

# Interpreters Everywhere and All the Time



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**Stefan Marr**  
PLISS, September 2023

Got a Question?  
Please Interrupt Me!





# Stefan Marr

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Tinkering with interpreters  
for 6,000 days, and counting



Potsdam.

# How it Started

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Course on  
Virtual Machines  
1<sup>st</sup> of April 2007



Course Project:  
Build a just-in-time  
compiler for CSOM

In between:  
Many Years of Brussels & Lille  
(and Linz, but without the chocolate)



This is THE BEST Chocolate!



# How it's Going

---

University of  
**Kent**

**SOM**  
SimpleObjectMachine

THE  
**ROYAL  
SOCIETY**

“something like a  
professor”

still tinkering with  
interpreters

# Who Are We?



# Please stand up, if you have used a programming language!



Please stand,  
if you have implemented  
a programming language!



Please stand,  
if it's a language  
we have or will hear of!



Please stand,  
if it's any kind of  
interpreter!



# Topics Discussed Today

- How are programming languages implemented?
- Types of interpreters
  - Abstract syntax tree
  - Bytecode
- Interpreter optimizations
  - Lookup caching
  - AST/byticode-level inlining
  - Library lowering,  
library intrinsification
  - Super nodes,  
super instructions
  - Self-optimization, bytecode quickening

Got a Question?  
Please Interrupt Me!



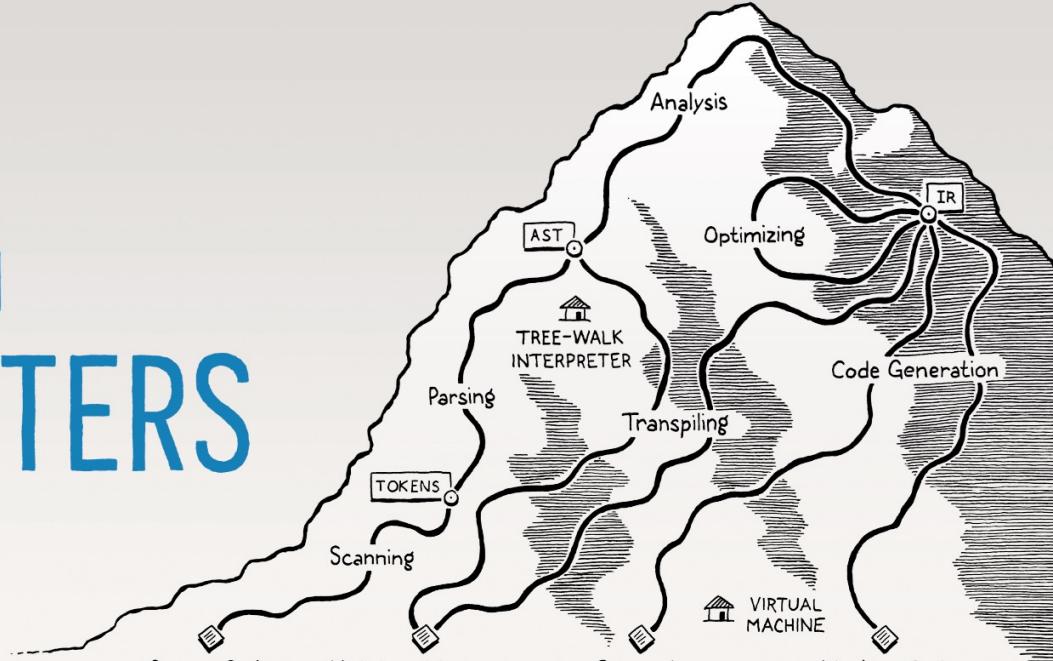
# Disclaimer!

Through the lens of Meta-compilation Systems  
Graal+Truffle and RPython



# CRAFTING INTERPRETERS

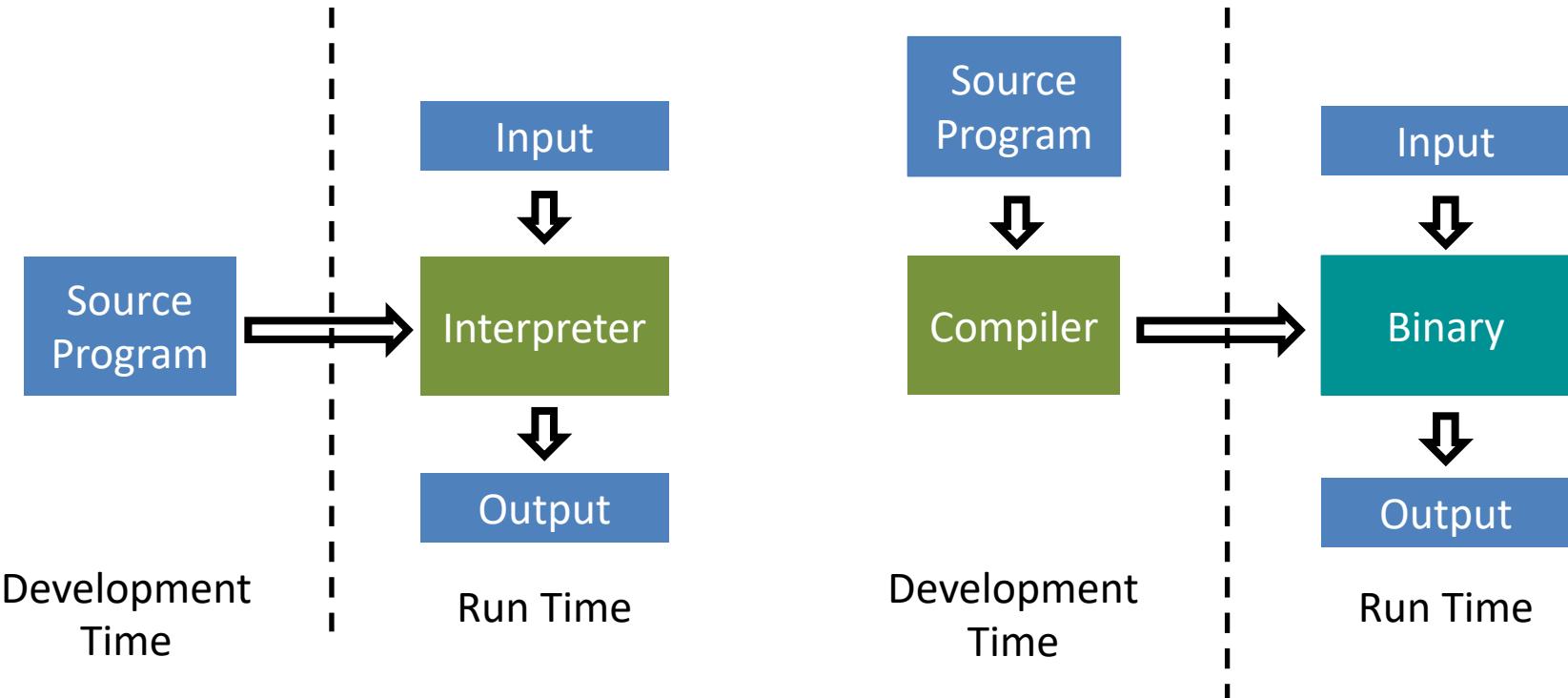
ROBERT NYSTROM



I am not using it today, but highly recommended: <https://craftinginterpreters.com/>

## HOW TO IMPLEMENT A PROGRAMMING LANGUAGE?

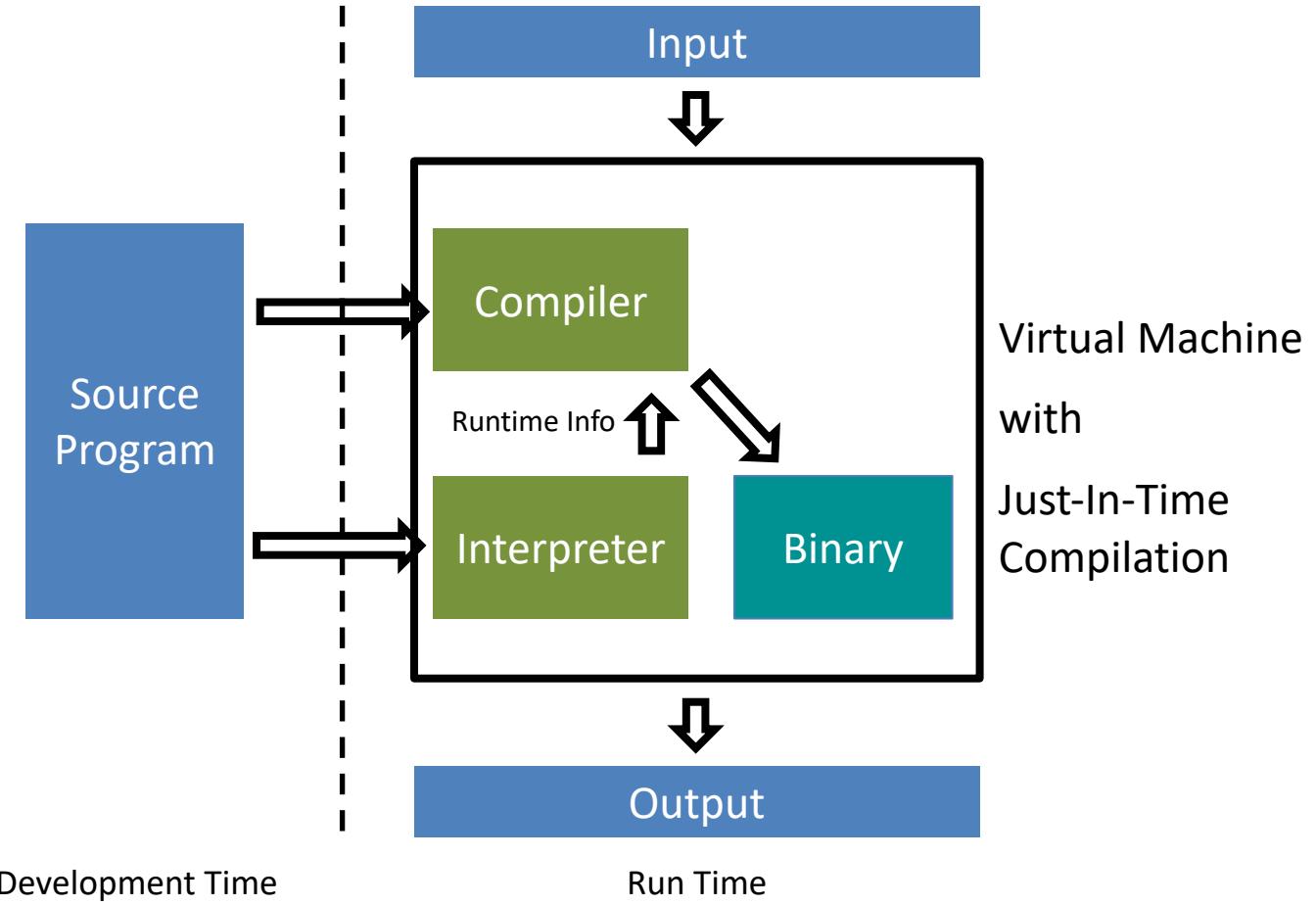
# Language Implementation Approaches



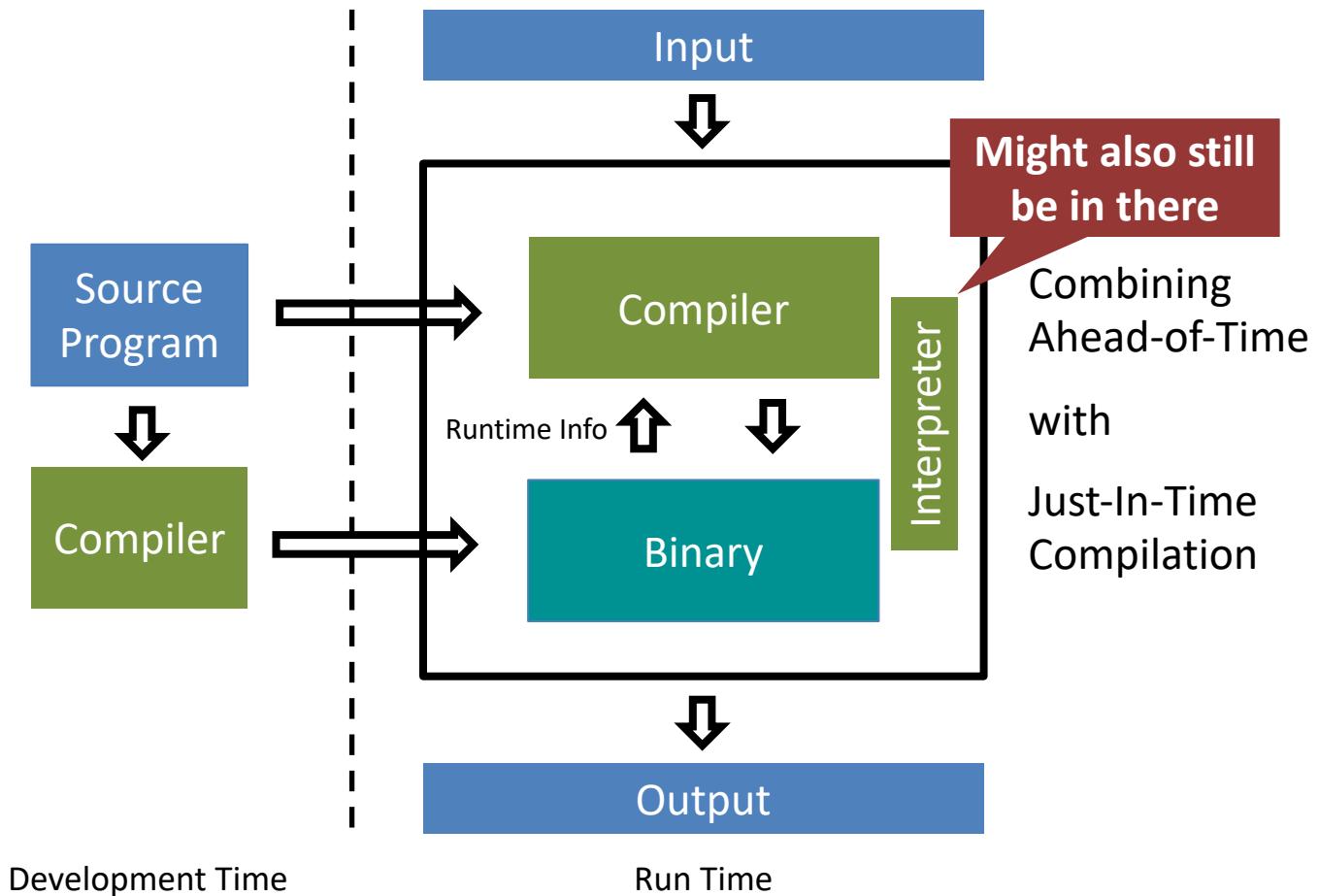
Simple, but often slow

More complex, but often faster  
Not ideal for all languages.

# Modern Virtual Machines



# Virtual Machines with Ahead-of-Time Compilation



# Pointers for this week



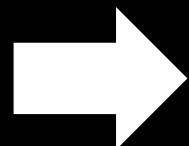
- JavaScript AOT Compilation  
Manuel Serrano
- Programming Languages on the Web  
Michael Lippautz
- A Brief Introduction to Just-in-Time  
Compilation  
by myself

# **TYPES OF INTERPRETERS**

# Code Convention



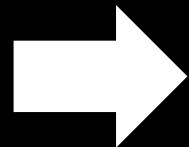
JavaScript-ish



Application Code

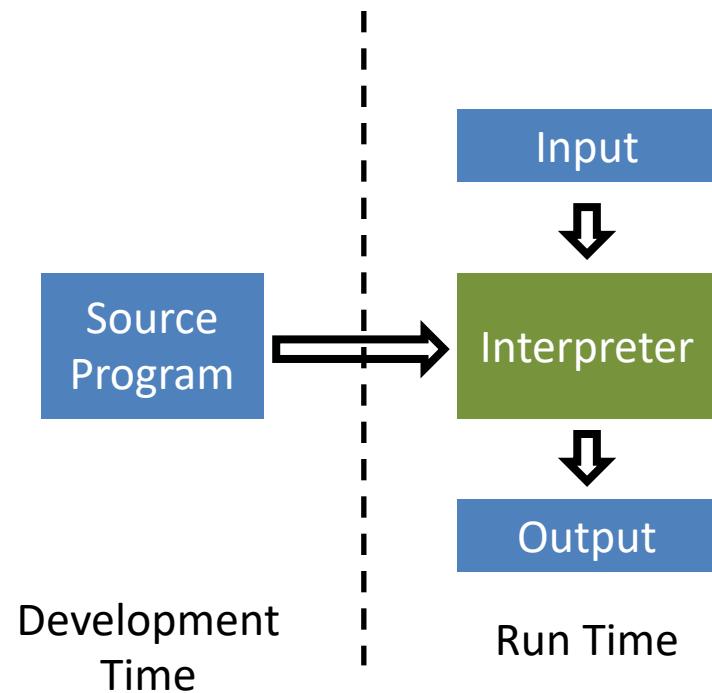


Python-ish  
with implicit typing



Interpreter Code

# Basic Interpretation



# From Program to Abstract Syntax Tree

```
function hello() {  
    log("Hello");  
}
```

programs are read by a parser

a parser is based on a grammar  
describing the language

**functionDeclaration:**

```
Function Identifier  
'(' formalParameterList? ')' '  
'{' functionBody '}' '
```

# From Program to Abstract Syntax Tree

```
function hello() {  
  log("Hello");  
}
```



Parser

focusing on the function body



**Tree Nodes represent semantic elements, and possibly operations**

CallFn

Constant

"Hello"

LoadGlobal

log

# A Simple Abstract Syntax Tree Interpreter

```
log("Hello");
```

root\_node



CallFn

Simple Interpreter:  
an execute method for  
each node

Constant

"Hello"

LoadGlobal

log

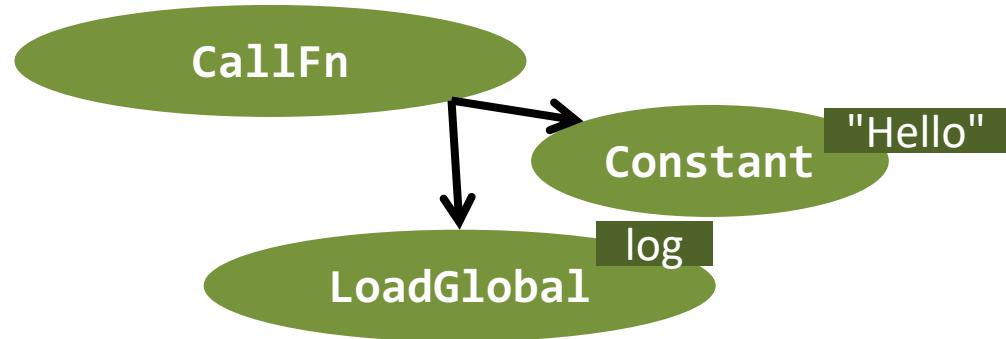
```
root_node = parse(file)  
root_node.execute(Frame())
```

# Implementing execute() methods

```
log("Hello");
```

```
class Constant:  
    final value  
    def execute(frame):  
        return value
```

```
class LoadGlobal:  
    final binding  
    def __init__(name):  
        binding = lookupGlobal(name)  
    def execute(frame):  
        return binding.value
```



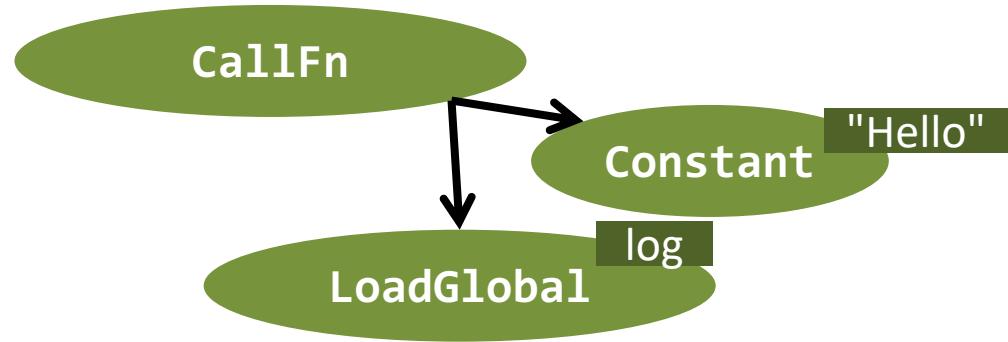
# Implementing execute() methods

```
log("Hello");
```

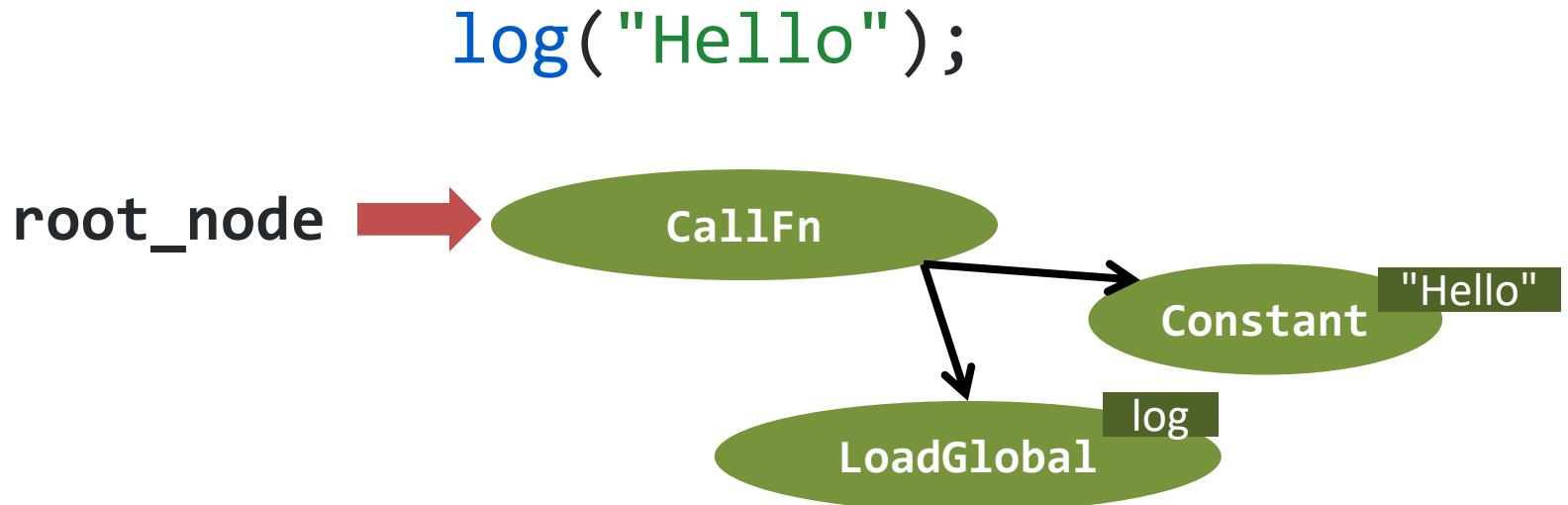
```
class Constant:  
    final value  
    def execute(frame):  
        return value
```

```
class LoadGlobal:  
    final binding  
    # ...  
    def execute(frame):  
        return binding.value
```

```
class CallFn:  
    node fn  
    node argument  
    def execute(frame):  
        fn = fn.execute(frame)  
        arg = argument.execute(frame)  
        return call(fn, arg)
```



# Abstract Syntax Tree Interpreters



Simple and Flexible

Any drawbacks?



# Abstract Syntax Tree Interpreters

## Benefits

- Very close to the source code
- Conceptually simple
- Makes building them “easy”

## Drawbacks

- A tree node per operation
  - Noncontiguous representation
  - High-ish memory use per node
- Many virtual methods (execution overhead)

# **BYTECODE INTERPRETERS**

# From Program to Bytecodes

```
log("Hello");
```



Parser



A linearized and  
compact representation

LoadGlobal

Constant

"Hello"

log



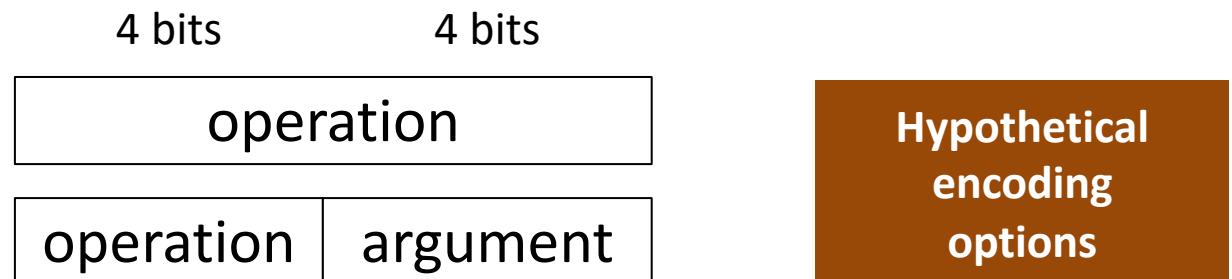
LoadGlobal "log"

LoadConstant "Hello"

CallFn

# Bytecode Encoding

- Represent common operations
- In a dense representation



LoadGlobal "log"  
LoadConstant "Hello"  
CallFn



translates to

1100 0000  
1110 0001  
1000 0000

# Classic Bytecode Interpreter

```
LoadGlobal "log"
LoadConstant "Hello"
CallFn

def interpret():
    idx = 0

    while idx < method.length:
        bc = method[idx]
        switch bc:
            case 0b1000_0000: # CallFn
                arg      = frame.pop()
                fn       = frame.pop()
                result  = call(fn, arg)
                frame.push(result)
```

# Bytecode Interpreters

## Benefits

- Compact and contiguous representation
- Traditionally more efficient than AST interpreters
- Bytecode can be more efficient to parse

## Drawbacks

- Requires extra design step
  - Bytecode format
  - Machine type
- Removes execution further from source code
  - Need to resolve mapping during debugging

For some languages,  
mapping is trivial, e.g.,  
Smalltalk

# Research and Literature

- Bytecode design
  - Stack vs. Register
    - Shi, Y., Gregg, D., Beatty, A. & Ertl, M. A. (2005). Virtual machine showdown: stack versus registers. *VEE'05*.
    - Ertl, M. A. (1995). Stack Caching for Interpreters. *PLDI'95*.
- Direct & indirect threaded interpretation
  - Ertl, M. A. & Gregg, D. (2003). The Structure and Performance of Efficient Interpreters.. *J. Instruction-Level Parallelism*, 5.
- Crafting Interpreters, Robert Nystrom  
<https://craftinginterpreters.com/>
- Java Virtual Machine specification  
<https://docs.oracle.com/javase/specs/index.html>
- Smalltalk's bytecode set  
[http://www.mirandabanda.org/bluebook/bluebook\\_chapter28.html](http://www.mirandabanda.org/bluebook/bluebook_chapter28.html)
- JavaScriptCore Bytecode Design  
<https://webkit.org/blog/9329/a-new-bytecode-format-for-javascriptcore/>
- V8 Ignition Interpreter design  
<https://v8.dev/docs/ignition>



But

**AST and Bytecode Interpreters**

are still

**“Often Slow” Interpreters!**

**WHY CARE FOR THESE  
“OFTEN SLOW”  
INTERPRETERS?**

## The Problem:

**Old assumptions don't hold for today's  
massive codebases and  
engineering approaches!**

For the last 30 years,  
we optimized for  
**Long-Running**  
Server Applications

# Large Applications, Frequent Changes



> 100 million  
lines of Hack/PHP

A new version every  
75 minutes<sup>1</sup>

1. HHVM Jump-Start: Boosting Both Warmup and Steady-State Performance at Scale. Guilherme Ottoni, Bin Liu. CGO'21
2. <https://shopify.engineering/automatic-deployment-at-shopify>
3. <https://instagram-engineering.com/continuous-deployment-at-instagram-1e18548f01d1>
4. <https://instagram-engineering.com/let-your-code-type-hint-itself-introducing-open-source-monkeytype-a855c7284881>

# Large Applications, Frequent Changes



> 100 million  
lines of Hack/PHP



*shopify*

> 3 million  
lines of Ruby

A new version every 30-40 times a day  
75 minutes<sup>1</sup> (every 36-48min)<sup>2</sup>

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# Large Applications, Frequent Changes



> 100 million  
lines of Hack/PHP



*shopify*

> 3 million  
lines of Ruby



> million  
lines of Python<sup>4</sup>

A new version every 30-40 times a day 30-50 times a day<sup>3</sup>  
75 minutes<sup>1</sup> (every 36-48min)<sup>2</sup>

1. HHVM Jump-Start: Boosting Both Warmup and Steady-State Performance at Scale. Guilherme Ottoni,  
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4. <https://instagram-engineering.com/let-your-code-type-hint-itself-introducing-open-source-monkeytype-a855c7284881>

# The Cost of Change

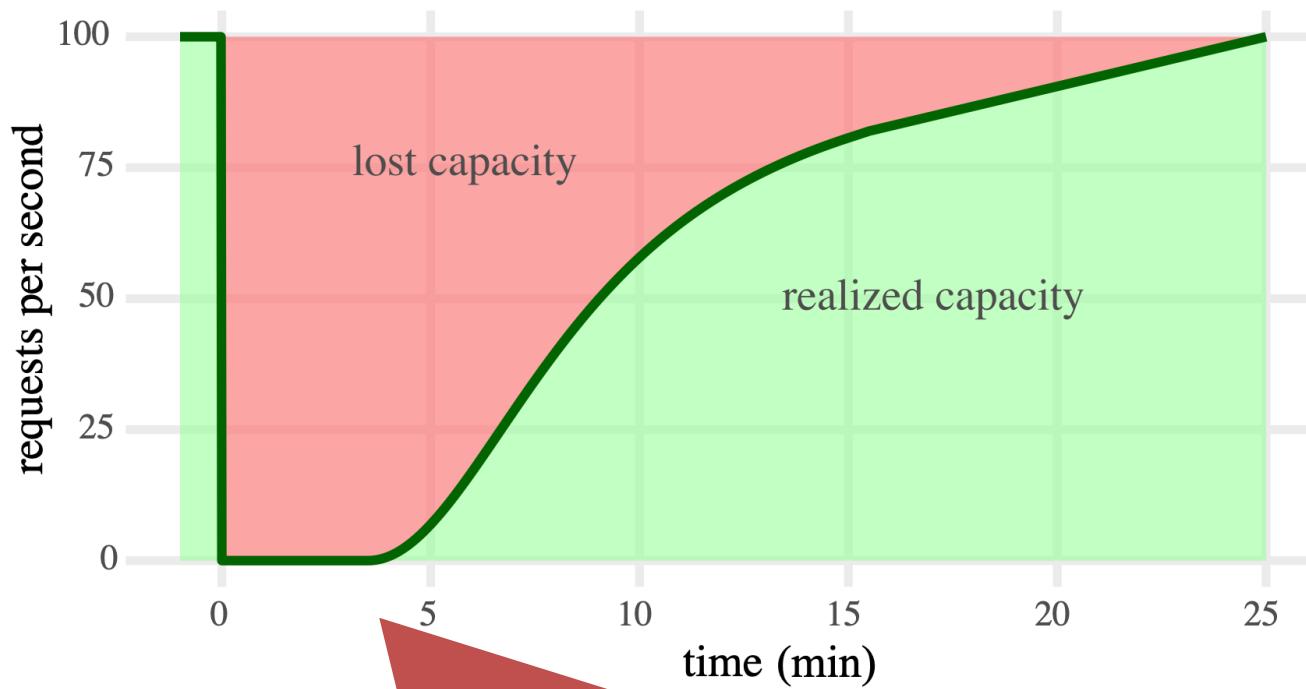


Fig.

**With millions of servers,  
every update “costs”!  
(need twice the capacity)**

Estimated Energy  
Bill, in 2020:  
£500,000,000  
(7,170,000 MWh)

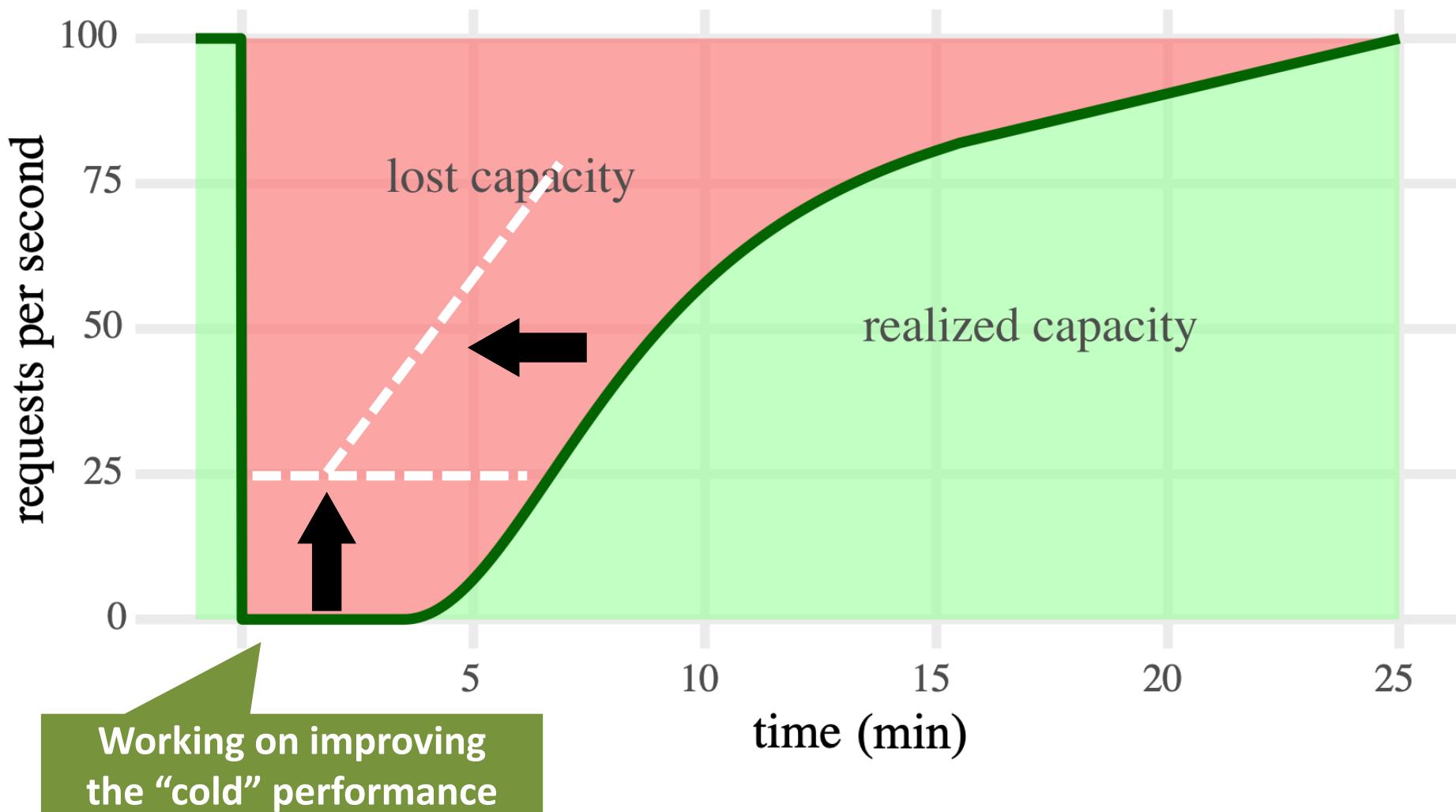


HHVM Jump-Start: Boosting both warmup and steady-state performance at scale.

Guilherme Ottoni, Bin Liu. CGO'21

<https://sustainability.fb.com/report-page/energy/>

# Building Faster Interpreters



Working on improving  
the “cold” performance

## SCIENCE ADVOCATES

Why is it important that we optimize interpreters?

Oh! They are used by many languages!  
Being fast reduces cost and increases productivity!



## SCIENTISTS

Why is it important that we optimize interpreters?

BECAUSE IT'S  
FUCKING  
AWESOME.

And, it's a  
very hard problem...

# How Many Interpreters/Compilers are Used During Execution?

- A Calculator Server

```
{"jsonrpc": "2.0",
"method": "subtract",
"params": [42, 23], "id": 1}
```
- Running on top of Node.js
- Running inside QEMU
- On top of a x86-64 processors?

2



4



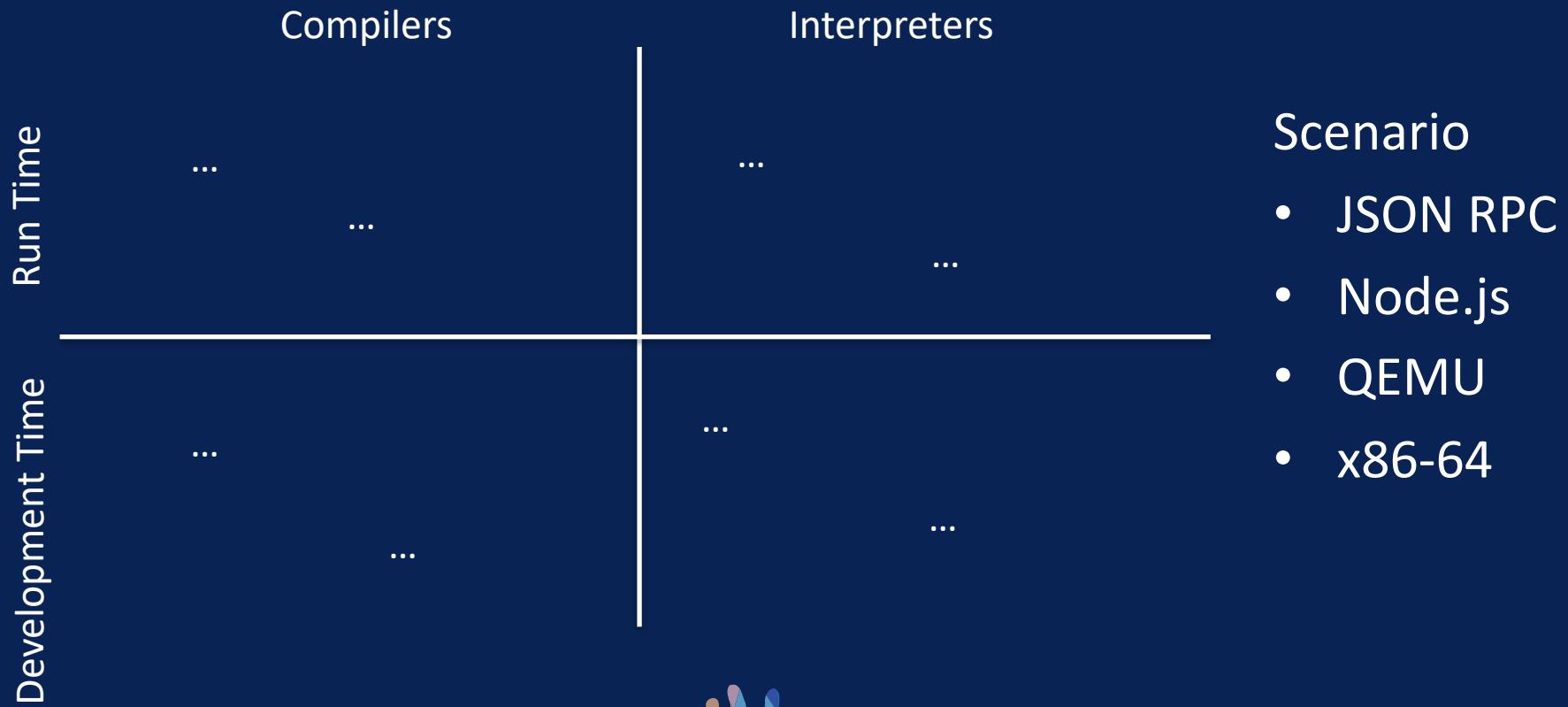
6



more



# Which Interpreters and Compilers are Involved to Create/Run this Stack?



# Interpreters

## Everywhere and All The Time!

# 5min Break

Next up

How can we optimize  
interpreters?

# **HOW CAN WE OPTIMIZE INTERPRETERS?**

# **LOOKUP CACHING**

# A Class Hierarchy Example

```
class Widget {  
    fitsInto(width) {  
        return this.width <= width;  
    }  
}  
  
class Button extends Widget {}  
class RadioButton extends Button {}  
  
function findAllThatFit(arr, width) {  
    const result = [];  
    for (const w of arr)  
        if (w.fitsInto(width))    What does this program do?  
            result.push(w)  
    return result;  
}
```



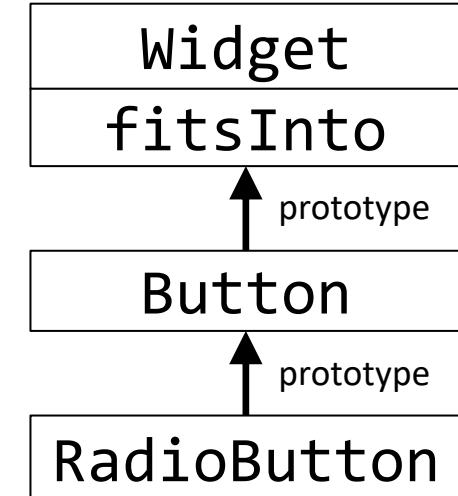
# Lookups can be frequent and costly

```
class Widget {  
    fitsInto(width) {  
        return this.width <= width  
    }  
}
```

```
class Button extends Widget {}  
class RadioButton extends Button {}
```

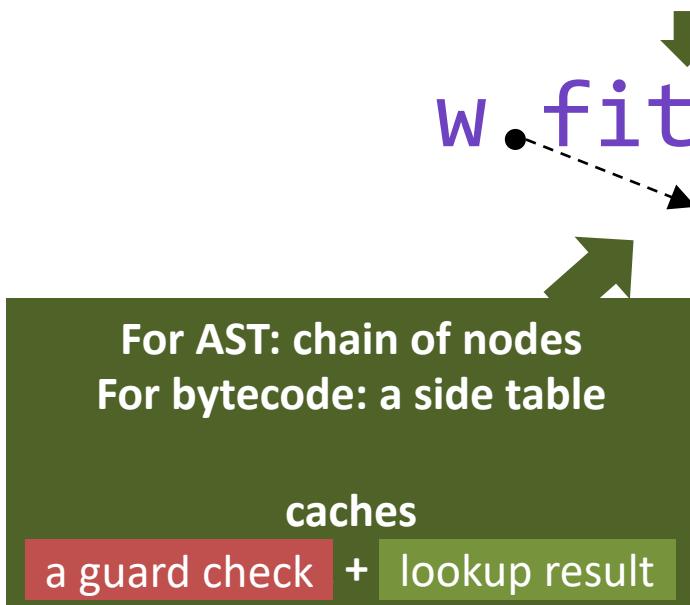
```
function findAllThatFit(arr, width) {  
    const result = [];  
    for (const w of arr)  
        if (w.fitsInto(width))  
            result.push(w)  
    return result;  
}
```

For each `fitsInto` call  
hasProperty: 3x  
getPrototypeOf: 2x



# Solution: Lookup Caching

could be various functions,  
but we don't need to do the same lookup repeatedly



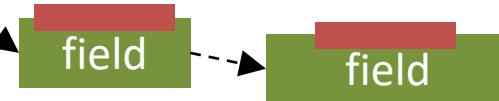
Any other language features  
we could use this for?



Useful when lookup is  
expensive, and essential  
for JIT compilation

# Lookup Caching Field Access

```
w.width = width // write  
width = w.width // read
```



e.g. for  
languages  
where one can  
add/remove  
fields

# Inlining Trivial Methods

```
class Widget {  
    #height;  
    getHeight() {  
        return this.height;  
    }  
    getAnswer() {  
        return 42;  
    }  
}  
const b = new Button;  
b.getHeight()  
b.getAnswer()
```

read field

42

Extra benefit:  
avoid call overhead

Already mentioned  
in the '91 paper...

# Lookup Cache

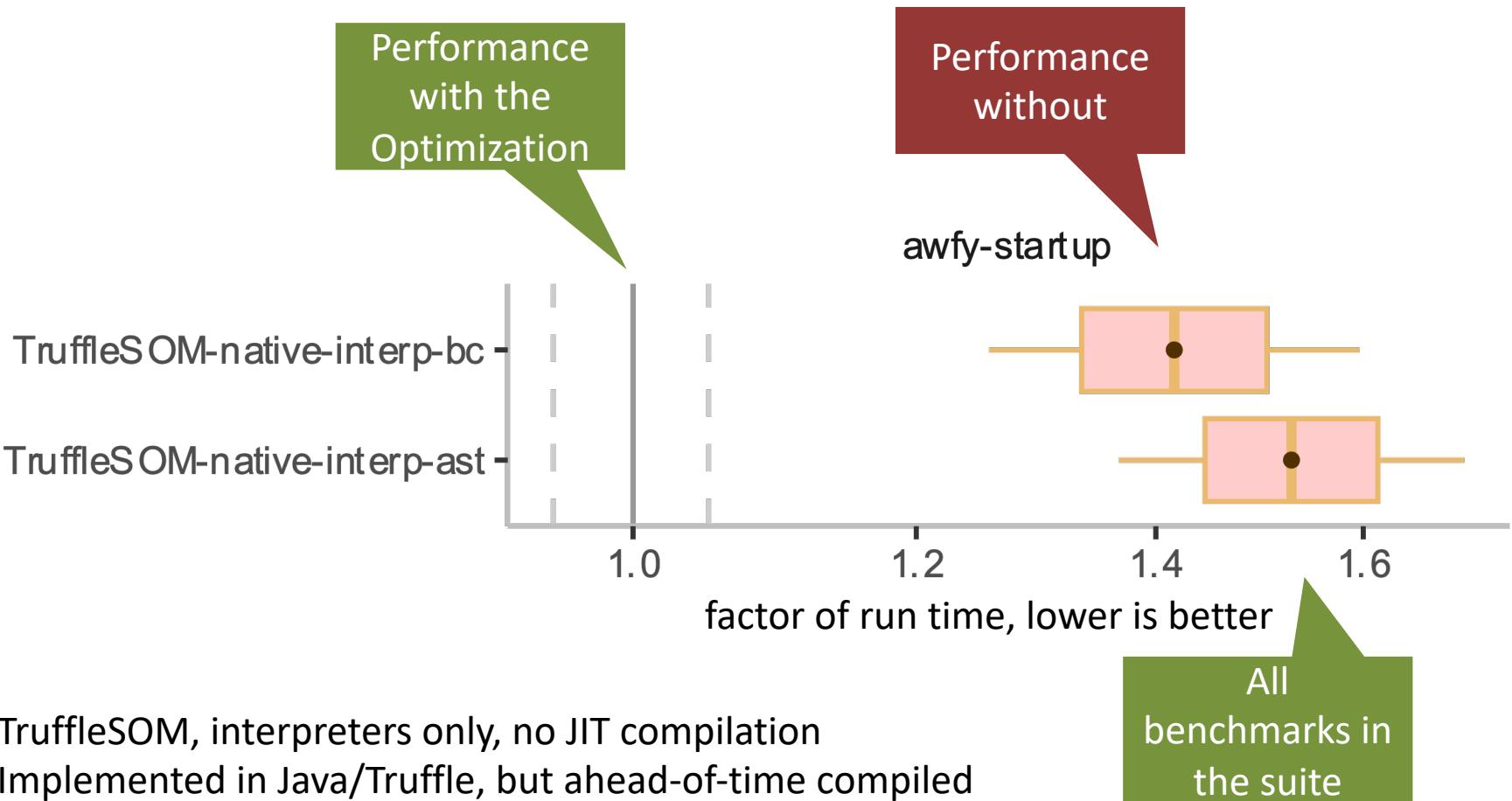
- Can be used for
  - Method/function lookup
  - Field access
  - Getters, setters, method returning constants, other “trivial” methods
  - Object creation when `new` is a method
  - Reflective operations, caches can be nested...

# Disclaimer!

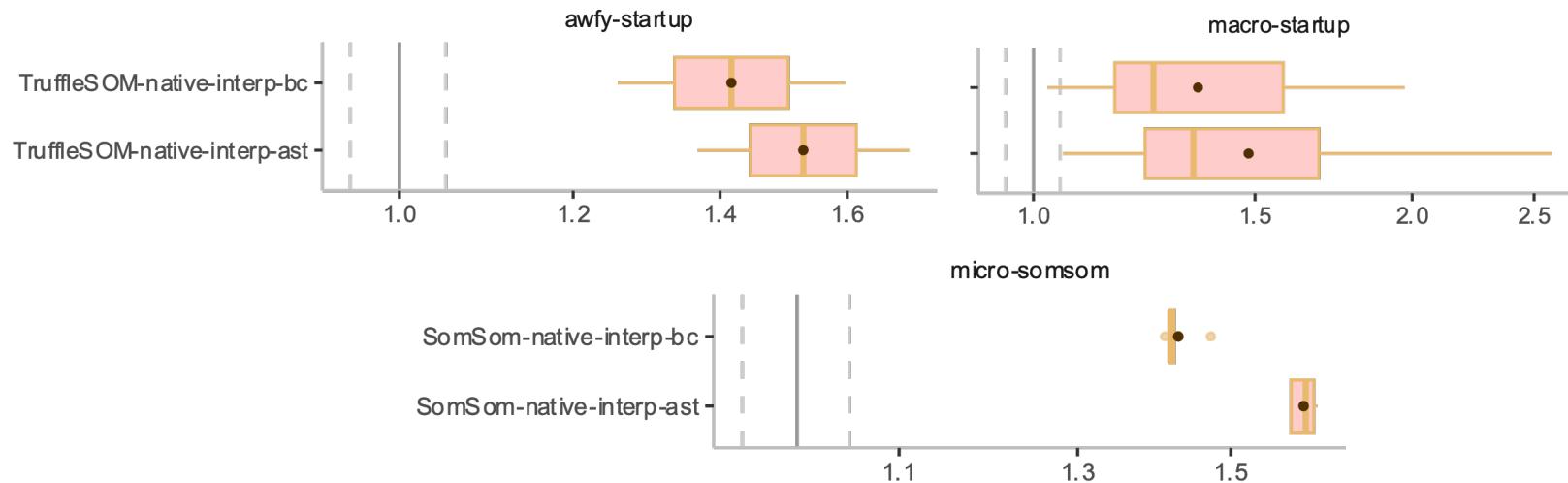
Through the lens of Meta-compilation Systems  
Graal+Truffle and RPython



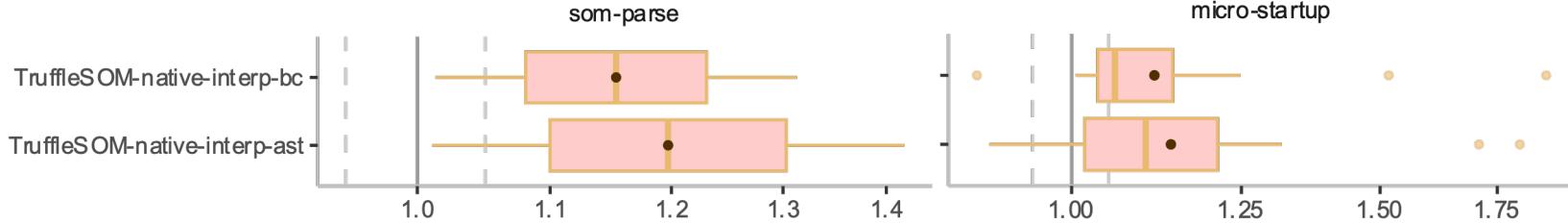
# Performance Benefit for the Interpreter



# Performance Benefit for the Interpreter



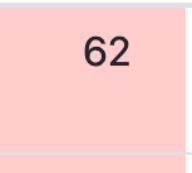
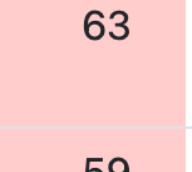
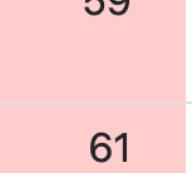
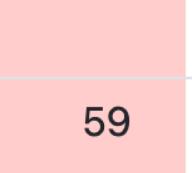
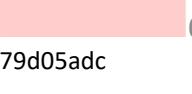
Larger Benchmarks take about 1.5x more time without Lookup Caching



# Performance Benefit, More Details

## micro-somsom

Executor: SomSom-native-interp-ast

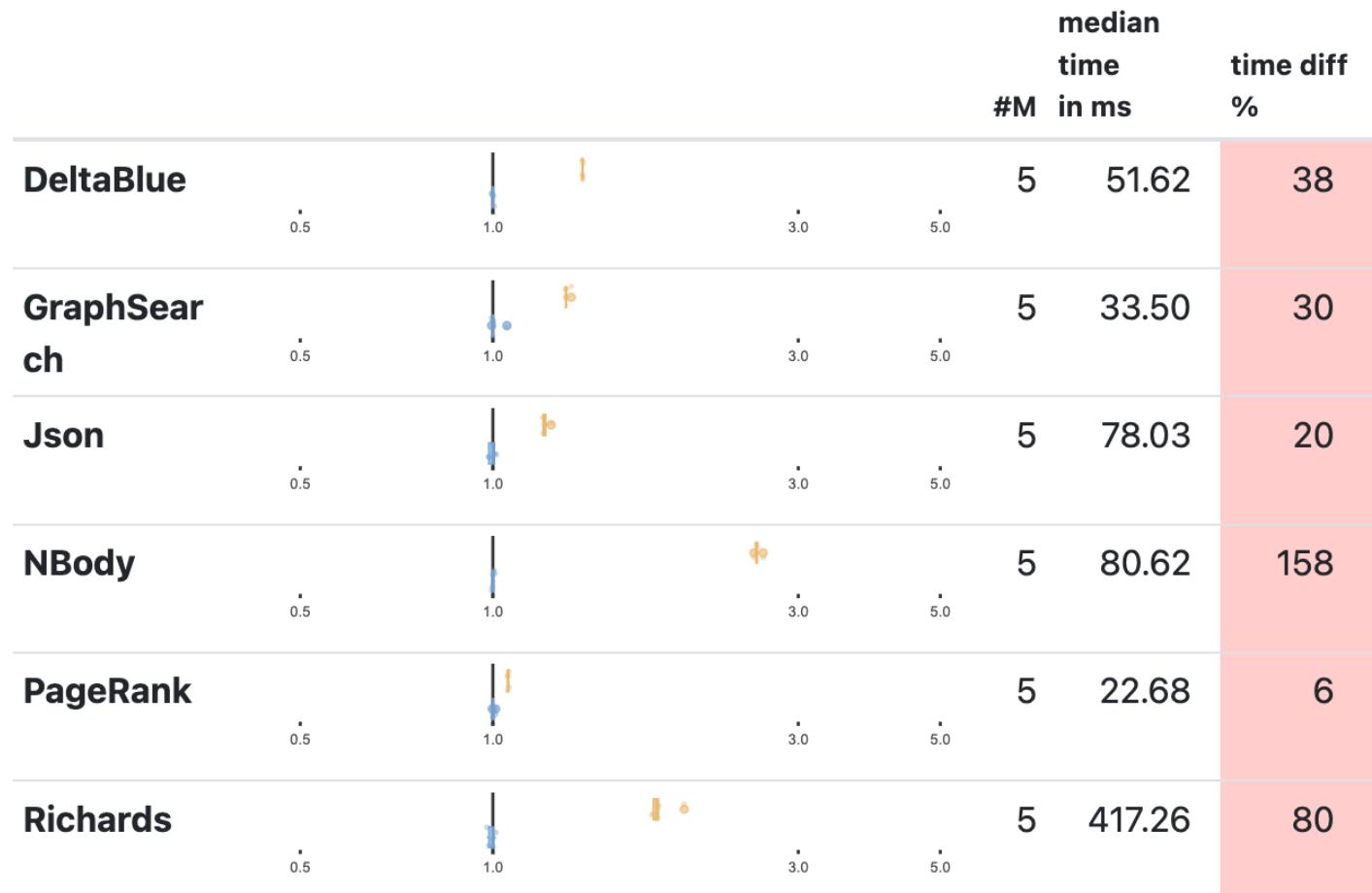
			median time	time diff
	#M	in ms	%	
List	1	20434.63	62	
Loop	1	14287.21	63	
Mandelbr ot	1	321.64	59	
Queens	1	12918.73	61	
Recurse	1	12934.17	59	

The table displays performance metrics for five benchmarks: List, Loop, Mandelbr ot, Queens, and Recurse. Each row includes a histogram showing the distribution of execution times, the median time in milliseconds, and the percentage difference from the baseline. The 'Recurse' benchmark shows the best performance, while 'List' and 'Loop' show the worst.

# Performance Benefit, More Details

## macro-startup

Executor: TruffleSOM-native-interp-ast



# Research and Literature

- **Efficient Implementation of the Smalltalk-80 System.**  
Deutsch, L. P. & Schiffman, A. M. (1984). *POPL'84*
- **Optimizing Dynamically-Typed Object-Oriented Languages With Polymorphic Inline Caches.**  
Hölzle, U., Chambers, C. & Ungar, D. (1991). *ECOOP'91*
- **Zero-Overhead Metaprogramming: Reflection and Metaobject Protocols Fast and without Compromises.**  
Marr, S., Seaton, C. & Ducasse, S. (2015). *PLDI'15*
- **Who You Gonna Call: Analyzing the Run-time Call-Site Behavior of Ruby Applications.**  
Kaleba, S., Larose, O., Jones, R., & Marr, S. (2022). *DLS'22*
- Optimizing prototypes in V8  
<https://mathiasbynens.be/notes/prototypes>
- <https://mathiasbynens.be/notes/shapes-ics>
- <https://mrale.ph/blog/2012/06/03/explaining-js-vms-in-js-inline-caches.html>



# **INLINING OF CONTROL STRUCTURES**

# Languages where Control Flow is Realized with Methods?

Languages	Examples of Methods
...	...
...	...
...	...
...	...
...	...
...	...
...	...
...	...



# Let's Go all In on Polymorphic Methods!

```
class True {  
    ifTrue(lambda) {  
        return lambda.call();  
    }  
    ifTrueElse(  
        trueL, falseL) {  
        return trueL.call();  
    }  
}  
  
true.ifTrue(() => {  
    // a bit clunky in JS syntax,  
    // but works...  
});
```

```
class False {  
    ifTrue(lambda) {  
        return null;  
    }  
    ifTrueElse(  
        trueL, falseL) {  
        return falseL.call();  
    }  
}
```

# Iterating with Methods

```
class Range {  
    each(lambda) {  
        let i = this.start;  
        while (i < this.end) {  
            lambda.call(i);  
            i += 1;  
        }  
        return null;  
    }  
}  
  
(1..10).each(n => log(n));
```

Can be applied to all control flow

- if/for/while
- each/map/filter
- computeIfAbsent

# Inlining the Simple Cases Only

constant lambda, right  
there in the code

```
debugMode.isTrue(() => {  
    log("Hello");  
});
```

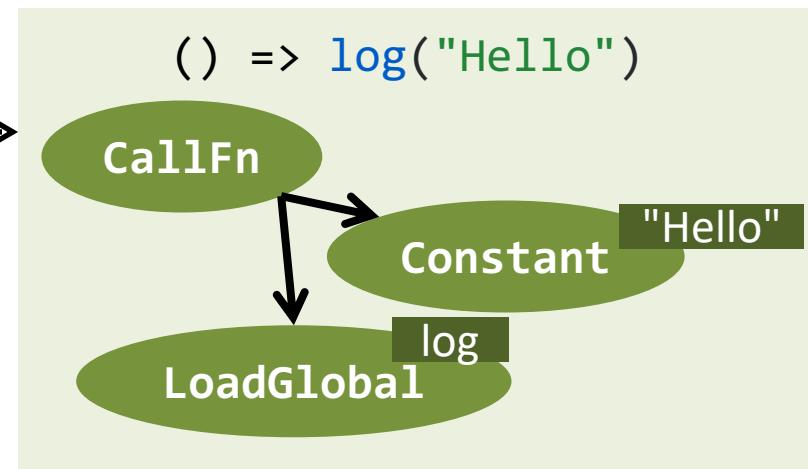
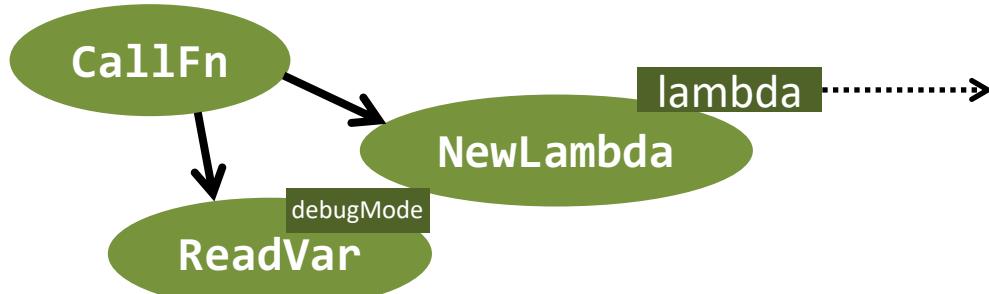
```
(1..10).each(n => log(n));
```

ensure correctness  
as we do with  
lookup caches

# Inlining ifTrue

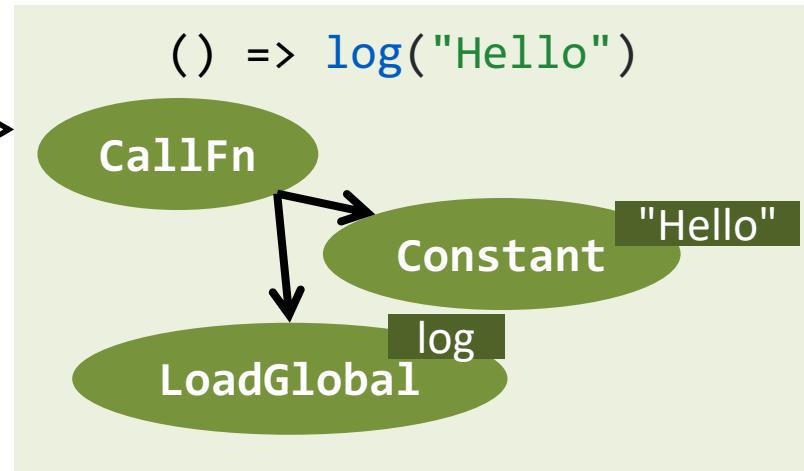
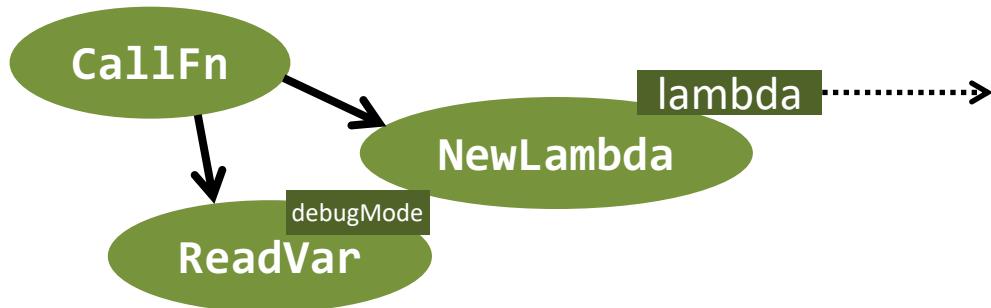
```
debugMode.ifTrue(() => {  
    log("Hello");  
});
```

```
debugMode.ifTrue(() => {...});
```



# Inlining ifTrue

```
debugMode.ifTrue(() => {...});
```



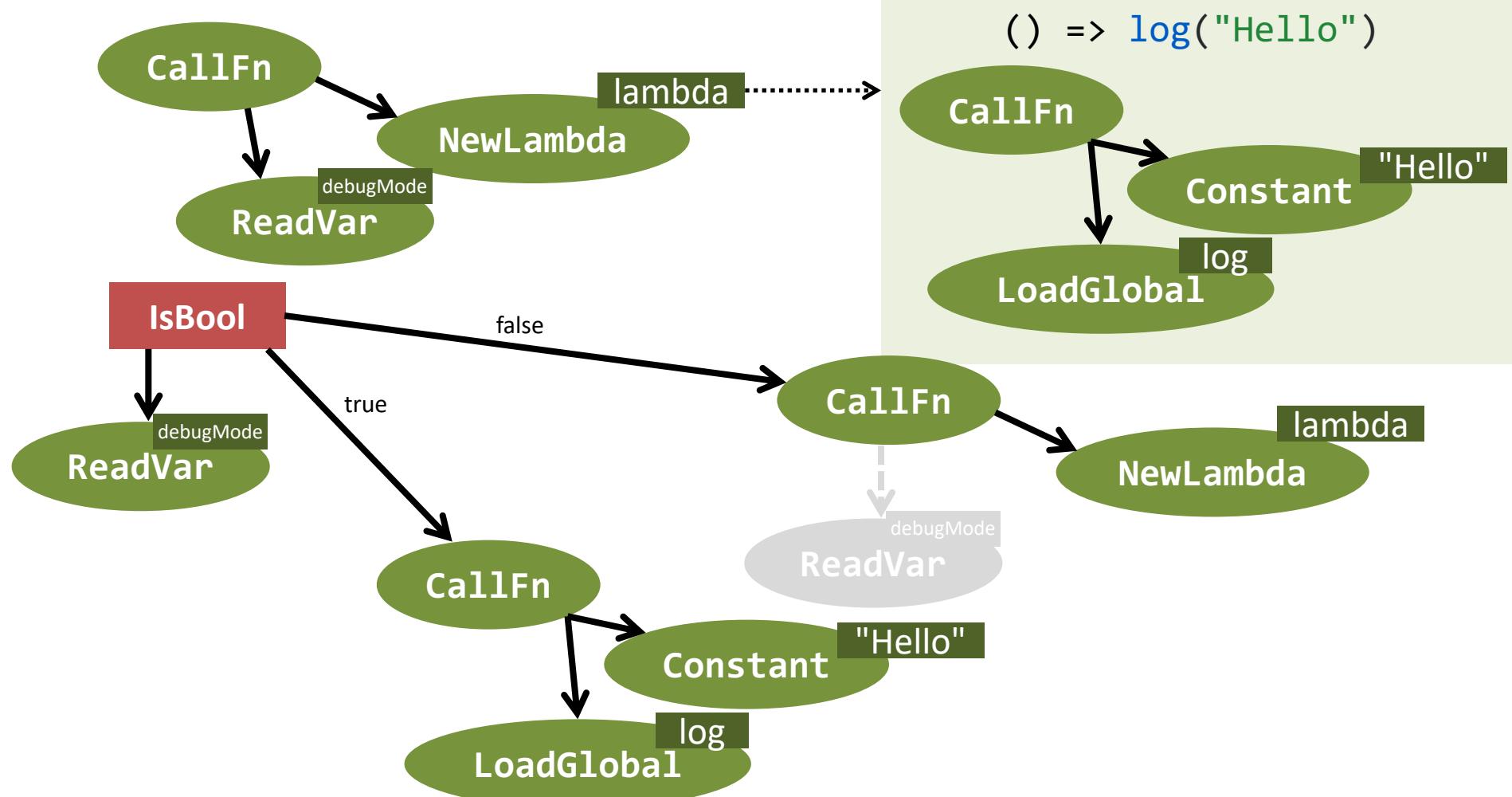
For inlining this, we need to

- Insert the guard
- Keep the old nodes as fallback
- Copy and adapt the body of

Get's more interesting  
with lexical scopes and  
variables

# Inlining ifTrue

```
debugMode.ifTrue(() => {...});
```



# Inlining ifTrue

```
debugMode.ifTrue(() => {...});
```

```
LoadVar debugMode  
NewLambda lambda  
CallFn ifTrue
```

## Inlined Version

```
LoadVar debugMode
```

```
TopIsBool  
JumpIfFalse -> fls
```

```
Pop // remove debugMode from stack
```

```
LoadGlobal "log"  
LoadConstant "Hello"  
CallFn
```

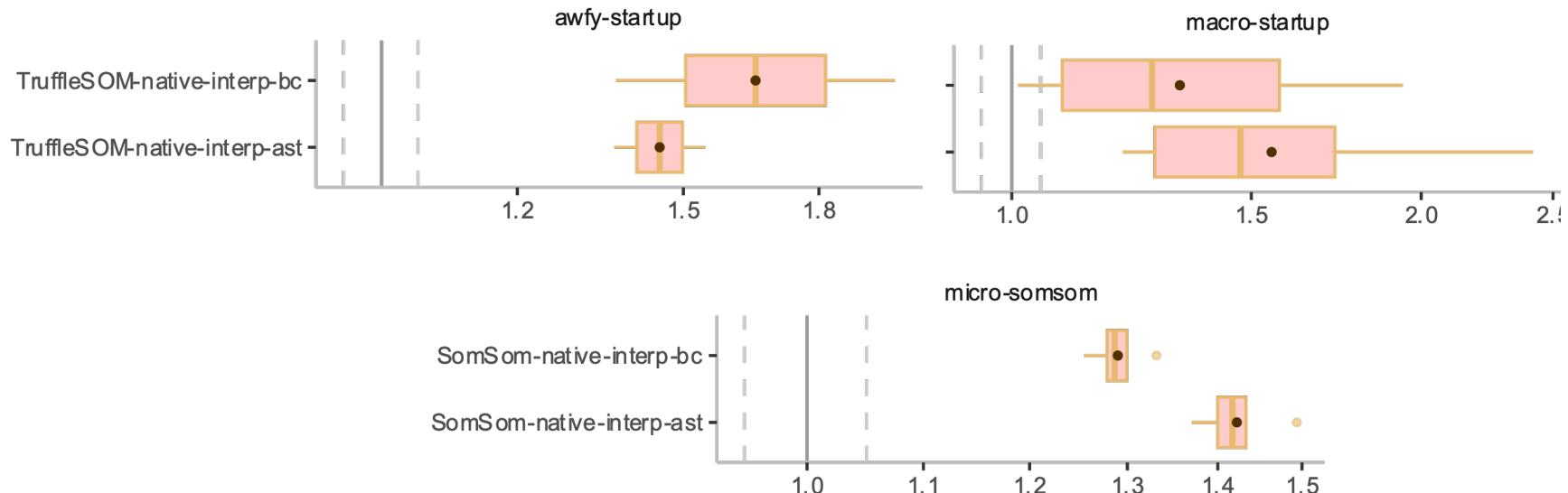
```
→ fls: NewLambda lambda  
CallFn ifTrue
```

```
() => log("Hello")
```

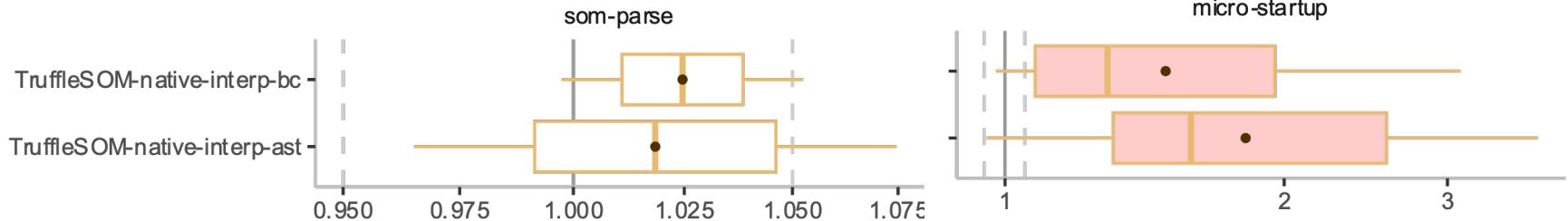
```
LoadGlobal "log"  
LoadConstant "Hello"  
CallFn
```

Looping gets more  
complicated,  
but works similarly

# Performance Benefit for the Interpreter



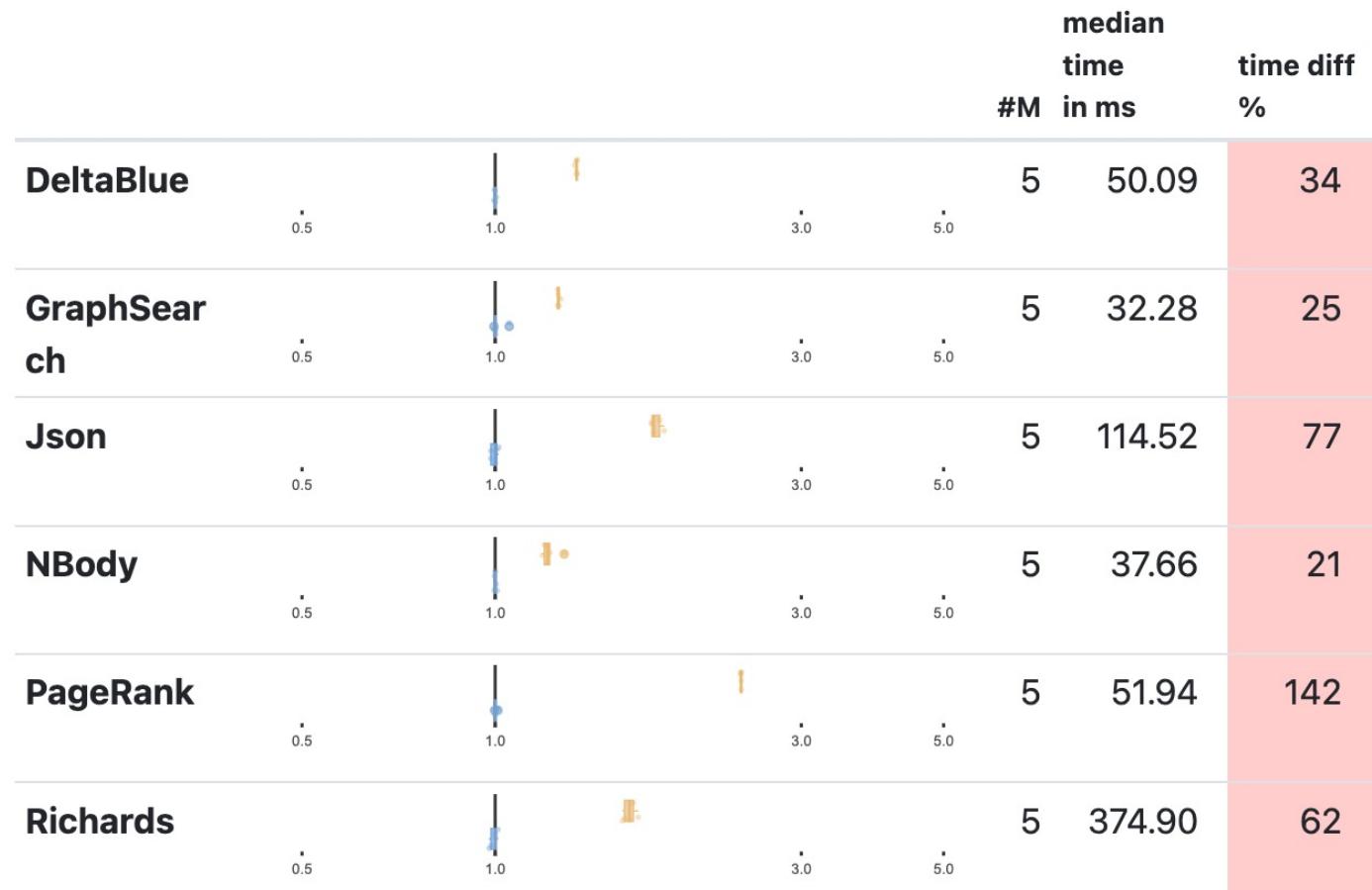
Larger Benchmarks take about 1.4x more time without Inlining of Control Structures



# Some More Details

## macro-startup

Executor: TruffleSOM-native-interp-ast



# micro-startup

Executor: TruffleSOM-native-interp-ast

	median time	#M in ms			time diff %	
		0.5	1.0	3.0	5.0	
Bounce	1.0	1.0	1.0	3.0	5.0	5 61.48 31
BubbleSort	1.0	1.0	1.0	3.0	5.0	5 69.00 161
Dispatch	1.0	1.0	1.0	3.0	5.0	5 64.02 62
Fannkuch	1.0	1.0	1.0	3.0	5.0	5 49.87 124
Fibonacci	1.0	1.0	1.0	3.0	5.0	5 80.89 55
FieldLoop	1.0	1.0	1.0	3.0	5.0	5 28.53 54
IntegerLoop	1.0	1.0	1.0	3.0	5.0	5 41.26 5
List	1.0	1.0	1.0	3.0	5.0	5 29.91 118
Loop	1.0	1.0	1.0	3.0	5.0	5 220.85 219
Mandelbrot	1.0	1.0	1.0	3.0	5.0	5 76.28 201
Permute	1.0	1.0	1.0	3.0	5.0	5 94.17 24
Queens	1.0	1.0	1.0	3.0	5.0	5 64.69 50
QuickSort	1.0	1.0	1.0	3.0	5.0	5 97.02 155
Recurse	1.0	1.0	1.0	3.0	5.0	5 70.86 53
Sieve	1.0	1.0	1.0	3.0	5.0	5 101.11 205
Storage	1.0	1.0	1.0	3.0	5.0	5 57.33 25
Sum	1.0	1.0	1.0	3.0	5.0	5 109.92 159
Test	1.0	1.0	1.0	3.0	5.0	10 237.39 6
TestGC	1.0	1.0	1.0	3.0	5.0	10 113.96 -4
Towers	1.0	1.0	1.0	3.0	5.0	5 38.65 32
TreeSort	1.0	1.0	1.0	3.0	5.0	5 87.05 120
WhileLoop	1.0	1.0	1.0	3.0	5.0	5 107.62 276

# **LIBRARY LOWERING LIBRARY INTRINSIFICATION**

# Integer class in SOM

```
class Integer {  
    = (argument) primitive  
    < (argument) primitive
```

Built into the interpreter

```
<>(argument) { return !(this = argument); }  
> (argument) { return (this >= argument).and(() => this <> argument); }  
=>(argument) { return !(this < argument); }  
<=(argument) { return (this < argument).or(() => this = argument); }  
negative() { return this < 0; }
```

```
max(otherInt) {  
    return (this < otherInt).ifTrueElse(otherInt, this);  
}
```

```
min(otherInt) {  
    return (this > otherInt).ifTrueElse(otherInt, this);  
}
```

```
}
```

Normal code, part of the standard library

# Array class in SOM

```
class Array {  
    copy() { return copy(1); }  
    copy(start) { return copy(start, this.length); }  
    copy(start, end) {  
        const result = Array.new(end - start + 1);  
        let i = 1.  
        (start..end).each((e) => {  
            result[i] = e;  
            i += 1;  
        });  
        return result;  
    }  
}
```

Array.copy, conceptually simple, but implemented by naïve library functions

How to get this to be fast?



# Implementing Library Methods in the Virtual Machine

```
class Integer {  
    <=(argument) { return (this < argument).or(() => this = argument); }  
  
    negative() { return this < 0; }  
  
    max(otherInt) {  
        return (this < otherInt).ifTrueElse(otherInt, this);  
    }  
}
```

Standard Library

```
class IntegerPrimitives:  
    def lessThanOrEqual(a: long, b: long):  
        return a <= b  
  
    def negative(a: long):  
        return a < 0  
  
    def max(a: long, b: long):  
        if a < b:  
            return b  
        return a
```

Implementation in VM

- Standard library should be “immutable”
- Changes to standard library won’t show effect
- Debuggers and profilers will show unexpected behavior
- Only works for standard library, optimization impact limited

Any drawbacks  
of this approach?



# Optimistic and Very Aggressive Library Lowering for Array class

```
createSomeArray() { return Array.new(1000, 'fast fast fast'); }
```

```
class Array {
    static new(size, lambda) {
        return new(size).setAll(lambda);
    }

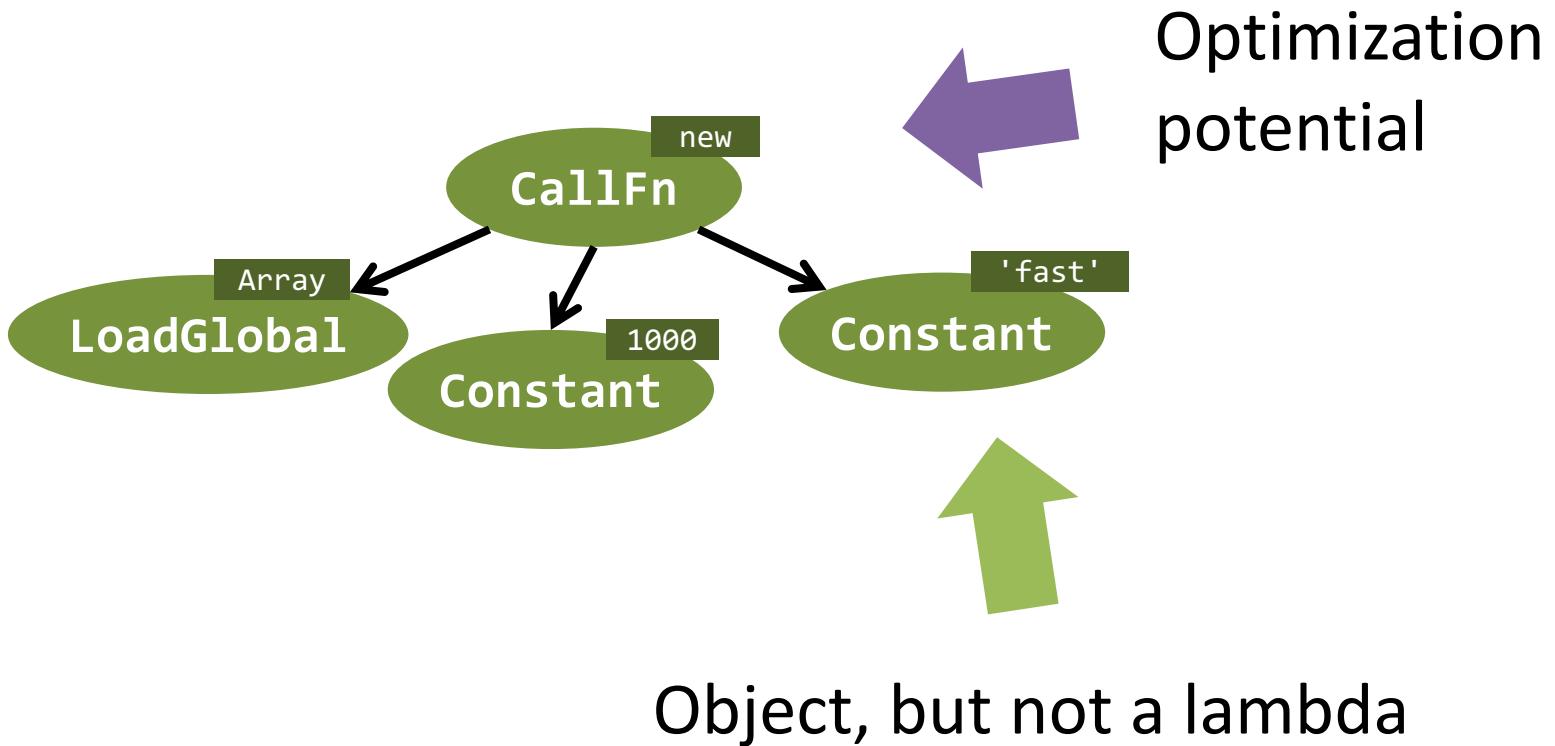
    setAll(lambda) {
        forEach((i, v) -> { this[i] = lambda.eval(); });
    }
}

class Object {
    eval() { return this; }
}
```

Some Object-Oriented Language

# Optimizing for Object Values

```
createSomeArray() { return Array.new(1000, 'fast fast fast'); }
```



# Self-optimizing new(size, lambda)

```
createSomeArray() { return Array.new(1000, 'fast fast fast'); }

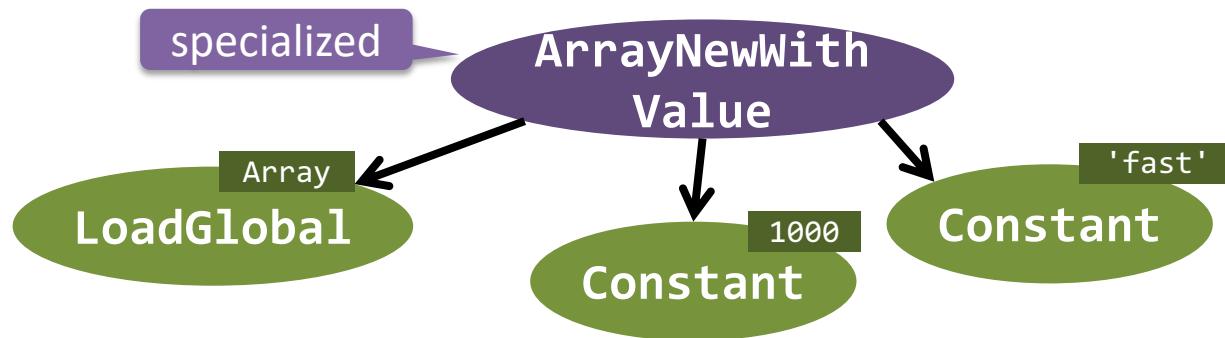
def UninitArrNew.execute(frame):
    size := size_expr.execute(frame)
    val  := val_expr.execute(frame)
    return specialize(size, val).

    execute_evaluated(frame, size, val)

def UninitArrNew.specialize(size, val):
    if val instanceof Lambda:
        return replace(StdMethodInvocation())
    else:
        return replace(ArrNewWithValue())
```

# Specialized new(size, lambda)

```
createSomeArray() { return Array.new(1000, 'fast fast fast'); }
```



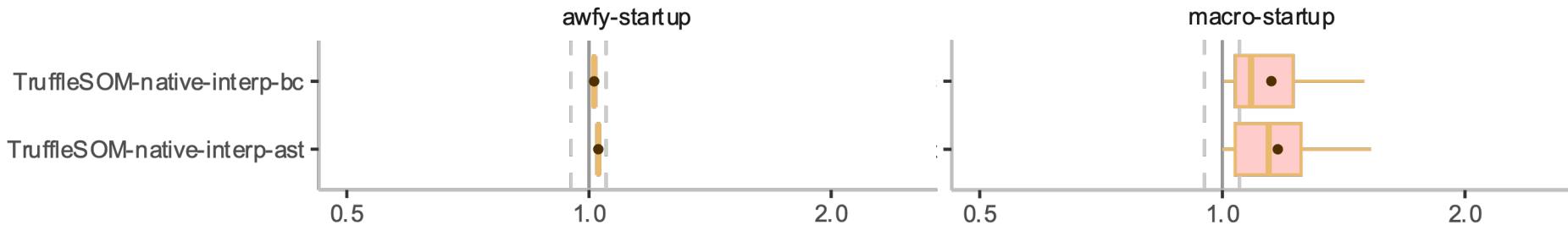
```
def ArrNewWithValue.execute_evaluated(frame, size,  
val):  
    return Array([val] * 1000)
```

1 specialized node

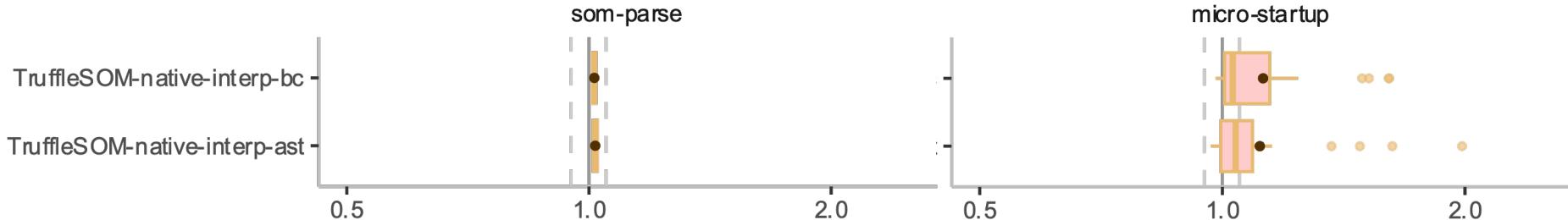
vs.

```
1000x `this[i] = lambda.eval()`  
1000x `eval() { return this; }`
```

# Performance Benefit

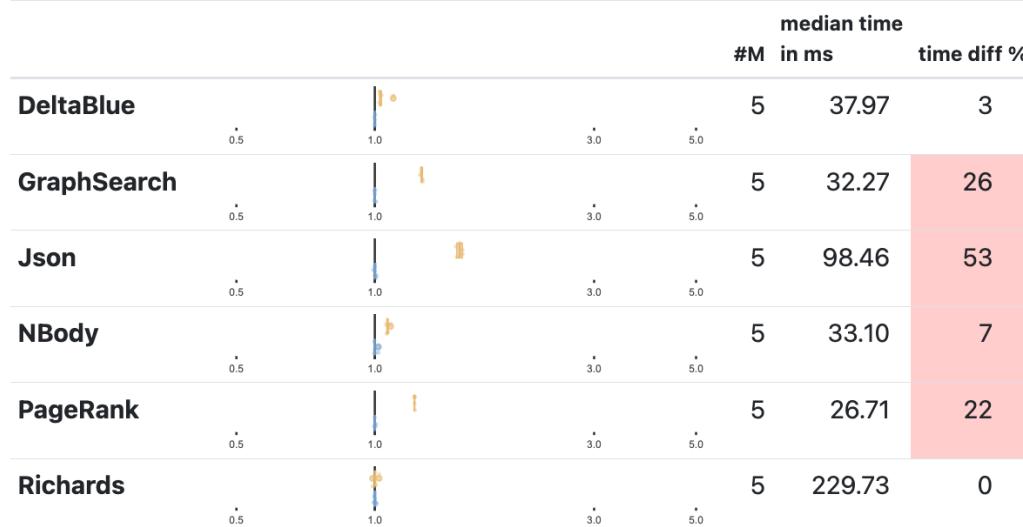


Reaching the point where optimizations are very specific.  
But still great for some benchmarks!



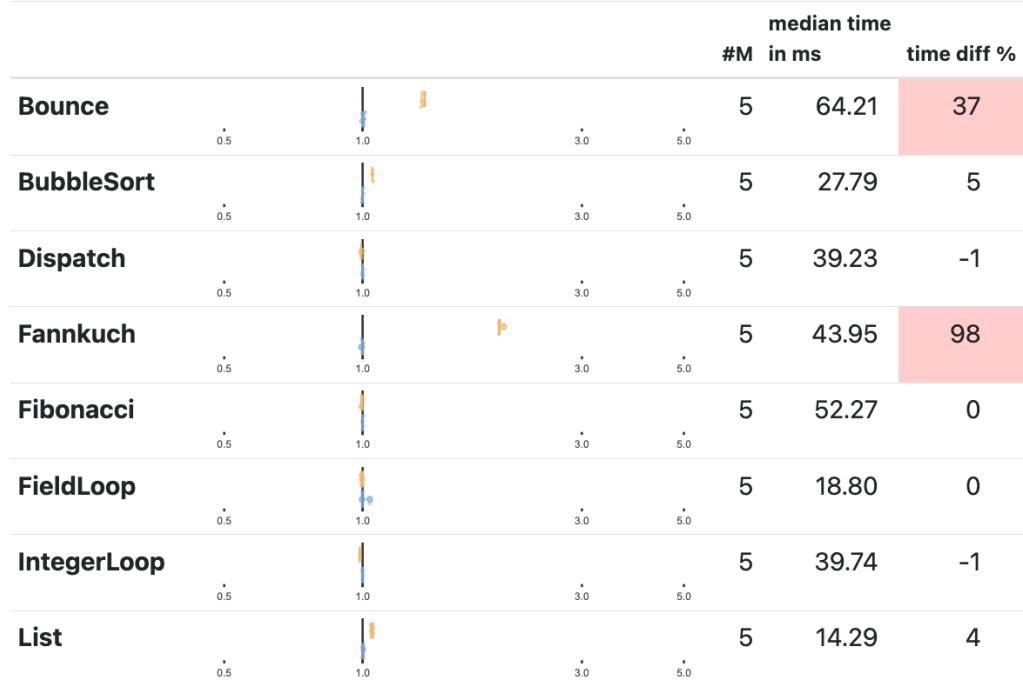
# macro-startup

Executor: TruffleSOM-native-interp-ast



# micro-startup

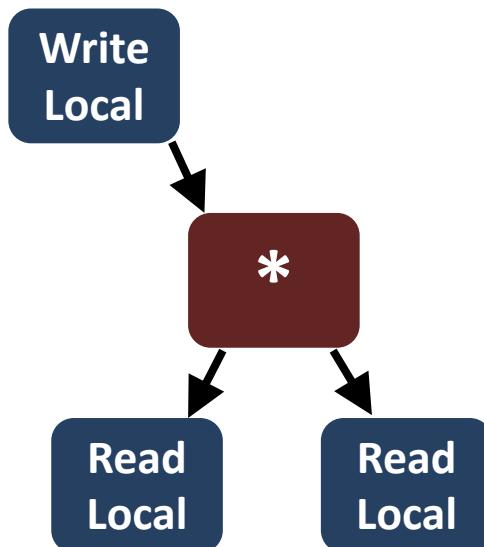
Executor: TruffleSOM-native-interp-ast



# **SUPERNODES AND SUPERINSTRUCTIONS**

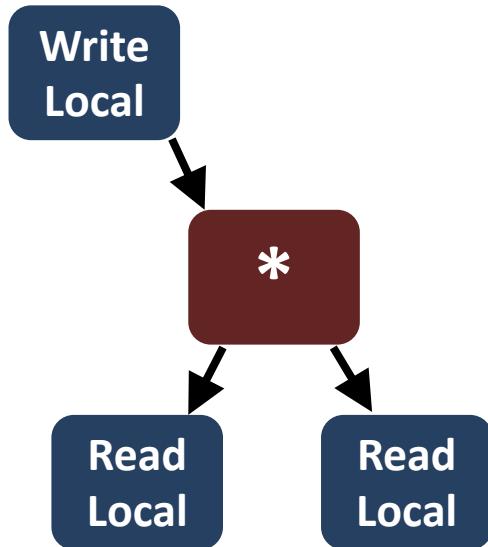
# Optimize Code Patterns

```
function mandelbrot() {  
    let zr = 0.0;  
    // while ... while ...  
    zrzr = zr * zr;  
    // ...  
}
```

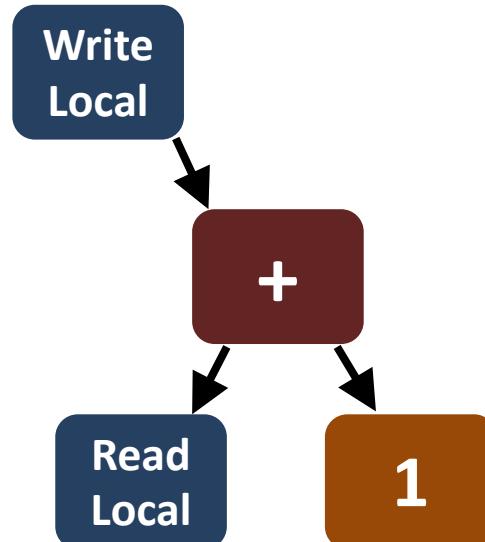


```
class ReadLocal(Node):  
    readonly slot_index  
    def execute(frame) {  
        return frame.read(slot_index)
```

# Applies to Various Node Patterns

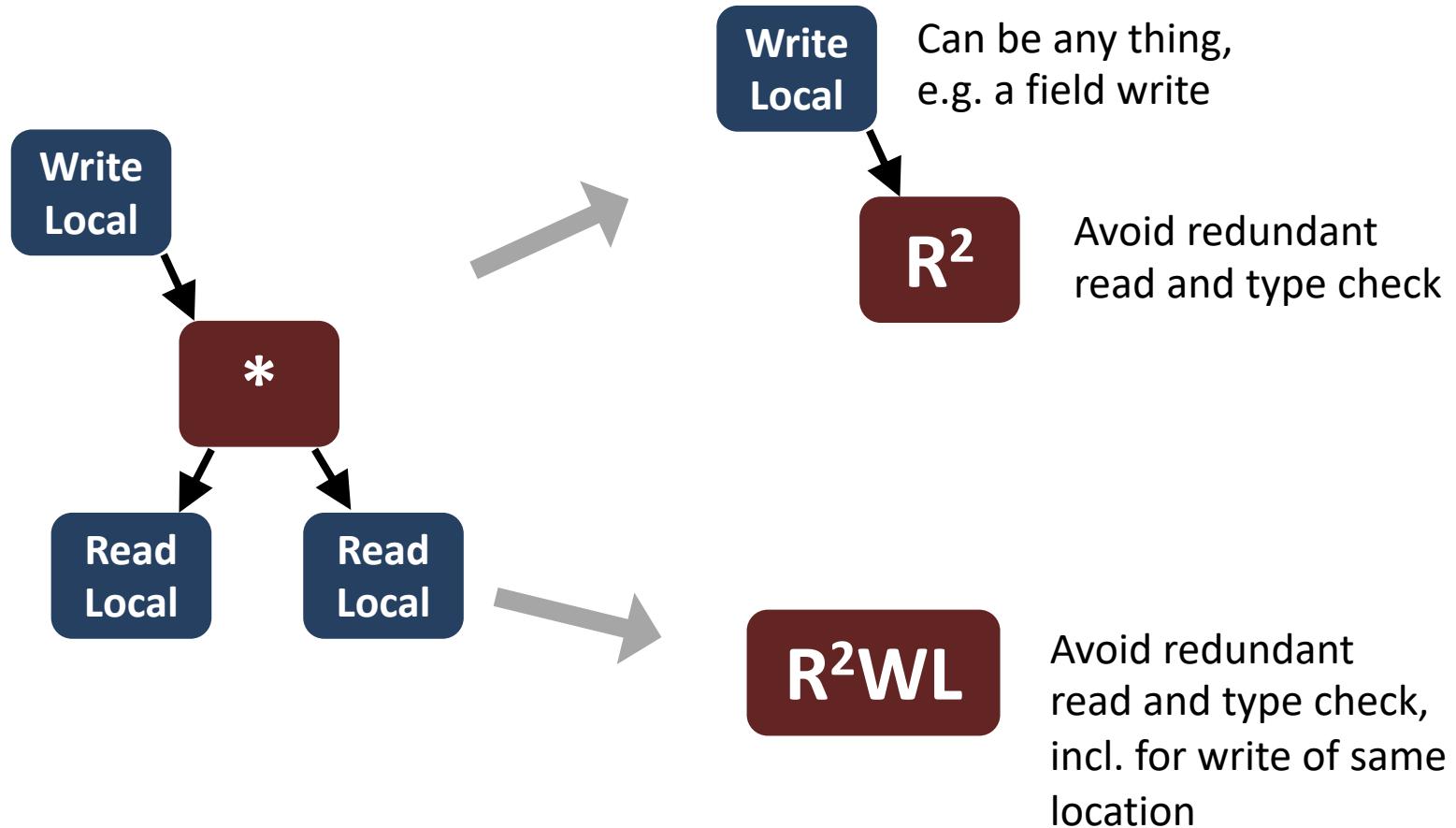


$$zr \cdot zr = zr * zr$$



$$i = i + 1$$

# Create Different Square Super Nodes for Different Situations



# 18 Super Nodes in Total

Variations of

- Method activation with **this** as argument
  - **this.method(arg1, arg2)**
- Increment/decrement operations
  - **var += n**
- Compute square
  - **var = d \* d**
- String equality with constant
  - **var == 'constant'**

Very “focused” experiment

# Results for Super Nodes

	median time in ms	time diff %
<b>DeltaBlue</b>	39.53	-2
<b>GraphSearch</b>	27.48	-3
<b>Json</b>	58.84	-14
<b>NBody</b>	33.62	-5
<b>PageRank</b>	24.94	-11
<b>Richards</b>	250.72	-11

	median time in ms	time diff %
<b>Bounce</b>	47.07	-11
<b>Dispatch</b>	36.34	-19
<b>Fibonacci</b>	49.37	-11
<b>FieldLoop</b>	19.78	-56
<b>Loop</b>	43.69	-56
<b>Mandelbrot</b>	30.52	-26
<b>Permute</b>	77.10	-10
<b>Queens</b>	45.96	-6
<b>QuickSort</b>	41.27	-9
<b>Recurse</b>	43.65	-10
<b>Sieve</b>	34.31	-15
<b>Sum</b>	32.81	-34
<b>Test</b>	184.28	-6
<b>Towers</b>	28.55	-10
<b>WhileLoop</b>	29.98	-29

- Optimizing small patterns can give a nice gain
- But not generalizable...

# For Bytecodes: Superinstructions

$i = i + 1$

bytecode to add 1



bytecode to add 1

to field



LoadField	i	LoadField	i	IncField	i
LoadConstant	1	Inc			
CallFn	+	StoreField	i		
StoreField	i				

# Research and Literature

- **Optimizing an ANSI C interpreter with superoperators**  
Proebsting, T. A. (1995). *POPL'95*
- **Combining Stack Caching with Dynamic Superinstructions**  
Ertl, M. A. & Gregg, D. (2004).  
*IVME'04*
- **Optimizing Indirect Branch Prediction Accuracy in Virtual Machine Interpreters.**  
Casey, K., Ertl, M. A. & Gregg, D.  
(2007). *TOPLAS'07*
- **Less Is More: Merging AST Nodes To Optimize Interpreters.**  
Larose, O., and Kaleba, S. & Marr,  
S. (2022). *MoreVMs'23*



# Dynamic Languages

cnt + 1

## ECMAScript Specification Sec. 11.6.1

```
left = ToPrimitive(GetValue(cnt))
right = ToPrimitive(GetValue(1))

if (IsString(left) || IsString(right)) {
    return ToString(left).concat(ToString(right))
}

return ToNumber(left) + ToNumber(right)
```



# Dynamic Languages

cnt + 1

## ECMAScript Specification Sec. 11.6.1

```
left = ToPrimitive(GetValue(cnt))  
right = ToPrimitive(GetValue(1))
```

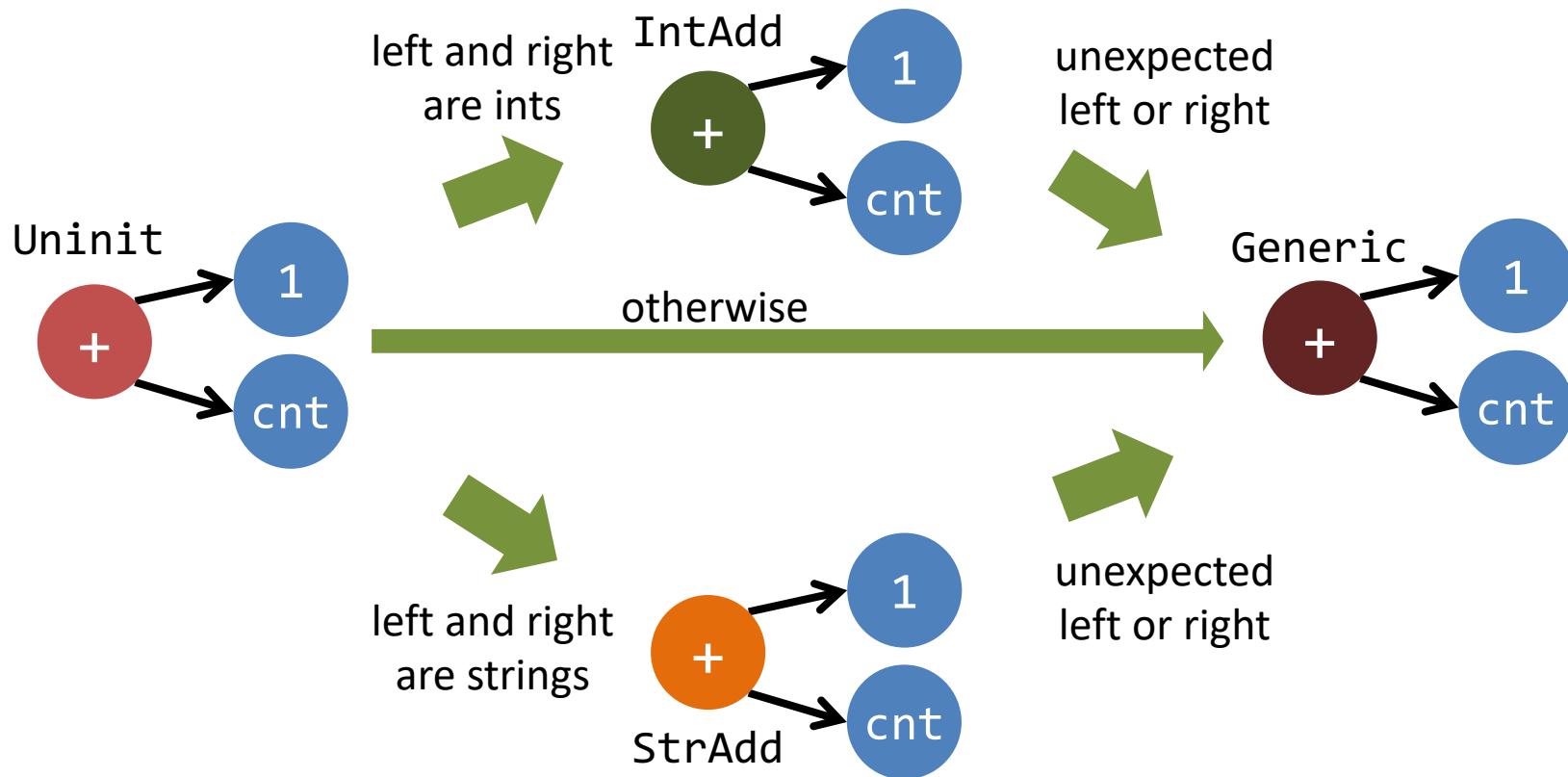
```
if (IsString(left) || IsString(right)) {  
    return ToString(left).concat Could we optimize this  
}  
                                         somehow?
```

```
return ToNumber(left) + ToNum
```



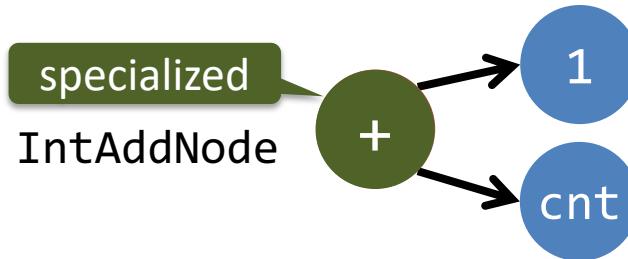
# **SPECULATIVE OPTIMIZATION**

# Self Optimization and Speculation



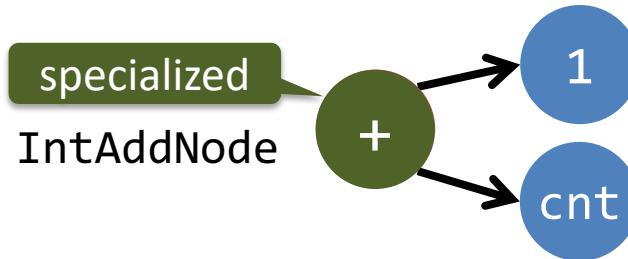
observe run-time value  
select optimization of first execution

# Self Optimization and Speculation



```
class UninitAdditionNode(BinaryNode):
    def execute(frame):
        lVal = left.execute(frame)
        rVal = right.execute(frame)
        if type(lVal) == int and type(rVal) == int:
            return replace(IntAddNode(self)).
                execute_evaluate(lVal, rVal)
```

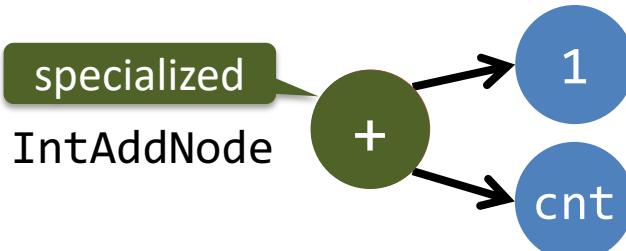
# Self Optimization and Speculation



```
class IntAddNode(BinaryNode):
    def execute(frame):
        lVal = left.execute(frame)
        rVal = right.execute(frame)
        if type(lVal) == int and type(rVal) == int:
            return lVal + rVal
        else:
            return deoptimize(lVal,
```

Simplified Version

# Self Optimization and Speculation



```
class IntAddNode(BinaryNode):
    def execute(frame):
        try:
            lVal = left.execute_int(frame)
        except UnexpectedResult, exp:
            return deoptimize(exp)
        try:
            rVal = right.execute_i
        except UnexpectedResult,
            return deoptimize(lVal)
        return lVal + rVal
```

Check moved down  
tree, and potentially  
eliminated

# Bytecode Quicken

$i = i + 1$

run with  $i$  being an int



LoadField     $i$   
LoadConstant  $1$   
**IntInc**  
StoreField     $i$

run with  $i$  being a str



LoadField     $i$   
LoadConstant  $1$   
CallFn         $+$   
StoreField     $i$

LoadField     $i$   
LoadConstant  $1$   
**StrAdd**  
StoreField     $i$

Rewriting constrained by  
bytecode format!  
And the linear format!

# Some Possible Self-Optimizations

- Type profiling and specialization

$\text{cnt} + 1$



- Lookup caching
- Value caching
- Operation inlining
- Library Lowering

→ function



# Summary

## Self Optimization

- For AST interpreters
- Specialize nodes
  - Fast run-time check
  - Avoid complex general case
  - A *local* optimization
- Uses deoptimization to fall back to general case

## Speculation

- Assume future behavior is same as past behavior
- Requires fallback strategy (deoptimization)
- Used with self optimization, quickening, and JIT compilation

# Research and Literature

- **AST interpreters**
  - **Self-Optimizing AST Interpreters**  
Würthinger, T., Wöß, A., Stadler, L., Duboscq, G., Simon, D. & Wimmer, C. (2012). *DLS'12*
- **Bytecode interpreters**
  - **Efficient Interpretation Using Quicken**  
Brunthaler, S. (2010). *DLS'10*
  - **Optimizing Indirect Branch Prediction Accuracy in Virtual Machine Interpreters**  
Casey, K., Ertl, M. A. & Gregg, D. (2007). *ACM Trans. Program. Lang. Syst.*, 29, 37.
  - **Multi-Level Quicken: Ten Years Later.**  
Brunthaler, S. (2021)  
arXiv:2109.02958v1
- **JIT compilers, deoptimization**
  - **An Efficient Implementation of SELF a Dynamically-Typed Object-Oriented Language Based on Prototypes.**  
Chambers, C., Ungar, D. & Lee, E. (1989). *OOPSLA'89*



# AST versus Bytecode



# Which one is faster?



Abstract Syntax Tree  
Interpreters



# Which one **is faster?**



Bytecode  
Interpreters

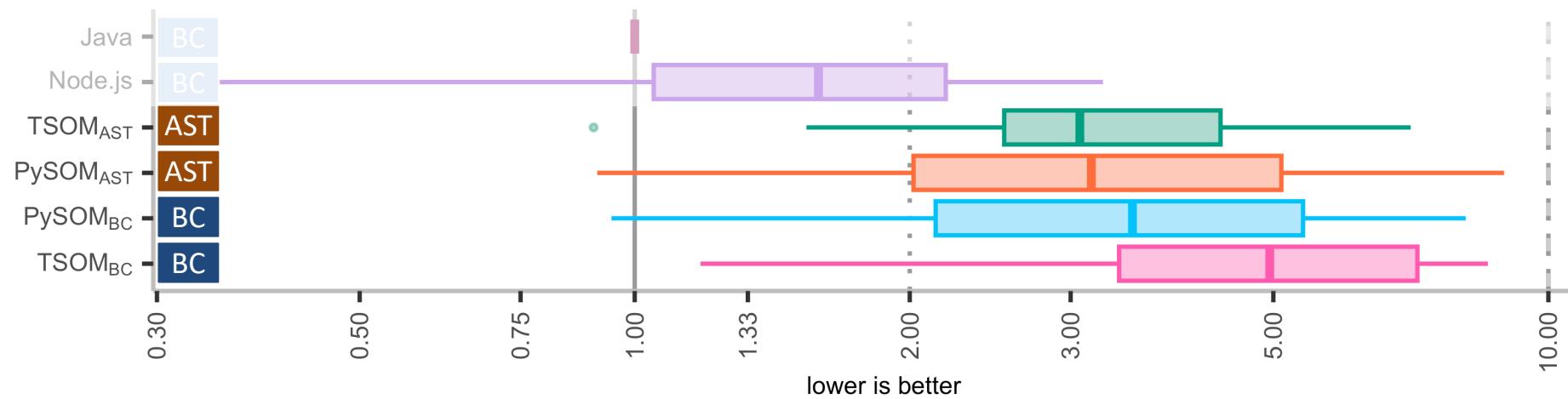


# Disclaimer!

Through the lens of Meta-compilation Systems  
Graal+Truffle and RPython



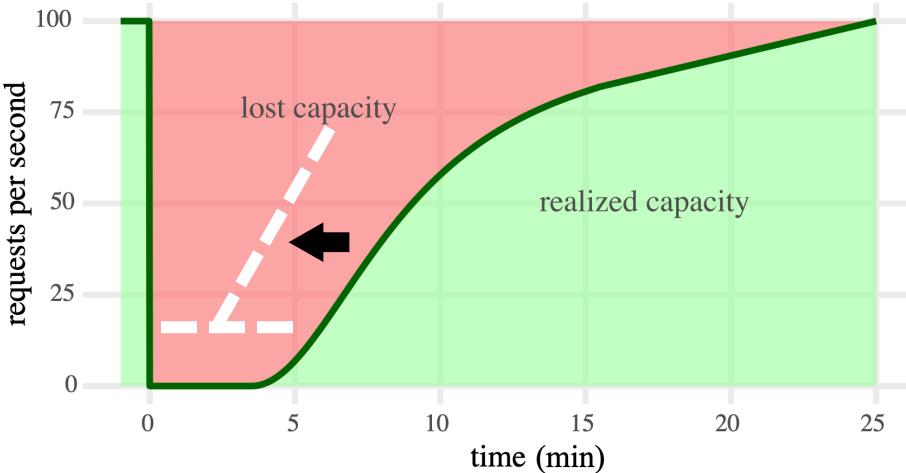
# AST vs Bytecode Interpreters for Metacompilation Systems



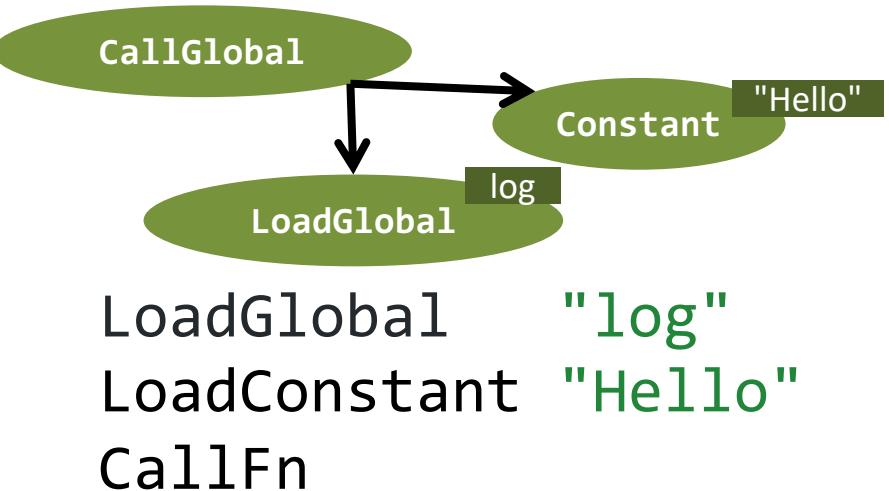
AST interpreters can be surprisingly fast!  
Though, bytecodes are much more compact in memory.

# WARP UP

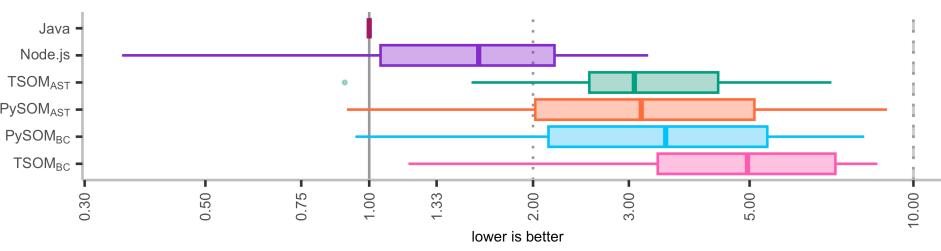
Interpreter performance is critical for overall performance



Most are AST or Bytecode Interpreters



Can reach similar performance (in metacompilation systems)



Lookup Caching is Generic, and Highly Effective

w • `fitsTnto(width)`  
function

Inlining also effective,  
but gets complicated quickly

Library Lowering/Intrinsicsification,  
Supernodes/instructions,  
... get more and more specific  
Should be driven by need!