

Training Binary Classifiers as Data Structure Invariants

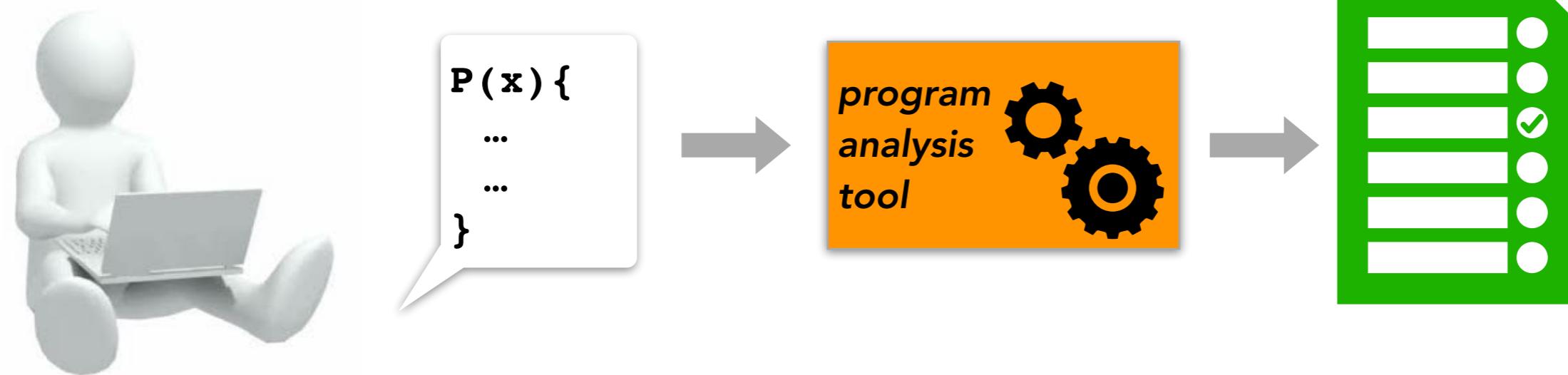
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CONICET, Argentina

*in collaboration with Renzo Degiovanni, Pablo Ponzio, Germán Regis, Nazareno Aguirre
and Marcelo Frias*

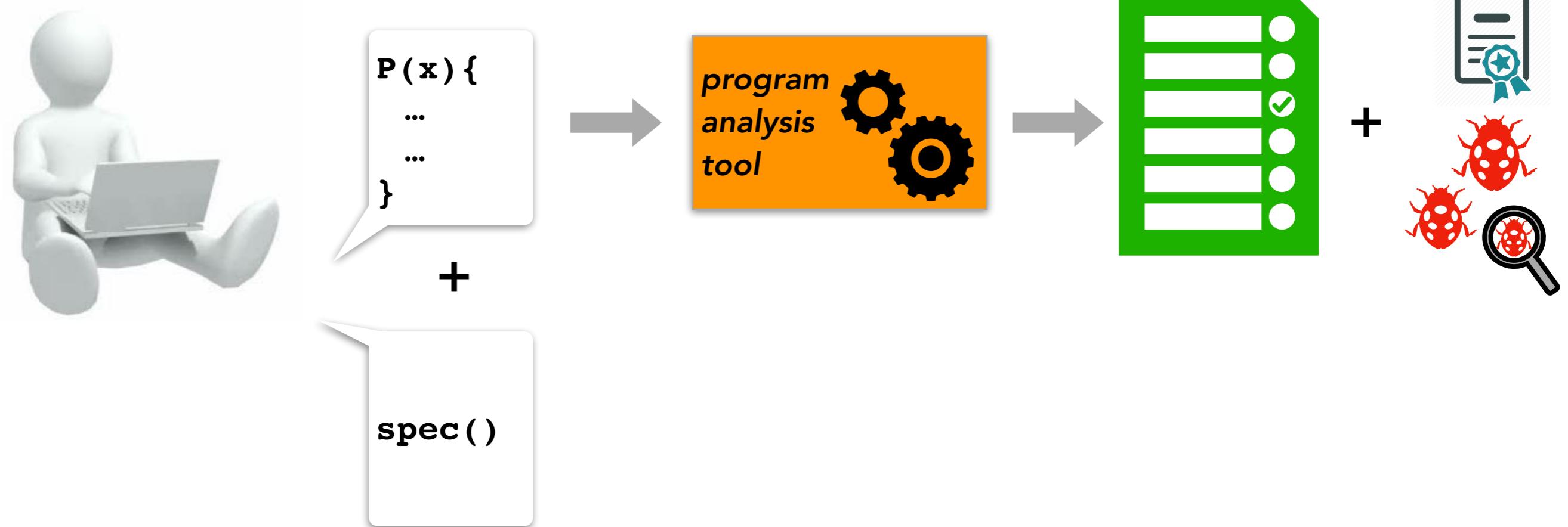


An Automated Analysis Scenario



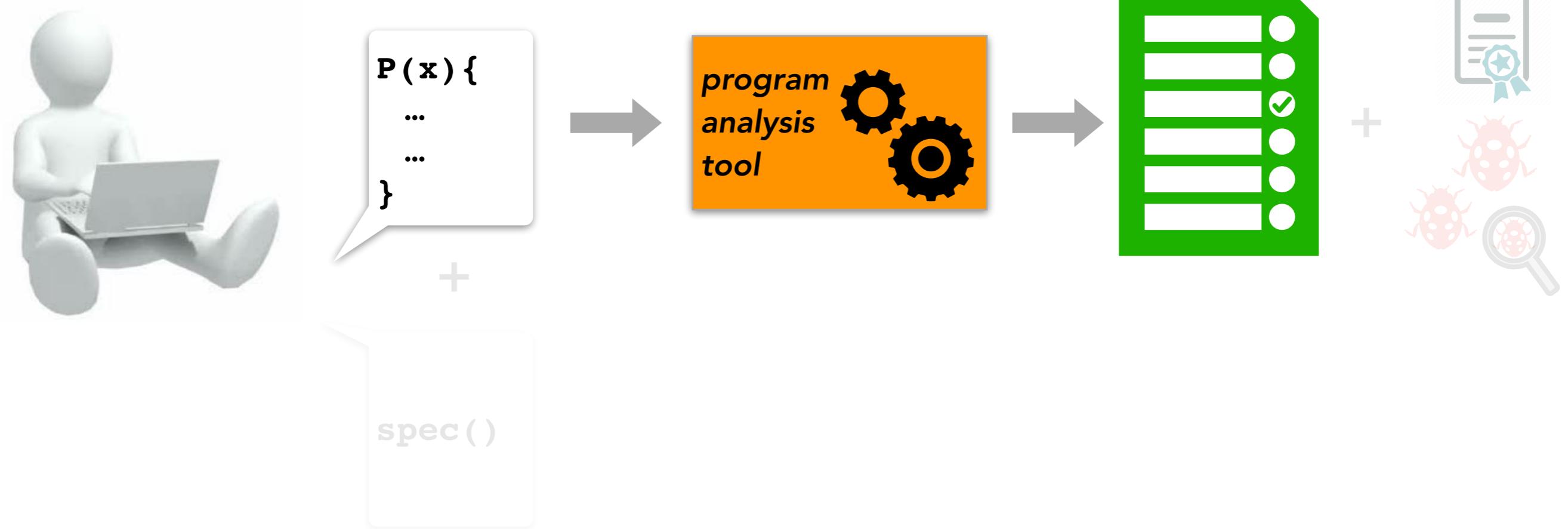
An Automated Analysis Scenario

improved analysis in the presence of specifications



An Automated Analysis Scenario

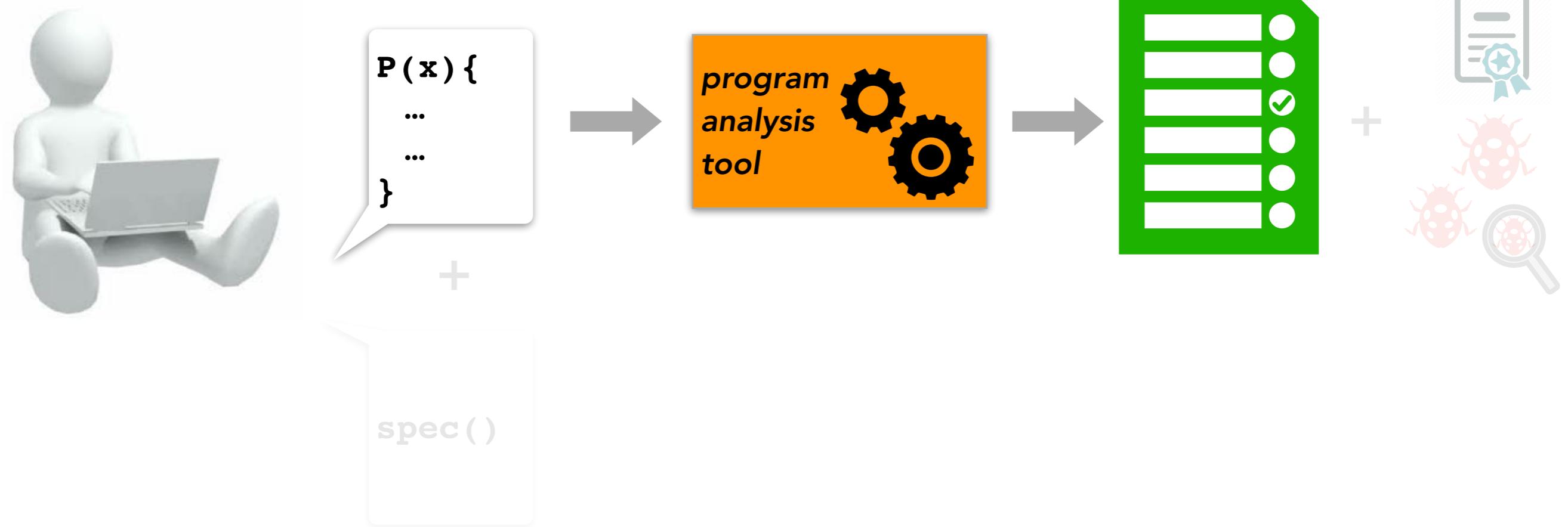
improved analysis in the presence of specifications



unfortunately, specifications
are sometimes unavailable

An Automated Analysis Scenario

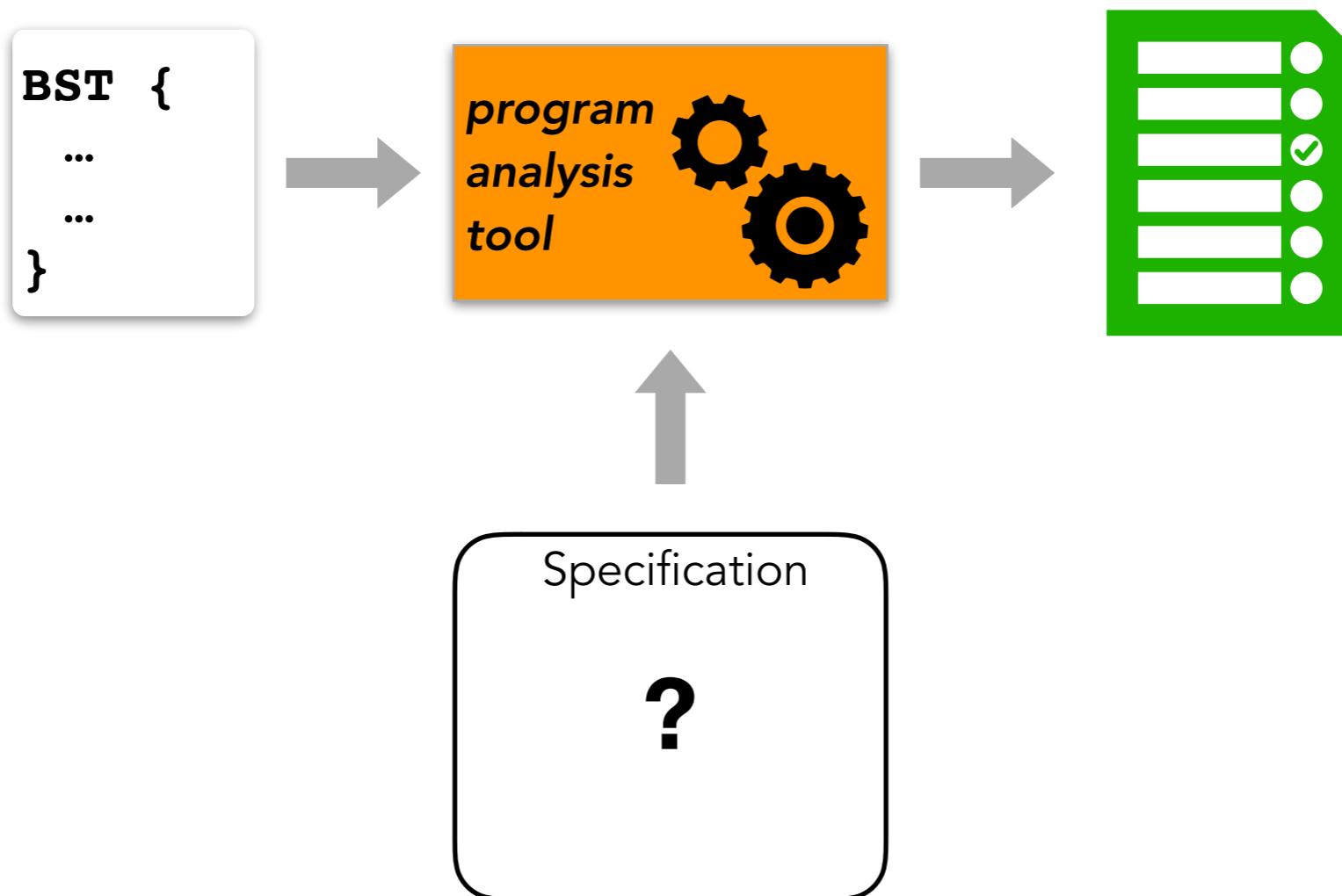
improved analysis in the presence of specifications



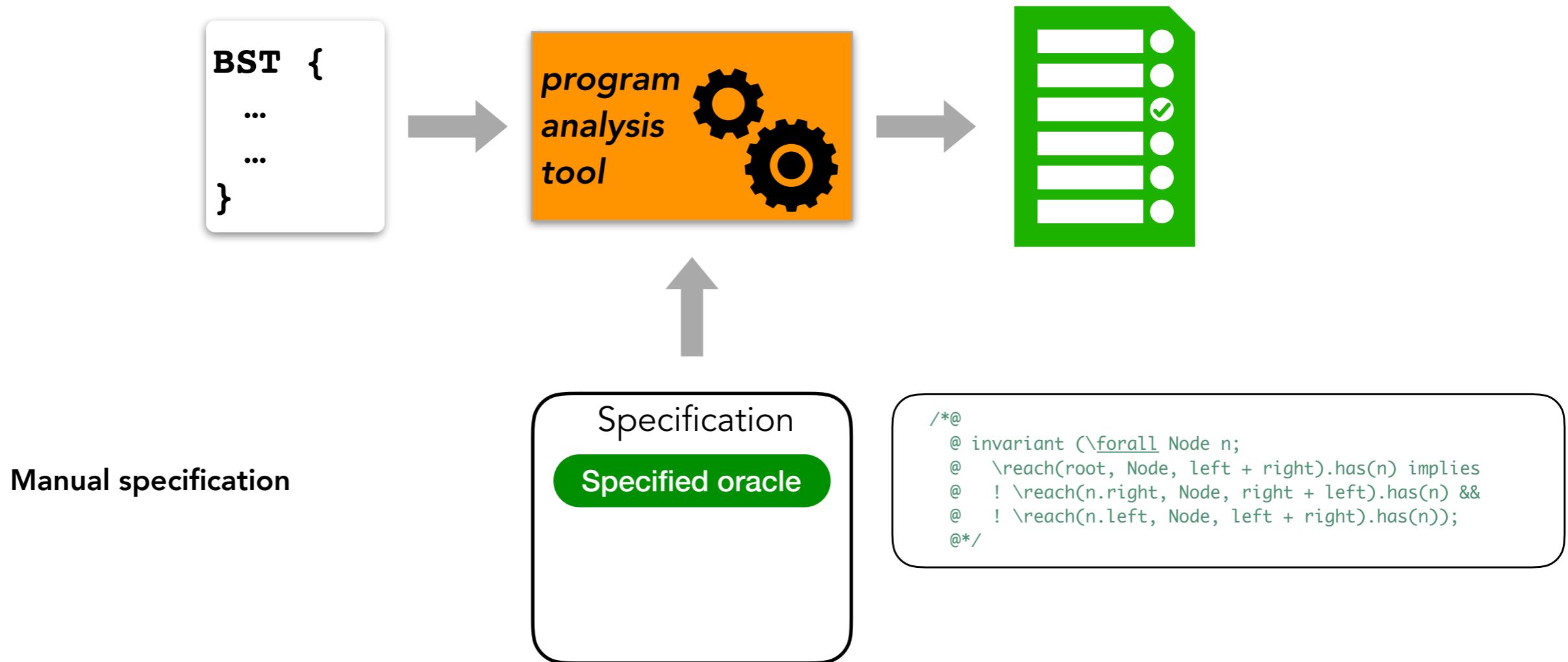
unfortunately, specifications
are sometimes unavailable

This emphasizes the relevance of the **oracle problem**

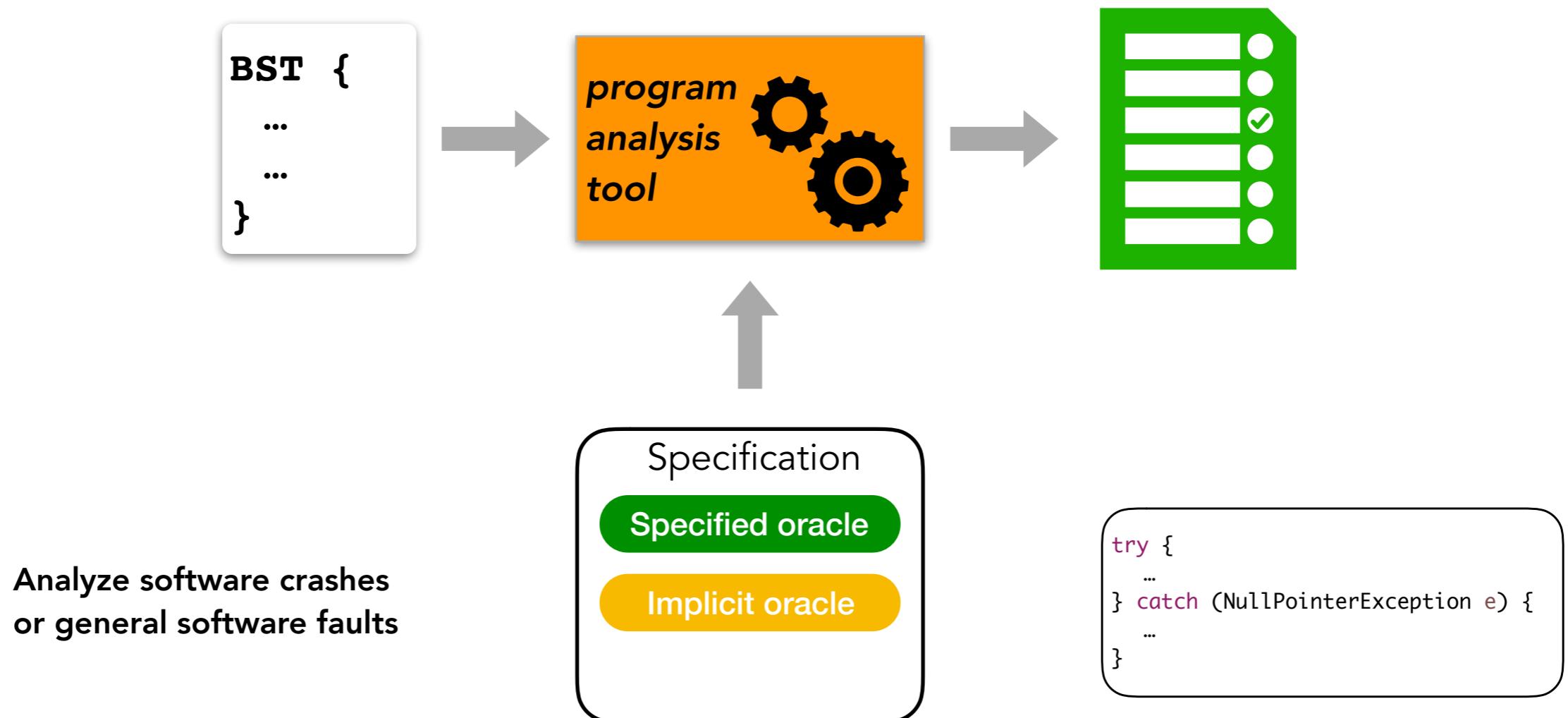
Approaches to the Oracle Problem



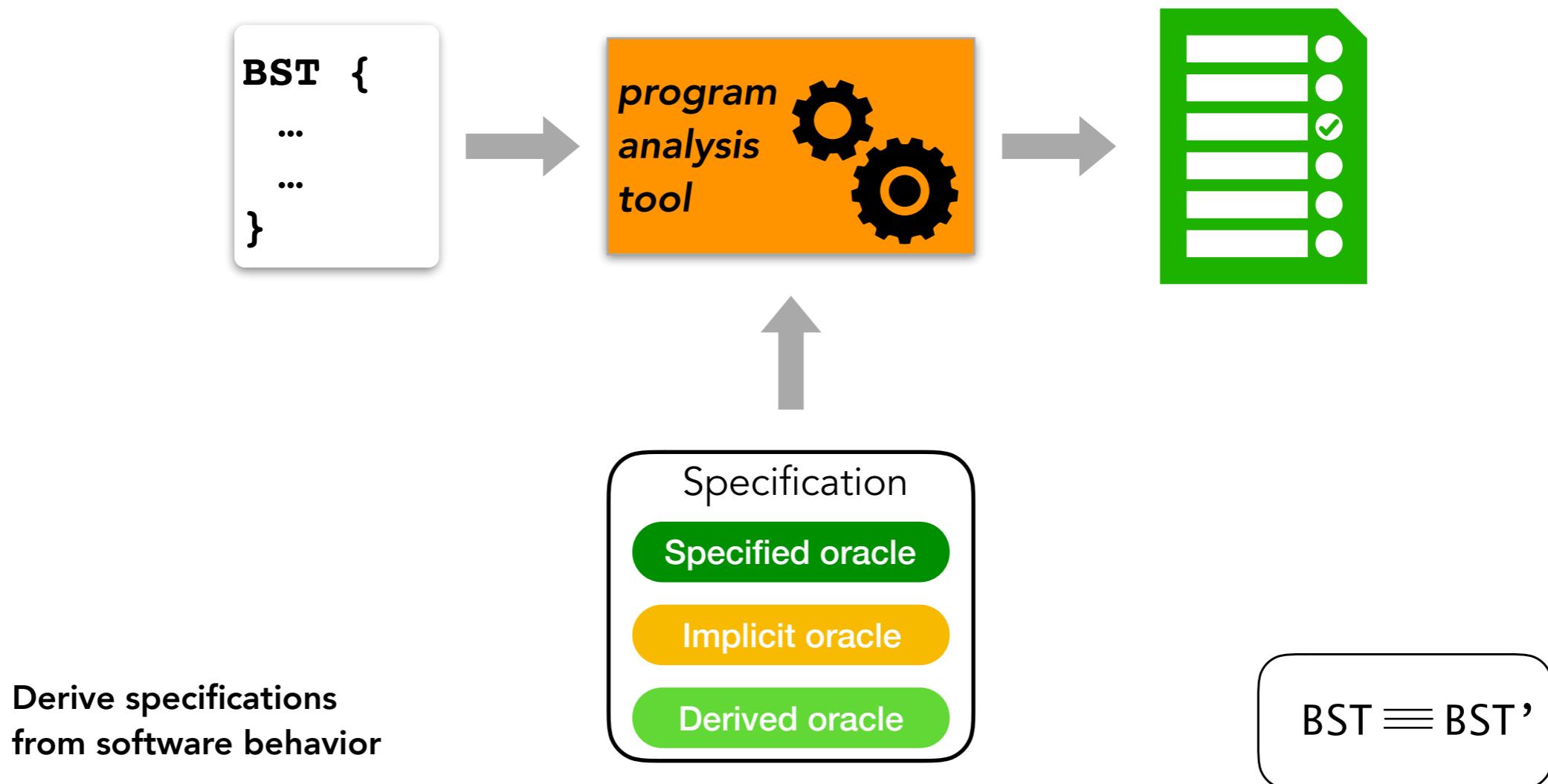
Approaches to the Oracle Problem



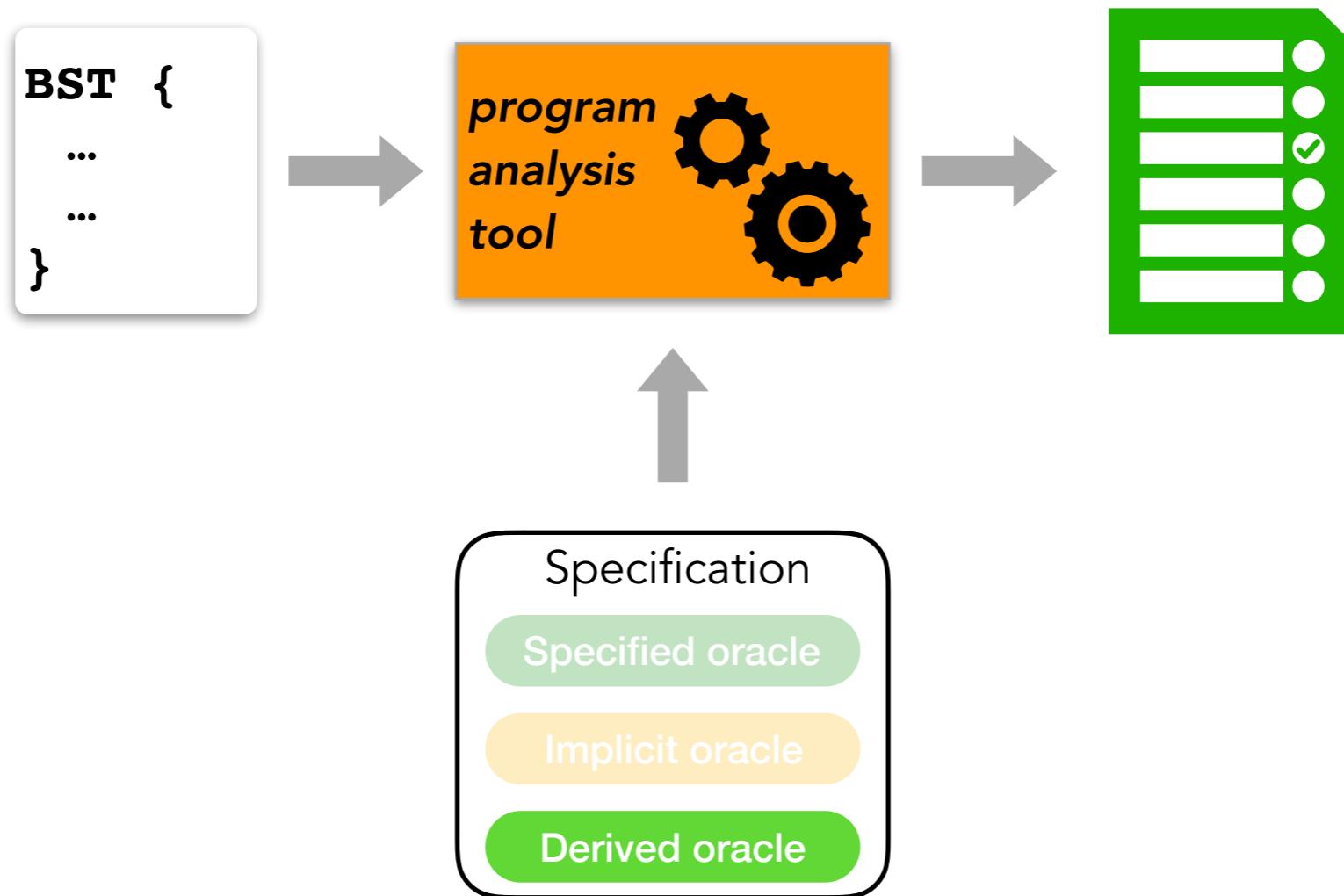
Approaches to the Oracle Problem



Approaches to the Oracle Problem



Approaches to the Oracle Problem



Invariants as Oracles

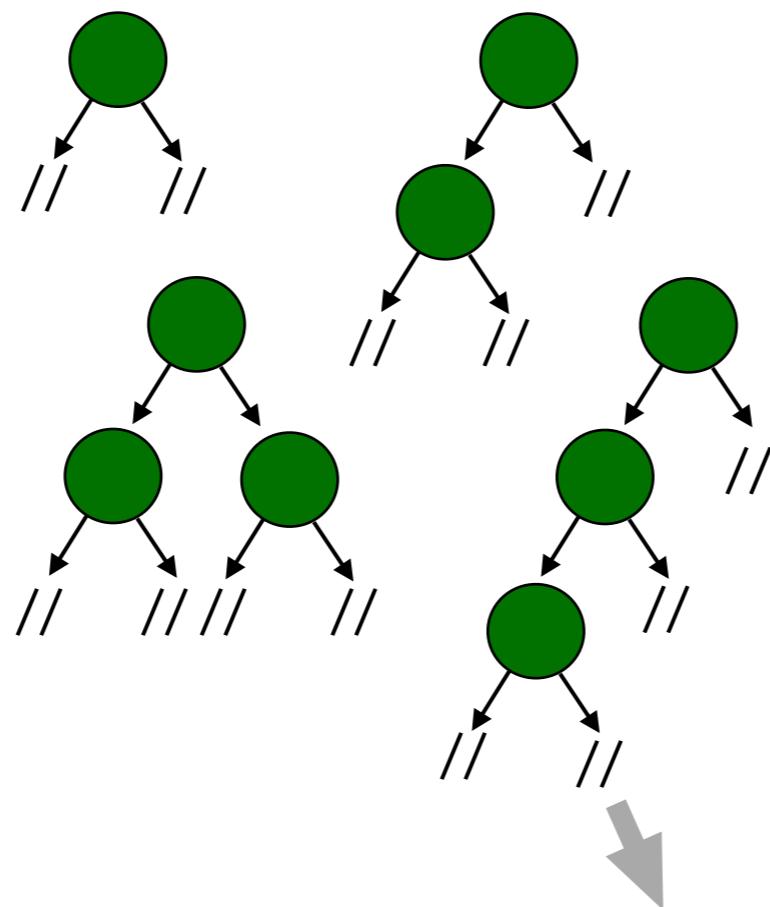
```
public class BST {  
  
    public void BST() {  
    }  
  
    public void  
    insert(...) {  
        ...  
    }  
  
    public void  
    remove() {  
        ...  
    }  
    ...  
}
```

```
/*@  
 @ invariant (\forall Node n;  
 @ \reach(root, Node, left + right).has(n) implies  
 @ ! \reach(n.right, Node, right + left).has(n) &&  
 @ ! \reach(n.left, Node, left + right).has(n));  
 @*/
```

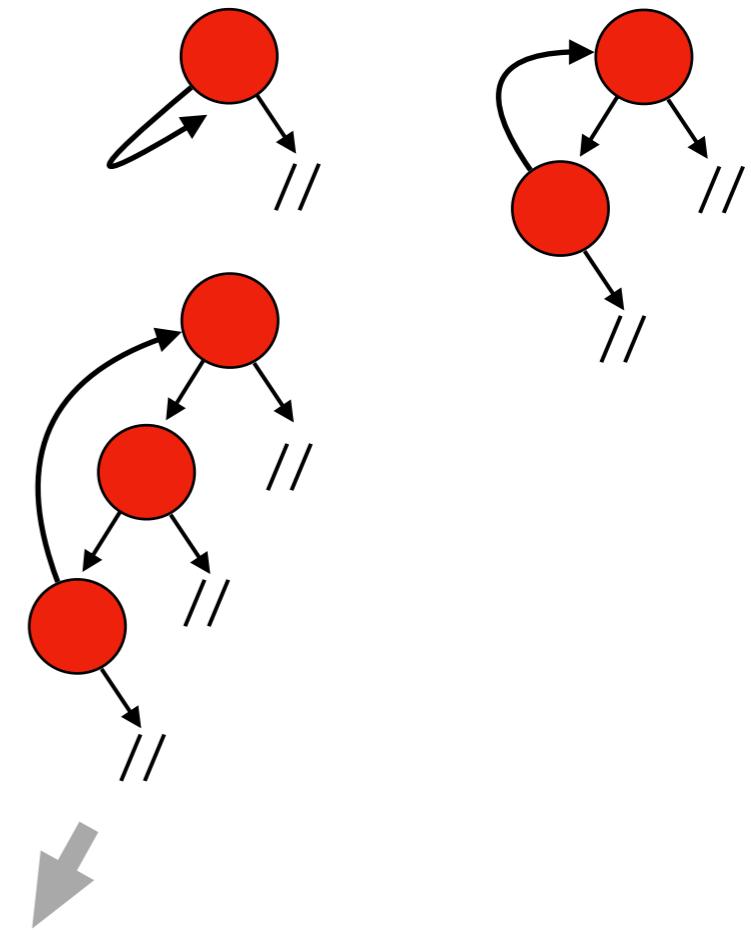
Invariants as Oracles

```
public class BST {  
    public void BST() {}  
    public void insert(...) {}  
    ...  
    public void remove() {}  
    ...  
}
```

correct program states



incorrect program states



```
/*@  
 @ invariant (\forall Node n;  
 @ \reach(root, Node, left + right).has(n) implies  
 @ ! \reach(n.right, Node, right + left).has(n) &&  
 @ ! \reach(n.left, Node, left + right).has(n));  
 @*/
```

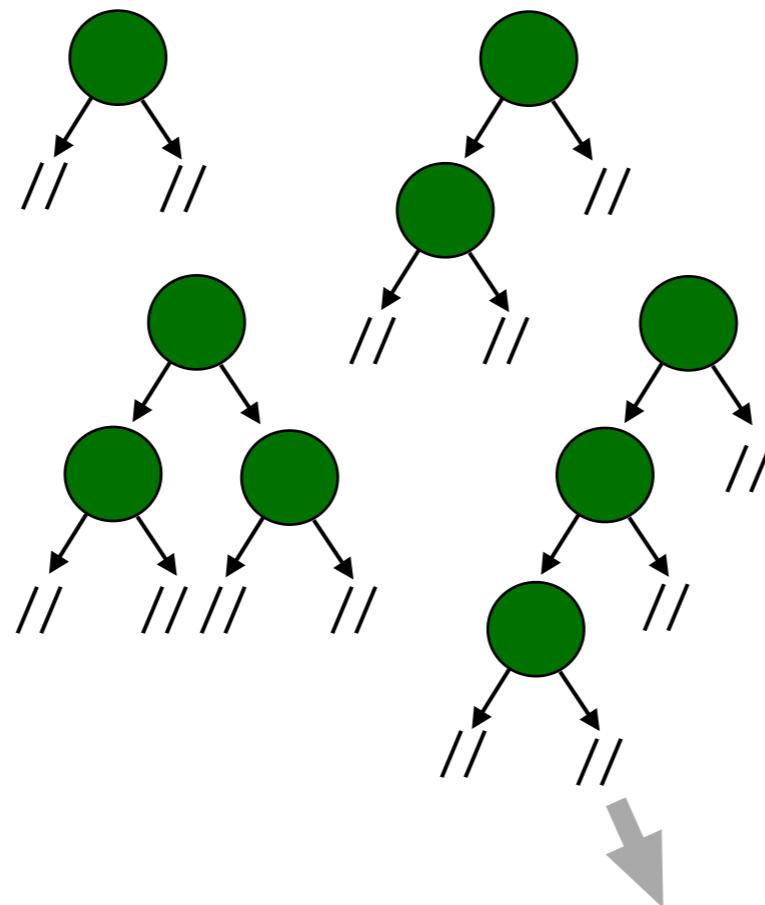
✓ return true

✗ return false

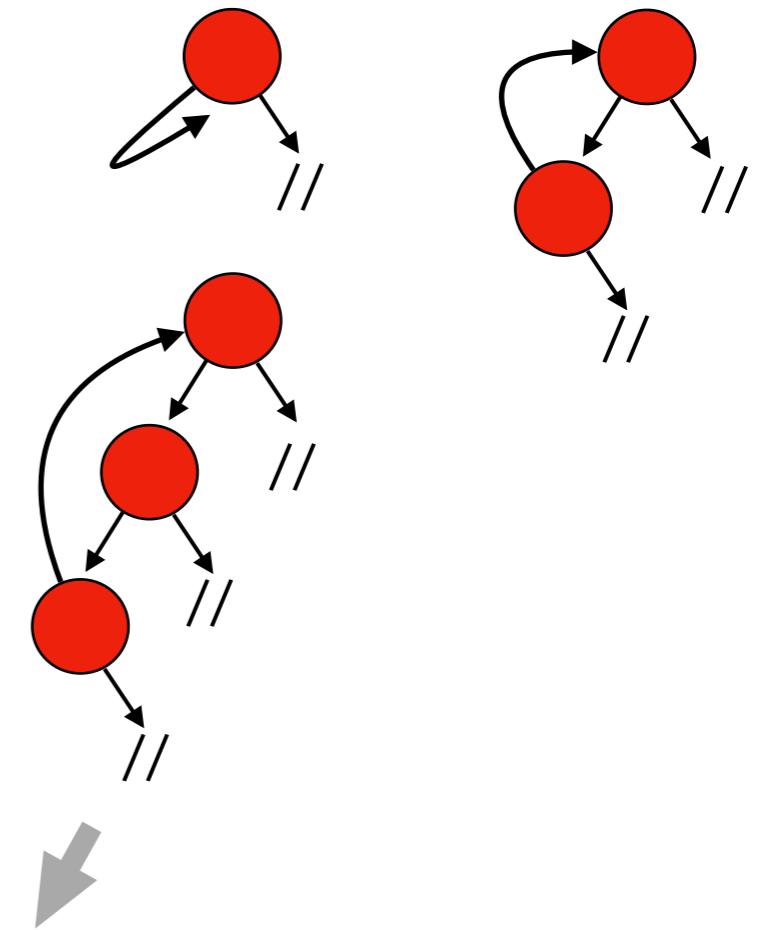
Invariants as Oracles

```
public class BST {  
    public void BST() {}  
    public void insert(...) {}  
    ...  
    public void remove() {}  
    ...  
}
```

correct program states



incorrect program states



✓ return true

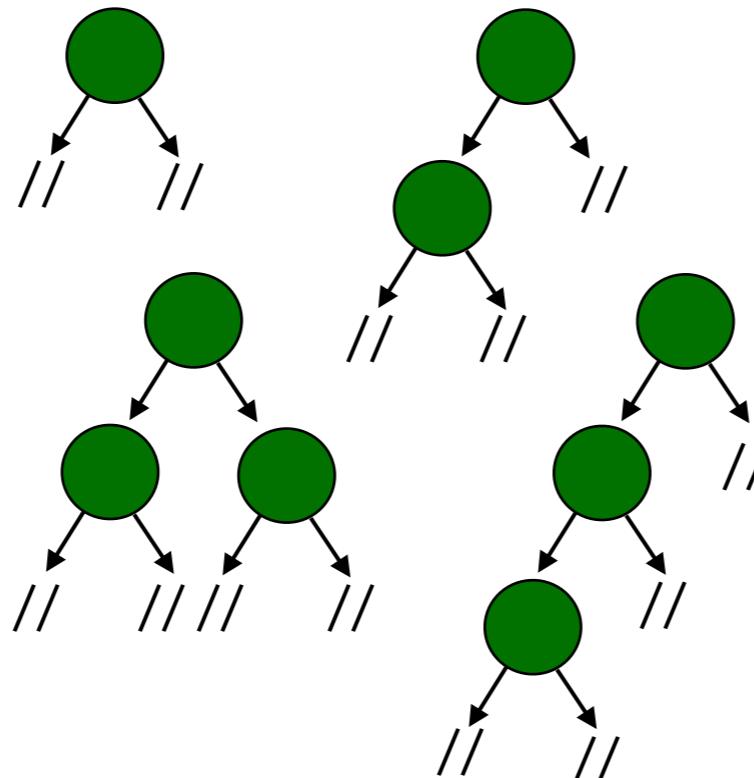


✗ return false

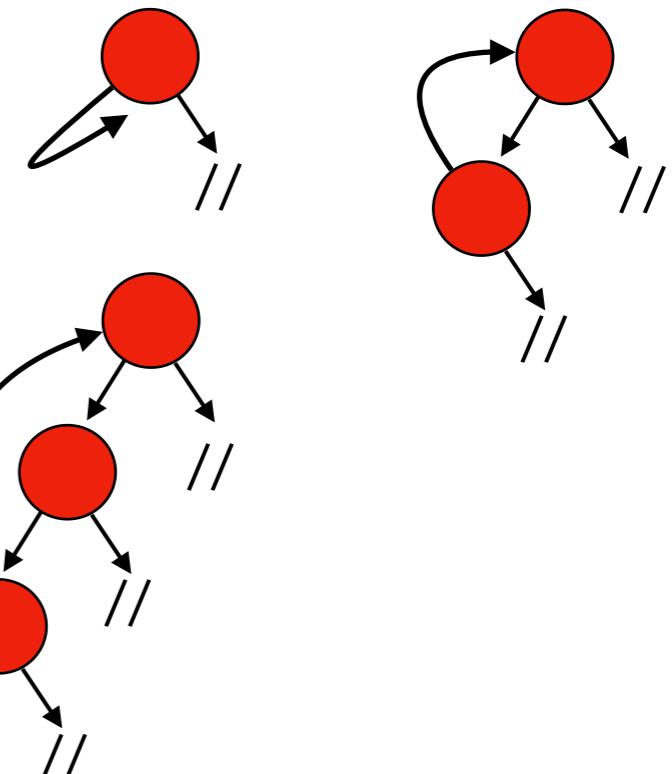
Invariants as Oracles

```
public class BST {  
    public void BST() {}  
    public void insert(...) {}  
    ...  
    public void remove() {}  
    ...  
}
```

correct program states



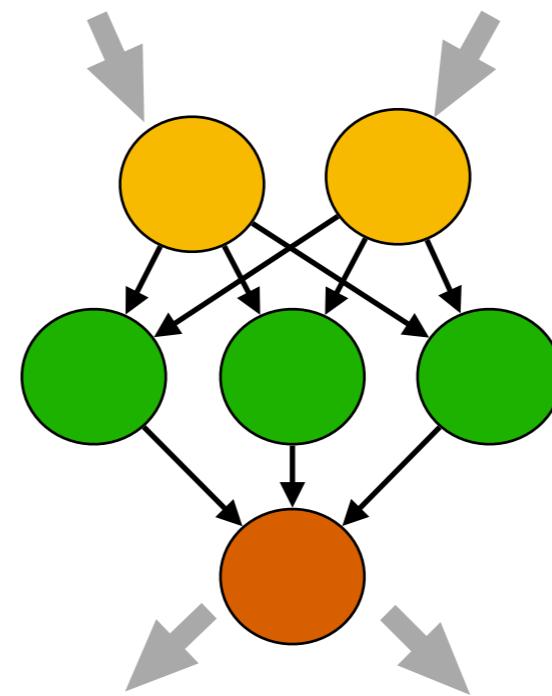
incorrect program states



Train a machine learning classifier in order to be used as a test oracle



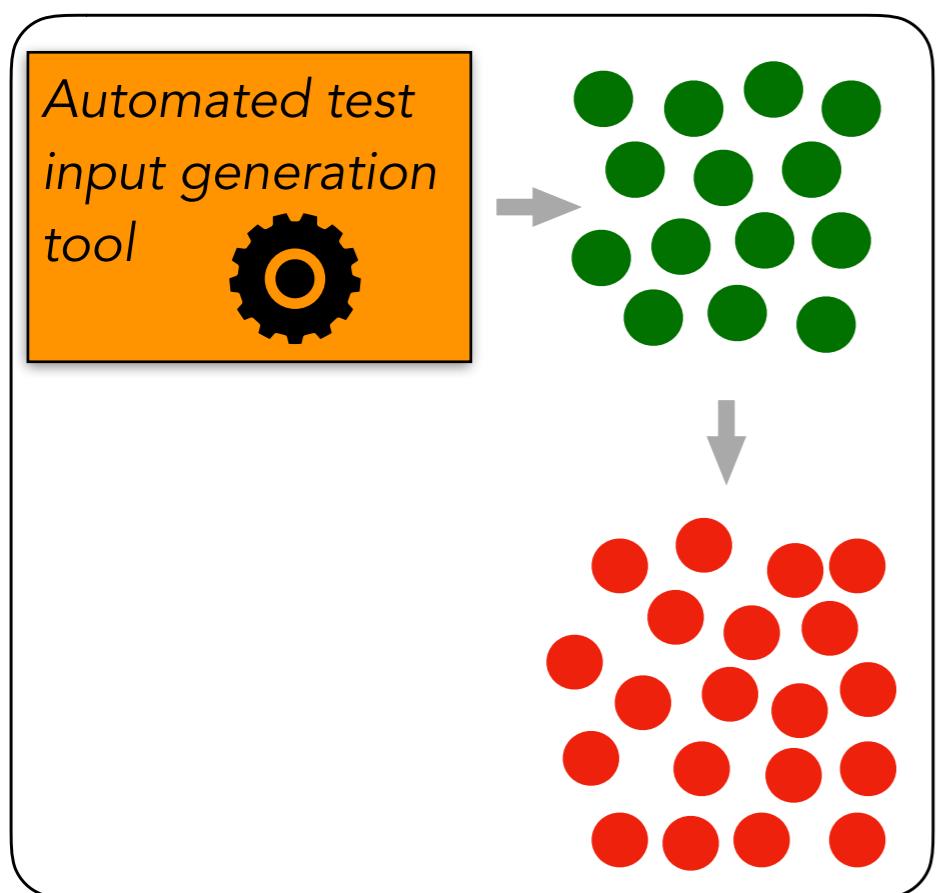
✓ return true



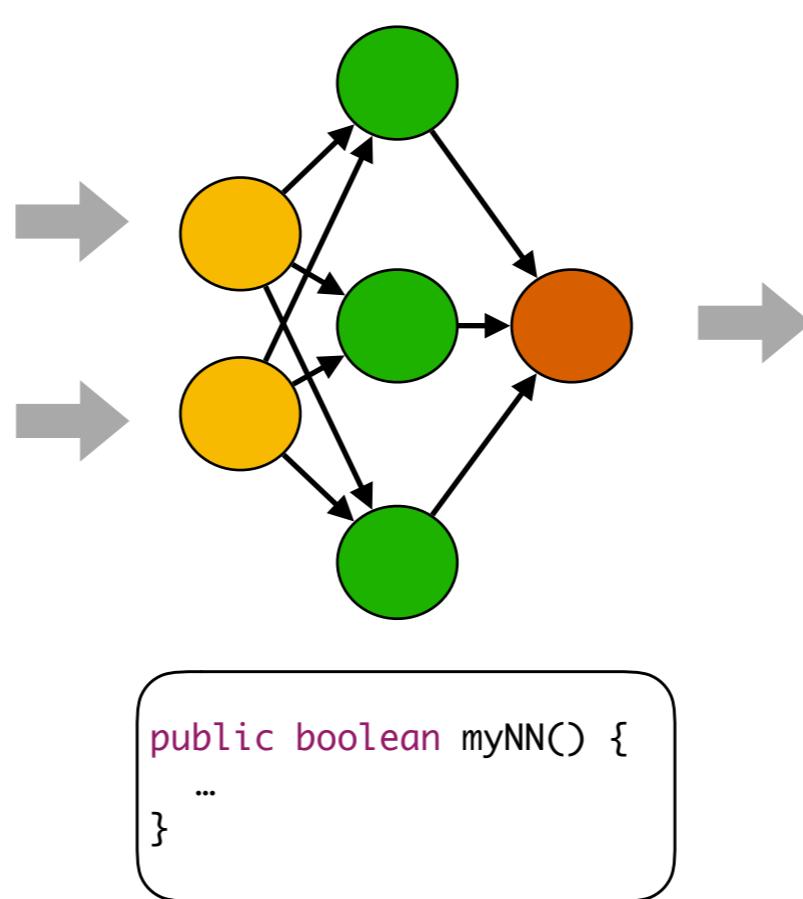
✗ return false

An Overview of the Approach

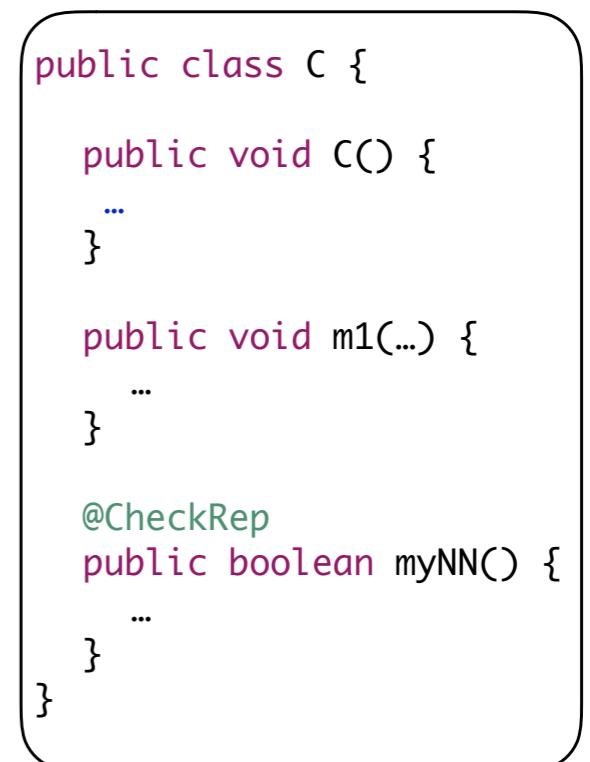
Instance Generation Mechanism



Neural Network Training



Runtime Checking



Instances Generation Mechanism

Positive instances: generated using assumed-correct building routines

```
public class C {  
    public void c() {  
        ...  
    }  
  
    public void m1() {  
        ...  
    }  
  
    public void m2() {  
        ...  
    }  
    ...  
}
```

Instances Generation Mechanism

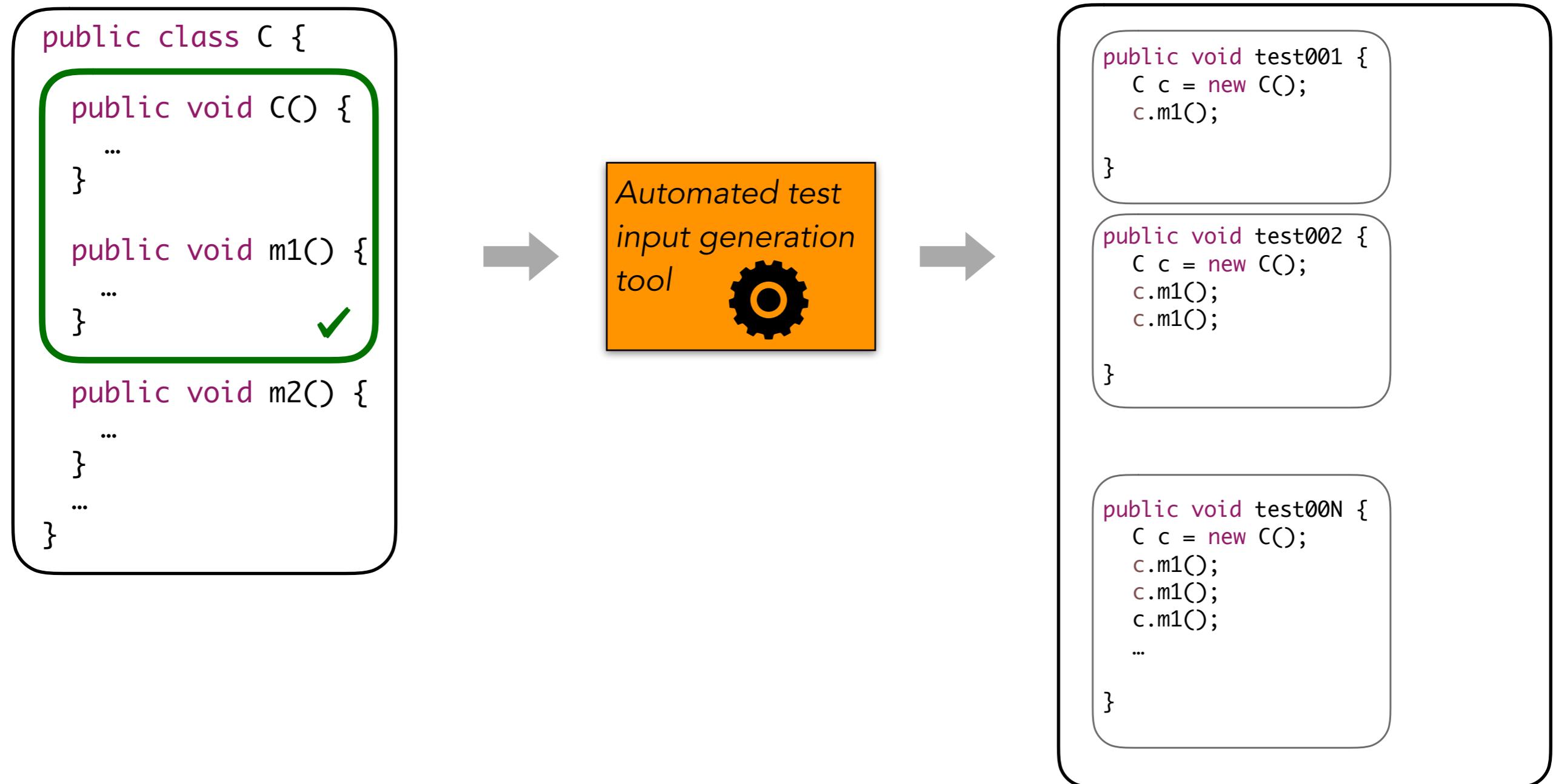
Positive instances: generated using assumed-correct building routines

Identification of assumed-correct builders

```
public class C {  
    public void c() {  
        ...  
    }  
  
    public void m1() {  
        ...  
    } ✓  
  
    public void m2() {  
        ...  
    }  
    ...  
}
```

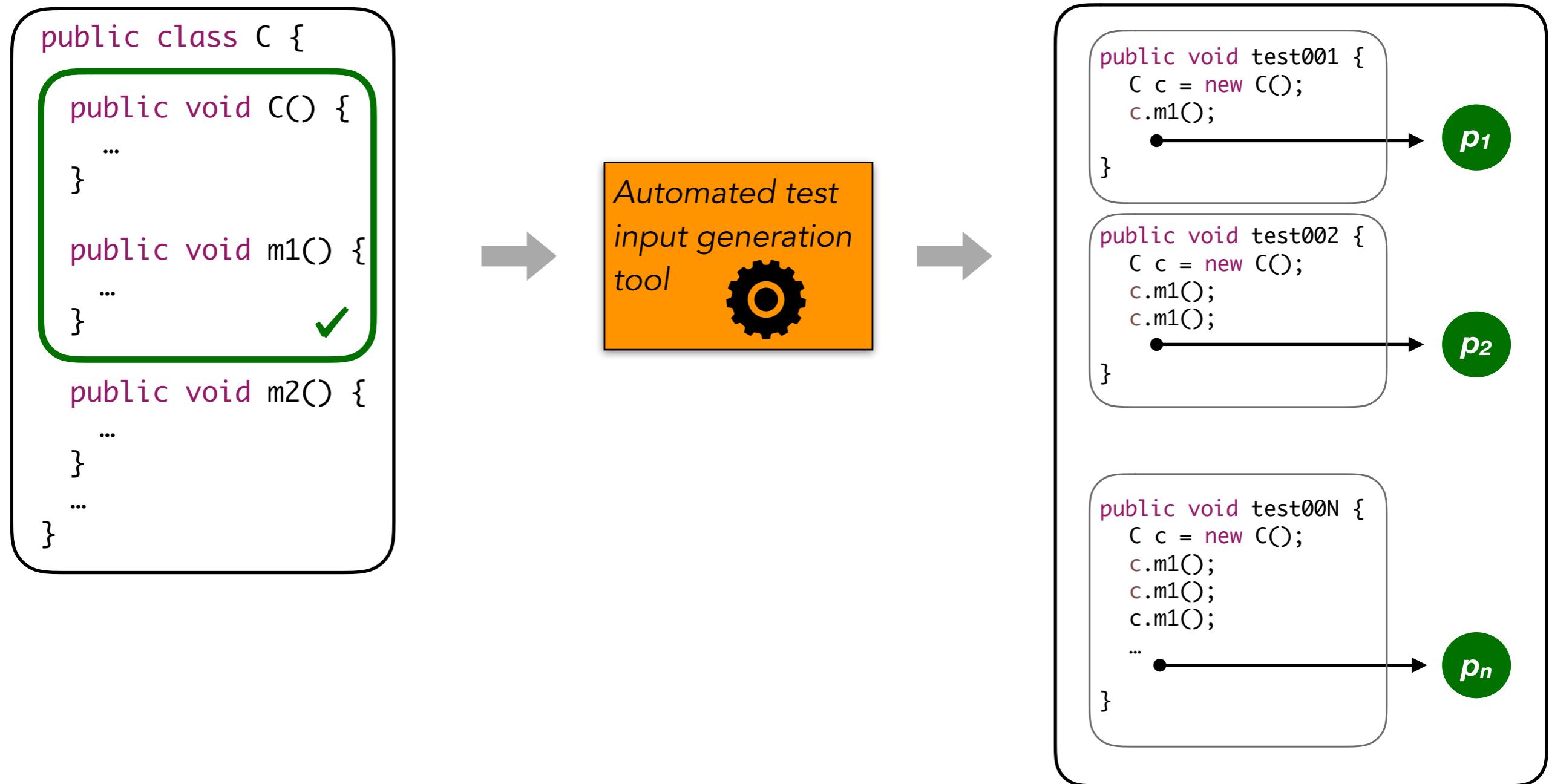
Instances Generation Mechanism

Positive instances: generated using assumed-correct building routines



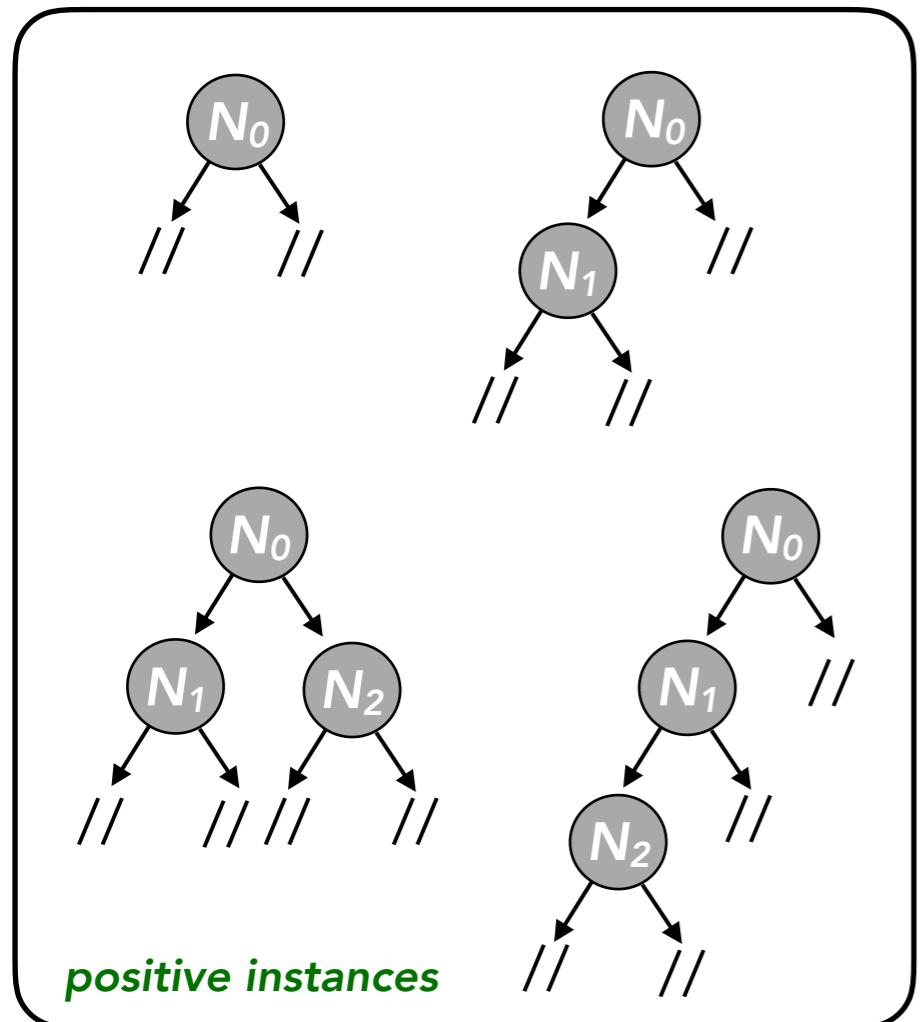
Instances Generation Mechanism

Positive instances: generated using assumed-correct building routines

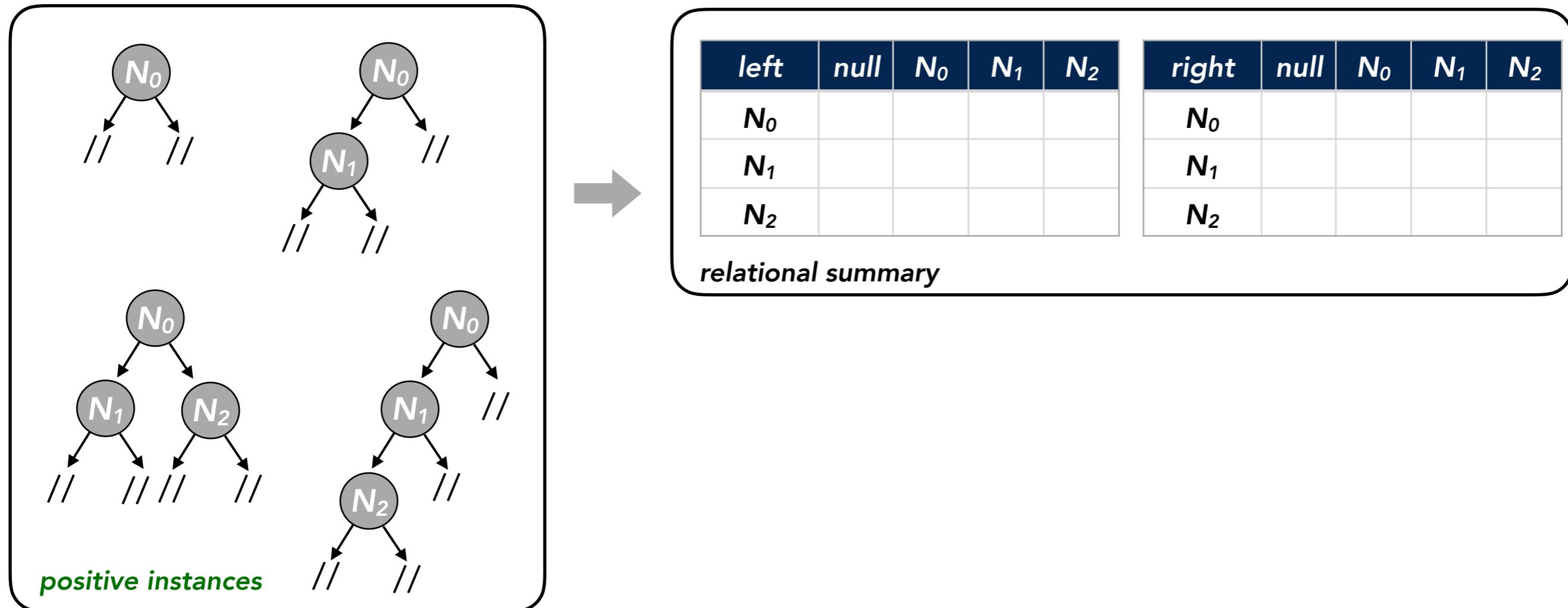


Negative instances: build a *relational summary* from valid instances, and mutate valid objects off the summary

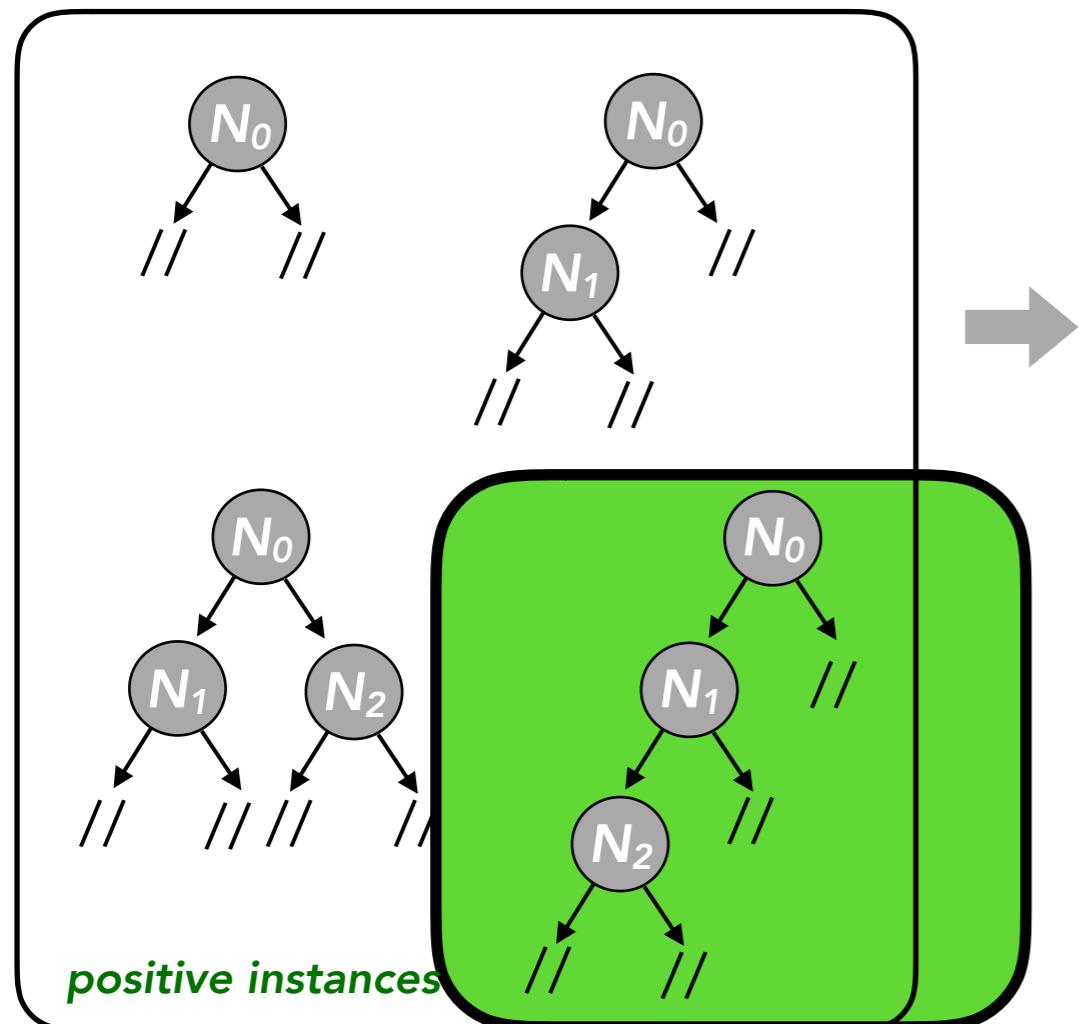
Negative instances: build a *relational summary* from valid instances, and mutate valid objects off the summary



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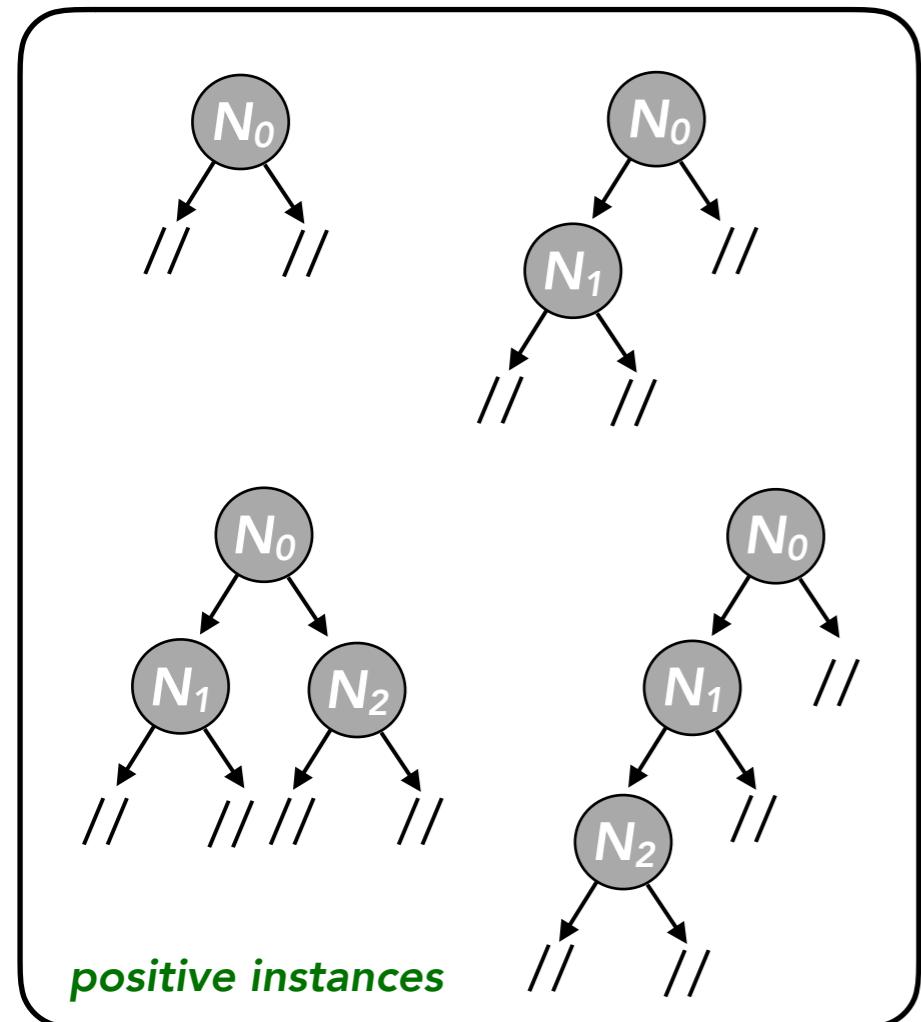
Negative instances: build a *relational summary* from valid instances, and mutate valid objects off the summary



relational summary

<i>left</i>	<i>null</i>	N_0	N_1	N_2	<i>right</i>	<i>null</i>	N_0	N_1	N_2
N_0	✓		✓		N_0	✓			✓
N_1	✓			✓	N_1	✓			
N_2	✓				N_2	✓			

Negative instances: build a *relational summary* from valid instances, and mutate valid objects off the summary

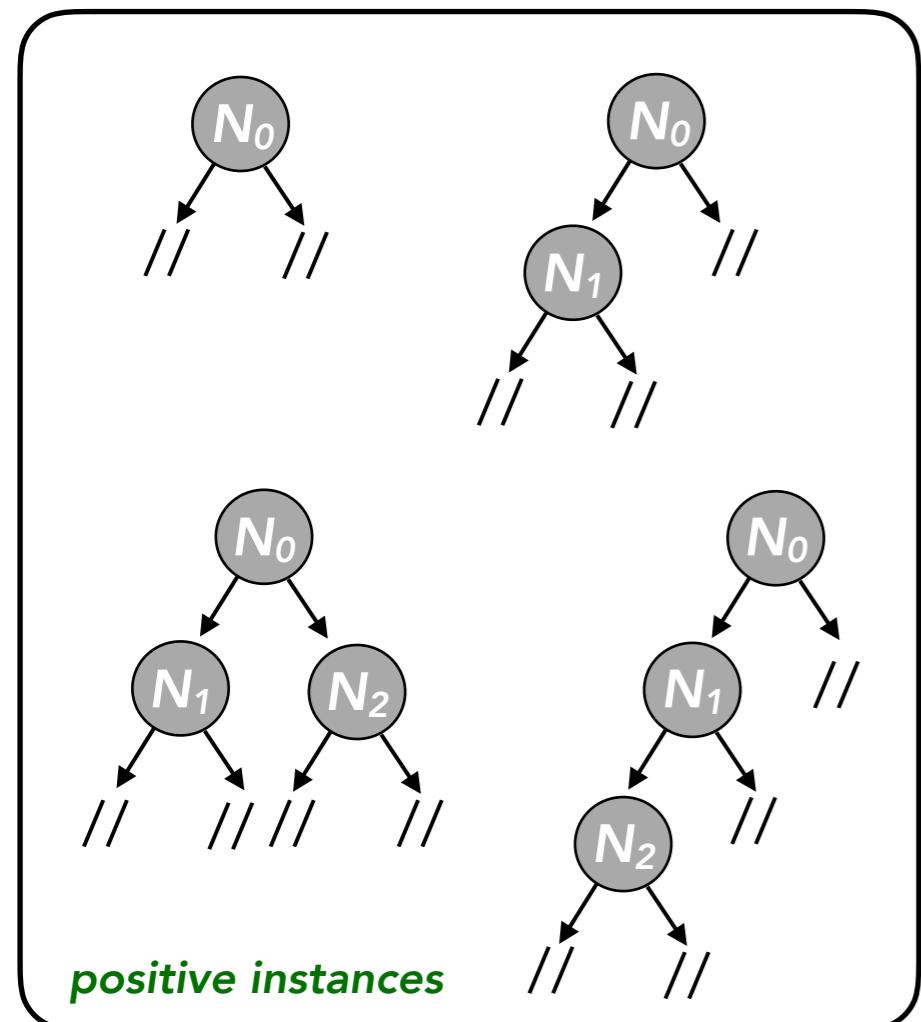


relational summary

left	null	N_0	N_1	N_2	right	null	N_0	N_1	N_2
N_0	✓		✓		N_0	✓			✓
N_1	✓			✓	N_1	✓			
N_2	✓				N_2	✓			



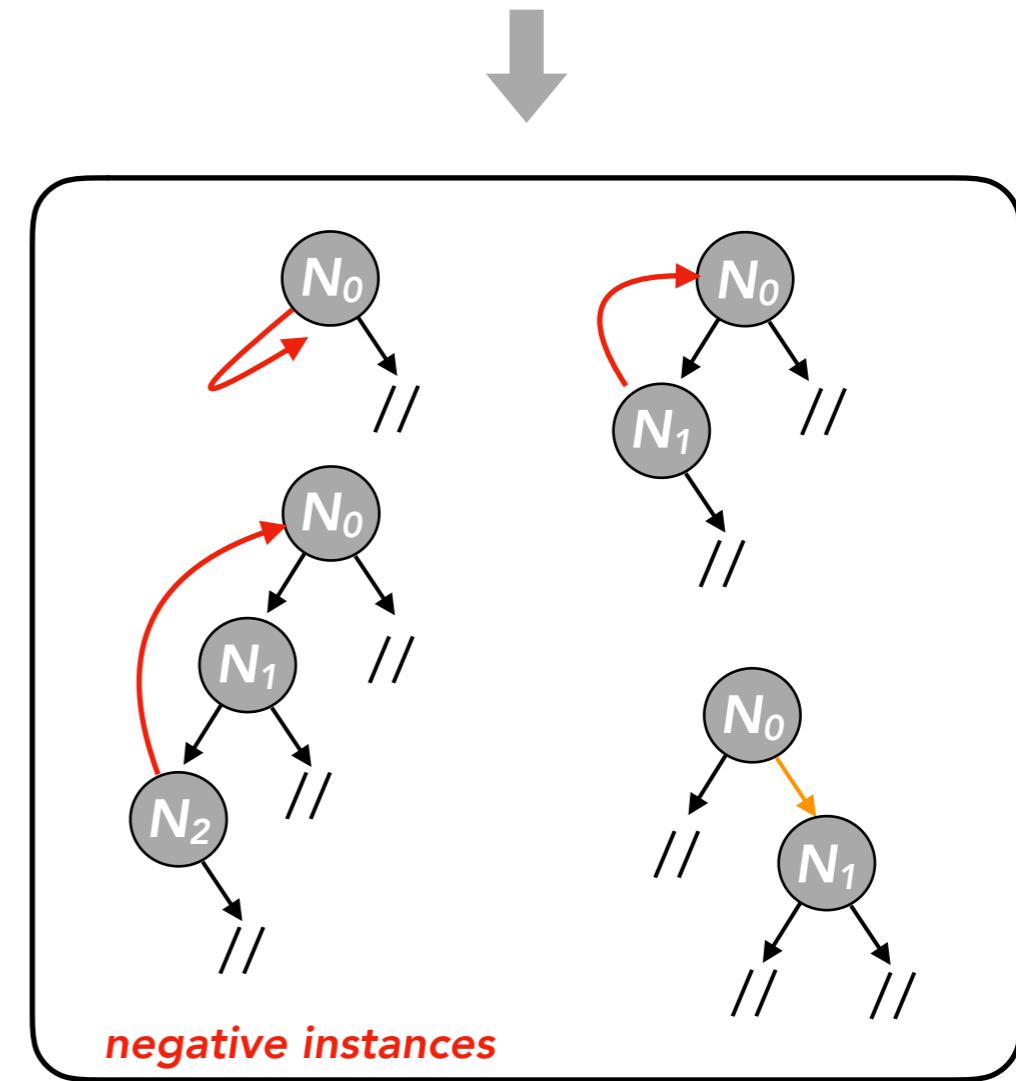
Negative instances: build a *relational summary* from valid instances, and mutate valid objects off the summary



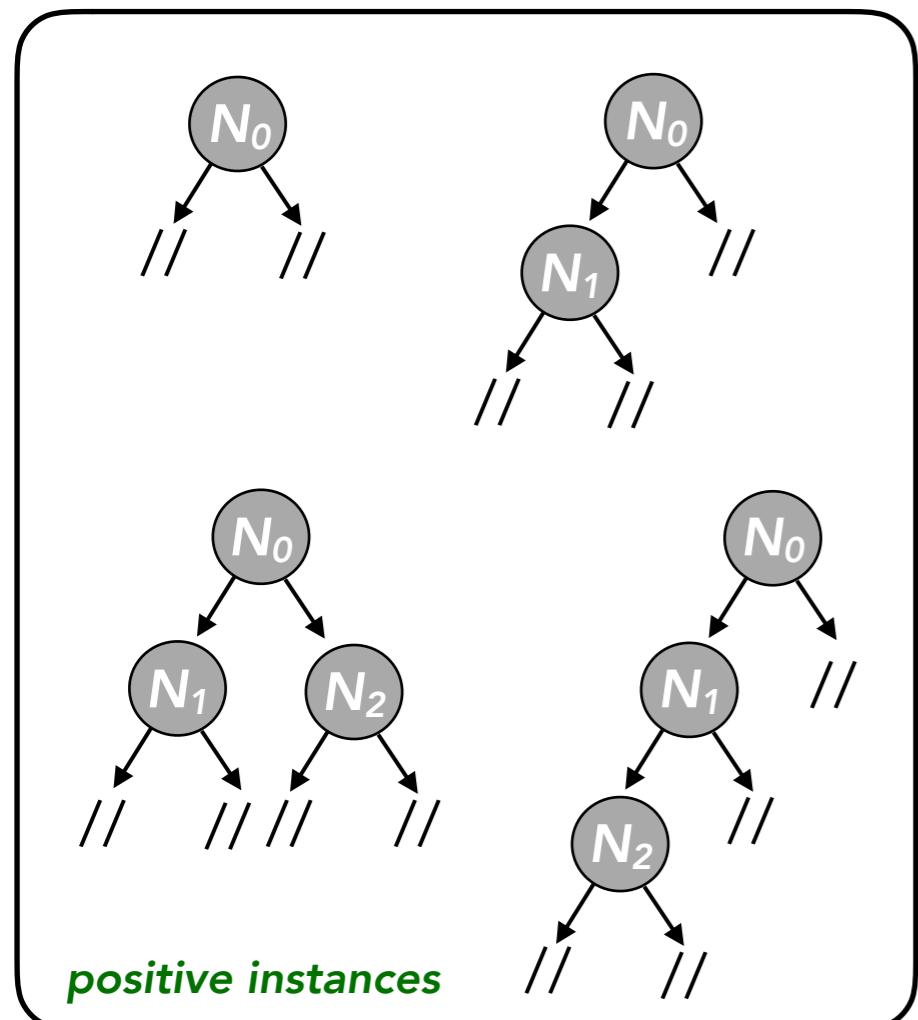
→

relational summary

left	null	N_0	N_1	N_2	right	null	N_0	N_1	N_2
N_0	✓	X	✓		N_0	✓		X	✓
N_1	✓	X		✓	N_1	✓			
N_2	✓	X			N_2	✓			

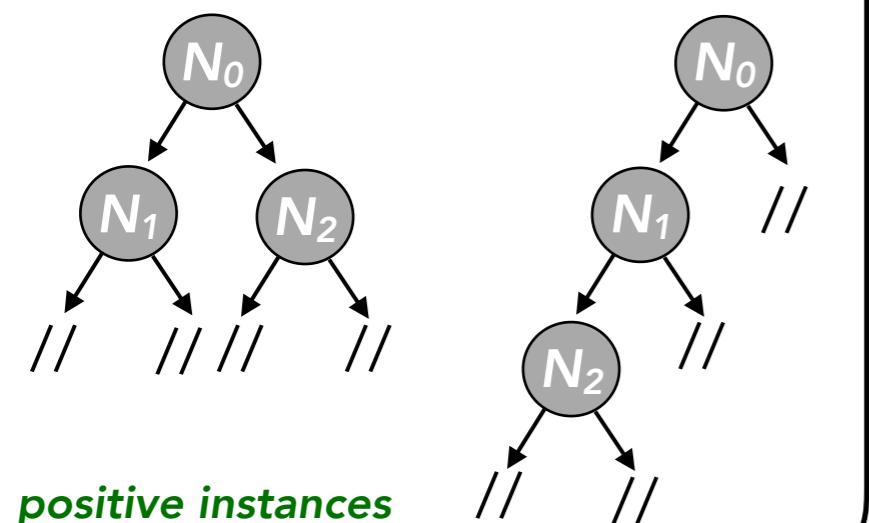


Negative instances: build a *relational summary* from valid instances, and mutate valid objects off the summary

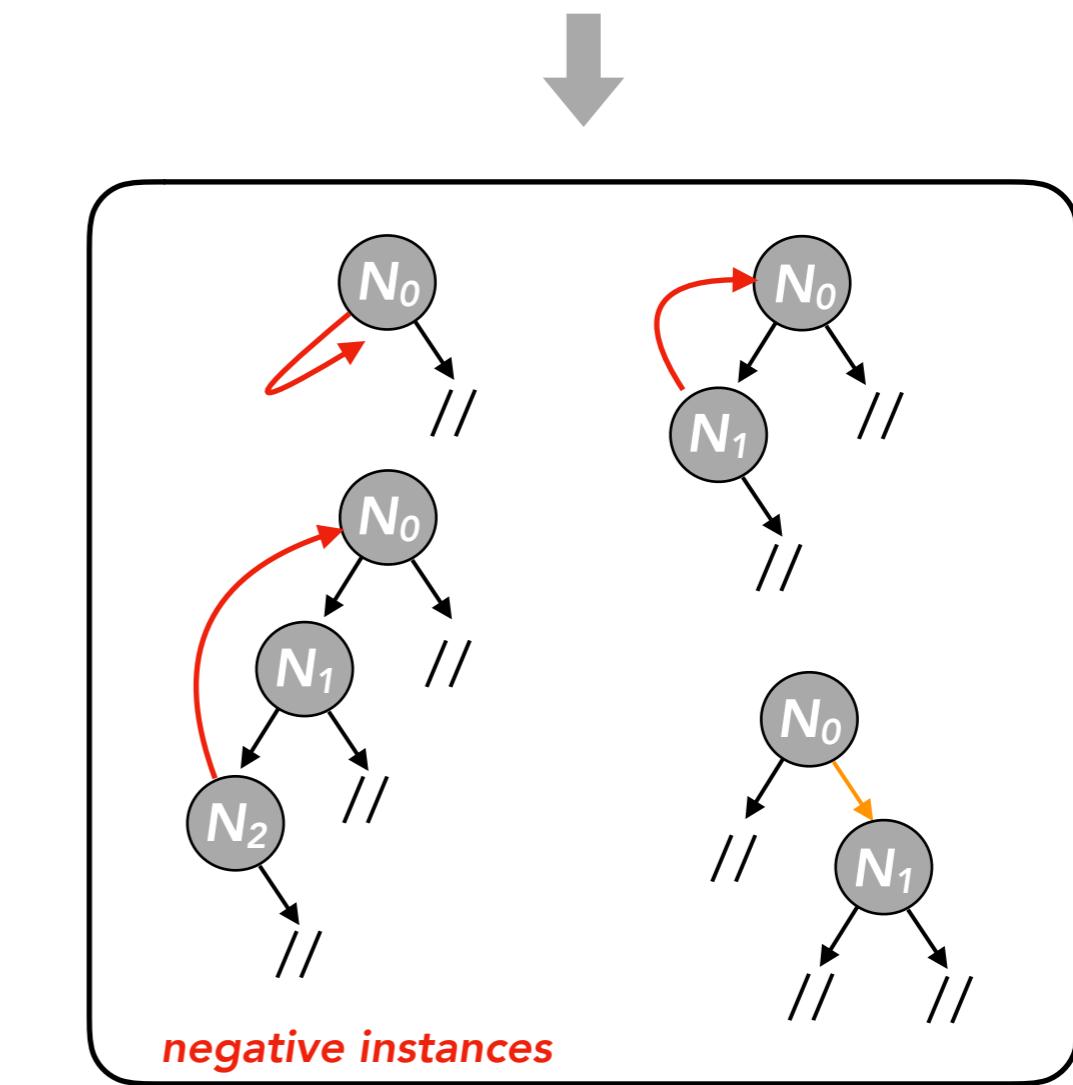


left	null	N_0	N_1	N_2	right	null	N_0	N_1	N_2
N_0	✓	X	✓		N_0	✓		X	✓
N_1	✓	X		✓	N_1	✓			
N_2	✓	X			N_2	✓			

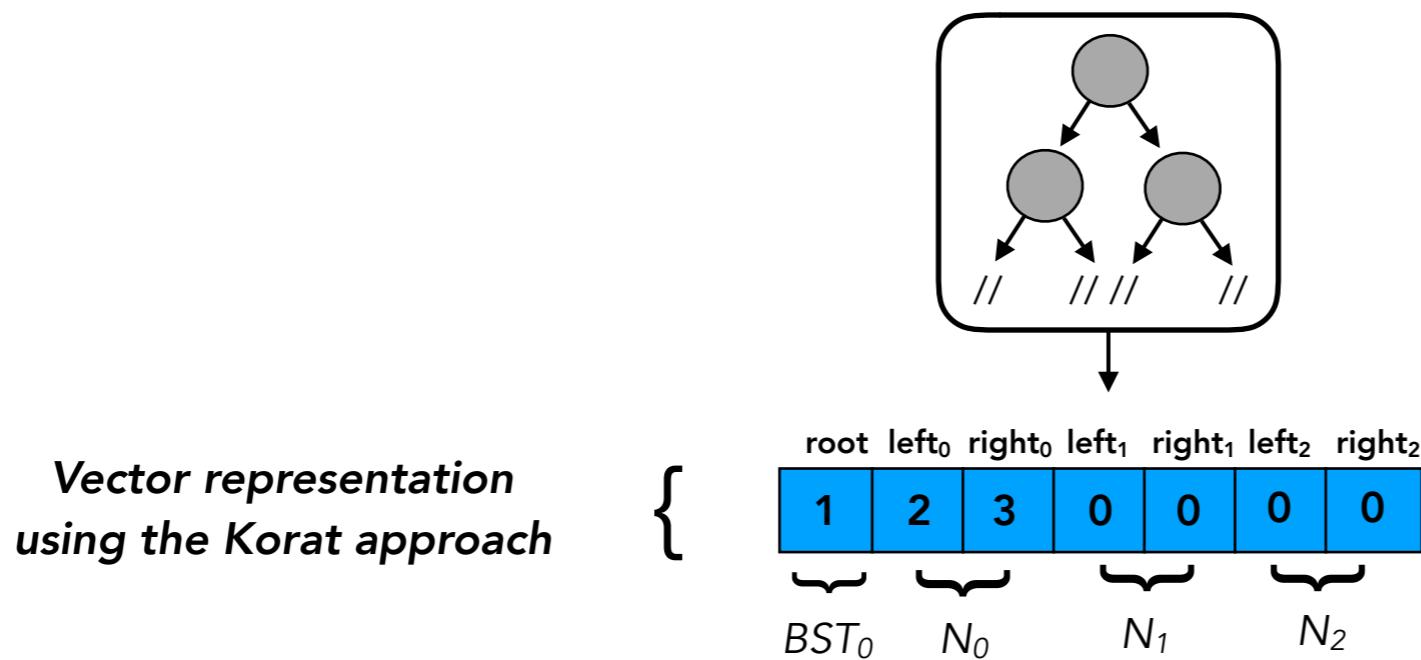
relational summary



False negative instances
may be generated

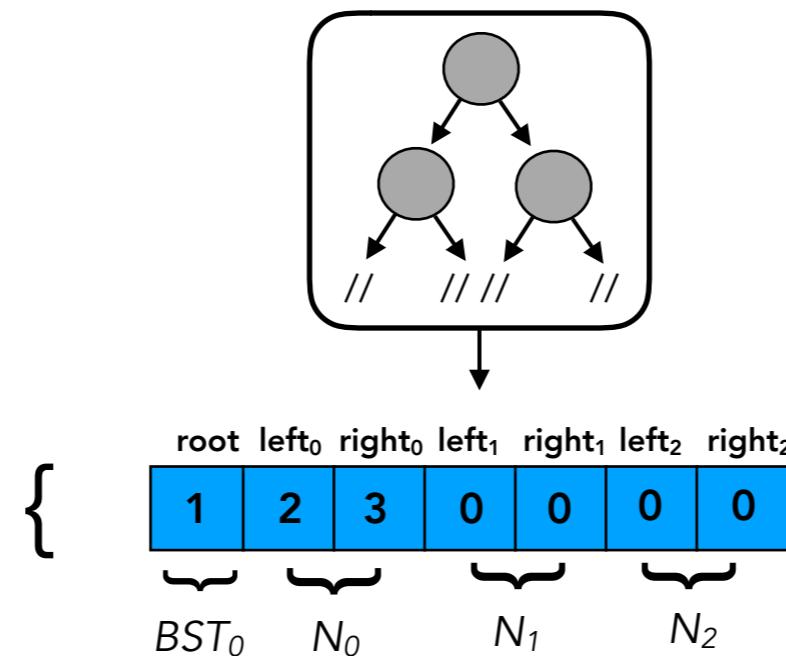


Vector Representation and Network Architecture



Vector Representation and Network Architecture

**Vector representation
using the Korat approach**

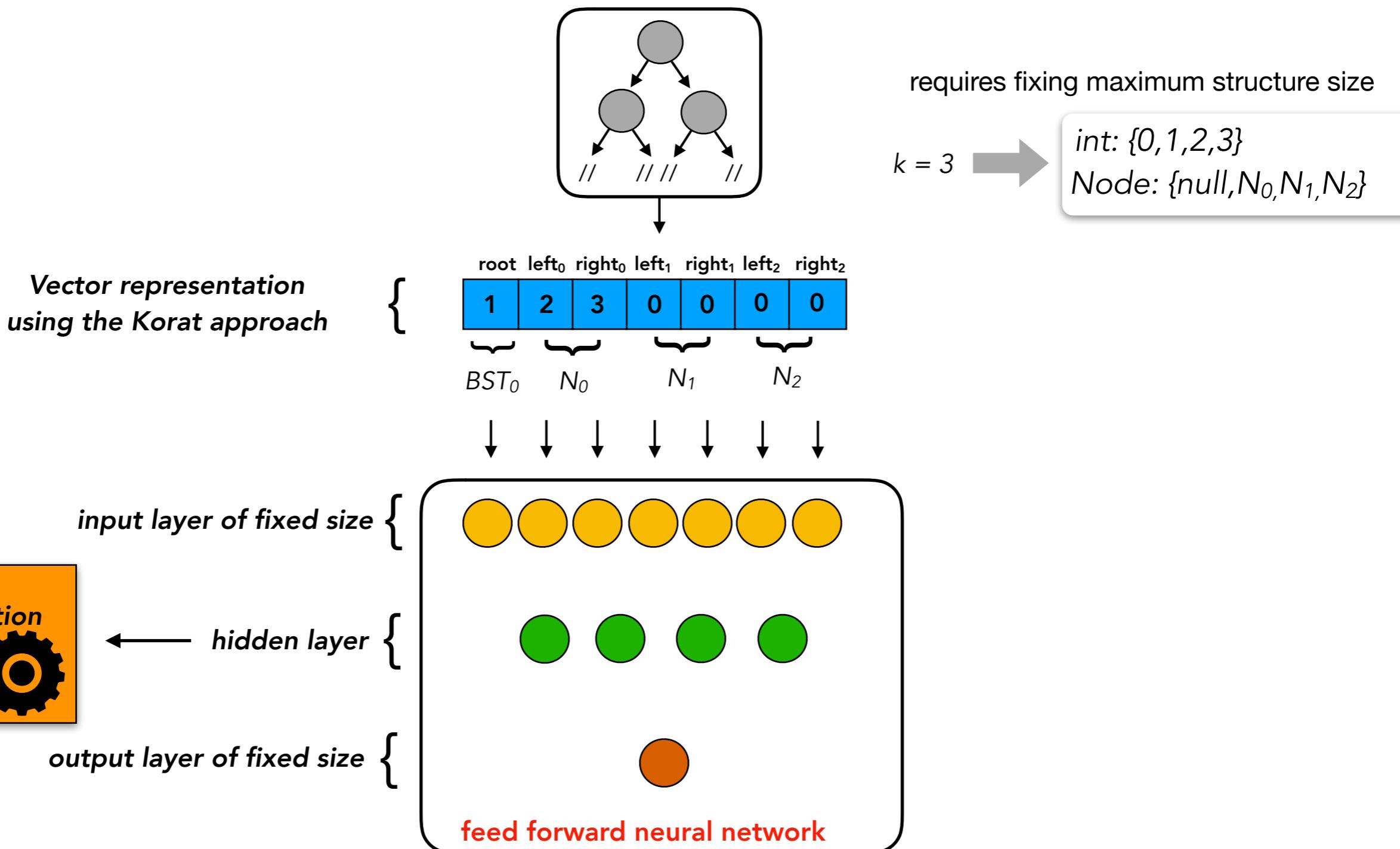


requires fixing maximum structure size

$k = 3$

int: {0,1,2,3}
Node: {null,N₀,N₁,N₂}

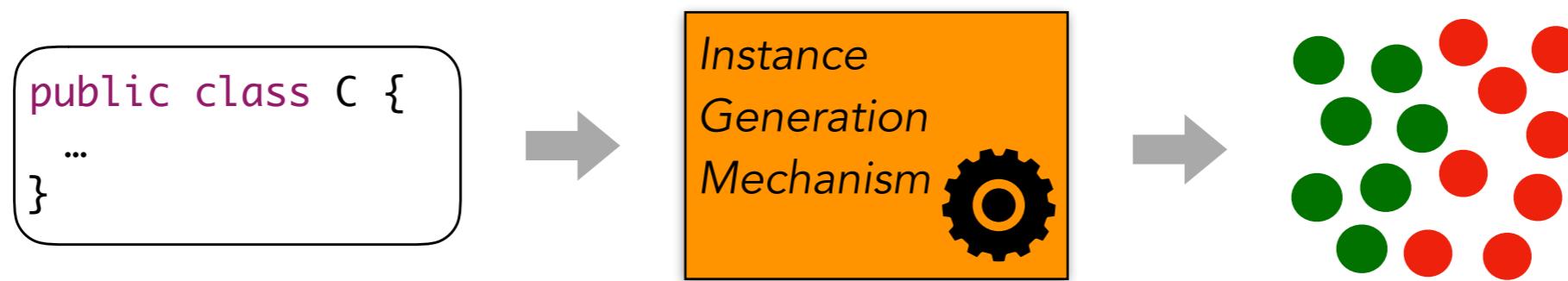
Vector Representation and Network Architecture



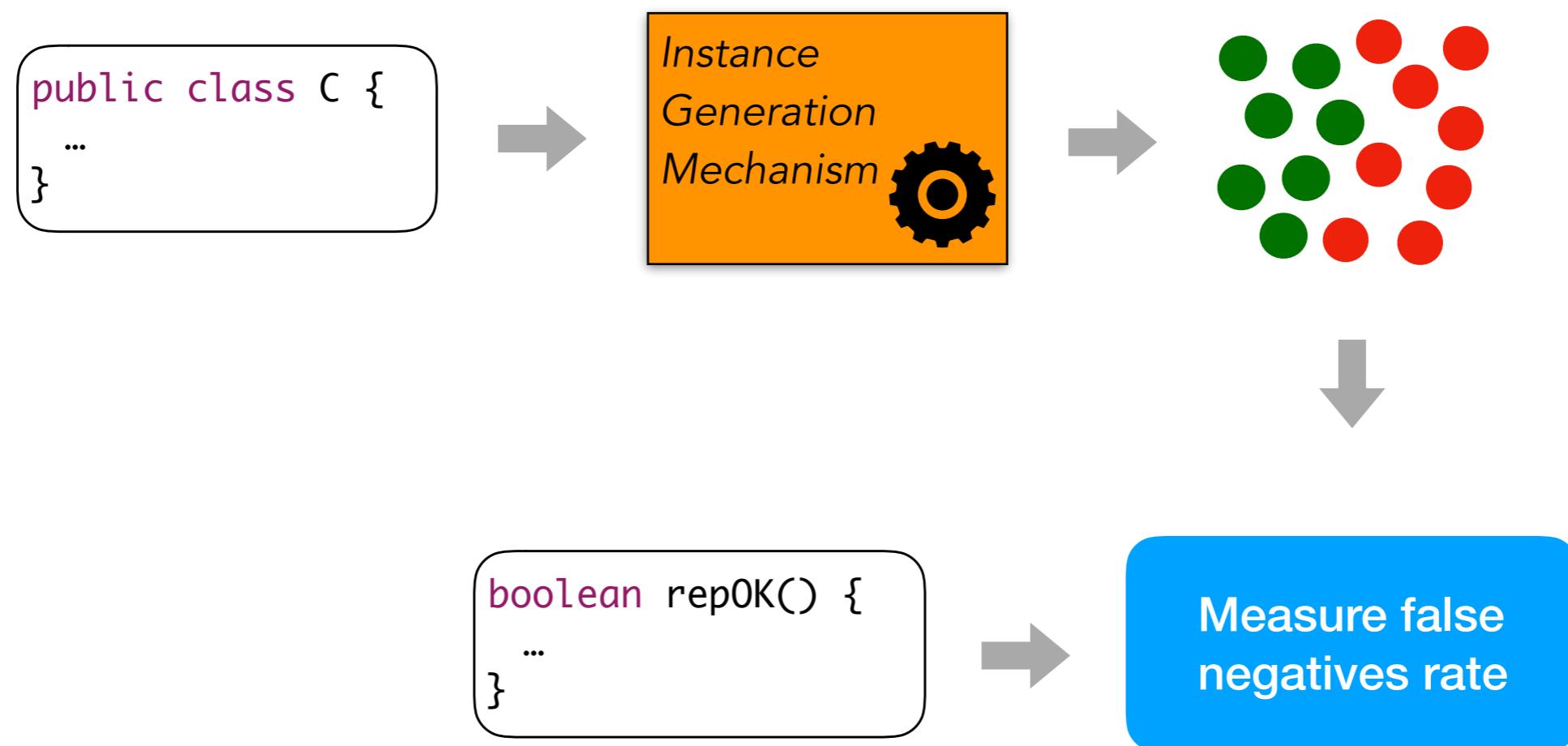
Evaluation

- *RQ1: is the mechanism for negative instance generation effective?*
- *RQ2: are neural networks precise in classifying valid/invalid data structure objects?*
- *RQ3: can automated analysis be improved by the use of NN invariants?*

RQ1: Measure of False Negatives Rate



RQ1: Measure of False Negatives Rate



RQ1: Measure of False Negatives Rate

Singly Sorted List

bound	negative instances	false negatives rate
3	75	6,67 %
4	455	4,18 %
5	2418	2,81 %
6	10306	2,32 %
7	33949	2,24 %
8	83131	2,36 %

Red Black Tree

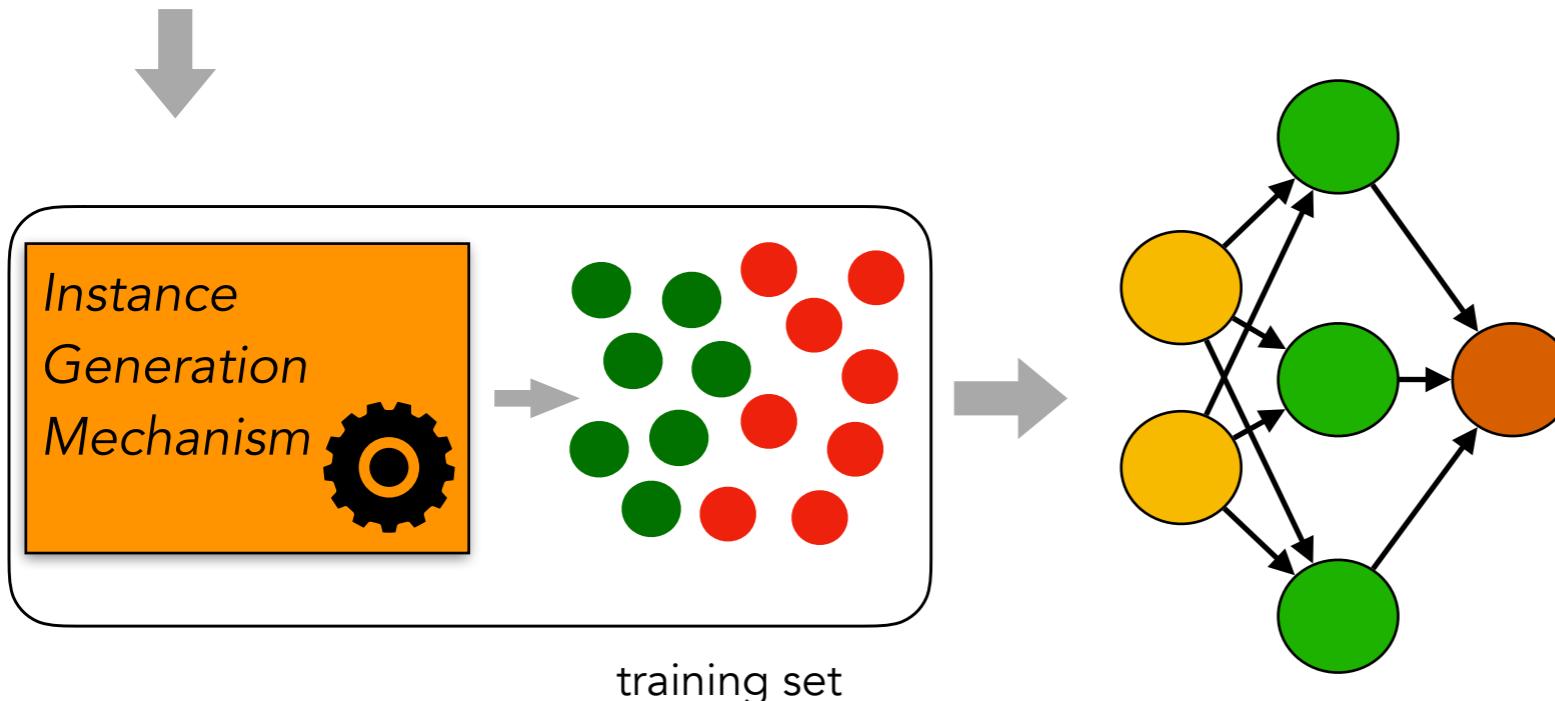
bound	negative instances	false negatives rate
3	144	5,56 %
4	604	3,31 %
5	2377	2,23 %
6	8535	1,53 %
7	26894	1,16 %
8	72099	0,96 %

Binary Search Tree

bound	negative instances	false negatives rate
3	200	4,50 %
4	1098	3,19 %
5	5089	0,26 %
6	18445	1,98 %
7	50808	1,65 %
8	106663	1,59 %

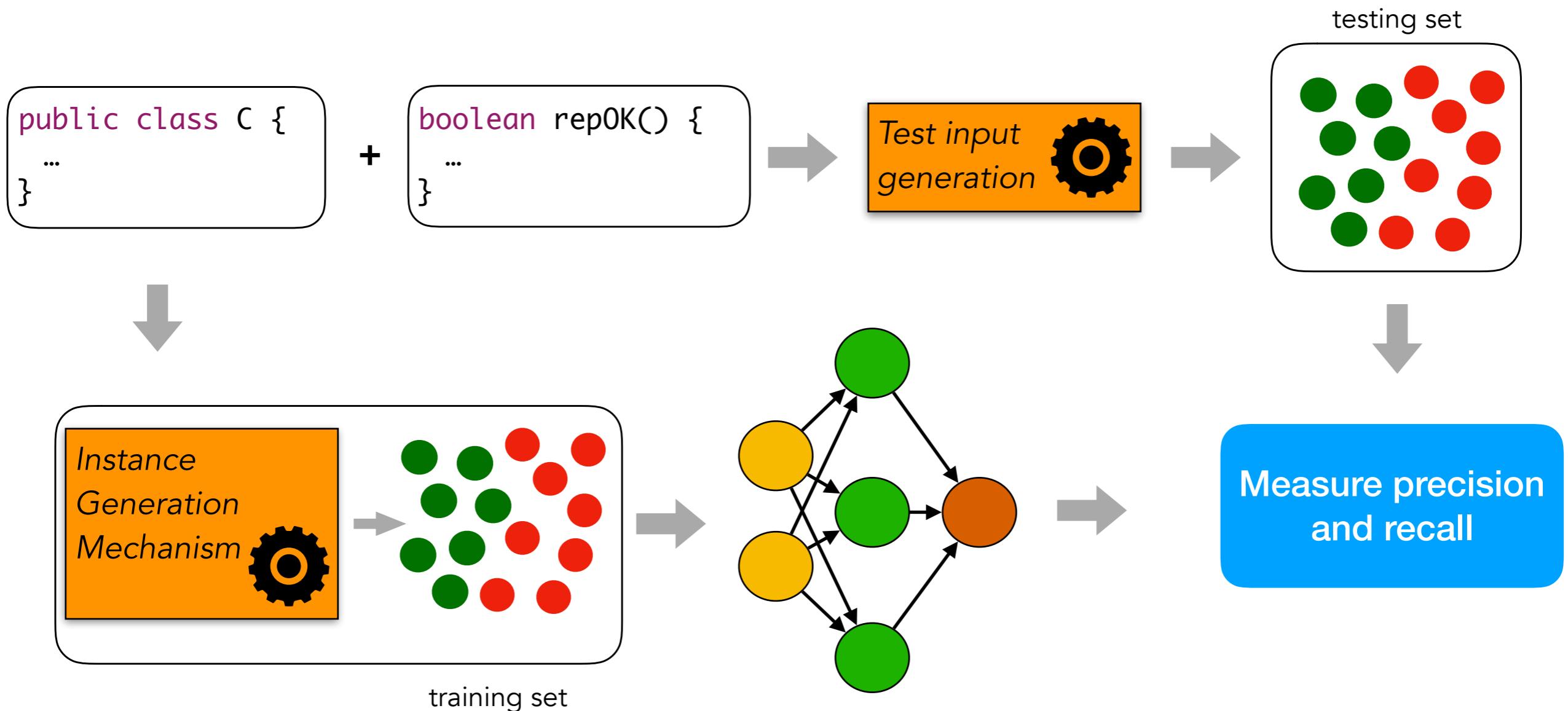
RQ2: Neural Networks Precision and Recall Classifying Data Structure Objects

```
public class C {  
    ...  
}
```



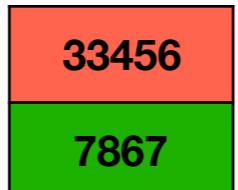
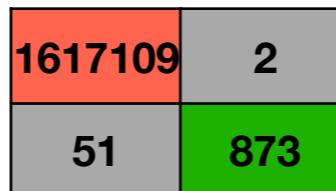
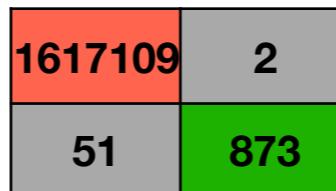
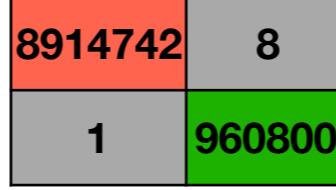
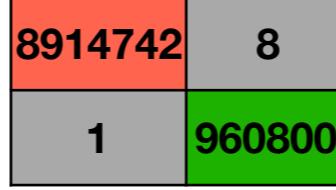
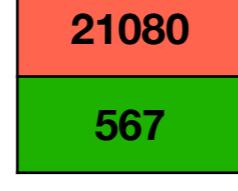
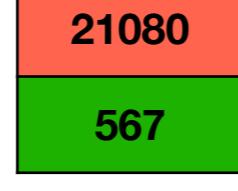
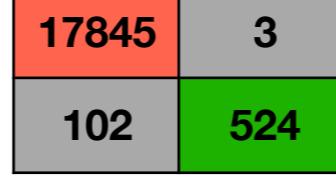
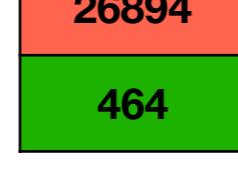
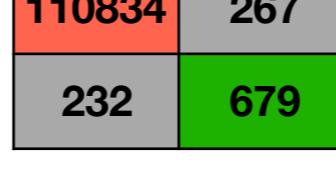
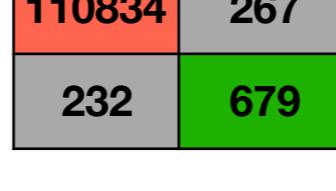
RQ2: Neural Networks Precision and Recall

Classifying Data Structure Objects



RQ2: Neural Networks Precision and Recall

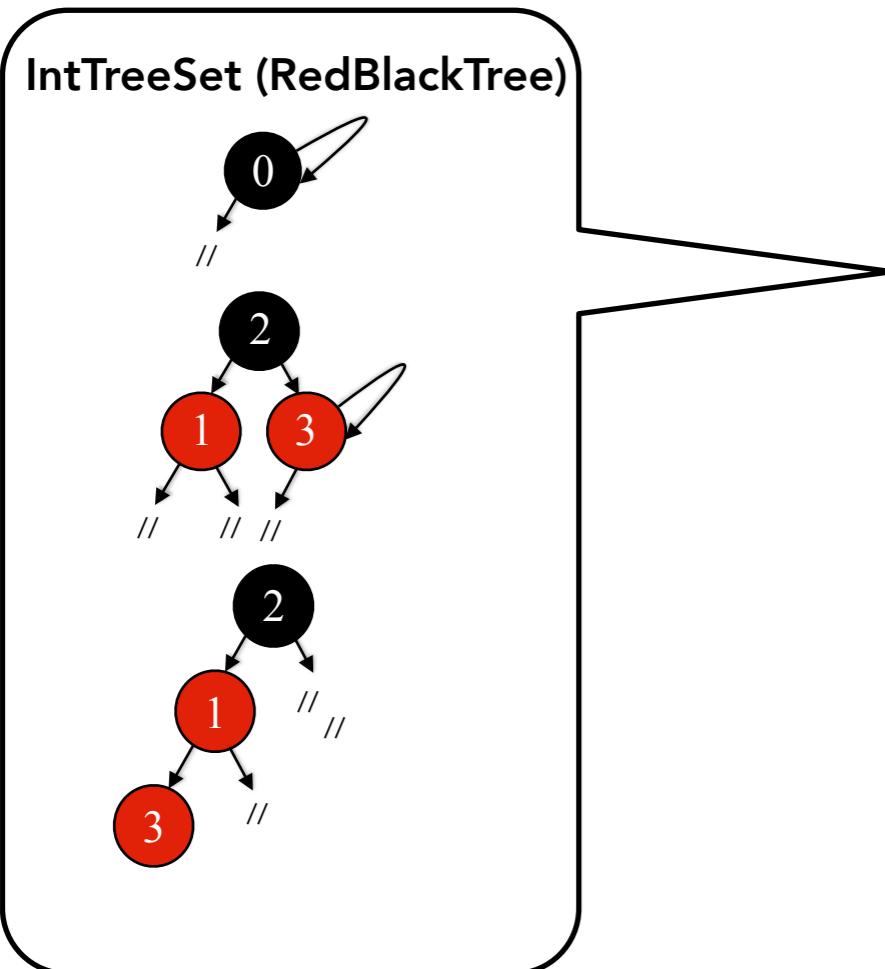
Classifying Data Structure Objects

	Training Set	Validation Set	Precision	Recall
		X ✓		
SinglyLinkedList - 7			99%	99%
			99%	99%
SinglySortedList - 7			99%	99%
			99%	94%
DoublyLinkedList - 7			99%	99%
			99%	99%
BinaryTree - 7			99%	99%
			99%	83%
RedBlackTree - 7			99%	99%
			71%	74%

RQ3: Improved Bug Finding?

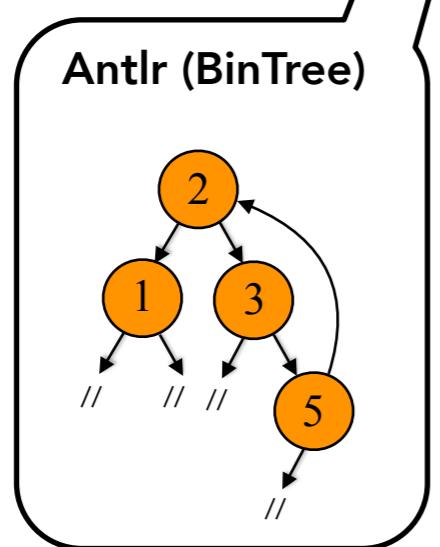
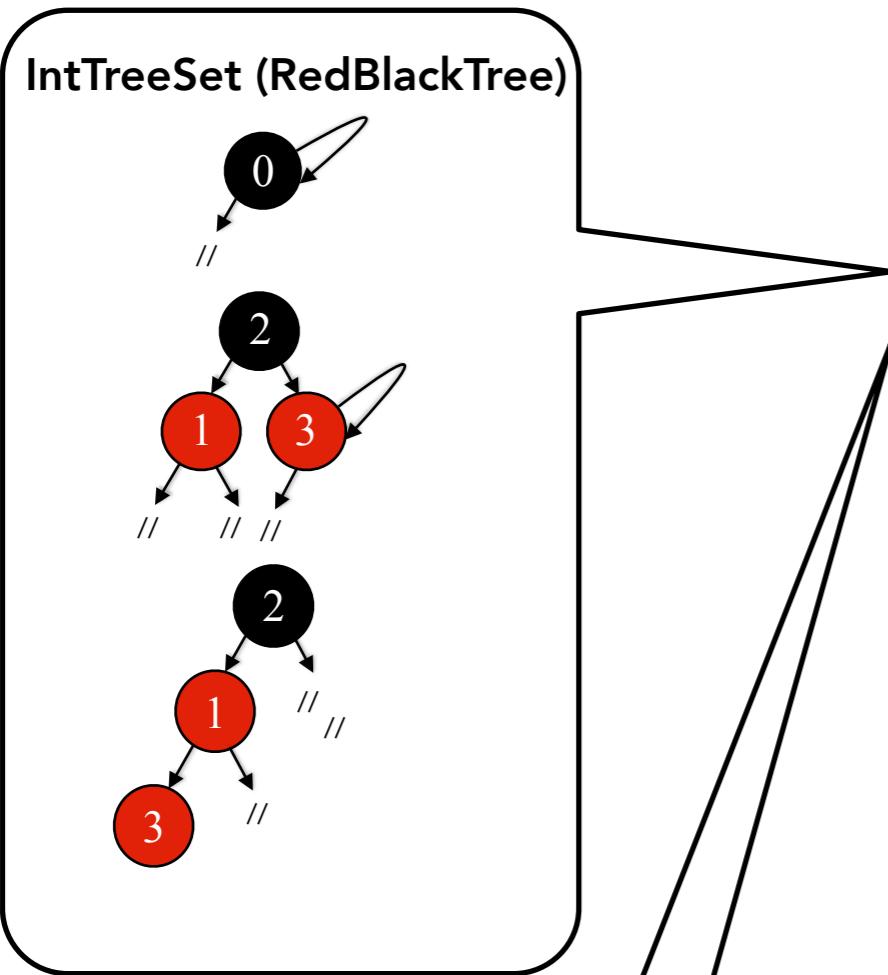
Case study	#bugs	#found without oracle	#found using NN
IntTreeSet	5	0	5
Antlr	1	0	1
FibHeap	1	0	1
BinTree	1	0	1
BinHeap	1	0	1
Schedule	8	3	4

RQ3: Improved Bug Finding?



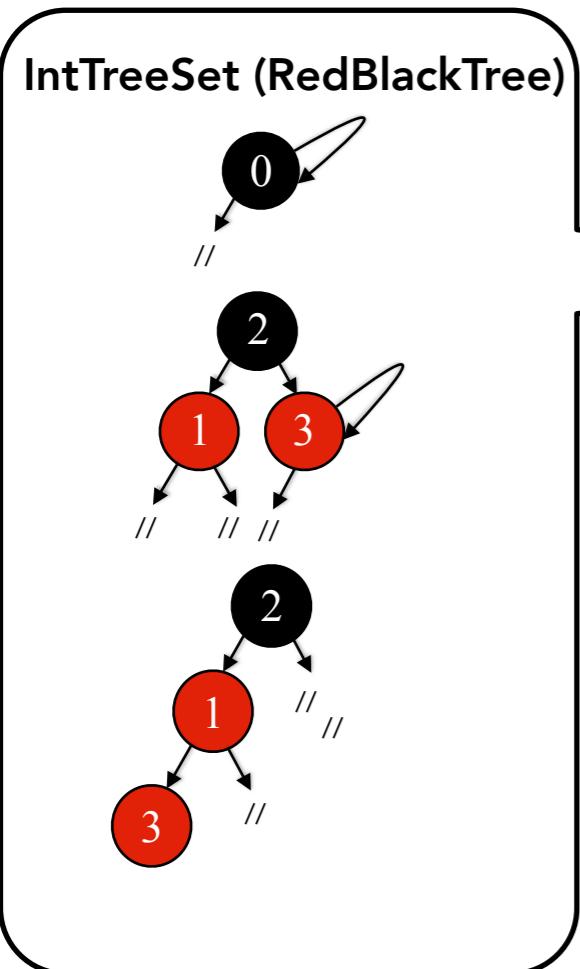
Case study	#bugs	#found without oracle	#found using NN
IntTreeSet	5	0	5
Antlr	1	0	1
FibHeap	1	0	1
BinTree	1	0	1
BinHeap	1	0	1
Schedule	8	3	4

RQ3: Improved Bug Finding?

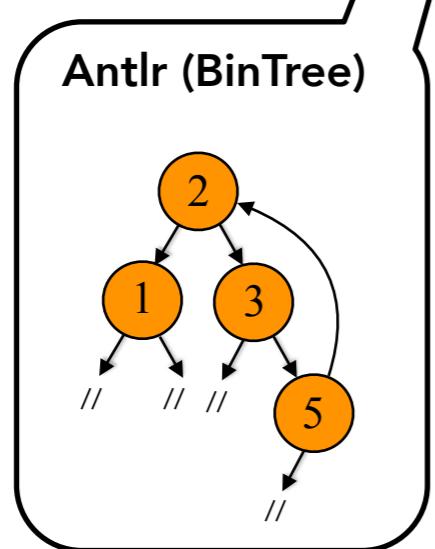


Case study	#bugs	#found without oracle	#found using NN
IntTreeSet	5	0	5
Antlr	1	0	1
FibHeap	1	0	1
BinTree	1	0	1
BinHeap	1	0	1
Schedule	8	3	4

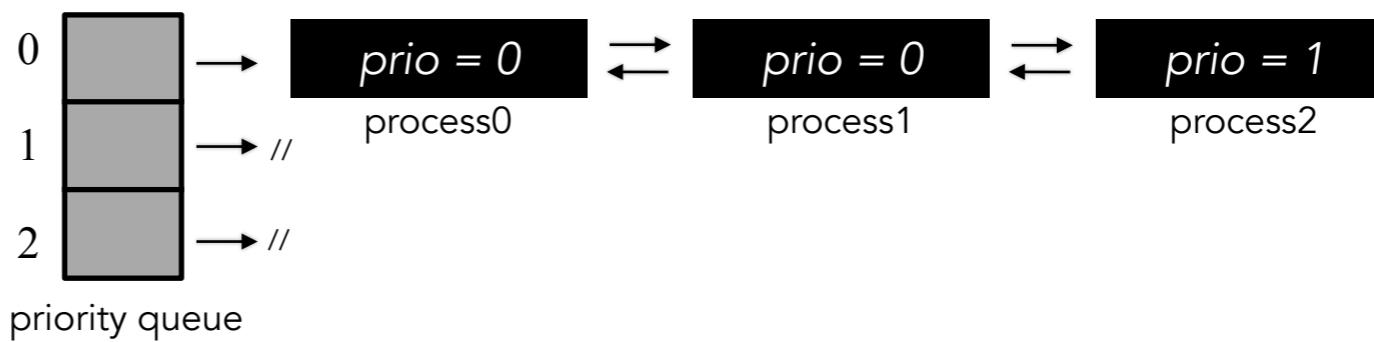
RQ3: Improved Bug Finding?



Case study	#bugs	#found without oracle	#found using NN
IntTreeSet	5	0	5
Antlr	1	0	1
FibHeap	1	0	1
BinTree	1	0	1
BinHeap	1	0	1
Schedule	8	3	4



Scheduler (Doubly Linked Lists indexed in an array)



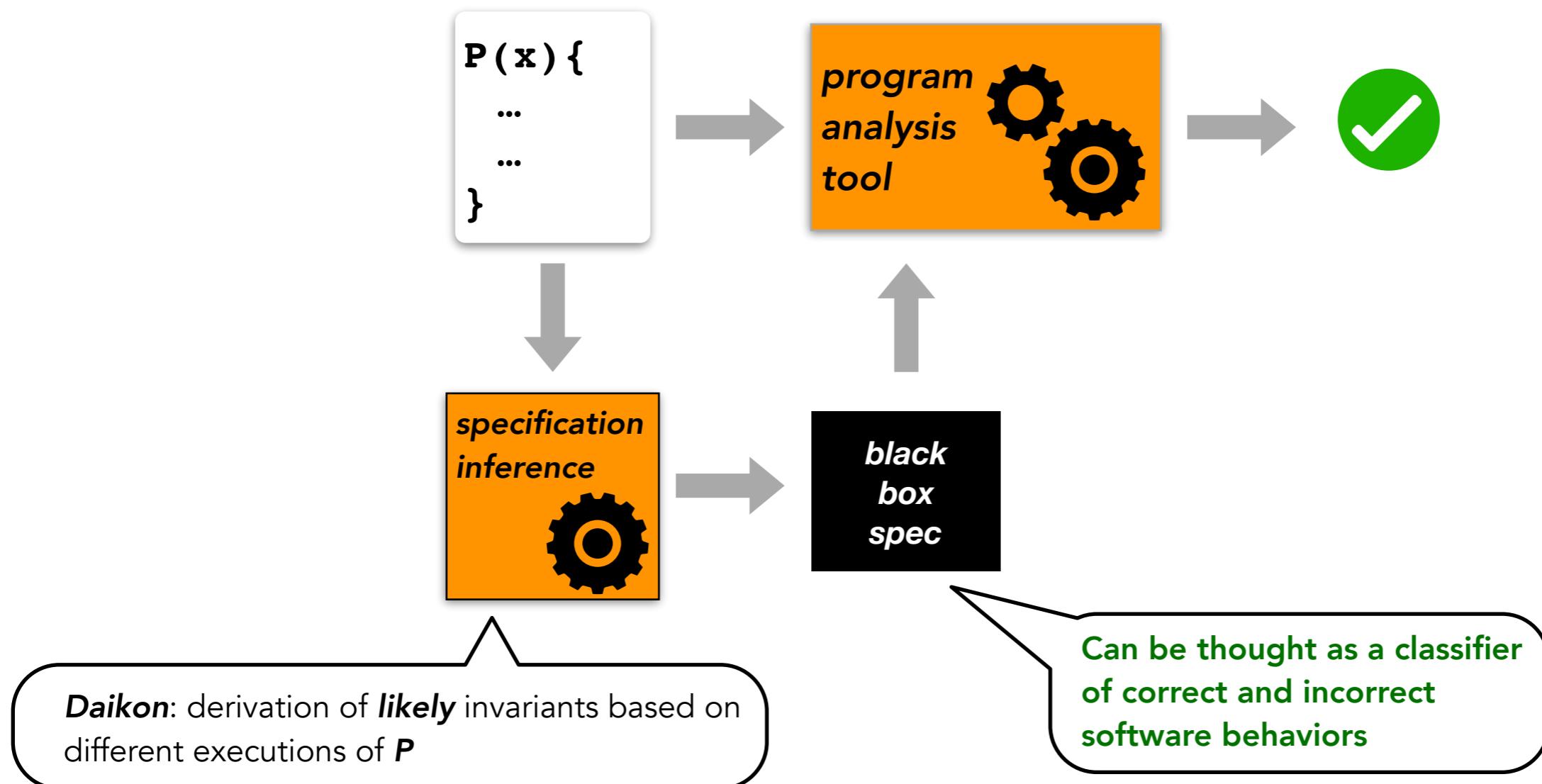
Remarks

- We defined a technique to generate valid and invalid objects from a set of assumed-correct object builders
- We proposed to use the generated objects to train a neural network to distinguish valid from invalid
- We used the trained neural network in place of an invariant for bug finding

Thank you :)

Questions?

Providing Specifications for Program Analysis

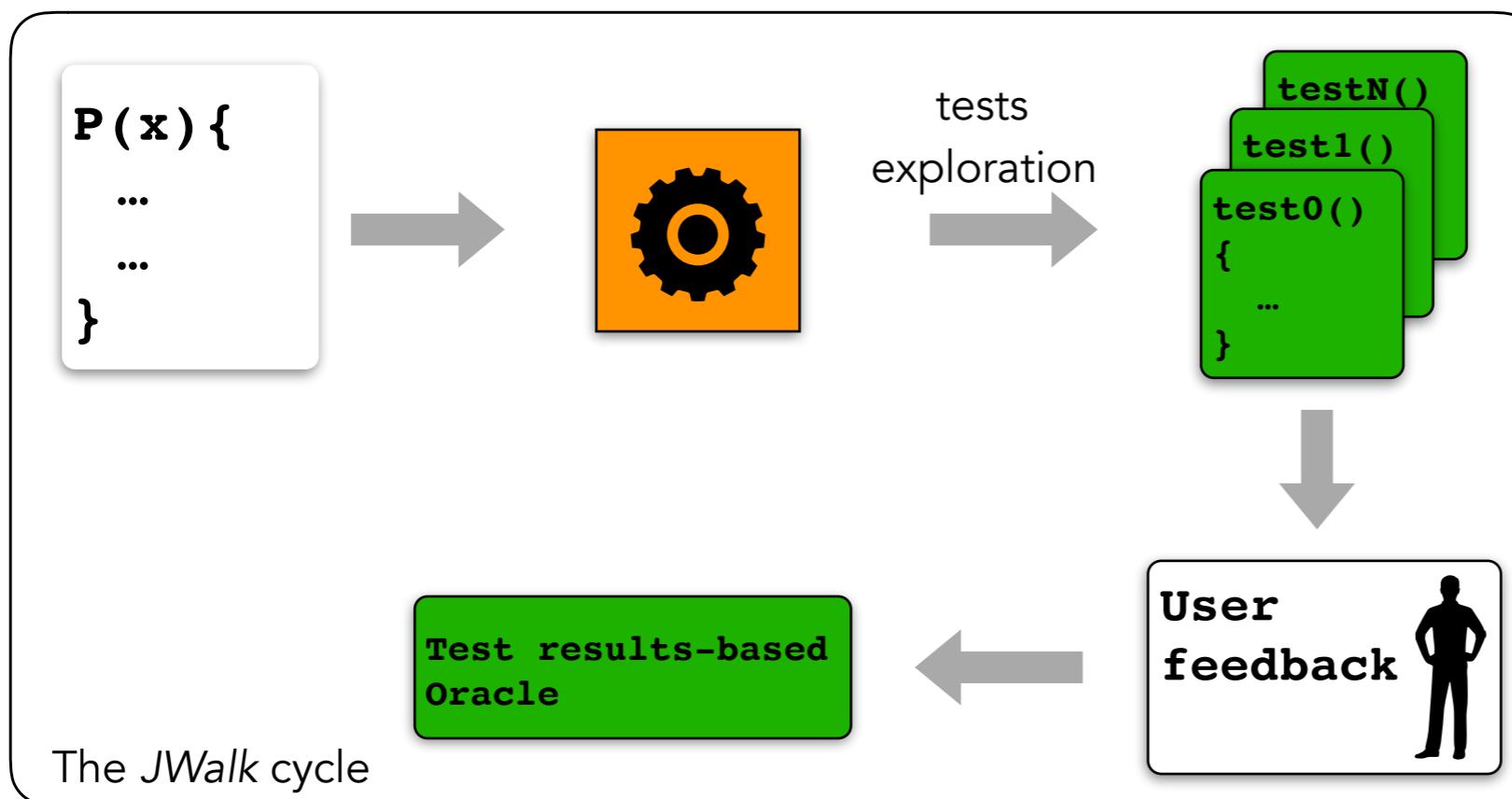


Current approaches to the Lack of Specifications

- Derivation of *likely* invariants based on different P executions



- JWalk**: allows dynamic inference of specifications with a feedback-based approach

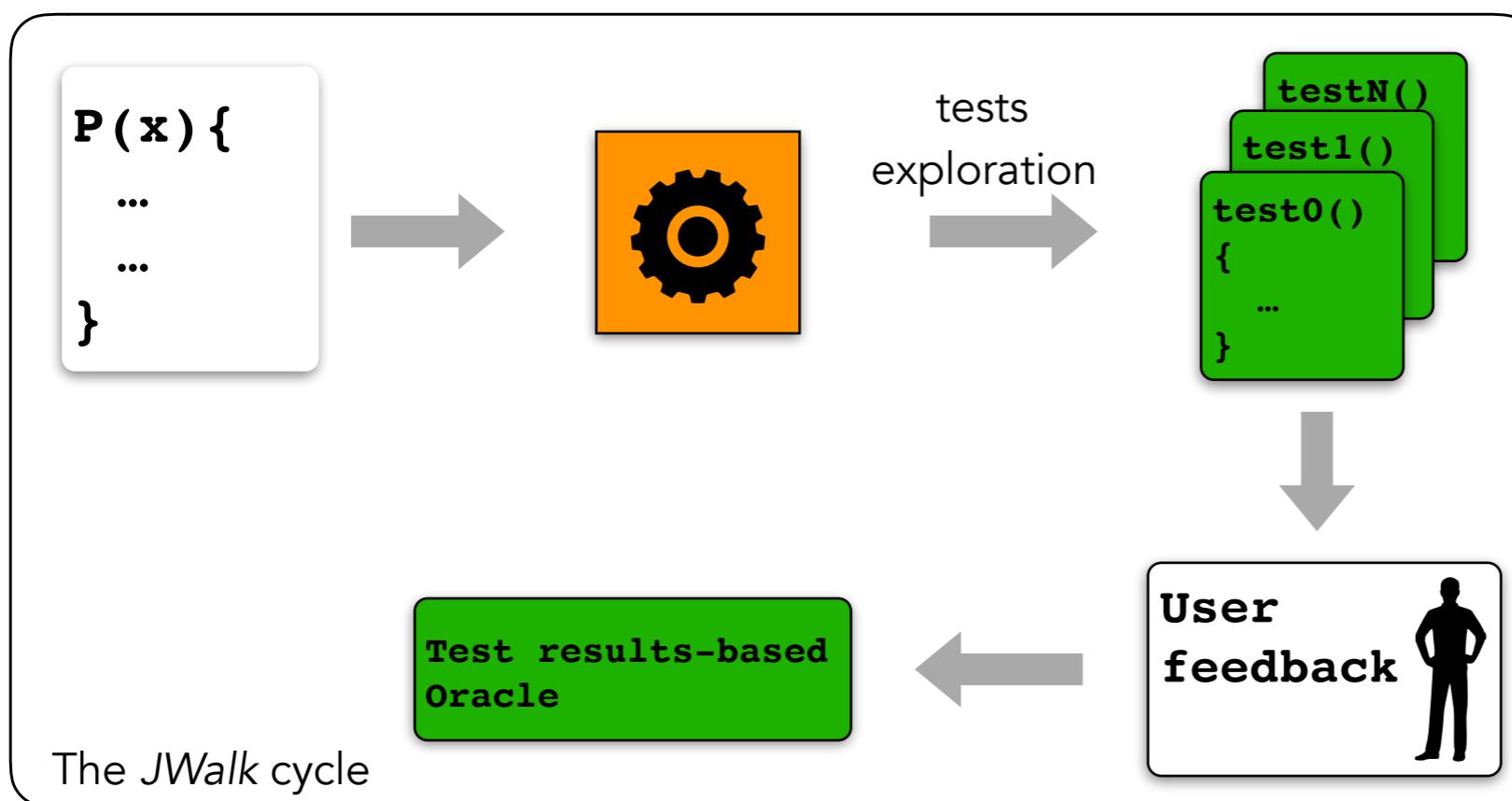


Current approaches to the Lack of Specifications

- Derivation of *likely* invariants based on different P executions



- JWalk**: allows dynamic inference of specifications with a feedback-based approach



- ✓ **human readable properties**
- ✗ **complex structural constraints are missed**
- ✗ **scenario specific oracles**

Daikon in the SinglyLinkedList case

```
public class SinglyLinkedList {  
    private Node header;  
    private int size;  
  
    public void SinglyLinkedList() {  
        header = new Node(0);  
        size = 0;  
    }  
  
    public void add(int n) {  
        ...  
    }  
  
    public class Node {  
        private int value;  
        private Node next;  
        ...  
    }  
}
```



```
this.header.value == 0;  
this.size >= 0;
```

- ✗ the acyclicity property is missed
- ✗ the size property is missed

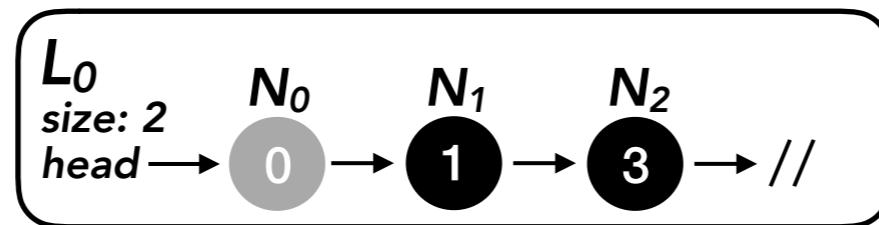
Precision and Recall vs the True Invariant for scope 7

```
public void test00100() {  
    public void test002() {  
        public void test001() {  
            SinglyLinkedList l = new SinglyLinkedList();  
            l.add(0);  
        }  
    }  
}
```

Validation Set		Precision	Recall
✗	✓		
✗	9	725964	100%
✓	0	55987	7%

Instances as Vectors

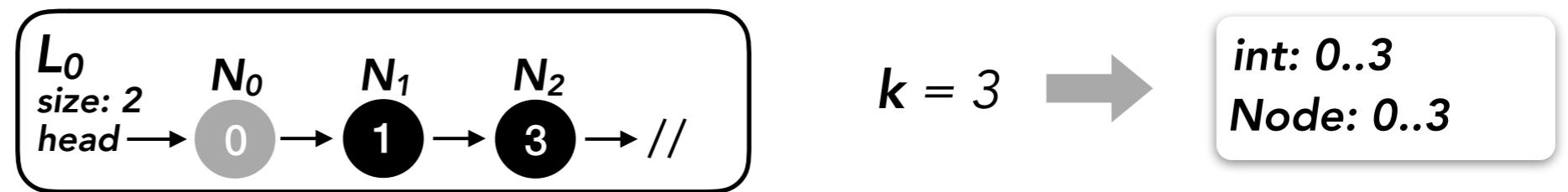
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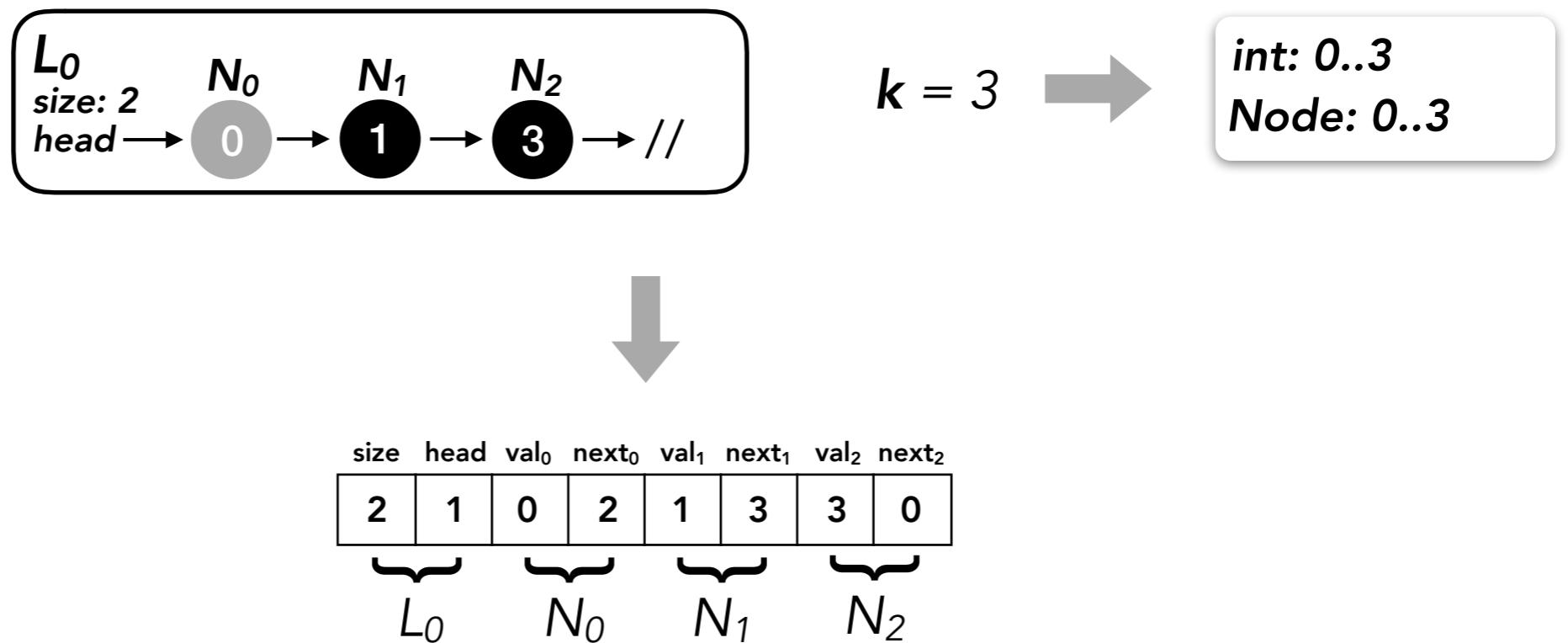
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Example:



RQ1: Measure of False Negatives Rate

