Formal Verification of Python-based Al libraries

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Motivating Example

main.py

```
import numpy as np
y = np.power(3, 21, dtype=np.int32) # overflow: 1870418611
```

Challenges in Python Verification

- Dynamic Features
 - Dynamic typing, introspection, and runtime checks.
- Optional Type System
- ► Limitations of Existing Approaches
 - Requires user expertise or code annotations.
 - Cannot handle Python's dynamic aspects.
 - Produces false positives and fails to handle external libraries
 - Often fails to guarantee full program coverage.

```
1    a = "1"
2    a = 1
3    b = MyClass()
4    if isinstance(a, int):
5     b.x = "str"
6    else:
7    b.x = 2
8    c = b.x
9    d = c.upper()
```

Research Questions

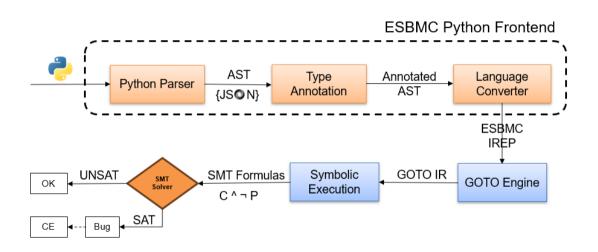
RQ1

Given the challenges of verifying Python programs and the limitations of current tools, is it possible to reuse a model checking framework to reduce time and effort while advancing the state of the art?

RQ2

Which verification properties are relevant for Python-based Al libraries, and how can a model checking framework be applied to verify them?

Our approach to verify Python Programs



JSON-Based Type Inference

Constant Values

$$x = 10 \Rightarrow x:int = 10$$

Referred Variables

$$y = x \Rightarrow y : int = x$$

Class Instances
z:MyClass = MyClass()

2:MyClass - MyClass()

► Function Calls

```
x = foo() \Rightarrow x:int = foo
```

```
{
  "_type": "Assign",
  "target": {
     "_type": "Name",
     "id": "x"
},
  "value": {
     "_type": "Constant",
     "value": 10
}
}
```

```
x = 10
```

```
"_type": "AnnAssign",
"target": {
  "_type": "Name",
  "id" · "v"
"annotation": {
  "_type": "Name",
  "id": "int"
"value": {
  "_type":"Constant".
  "value": 10
```

x:int = 10

Research Progress

- Python features
 - Language basics: Data types, conditionals, loops, functions.
 - OOP: classes, inheritance, polymorphism
 - Containers: Strings, Lists
- ▶ Non-determinism: Used to model unspecified inputs.
- Operational models: Simplified stubs to reduce external libraries complexity.
- Verification Properties
 - Arithmetic overflow, out-of-bounds access, division-by-zero, user assertions.
- Case Studies
 - Ethereum Consensus Verification: 1 bug confirmed; 7 bugs in analysis.
 - Numpy Verification: Arrays construction, math correctness.

Verification of Ethereum Specifications

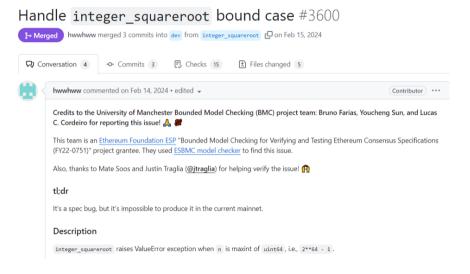
- **Consensus Specification**: A set of runnable specifications in Python.
- ▶ Detected overflow and div-by-zero in a function call.

main.py

```
def integer_squareroot(n: uint64)
   -> uint64:
   x = n
   y = (x + 1) // 2
   while y < x:
        x = y
        y = (x + n // x) // 2
   return x</pre>
```

```
[Counterexample]
State 1 main.pv line 2
State 2 main.py line 3
y = 0
State 3 main.py line 5
x = 0
State 4 main.py line 6
Violated property:
division by zero
x != 0
```

Verification of Ethereum Specifications



Software Vulnerabilities in Python Al Libraries

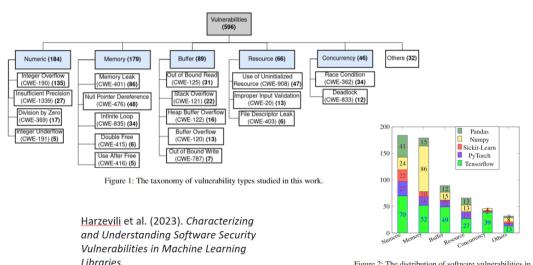


Figure 2: The distribution of software vulnerabilities in different ML libraries.

NumPy Properties of Interest

Property	Description	Check
Array shape consistency	Ensure operations on arrays are per-	Assert that shapes on input arrays match
	formed on arrays with compatible sizes.	before performing operations.
Array element type consis-	Ensure array elements are of the expected	Use assertions or type checks to enforce
tency	type (e.g. all elements should be integers,	element type consistency.
	float).	
Index out of bounds	Ensure that all accesses to arrays are	Assert that indices used for accessing ar-
	within bounds.	rays are within the bounds.
Mathematical Correctness	Ensure the correctness of numerical op-	Use assertions to check that the result of
	erations (addition, multiplication, etc)	a calculation matches an expected value.
Abscence of Arithmetic Over-	Ensure numeric operations do not	Use symbolic execution to track integer
flow	cause overflow, particularly with custom	values and ensure they stay within types
	bounded integer types.	bounds.

Verifying NumPy Programs with ESBMC

- ▶ Black-Box Verification with ESBMC
 - Analyze library behaviour via assertions from function calls.

main.py

```
import numpy as np
x = np.add(2147483647, 1, dtype=np.int32)
```

```
add-overflow git:(master) ② more main.py
import numpy as np

x = np.add(2147483647, 1, dtype=np.int32)
add-overflow git:(master) ②
add-overflow git:(master) ② python main.py
add-overflow git:(master) ③
```

```
[Counterexample]
State 1 file main.py line 2
------
Violated property:
  file main.py line 2 column 0
  arithmetic overflow on add
  !overflow("+", 2147483647, 1)

VERIFICATION FAILED
```

Takeway messages

- ▶ Python presents several verification challenges, but we are already able to handle its static aspects and have validated our approach by reusing a BMC framework.
- ▶ NumPy and AI contain real-world issues, and our approach has the potential to uncover new ones.
- Our next steps focus on expanding language coverage, handling the dynamic aspects and integration with native libraries.

Thank you