

Static Type and Value Analysis by Abstract Interpretation of Python Programs with Native C Libraries

Raphaël Monat, Abdelraouf Ouadjaout, Antoine Miné

MTV Seminar
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Introduction

Whoami

- ▶ Member of APR, LIP6, Sorbonne Université & CNRS
- ▶ September 2018 – August 2021: PhD Student
- ▶ September 2021 – August 2022: ATER

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- ▶ Value analyses (relational)

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Bonus: Around the tax code

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Static Program Analysis

```
average.py
```

```
1 def average(l):
2     m = 0
3     for i in range(len(l)):
4         m = m + l[i]
5     m = m // (i + 1)
6     return s
7
8 r1 = average([1, 2, 3])
9 r2 = average(['a', 'b', 'c'])
```

TypeError: unsupported operand type(s) for '+': 'int' and 'str'

```
argslen.c
```

```
1 #include <string.h>
2
3 int main(int argc, char *argv[]) {
4     int i = 0;
5     for (char **p = argv; *p; p++) {
6         strlen(*p); // valid string
7         i++; // no overflow
8     }
9     return 0;
10 }
```

No alarm

Specifications of the analyzer

Inference of program properties such as the absence of run-time errors.

Automatic no expert knowledge required.

Semantic based on a formal modelization of the language.

Sound cover all possible executions.

Growing popularity

JavaScript #1, Python #2 on GitHub¹

¹<https://octoverse.github.com/#top-languages>

Dynamic programming languages

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New features

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- ▶ Introspection operators,
- ▶ eval.

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Outline

1 Introduction

2 A Taste of Python

3 Analyzing Python Programs

4 Analyzing Python Programs with C Libraries

5 Conclusion

A Taste of Python

No standard

- ▶ CPython is the reference
 - ⇒ manual inspection of the source code and handcrafted tests

Python's specificities

No standard

- ▶ CPython is the reference
⇒ manual inspection of the source code and handcrafted tests

Operator redefinition

- ▶ Calls, additions, attribute accesses
- ▶ Operators eventually call overloaded `__methods__`

Protected attributes

```
1 class Protected:  
2     def __init__(self, priv):  
3         self._priv = priv  
4     def __getattribute__(self, attr):  
5         if attr[0] == "_": raise AttributeError("protected")  
6         return object.__getattribute__(self, attr)  
7     a = Protected(42)  
8 a._priv # AttributeError raised
```

Python's specificities (II)

Dual type system

- ▶ Nominal (classes, MRO)

Fspath (from standard library)

```
1 class Path:
2     def __fspath__(self): return 42
3
4     def fspath(p):
5         if isinstance(p, (str, bytes)):
6             return p
7         elif hasattr(p, "__fspath__"):
8             r = p.__fspath__()
9             if isinstance(r, (str, bytes)):
10                 return r
11             raise TypeError
12
13 fspath("/dev" if random() else Path())
```

Barrett et al. "A Monotonic Superclass Linearization for Dylan". OOPSLA 1996

Python's specificities (II)

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- ▶ Nominal (classes, MRO)
- ▶ Structural (attributes)

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Exceptions

Exceptions rather than specific values

- ▶ `1 + "a" ~> TypeError`
- ▶ `l[len(l) + 1] ~> IndexError`

Fspath (from standard library)

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Previous works on Python 3

Guth. "A formal semantics of Python 3.3". 2013

Implementation within the K framework.

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Complex desugaring into λ_{π} .

May incur losses of precision in the abstract interpreter.

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Semantics of Python, using a Python framework, developed concurrently.

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Different goal

These works focus on the concrete semantics. This is not our endgoal.

Previous works on Python 3

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Politz et al. "Py Moving to our own semantics"

Complex desugaring of Python code.

May incur losses of precision.

- ▶ Cost of understanding the code (vs CPython)
- ▶ Trust in the code (CPython's tests?)
- ▶ Insights of the papers

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Our approach

Interpreter-like semantics

Easily convertible to an abstract interpreter.

²Fromherz, Ouadjaout, and Miné. “Static Value Analysis of Python Programs by Abstract Interpretation”. NFM 2018.

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Interpreter-like semantics

Easily convertible to an abstract interpreter.

Major extension of the work of Fromherz, Ouadjaout, and Miné²

- ▶ Separation between core and builtins
- ▶ 2.3× more cases (**with** statement, bidirectional generators, ...)
- ▶ Improved some cases (+, boolean casts of conditionals, data descriptors, ...)

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Correctness

- ▶ Strived to make it auditable (with links to the source).
- ▶ Tested only through the abstract analysis yet (no concrete execution).

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Example – attribute access

```
Ecur[(x.s)](cur, e, h)  $\stackrel{\text{def}}{=}$  LOAD_ATTR PyObject_GetAttr (slot_tp_getattr_hook)  
letb (cur, e, h), @x = E[x](cur, e, h) in  
letb (cur, e, h), @c = E[mro_search(type(@x), "__getattribute__")](cur, e, h) in  
letcases (f, e, h), @x.s = E[@c(@x, s)](cur, e, h) in  
match f with  
• exn @exc when isinstance(@exc, AttributeError)  $\Rightarrow$   
  let (f, e, h), @d = E[mro_search(type(@x), "__getattr__")](f, e, h) in  
    if d  $\neq \perp$  then return E[@d(@x, s)](cur, e, h)  
    else return (f, e, h),  $\perp$   
• _  $\Rightarrow$  return (f, e, h), @x.s
```

Example – attribute access

```
Ecur[(x.s)](cur, e, h)  $\stackrel{\text{def}}{=}$  LOAD_ATTR PyObject_GetAttr (slot_tp_getattr_hook)
letb (cur, e, h), @x = E[x](cur, e, h) in
  ...
Ecur[(object.__getattribute__)(obj, name)](cur, e, h)  $\stackrel{\text{def}}{=}$ 
  tp_field _PyObject_GenericGetAttrWithDict
letb (cur, e, h), @o = E[obj](cur, e, h) in
letb (cur, e, h), @n = E[name](cur, e, h) in
  if  $\neg$ isinstance(@n, str) then return S[raise TypeError](cur, e, h), ⊥ else
    let str(n) = fst oh(@n) in
      letcases (f, e, h), @descr = E[mro_search(type(@o), n)](f, e, h) in
        if @descr  $\neq$  ⊥ then
          if hasattr(type(@descr), "__get__")  $\wedge$ 
            (hasattr(type(@descr), "__set__")  $\vee$  hasattr(type(@descr), "__delete__")) then
              return E[type(@descr).__get__(@descr, @o, type(@o))](f, e, h)
```

Example – attribute access

$E_{cur}[\![x.s]\!](cur, e, h) \stackrel{\text{def}}{=} \text{LOAD_ATTR PyObject_GetAttr}(\text{slot_tp_getattr_hook})$

letb $(cur, e, h), @_x = E[\![x]\!](cur, e, h)$ in

$E_{cur}[\![\text{object}.__getattribute__}\!(\text{obj}, \text{name})]\!](cur, e, h) \stackrel{\text{def}}{=} \text{tp_field PyObject_GenericGetAttrWithDict}$

letb $(cur, e, h), @_o = E[\![\text{obj}]\!](cur, e, h)$ in

letb $(cur, e, h), @_n = E[\![\text{name}]\!]$ in

$E_{cur}[\![\text{type}.__getattribute__}\!(\text{typ}, \text{name})]\!](cur, e, h) \stackrel{\text{def}}{=} \text{tp_field type_getattro}$

letb $(cur, e, h), @_typ = E[\![\text{typ}]\!](cur, e, h)$ in

letb $(cur, e, h), @_name = E[\![\text{name}]\!](cur, e, h)$ in

$E_{cur}[\![\text{mro_search}(\text{type}(@_typ), @_name)]]\!](cur, e, h)$ in

letb $(cur, e, h), @_meta = E[\![\text{mro_search}(\text{type}(@_typ), @_name)]]\!](cur, e, h)$ in

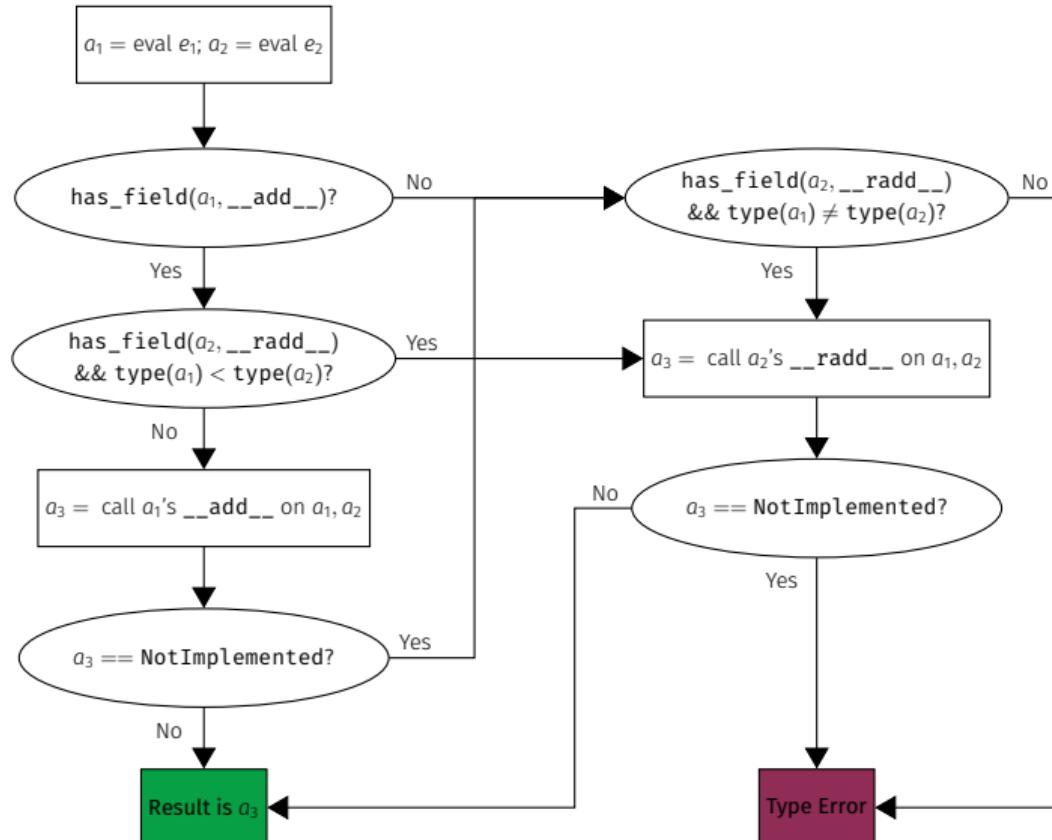
if $@_meta \neq \perp$ then

if $\text{hasattr}(\text{type}(@_meta), \text{"__get__"}) \wedge$

$\text{hasattr}(\text{type}(@_meta), \text{"__set__"}) \vee \text{hasattr}(\text{type}(@_meta), \text{"__delete__"})$ then

$E_{cur}[\![\text{type}(@_typ), \text{type}(@_typ)]]\!](cur, e, h)$

Example – binary operators



Crazy Python

Custom infix operators

```
1 class Infix(object):
2     def __init__(self, func): self.func = func
3     def __or__(self, other): return self.func(other)
4     def __ror__(self, other): return Infix(lambda x: self.func(other, x))
5
6 instanceof = Infix(isinstance)
7 b = 5 |instanceof| int
8
9 @Infix
10 def padd(x, y):
11     print("{x} + {y} = {x + y}")
12     return x + y
13 c = 2 |padd| 3
```

Credits tomerfiliba.com/blog/Infix-Operators/

Analyzing Python Programs

Analysis | Overview

Goal

Detect runtime errors: uncaught raised exceptions

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Supported constructs

Our analysis supports:

- ▶ Objects
- ▶ Exceptions
- ▶ Dynamic typing
- ▶ Introspection
- ▶ Permissive semantics
- ▶ Dynamic attributes
- ▶ Generators
- ▶ `super`
- ▶ Metaclasses

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Unsupported constructs

- ▶ Recursive functions
- ▶ `eval`
- ▶ Finalizers

Analysis | Domains required

Avering numbers

```
1 def average(l):
2     m = 0
3     for i in range(len(l)):
4         m = m + l[i]
5     m = m // (i + 1)
6     return m
7
8 l = [randint(0, 20)
9      for i in range(randint(5, 10))]
10 m = average(l)
```

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Searching for a loop invariant (l. 4)

Environment abstraction

$$m \mapsto @_{\text{int}^{\#}} \quad i \mapsto @_{\text{int}^{\#}}$$

Proved safe?

- ▶ $m // (i+1)$
- ▶ $l[i]$

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Searching for a loop invariant (l. 4)

Stateless domains: **list content**,

Environment abstraction

$$m \mapsto @_{\text{int}^{\#}} \quad i \mapsto @_{\text{int}^{\#}} \quad \underline{\text{els}}(l) \mapsto @_{\text{int}^{\#}}$$

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Numeric abstraction (intervals)

$$m \in [0, +\infty) \quad \underline{\text{els}}(l) \in [0, 20] \quad i \in [0, +\infty)$$

Analysis | Domains required

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- ▶ $l[i]$

Searching for a loop invariant (l. 4)

Stateless domains: list content, **list length**

Environment abstraction

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$$m \in [0, +\infty) \quad \underline{\text{els}}(l) \in [0, 20]$$

$$\underline{\text{len}}(l) \in [5, 10] \quad i \in [0, 10]$$

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$$m \mapsto @_{\text{int}^{\#}}^{\#} \quad i \mapsto @_{\text{int}^{\#}}^{\#} \quad \underline{\text{els}}(l) \mapsto @_{\text{int}^{\#}}^{\#}$$

Numeric abstraction (polyhedra)

$$m \in [0, +\infty) \quad \underline{\text{els}}(l) \in [0, 20]$$

$$0 \leq i < \underline{\text{len}}(l) \quad 5 \leq \underline{\text{len}}(l) \leq 10$$

Analysis | Domains required

Averaging tasks

```
1 class Task:
2     def __init__(self, weight):
3         if weight < 0: raise ValueError
4         self.weight = weight
5
6     def average(l):
7         m = 0
8         for i in range(len(l)):
9             m = m + l[i].weight
10            m = m // (i + 1)
11        return m
12
13    l = [Task(randint(0, 20))
14        for i in range(randint(5, 10))]
15    m = average(l)
```

Proved safe?

- ▶ $m // (i+1)$
- ▶ $l[i].weight$

Searching for a loop invariant (l. 4)

Stateless domains: list content, list length

Environment abstraction

$$\begin{aligned}m &\mapsto @_{\text{int}\#}^{\sharp} \quad i \mapsto @_{\text{int}\#}^{\sharp} \quad \underline{\text{els}(l)} \mapsto @_{\text{Task}}^{\sharp} \\@_{\text{Task}}^{\sharp} \cdot \underline{\text{weight}} &\mapsto @_{\text{int}\#}^{\sharp}\end{aligned}$$

Numeric abstraction (polyhedra)

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Attributes abstraction

$$@_{\text{Task}}^{\sharp} \mapsto (\{\text{weight}\}, \emptyset)$$

Analysis | Domains required

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Conclusion

- ▶ Different domains depending on the precision
- ▶ Use of auxiliary variables (underlined)

Proved safe?

- ▶ $m // (i+1)$
- ▶ $l[i].weight$

Searching for a loop invariant (l. 4)

Stateless domains: list content, list length

Environment abstraction

$m \mapsto @_{Task}^{\#} \quad ; \quad @_{Task}^{\#}$

- ▶ $l \mapsto \{l[i] \mid 0 \leq i < \text{len}(l)\}$
- ▶ $0 \leq i < \underline{\text{len}}(l) \quad 5 \leq \underline{\text{len}}(l) \leq 10$
- ▶ $0 \leq \underline{@_{Task}^{\#} \cdot \text{weight}} \leq 20$

Attributes abstraction

$@_{Task}^{\#} \mapsto (\{\text{weight}\}, \emptyset)$



Modular Open Platform for Static Analysis³

³Journault, Miné, Monat, and Ouadjaout. "Combinations of reusable abstract domains for a multilingual static analyzer". VSTTE 2019.



Modular Open Platform for Static Analysis³

- ▶ One AST to analyze them all
 - 🚩 Multilanguage support
 - 📄 Expressiveness
 - ♻️ Reusability

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- ▶ Unified domain signature
 - 📝 Semantic rewriting
 - 🧩 Loose coupling
 - ⌚ Observability

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 - ⌚ Observability
- ▶ DAG of abstract domains
 - 📦 Composition
 - 💬 Cooperation

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Mopsa | Dynamic, semantic iterators with delegation

Universal.Iterators.Loops

Matches `while(...){...}`

Computes fixpoint using widening

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for(init; cond; incr) body
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C.iterators.loops

Rewrite and analyze recursively

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C.iterators.loops

Rewrite and analyze recursively

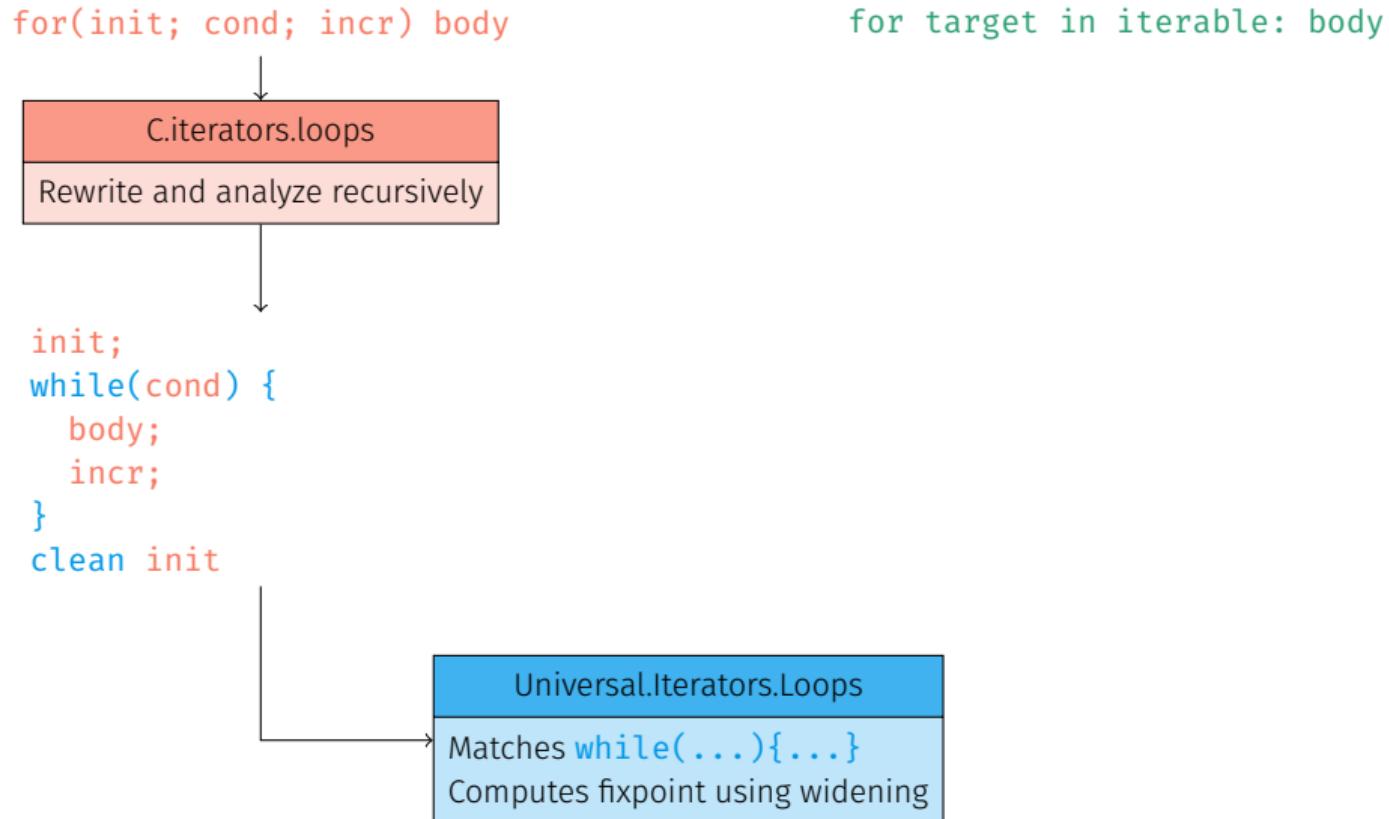
```
init;  
while(cond) {  
    body;  
    incr;  
}  
clean init
```

Universal.Iterators.Loops

Matches `while(...){...}`

Computes fixpoint using widening

Mopsa | Dynamic, semantic iterators with delegation



Mopsa | Dynamic, semantic iterators with delegation

```
for(init; cond; incr) body
```



C.iterators.loops

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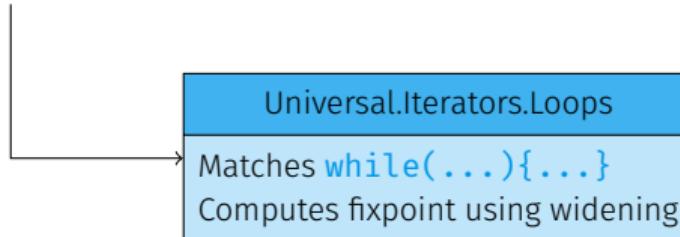
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init;  
while(cond) {  
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clean init
```

```
for target in iterable: body
```



Python.Desugar.Loops

- o Rewrite and analyze recursively
- o Optimize for some semantic cases



Mopsa | Dynamic, semantic iterators with delegation

```
for(init; cond; incr) body
```



C.iterators.loops

Rewrite and analyze recursively

```
init;  
while(cond) {  
    body;  
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clean init
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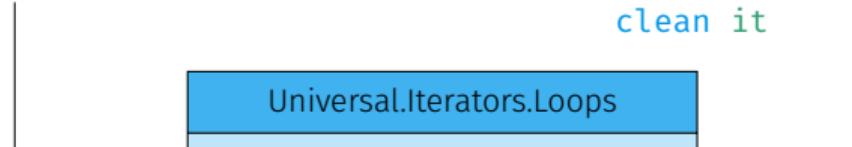
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for target in iterable: body
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Python.Desugar.Loops

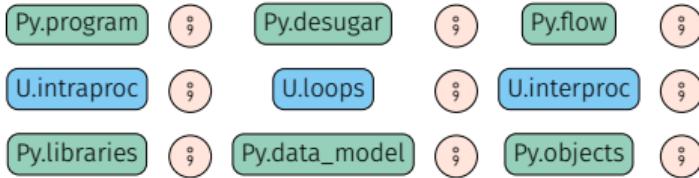
- o Rewrite and analyze recursively
- o Optimize for some semantic cases

```
it = iter(iterable)  
while(1) {  
    try: target = next(it)  
    except StopIteration: break  
    body  
}  
clean it
```



- ▶ Dynamicity:
type inference first
- ▶ Flow-sensitive
- ▶ Context-sensitive

Types | Analysis



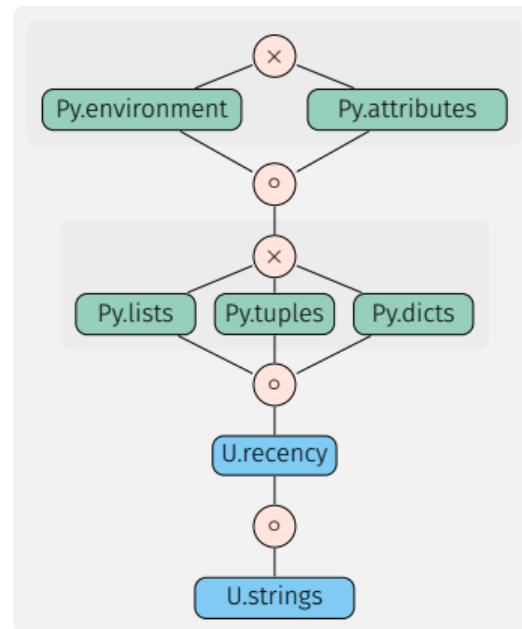
Sequence

Cartesian product

Composition

Universal

Python specific



- ▶ Dynamicity:
type inference first
- ▶ Flow-sensitive
- ▶ Context-sensitive

Types | Related work

- ▶ Similar in essence to TAJS.⁴
- ▶ Dataflow analysis by Fritz and Hage.⁵
- ▶ Typpete: SMT-based type inference.⁶
- ▶ Pytype, type inference tool used by Google.⁷
- ▶ RPython: efficient compilation of a static subset of Python.⁸
- ▶ Value analysis by Fromherz et al.⁹

⁴Jensen, Møller, and Thiemann. “Type Analysis for JavaScript”. SAS 2009.

⁵Fritz and Hage. “Cost versus precision for approximate typing for Python”. PEPM 2017.

⁶Hassan, Urban, Eilers, and Müller. “MaxSMT-Based Type Inference for Python 3”. CAV 2018.

⁷Kramm et al. Pytype. 2019.

⁸Ancona, Ancona, Cuni, and Matsakis. “RPython: a step towards reconciling dynamically and statically typed OO languages”. DLS 2007.

⁹Fromherz, Ouadjaout, and Miné. “Static Value Analysis of Python Programs by Abstract Interpretation”. NFM 2018.

Types | Experimental evaluation

Name	LOC	Mopsa	⚠	Fritz & Hage	Pypyte	Typpete	Fromherz et al.	RPython
bellman_ford.py	61	0.24s	0†	1.4s	0.99s	1.4s	2.4m	7.1s
float.py	63	82ms	0†	1.7s	0.92s	1.3s	0.84s	5.6s
coop_concat.py	64	43ms	0†	1.8s	0.81s	1.3s	20ms	⌚⌚
crafting.py	132	0.41s	0†🔑	1.6s	0.97	1.7s	⌚⌚	⌚⌚
nbody.py	157	0.80s	1†🔑*	1.7s	1.3s	⌚⌚	⌚⌚	⌚⌚
chaos.py	324	2.3s	0†*	13s	11s	⌚⌚	⌚⌚	⌚⌚
scimark.py	416	0.55s	2†	8.5s	4.4s	⌚⌚	⌚⌚	⌚⌚
richards.py	426	5.0s	2†*	38s	2.4s	⌚⌚	⌚⌚	7.8s
unpack_seq.py	458	4.2s	0*	1.1s	7.4s	2.7s	14s	⌚⌚
go.py	461	15s	32†*	8.5s	3.4s	⌚⌚	⌚⌚	⌚⌚
hexiom.py	674	22s	25†🔑*	⌚⌚	4.2s	⌚⌚	⌚⌚	⌚⌚
regex_v8.py	1792	15s	0†	4.9s	⌚	1.7m	⌚⌚	⌚⌚
processInput.py	1417	4.8s	7†🔑*	2.4s	11s	⌚⌚	⌚⌚	⌚⌚
choose.py	2562	46s	17🔑†*	1.7s	15s	⌚⌚	⌚⌚	⌚⌚

⌚ unsupported by the analyzer (crash) ⌄ timeout (after 1h)

Smashed exceptions: KeyError 🔑, IndexError †, ValueError *

Name	LOC	Mopsa	⚠	Fritz & Hage	Ptype	Typpete	Fromherz et al.	RPython
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crafting.py	132	0.41s	0†🔑	1.6s	0.71s	1.6s	⌚⌚	⌚⌚
nbody.py	157						⌚⌚	⌚⌚
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regex_v8.py	1792	15s	0†	4.9s	⌚	1.7m	⌚⌚	⌚⌚
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Conclusion

- ▶ Handling Python's dynamicity
- ▶ Good scalability (w.r.t. other semantic tools)

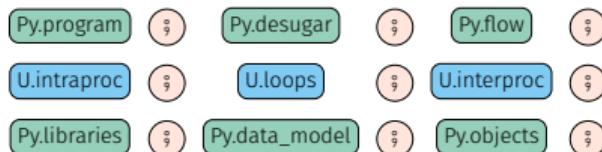
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Types \rightsquigarrow values | Configurations

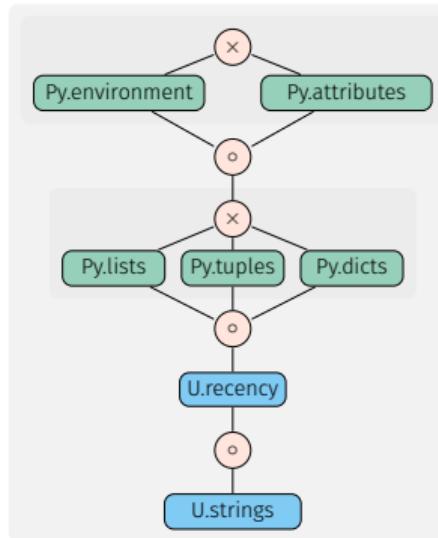
Thanks to Mopsa, switching from types to values is straightforward!

Types \rightsquigarrow values | Configurations

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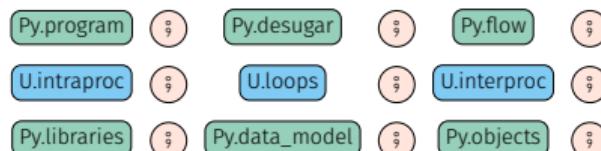
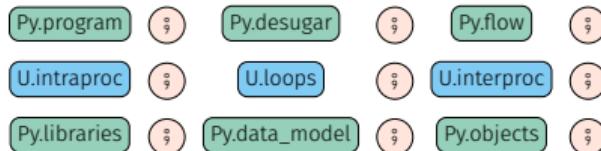


- ∅ Sequence
- × Cartesian product
- Composition
- Universal
- Python specific

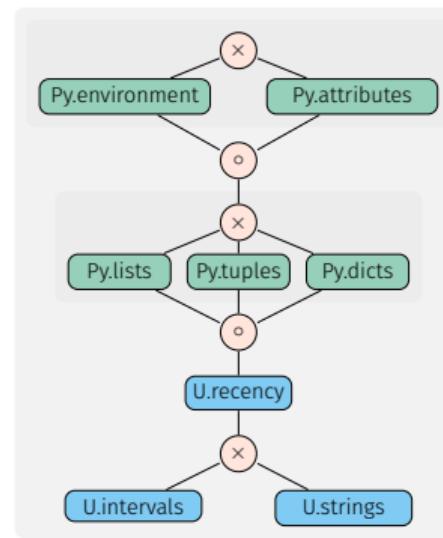
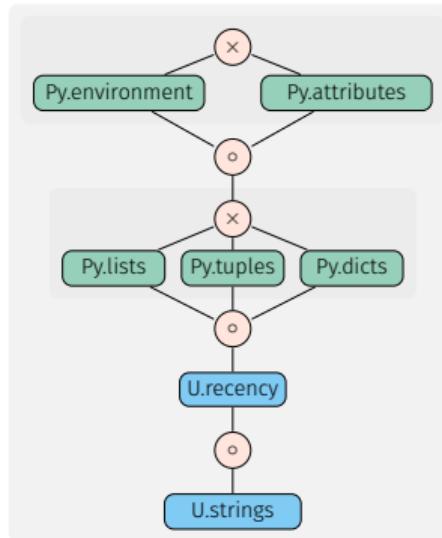


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- (?) Sequence
- (x) Cartesian product
- (o) Composition
- (blue) Universal
- (green) Python specific



Types \rightsquigarrow values | Comparing the analyses

```
Averaging tasks
1 class Task:
2     def __init__(self, weight):
3         if weight < 0: raise ValueError
4         self.weight = weight
5
6     def average(l):
7         m = 0
8         for i in range(len(l)):
9             m = m + l[i].weight
10        m = m // (i + 1)
11        return m
12
13 l = []
14 for i in range(randint(5, 10)):
15     l.append(Task(randint(0, 20)))
16 m = average(l)
```

Type analysis

- **ValueError** (l. 3)

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- ▶ **IndexError** (l. 9)

Types ↵ values | Comparing the analyses

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Non-relational value analysis

IndexError (l. 9)

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Non-relational value analysis

IndexError (l. 9)

Relational value analysis

No alarm!

Types \rightsquigarrow values | Comparing the analyses (II)

Name	LOC	Type Analysis						Non-relational Value Analysis						
		Time	Mem.	Exceptions detected			Time	Mem.	Exceptions detected			Type	Index	Key
				Type	Index	Key			Type	Index	Key			
nbbody.py	157	1.5s	3MB	0	22	1	5.7s	9MB	0	1	1			
scimark.py	416	1.4s	12MB	1	1	0	3.4s	27MB	1	0	0			
richards.py	426	13s	112MB	1	4	0	17s	149MB	1	2	0			
unpack_seq.py	458	8.3s	7MB	0	0	0	9.4s	6MB	0	0	0			
go.py	461	27s	345MB	33	20	0	2.0m	1.4GB	33	20	0			
hexiom.py	674	1.1m	525MB	0	46	3	4.7m	3.2GB	0	21	3			
regex_v8.py	1792	23s	18MB	0	2053	0	1.3m	56MB	0	145	0			
processInput.py	1417	10s	64MB	7	7	1	12s	85MB	7	4	1			
choose.py	2562	1.1m	1.6GB	12	22	7	2.9m	3.7GB	12	13	7			
Total	9294	4.0m	2.8GB	59	2214	12	13m	9.1GB	59	228	12			

Types \rightsquigarrow values | Comparing the analyses (II)

Name	LOC	Type Analysis			Non-relational Value Analysis				
		Time	Mem.	Exceptions detected	Time	Mem.	Exceptions detected	Index	Key
		Type	Index	Key					
nbbody.py	157	1.5s	—	—	—	—	—	1	1
scimark.py	416	1.4s	—	—	—	—	—	0	0
richards.py	426	13s	—	—	—	—	—	2	0
unpack_seq.py	458	8.3s	—	—	—	—	—	0	0
go.py	461	27s	—	—	—	—	—	20	0
hexiom.py	674	1.1m	—	—	—	—	—	21	3
regex_v8.py	1792	23s	—	—	—	—	—	145	0
processInput.py	1417	10s	—	—	—	—	—	4	1
choose.py	2562	1.1m	—	—	—	—	—	13	7
Total	9294	4.0m	2.8GB	59	2214	12	13m	9.1GB	59
								228	12

Conclusion

The non-relational value analysis

- ▶ does not remove false type alarms
- ▶ significantly reduces index errors
- ▶ is $\simeq 3\times$ costlier

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		Time	Mem.	Exceptions detected	Time	Mem.	Exceptions detected	Index	Key
Type	Index	Key	Type	Index	Key	Index	Key	Index	Key
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scimark.py	416	1.4s	—	—	—	—	—	0	0
richards.py	426	13s	—	—	—	—	—	2	0
unpack_seq.py	458	8.3s	—	—	—	—	—	0	0
go.py	461	27s	—	—	—	—	—	20	0
hexiom.py	674	1.1m	—	—	—	—	—	21	3
regex_v8.py	1792	23s	—	—	—	—	—	145	0
processInput.py	1417	10s	—	—	—	—	—	4	1
choose.py	2562	1.1m	—	—	—	—	—	13	7
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Conclusion

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Heuristic packing and relational analyses

- ▶ Static packing, using function's scope
- ▶ Rules out all 145 alarms of `regex_v8.py` (1792 LOC) at $2.5\times$ cost

Selectivity of the non-relational value analysis

Name	Attributes	Types	Indexes	Keys	Values	Overflows	Divisions
scimark.py	746/746	844/844	2/5		29/30	21/43	20/21
richards.py	352/353	389/389	2/4		2/3		2/2
unpack_seq.py	807/807	1210/1210			1/1		
go.py	664/697	728/728	2/20		7/7	6/12	4/6
hexiom.py	598/598	672/672	10/32	0/3	23/24		
regex_v8.py	7357/7357	8349/8349	1913/2057		63/63		
processInput.py	617/619	790/792	12/12	0/1	0/1	2/2	
choose.py	2519/2521	2997/2999	28/39	4/8	9/24	7/17	

Selectivity of the analysis on some classes of exceptions

Selectivity = Number of proved safe operations / Total number of checks

An empty cell denotes a program where the kind of exception cannot happen

Two soundnesses

- ▶ Modelization of the semantics from CPython
- ▶ Implementation of this semantics within Mopsa

Soundness

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- ▶ Implementation of this semantics within Mopsa

Our approach

- ▶ Test only in the abstract
- ▶ Issue of overapproximations and unproved assertions

Unsupported constructs

- ▶ eval
- ▶ Recursive functions
- ▶ Finalizers

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Unsupported constructs

- ▶ eval
- ▶ Recursive functions
- ▶ Finalizers

Tests from previous works

- ▶ 450/586 tests supported
- ▶ 268/586 assertions proved

Official tests from CPython

- ▶ 325/416 tests supported (17 chosen files)
- ▶ 389/702 assertions proved

Analyzing Python Programs with C Libraries

Combining C and Python – motivation

One in five of the top 200 Python libraries contains C code

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- ▶ Different values (arbitrary-precision integers in Python, bounded in C)

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Pitfalls

- ▶ Different values (arbitrary-precision integers in Python, bounded in C)
- ▶ Different object representations (Python objects, C structs)
- ▶ Different runtime-errors (exceptions in Python)
- ▶ Garbage collection

A combined static analysis of C/Python¹⁰

- ▶ Targeting C extensions using the CPython API

¹⁰ Monat, Ouadjaout, and Miné. “A Multilanguage Static Analysis of Python Programs with Native C Extensions”. SAS 2021.

A combined static analysis of C/Python¹⁰

- ▶ Targeting C extensions using the CPython API
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- ▶ Targeting C extensions using the CPython API
 - ▶ To detect runtime errors (in C, Python, and the “glue”)
 - ▶ Observations
 - allocated objects are shared in the memory
 - but each language has different abstractions
- ⇒ Share universal domains and synchronize abstractions

¹⁰ Monat, Ouadjaout, and Miné. “A Multilanguage Static Analysis of Python Programs with Native C Extensions”. SAS 2021.

Combining C and Python – example

counter.c

```
1 typedef struct {
2     PyObject_HEAD;
3     int count;
4 } Counter;
5
6 static PyObject*
7 CounterIncr(Counter *self, PyObject *args)
8 {
9     int i = 1;
10    if(!PyArg_ParseTuple(args, "|i", &i))
11        return NULL;
12
13    self->count += i;
14    Py_RETURN_NONE;
15 }
16
17 static PyObject*
18 CounterGet(Counter *self)
19 {
20     return Py_BuildValue("i", self->count);
21 }
```

count.py

```
1 from counter import Counter
2 from random import randrange
3
4 c = Counter()
5 power = randrange(128)
6 c.incr(2**power-1)
7 c.incr()
8 r = c.get()
```

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```

► $\text{power} \leq 30 \Rightarrow r = 2^{\text{power}}$

Combining C and Python – example

counter.c

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1 typedef struct {
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13    self->count += i;
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count.py

```
1 from counter import Counter
2 from random import randrange
3
4 c = Counter()
5 power = randrange(128)
6 c.incr(2**power-1)
7 c.incr()
8 r = c.get()
```

- ▶ $\text{power} \leq 30 \Rightarrow r = 2^{\text{power}}$
- ▶ $32 \leq \text{power} \leq 64$: OverflowError:
signed integer is greater than maximum
- ▶ $\text{power} \geq 64$: OverflowError:
Python int too large to convert to C long

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How to analyze multilanguage programs?

Type annotations

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class Counter:  
    def __init__(self): ...  
    def incr(self, i: int = 1): ...  
    def get(self) -> int: ...
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How to analyze multilanguage programs?

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- ▶ Typeshed: type annotations for the standard library, used in the single-language analysis before

How to analyze multilanguage programs?

Type annotations

Rewrite into Python code

```
class Counter:  
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```

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```

- ▶ No integer wrap-around in Python
- ▶ Some effects can't be written in pure Python (e.g., read-only attributes)

How to analyze multilanguage programs?

Type annotations

Rewrite into Python code

Drawbacks of the current approaches

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Our approach

- ▶ Analyze both the C and Python sources
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- ▶ Reuse previous analyses of C and Python
- ▶ Detect runtime errors in Python, in C, and at the boundary

Analysis result

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counter.c	count.py
<pre>1 typedef struct { 2 PyObject_HEAD; 3 int count; 4 } Counter; 5 6 static PyObject* 7 CounterIncr(Counter *self, 13: self->count += i; 8 { 9 int i = 1; 10 if(!PyArg_ParseTuple(a 11 return NULL; 12 13 self->count += i; 14 Py_RETURN_NONE; 15 } 16 17 static PyObject* 18 CounterGet(Counter *self) 6: c.incr(2**p-1) 19 { 20 return Py_BuildValue("U"); 21 }</pre>	<pre>1 from counter import Counter 2 from random import randrange 3 4 △ Check #430: 5 ./counter.c: In function 'CounterIncr': 6 ./counter.c:13.2-18: warning: Integer overflow 7 8 Callstack: 9 from count.py:8.0-8: CounterIncr 10 11 ('self->count + i)' has value [0,2147483648] that is larger 12 than the range of 'signed int' = [-2147483648,2147483647] 13 14 Callstack: 15 from count.py:8.0-8: CounterIncr 16 17 X Check #506: 18 count.py: In function 'PyErr_SetString': 19 count.py:6.0-14: error: OverflowError exception 20 21 Callstack: 22 from ./counter.c:17.6-38::convert_single[0]: PyParseTuple_int 23 from count.py:7.0-14: CounterIncr 24 25 +1 other callstack</pre>

Concrete definition

- ▶ Builds upon the Python and C semantics

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- ▶ C access to Python objects only through the API (verified by Mopsa)

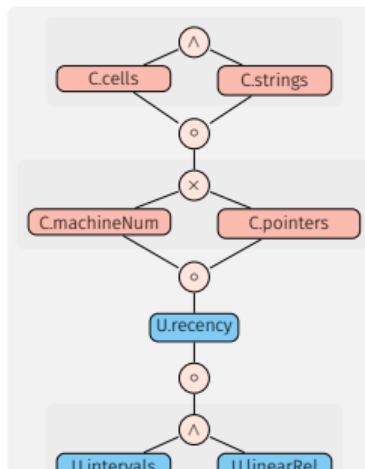
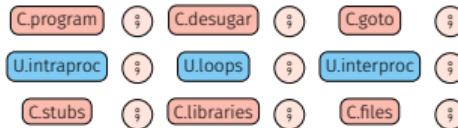
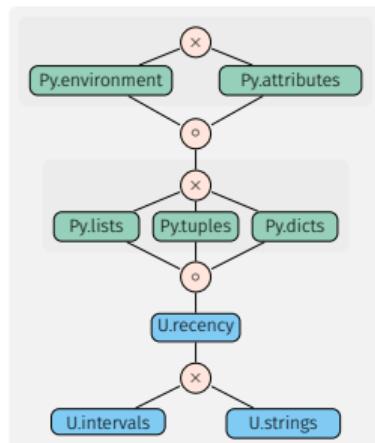
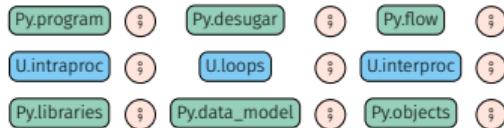
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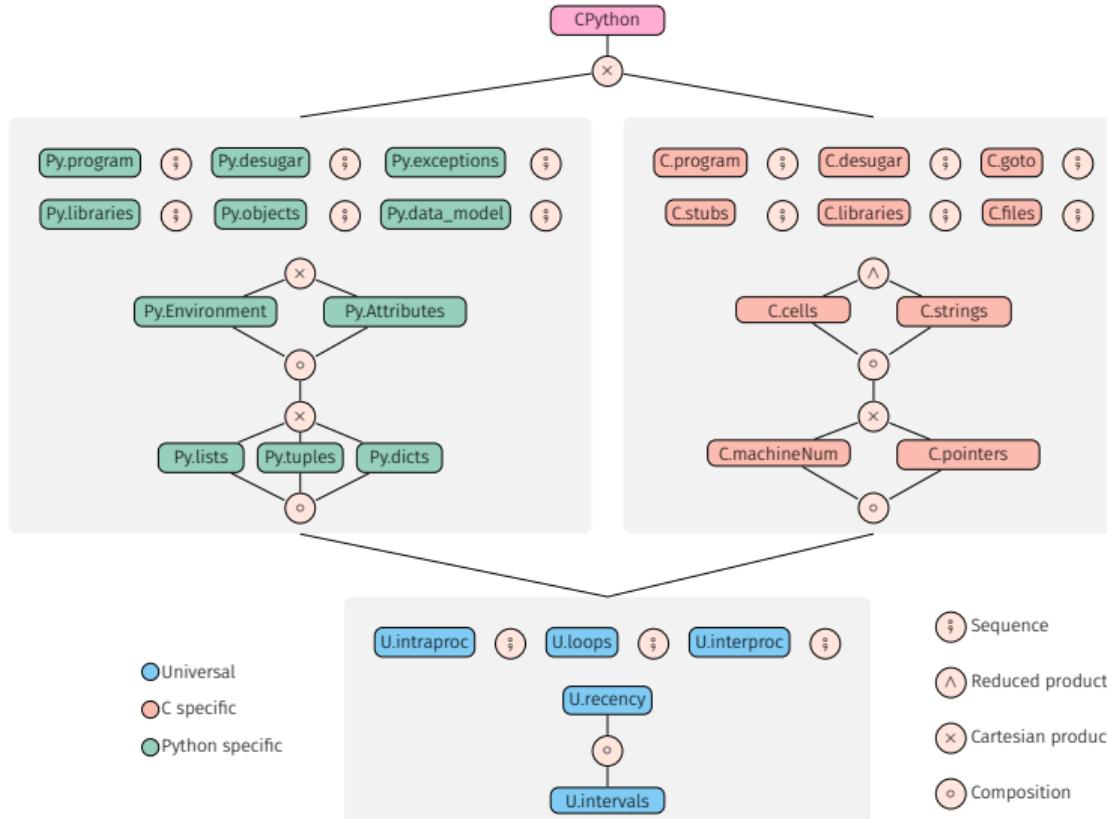
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- ▶ Manual modelization from CPython's source code

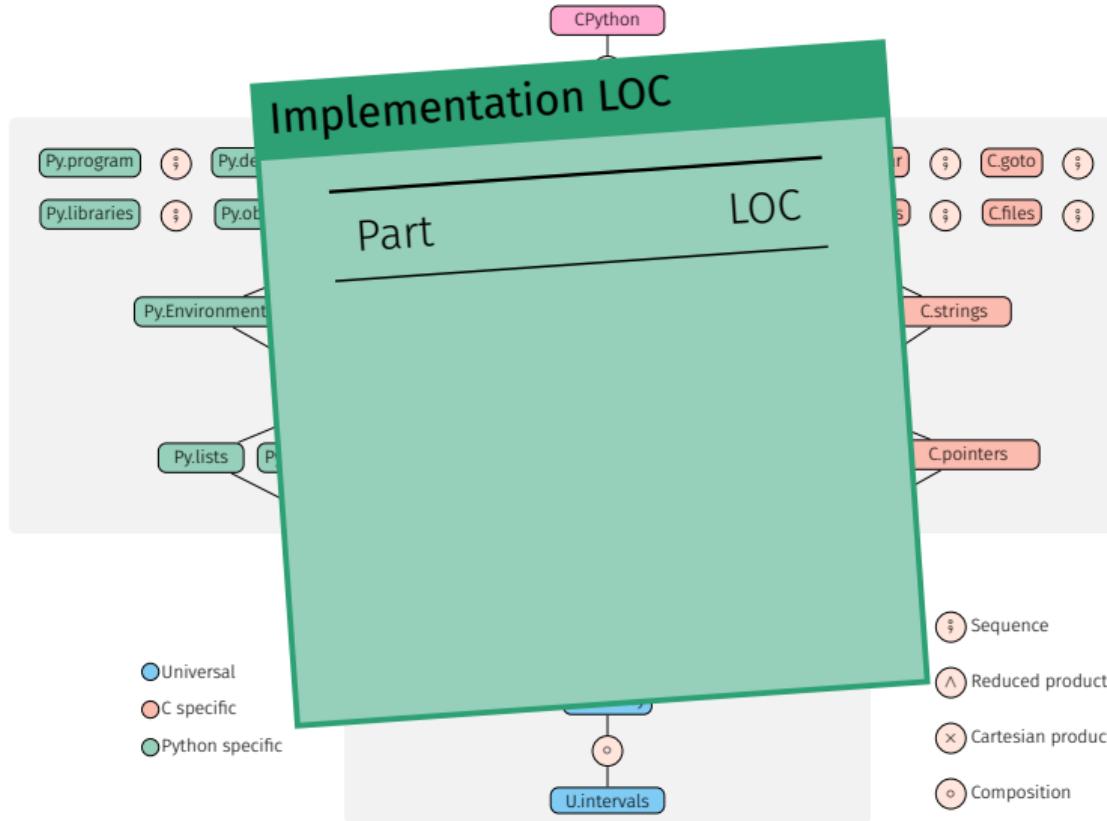
From distinct Python and C analyses...



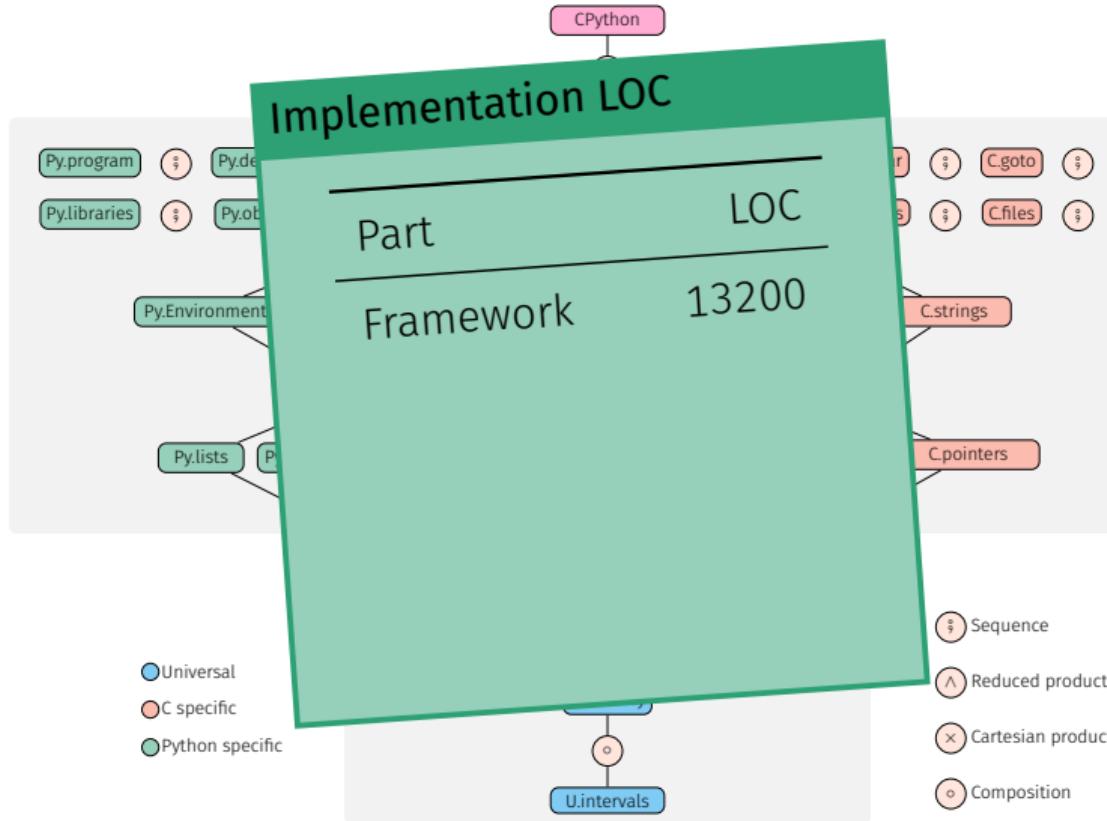
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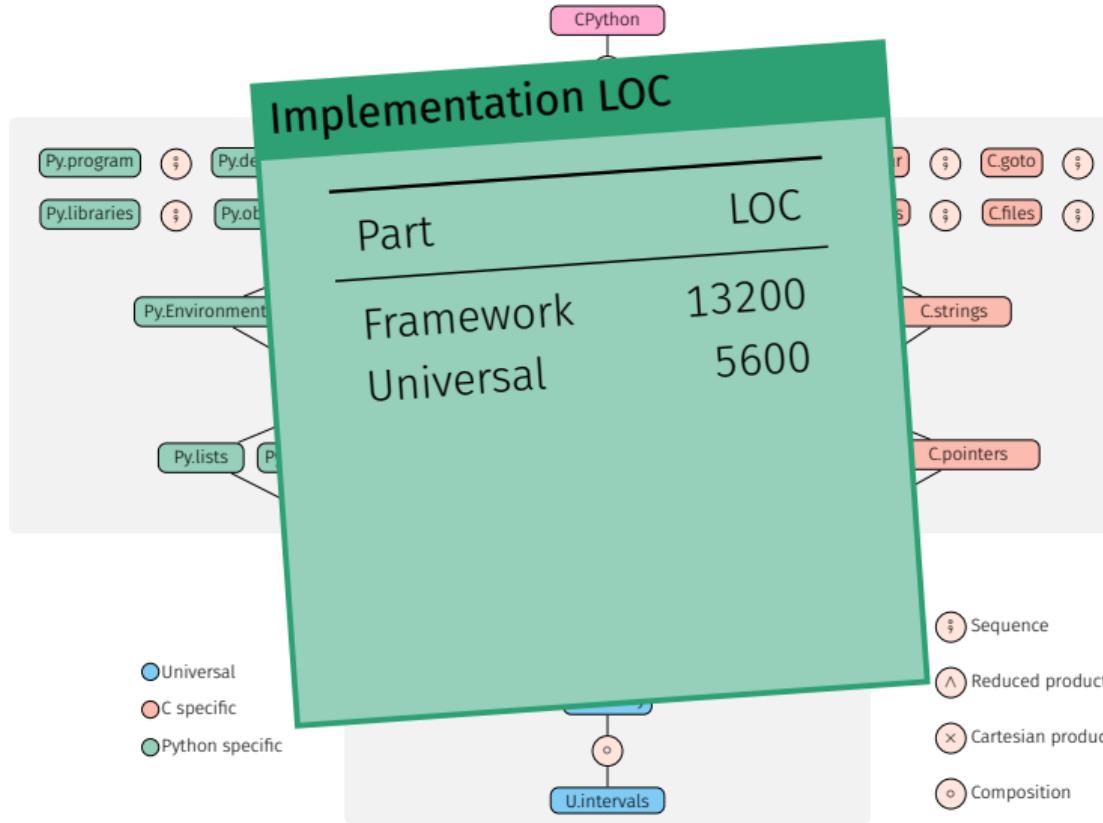
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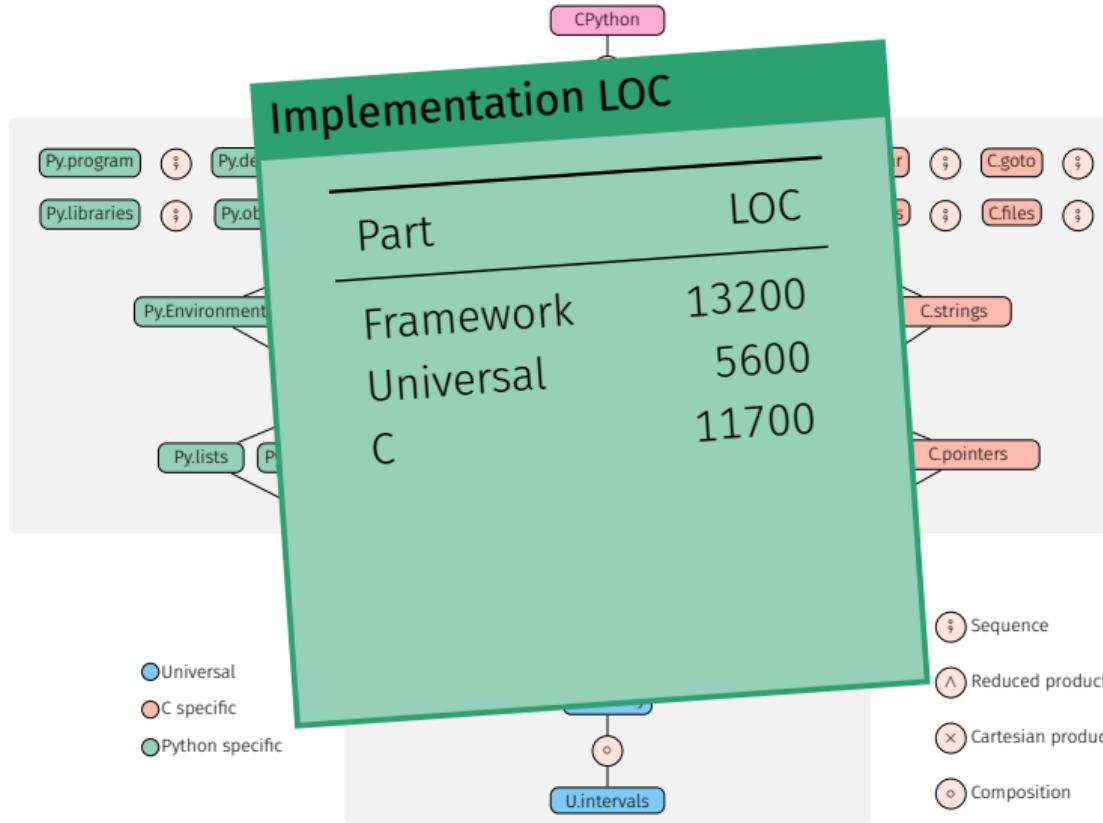
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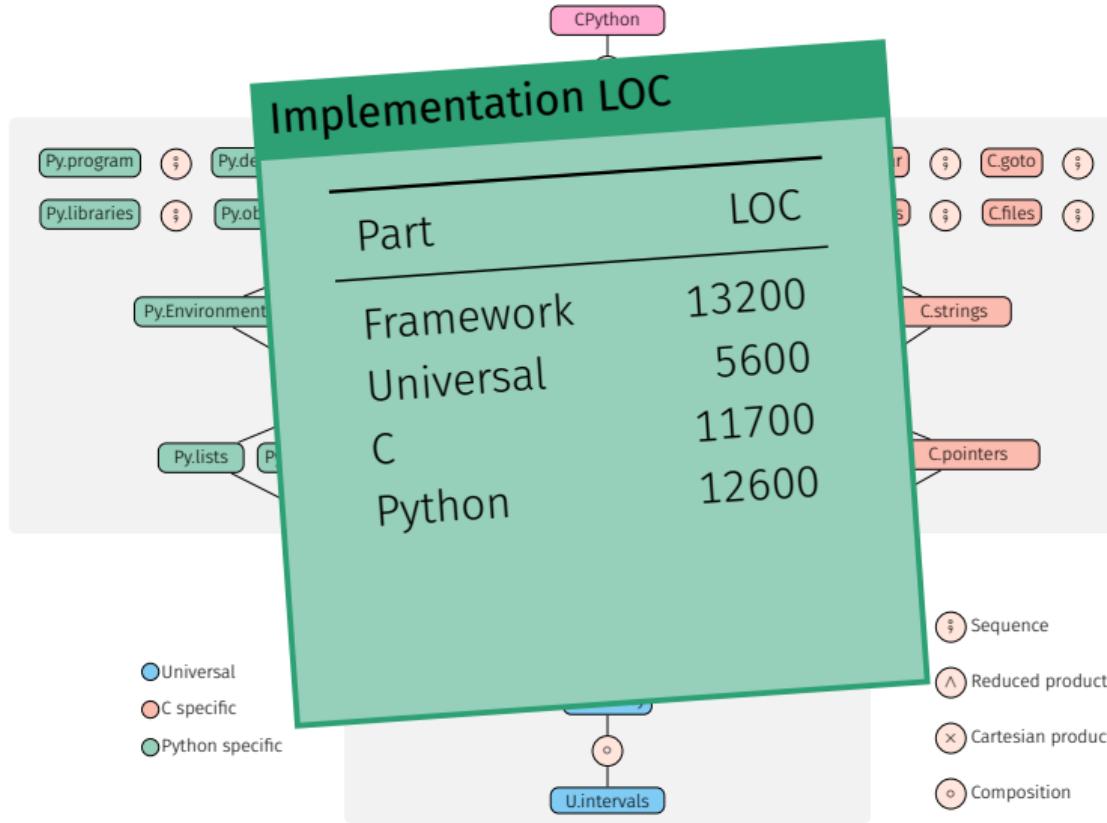
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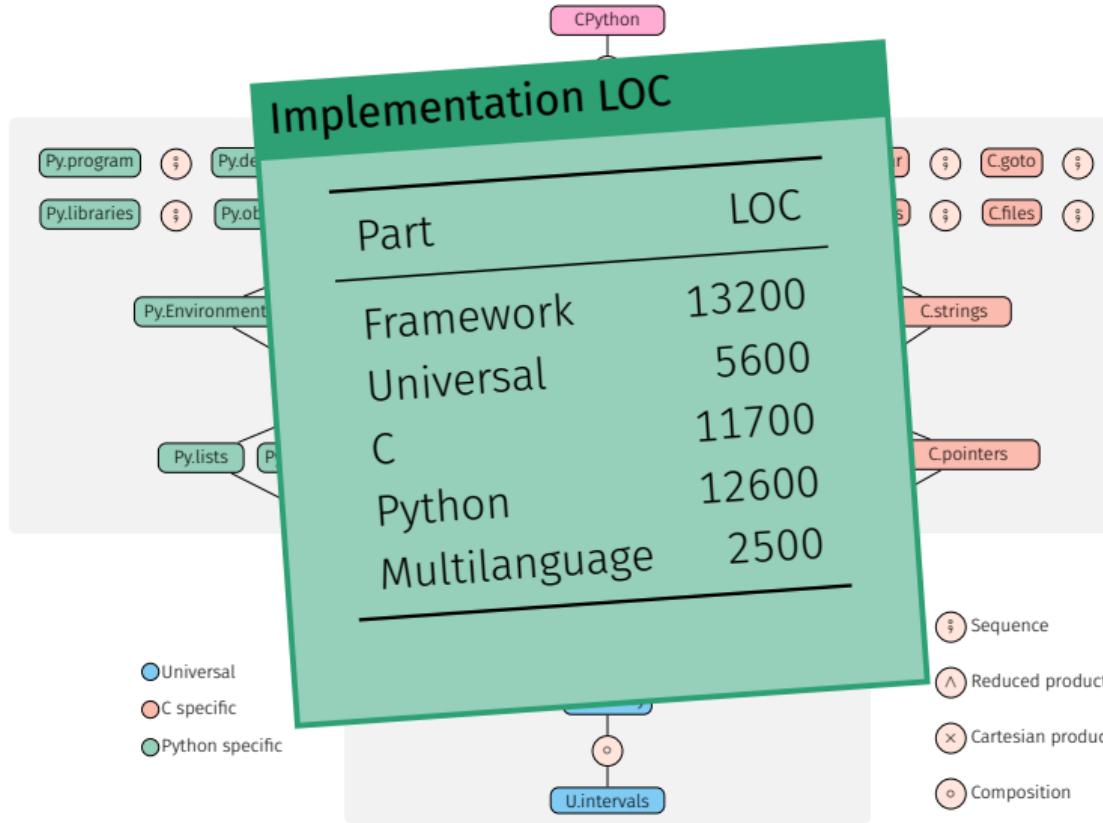
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Corpus selection

- ▶ Popular, real-world libraries available on GitHub, averaging 412 stars.
- ▶ Whole-program analysis: we use the tests provided by the libraries.

Library	C	Py	Tests	⌚	✖	✓	Assertions	Py ↔ C
noise	722	675	15/15	18s	99.6%	(4952)	100.0%	(1738) 0/21 6.5
ahocorasick	3541	1336	46/92	54s	93.1%	(1785)	98.0%	(4937) 30/88 5.4
levenshtein	5441	357	17/17	1.5m	79.9%	(3106)	93.2%	(1719) 0/38 2.7
cdistance	1433	912	28/28	1.9m	95.3%	(1832)	98.3%	(11884) 88/207 8.7
llist	2829	1686	167/194	4.2m	99.0%	(5311)	98.8%	(30944) 235/691 51.7
bitarray	3244	2597	159/216	4.2m	96.3%	(4496)	94.6%	(21070) 100/378 14.8

$$\frac{\text{safe C checks}}{\text{total C checks}} \%$$

total C checks

average # transitions
between Python and C
per test

Theoretical frameworks

- ▶ Matthews and Findler¹¹ boundary functions as value conversions between two languages.
- ▶ Buro, Crole, and Mastroeni¹² generic framework for combining analyses of different languages.

¹¹ Matthews and Findler. "Operational semantics for multi-language programs". 2009.

¹² Buro, Crole, and Mastroeni. "On Multi-language Abstraction - Towards a Static Analysis of Multi-language Programs". SAS 2020.

Around the Java Native Interface (JNI)

Static translation of some of C's effects, injected back into the Java analysis.

- ▶ Effects of C code on Java heap modelized using JVML¹³
- ▶ Type inference of Java objects in C code¹⁴
- ▶ Extraction of C callbacks to Java¹⁵

- ▶ Modular analyses
- ▶ No numeric information
- ▶ Missing C runtime errors

¹³Tan and Morrisett. "Ilea: inter-language analysis across Java and C". OOPSLA 2007.

¹⁴Furr and Foster. "Checking type safety of foreign function calls". 2008.

¹⁵Lee, Lee, and Ryu. "Broadening Horizons of Multilingual Static Analysis: Semantic Summary Extraction from C Code for JNI Program Analysis". ASE 2020.

Conclusion

Contribution: concrete semantics of Python

Difficulties

- ▶ Size of the semantics
- ▶ CPython's source code

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Our results

- ▶ Semantics suitable for abstract interpretation
- ▶ Written and explained in the manuscript (70 cases)
- ▶ Backreferences to the source code
- ▶ Preliminary tests using CPython's suite

Contribution: type & value analyses of Python

Difficulties

- ▶ Dynamicity
- ▶ Dual type system
- ▶ Size of the semantics

¹⁶Monat, Ouadjaout, and Miné. “Static Type Analysis by Abstract Interpretation of Python Programs”. ECOOP 2020.

¹⁷Monat, Ouadjaout, and Miné. “Value and allocation sensitivity in static Python analyses”. SOAP@PLDI 2020.

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Our results

- ▶ Type analysis¹⁶,
- ▶ Numeric value analysis & new sensitivities for the recency abstraction¹⁷
- ▶ Relational value analysis with packing (manuscript)
- ▶ Scale to small, real-world benchmarks

¹⁶Monat, Ouadjaout, and Miné. “Static Type Analysis by Abstract Interpretation of Python Programs”. ECOOP 2020.

¹⁷Monat, Ouadjaout, and Miné. “Value and allocation sensitivity in static Python analyses”. SOAP@PLDI 2020.

Contribution: multilanguage Python/C analysis

Difficulties

- ▶ Concrete semantics
- ▶ Memory interaction

Monat, Ouadjaout, and Miné. “A Multilanguage Static Analysis of Python Programs with Native C Extensions”. SAS 2021

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Our results

- ▶ Careful separation of the states and modelization of the API
- ▶ Lightweight domain on top of off-the-shelf C and Python analyses
- ▶ Shared underlying abstractions (numeric, recency)
- ▶ Scale to small, real-world libraries (using client code)

Monat, Ouadjaout, and Miné. “A Multilanguage Static Analysis of Python Programs with Native C Extensions”. SAS 2021

Some future works

Formalizing complex semantics

Use a concrete semantics framework such as

- ▶ Skeletal semantics
- ▶ Interaction trees

In order to

- ▶ Formalize the concrete semantics of Python
- ▶ Automatically generate static analyzers

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Improving the scalability of static analyses

- ▶ Incremental analyses
- ▶ Variable packing refinement algorithm
- ▶ Compositional analyses

Static Type and Value Analysis by Abstract Interpretation of Python Programs with Native C Libraries

Questions

xkcd.com/353

Raphaël Monat, Abdelraouf Ouadjaout, Antoine Miné

MTV Seminar
16 February 2022

