

Computer-Aided Reasoning for Software

# **Symbolic Execution**

# CSE507

[courses.cs.washington.edu/courses/cse507/14au/](https://courses.cs.washington.edu/courses/cse507/14au/)

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# Today

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## Last lecture

- Bounded verification: forward VCG for finitized programs

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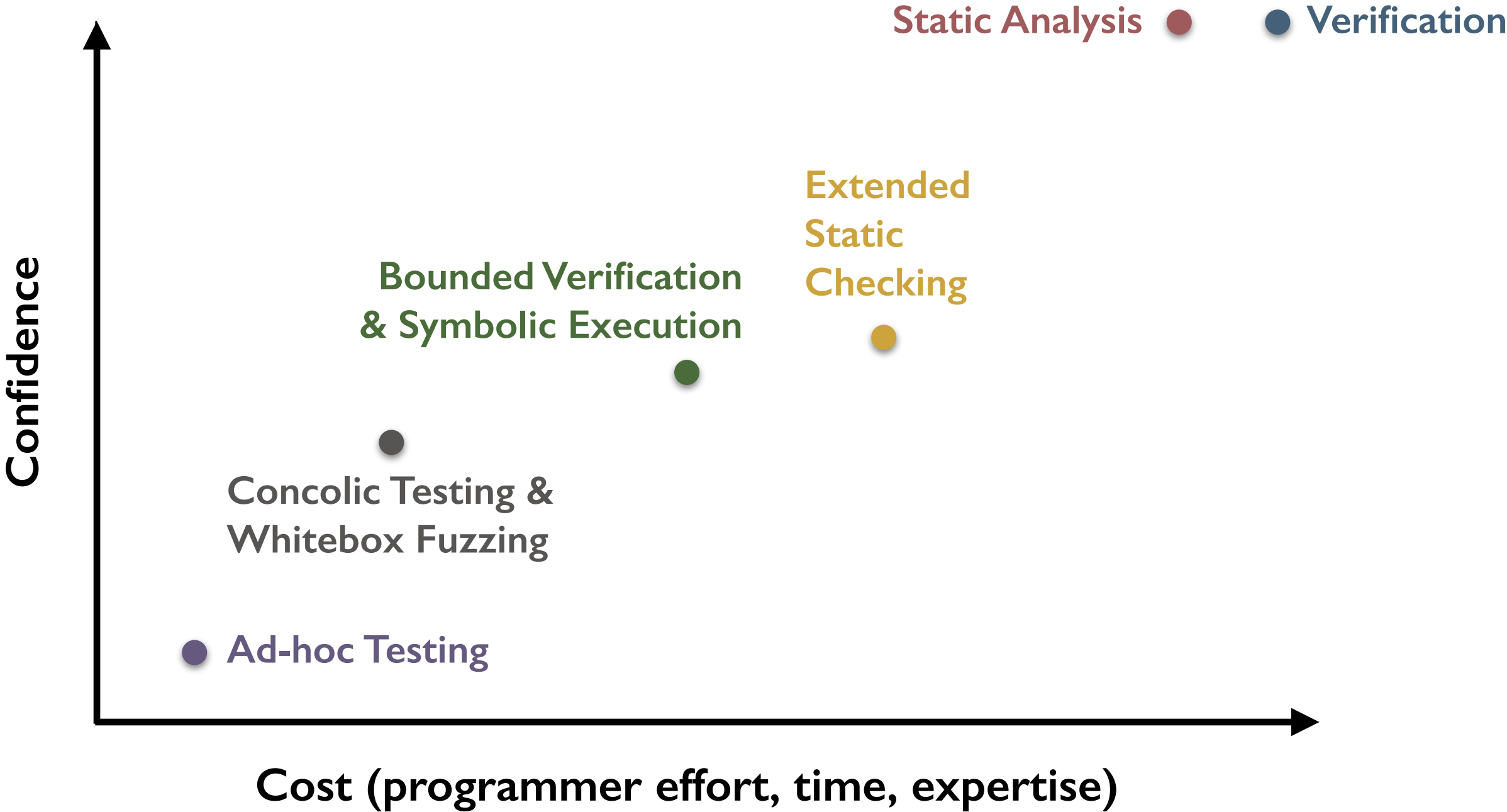
## Last lecture

- Bounded verification: forward VCG for finitized programs

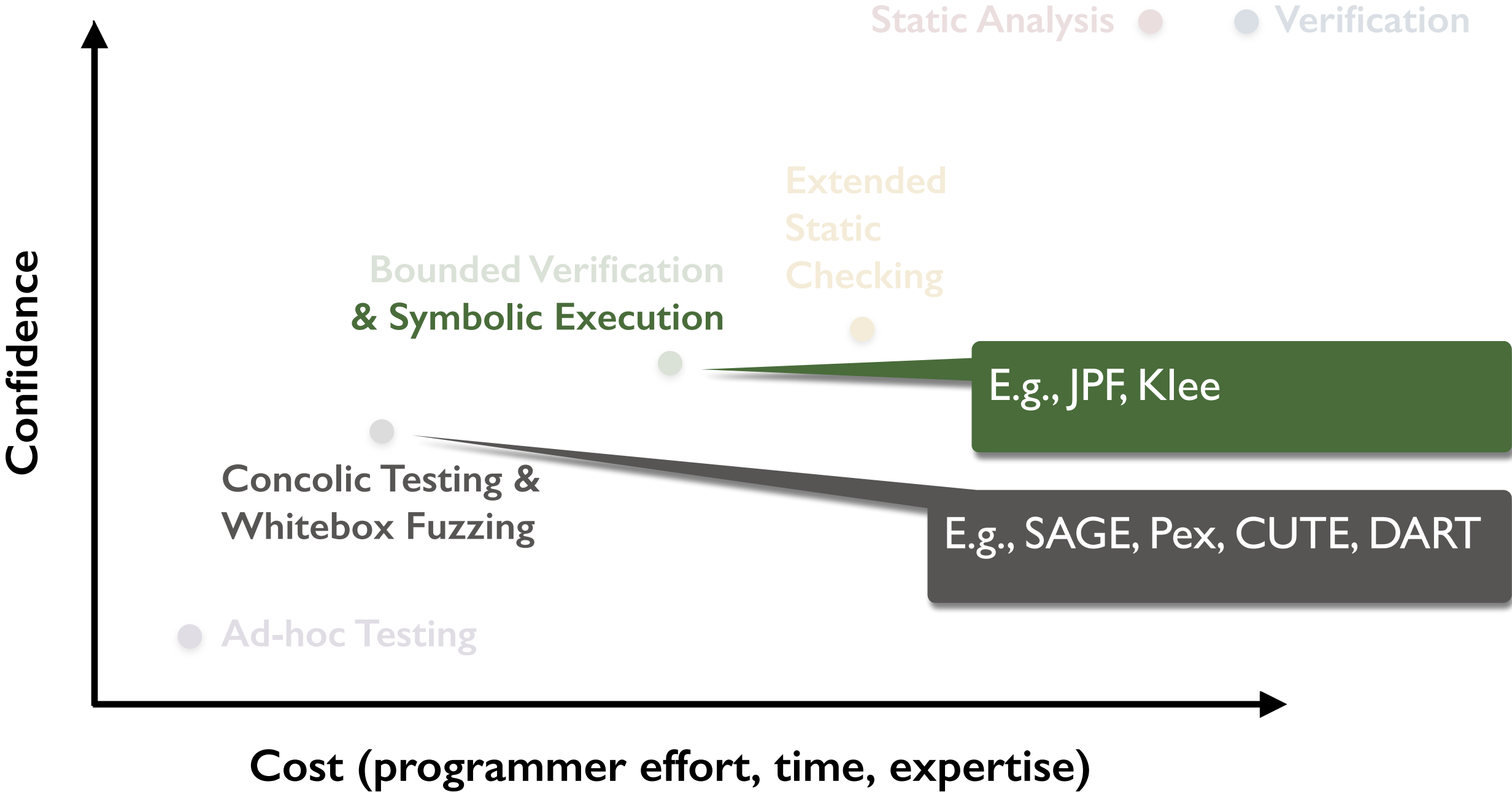
## Today

- Symbolic execution: a path-based translation
- Concolic testing

# The spectrum of program validation tools



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# Symbolic execution

**1976:** *A system to generate test data and symbolically execute programs* (Lori Clarke)

**1976:** *Symbolic execution and program testing* (James King)

**2005-present:** practical symbolic execution

- Using SMT solvers
- Heuristics to control exponential explosion
- Heap modeling and reasoning about pointers
- Environment modeling
- Dealing with solver limitations

# Classic symbolic execution

```
def f (x, y):  
    if (x > y):  
        x = x + y  
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$x \mapsto X$   
 $y \mapsto Y$

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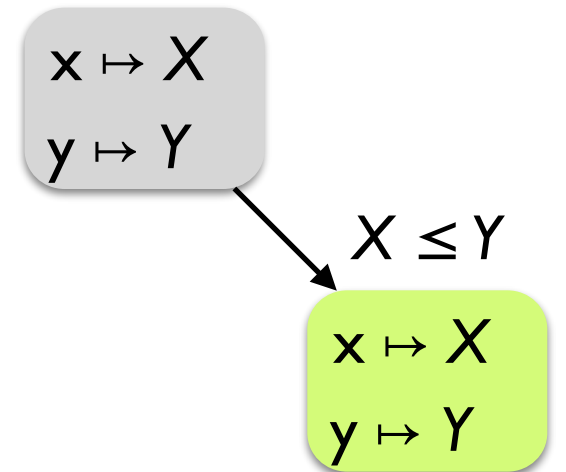
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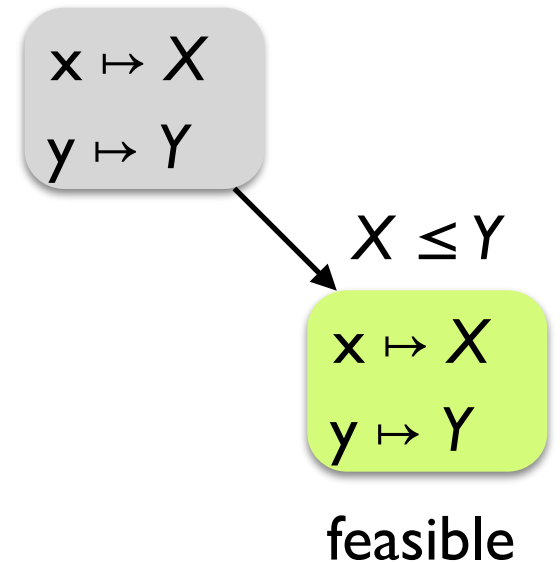
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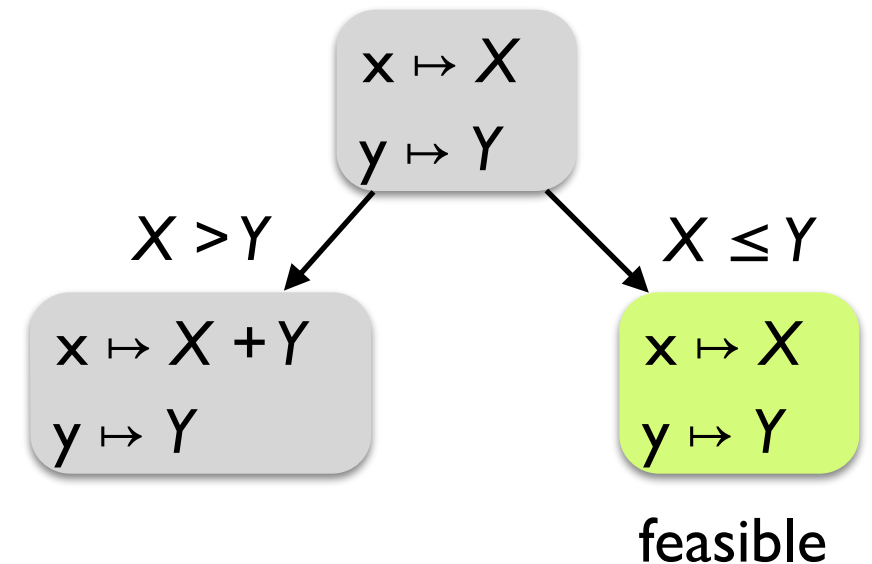
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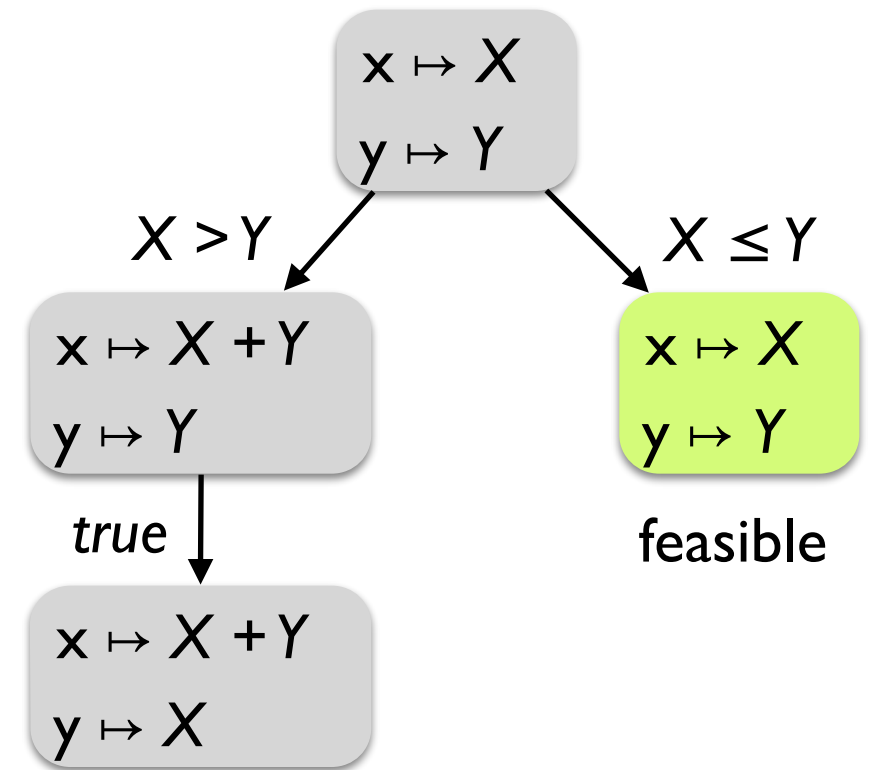
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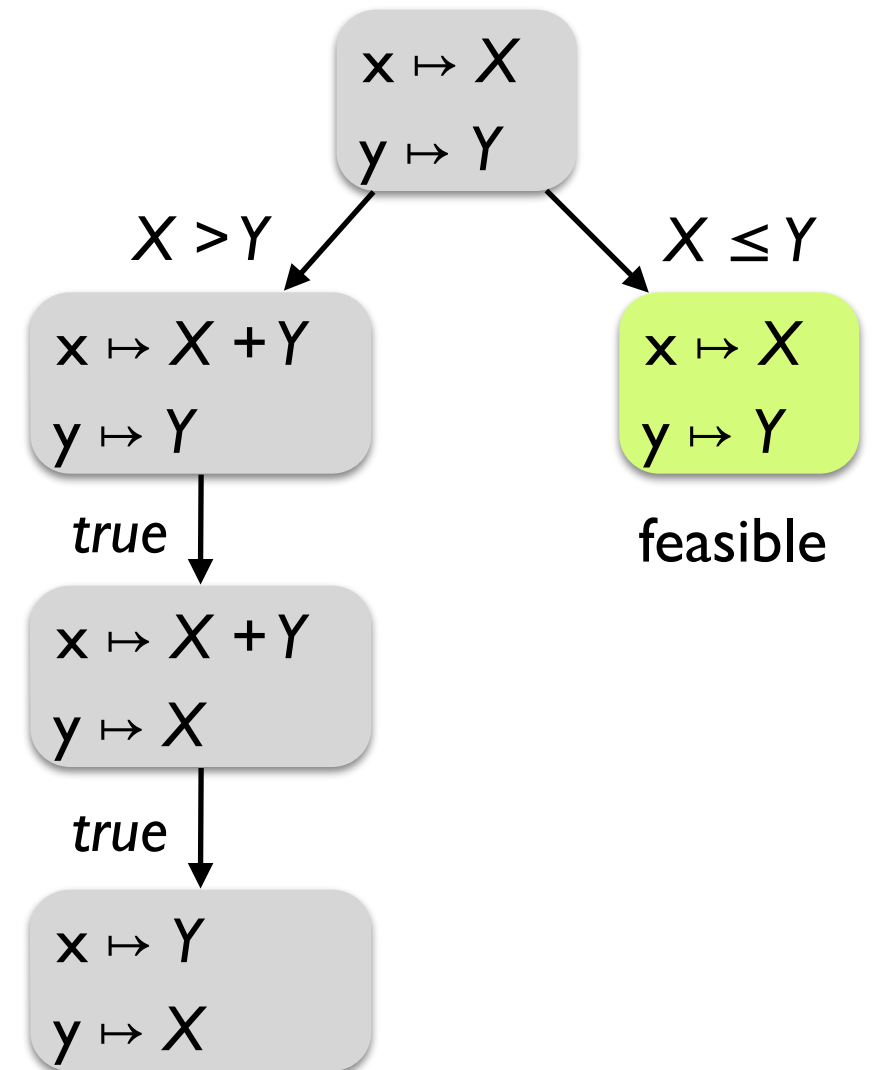
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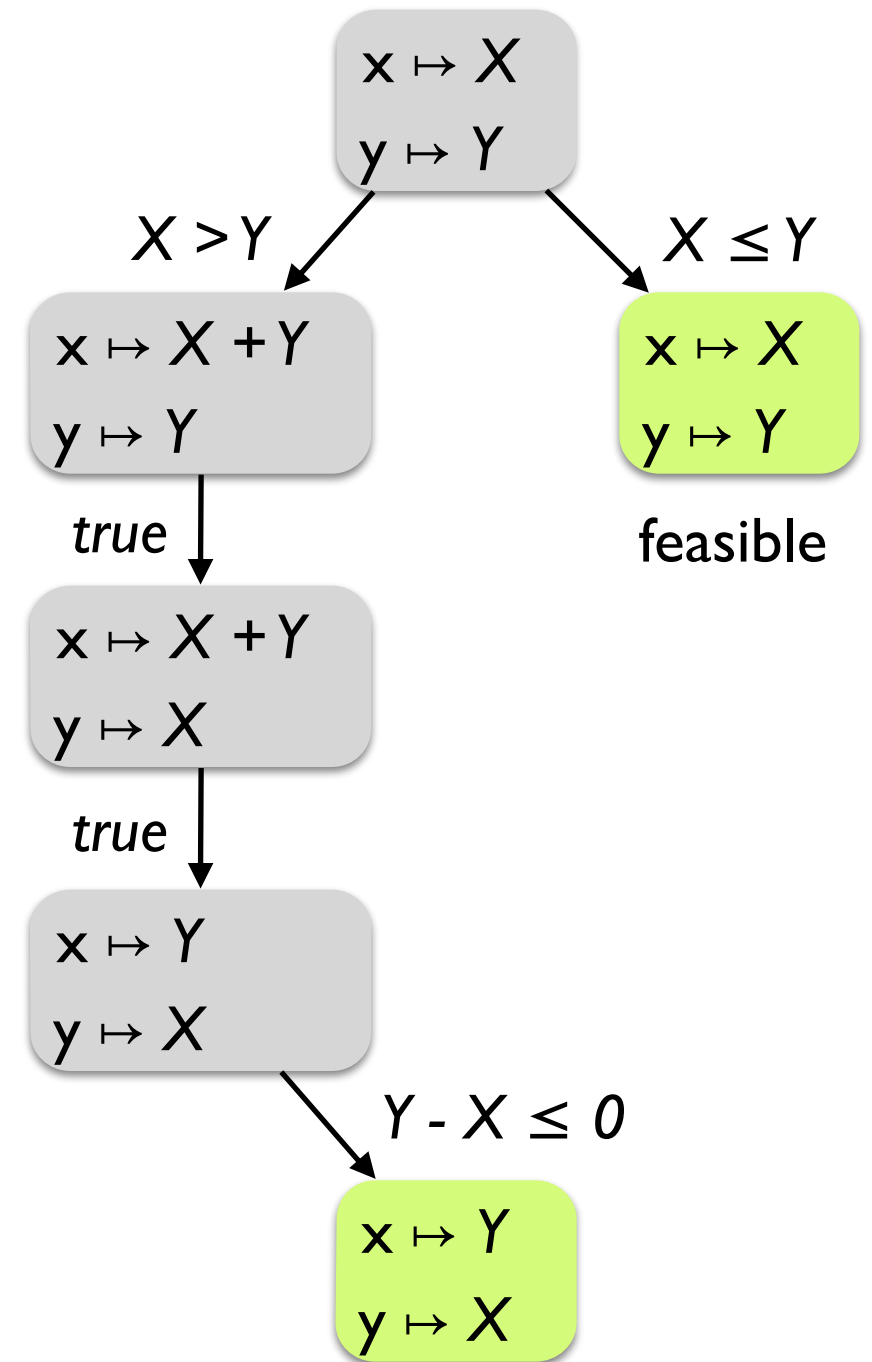
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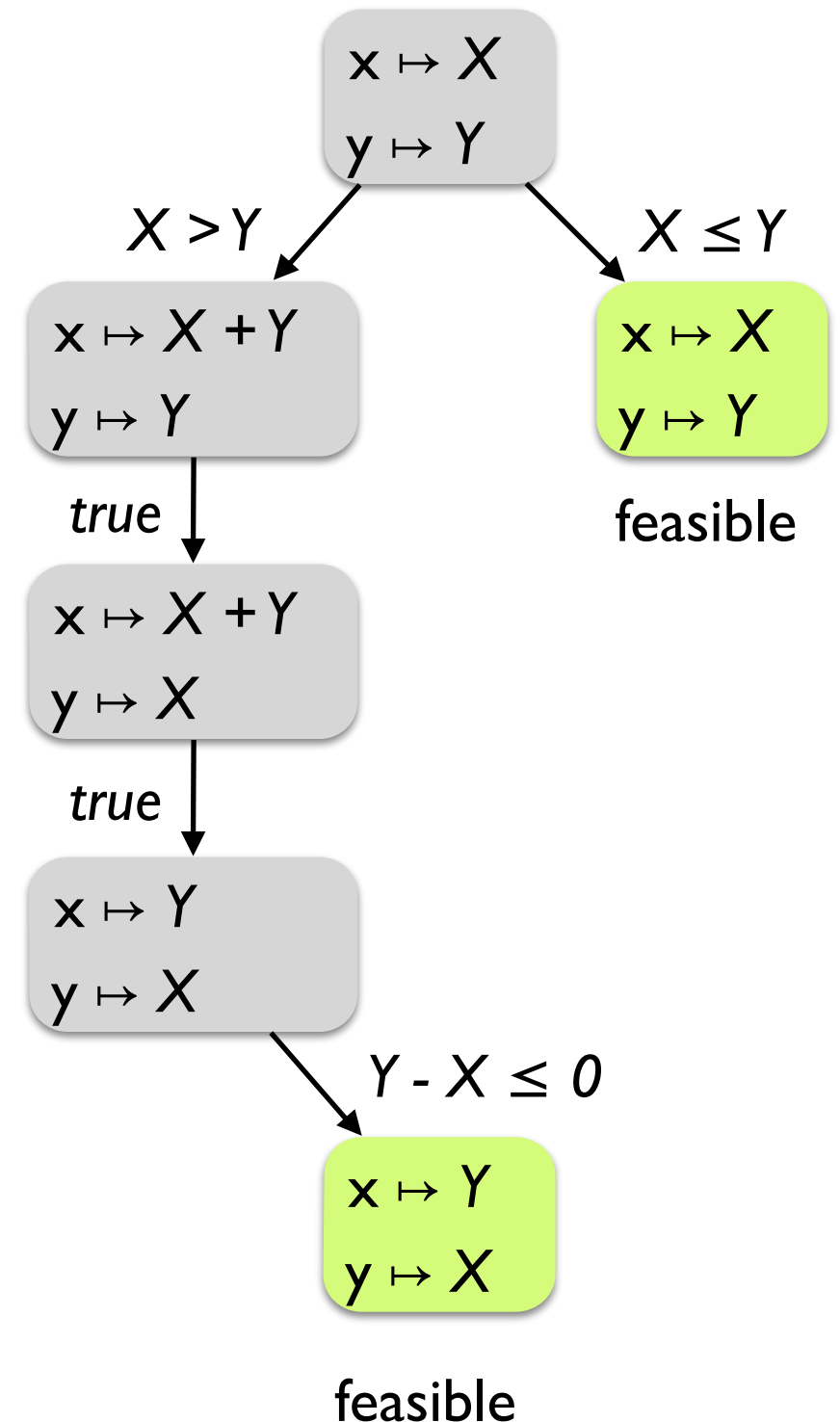
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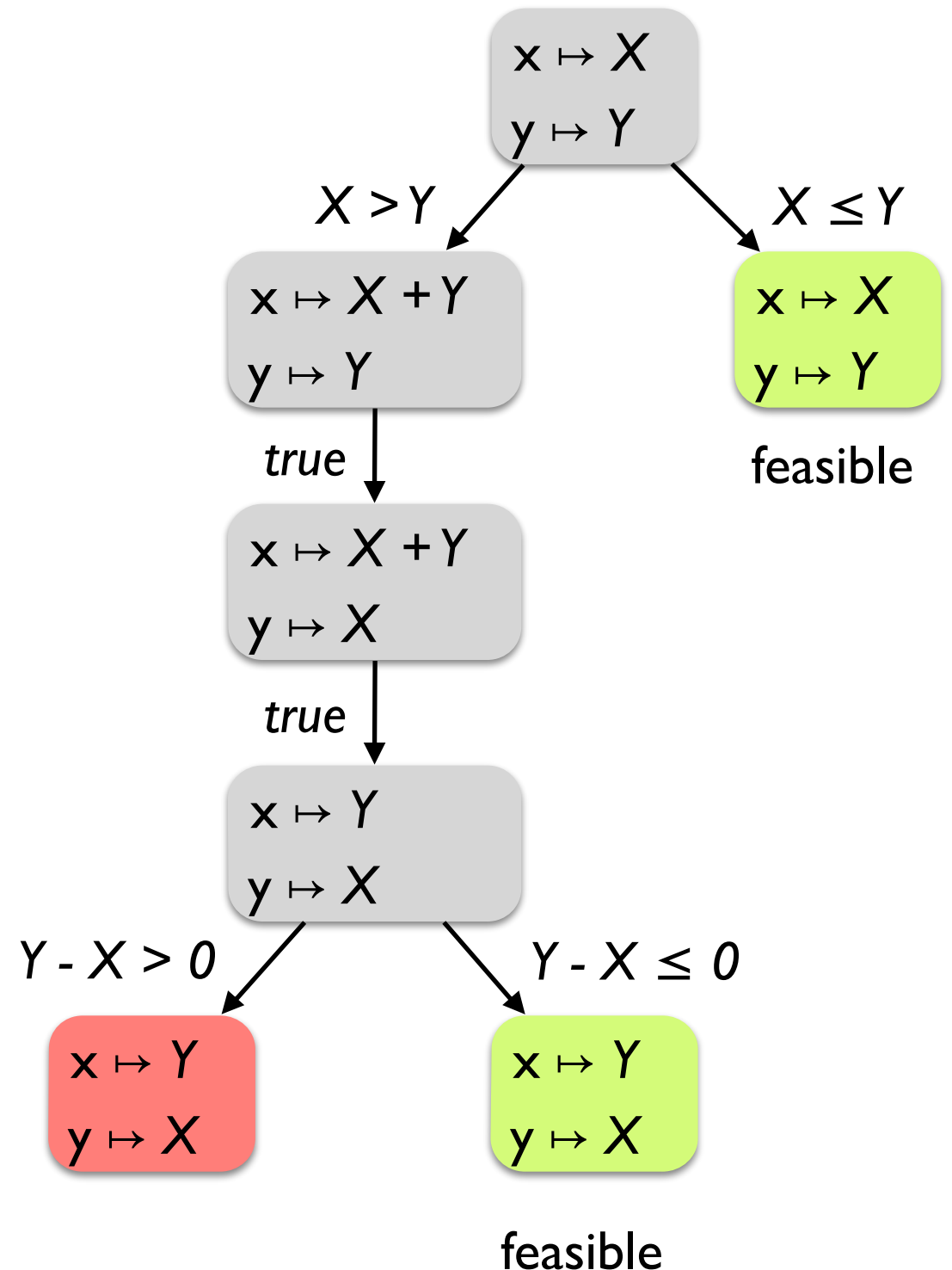
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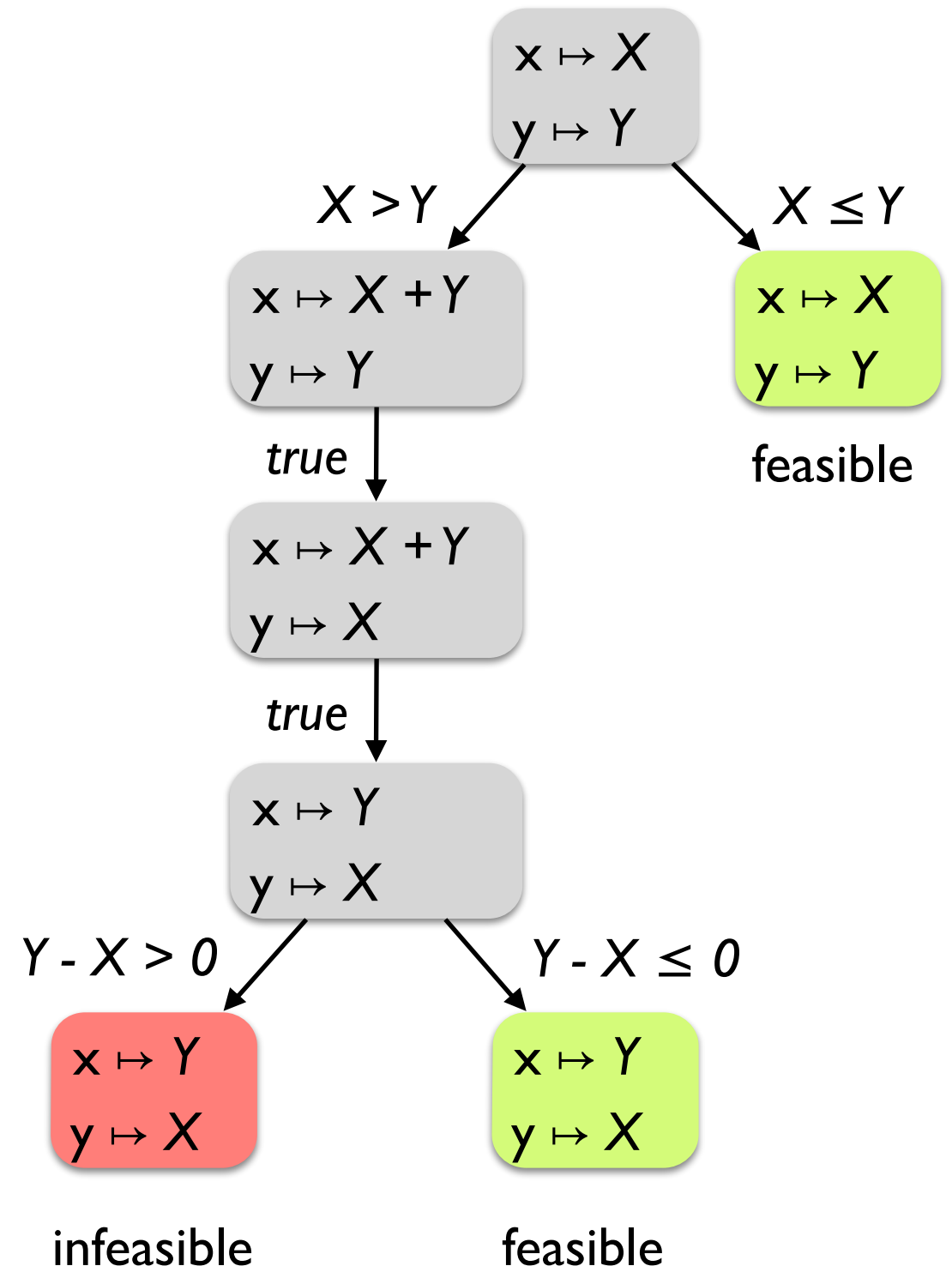
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**Environment modeling:** dealing with native / system / library calls

# Classic symbolic execution: practical issues

**Loops and recursion:** infinite execution trees

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**Heap modeling:** symbolic data structures and pointers

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# Loops and recursion

Dealing with infinite execution trees:

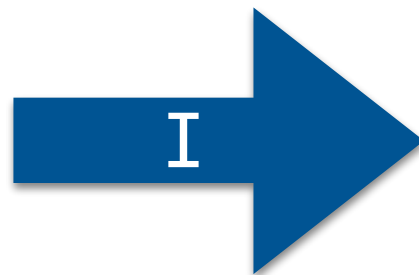
- Finitize paths by limiting the size of PCs (bounded verification)
- Use loop invariants (verification)

# Loops and recursion

Dealing with infinite execution trees:

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- Use loop invariants (verification)

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init;  
while (C) {  
    B;  
}  
assert P;
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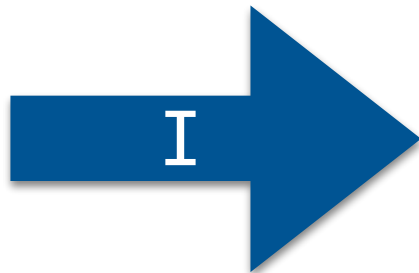


# Loops and recursion

Dealing with infinite execution trees:

- Finitize paths by limiting the size of PCs (bounded verification)
- Use loop invariants (verification)

```
init;  
while (C) {  
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}  
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```



```
init;  
assert I;  
makeSymbolic(targets(B));  
assume I;  
if (C) {  
    B;  
    assert I;  
} else  
    assert P;
```

# Path explosion

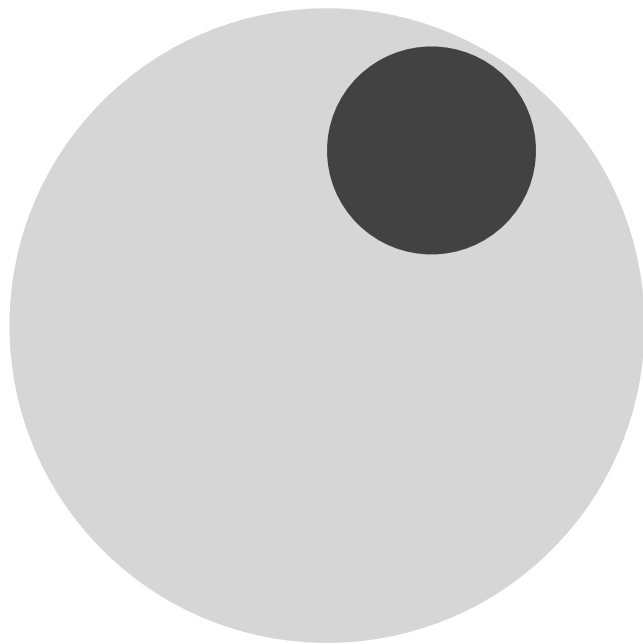
Achieving good coverage in the presence of exponentially many paths:

- Select next branch at random
- Select next branch based on coverage
- Interleave symbolic execution with random testing

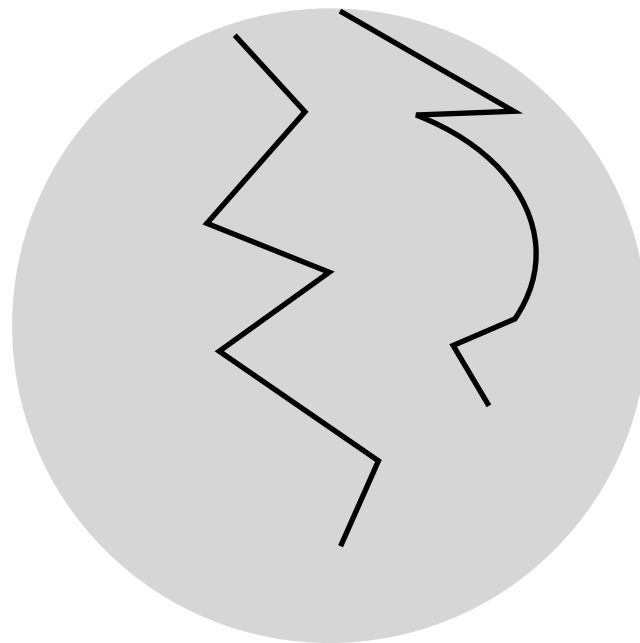
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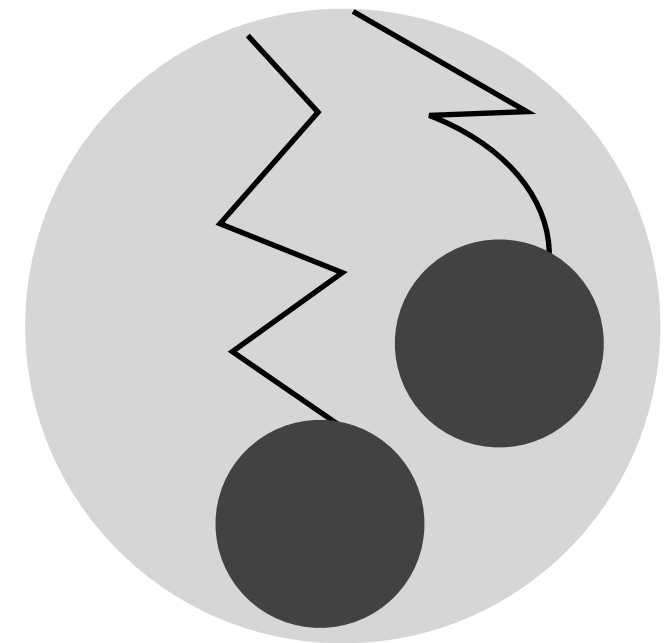
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symbolic execution



random testing



interleaved execution

# Heap modeling

Modeling symbolic heap values and pointers

- Segmented address space via the theory of arrays (Klee)
- Lazy concretization (JPF)
- Concolic lazy concretization (CUTE)



# General lazy concretization

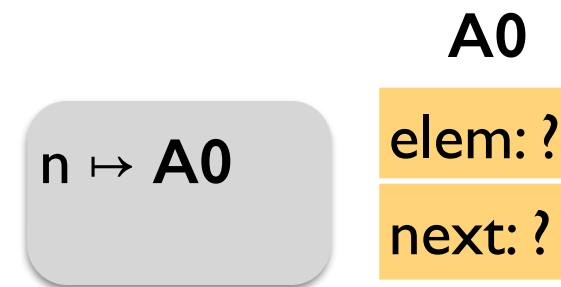
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class Node {  
    int elem;  
    Node next;  
}
```

```
n = symbolic(Node);  
x = n.next;
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# General lazy concretization

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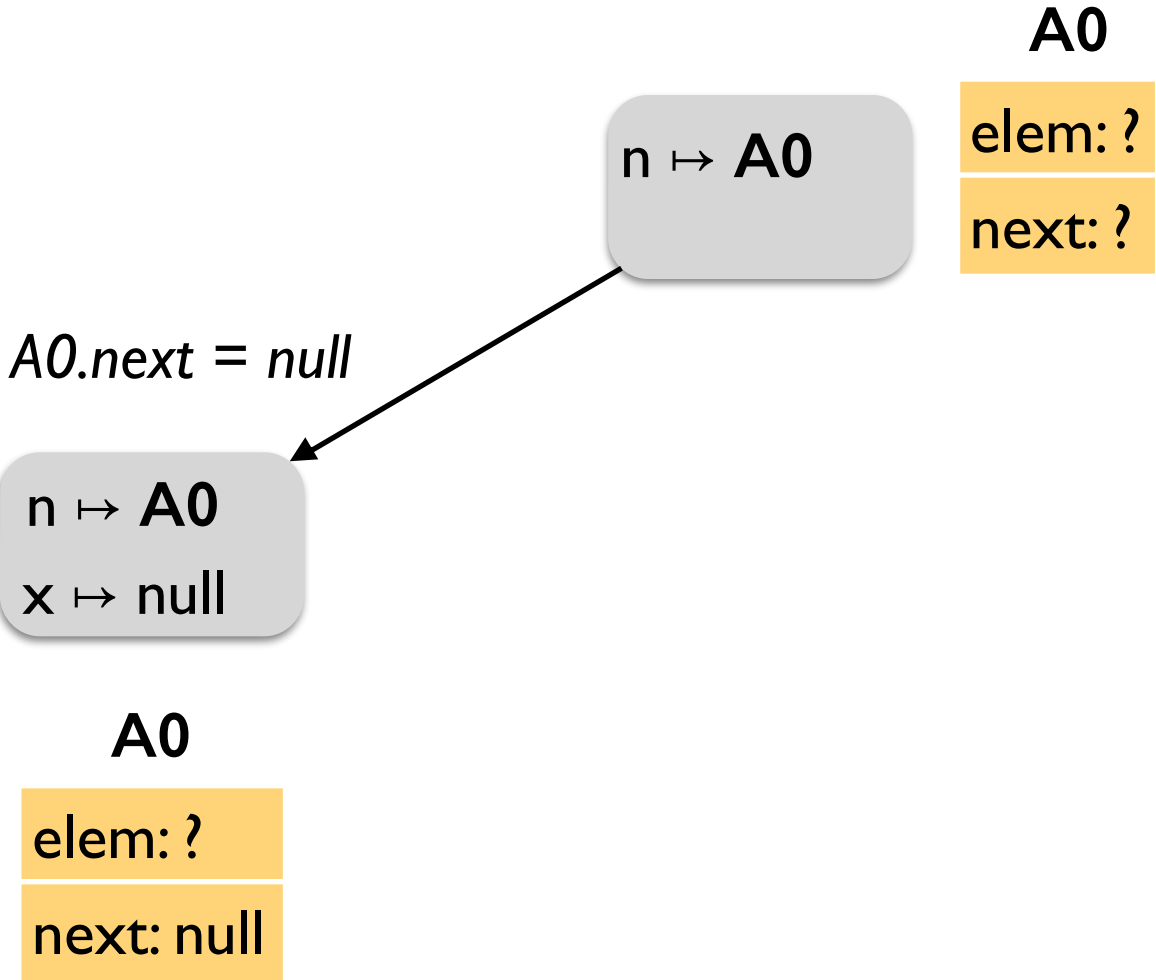
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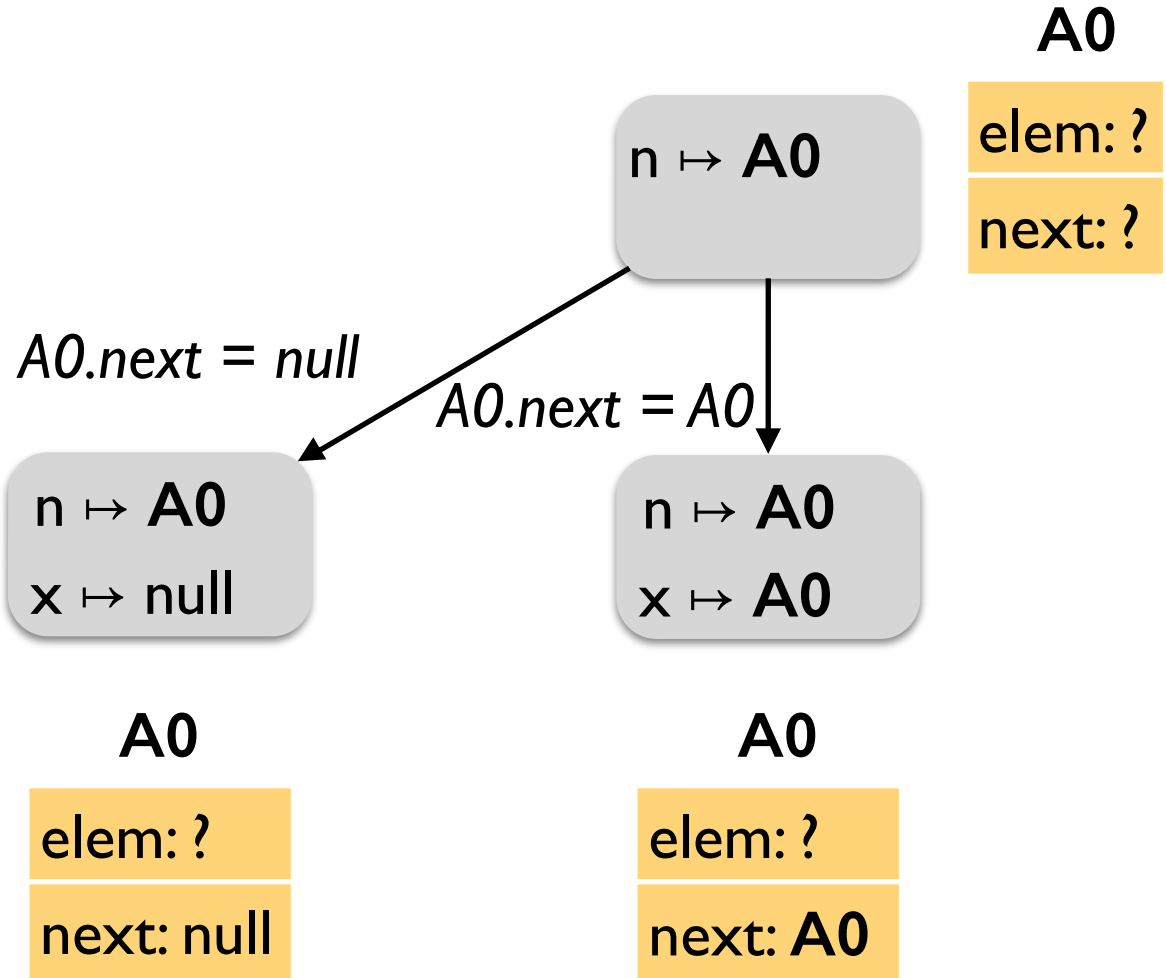
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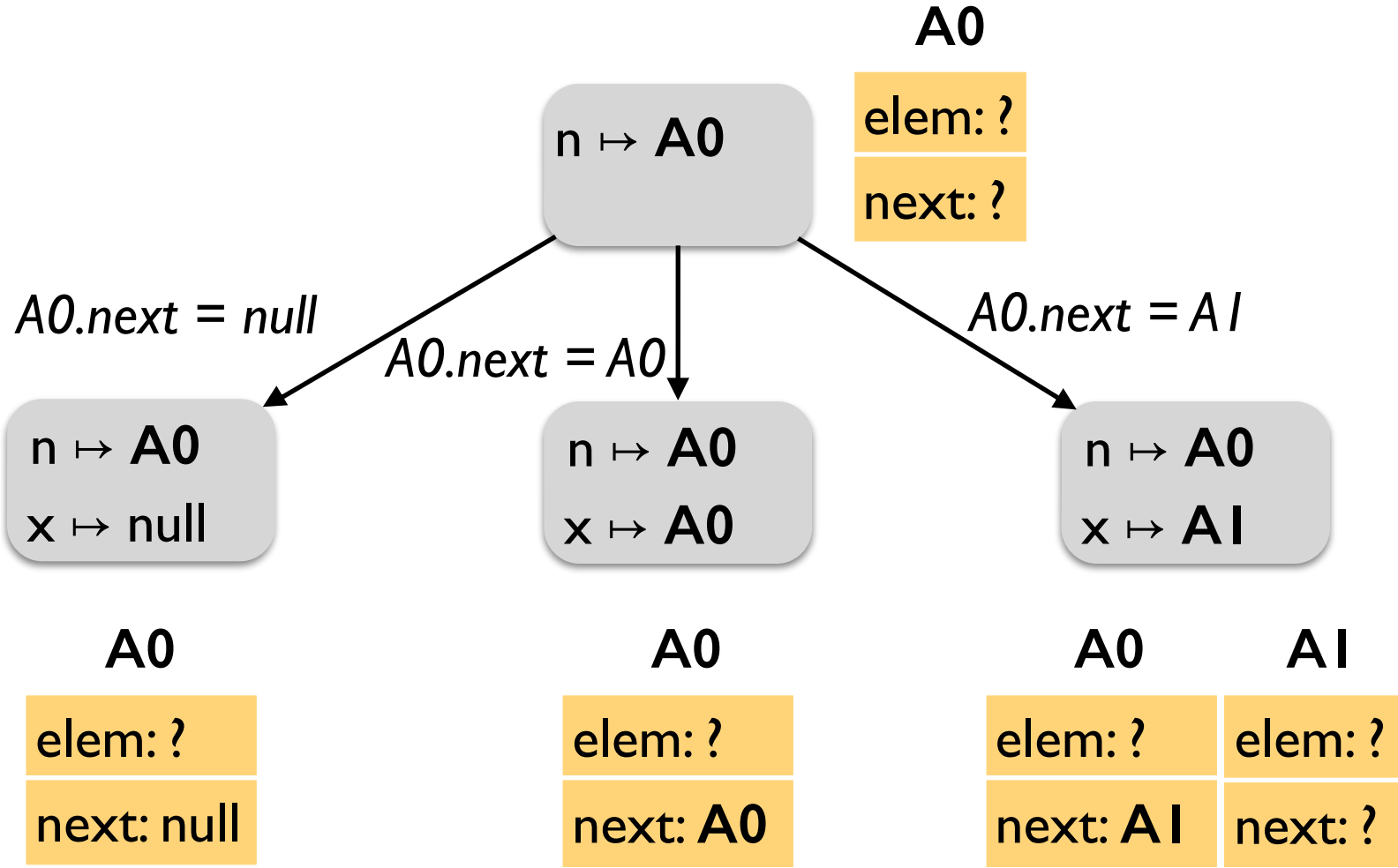
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# Concolic testing

```
typedef struct cell {
    int v;
    struct cell *next;
} cell;

int f(int v) {
    return 2*v + 1;
}

int testme(cell *p, int x) {
    if (x > 0)
        if (p != NULL)
            if (f(x) == p->v)
                if (p->next == p)
                    abort();
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Concrete

$p \mapsto \text{null}$   
 $x \mapsto 236$

PC

$x > 0 \wedge p = \text{null}$

Execute concretely and symbolically.

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next: null

v: 634

$p \mapsto \text{A0}$   
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$x > 0 \wedge p \neq \text{null} \wedge$   
 $p.v \neq 2x + 1$

Execute concretely and symbolically.

Negate last decision and solve for new inputs.



# Concolic testing

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next: null

v: 3

$p \mapsto \text{A0}$   
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Execute concretely and symbolically.

Negate last decision and solve for new inputs.

# Solver limitations

Reducing the demands on the solver:

- On-the-fly expression simplification
- Incremental solving
- Solution caching
- Substituting concrete values for symbolic in complex PCs (CUTE)

# Environment modeling

Dealing with system / native / library calls:

- Partial state concretization
- Manual *models* of the environment (Klee)

# Summary

## Today

- Practical symbolic execution and concolic testing

## Next lecture

- Basics of model checking