

# Type-Targeted Testing

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```
data Tree
  = Leaf
  | Node Int Tree Tree
```

```
insert :: Int -> Tree -> Tree
delete :: Int -> Tree -> Tree
```

```
data Tree
  = Leaf
  | Node Int Tree Tree

insert :: Int -> Tree -> Tree
delete :: Int -> Tree -> Tree
```

Did I get it "right"?

# An Oracle for Testing

```
isBST t = case t of
  Leaf      -> True
  Node x l r -> abs (height l - height r) <= 1
              && all (< x) l && all (> x) r
              && isBST l      && isBST r
```

# QuickCheck

```
prop_insert_bst x t
  = isBST t ==> isBST (insert x t)
```

```
>>> quickCheck prop_insert_bst
+++ OK, passed 100 tests.
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```

How is this possible?

Valid trees are a **sparse subset** of all trees!

# QuickCheck With Statistics

```
prop_insert_bst x t
  = isBST t
    ==> collect (size t)
              (isBST (insert x t))
```

```
>>> quickCheck prop_insert_bst
+++ OK, passed 100 tests:
73% 0
21% 1
 6% 2
```

# QuickCheck With Statistics

```
prop_insert_bst x t
  = isBST t
    ==> collect (size t)
              (isBST (insert x t))
```

```
>>> quickCheck prop_insert_bst
+++ OK, passed 100 tests:
73% 0
21% 1
 6% 2
```

73% of trees were **empty** and  
21% only had one element



# QuickCheck: Non-Trivial Trees

```
prop_insert_bst x t
  = isBST t && size t > 1
  ==> isBST (insert x t)
```

```
>>> quickCheck prop_insert_bst
*** Gave up! Passed only 37 tests.
```

# QuickCheck: Non-Trivial Trees

```
prop_insert_bst x t
  = isBST t && size t > 1
  ==> isBST (insert x t)
```

```
>>> quickCheck prop_insert_bst
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```

Less than 10% of generated trees were valid

# Custom Generators

```
newtype BST = BST Tree
```

```
instance Arbitrary BST where  
  arbitrary = ...
```

```
prop_insert_bst x (BST t) = ...
```

# Custom Generators

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newtype BST = BST Tree
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```
instance Arbitrary BST where  
  arbitrary = ...
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```
prop_insert_bst x (BST t) = ...
```

Must repeat for any further restrictions on  
input domain!

# SmallCheck

```
prop_insert_bst x t
  = isBST t ==> isBST (insert x t)
```

```
>>> smallCheck 3 prop_insert_bst
Completed 567 tests without failure.
But 434 did not meet ==> condition.
```

# SmallCheck

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prop_insert_bst x t
  = isBST t ==> isBST (insert x t)
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```
>>> smallCheck 3 prop_insert_bst
Completed 567 tests without failure.
But 434 did not meet ==> condition.
```

Only **133** input trees were valid!

# SmallCheck: How Small?

```
prop_insert_bst x t
  = isBST t ==> isBST (insert x t)
```

```
>>> smallCheck 4 prop_insert_bst
.....
.....
```

# SmallCheck: How Small?

```
prop_insert_bst x t
  = isBST t ==> isBST (insert x t)
```

```
>>> smallCheck 4 prop_insert_bst
.....
.....
```

Exponential blowup in input space  
confines search to **very small** inputs



How can we **systematically**  
generate only **valid** inputs?

# Target

Generates tests from **refinement types** via query-decode-check loop

1. Translate input types into SMT **query**
2. **Decode** SMT model into concrete values
3. Run function and **check** that result inhabits output type

# Refinement Types

$$\{v : t \mid p\}$$

The set of values  $v$  of type  $t$  satisfying a predicate  $p$

# Refinement Types

```
type Nat      = {v:Int | 0 <= v}
type Pos      = {v:Int | 0 <  v}
type Rng N    = {v:Int | 0 <= v && v < N}
```

The natural numbers, positive integers,  
and integers in a range

# Refinement Types

$x : \text{Nat} \rightarrow \{v : \text{Nat} \mid v = x + 1\}$

Functions that take a natural number  
and increment it by one

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# Step 1: Query

```
type Nat    = {v:Int | 0 <= v}
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```
rescale :: r1:Nat -> r2:Nat -> s:Rng r1 -> Rng r2
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Represent preconditions directly in logic



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Represent preconditions directly in logic

$$C_0 \doteq 0 \leq r_1 \wedge 0 \leq r_2 \wedge 0 \leq s < r_1$$

# Step 2: Decode

```
type Nat      = {v:Int | 0 <= v}
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rescale :: r1:Nat -> r2:Nat -> s:Rng r1 -> Rng r2
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Represent preconditions directly in logic

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A model  $[r_1 \mapsto 1, r_2 \mapsto 1, s \mapsto 0]$  maps to a concrete test case

```
>>> rescale 1 1 0
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# Step 3: Check

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type Nat    = {v:Int | 0 <= v}
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Postcondition is:  $\{v:Int \mid 0 \leq v \ \&\& \ v < r_2\}$

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After substituting  $v$  and  $r_2$ :  $0 \leq 0 \ \wedge \ 0 < 1$       **VALID**

Force new test by adding refutation constraint  $\neg(r_1 = 1 \wedge r_2 = 1 \wedge s = 0)$

# Repeat With New Test

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type Nat    = {v:Int | 0 <= v}
```

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type Rng N = {v:Int | 0 <= v && v < N}
```

```
rescale :: r1:Nat -> r2:Nat -> s:Rng r1 -> Rng r2
```

Represent preconditions directly in logic, excluding 1<sup>st</sup> test

$$C_1 \doteq 0 \leq r_1 \wedge 0 \leq r_2 \wedge 0 \leq s < r_1 \wedge \neg(r_1 = 1 \wedge r_2 = 1 \wedge s = 0)$$

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After substituting  $v$  and  $r2$ :  $0 \leq 0 \ \wedge \ 0 < 0$  **INVALID**

`rescale 1 0 0` fails the postcondition check!

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How should we handle **structured data**?

# Containers

```
type Weight = Pos
type Score  = Rng 100

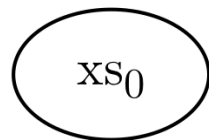
average :: [(Weight, Score)] -> Score
```

How to generate lists via SMT solver?



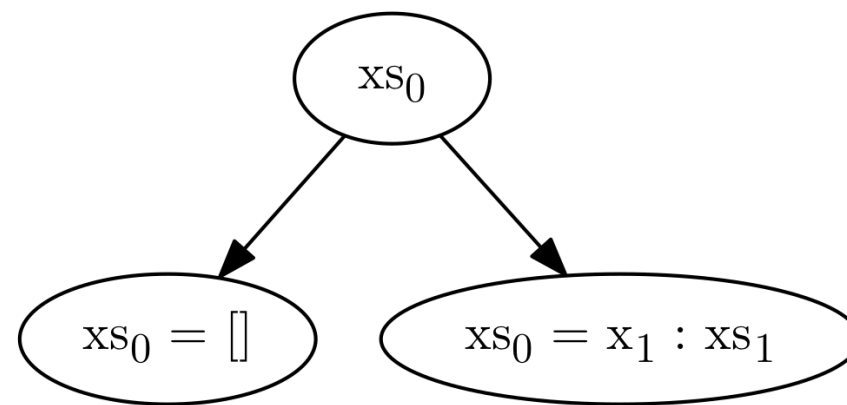
# Containers: Query

A **single** set of constraints describes **all possible** inputs



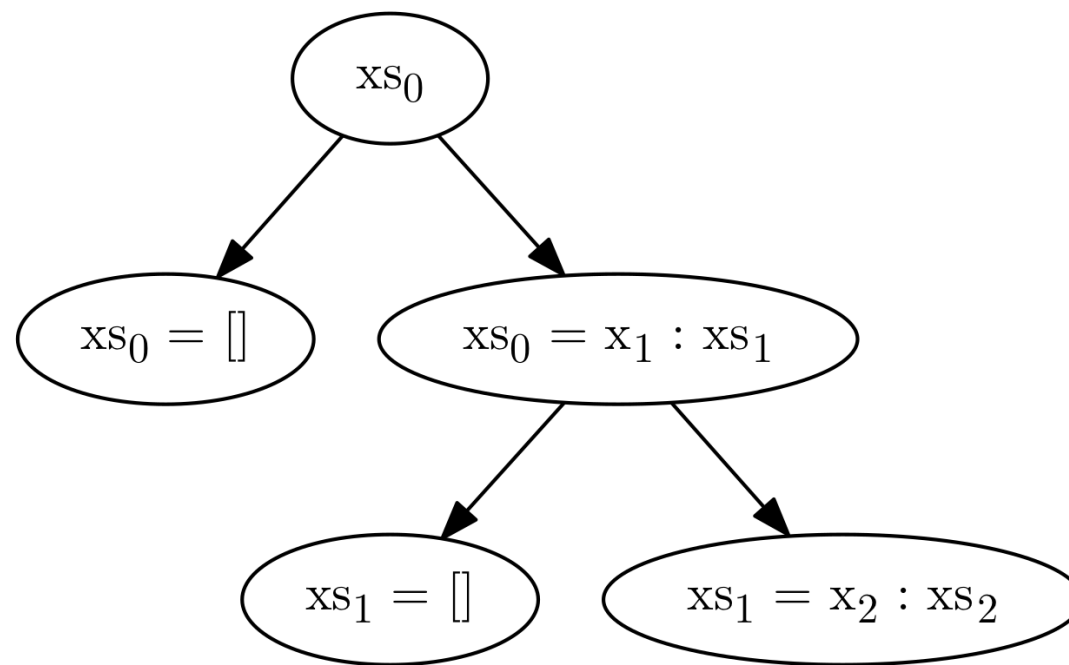
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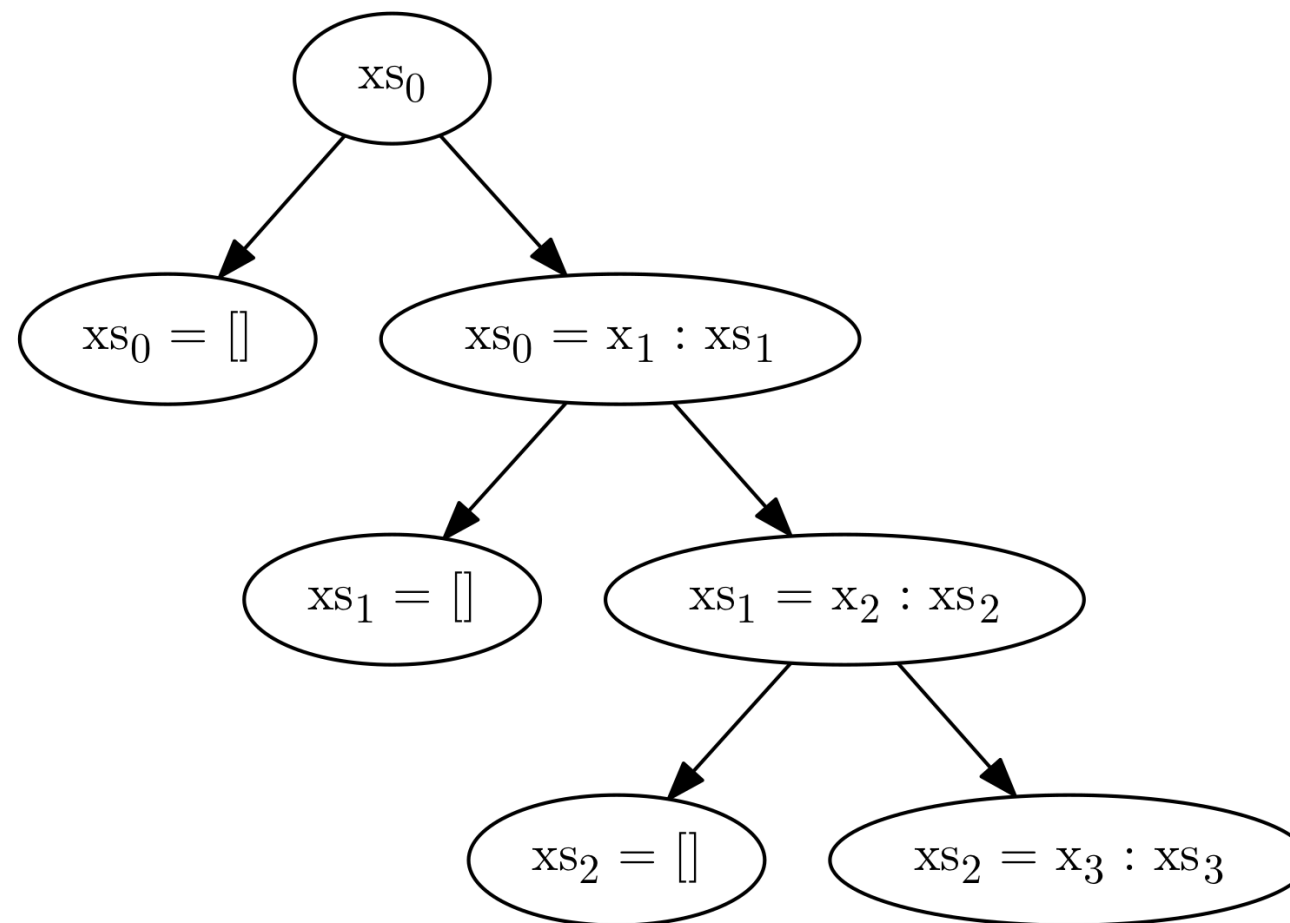
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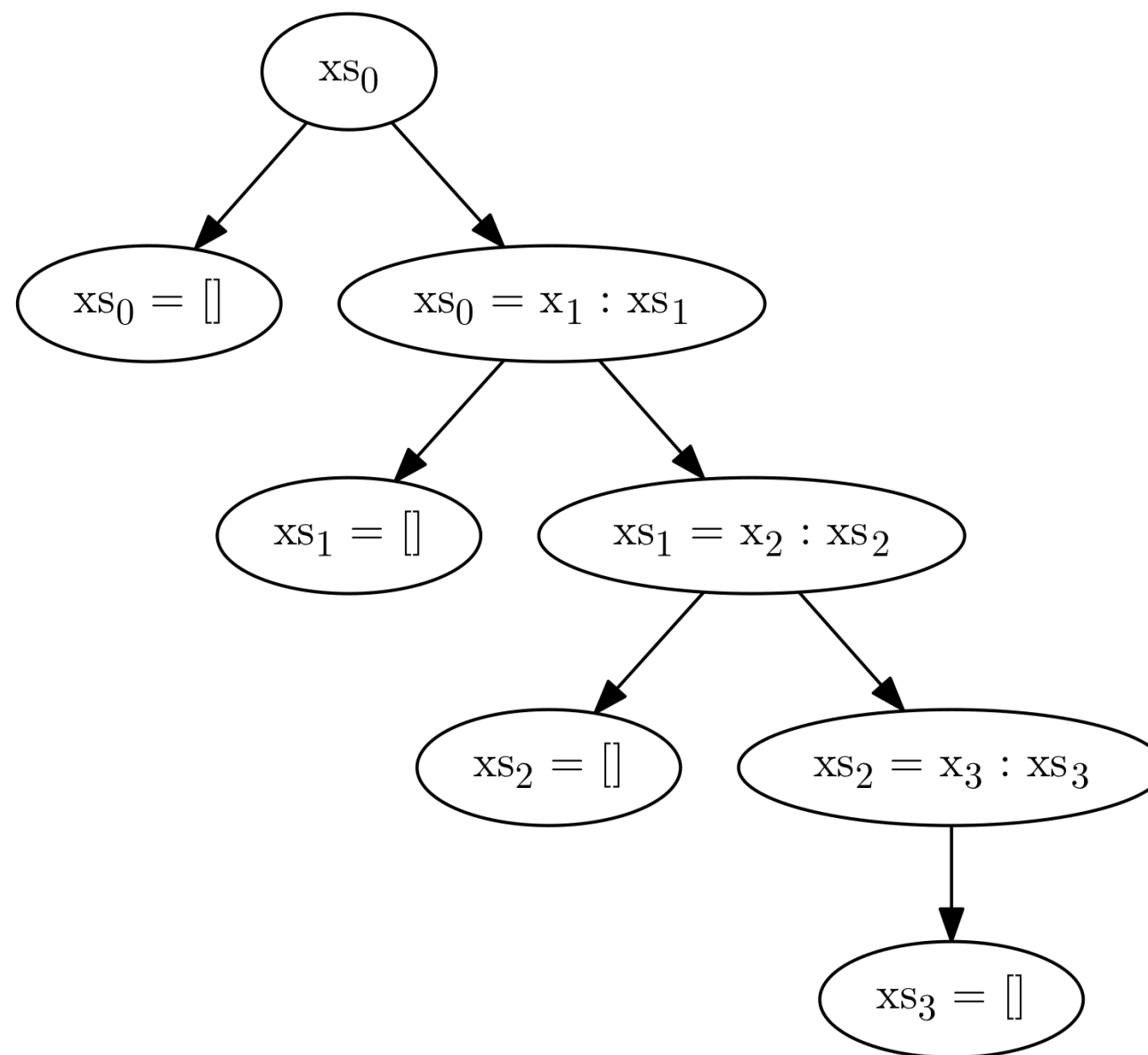
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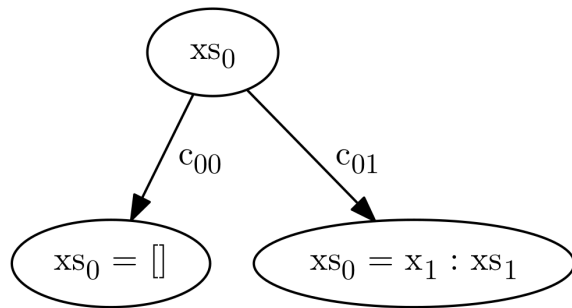
$xs_0$

Choice variables  $c$  **guard** other constraints

$C_{\text{list}} \doteq$

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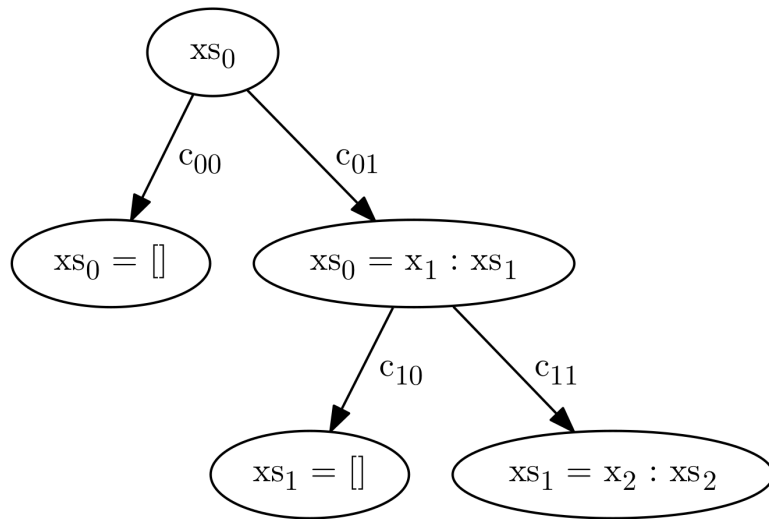


Choice variables  $c$  **guard** other constraints

$$C_{\text{list}} \doteq (c_{00} \Rightarrow xs_0 = []) \wedge (c_{01} \Rightarrow xs_0 = x_1 : xs_1) \wedge (c_{00} \oplus c_{01})$$

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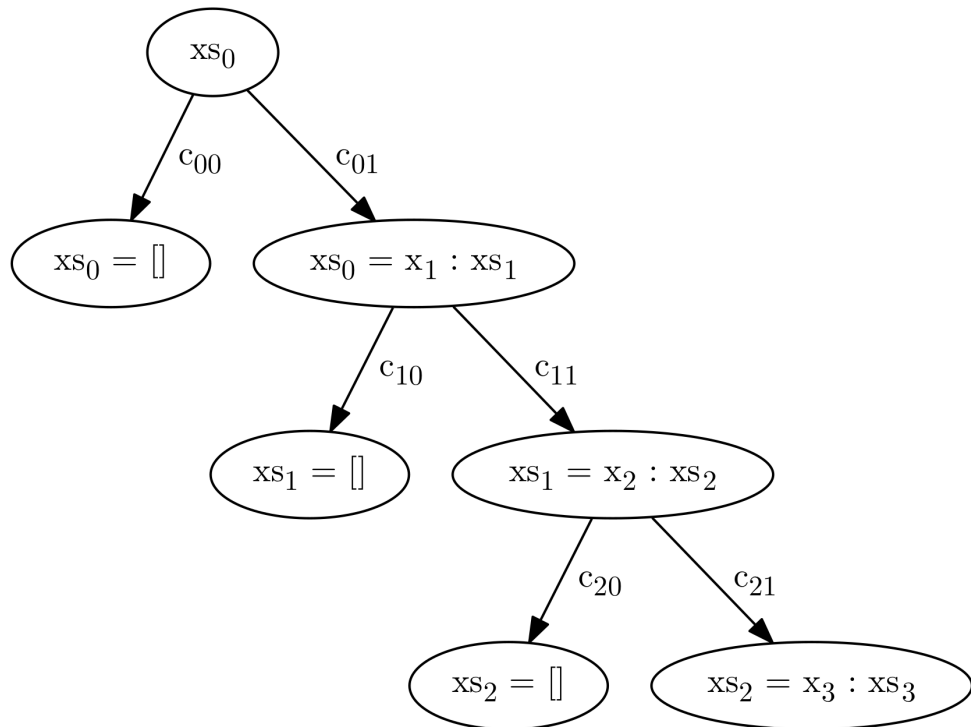
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$$\begin{aligned} C_{\text{list}} \doteq & (c_{00} \Rightarrow xs_0 = []) \wedge (c_{01} \Rightarrow xs_0 = x_1 : xs_1) \wedge (c_{00} \oplus c_{01}) \\ & \wedge (c_{10} \Rightarrow xs_1 = []) \wedge (c_{11} \Rightarrow xs_1 = x_2 : xs_2) \wedge (c_{01} \Rightarrow c_{10} \oplus c_{11}) \end{aligned}$$



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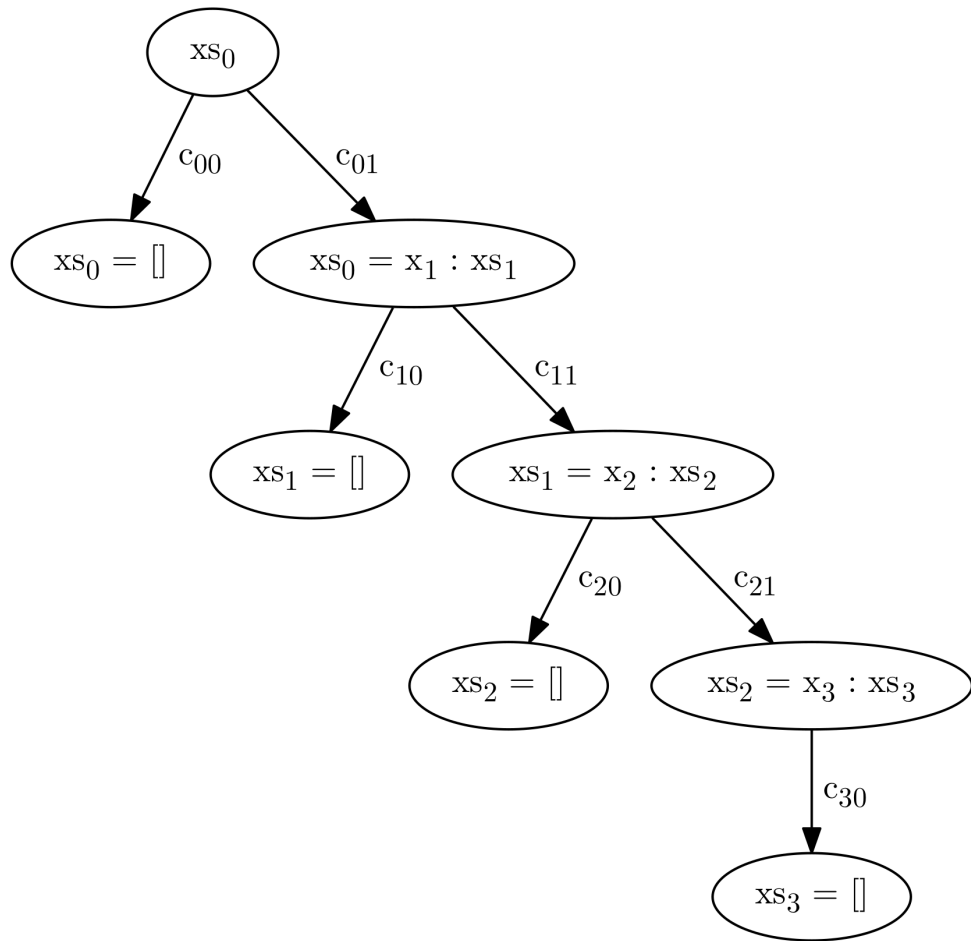


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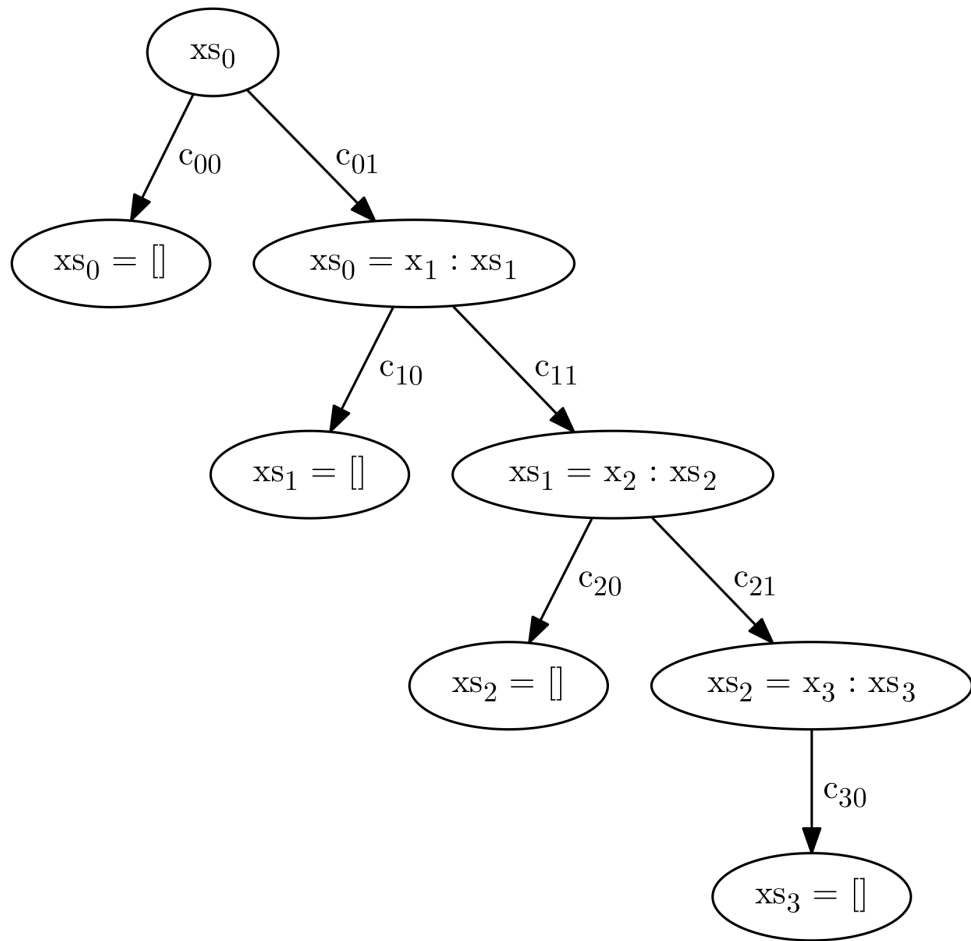


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 & \wedge (c_{30} \Rightarrow xs_3 = []) \wedge (c_{21} \Rightarrow c_{30})
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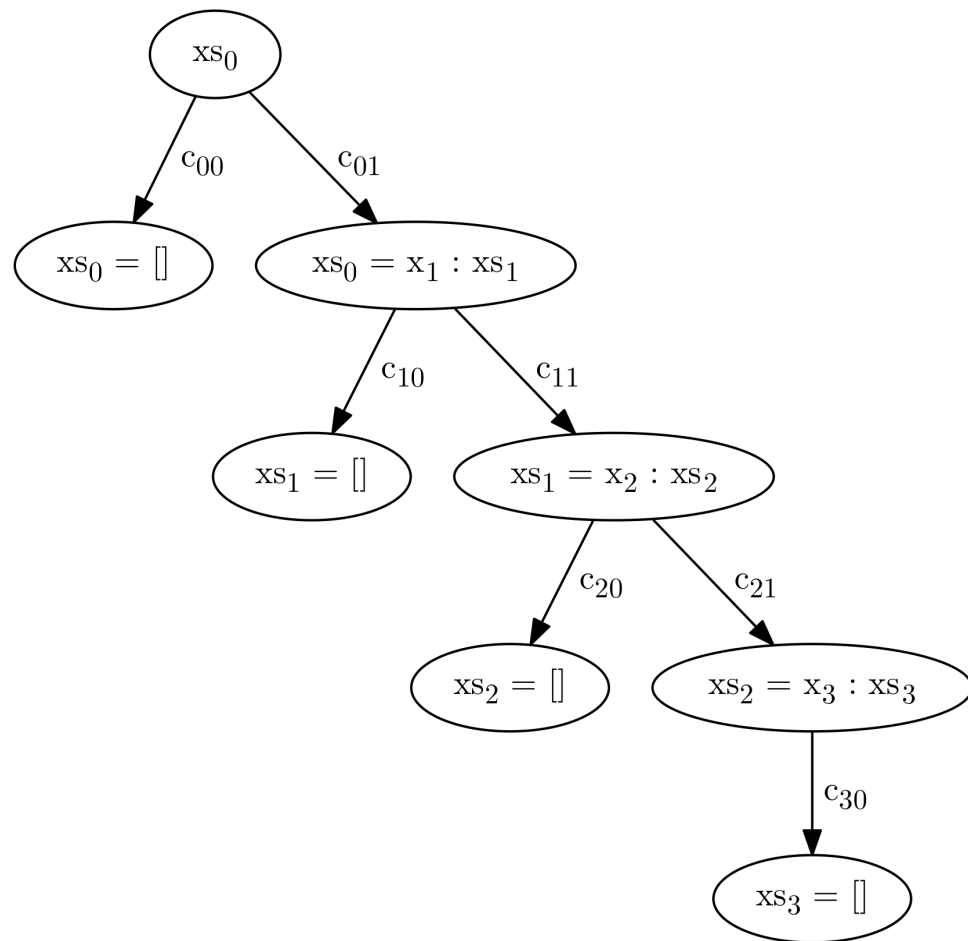
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 \end{aligned}$$

$$\begin{aligned}
 C_{\text{data}} \doteq & (c_{01} \Rightarrow x_1 = (w_1, s_1) \wedge 0 < w_1 \wedge 0 \leq s_1 < 100) \\
 & \wedge (c_{11} \Rightarrow x_2 = (w_2, s_2) \wedge 0 < w_2 \wedge 0 \leq s_2 < 100) \\
 & \wedge (c_{21} \Rightarrow x_3 = (w_3, s_3) \wedge 0 < w_3 \wedge 0 \leq s_3 < 100)
 \end{aligned}$$

Full constraint  $C \doteq C_{\text{list}} \wedge C_{\text{data}}$

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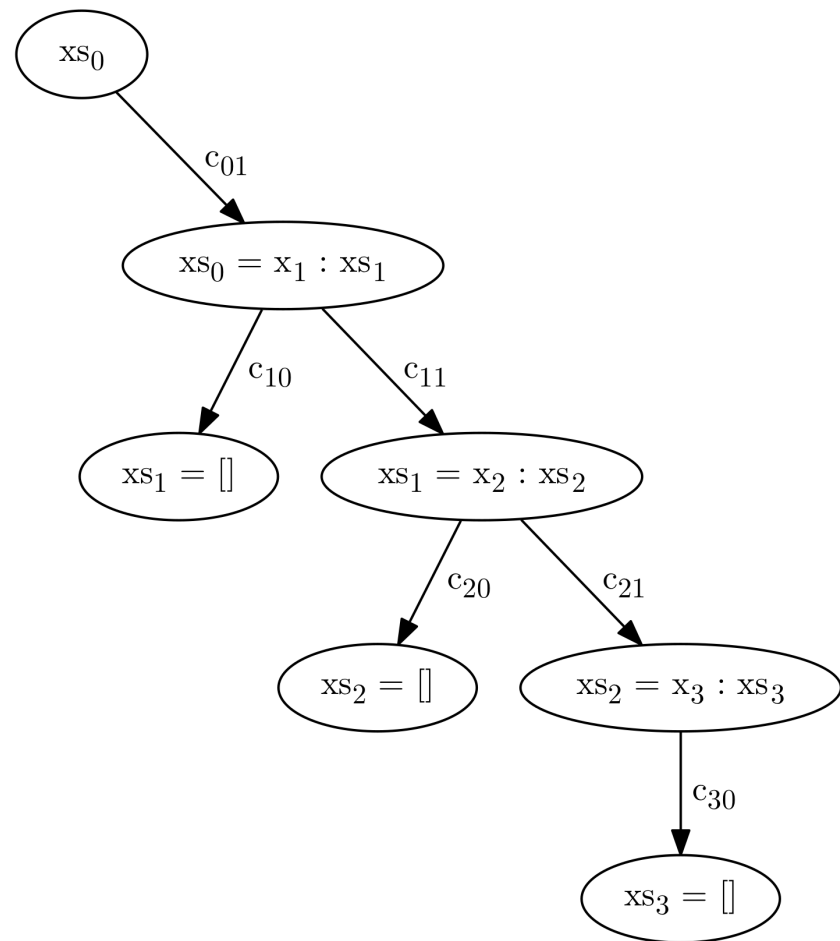


Follow choice variables to reconstruct the list

[  $c_{00} \mapsto \text{false}$ ,  $c_{01} \mapsto \text{true}$ ,  $x_1 \mapsto (w_1, s_1)$ ,  $w_1 \mapsto 1$ ,  $s_1 \mapsto 2$ ,  
 $c_{10} \mapsto \text{true}$ ,  $c_{11} \mapsto \text{false}$ ,  $x_2 \mapsto (w_2, s_2)$ ,  $w_2 \mapsto 3$ ,  $s_2 \mapsto 4, \dots$  ]

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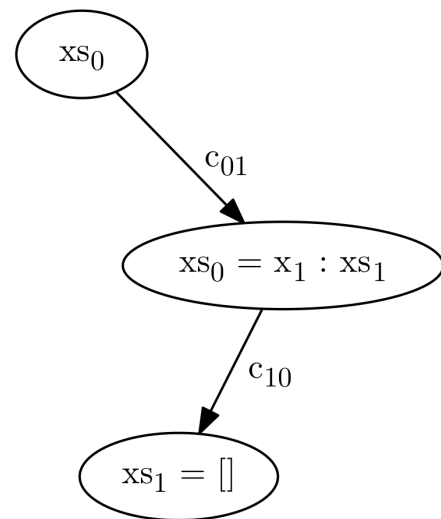
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$c_{01} \mapsto \text{true} \Rightarrow xs_0 = x_1 : xs_1$

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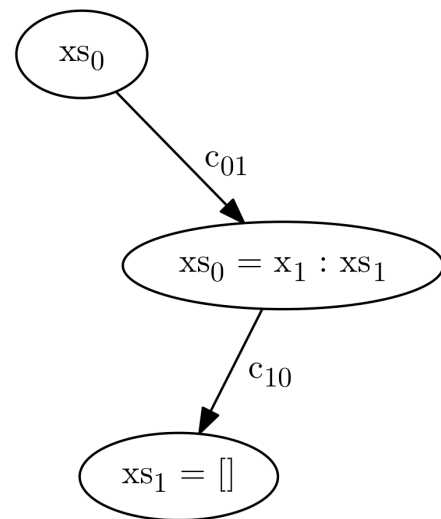
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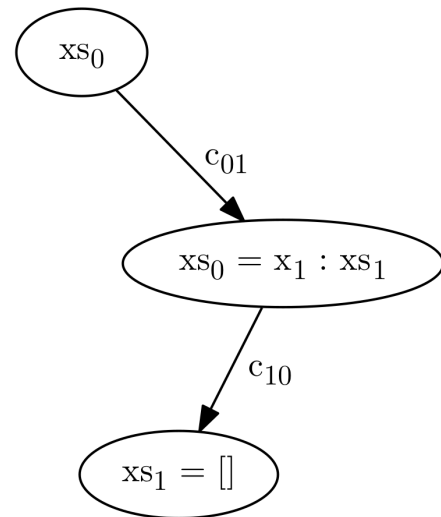
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Realized value: [ (1, 2) ]

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$c_{01} \mapsto \text{true} \Rightarrow xs_0 = x_1 : xs_1$

$c_{10} \mapsto \text{true} \Rightarrow xs_1 = []$

Realized value: [ (1, 2) ]

**Only** refute constraints that contribute to **realized** value

$\neg(c_{00} = \text{false} \wedge c_{01} = \text{true} \wedge x_1 = (w_1, s_1) \wedge w_1 = 1 \wedge s_1 = 2 \wedge c_{10} = \text{true})$



# Structured Containers

```
best :: k:Nat -> {xs:[Score] | k <= len xs}  
      -> {v:[Score] | k = len v}
```

**best** takes a list of **at least** **k** scores,  
and returns a list with **exactly** **k** scores

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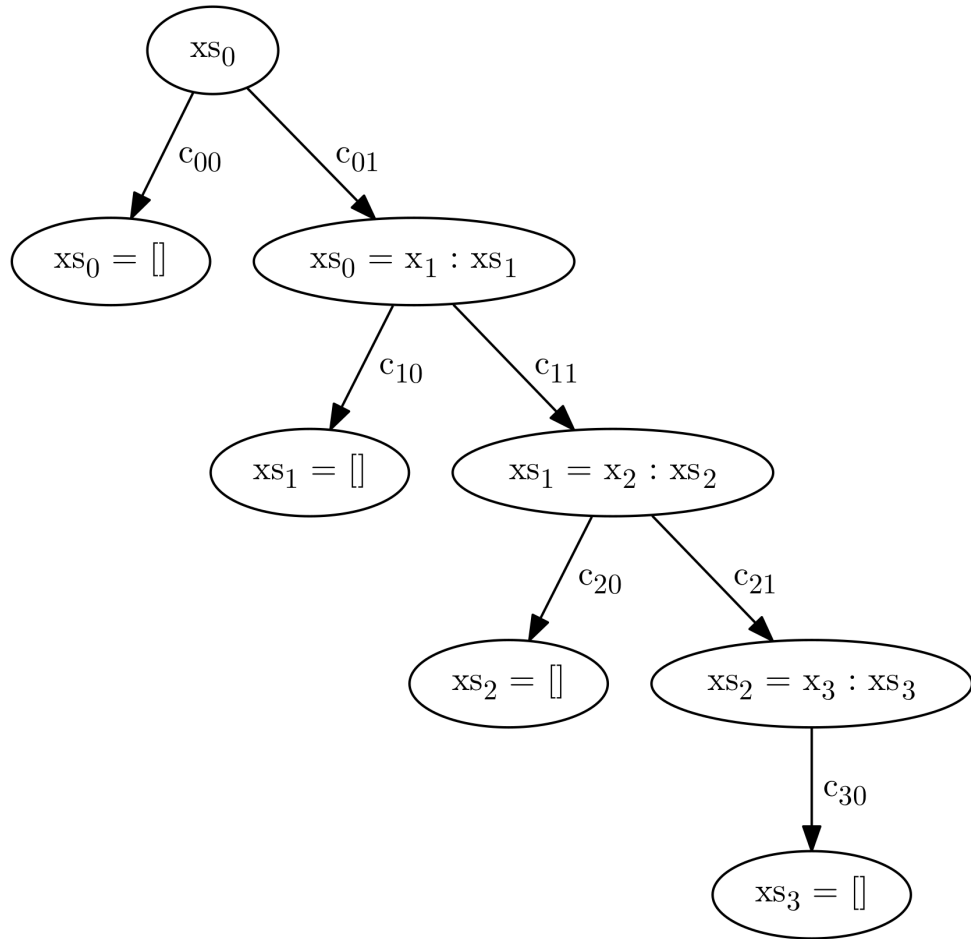
```
measure len :: [a] -> Nat  
len []      = 0  
len (x:xs)  = 1 + len xs
```

**len** is a **logical function** that describes the length of a list.

We instantiate measure definition each time we unfold **[]** or **(:)**

# Structured Containers

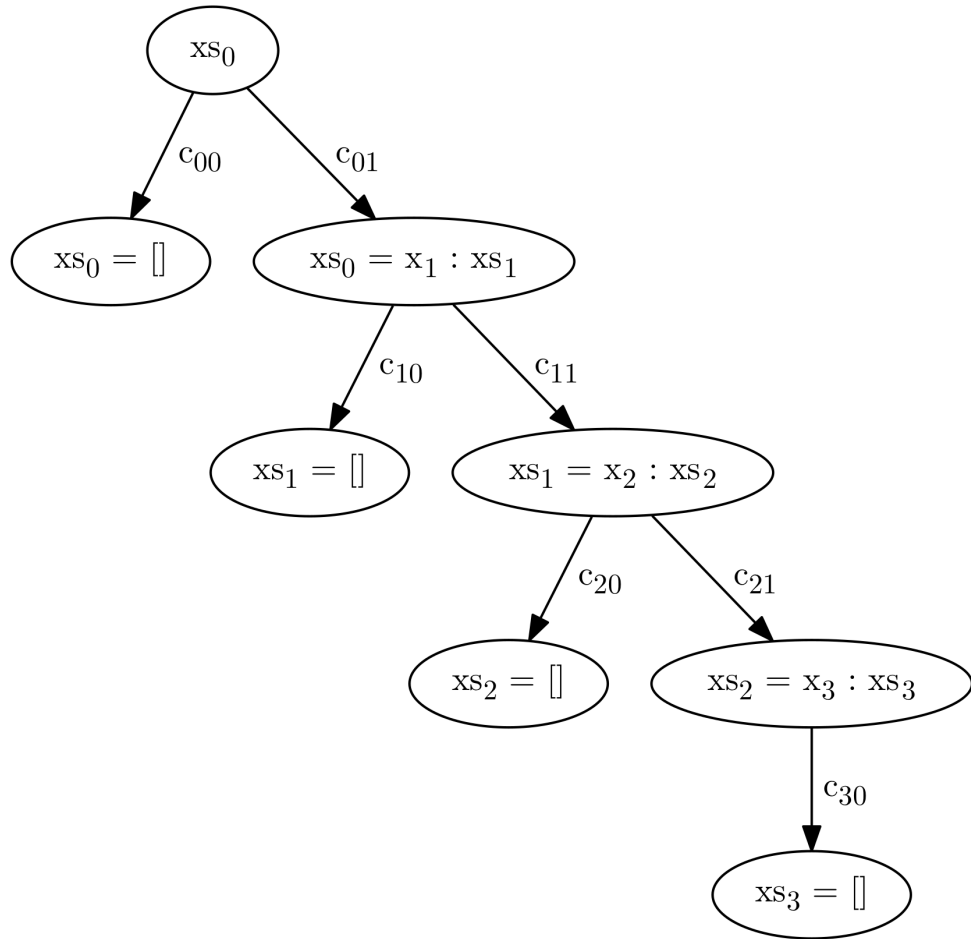
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best :: k:Nat -> {xs:[Score] | k <= len xs}
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```



$$\begin{aligned}
 C_{\text{list}} \doteq & (c_{00} \Rightarrow xs_0 = []) \wedge (c_{01} \Rightarrow xs_0 = x_1 : xs_1) \wedge (c_{00} \oplus c_{01}) \\
 & \wedge (c_{10} \Rightarrow xs_1 = []) \wedge (c_{11} \Rightarrow xs_1 = x_2 : xs_2) \wedge (c_{01} \Rightarrow c_{10} \oplus c_{11}) \\
 & \wedge (c_{20} \Rightarrow xs_2 = []) \wedge (c_{21} \Rightarrow xs_2 = x_3 : xs_3) \wedge (c_{11} \Rightarrow c_{20} \oplus c_{21}) \\
 & \wedge (c_{30} \Rightarrow xs_3 = []) \wedge (c_{21} \Rightarrow c_{30})
 \end{aligned}$$

# Structured Containers

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best :: k:Nat -> {xs:[Score] | k <= len xs}
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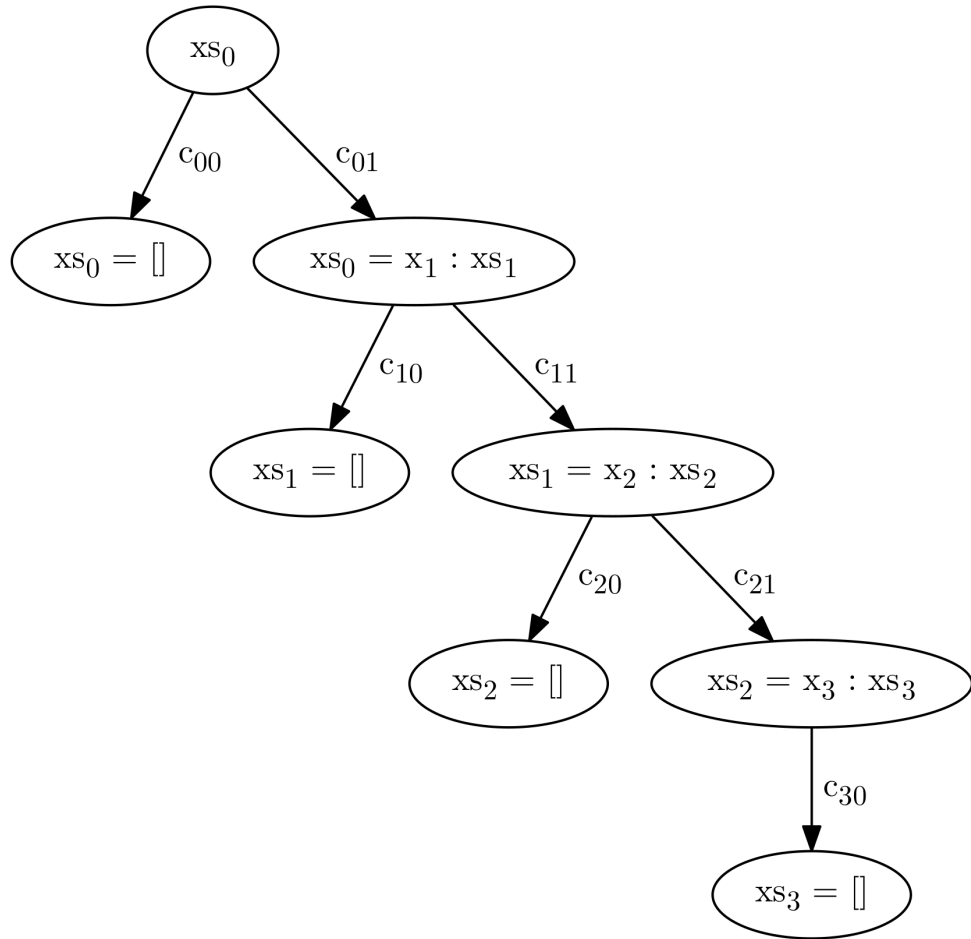


$$\begin{aligned}
 C_{\text{list}} \doteq & (c_{00} \Rightarrow xs_0 = []) \wedge (c_{01} \Rightarrow xs_0 = x_1 : xs_1) \wedge (c_{00} \oplus c_{01}) \\
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 \end{aligned}$$

$$\begin{aligned}
 C_{\text{size}} \doteq & (c_{00} \Rightarrow \text{len } xs_0 = 0) \wedge (c_{01} \Rightarrow \text{len } xs_0 = 1 + \text{len } xs_1) \\
 & \wedge (c_{10} \Rightarrow \text{len } xs_1 = 0) \wedge (c_{11} \Rightarrow \text{len } xs_1 = 1 + \text{len } xs_2) \\
 & \wedge (c_{20} \Rightarrow \text{len } xs_2 = 0) \wedge (c_{21} \Rightarrow \text{len } xs_2 = 1 + \text{len } xs_3) \\
 & \wedge (c_{30} \Rightarrow \text{len } xs_3 = 0)
 \end{aligned}$$

# Structured Containers

```
best :: k:Nat -> {xs:[Score] | k <= len xs}
      -> {v:[Score] | k = len v}
```



$$C_{\text{list}} \doteq (c_{00} \Rightarrow xs_0 = []) \wedge (c_{01} \Rightarrow xs_0 = x_1 : xs_1) \wedge (c_{00} \oplus c_{01}) \\ \wedge (c_{10} \Rightarrow xs_1 = []) \wedge (c_{11} \Rightarrow xs_1 = x_2 : xs_2) \wedge (c_{01} \Rightarrow c_{10} \oplus c_{11}) \\ \wedge (c_{20} \Rightarrow xs_2 = []) \wedge (c_{21} \Rightarrow xs_2 = x_3 : xs_3) \wedge (c_{11} \Rightarrow c_{20} \oplus c_{21}) \\ \wedge (c_{30} \Rightarrow xs_3 = []) \wedge (c_{21} \Rightarrow c_{30})$$

$$C_{\text{size}} \doteq (c_{00} \Rightarrow \text{len } xs_0 = 0) \wedge (c_{01} \Rightarrow \text{len } xs_0 = 1 + \text{len } xs_1) \\ \wedge (c_{10} \Rightarrow \text{len } xs_1 = 0) \wedge (c_{11} \Rightarrow \text{len } xs_1 = 1 + \text{len } xs_2) \\ \wedge (c_{20} \Rightarrow \text{len } xs_2 = 0) \wedge (c_{21} \Rightarrow \text{len } xs_2 = 1 + \text{len } xs_3) \\ \wedge (c_{30} \Rightarrow \text{len } xs_3 = 0)$$

Enforce relation between **k** and **xs** by adding  $k \leq \text{len } xs_0$

$$C \doteq C_{\text{list}} \wedge C_{\text{data}} \wedge C_{\text{size}} \wedge 0 \leq k \leq \text{len } xs_0$$

# Demo: Targeting **BSTs**

# Evaluation

## Our Claims

1. Target handles highly structured inputs automatically
2. Target generates tests that provide high code coverage

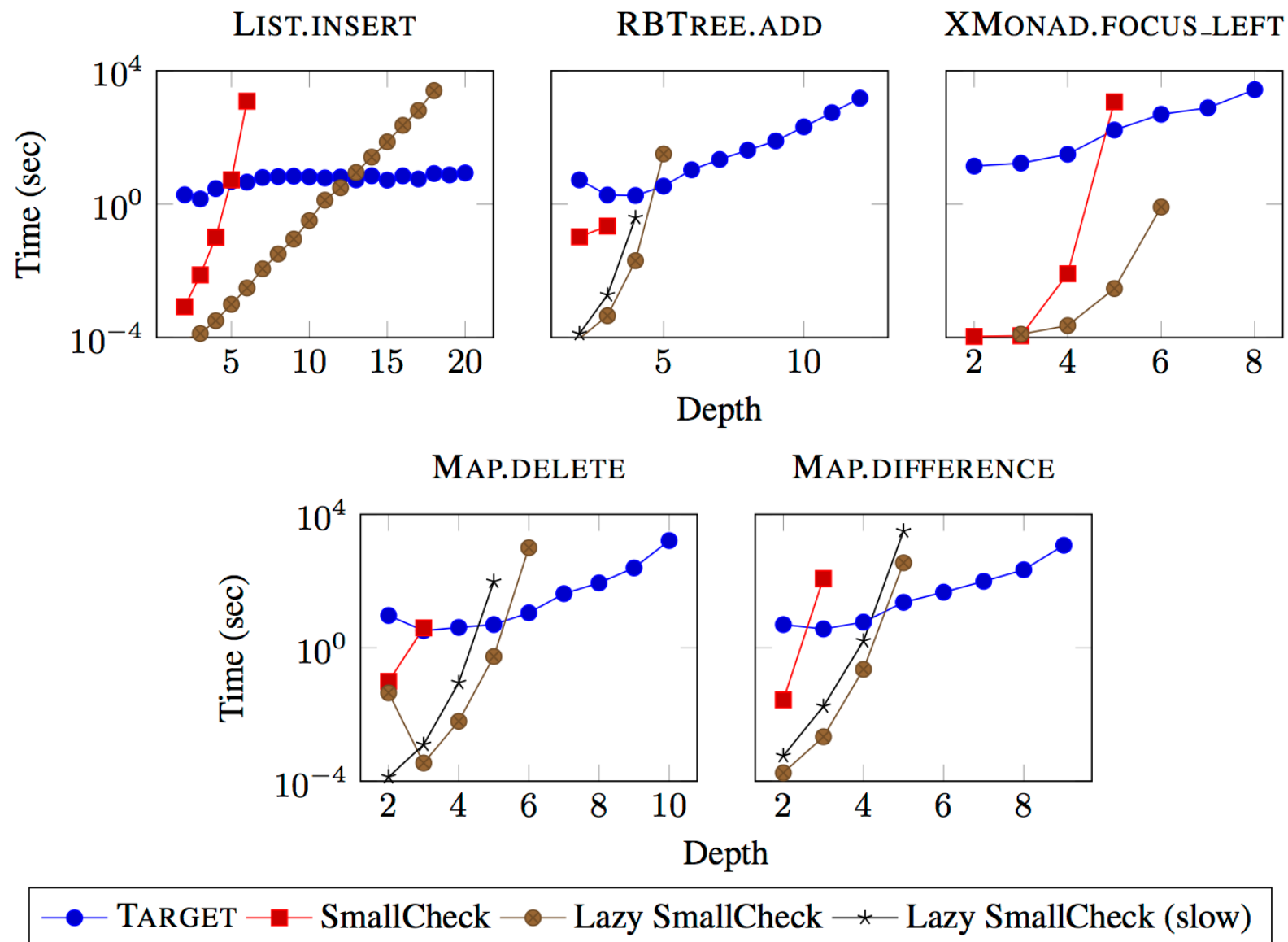
## Benchmarks

1. **Data.Map**: checked balancing and ordering invariants
2. **RBTree**: checked red-black and ordering invariants
3. **XMonad.StackSet**: checked uniqueness of windows

Compared Target against QuickCheck and SmallCheck

# Evaluation: Results

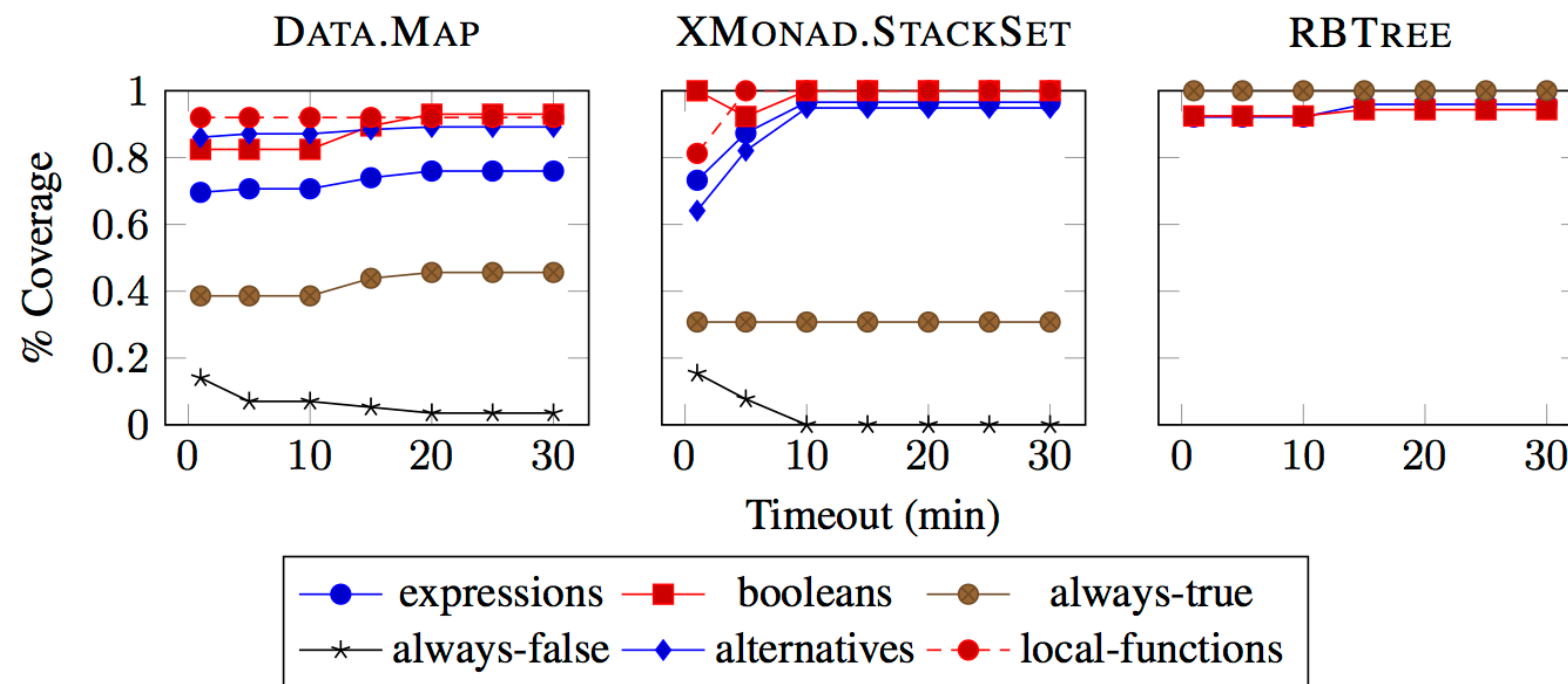
Target checks larger inputs than brute-force





# Evaluation: Results

Target provides high coverage with low investment



# Takeaway

**Target** - a new approach for automatically testing functions with preconditions.

- **guarantees** inputs satisfy preconditions
- vs **QuickCheck**: does not require custom generators
- vs **SmallCheck**: defers the onset of input explosion