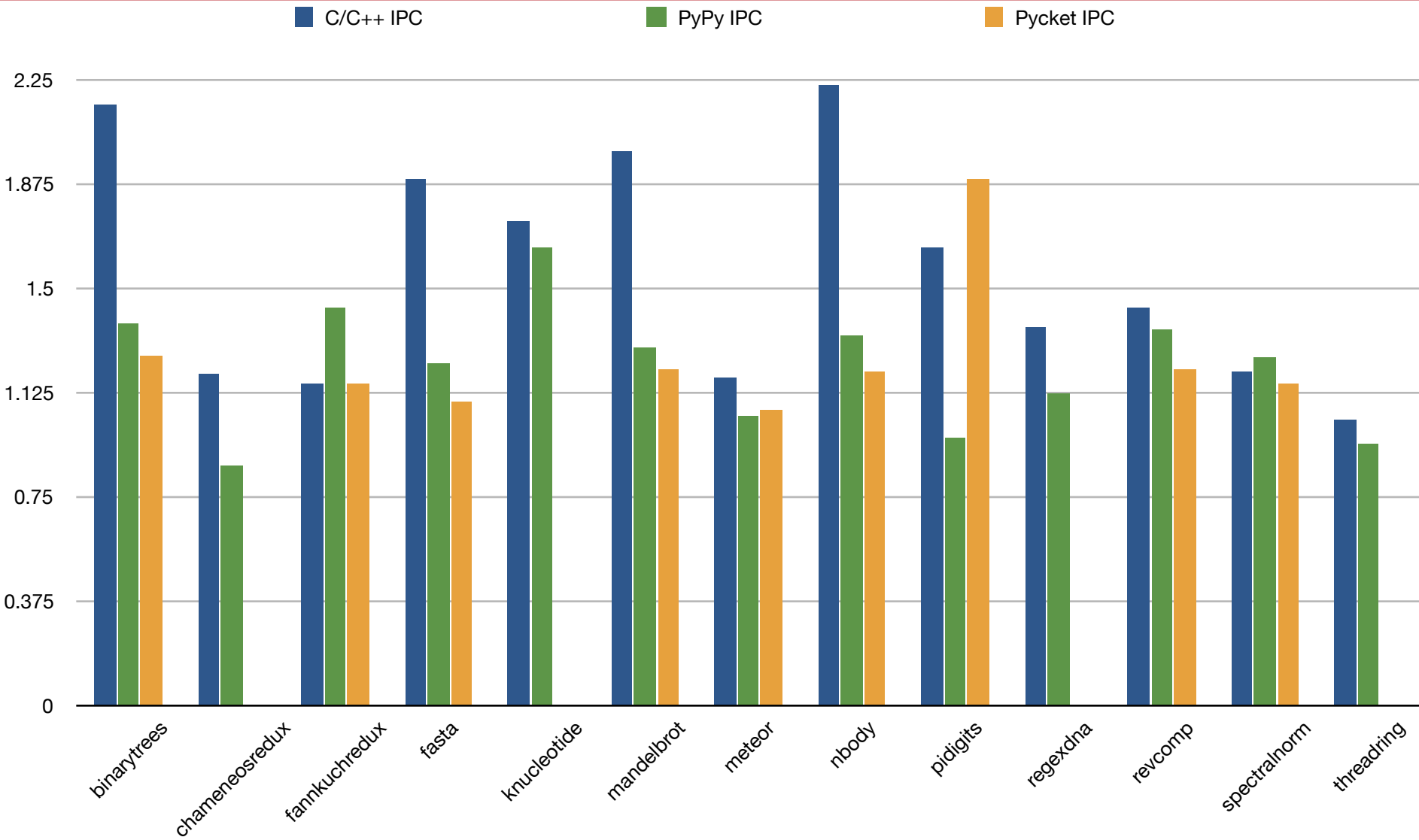
## Iron law of processor performance:

Does meta-tracing VM code execute poorly in addition to more instructions?

$$\frac{\text{Time}}{\text{Program}} = \frac{\text{Instructions}}{\text{Program}} \times \frac{\text{Cycle}}{\text{Instructions}} \times \frac{\text{Time}}{\text{Cycle}}$$



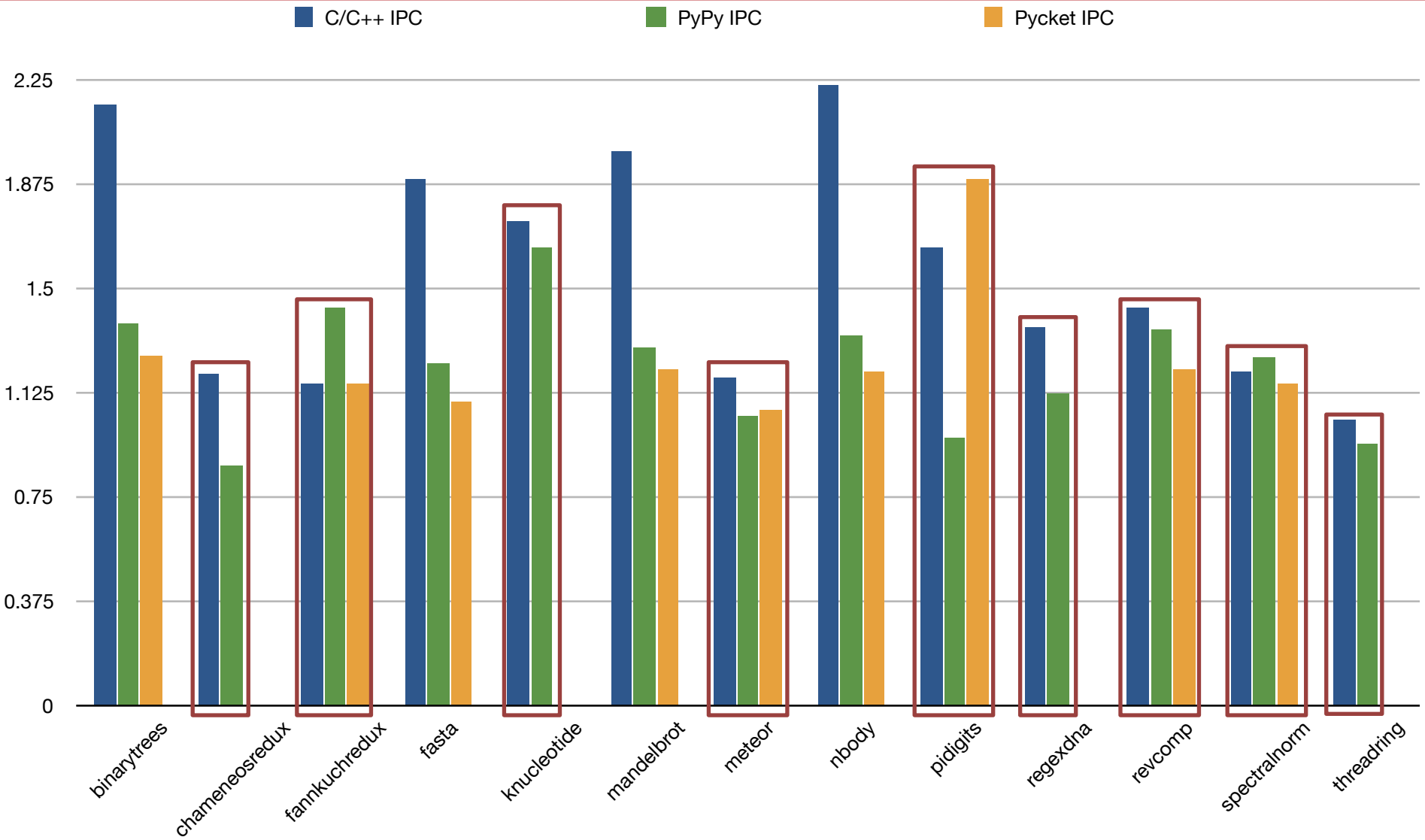
# Comparing meta-tracing JIT IPC to C/C++: Meta-tracing has a similar IPC for most benchmarks



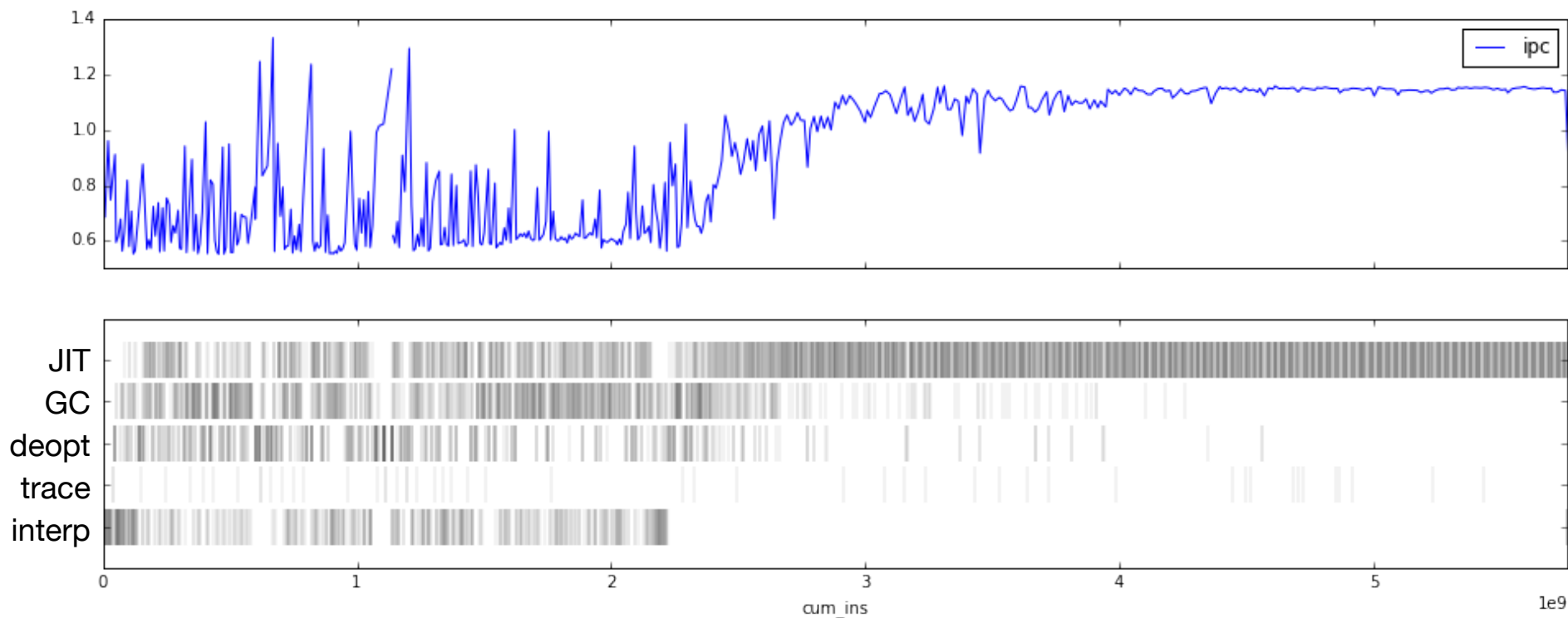


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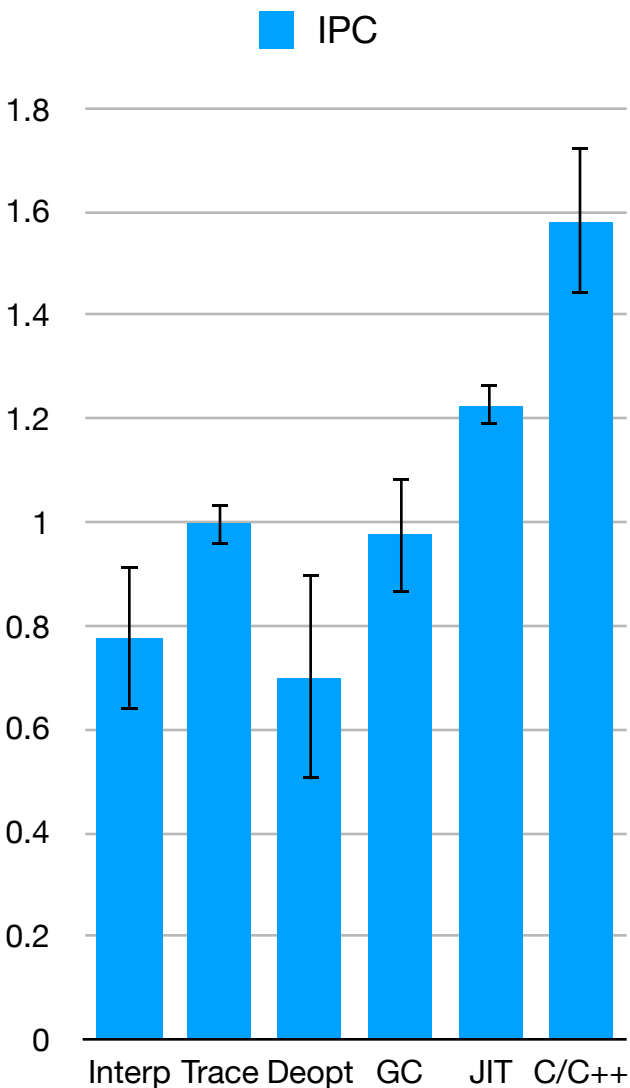


# IPC measurements can be accurately matched against VM phases



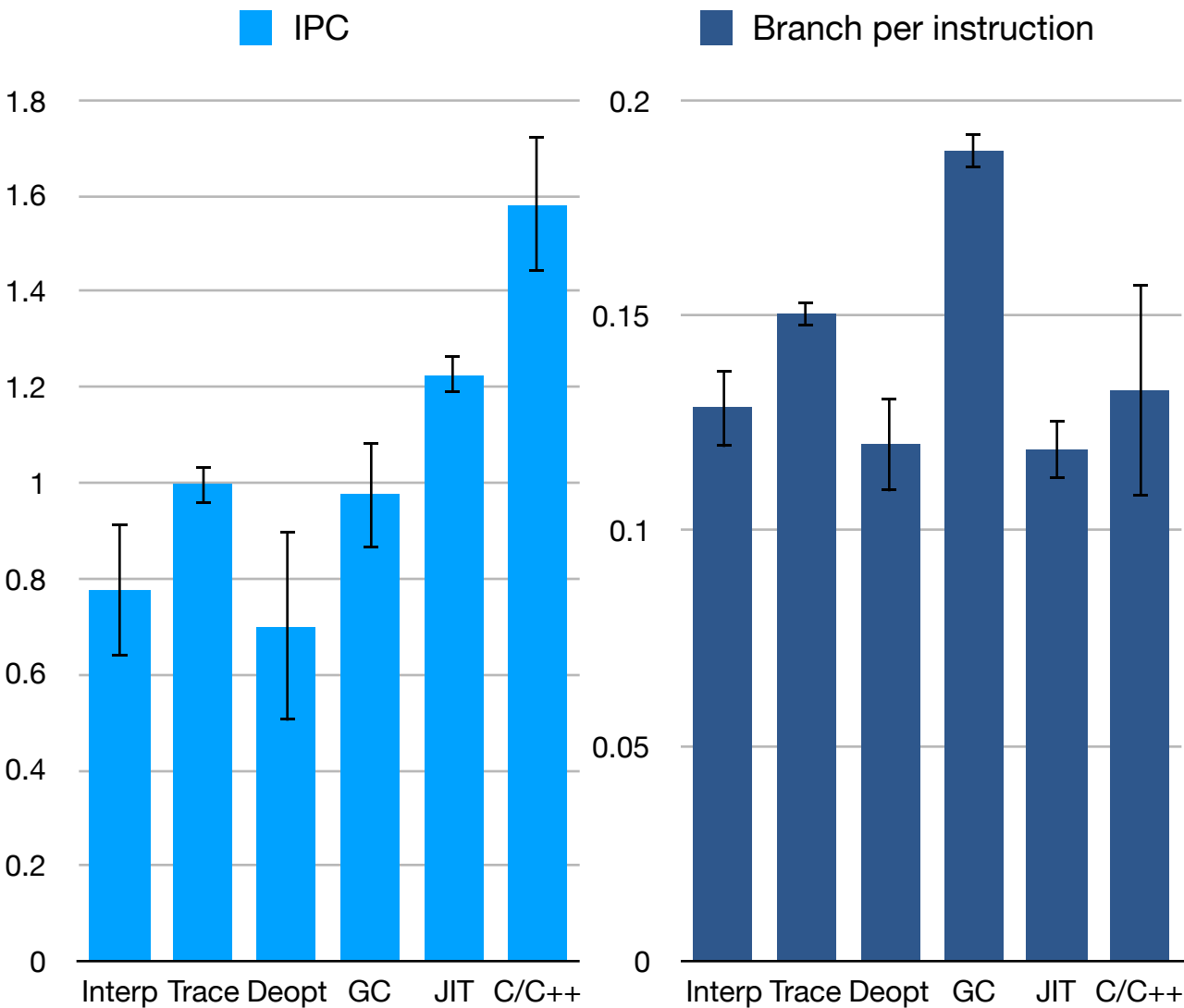
# Microarchitectural characterization by the VM phase:

## Meta-tracing-JIT-compiled code has a similar IPC, fewer branches and mispredictions

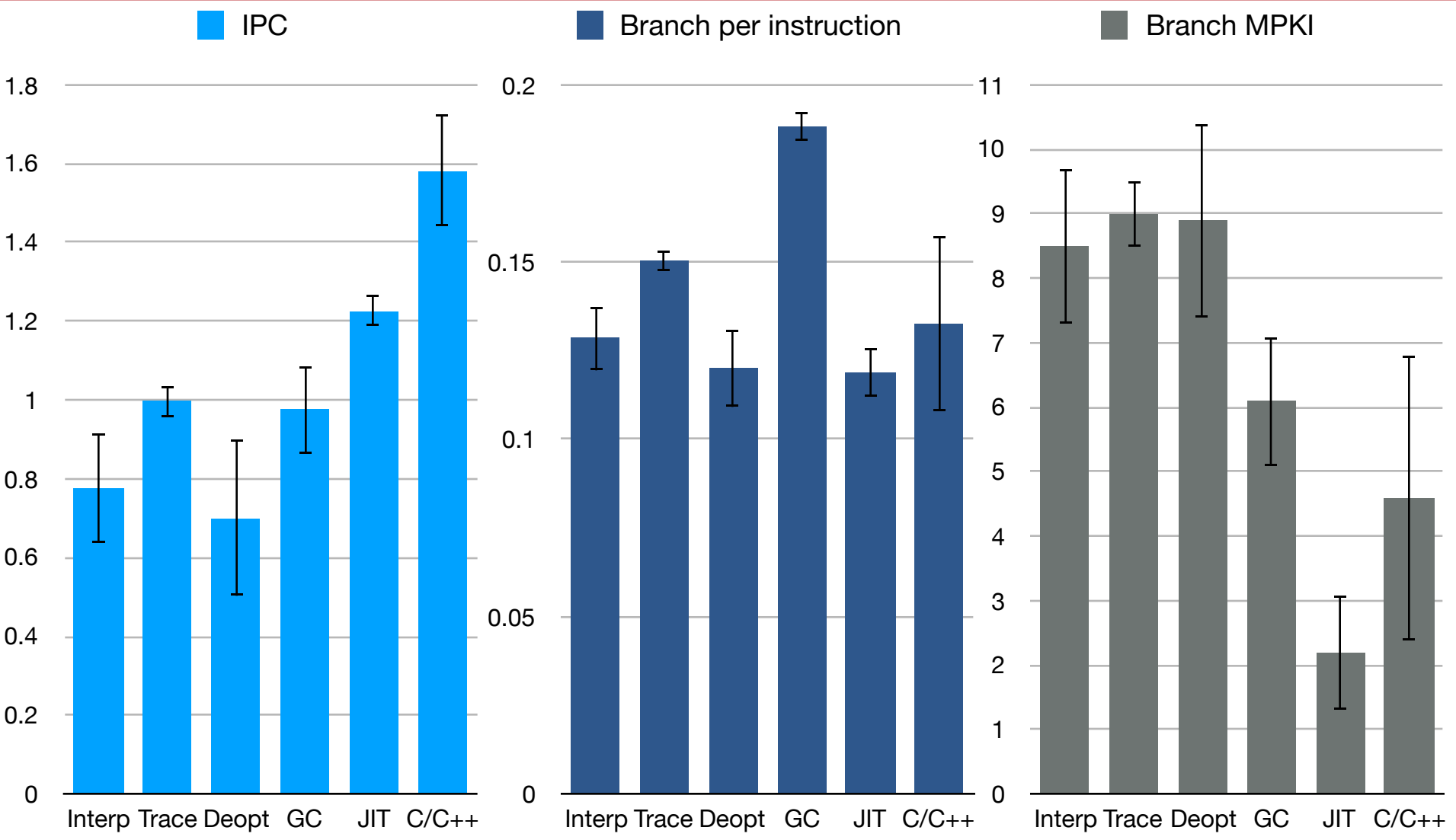


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  - *The problem is more instructions, not instructions that execute poorly*
  - *There is no silver bullet in addressing the performance gap*





Cornell University  
Computer Systems Laboratory



# Cross-Layer Workload Characterization of Meta-Tracing JIT VMs

**Berkin Ilbeyi<sup>1</sup>, Carl Friedrich Bolz-Tereick<sup>2</sup>,  
and Christopher Batten<sup>1</sup>**

**<sup>1</sup> Cornell University, <sup>2</sup> Heinrich-Heine-Universität Düsseldorf**