

A Scalable Segmented Decision Tree Abstract Domain

Dedicated to Amir

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joint work with Radhia Cousot and Laurent Mauborgne

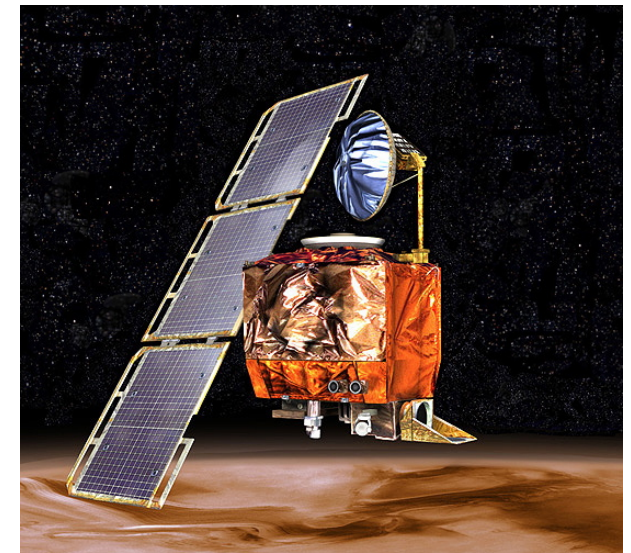
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Motivation

Computer scientists have made great contributions to the failure of complex systems



Ariane 5.01 failure Patriot failure Mars orbiter loss
(overflow) (float rounding) (unit error)

- On-board checking the **presence** of bugs is great!
- Proving their **absence** automatically by static analysis is even better!!!

Static analysis

- Automatic static analysis is extremely easy, but for several serious problems:
 - Formally defining the **semantics** of programming languages and machines
 - Minimizing **efforts of developers and end-users**
 - **Scaling up** with enough **precision**

Making static analysis very easy

- Choose a **simple semantic model** (e.g. transition systems)
- Choose a **uniform representation of properties** (e.g. *terms* in deductive methods, *BDDs* in model-checking)
- **Problems:**
 - Manuel assistance, and/or
 - Combinatorial explosion, and/or
 - Non-termination, and/or
 - Unsoundness, and/or
 - Imprecision (models are not programs)

Origin of the combinatorial explosion: disjunctions

- We have to compute iteratively

$$\text{lfp}_{\perp}^{\sqsubseteq} F$$

where $F \triangleq \bigsqcup_{i \in \Delta} F_i$ is continuous on a cpo

that is

$$\text{lfp}_{\perp}^{\sqsubseteq} F = X^{\omega} = \bigsqcup_{n \geq 0} X^n =$$
$$\bigsqcup_{n \geq 0} \bigsqcup_{i_1, \dots, i_n \in \Delta^n} F_{i_1} \circ \dots \circ F_{i_n}(\perp)$$

combinatorial explosion!

Abstract interpretation

- Sound approximations of disjunctions (Galois connection, widening/narrowing, etc)
- Abstract domains (efficient machine representation of a class of abstract program properties & efficient algorithms for implementing abstract operations and transformers)
- Abstract domain functors for combining abstract domains (e.g. reduced product, reduced cardinal power, etc)

Contribution

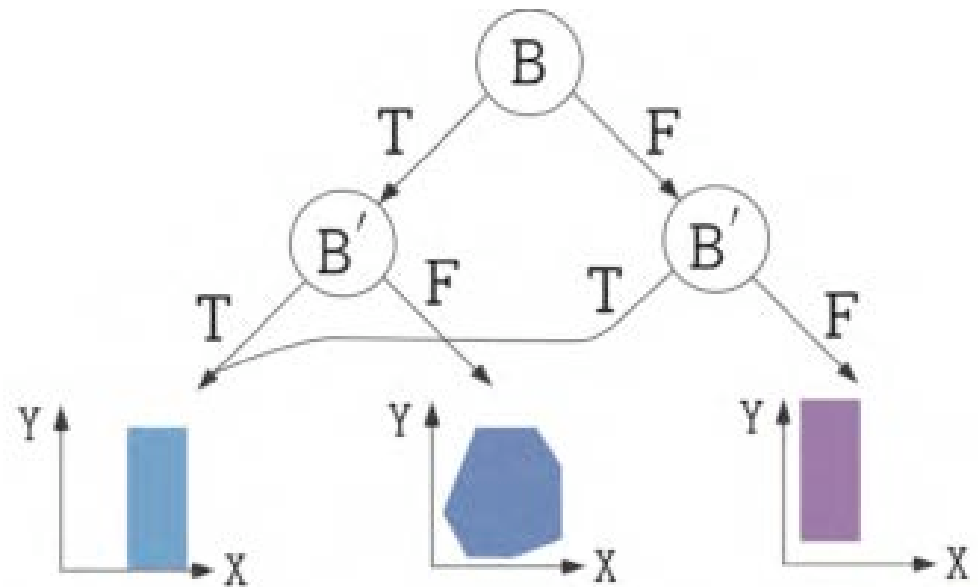
Segmented decision tree functor

- A new abstract domain functor generalizing
 - The **binary decision tree** functor (L. Mauborgne)
 - The **array segmentation** functor (P. Cousot, R. Cousot & F. Logozzo)

to approximate **disjunctions** efficiently with reasonable **expressivity**

The binary decision tree functor

```
/* boolean.c */
typedef enum {F=0,T=1} BOOL;
BOOL B;
void main () {
  unsigned int X, Y;
  while (1) {
    ...
    B = (X == 0);
    ...
    if (!B) {
      Y = 1 / X;
    }
    ...
  }
}
```



The boolean decision tree abstract domain functor is parameterized by the maximal height of the decision tree (an analyzer option) and the abstract domain at the leaves

Implemented in *Astrée*, <http://www.astree.ens.fr/>, <http://www.absint.com/astree/>

The array segmentation functor

```
int n = 10;
int i, A[n];
i = 0;
/* 1: */
while /* 2: */ (i < n) {
    p2 = [ A: <{0} [0,0] {i}? [-∞,+∞] {n,10}?>
          i: [0,+∞] n: [10,10] ]
/* 3: */
    A[i] = 0;
/* 4: */
    i = i + 1;
/* 5: */
}
/* 6: */
p6 = [ A: <{0} [0,0] {n,10,i}> i: [10 +∞] n: [10,10] ]
```

Array A is initialized to 0

Loop invariant at /* 2 */:

if $i = 0$; then

block is empty (so
array A is not
initialized)

else if $i > 0$ then

$A[0] = \dots = A[i-1] = 0$

else (* $i < 0$ *)

Impossible

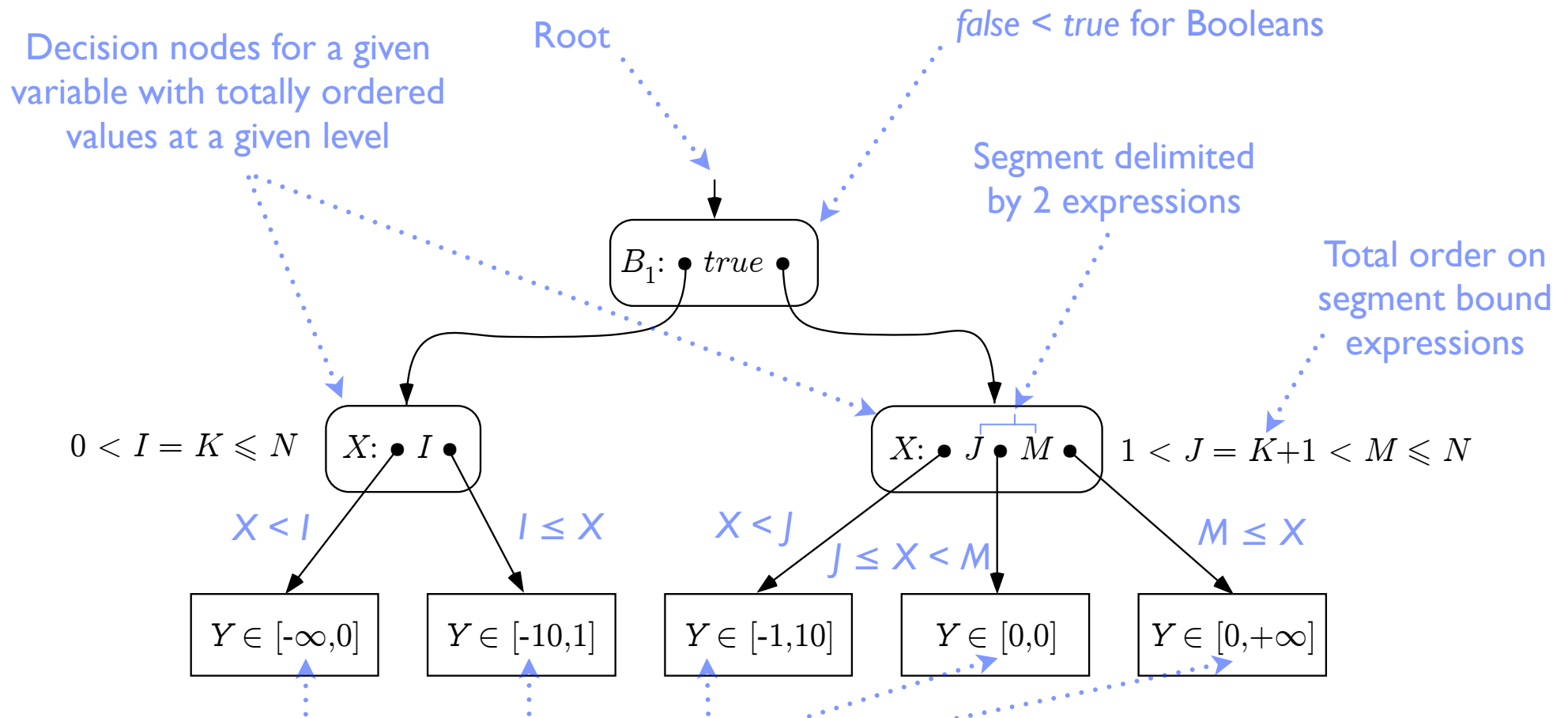
Implemented in Clousot, <http://research.microsoft.com/apps/pubs/default.aspx?id=70614>

The segmented decision tree functor

The segmented decision tree functor

I) Abstract properties

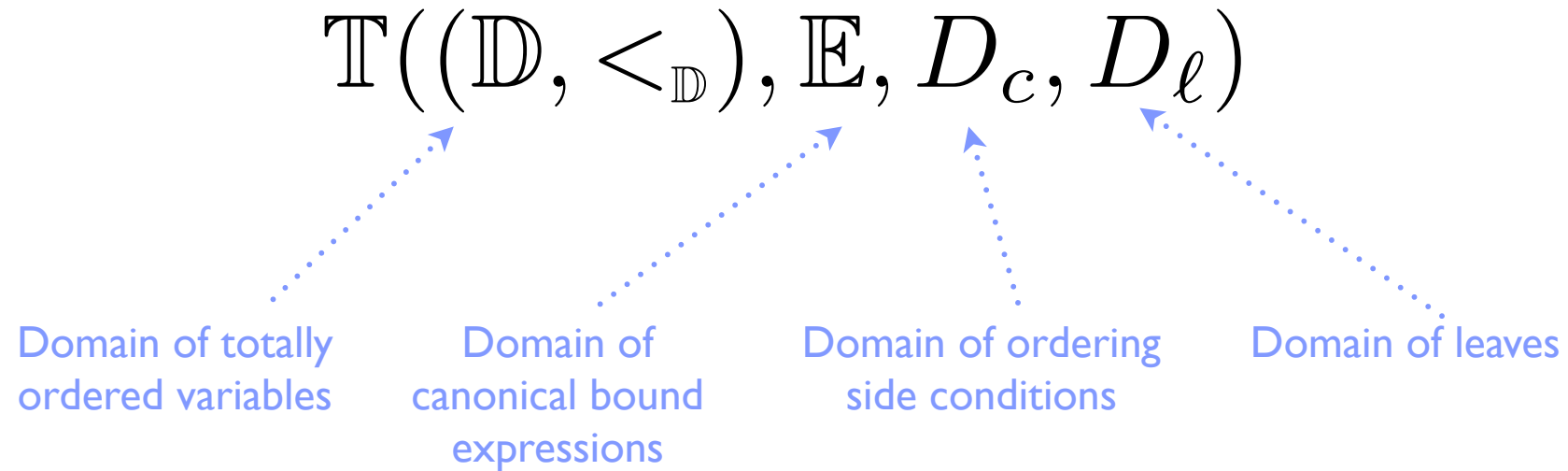
An example of segmented decision tree



The abstract domain at the leaves is a parameter of the functor (here intervals)

This segmented decision tree encodes the fact that if B_1 is false (i.e. $B_1 < \text{true}$) then if $X < I$ then Y is non-positive while if $X \geq I$ then $-10 \leq Y \leq 1$. Similarly, if B_1 is true (i.e. $B_1 \geq \text{true}$) then either $X < J$ and $-1 \leq Y \leq 10$, or $J \leq X < M$ and Y is null, or $X > M$ and Y is non-negative.

The segmented decision tree abstract functor



Controlling costs

- The **time and memory cost** of relational abstract domains grows polynomially/exponentially in the number n of variables
- For **segmented decision trees**:
 - Limit the bound expressions to a **simple canonical form** (e.g. octagons)
 - Limit the **height** of trees (e.g. 3/4)
 - **Variable packing** (*) for side expressions

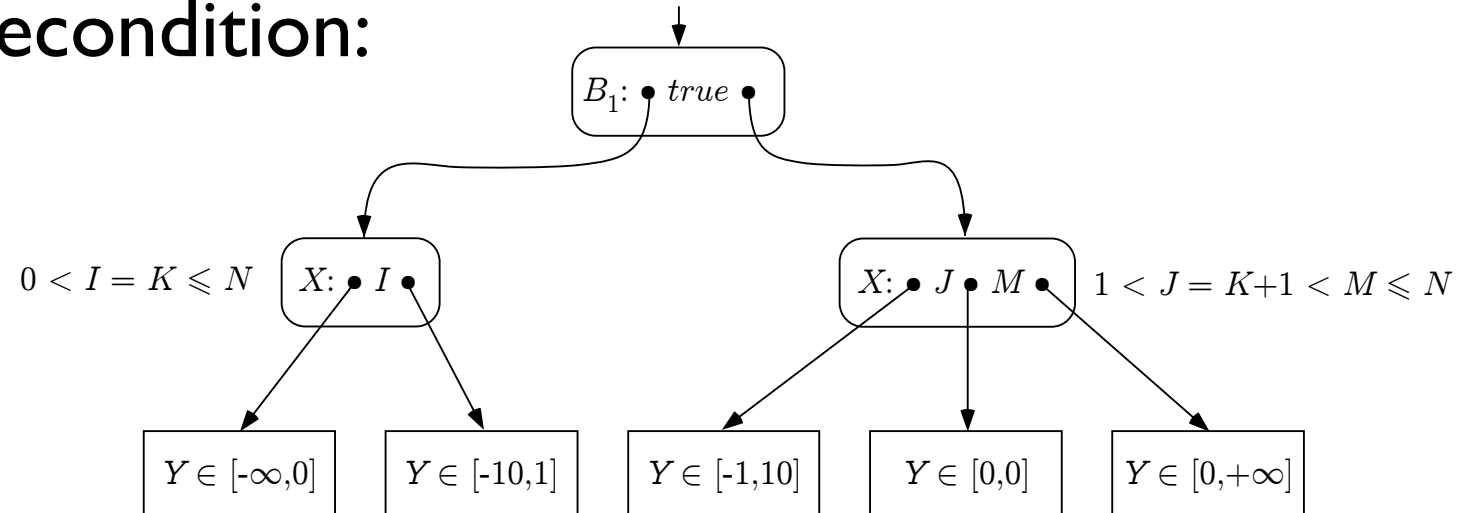
(*) a simple and cheap pre-analysis that groups interdependent variables into packs, leaving unrelated variables in separate packs

The segmented decision tree functor

II) Abstract operations

Segment unification

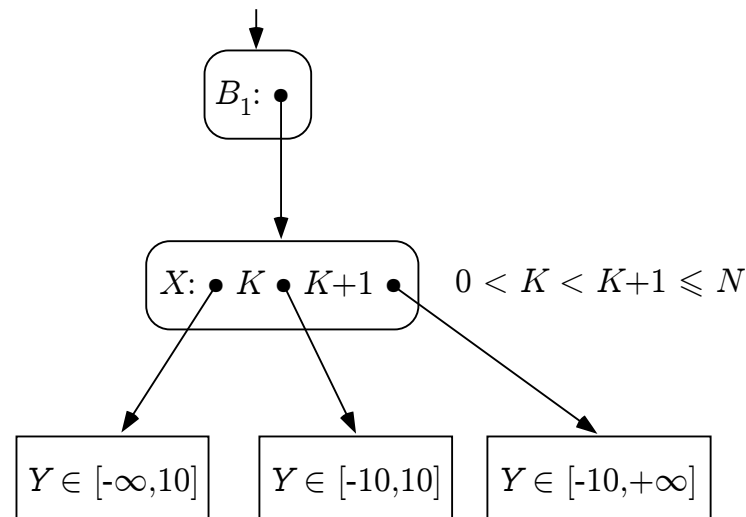
- Abstract precondition:



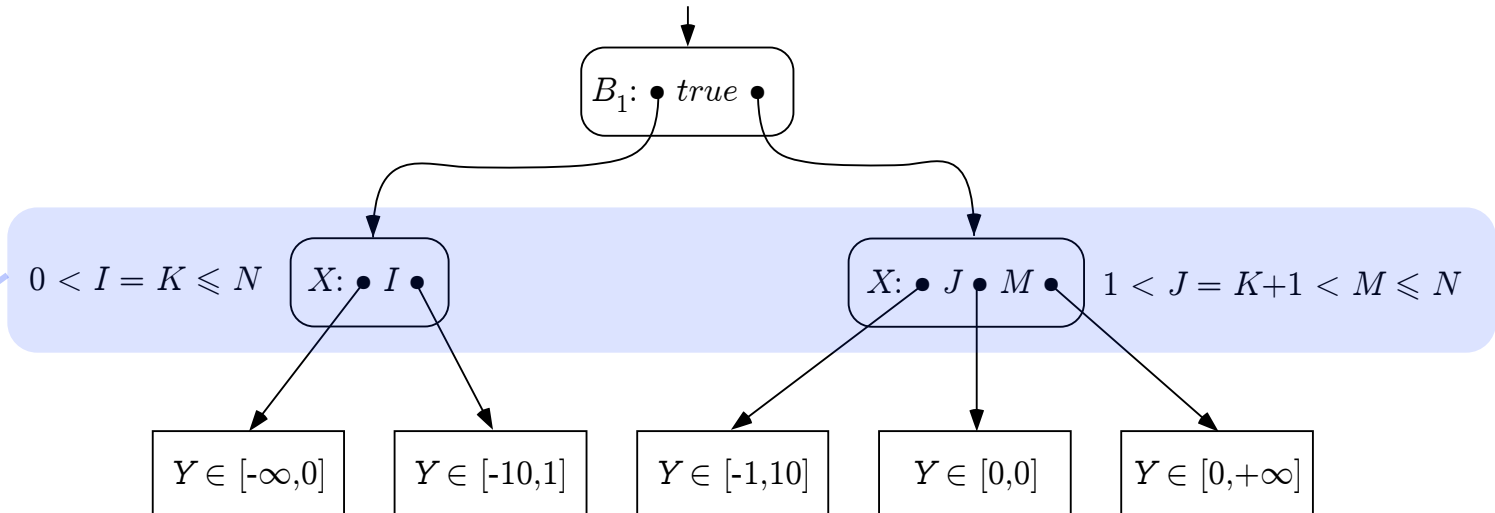
- Assignment:

$$B_1 = ?$$

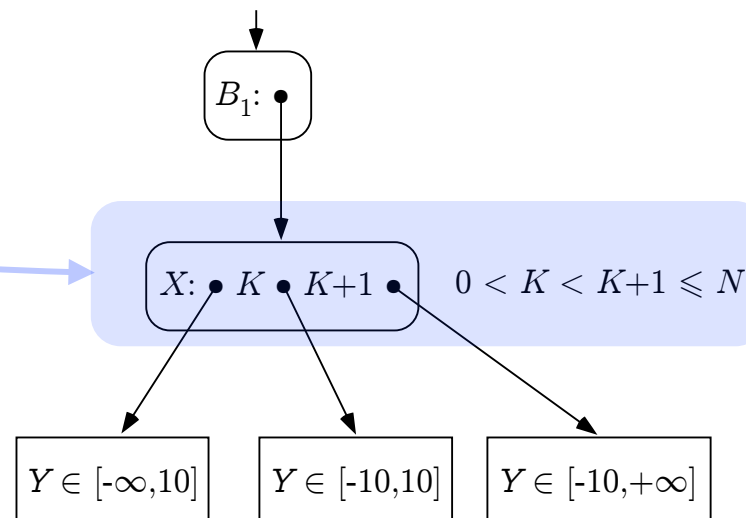
- Abstract postcondition:



Segment unification (cont'd)



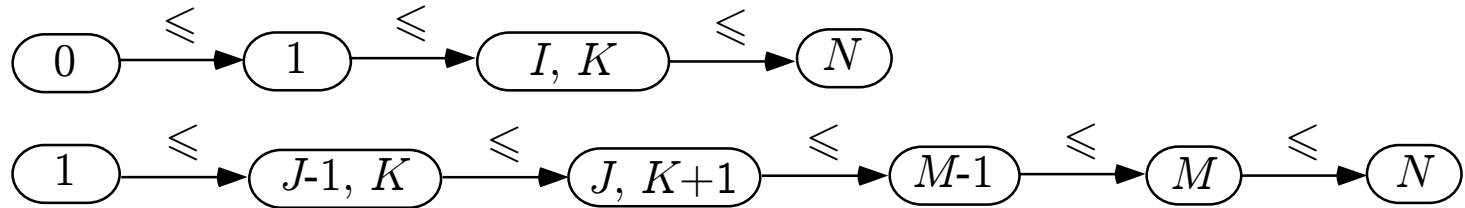
Segment unification



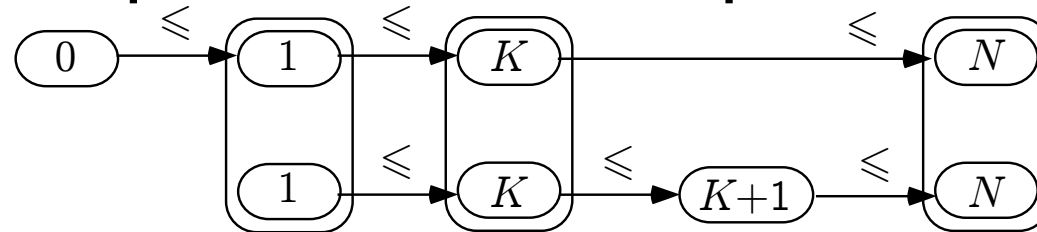
- Given two segments to unify:



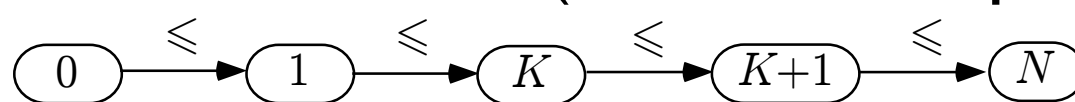
- Build pre-orders with bounds and side conditions



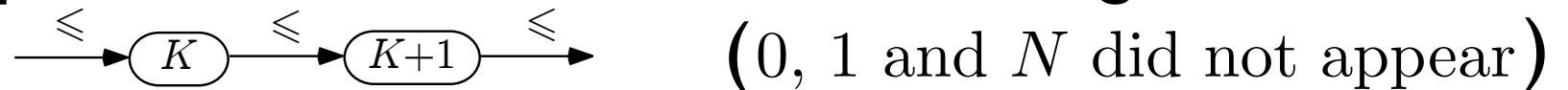
- Eliminate expressions not comparable in both pre-orders:



- Choose a maximal chain (valid in both pre-orders)



- Keep representatives of bounds in either segment



- Merge the corresponding sub-trees

Union, intersection, comparison, etc

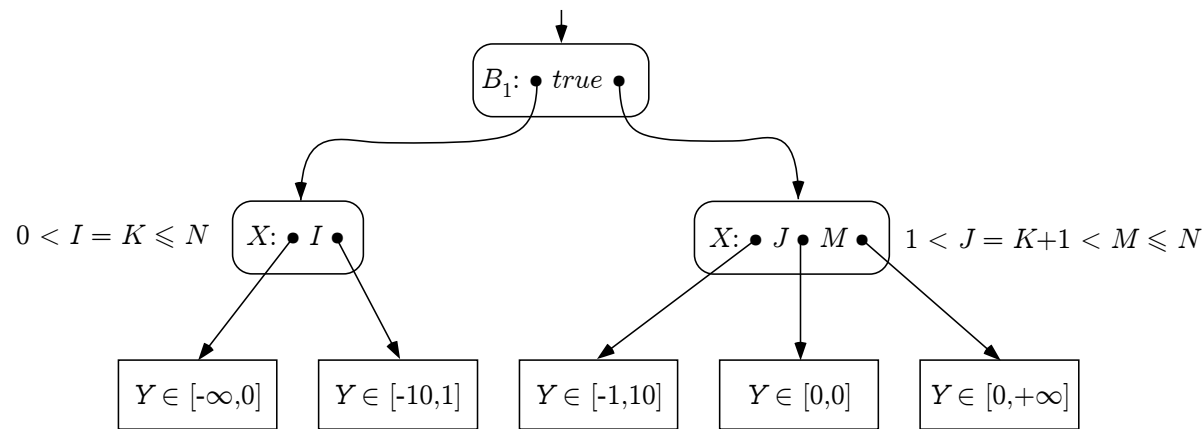
- **Unify** segmentations
- Perform operation **segmentwise** at the leaves

Widening

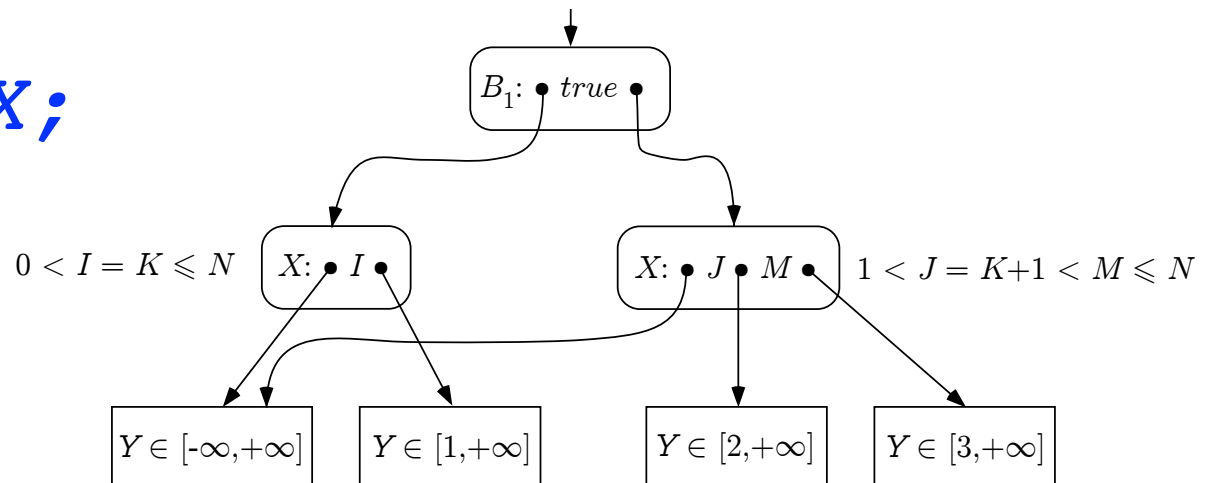
- **Unify** segments **using only common expressions** in both segmentations
 - Use the side-conditions and leave abstract domain widenings
 - The number of expressions in segmentations can only decrease and each segment is widened
- ⇒ **termination**

Assignment to leave variables

- Determine the feasible paths
- Perform assignments at the leaves (opportunistic sharing)



$$Y = X;$$

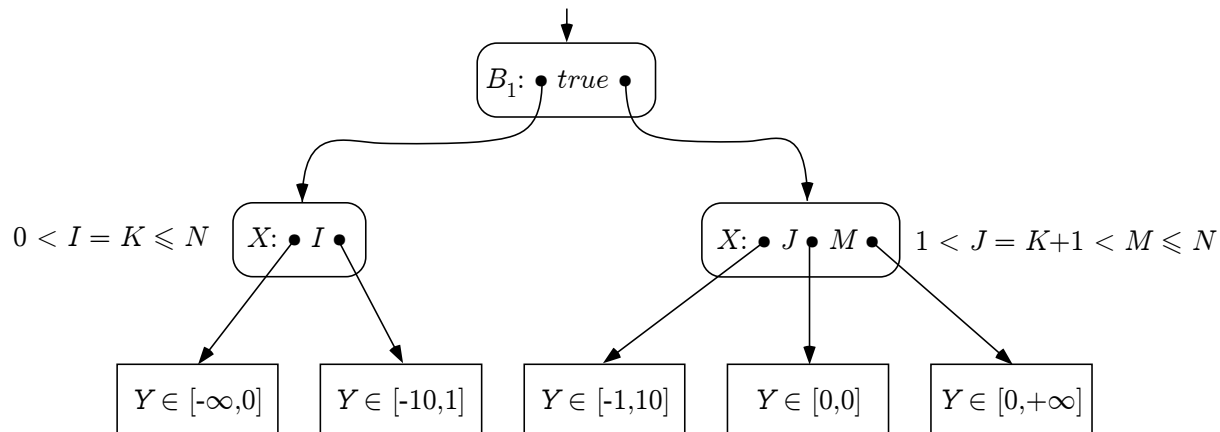


Assignments to variables in segment bounds

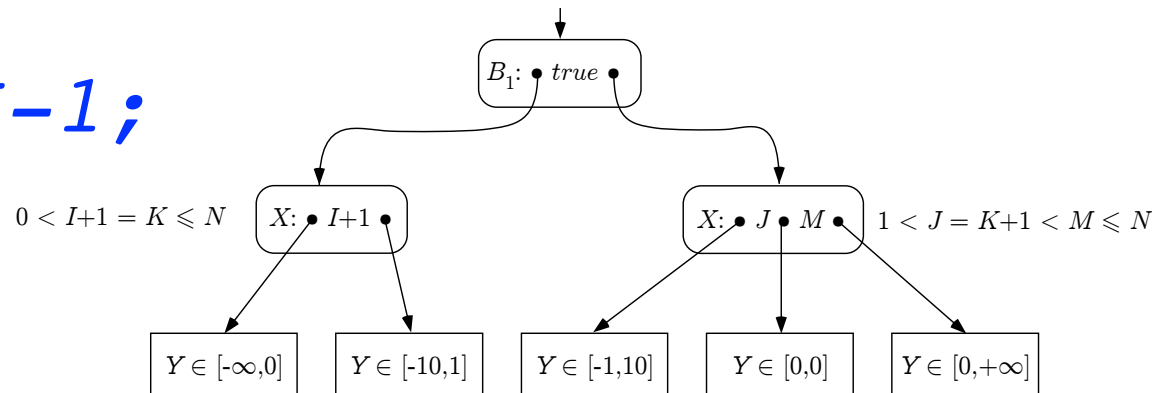
- Invertible assignments:

$$i' = f(i) \Rightarrow i = f^{-1}(i')$$

- Replace i by $f^{-1}(i)$ in each segment bound expression (and side conditions)



$I = I - 1;$



Assignments to variables in segment bounds

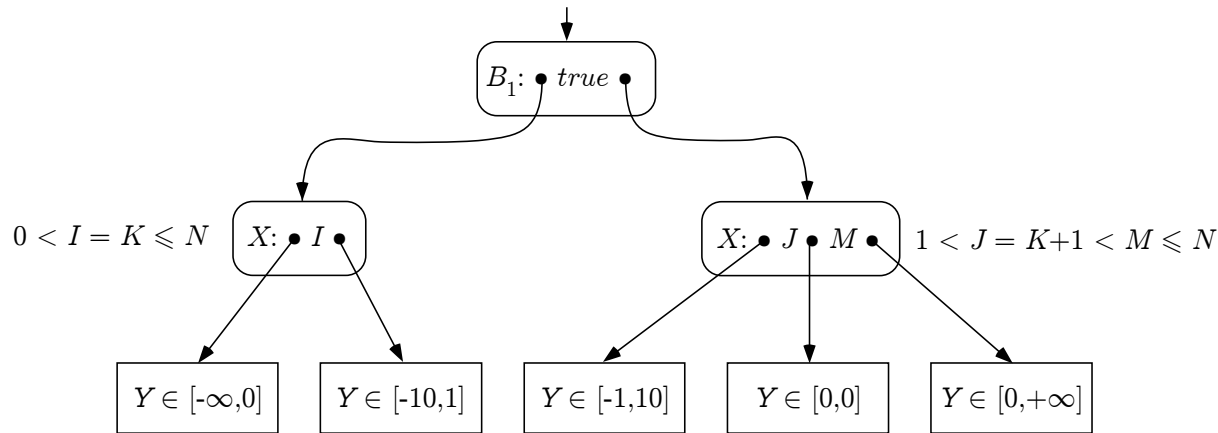
- **Non-invertible assignments:**
 - Replace expressions with that variable by an equal one, if any in side condition
 - Otherwise eliminate the segment bounds and merge segments
 - Take assignment into account in side conditions

Assignments to decision variables

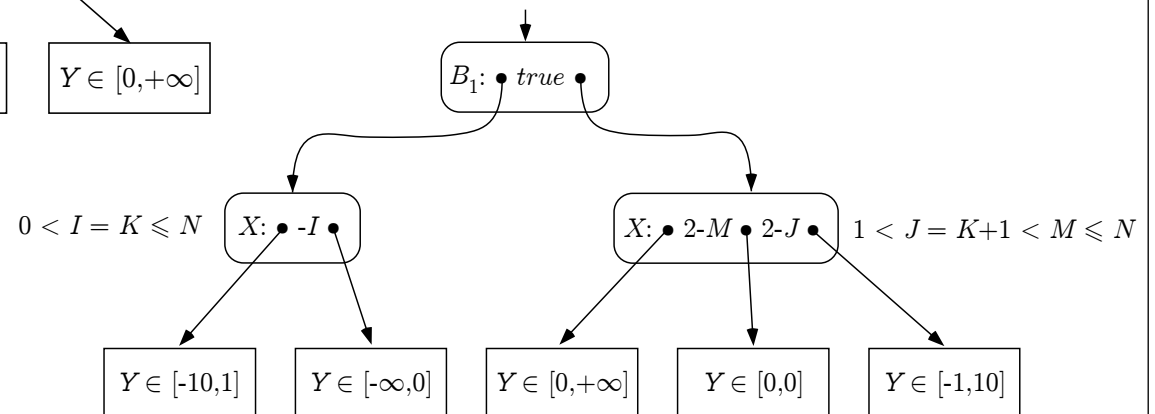
- Invertible assignments:

$$x' = f(x) \Rightarrow x = f^{-1}(x')$$

- Replace $e \leq x$ by $e \leq f^{-1}(x')$ that is
 - $f(e) \leq x'$ when f increasing
 - $f(e) \geq x'$ when f decreasing

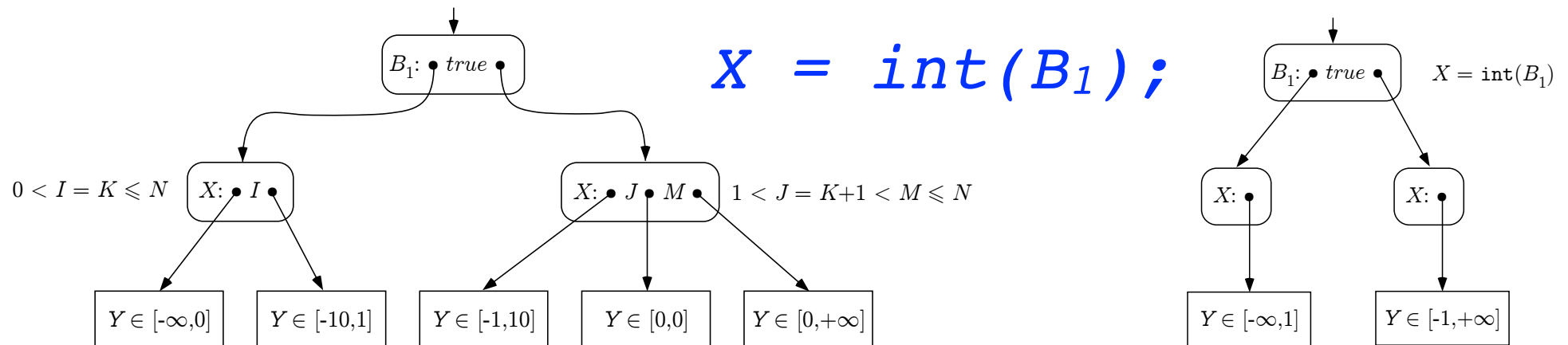


$$X = \text{int}(B_1) - X;$$



Assignments to decision variables

- **Non-invertible assignment:**
 - merge segments related to assigned variable
 - possible preserve information in side-conditions



Abstracting functions and arrays

- $f(x_1, \dots, x_n)$: values at leaves are function of side-conditions on decision variables x_1, \dots, x_n

$\sin x, x \in [0, 2\pi]$ is $\llbracket x \{0 \leq x \leq 2\pi\} : (\sin x : [0, 1]) \ \pi \ (\sin x : [-1, 0]) \rrbracket$

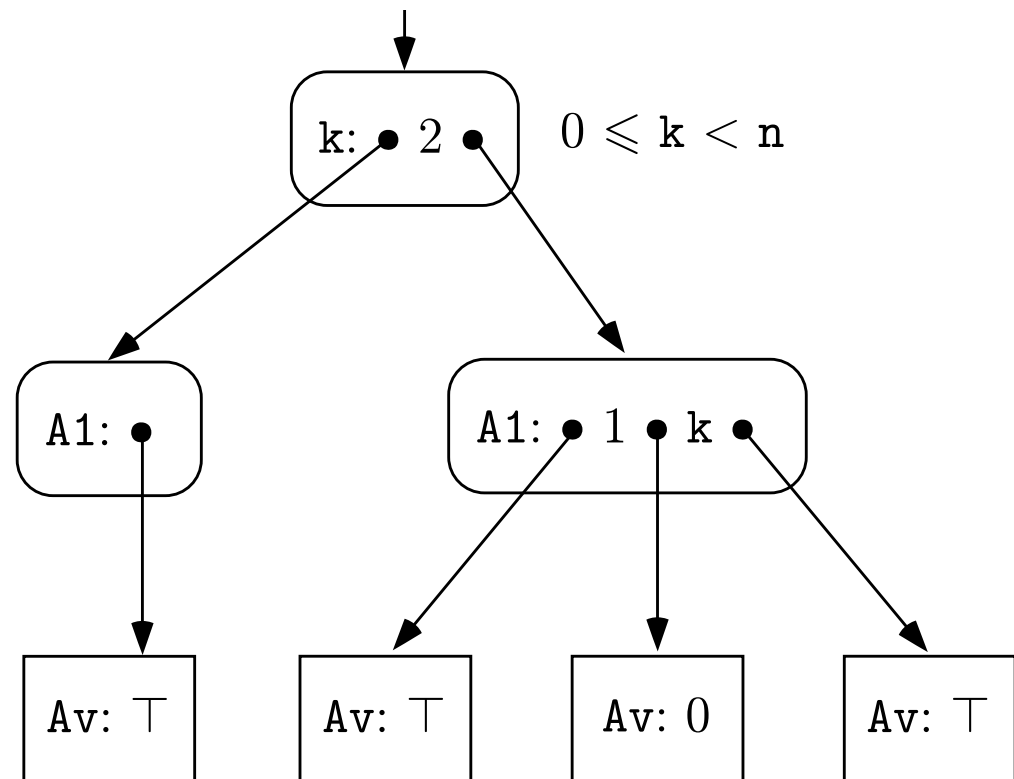
- Arrays A map the indexes (denoted A_i for dimension $i, i = 1, \dots, n$) to values (denoted A_v)

Examples

Partial array initialization

```
int n; /* n > 0 */
int k, A[n];
/* 0: */ k = 0;
/* 1: */ while /* 2: */ (k < n) {
/* 3: */     if (k > 0) {
/* 4: */         A[k] = 0;
/* 5: */     };
/* 6: */     k = k+1;
/* 7: */ };
/* 8: */
```

Loop invariant at /* 3 */:



Partial matrix initialization

```

int m, n; /* m, n > 0 */
int i, j, M[m,n];
/* 0: */ i = 0;
/* 1: */ while /* 2: */ (i < m) {
/* 3: */     j = i+1;
/* 4: */     while /* 5: */ (j < n) {
/* 6: */         M[i,j] = 0;
/* 7: */         j = j+1;
/* 8: */     };
/* 9: */     i = i+1;
/* 10: */ };
/* 11: */

```

11: $\llbracket M1 \{0 < m = i\} : \llbracket M2 : (Mv : \top) M1 + 1 (Mv : 0) \rrbracket \rrbracket$

⊤	0	0	0	0	0	0	0	0	0
⊤	⊤	0	0	0	0	0	0	0	0
⊤	⊤	⊤	0	0	0	0	0	0	0
⊤	⊤	⊤	⊤	0	0	0	0	0	0
⊤	⊤	⊤	⊤	⊤	0	0	0	0	0
⊤	⊤	⊤	⊤	⊤	⊤	0	0	0	0

The analysis computation is automatic, precise and efficient

0: $\llbracket M1 : \llbracket M2 : (Mv : \top) \rrbracket \rrbracket$ $\{ \text{program precondition: } i, j, \text{ and } A \text{ uninitialized} \}$
 ℓ : \perp $\{ \ell = 1, \dots, 11, \text{ infimum} \}$
1:, 2:, 3: $\llbracket M1 \{i = 0\} : \llbracket M2 : (Mv : \top) \rrbracket \rrbracket$ $\{ 0: \text{ with } i = 0, i < m \}$
4:, 5:, 6: $\llbracket M1 \{i = 0, j = i + 1 = 1 < n\} : \llbracket M2 : (Mv : \top) \rrbracket \rrbracket$ $\{ 3: \text{ with } j = i + 1; j < n \}$
7: $\llbracket M1 \{i = 0, j = i + 1 = 1 < n\} :$
 $\llbracket M2 : (Mv : \top) j (Mv : 0) j + 1 (Mv : \top) \rrbracket i + 1 \llbracket M2 : (Mv : \top) \rrbracket$
 \rrbracket $\{ 6: \text{ with } M[i, j] = 0; \}$
8: $\llbracket M1 \{i = 0, j = i + 2 = 2 \leq n\} :$
 $\llbracket M2 : (Mv : \top) j - 1 (Mv : 0) j (Mv : \top) \rrbracket i + 1 \llbracket M2 : (Mv : \top) \rrbracket$
 \rrbracket $\{ 7: \text{ with } j = j + 1; \}$
4: \sqcup_t 8: $\llbracket M1 \{i = 0, i + 1 \leq j \leq i + 2 \leq n\} :$
 $\llbracket M2 : (Mv : \top) 1 (Mv : 0) j (Mv : \top) \rrbracket i + 1 \llbracket M2 : (Mv : \top) \rrbracket$
 \rrbracket $\{ \text{join of 4: and 8:} \}$
5: $\llbracket M1 \{i = 0, i + 1 \leq j \leq n\} :$
 $\llbracket M2 : (Mv : \top) 1 (Mv : 0) j (Mv : \top) \rrbracket i + 1 \llbracket M2 : (Mv : \top) \rrbracket$
 \rrbracket $\{ 5: \nabla (4: \sqcup_t 8:) \}$
9: $\llbracket M1 \{i = 0, i + 1 \leq j = n\} : \llbracket M2 : (Mv : \top) 1 (Mv : 0) i + 1$
 $\llbracket M2 : (Mv : \top) \rrbracket \rrbracket$ $\{ 5: \text{ and } j \geq n \}$
10: $\llbracket M1 \{i = 1, i \leq j = n\} : \llbracket M2 : (Mv : \top) 1 (Mv : 0) \rrbracket i \llbracket M2 : (Mv : \top) \rrbracket \rrbracket$
 $\{ 9: \text{ and } i = i + 1; \}$
1: \sqcup_t 10: $\llbracket M1 \{i = 1, i \leq j = n\} : \llbracket M2 : (Mv : \top) 1 (Mv : 0) \rrbracket i \llbracket M2 : (Mv : \top) \rrbracket \rrbracket$
 $\{ \text{join of 1: and 10:} \}$
2: $\llbracket M1 \{0 \leq i\} : \llbracket M2 : (Mv : \top) 1 (Mv : 0) \rrbracket i \llbracket M2 : (Mv : \top) \rrbracket \rrbracket$
 $\{ 2: \nabla (1: \sqcup_t 10:) \}$
3: $\llbracket M1 \{0 \leq i < m\} : \llbracket M2 : (Mv : \top) 1 (Mv : 0) \rrbracket i \llbracket M2 : (Mv : \top) \rrbracket \rrbracket$
 $\{ 2: \text{ and } j < n \}$
4:, 5:, 6: $\llbracket M1 \{0 \leq i < m, j = i + 1 < n\} :$
 $\llbracket M2 : (Mv : \top) 1 (Mv : 0) \rrbracket i \llbracket M2 : (Mv : \top) \rrbracket$
 \rrbracket $\{ 3: , j = i + 1; \text{ and } j < n \}$

7: $\llbracket M1 \{0 \leq i < m, j = i + 1 < n\} : \llbracket M2 : (Mv : \top) 1 (Mv : 0) \rrbracket i$
 $\llbracket M2 : (Mv : \top) j (Mv : 0) j + 1 (Mv : \top) \rrbracket i + 1 \llbracket M2 : (Mv : \top) \rrbracket$
 \rrbracket $\{ 6: \text{ and } M[i, j] = 0; \}$
8: $\llbracket M1 \{0 \leq i < m, j = i + 2 \leq n\} : \llbracket M2 : (Mv : \top) 1 (Mv : 0) n \rrbracket i$
 $\llbracket M2 : (Mv : \top) j - 1 (Mv : 0) j (Mv : \top) \rrbracket i + 1 \llbracket M2 : (Mv : \top) \rrbracket$
 \rrbracket $\{ 7: \text{ with } j = j + 1; \}$
4: \sqcup_t 8: $\llbracket M1 \{0 \leq i < m, i + 1 \leq j \leq i + 2 \leq n\} : \llbracket M2 : (Mv : \top) 1 (Mv : 0) \rrbracket i$
 $\llbracket M2 : (Mv : \top) i + 1 (Mv : 0) j (Mv : \top) \rrbracket i + 1 \llbracket M2 : (Mv : \top) \rrbracket$
 \rrbracket $\{ \text{join of 4: and 8:} \}$
5: $\llbracket M1 \{0 \leq i < m, i + 1 \leq j \leq n\} : \llbracket M2 : (Mv : \top) 1 (Mv : 0) \rrbracket i$
 $\llbracket M2 : (Mv : \top) i + 1 (Mv : 0) j (Mv : \top) \rrbracket i + 1 \llbracket M2 : (Mv : \top) \rrbracket$
 \rrbracket $\{ 5: \nabla (4: \sqcup_t 8:) \}$
9: $\llbracket M1 \{0 \leq i < m, i + 1 \leq j = n\} : \llbracket M2 : (Mv : \top) 1 (Mv : 0) \rrbracket i$
 $\llbracket M2 : (Mv : \top) i + 1 (Mv : 0) \rrbracket i + 1 \llbracket M2 : (Mv : \top) \rrbracket$
 \rrbracket $\{ 5: \text{ and } j \geq n \}$
10: $\llbracket M1 \{0 < i \leq m, i \leq j = n\} : \llbracket M2 : (Mv : \top) 1 (Mv : 0) \rrbracket i - 1$
 $\llbracket M2 : (Mv : \top) i (Mv : 0) \rrbracket i \llbracket M2 : (Mv : \top) \rrbracket$
 \rrbracket $\{ 9: \text{ and } i = i + 1; \}$
1: \sqcup_t 10: $\llbracket M1 \{0 \leq i \leq m\} : \llbracket M2 : (Mv : \top) M1 + 1 (Mv : 0) \rrbracket i \llbracket M2 : (Mv : \top) \rrbracket \rrbracket$
 $\{ \text{join of 1: and 10: (segments unification yields } 1 \leq M1 + 1 \leq i \text{ for subtree merges)} \}$
2: $\llbracket M1 \{0 \leq i\} : \llbracket M2 : (Mv : \top) M1 + 1 (Mv : 0) \rrbracket i \llbracket M2 : (Mv : \top) \rrbracket \rrbracket$
 $\{ 2: \sqsubseteq (1: \sqcup_t 10:), \text{ stabilization at a fixpoint} \}$
11: $\llbracket M1 \{0 < m = i\} : \llbracket M2 : (Mv : \top) M1 + 1 (Mv : 0) \rrbracket \rrbracket$
 $\{ 2: \text{ and } i \geq m, \text{ program postcondition.} \}$

Conclusion

Abstract domain (functors)

- **Abstract domains** efficiently encode classes of program properties and operations on these properties
- The approach requires more work than universal representations but is much more efficient
- **Abstract domain functors** combine abstract domains to produce many instantiated powerful abstract domain at various levels of cost/precision
- Key to **scalability** with **precision** in abstract interpretation

The End, Thank You