

ON THE COST OF TYPE-TAG SOUNDNESS

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ON THE COST OF TYPE-TAG SOUNDNESS

1. Tag soundness
2. Performance cost of soundness
3. Evaluation method
4. Conclusions

TYPE-TAG SOUNDNESS

Type Soundness

If $\vdash e : \tau$ then either:

- $e \xrightarrow{*} v$ and $\vdash v : \tau$
- e diverges
- $e \xrightarrow{*} \text{Error}$ (division by zero, etc.)

No undefined behavior
Type-based reasoning

Type Soundness

If $\vdash e : \tau$ then either:

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Tag Soundness

If $\vdash e : \tau$ then either:

- $e \xrightarrow{*} v$ and $\vdash v : \lfloor \tau \rfloor$
- e diverges
- $e \xrightarrow{*} \text{Error}$ (division by zero, etc.)

Tag Soundness

If $\vdash e : \tau$ then either:

- $e \xrightarrow{*} v$ and $\vdash v : \lfloor \tau \rfloor$
- e diverges
- $e \xrightarrow{*} \text{Error}$ (division by zero, etc.)

$$\lfloor \tau \rfloor = K$$

$$\lfloor \text{Int} \rfloor = \text{Int}$$

$$\lfloor \tau \times \tau' \rfloor = \text{Pair}$$

$$\lfloor \tau \rightarrow \tau' \rfloor = \text{Fun}$$

...

Tag Soundness

If $\vdash e : \tau$ then either:

- $e \xrightarrow{*} v$ and $\vdash v : \lfloor \tau \rfloor$
- e diverges
- $e \xrightarrow{*} \text{Error}$

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...

No undefined behavior
Tag-based reasoning

Types vs. Tags

If $\vdash e : \text{Int} \times \text{Int}$ and $e \rightarrow^* v$ then v might be:

Type Soundness

(0, 0)

(3, 2)

(-7, 9)

Tag Soundness

(0, 0)

(3, 2)

(-7, 9)

("A", 0)

(0, (1, 2))

Types vs. Tags

If $\vdash e : \text{Int} \times \text{Int}$ and $e \rightarrow^* v$ then v might be:

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(3, 2)

(-7, 9)

("A", 0)

(0, (1, 2))



* **fast**

* **slow**

Type Sound?



Tag Sound?

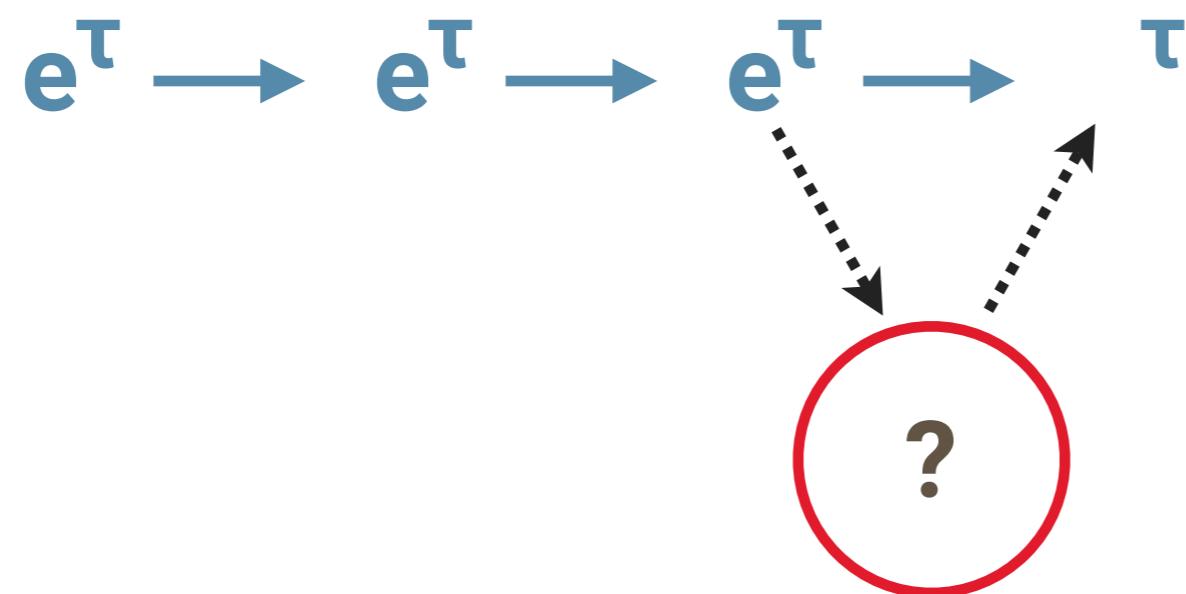


→* **fast**

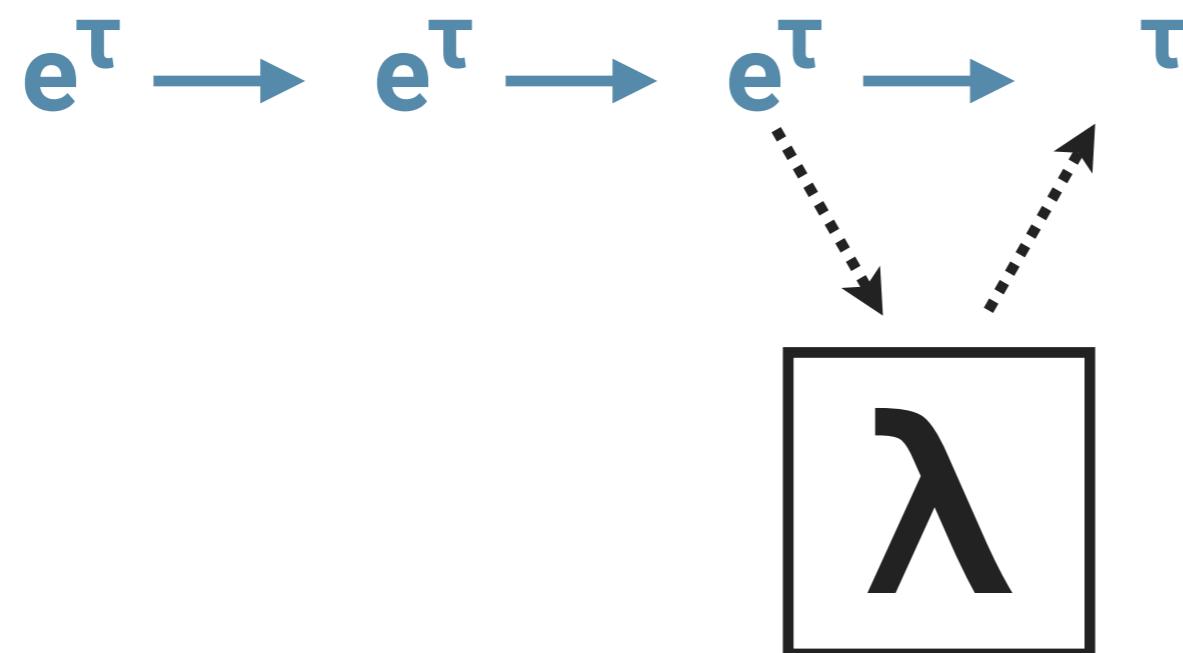
→* **slow**

PERFORMANCE COST OF SOUNDNESS

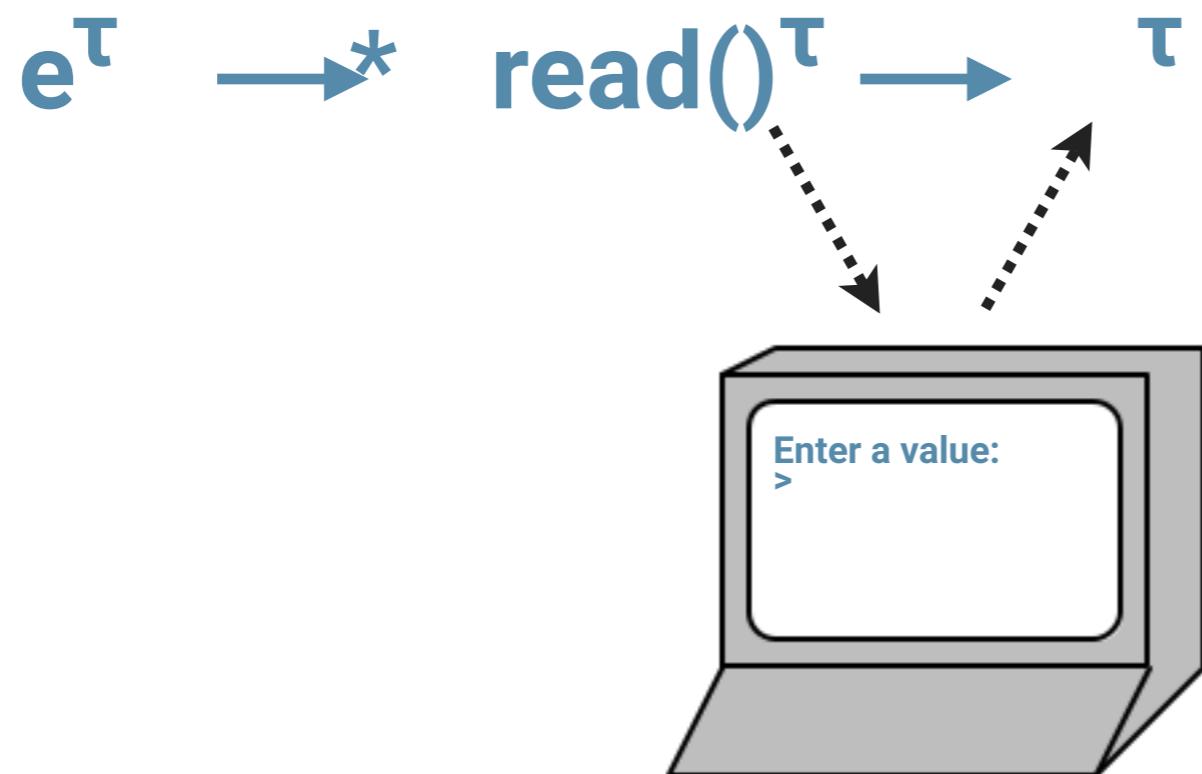
Problem: Safe Interaction



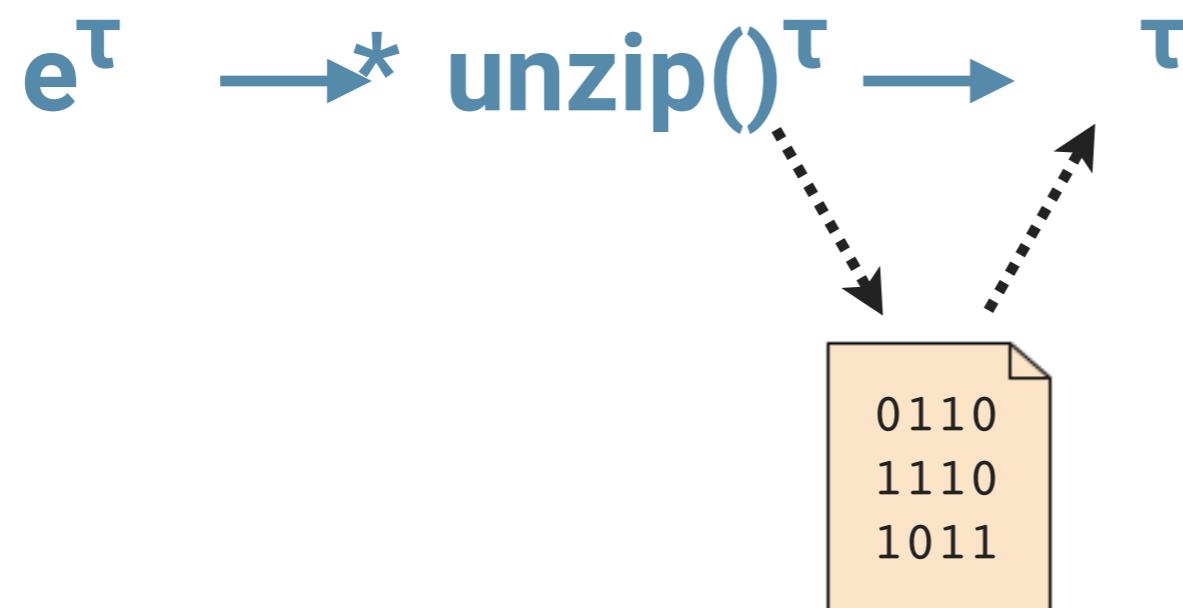
Gradual Typing



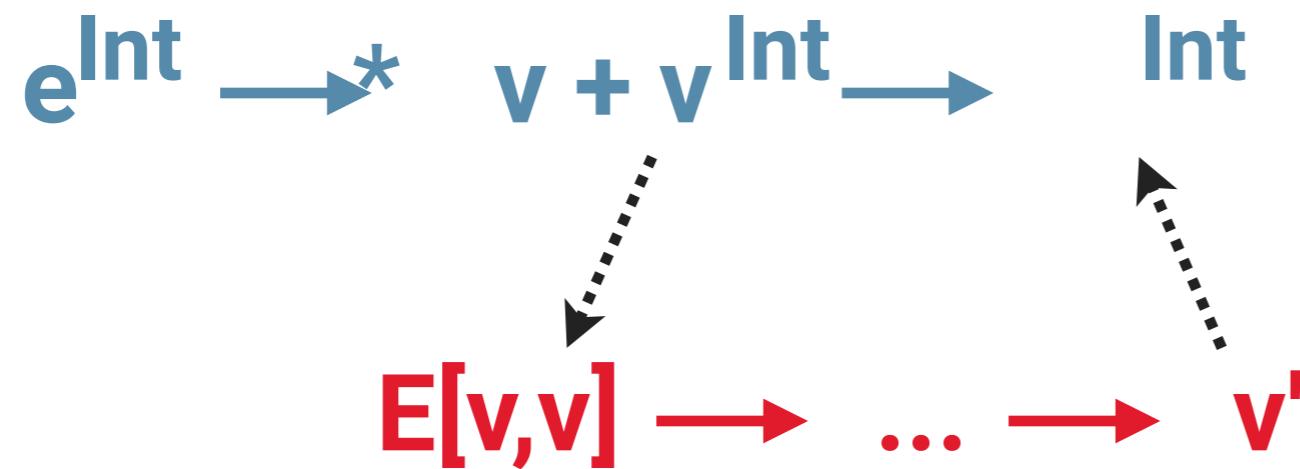
User Input



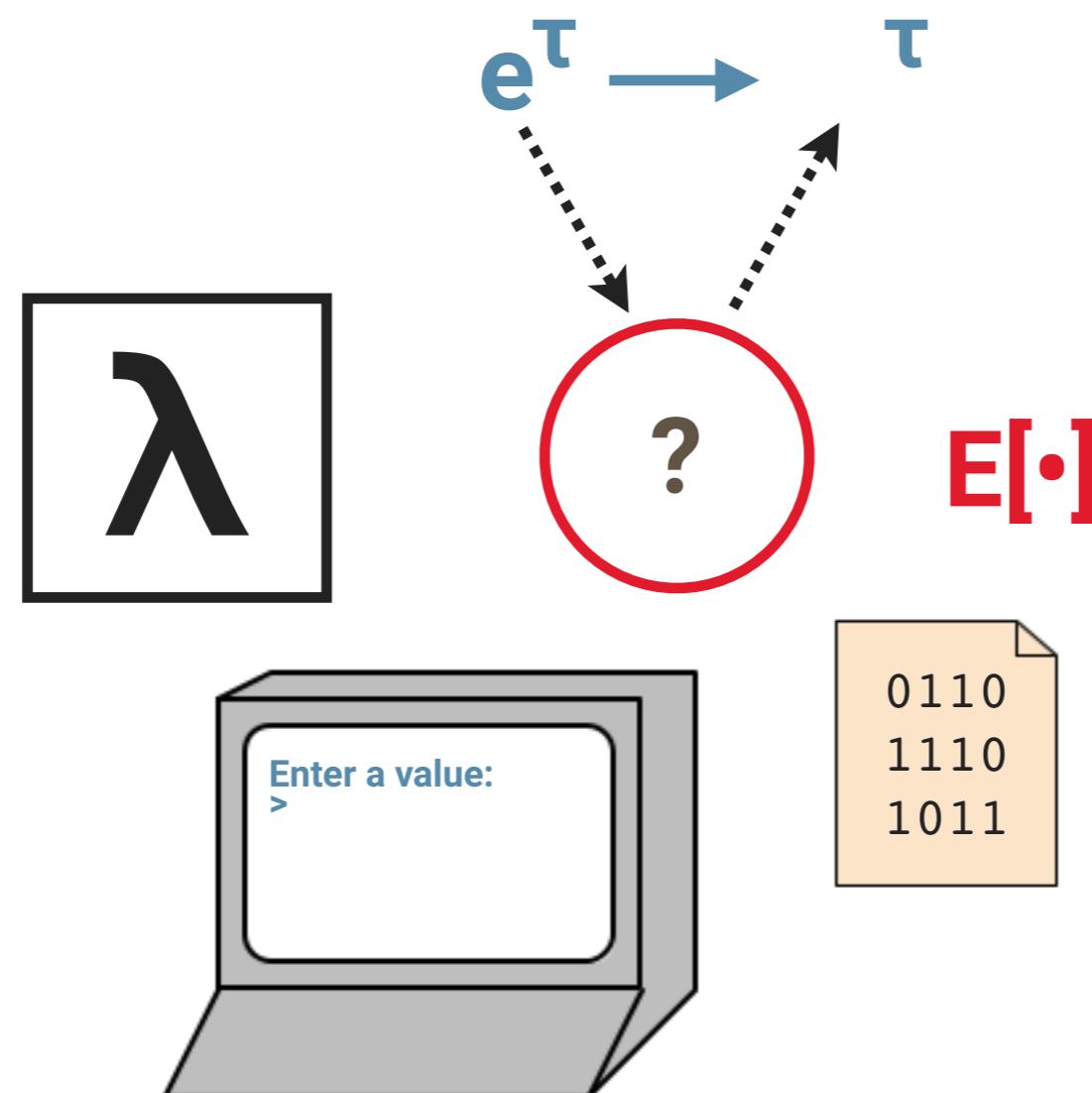
Deserialization



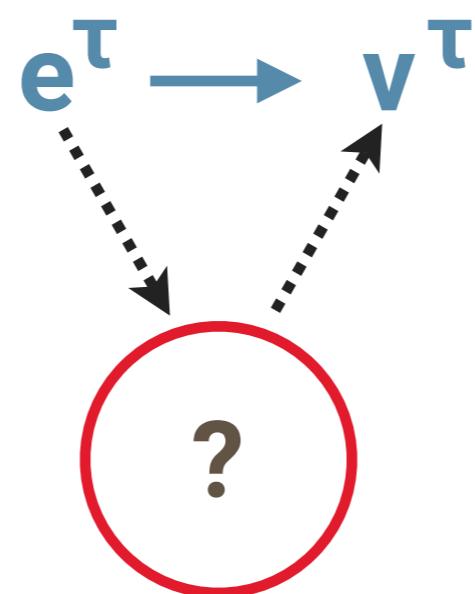
Primitive Operations (δ)



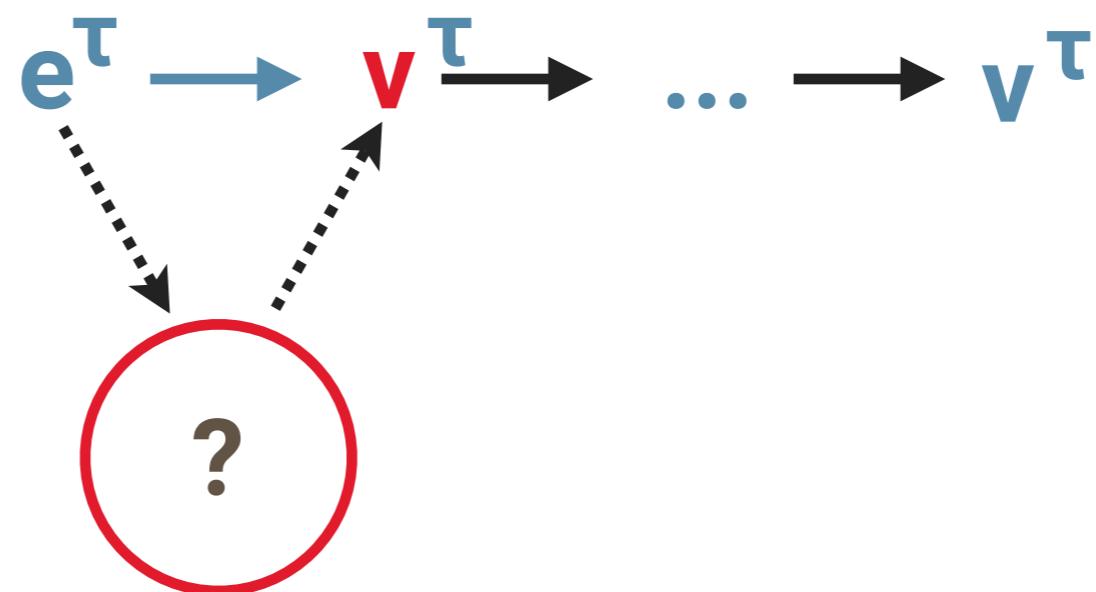
Unreliable Source



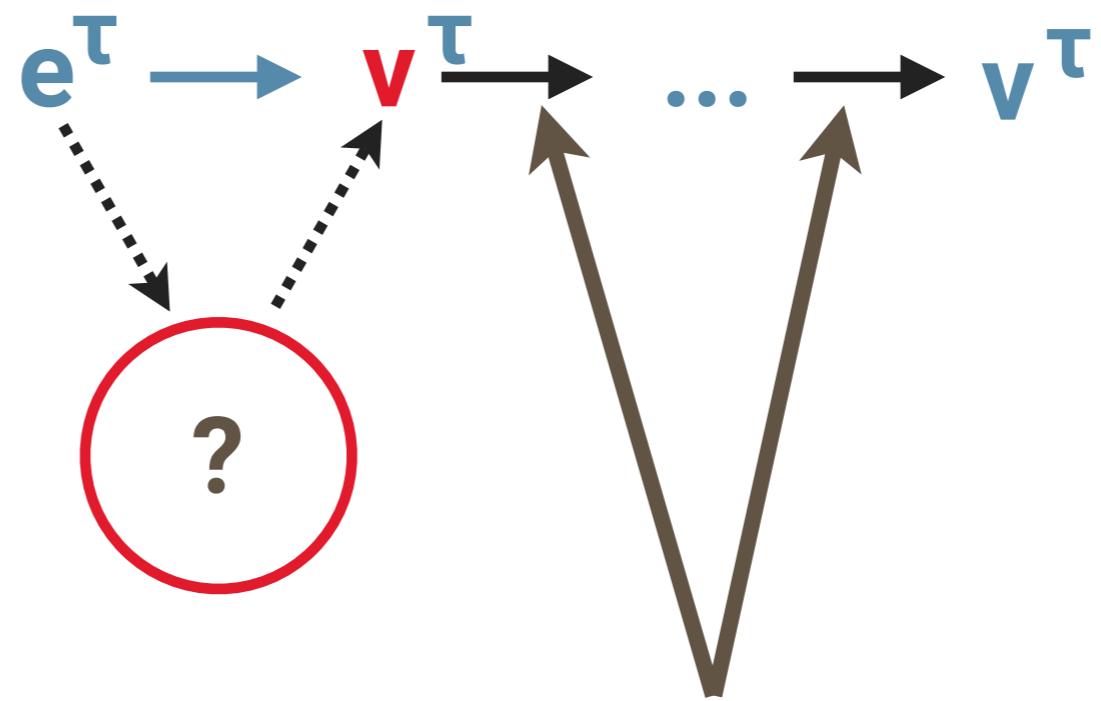
Option 1: Trust



Option 2: Check

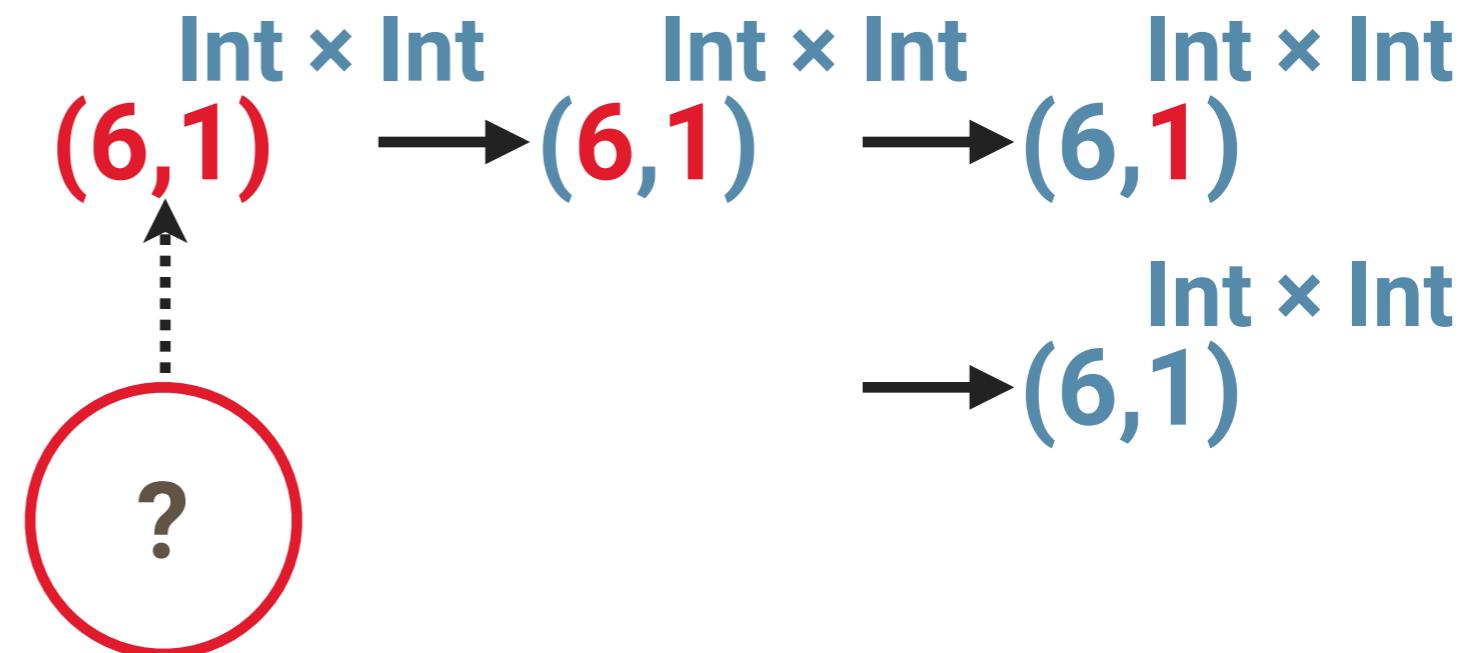


Option 2: Check

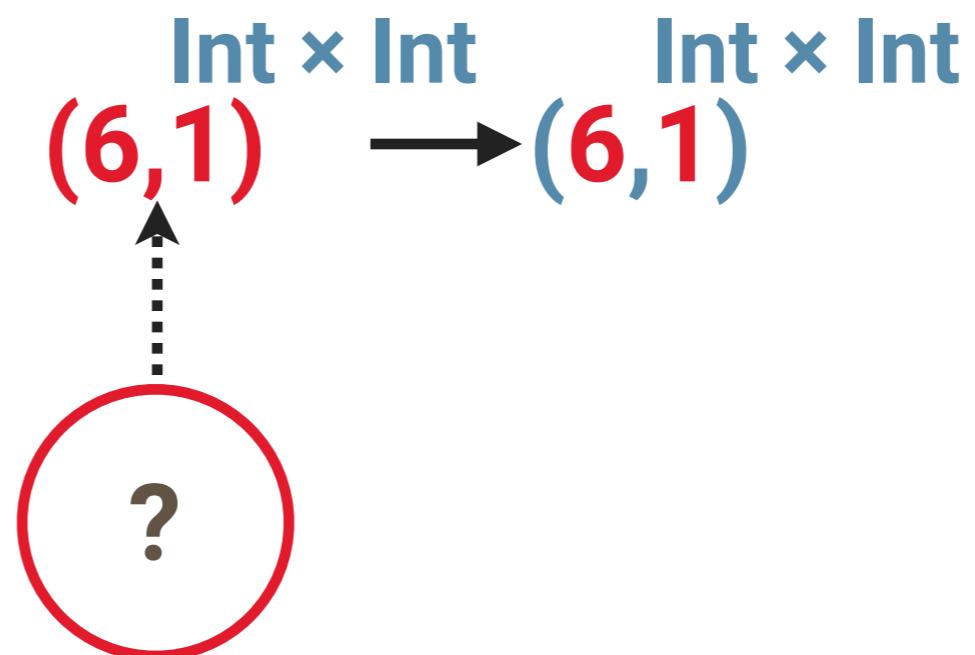


COST OF SOUNDNESS

Cost of Types ($\xrightarrow{*} \text{slow}$)

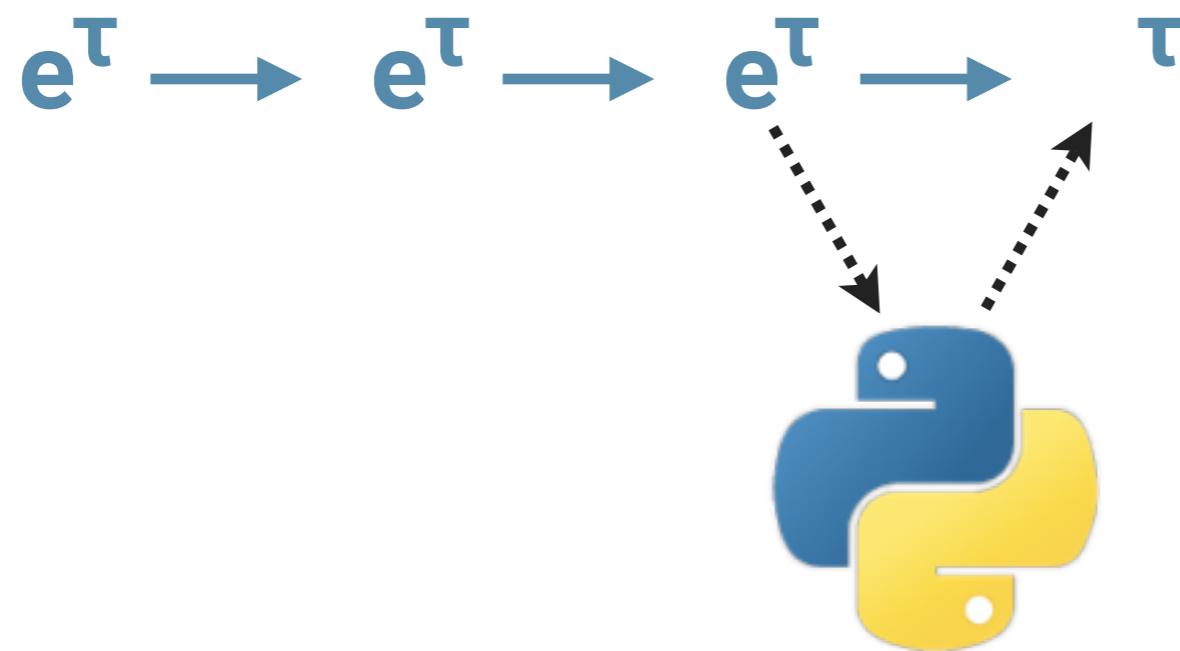


Cost of Tags ($\xrightarrow{*} \text{fast}$)



COST OF SOUNDNESS IN RETICULATED

Retic vs. Python



Reticulated

```
def dist(pt : Tuple[Int, Int]) -> Int:  
    x = pt[0]  
    y = pt[1]  
    return abs(x + y)
```

Reticulated

```
def dist(pt : Tuple[Int, Int]) -> Int:  
    x = pt[0]  
    y = pt[1]  
    return abs(x + y)
```

dist((0, 0)) →* 0

Reticulated

```
def dist(pt : Tuple[Int, Int]) -> Int:  
    x = pt[0]  
    y = pt[1]  
    return abs(x + y)
```

dist("NaN") →* Expected Tuple

Reticulated

```
def dist(pt : Tuple[Int, Int]) -> Int:  
    x = pt[0]  
    y = pt[1]  
    return abs(x + y)
```

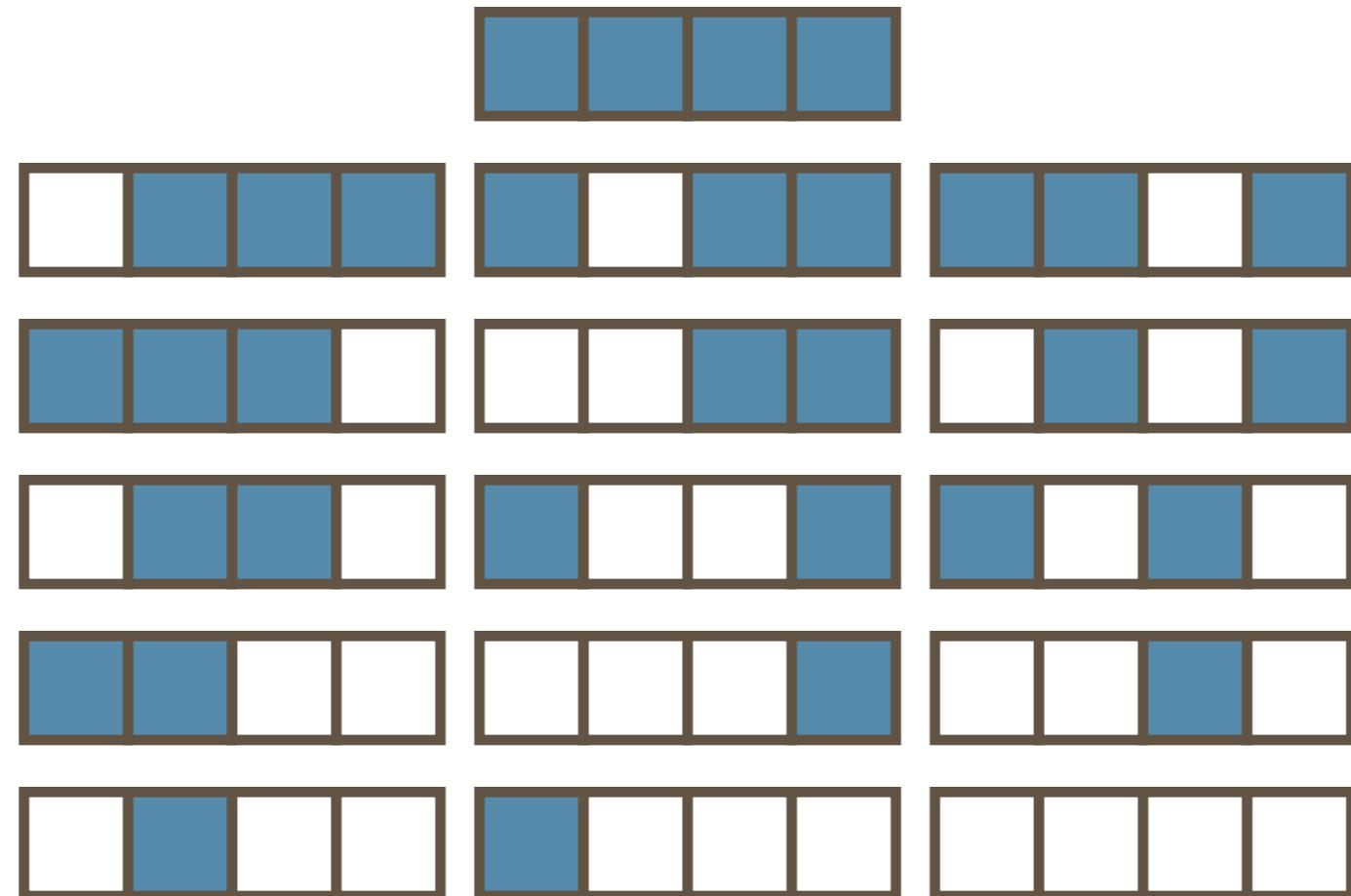
dist((0, "NaN")) →* Expected Int

Evaluation Method

1. Fully-Typed



2. Configurations



3. Measure

	11s	
7s	9s	2s
5s	24s	9s
14s	5s	21s
9s	6s	9s
8s	4s	5s

What % have at most Dx overhead?

D = 4, vs.



11s

7s

9s

2s

5s

24s

9s

14s

5s

21s

9s

6s

9s

8s

4s

5s

What % have at most
Dx overhead?

D = 4, vs.



Evaluation Method



EXPERIMENT & RESULTS

Benchmarks

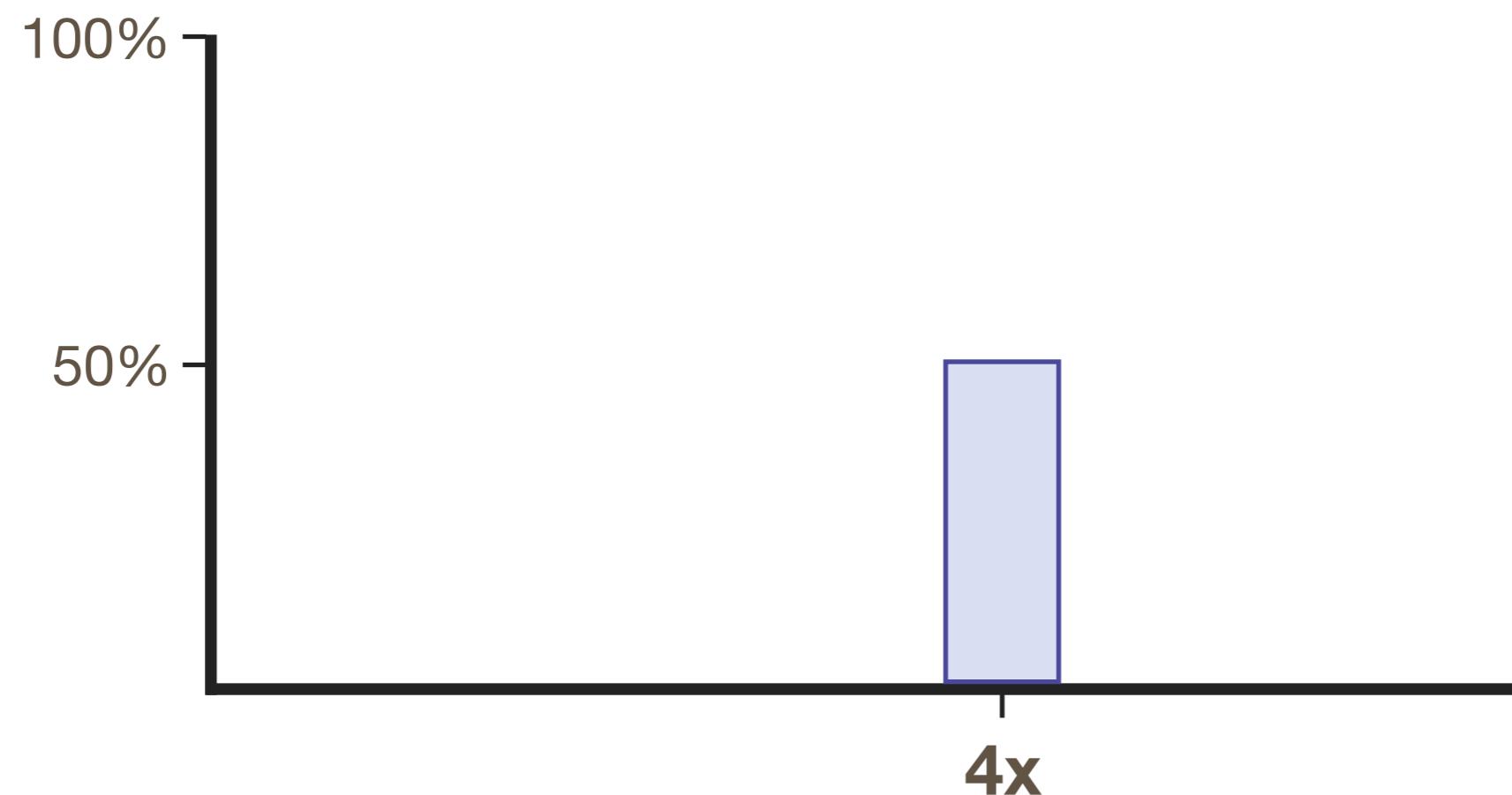
DLS 2014	POPL 2017	PEPM 2018
futen	call_method	espionage
http2	call_simple	pythonflow
slowSHA	chaos	take5
aespython	fannkuch	sample_fsm
stats	go	
	meteor	
	nbody	
	nqueens	
	pidigits	
	pystone	
	spectralnorm	

Typed Components

DLS 2014	POPL 2017	PEPM 2018
15	7	12
4	6	12
17	15	16
34 *	1	19 *
79 *	7	
	8	
	5	
	2	
	5	
	14	
	5	

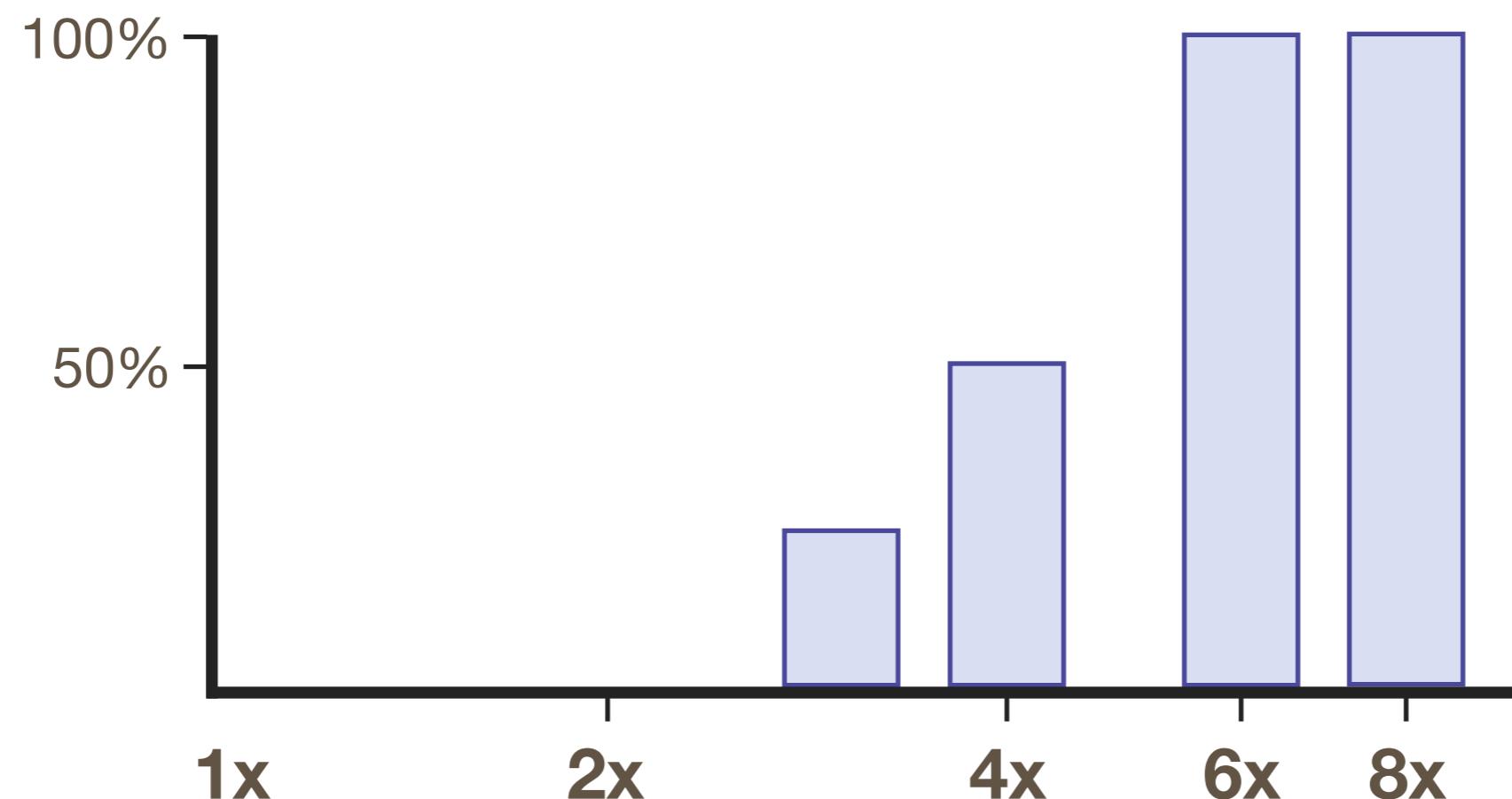
Exhaustive Results

What % of configurations have at most **4x** overhead?



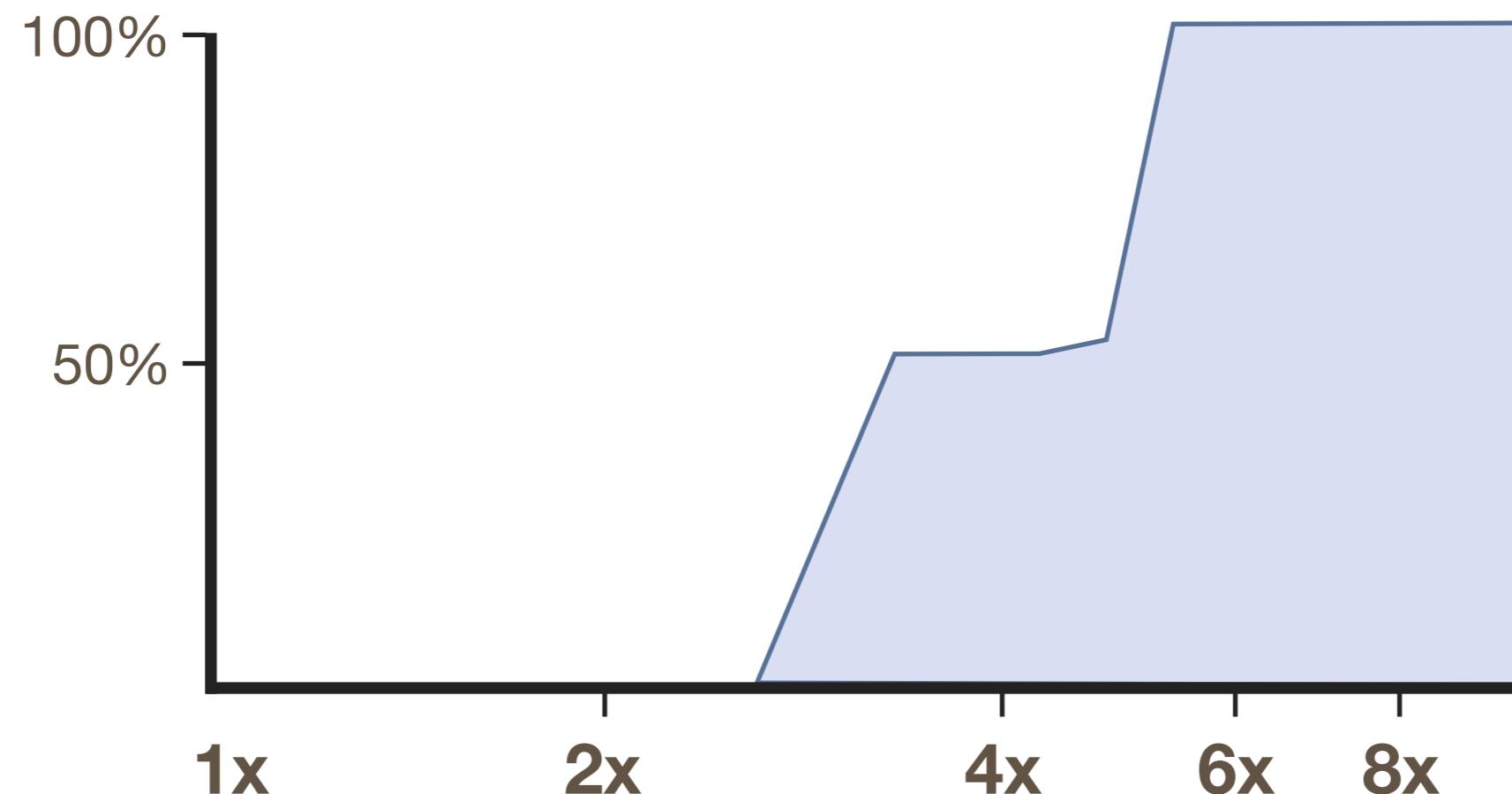
Exhaustive Results

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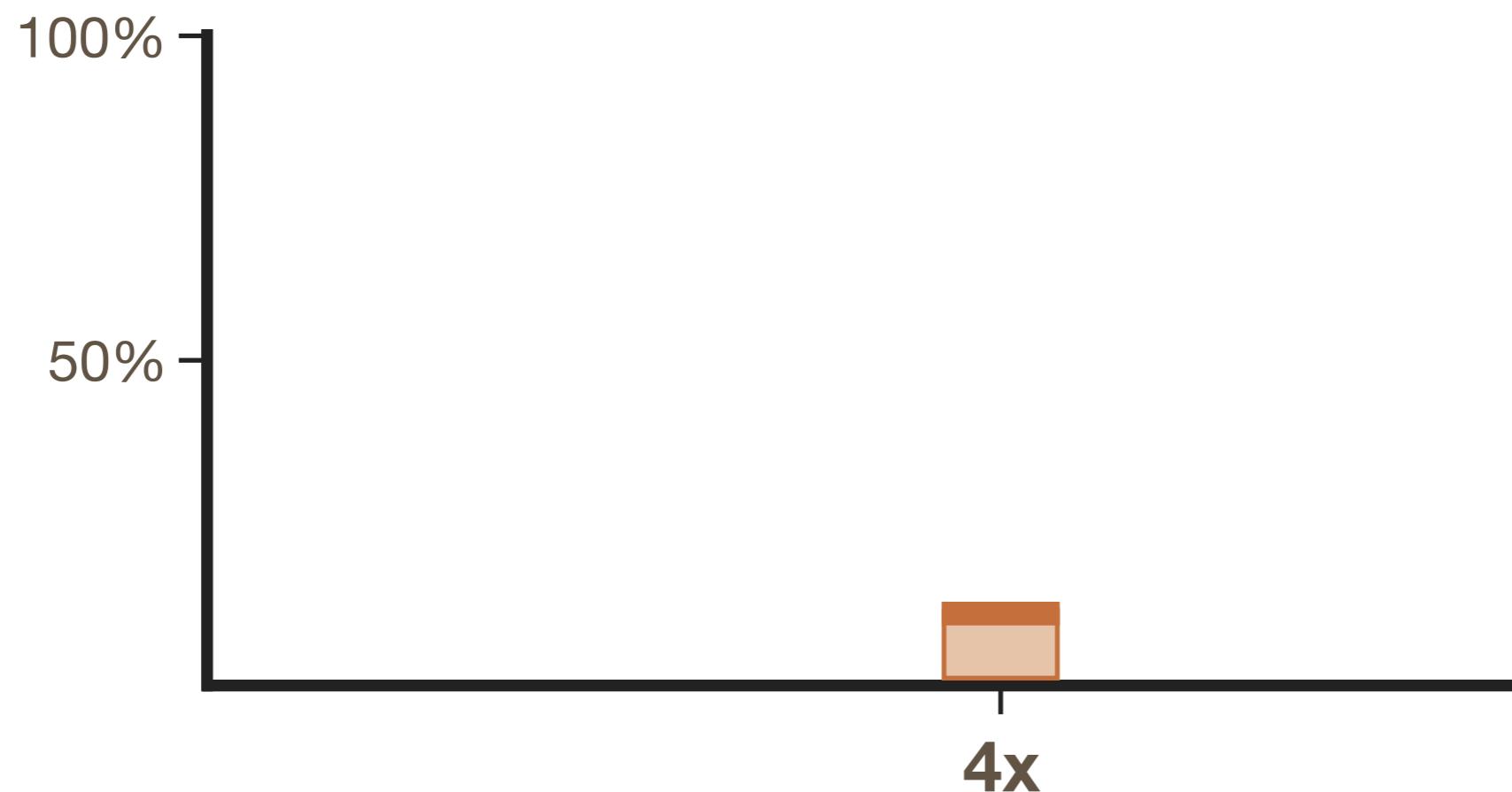
Exhaustive Results

What % of configurations have at most **Dx** overhead?



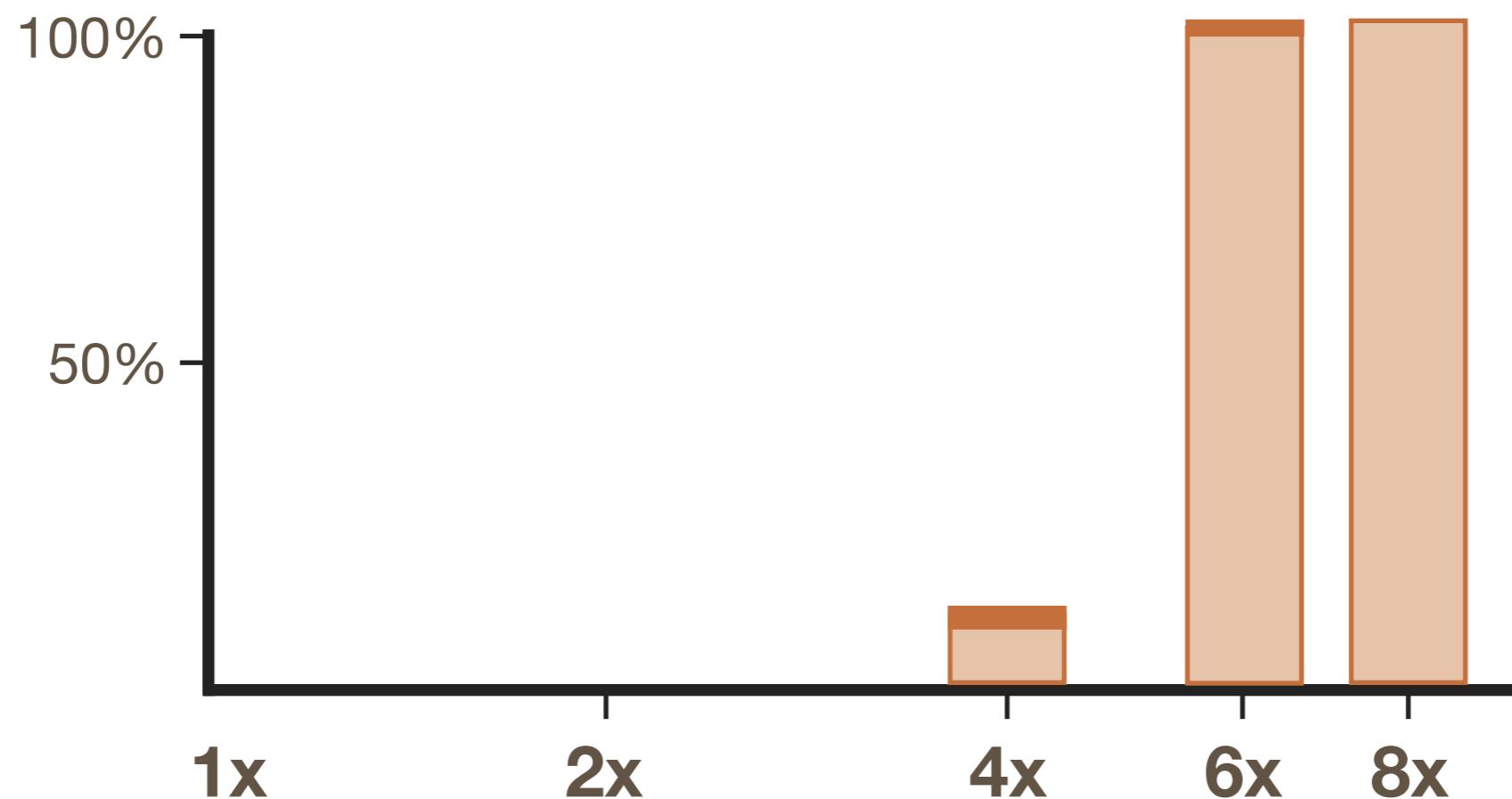
Approximate Results

What % of configurations have at most **4x** overhead,
based on **R** samples of **S** configurations each?



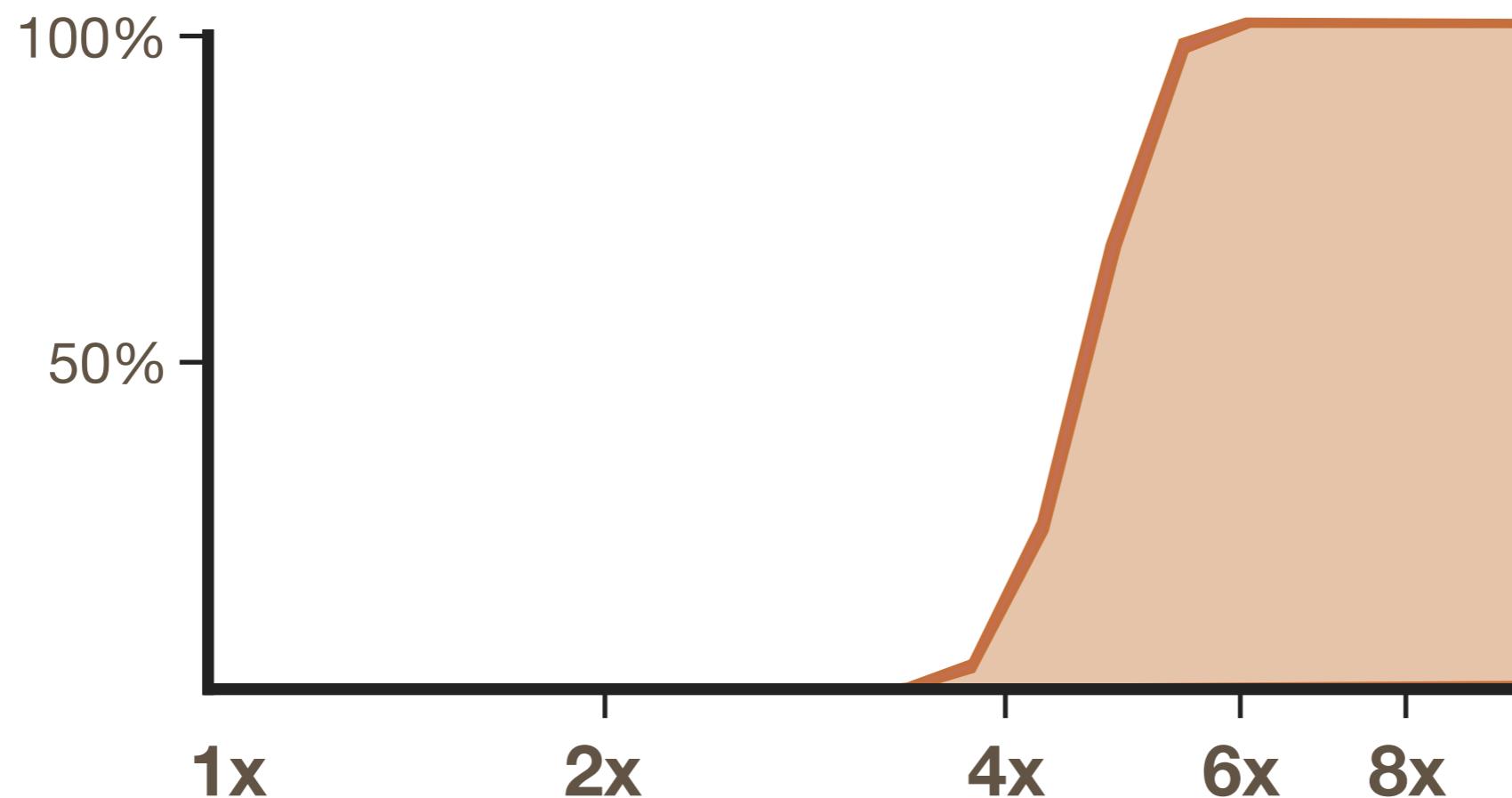
Approximate Results

What % of configurations have at most **Dx** overhead,
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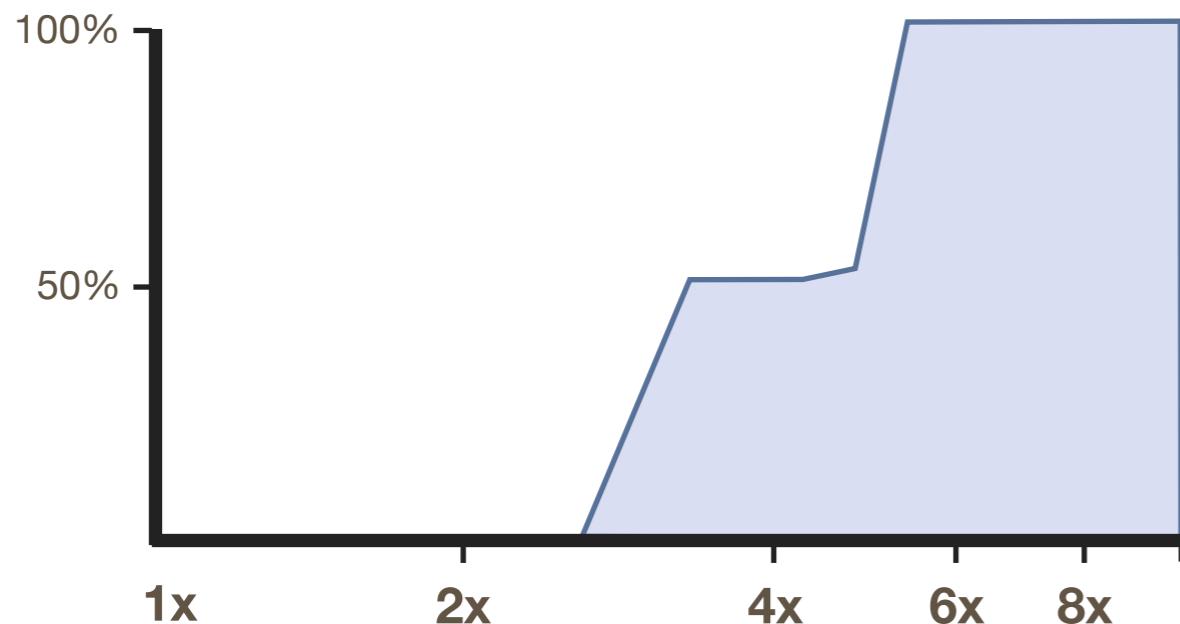
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What % of configurations have at most **Dx** overhead,
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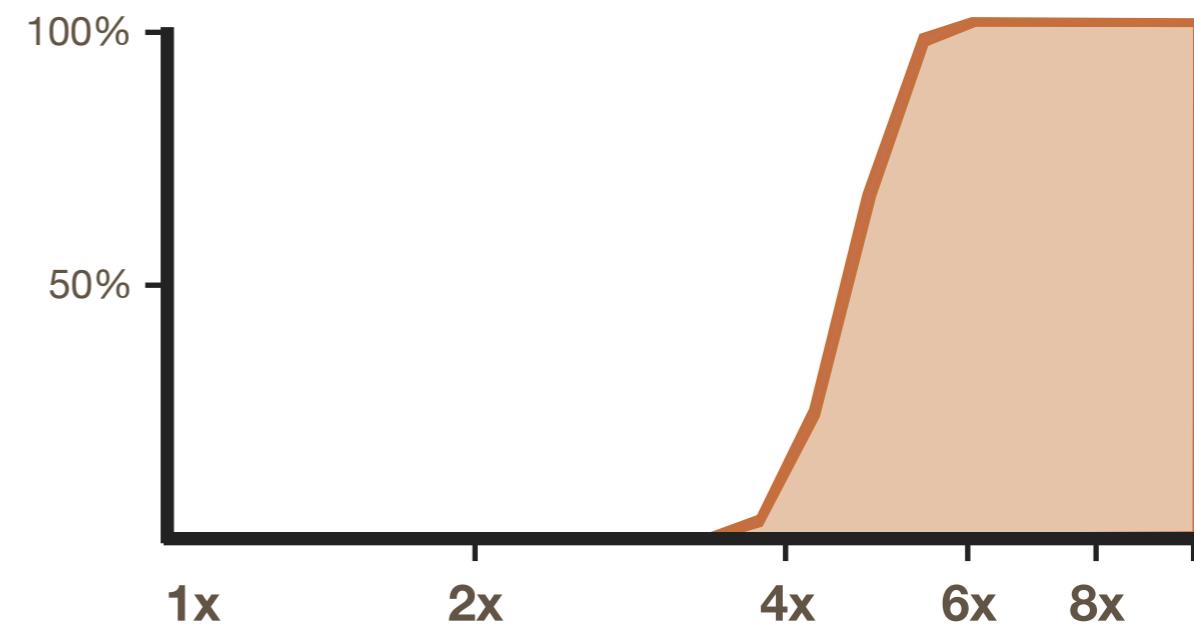
espionage

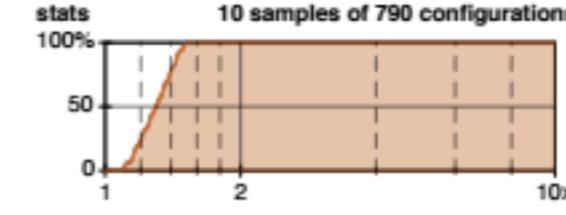
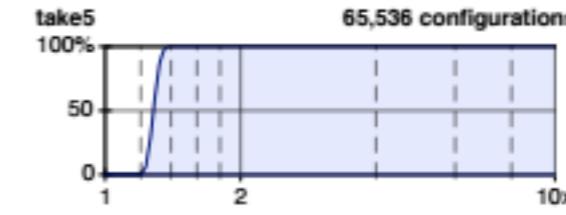
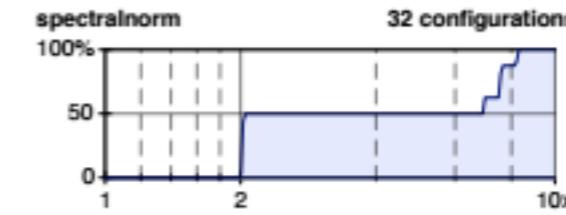
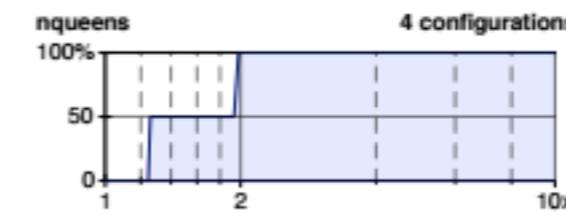
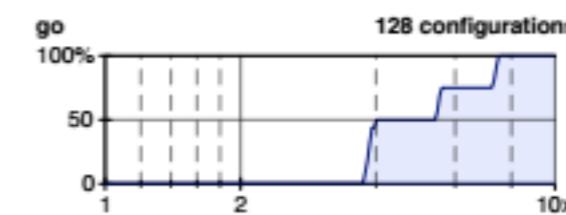
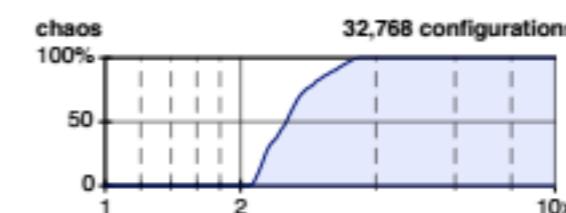
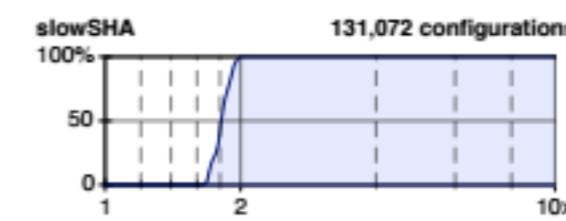
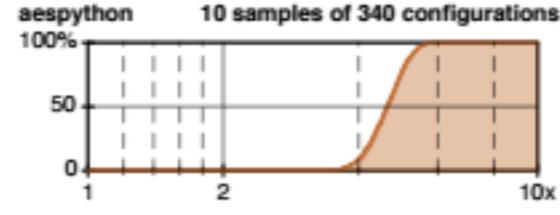
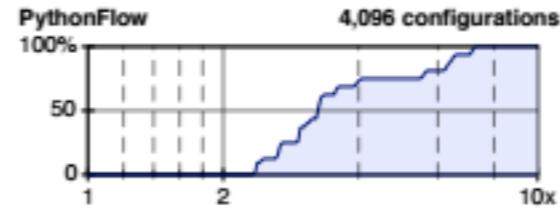
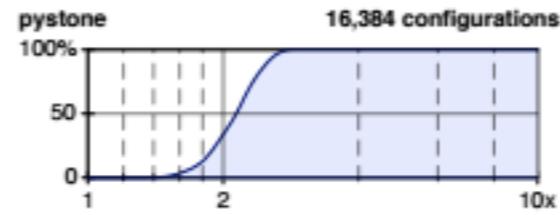
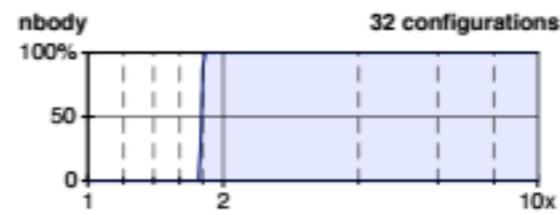
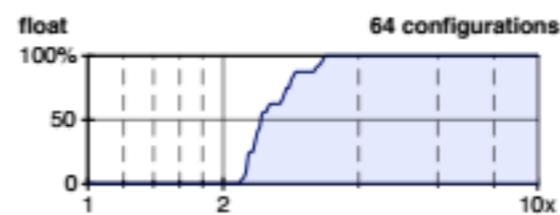
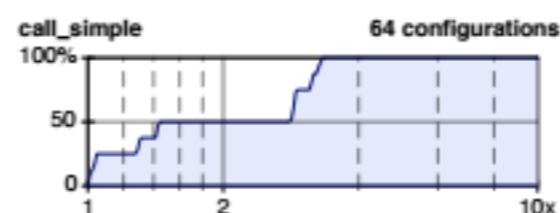
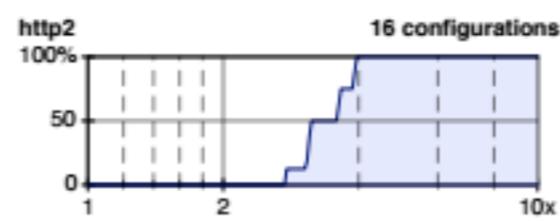
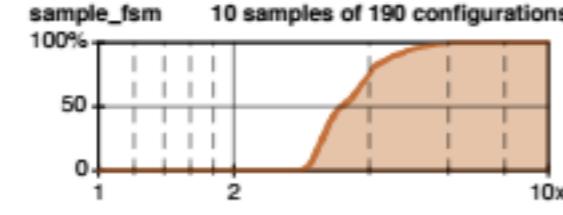
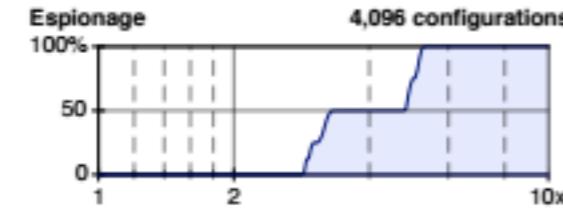
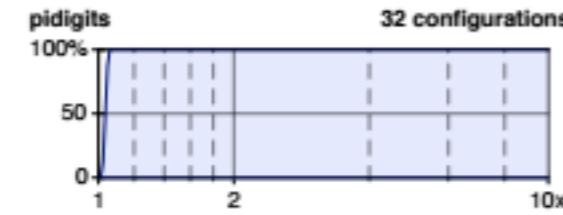
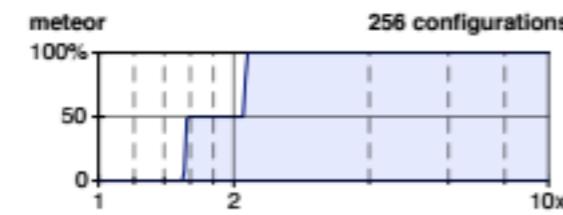
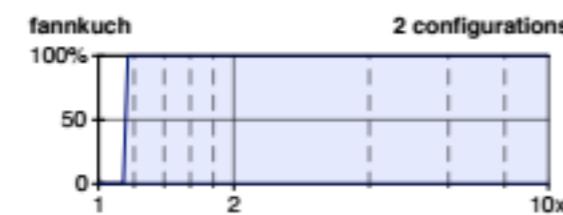
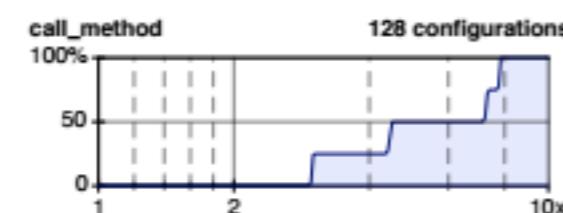
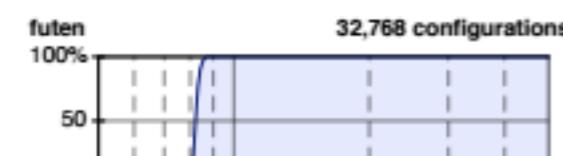
4,096 configurations



aespython

**10 samples
of 340 configurations**





Cost of Tag Soundness

- Worst-case overhead: under 10x

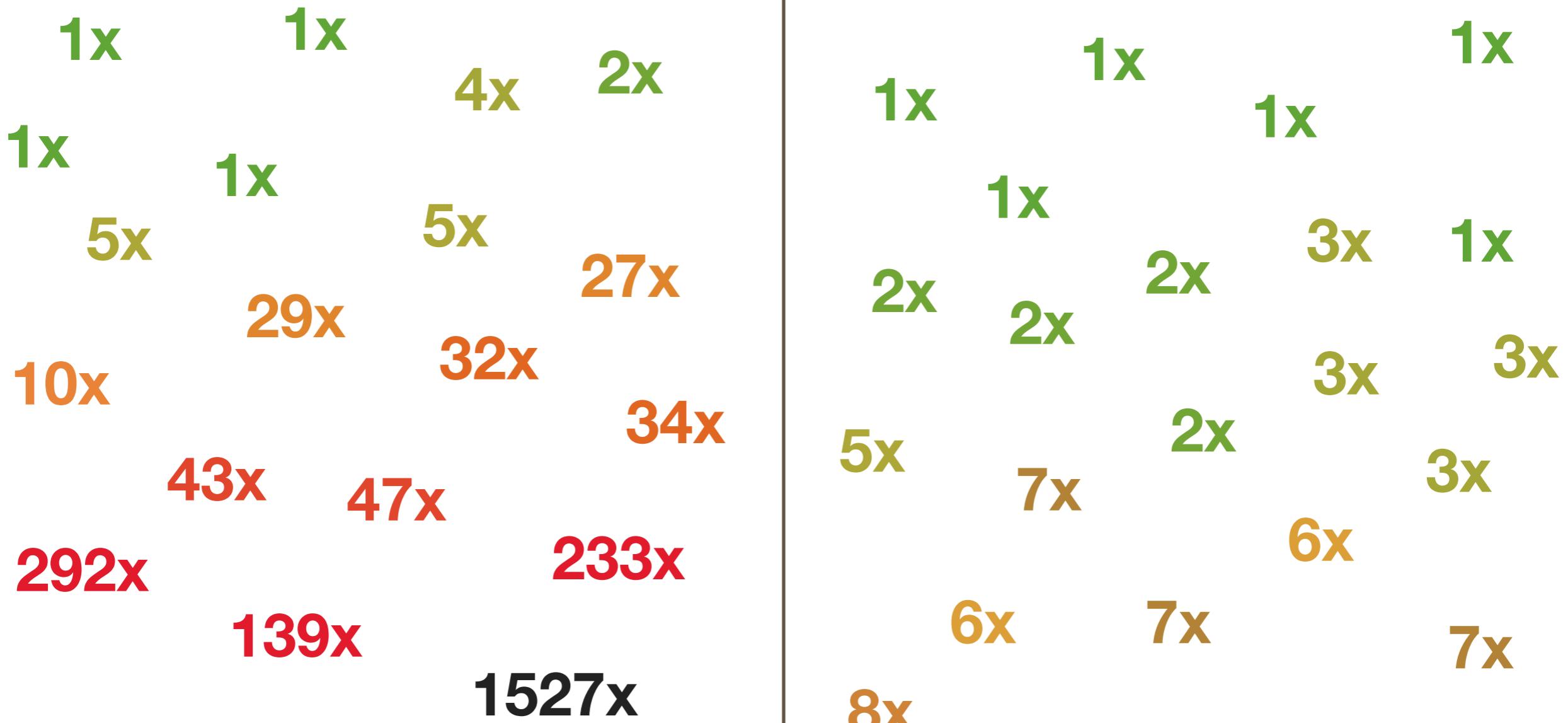


This is an APPLES to ORANGES comparison!



Type Soundness

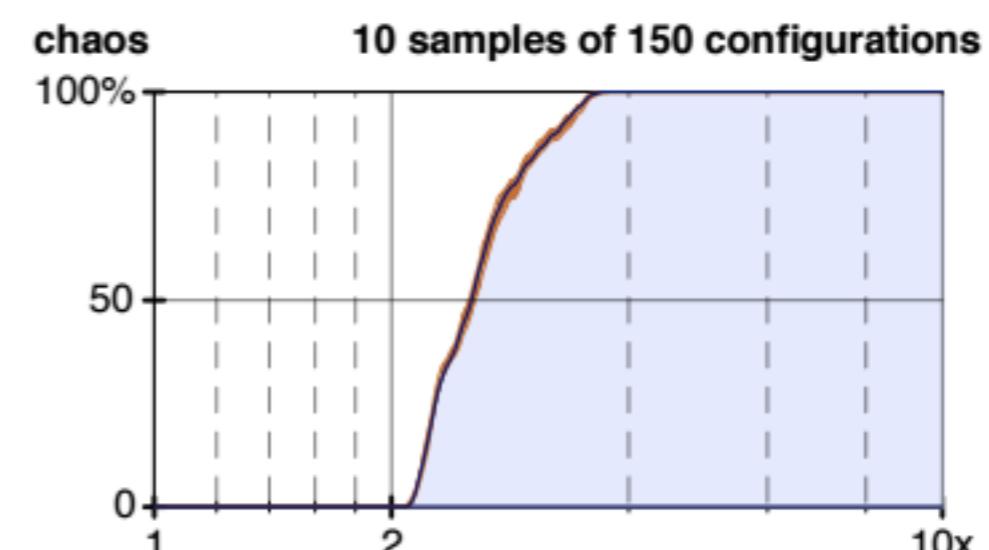
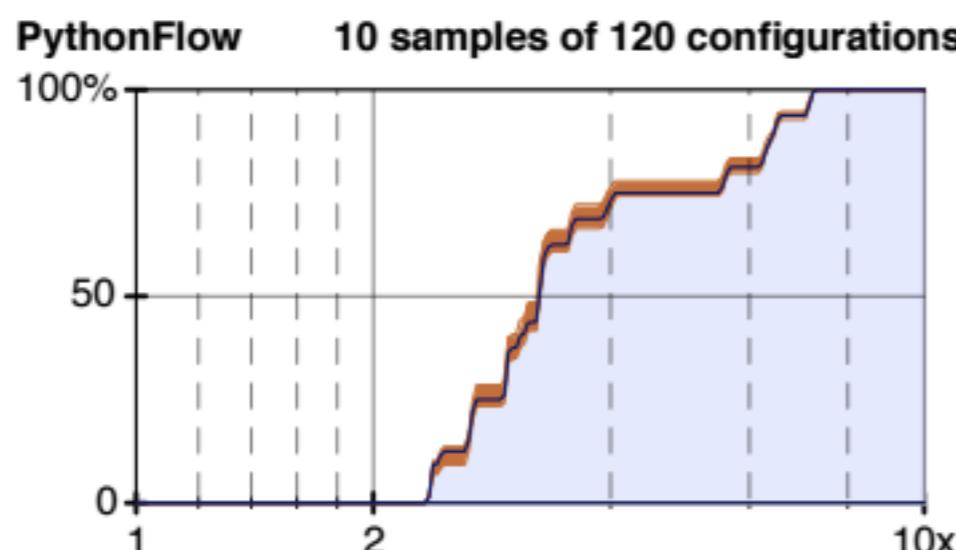
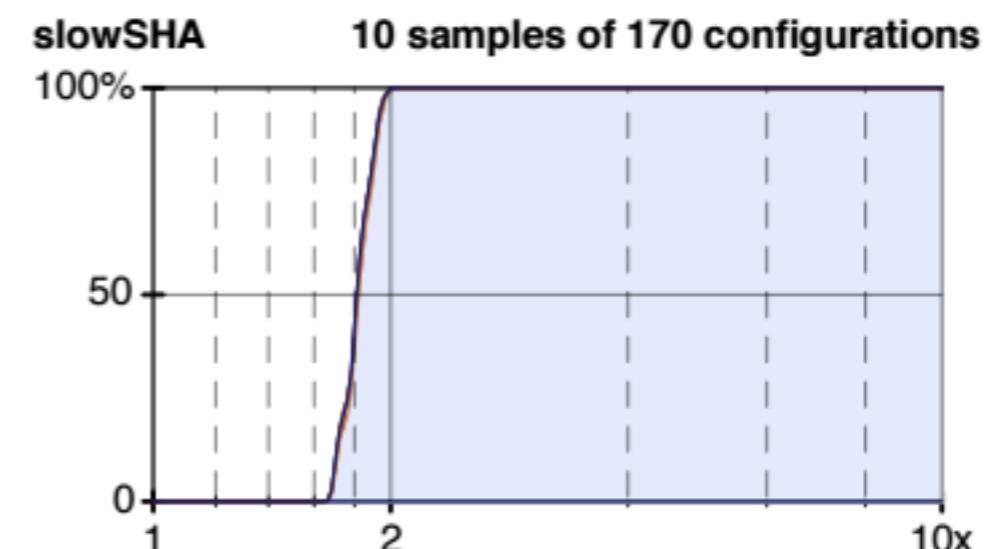
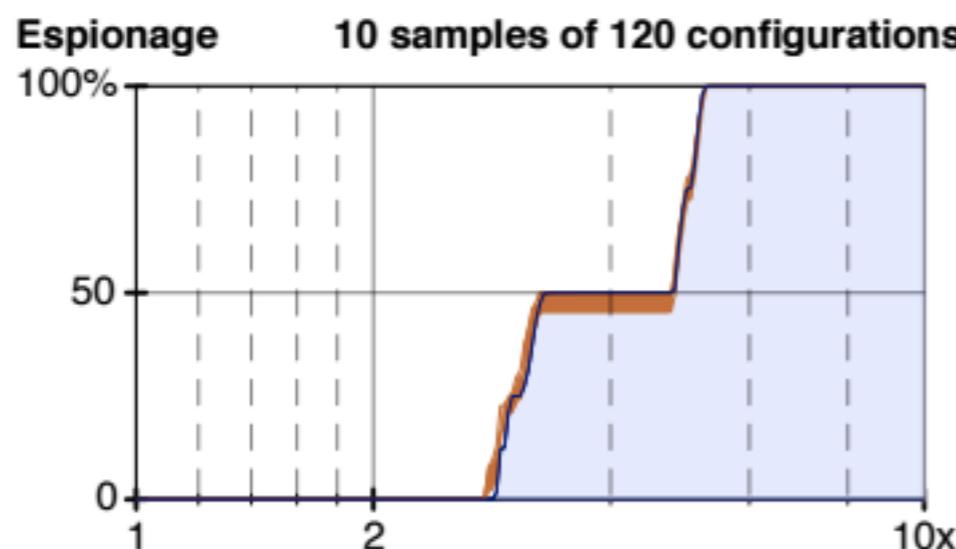
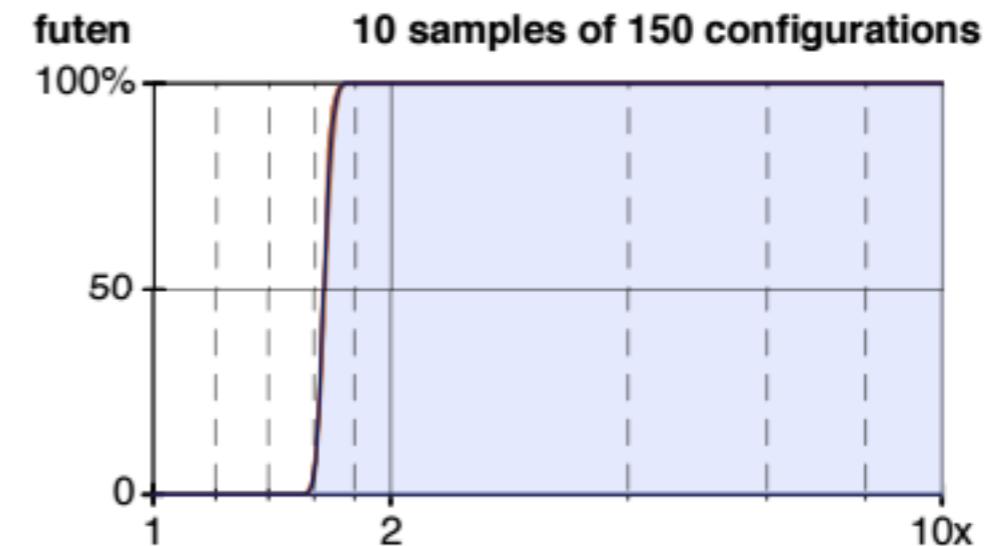
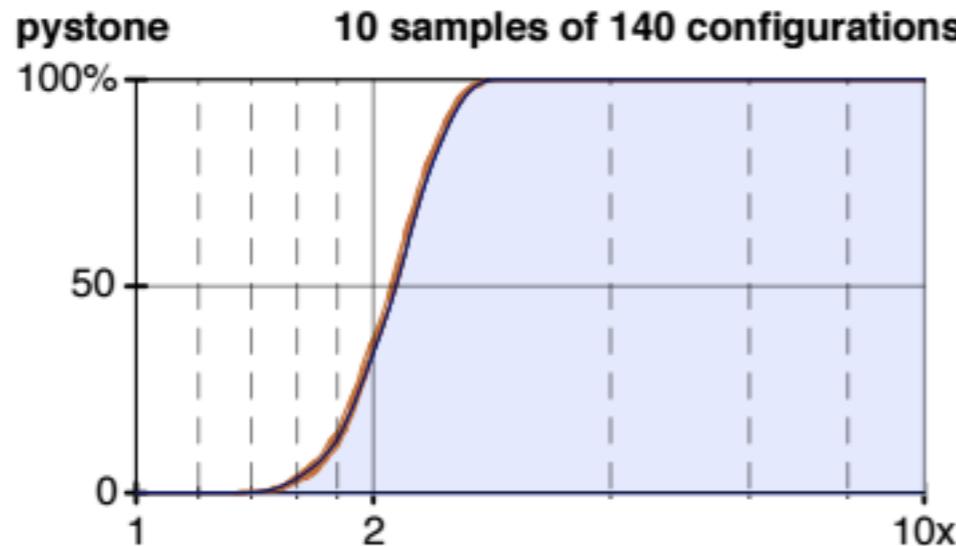
Tag Soundness



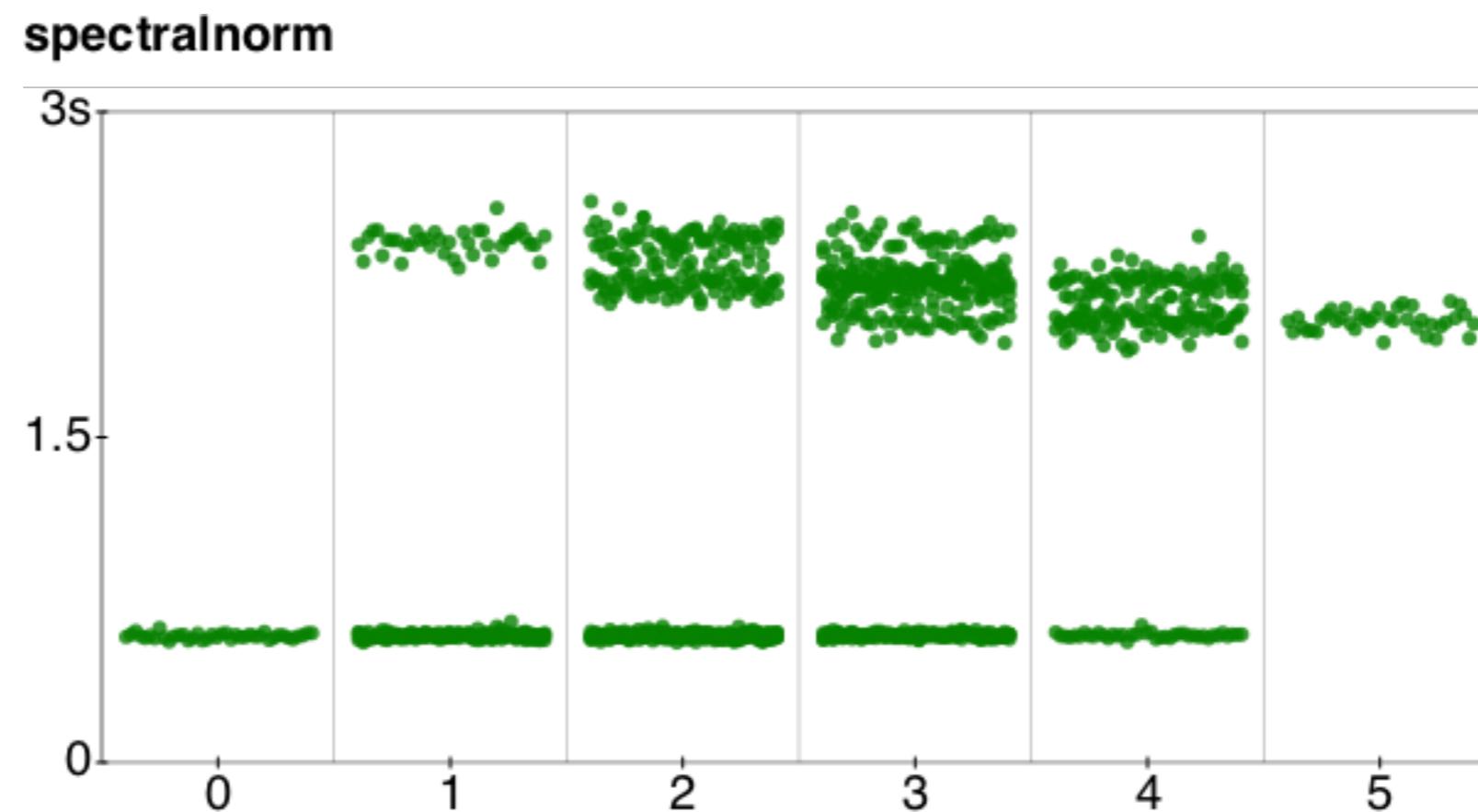
Cost of Tag Soundness

- Worst-case overhead: under 10x
- Best-case overhead: 1x -- 4x
 - adding types never* improves performance
- Slowest configuration: fully-typed
 - Overhead \propto number of type annotations

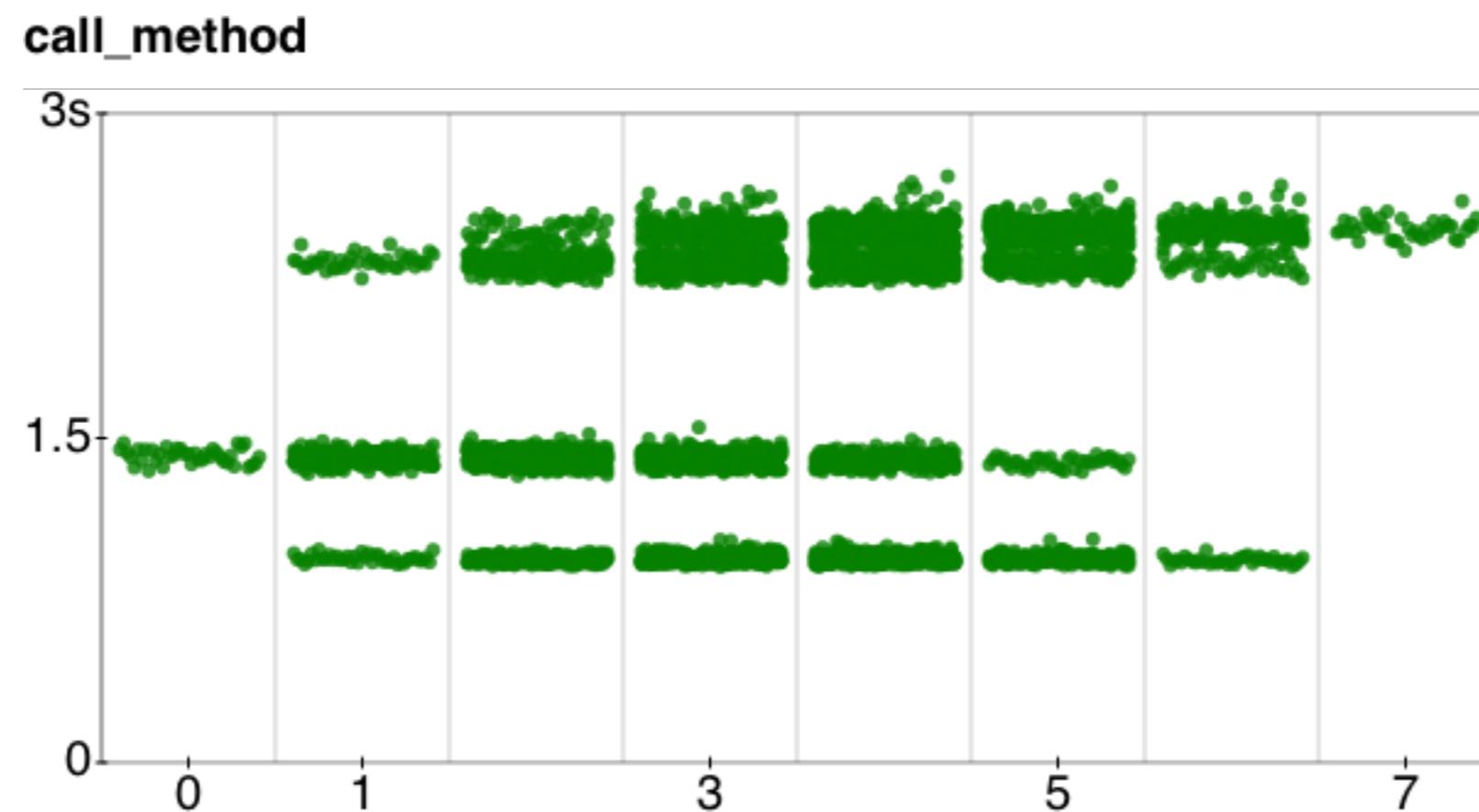




Runtime vs. # Types

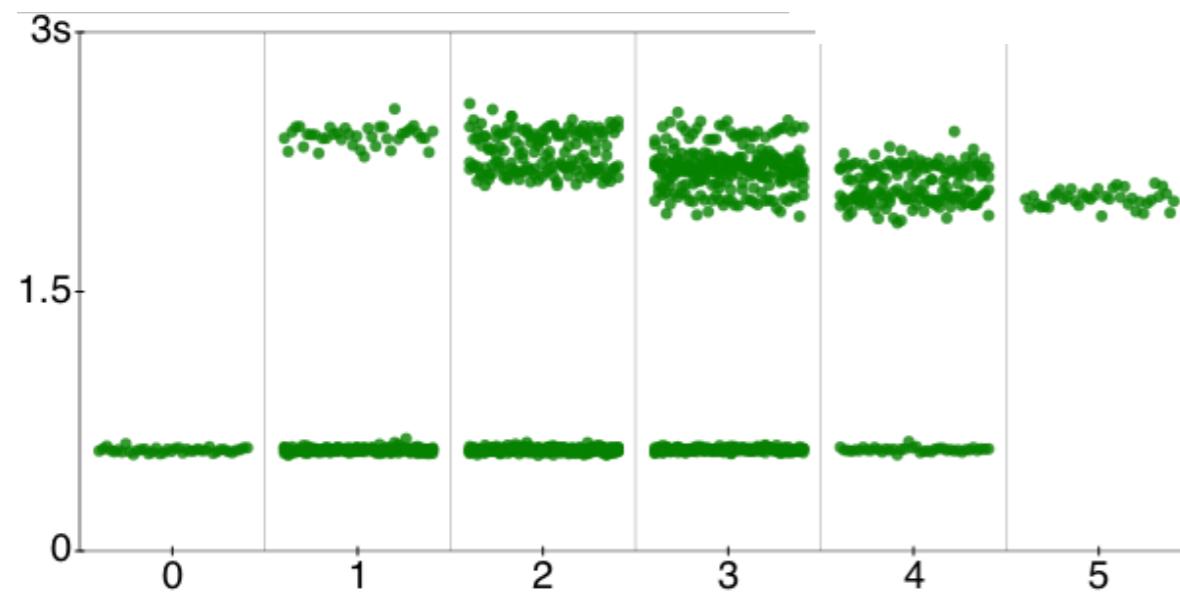


Runtime vs. # Types

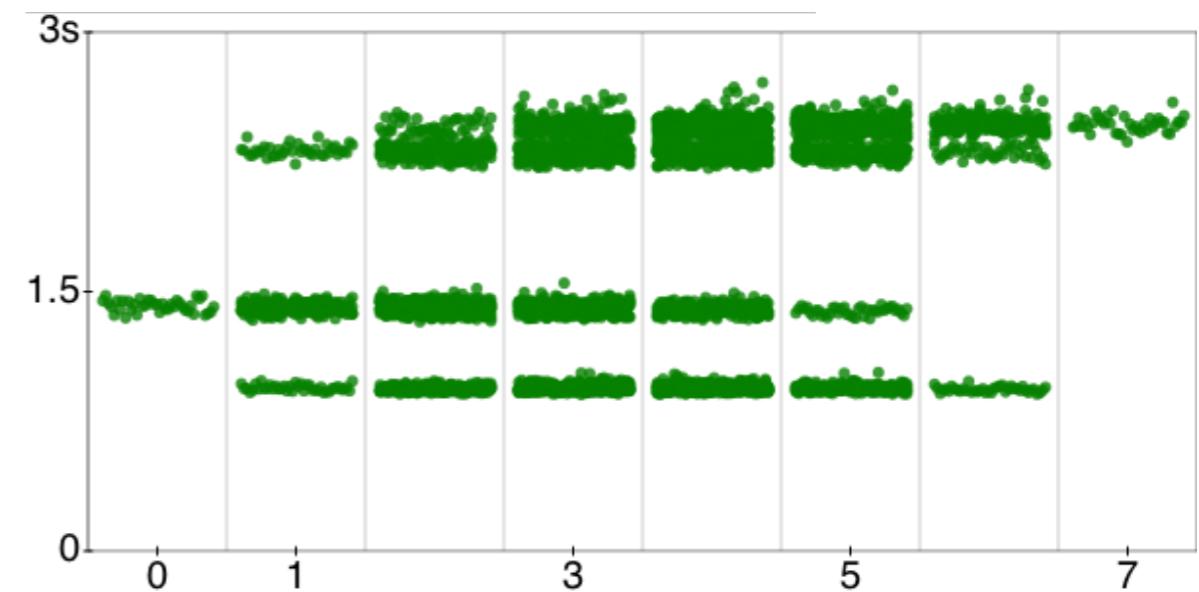


Speedup?

spectralnorm

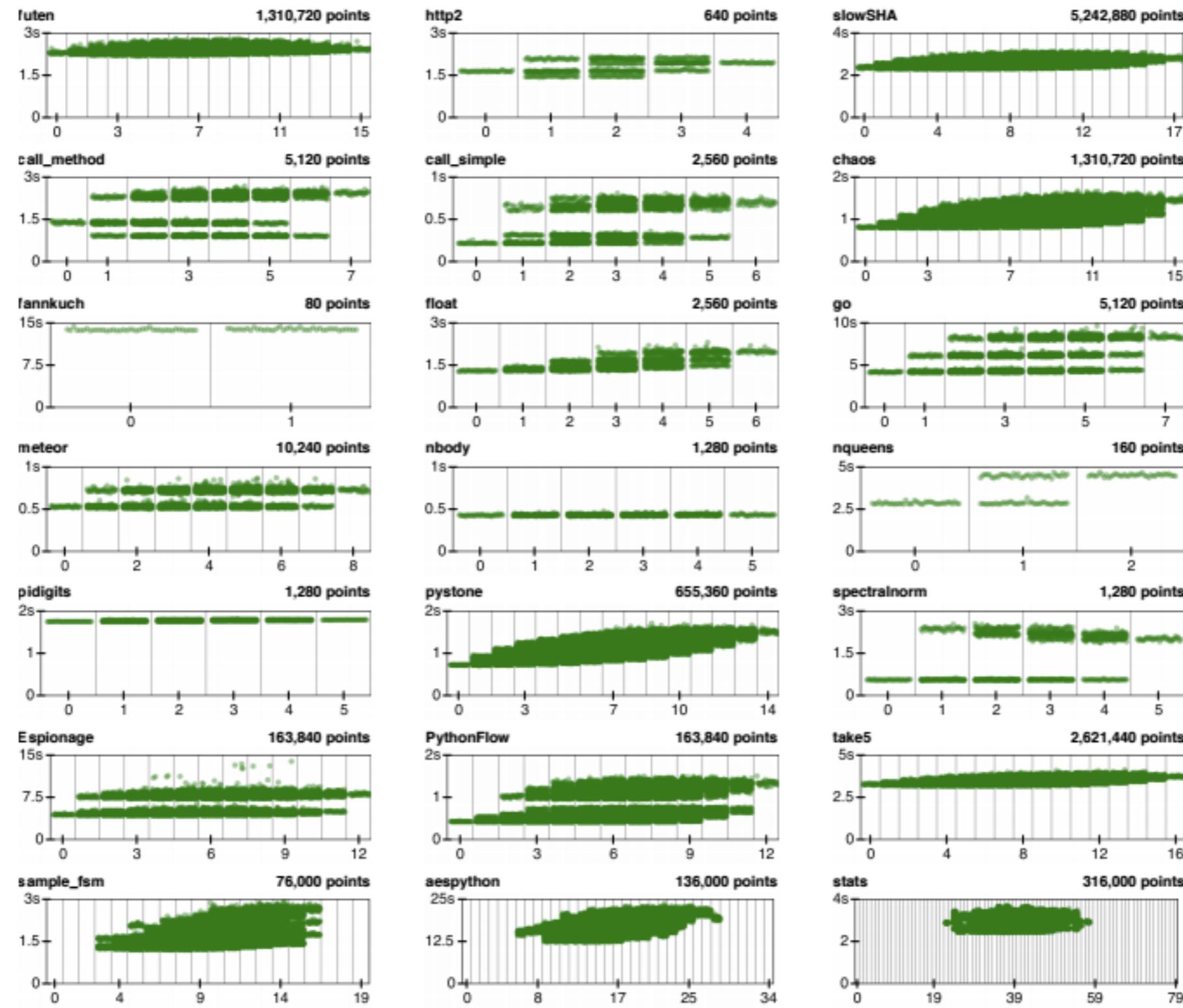


call_method



- Unsound optimization for read-only values (tuples)
- Double-checks method calls

Runtime vs. # Types



Experiment

- granularity: functions & class-fields
- 10 samples of $[10 * (F + C)]$ configurations
- Karst at Indiana University cluster (32GB RAM, 250GB other)
- Reticulated, master branch, commit e478343
- Python 3.4.3
- 40 iterations per configuration, report average
- 200 values of D on x-axis

POPL 2017

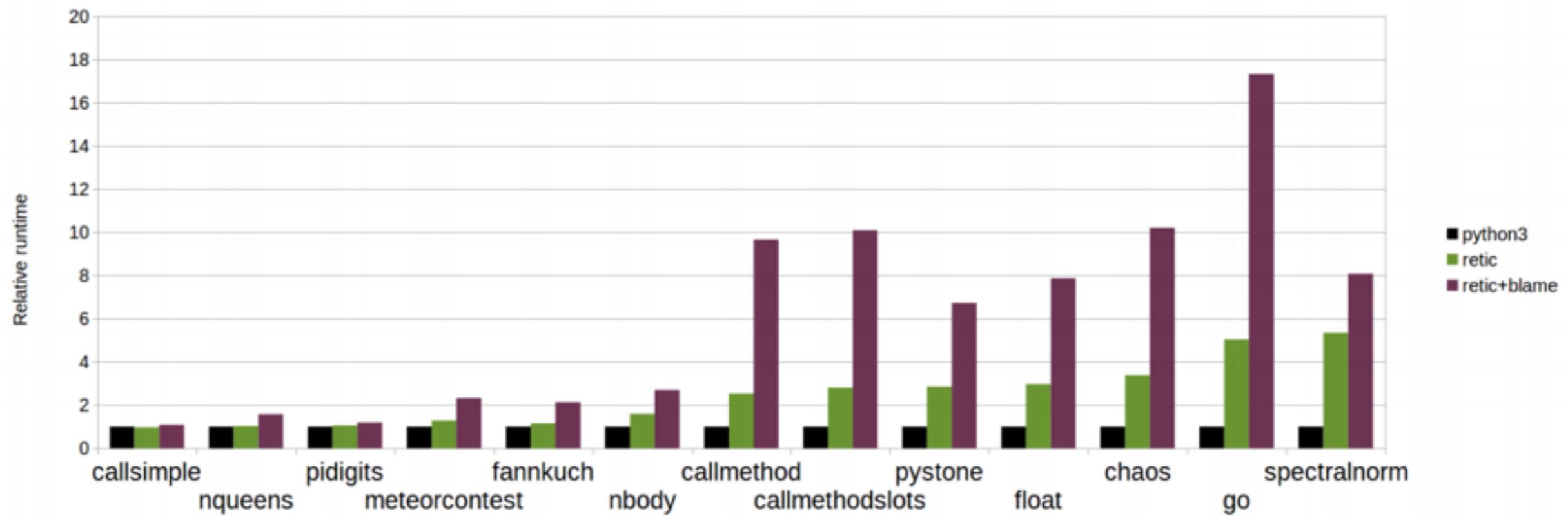


Figure 11. Runtime comparison of Reticulated Python to standard Python 3.4. Experiments were performed on an Ubuntu 14.04 laptop with a 2.8GHz Intel i7-3840QM CPU and 16GB memory.

Module Marshal

```
module Marshal: sig .. end
```

Marshaling of data structures.

This module provides functions to encode arbitrary data structures as sequences of bytes, which can then be written on a file or sent over a pipe or network connection. The bytes can then be read back later, possibly in another process, and decoded back into a data structure. The format for the byte sequences is compatible across all machines for a given version of OCaml.

Warning: marshaling is currently not type-safe. The type of marshaled data is not transmitted along the value of the data, making it impossible to check that the data read back possesses the type expected by the context. In particular, the result type of the `Marshal.from_*` functions is given as `'a`, but this is misleading: the returned OCaml value does not possess type `'a` for all `'a`; it has one, unique type which cannot be determined at compile-time. The programmer should explicitly give the expected type of the returned value, using the following syntax:

- (`Marshal.from_channel chan : type`). Anything can happen at run-time if the object in the file does not belong to the given type.

References

- Vitousek, Swords, Siek. *Big Types in Little Runtime: Open-World Soundness and Collaborative Blame for Gradual Type Systems.* POPL 2017
- Takikawa, Feltey, Greenman, New, Vitek, Felleisen. *Is Sound Gradual Typing Dead?* POPL 2016.

UNUSED SLIDES

???

⊤ e' : τ'

⋮ ⋮

⊤ e : τ

Takikawa Method

- granularity
- experimental modules, fixed modules
- configurations
- baseline
- performance ratio