

# Generation of Test Data Structures Using Constraint Logic Programming

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# Software testing

- Increasing **confidence** in software correctness

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- Focus: how to **generate** input-output pairs to check the behavior of a **function/module**

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Input = generate()  
Output = my_function(Input)  
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- **White-box** (code-aware)  
select input-output pairs to **maximize** a code **coverage** criteria
- **Black-box** (model-based)  
select input-output pairs using a given **model** (specification)

# Developing Generators for Black-box Testing

**Black-box** : select input-output pairs w.r.t. a given **model**

(A) **imperative** (direct generation) [e.g., ASTGen]  
*specifies **how** to compute tests*

## PROS

- **fast** (ad-hoc)

## CONS

- **high effort** in development
- **error-prone**

(B) **declarative** (filtering-based) [e.g., Korat, TestEra]  
*specifies **which** tests to compute*

## PROS

- **low effort** in development
- **correctness** improved

## CONS

- **slow** (generate-and-test)

# Bounded-Exhaustive Testing (BET)

Small scope hypothesis [Jackson]:

*a large portion of faults is likely to be revealed  
already by testing all inputs up to a small scope*

BET: testing software for **all valid inputs** up to a given **size bound**

**Challenge:** **efficient** & **correct** generators of **complex data structures**

**Applications:** testing of

- data-structures manipulating programs
- web-based applications (accepting XML, PHP, HTML, . . . inputs)
- . . .

# Constraint Logic Programming -based Generation

## Constraint Logic Programming :

- **declarative** programming
- **constraint** solving
- **symbolic** execution
- checkpoint-based exploration  
(**backtracking** + **unification**)

} efficient & correct generators  
of complex data structures

- Constraint-based  
[DeMillo-Offutt '91, Meudec (ATGen) '01, Euclide '09]
- Constraint Logic Programming -based  
[PathCrawler '05, Charreteur-Botella-Gotlieb '01, jPET '11]

# Plan

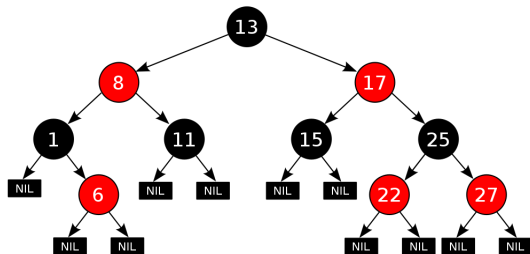
1. Running Example and Korat
2. Use of the Constrain-and-Generate Paradigm
3. Optimizations



# Red-Black Tree

Nodes are:

- colored (red or black)
- marked (by a key)



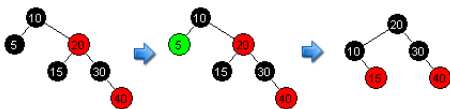
A binary tree satisfying the following invariants:

- $(I_1)$  **red** nodes have **black children**
- $(I_2)$  every **root-to-leaf path** has the same **number of black nodes**
- $(I_3)$  keys are **ordered** left-to-right

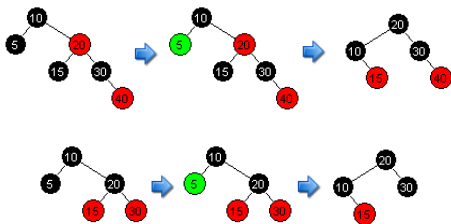
Efficient worst-case insertion/deletion/retrieval

# Red-Black Tree - Node Removal

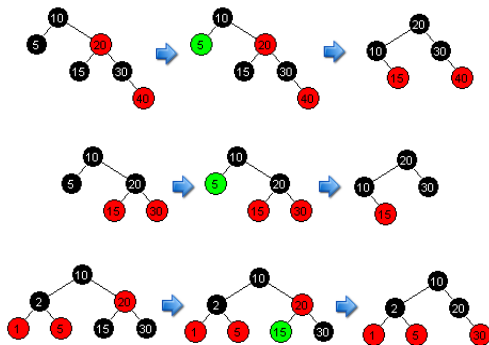
# Red-Black Tree - Node Removal



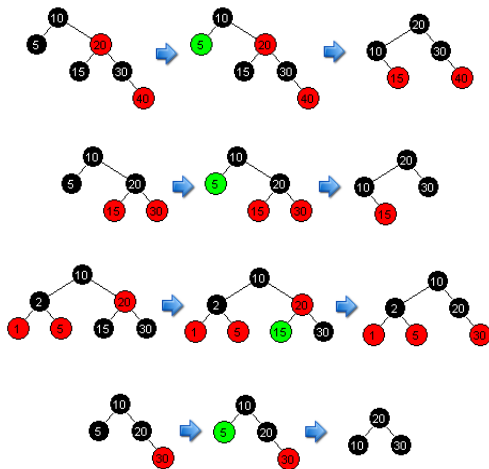
# Red-Black Tree - Node Removal



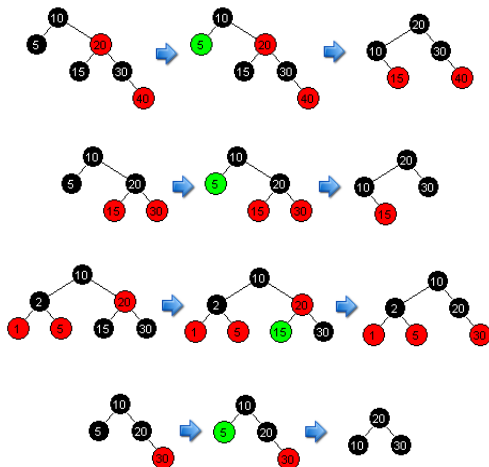
# Red-Black Tree - Node Removal



# Red-Black Tree - Node Removal



# Red-Black Tree - Node Removal



~ 300 line of Java code for node deletion...

# KORAT: filter-based test-case generation

```
public class RedBlackTree {
    Node root;
    int size;
    class Node {
        int key, value;
        Node left, right, parent;
        int color;
    }
}
```

```
boolean repOK() { ... }
```

1. build a tuple of the variables
2. fix a finite domain for variables
3. enumerate concrete tuples
4. select valid structures using repOK
5. pruning irrelevant instances by monitoring fields accessed by repOK

Keys in {0,1,2,3}

Size = 3     $\langle root, size, \underbrace{key_1, value_1, left_1, right_1, parent_1, color_1}_{\text{Node 1}}, \underbrace{\dots}_{\text{Node 2}}, \underbrace{\dots}_{\text{Node 3}} \rangle$

<i>domain</i>	2147483648	(combinatorial)
<i>explored</i>	248	(with pruning)
<i>found</i>	12	(desired)

$color_1, color_2, color_3 \in \{\text{RED, BLACK}\}$   
 $left_1, left_2, left_3 \in \{\text{null, 1, 2, 3}\}$   
 $key_1, key_2, key_3 \in \{0, 1, 2, 3\} \dots$



## KORAT on Red-Black Trees

size	<i>domain</i>	<i>explored</i>	<i>found</i>	time (s)
0	1	1	1	0.20
1	128	10	4	0.21
2	236196	52	6	0.22
3	2147483648	248	12	0.23
4	$6.10 \cdot 10^{13}$	1521	40	0.30
5	$4.21 \cdot 10^{18}$	6827	84	0.46
6	$6.01 \cdot 10^{23}$	25727	140	0.58
7	$1.58 \cdot 10^{29}$	109749	280	0.95
8	$7.12 \cdot 10^{34}$	481398	576	1.91
9	$5.12 \cdot 10^{40}$	2184788	1220	6.12
10	$5.61 \cdot 10^{46}$	10693872	2860	30.01
11	$8.99 \cdot 10^{52}$	54605852	7032	168.33
12	$2.04 \cdot 10^{59}$	279610078	16848	878.13
13	$6.37 \cdot 10^{65}$	1416658806	37912	5634.82

- Korat has an **efficient pruning** mechanism
- The problem is **inherently hard**
- How do we scale-up?

# Constraint Logic Programming

Programs as sets of rules (clauses) of the form:

$H :- c \wedge B$  (meaning, H holds if  $c$  is satisfiable in  $\mathcal{T}$  and B holds)

Example:

`ordered([]).`

`ordered([X]).`

`ordered([X1, X2 | L]) :-  $\underbrace{X_1 \leq X_2}_{\text{solver for } \mathcal{T}} \wedge \underbrace{\text{ordered}([X_2 | L])}_{\text{resolution}}.$`

Query evaluation:

$d \wedge G \xRightarrow{\rho^k} c_1 \wedge \dots \wedge c_n \wedge \square$  (with  $c_1 \wedge \dots \wedge c_n$   $\mathcal{T}$ -satisfiable)  
(and  $\rho = \vartheta_1 \cdot \dots \cdot \vartheta_k$ )

1.  $d \wedge G = d \wedge A \wedge R$
2. find unifying head ( $A = H$ ) $\vartheta$
3. rewrite into  $(d \wedge c \wedge B \wedge R)\vartheta$

# CLP Evaluation for Test Generation

ordered([]).

ordered([X]).

ordered([X<sub>1</sub>, X<sub>2</sub> | L]) :-  $\underbrace{X_1 \leq X_2}_{\text{solver for } \mathcal{T}} \wedge \underbrace{\text{ordered}([X_2 | L])}_{\text{resolution}}.$

As a generator:

ordered(L).



L = []



L = [X]



L = [X<sub>1</sub>, X<sub>2</sub>] with  $X_1 \leq X_2$



...

L = [X<sub>1</sub>, X<sub>2</sub>, X<sub>3</sub>] with  $X_1 \leq X_2 \wedge X_2 \leq X_3$

# A CLP-based Encoding of Red-Black Trees

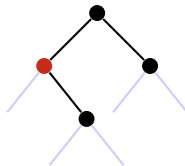
```
rbtree (T, MinSize, MaxSize, NumKeys) :-  
  % Preamble  
  MinSize#=<S, S#=<MaxSize, fd_labeling([S]),  
  length(Keys, S), length(Colors, S), Max#=NumKeys-1,  
  fd_domain(Keys, 0, Max), fd_domain(Colors, 0, 1),  
  % Symbolic Definition  
  lbt (T, S, Keys, [ ]),           % data structure shape  
  pi (T, D), ci (T, Colors, [ ]), % filters  
  ordered (T, 0, NumKeys),       % filters  
  % Instantiation  
  fd_labeling (Keys), fd_labeling (Colors).  
  
  lbt (T, S, Keys, [ ]) if T is labeled binary tree with S nodes  
    pi (T, D) if the tree T satisfies the path invariant  
    ci (T, Colors) if the tree T satisfies the color invariant  
ordered (T, 0, NumKeys) if the labels (keys) in T are ordered left-to-right  
  fd_labeling (X) if the variable X is instantiated to a feasible value
```

# On Specifying Invariants

KORAT

```
private boolean path_invariant(){
    int numberOfBlack = -1;
    workList = new java.util.LinkedList();
    workList.add(new Pair(root, 0));

    while (!workList.isEmpty()) {
        Pair p = (Pair) workList.removeFirst();
        Node e = p.e;
        int n = p.n;
        if (e != null && e.color == BLACK)
            n++;
        if (e == null) {
            if (numberOfBlack == -1)
                numberOfBlack = n;
            else if (numberOfBlack != n)
                return debug("failure");
        } else {
            workList.add(new Pair(e.left, n));
            workList.add(new Pair(e.right, n));
        }
    }
    return true;
}
```

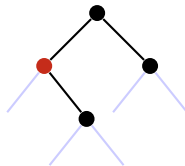


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numberOfBlack, workList

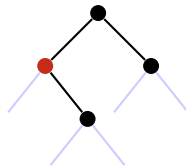
-1 (root,0)  
-1 (l,1), (r,1)  
-1 (e,1), (lr,1), (r,1)  
**1** (lr,1), (r,1)  
**1** (e,2), (e,2), (r,1)

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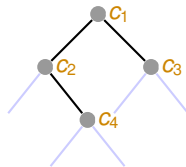
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1 ≠ 2 **fail**

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$C_1, \dots, C_4$  in  $\{0, 1\}$

1.  $\text{pi}(e, D) :- D \# = 0.$
2.  $\text{pi}(t(C, K, L, R), D) :- ND \# \geq 0, D \# = ND + C, \text{pi}(L, ND), \text{pi}(R, ND).$

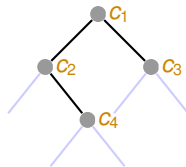
CLP



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$C_1, \dots, C_4$  in  $\{0, 1\}$

$$C_1 + C_2 = C_1 + C_2 + C_4 = C_1 + C_3$$

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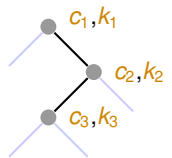
CLP

# Shape Rejection

Tree ::= e | **Color** × **Key** × Tree × Tree

with **Color** in {0, 1} (red, black)  
and **Key** in {0, ..., MaxKey}

**Size 3** (a possible solution):



structure shape

$\wedge$

$$\begin{array}{c} c_1 = c_1 + c_2 \\ \wedge \\ c_1 + c_2 = c_1 + c_2 + c_3 \end{array}$$

path invariant

$\wedge$

$$\begin{array}{c} c_1 + c_2 > 0 \\ \wedge \\ c_2 + c_3 > 0 \end{array}$$

color invariant

$\wedge$

$$\begin{array}{c} k_1 < k_2 \\ \wedge \\ k_1 < k_3 \\ \wedge \\ k_3 < k_2 \end{array}$$

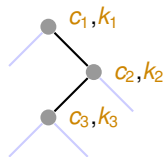
ordering

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path invariant

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color invariant

∧

$$\begin{array}{c} k_1 < k_2 \\ \wedge \\ k_1 < k_3 \\ \wedge \\ k_3 < k_2 \end{array}$$

ordering

lbt(T,S,Keys,[ ]) ∧

pi(T,D)

∧

ci(T,Colors)

∧

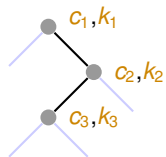
ordered(T,...)

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color invariant

$\wedge$

$$\begin{array}{c} k_1 < k_2 \\ \wedge \\ k_1 < k_3 \\ \wedge \\ k_3 < k_2 \end{array}$$

ordering

UNFEASIBLE

$$\text{lbt}(T, S, \text{Keys}, []) \wedge \text{pi}(T, D) \wedge \text{ci}(T, \text{Colors}) \wedge \text{ordered}(T, \dots)$$

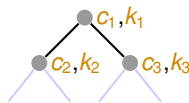
**No instantiation possible**  $\Rightarrow c_2 = 0 \wedge c_3 = 0$

# Shape Concretization

Tree ::= e | Color × Key × Tree × Tree

with Color in {0, 1} (red, black)  
and Key in {0, ..., MaxKey}

Size 3 (another solution) :



∧

$$c_1 + c_2 = c_1 + c_3$$

∧

$$c_1 + c_2 > 0$$

∧

$$c_1 + c_3 > 0$$

$$k_2 < k_1$$

∧

$$k_1 < k_3$$

FEASIBLE

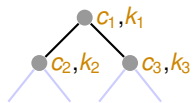
Instantiations:

# Shape Concretization

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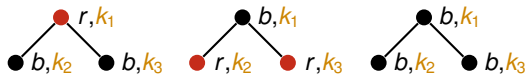
**Size 3** (another solution) :



$$\wedge \quad c_1 + c_2 = c_1 + c_3 \quad \wedge \quad \begin{array}{l} c_1 + c_2 > 0 \\ \wedge \\ c_1 + c_3 > 0 \end{array} \quad \wedge \quad \begin{array}{l} k_2 < k_1 \\ \wedge \\ k_1 < k_3 \end{array}$$

FEASIBLE

**Instantiations:**

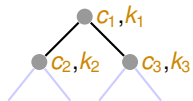


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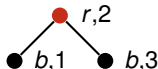
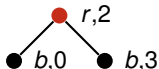
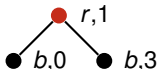
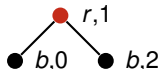
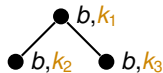
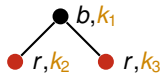
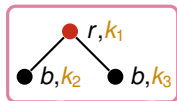
**Size 3** (another solution) :



$$\wedge \quad c_1 + c_2 = c_1 + c_3 \quad \wedge \quad \begin{matrix} c_1 + c_2 > 0 & k_2 < k_1 \\ \wedge & \wedge \\ c_1 + c_3 > 0 & k_1 < k_3 \end{matrix}$$

FEASIBLE

**Instantiations:**

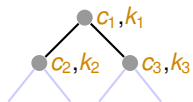


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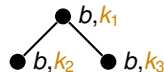
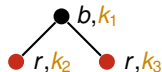
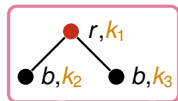


$$\wedge c_1 + c_2 = c_1 + c_3 \wedge$$

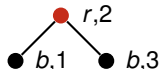
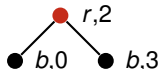
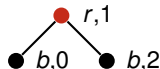
$$\begin{array}{cc} c_1 + c_2 > 0 & k_2 < k_1 \\ \wedge & \wedge \\ c_1 + c_3 > 0 & k_1 < k_3 \end{array}$$

FEASIBLE

**Instantiations:**



5 shapes  
1 feasible  
12 instantiations



...

we let the **solver** choose an instantiation order to **minimize backtracking**



size	<i>domain</i>	<i>explored</i>	<i>found</i>	time (s)
0	1	1	1	0.20
1	128	10	4	0.21
2	236196	52	6	0.22
3	2147483648	248	12	0.23
4	$6.10 * 10^{13}$	1521	40	0.30
5	$4.21 * 10^{18}$	6827	84	0.46
6	$6.01 * 10^{23}$	25727	140	0.58
7	$1.58 * 10^{29}$	109749	280	0.95
8	$7.12 * 10^{34}$	481398	576	1.91
9	$5.12 * 10^{40}$	2184788	1220	6.12
10	$5.61 * 10^{46}$	10693872	2860	30.01
11	$8.99 * 10^{52}$	54605852	7032	168.33
12	$2.04 * 10^{59}$	279610078	16848	878.13
13	$6.37 * 10^{65}$	1416658806	37912	5634.82

## KORAT

size	<i>domain</i>	symb.expl.	<i>feasible</i>	<i>found</i>	time (s)
0	1	1	1	1	0.01
1	4	1	1	4	0.01
2	72	2	2	6	0.01
3	2560	5	1	12	0.01
4	140000	14	4	40	0.01
5	10450944	42	6	84	0.01
6	993898752	132	12	140	0.01
7	$1.15 * 10^{11}$	429	29	280	0.01
8	$1.57 * 10^{14}$	1430	40	576	0.03
9	$2.49 * 10^{15}$	4862	46	1220	0.11
10	$4.46 * 10^{17}$	16796	76	2860	0.40
11	$8.94 * 10^{19}$	58786	190	7032	1.63
12	$1.98 * 10^{22}$	208012	456	16848	6.20
13	$4.83 * 10^{24}$	742900	904	37912	23.31

## Constrain-and-Generate

size	<i>domain</i>	<i>explored</i>	<i>found</i>	time (s)
0	1	1	1	0.20
1	128	10	4	0.21
2	236196	52	6	0.22
3	2147483648	248	12	0.23
4	$6.10 \times 10^{13}$	1521	40	0.30
5	$4.21 \times 10^{18}$	6827	84	0.46
6	$6.01 \times 10^{23}$	25727	140	0.58
7	$1.58 \times 10^{29}$	109749	280	0.95
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13	$6.37 \times 10^{65}$	1416658806	37912	5634.82

**KORAT**

Graphs

size	<i>domain</i>	symb.expl.	<i>feasible</i>	<i>found</i>	time (s)
0	1	1	1	1	0.01
1	4	1	1	4	0.01
2	72	2	2	6	0.01
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**Constrain-and-Generate**

Trees

size	<i>domain</i>	<i>explored</i>	<i>found</i>	time (s)
0	1	1	1	0.20
1	128	10	4	0.21
2	236196	52	6	0.22
3	2147483648	248	12	0.23
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**KORAT**

Graphs

size	<i>domain</i>	symp.expl.	<i>feasible</i>	<i>found</i>	time (s)
0	1	1	1	1	0.01
1	4	1	1	4	0.01
2	72	2	2	6	0.01
3	2560	5	1	12	0.01
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**Constrain-and-Generate**

Trees

Symbolic + instantiation

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## KORAT

Graphs

Pruning

(ratio ~6 oom)

size	<i>domain</i>	symp.expl.	<i>feasible</i>	<i>found</i>	time (s)
0	1	1	1	1	0.01
1	4	1	1	4	0.01
2	72	2	2	6	0.01
3	2560	5	1	12	0.01
4	140000	14	4	40	0.01
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## Constrain-and-Generate

Trees

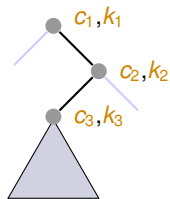
Symbolic + instantiation

Pruning symbolic

(ratio ~6.5 oom)

# Optimizing Generators by Transformation

Size 20



No way of placing the remaining  
17 nodes to build a feasible tree

$$c_1 = c_1 + c_2 \wedge c_1 = c_1 + c_2 + c_3 + X \wedge \\ X \geq 0 \wedge c_1 + c_2 > 0 \wedge c_2 + c_3 > 0$$

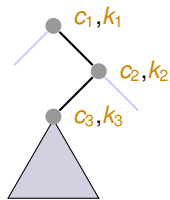
UNFEASIBLE

Yet, there would be **35357670 shapes**  
(and corresponding feasibility tests)

**Idea:** apply filter earlier

# Optimizing Generators by Transformation

Size 20



No way of placing the remaining 17 nodes to build a feasible tree

$$c_1 = c_1 + c_2 \wedge c_1 = c_1 + c_2 + c_3 + X \wedge X \geq 0 \wedge c_1 + c_2 > 0 \wedge c_2 + c_3 > 0$$

UNFEASIBLE

Yet, there would be **35357670 shapes** (and corresponding feasibility tests)

**Idea:** apply filter earlier

Program Transformation = **rewrite rules** + **strategies**

easy to prove correct

$$P_0 \quad \longmapsto \quad \dots \quad \longmapsto \quad P_n$$
$$M(P_0) \quad = \quad \dots \quad = \quad M(P_n)$$

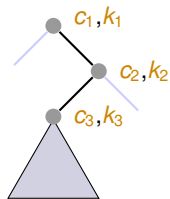
efficient

rule-based  
model-preserving  
(local) rewriting

A flexible deductive framework: applied to **Verification** & **SW Model Checking**

# Optimizing Generators by Transformation

Size 20



No way of placing the remaining  
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$$c_1 = c_1 + c_2 \wedge c_1 = c_1 + c_2 + c_3 + X \wedge \\ X \geq 0 \wedge c_1 + c_2 > 0 \wedge c_2 + c_3 > 0$$

UNFEASIBLE

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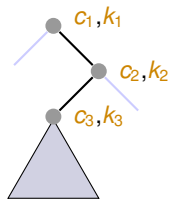
**Idea:** apply filter earlier

The **Synchronization** transformation strategy

- **filter promotion + tupling + folding** (local optimization inductively propagated)
- force the generator to have the desired behavior
- reduces don't care nondeterminism (less **failures**)

# Optimizing Generators by Transformation

Size 20



No way of placing the remaining  
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$$c_1 = c_1 + c_2 \wedge c_1 = c_1 + c_2 + c_3 + X \wedge \\ X \geq 0 \wedge c_1 + c_2 > 0 \wedge c_2 + c_3 > 0$$

UNFEASIBLE

Yet, there would be **35357670 shapes**  
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The **Synchronization** transformation strategy

- **filter promotion + tupling + folding** (local optimization inductively propagated)
- force the generator to have the desired behavior
- reduces don't care nondeterminism (less **failures**)

Related techniques : **compiling control**, **co-routining**



# The synch program

1. **sync** ( **e** , A, B, B, C, C, \_, \_, 0 ) :- A#=0.
2. **sync** ( **t**(A, B, **e**, **e**) , C, [B|D], D, [A|E], E, F, G, A ) :-  
C#=1, F#=<B, B#<G.
3. **sync** ( **t**(A, B, **e**, **t**(C, D, E, F)) , G, [B|H], I, [A|J], K, L, M, A ) :-  
G#>=2, O#=G-1, A+C#>0, L#=<B, B#<M, P#=B+1,  
**sync**(**t**(C, D, E, F), O, H, I, J, K, P, M, 0) .
4. **sync** ( **t**(A, F, **t**(B, C, D, E), **e**) , G, [F|H], I, [A|J], K, L, M, A ) :-  
G#>=2, O#=G-1, A+B#>0, L#=<F, F#<M,  
**sync**(**t**(B, C, D, E), O, H, I, J, K, L, F, 0) .
5. **sync** ( **t**(A, F, **t**(B, C, D, E), **t**(G, H, I, J)) , K, [F|L], M, [A|N], O, P, Q, R ) :-  
K#>=3, S#=K-1, T#>0, U#>0, A+B#>0, A+G#>0, V#>=0,  
R#=V+A, P#=<F, F#<Q, W#=F+1, fdsum(S, T, U),  
**sync**(**t**(B, C, D, E), T, L, X, N, Y, P, F, V),  
**sync**(**t**(G, H, I, J), U, X, M, Y, O, W, Q, V) .

The new generator intertwines **tree construction** and **invariant construction**

# Experiments

Size	<i>RB Trees</i>	Time		
		Original	Synchronized	Korat
6	20	0	0	0.47
7	35	0.01	0	0.63
8	64	0.02	0	1.49
9	122	0.08	0	4.51
10	260	0.29	0.01	21.14
11	586	1.07	0.03	116.17
12	1296	3.98	0.06	-
13	2708	14.85	0.12	-
14	5400	55.77	0.26	-
15	10468	-	0.55	-
...	...	...	...	...
20	279264	-	25.90	-

Size = number of nodes

RB Trees = number of structures found

Time = (in seconds) for generating all the structures

Zero means less than 10 ms and (-) means more than 200 seconds.

# Experiments (cont.)

Size	Sorted Lists	Time	
		C&G	Korat
8	6435	0.00	0.61
9	24310	0.00	1.08
10	92378	0.02	1.83
11	352716	0.09	6.37
12	1352078	0.36	24.95
13	5200300	1.40	125.73
14	20058300	5.40	-
15	77558760	21.16	-
16	300540195	82.22	-

Size	Heaparrays	Time	
		C&G	Korat
6	13139	0.00	0.30
7	117562	0.03	0.86
8	1005075	0.17	3.41
9	10391382	1.66	34.103
10	111511015	17.57	-

Size	Search Trees	Time	
		C&G	Korat
7	429	0.01	0.87
8	1430	0.02	4.43
9	4862	0.08	33.99
10	16796	0.28	-
11	58786	1.11	-
12	208012	4.43	-
13	742900	17.68	-
14	2674440	70.75	-

Size	Disjoint Sets	Time	
		C&G	Korat
6	203	0.00	1.60
7	877	0.00	37.10
8	4140	0.01	-
9	21147	0.10	-
10	115975	0.61	-
11	678570	3.90	-
12	4213597	26.63	-
13	27644437	189.42	-

**No optimizations**

# From CLP Terms to Java Objects

A solution using Java (de)serialization:

1. generate XML encodings of (found) objects
2. XML description  $\rightarrow$  Java object  
by using libraries such as XStream/Simple

XML  $\longleftrightarrow$  Java-objects given at specification-time, inductively

Triggered only on structures of actual interest for testing,  
with negligible overhead

## Work in Progress

- Experiments with graph-like structures
- Experiments with ASP solvers
- Full automation of the optimization algorithm
- Automatic extraction of CLP generators from contracts
- Testing web-based applications