

# Domain approximations for finite set constraint variables

An integrated approach

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## 1 Introduction

- Recapitulation constraint variable
- Domain Approximations
- Set Bounds
- Cardinality Set Bounds
- Full domain
- ROBDDs as data structure for full domain

## 2 Connecting approximations with variable views

- Views as adaptors
- Views as propagation interface - domain lookup
- Simulation of non-existing data structures
- Simulation of non-existing propagators

## 3 Summary

## 4 References

## 4 Additional Slides

# Mainstream

## Constraint Programming

- largely restricted to finite domain variables (FDVar)

### Example

- variables:
  - $x \in [1..6]$
  - $y \in \{1, 3, 6, 8, 12\}$
  - $z \in \{10\}$
- constraints as relations between variables
  - $x + y = z$
  - $x \neq y$

# Beyond finite domains

## When do we use sets?

- constraints are domain specific
- interested in collection of elements
- symmetries among elements have to be avoided
  - students in tutorial groups
  - players in a team
  - workers at a shift
- use finite set variables (FSVar)

# Beyond finite domains

## Example

- variables:
  - $g \in \{\{1, 3\}, \{2, 7, 12\}, \{11, \dots, 14\}\}$
  - $h \in \{\{1, 3, 5, 6\}, \{7, 9, 13\}, \{1, \dots, 20\}\}$
  - $u \in \{\emptyset, \dots, \{1, \dots, 20\}\}$
- constraints:
  - $g \subset h$
  - $|h| = 4$
  - $u = g \cup h$

# Set variables as computation domain

## Finite domain constraint variable $x : D$

### Definition

- Variable  $x \in \text{Var}$
- associated with finite domain  $D \in \text{Dom}$

### Components for integer variable

- Finite set of *variables*  $\text{Var}$ .
- Finite universe  $\mathcal{U} \subset \mathbb{Z}$ .
- Finite set of *values*  $\text{Val} = \mathcal{U}$ .
- Finite set of possible *domains*  $\text{Dom} = \mathcal{P}(\text{Val})$ .

# Set variables as computation domain

## Finite domain constraint variable $x : D$

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- Finite set of *variables*  $\text{Var}$ .
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- Finite set of *values*  $\text{Val} = \mathcal{P}(\mathcal{U})$ .
- Finite set of possible *domains*  $\text{Dom} = \mathcal{P}(\text{Val})$ .

# Representation problem

## Size Issue

- Assume set variable  $x : D = \mathcal{P}(\{1, \dots, 400\})$
- $|D| = 2^{400}$
- Naive enumeration of all values  $\Rightarrow$  exponential size  $\mathcal{O}(2^N)$
- impracticable representation



# Domain Approximation - a viable solution

## Domain Approximation $\mathcal{A}$

- theoretical framework by Benhamou [Ben96]
- representative subset  
 $\mathcal{A} \subseteq Dom$
- closed under intersection  
 $\forall A, B \in \mathcal{A} : (A \cap B) \in \mathcal{A}$
- Elements of  $\mathcal{A}$  are called *approximate domains*

### Required approximate domains

$\emptyset$	set with no values
$\text{Val}$	set with all values
$D \in Dom,  D  = 1$	sets containing a single value

# What approximations are there?

## Overview of approximations

- $(\mathcal{S})$  Set bounds approximation
- $(\mathcal{C})$  Cardinality set bounds approximation
- $(\mathcal{F})$  Full domain approximation

# Set bounds approximation ( $\mathcal{S}$ )

## Theoretical foundations of ( $\mathcal{S}$ )

- Puget in [Pug92]
  - First to introduce set bounds representation in constraint programming
- Gervet in [Ger95, chp. 4]
  - Describes representation in full detail
  - Reference implementation for set constraint solver *Conjunto* [Ger94]

# Set bounds approximation $(\mathcal{S})$

## Convex Set Bounds $(\mathcal{S})$

- approximate a domain  $D \in Dom$
- $E \in (\mathcal{S})$  is the smallest convex interval with respect to  $\subseteq$  containing  $D$

$$\begin{aligned}
 E &= [[E]..[E]]_{\subseteq} \\
 &= \{T \in Dom \mid \inf(D) \subseteq T \subseteq \sup(D)\} \\
 &= [\inf(D).. \sup(D)]_{\subseteq} \\
 &= \left[ \bigcap_{d \in D} d .. \bigcup_{d \in D} d \right]_{\subseteq}
 \end{aligned}$$

- $(\mathcal{S}) \stackrel{\text{def}}{=} \{E \in Dom \mid E \text{ is convex wrt. } \subseteq\}$

# Set Bounds Approximation $(\mathcal{S})$ - Pros and Cons

## Pros $(\mathcal{S})$

- guaranteed linear size
- space efficiency by definition of  $(\mathcal{S})$ 
  - represent only two sets  $\lfloor E \rfloor, \lceil E \rceil$  instead of exponentially many
- *extension* property identified by Gervet in [Ger95, sect.4.2.3 p.45]:
  - set variable  $x : E, E \in (\mathcal{S})$
  - variable assignment  $\alpha \in \text{Var} \rightarrow \text{Val}$ :
    - $\forall v \in \lfloor E \rfloor \Rightarrow v \in \alpha(x)$
    - $\forall v \notin \lceil E \rceil \Rightarrow v \notin \alpha(x)$

Set Bounds Approximation ( $\mathcal{S}$ ) - Pros and ConsPros ( $\mathcal{S}$ )

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Cons ( $\mathcal{S}$ )

- $\lfloor E \rfloor$  represented twice, since  $\lfloor E \rfloor \subseteq \lceil E \rceil$ .

Cardinality set bounds approximation ( $\mathcal{C}$ )More fine grained version of ( $\mathcal{S}$ )

- Used in most constraint solvers:
  - *Mozart* [The06b]
  - *Gecode* [The06a]
  - *ILOG* [ILO00]
- based on set bounds approximation ( $\mathcal{S}$ )
- imposes **additional cardinality restrictions**

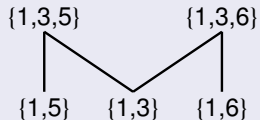
Cardinality set bounds approximation  $(\mathcal{C})$ Extending  $(\mathcal{S})$  to  $(\mathcal{C})$ 

- set variable  $x : E$  and  $E \in (\mathcal{S})$
- adding basic cardinality constraints  $l \leq |x| \leq r$
- translate into cardinality restrictions for  $\lfloor E \rfloor$  and  $\lceil E \rceil$ :  
 $\text{card}(E, l, r) = l \leq \|\lfloor E \rfloor\| \wedge \|\lceil E \rceil\| \leq r$
- $(\mathcal{C})_l^r \stackrel{\text{def}}{=} (\mathcal{S}) \cap \{T \in (\mathcal{S}) \mid \text{card}(T, l, r)\}$



Representing a set variable  $x : D$  in  $(C)$ Set variable  $x : D$  $D = \{\{1, 3\}, \{1, 5\}, \{1, 6\}, \{1, 3, 5\}, \{1, 3, 6\}\}$ 

Corresponding Hesse-Diagram

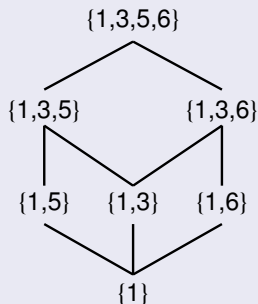


# Representing a set variable $x : D$ in $(C)$

Approximate domain  $E \in (S)$

$$E = [\{1\}.. \{1, 3, 5, 6\}]_{\subseteq}$$

Corresponding Hesse-Diagram

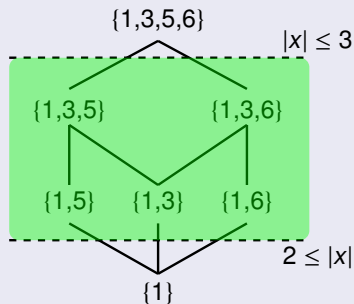


# Representing a set variable $x : D$ in $(C)$

Approximate domain  $F \in (C)_2^3$

$$F = E \cap \{T \in (S) \mid \text{card}(T, 2, 3)\}$$

## Corresponding Hasse-Diagram



# Full domain approximation $(\mathcal{F})$

## From Dom to $(\mathcal{F})$

- choose  $Dom$  itself as approximation of  $Dom$
- approximate a domain  $D \in Dom$  by  $D$
- $(\mathcal{F}) \stackrel{\text{def}}{=} Dom$

# Full domain approximation ( $\mathcal{F}$ ) - Pros and Cons

## Pros ( $\mathcal{F}$ )

- Exact representation of the complete domain
- Encodes characteristic function

$$\chi_D : \mathbb{Z} \mapsto \mathbb{B} : \chi_D(i) = \begin{cases} 1 & \text{if } i \in D \\ 0 & \text{otherwise} \end{cases}$$

to represent a set  $D$

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to represent a set  $D$

## Cons ( $\mathcal{F}$ )

- Space efficiency must be obtained by choice of data structure
- With full approximation still exponential size possible

# Efficient data structure for $(\mathcal{F})$

## Theoretical foundations

- Hawkins Lagoon and Stuckey in [HLS04]
  - First to introduce a full domain approximation
- Use reduced ordered binary decision diagrams (ROBDDs)

Efficient data structure for  $(\mathcal{F})$ Representing a set variable  $x : D$  in  $(\mathcal{F})$ 

- $D = \{\{1, 3\}, \{1, 5\}, \{1, 6\}, \{1, 3, 5\}, \{1, 3, 6\}\}$
- vector of Boolean variables  $b = \langle b_1, b_2, b_3, b_4, b_5, b_6 \rangle$
- ROBDD representing all valuations of formula  $\phi$

$\phi$	$=$	$\neg(b_1 \wedge \neg b_2 \wedge \neg b_3 \wedge \neg b_4 \wedge \neg b_5 \wedge \neg b_6)$	$\neg\{1\}$
	$\vee$	$b_1 \wedge \neg b_2 \wedge b_3 \wedge \neg b_4 \wedge \neg b_5 \wedge \neg b_6$	$\{1, 3\}$
	$\vee$	$b_1 \wedge \neg b_2 \wedge \neg b_3 \wedge \neg b_4 \wedge b_5 \wedge \neg b_6$	$\{1, 5\}$
	$\vee$	$b_1 \wedge \neg b_2 \wedge \neg b_3 \wedge \neg b_4 \wedge \neg b_5 \wedge b_6$	$\{1, 6\}$
	$\vee$	$b_1 \wedge \neg b_2 \wedge b_3 \wedge \neg b_4 \wedge b_5 \wedge \neg b_6$	$\{1, 3, 5\}$
	$\vee$	$b_1 \wedge \neg b_2 \wedge b_3 \wedge \neg b_4 \wedge \neg b_5 \wedge b_6$	$\{1, 3, 6\}$
	$\vee$	$\neg(b_1 \wedge \neg b_2 \wedge b_3 \wedge \neg b_4 \wedge b_5 \wedge b_6)$	$\neg\{1, 3, 5, 6\}$



Efficient data structure for  $(\mathcal{F})$ Representing a set variable  $x : D$  in  $(\mathcal{F})$ 

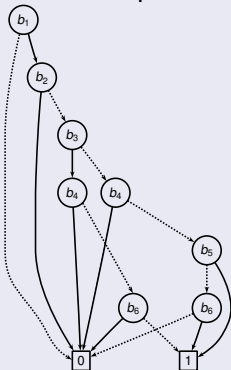
- $D = [\{1\}.. \{1, 3, 5, 6\}]_{\subseteq} \setminus \{\{1\}, \{1, 3, 5, 6\}\}$
- vector of Boolean variables  $b = \langle b_1, b_2, b_3, b_4, b_5, b_6 \rangle$
- ROBDD representing all valuations of formula  $\phi$

$$\begin{aligned} \phi &= b_1 \wedge \neg b_2 \wedge \neg b_4 \\ &\wedge \neg (b_1 \wedge \neg b_2 \wedge \neg b_3 \wedge \neg b_4 \wedge \neg b_5 \wedge \neg b_6) \\ &\wedge \neg (b_1 \wedge \neg b_2 \wedge b_3 \wedge \neg b_4 \wedge b_5 \wedge b_6) \end{aligned}$$

# Efficient data structure for $(\mathcal{F})$

## Representing a set variable $x : D$ in $(\mathcal{F})$

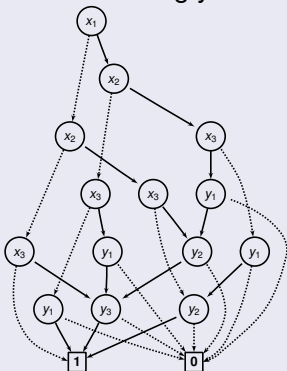
- $D = [\{1\}.. \{1, 3, 5, 6\}]_{\subseteq} \setminus \{\{1\}, \{1, 3, 5, 6\}\}$
- vector of Boolean variables  $b = \langle b_1, b_2, b_3, b_4, b_5, b_6 \rangle$
- *ROBDD* representing all valuations of formula  $\phi$



# Set constraints in $(\mathcal{F})$

## Modeling advantage: constraints as ROBDDs

- constraint  $x \subseteq y, x, y : \mathcal{P}(\{1, 2, 3\})$
- naive modeling yields:



# Set constraints in $(\mathcal{F})$

## Modeling advantage: constraints as ROBDDs

- constraint  $x \subseteq y, x, y : \mathcal{P}(\{1, 2, 3\})$
- Choosing the variable order as
  - 1  $\mathcal{V} = \{v_1, \dots, v_m\}$
  - 2 associated vectors of Boolean variables  $\langle v_{i,1}, \dots, v_{i,N} \rangle$ , where  $i \in \{1, \dots, m\}$  and  $N \stackrel{\text{def}}{=} |\mathcal{U}|$
  - 3 Fix variable order ( $<$ ) as

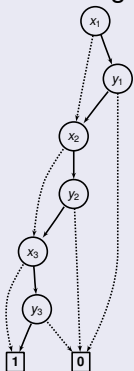
$$v_{1,1} < v_{1,m} < v_{1,2} < \dots < v_{1,N} < \dots < v_{m,N}$$

- 4 *guarantees* linear size of ROBDD except cardinality constraints [LS04]

Set constraints in  $(\mathcal{F})$ 

## Modeling advantage: constraints as ROBDDs

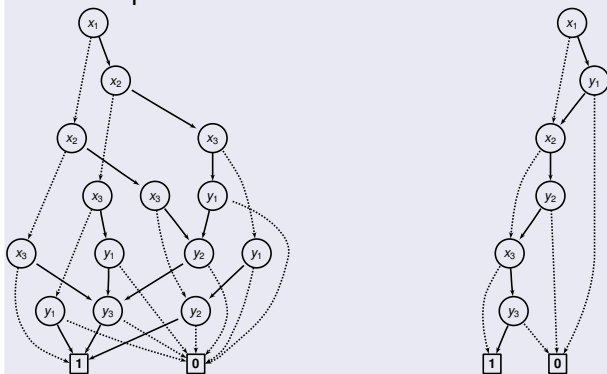
- constraint  $x \subseteq y, x, y : \mathcal{P}(\{1, 2, 3\})$
- According to specified order:



Set constraints in  $(\mathcal{F})$ 

## Modeling advantage: constraints as ROBDDs

- constraint  $x \subseteq y, x, y : \mathcal{P}(\{1, 2, 3\})$
- Comparison



# Different domain approximations available

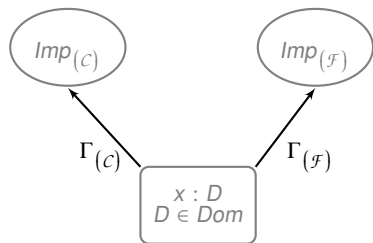
$Imp_C$

$Imp_{\mathcal{F}}$

$x : D$   
 $D \in Dom$

- Howto connect them ?

# Different domain approximations available



- Howto connect them ?
- Using **variable views**



# Variable Views

## Variable Views [ST06]

- 1 Mapping  $V : \mathcal{A} \rightarrow \mathcal{B}$  between domain approximations  $\mathcal{A}$  and  $\mathcal{B}$
- 2 Adaptor for given domain approximation  $\mathcal{A}$ 
  - map  $D \in Dom$  to  $A \in \mathcal{A}$
  - prescribe internal representation (data structures)
- 3 Propagation interface providing propagation services
  - domain lookup
  - domain update
- 4 Simulating non-existing variable representations using existing variable representations

# Using views to connect approximations

## Adaptor functionality

- Set bounds view  $\Gamma_{(S)} : \mathcal{A} \rightarrow (S)$

$$\Gamma_{(S)}(A) = \left[ \bigcap_{a \in A} a .. \bigcup_{a \in A} a \right]_{\subseteq}$$

- Cardinality set bounds view  $\Gamma_{(C)} : \mathcal{A} \rightarrow (C)$

$$\Gamma_{(C)}(A) = \Gamma_{(S)}(A) \cap \{T \in (S) \mid \text{card}(T, l, r)\}$$

- Full domain view  $\Gamma_{(\mathcal{F})} : \mathcal{A} \rightarrow (\mathcal{F})$

$$\Gamma_{(\mathcal{F})}(A) = A$$

# Using views as propagation interface

## Modeling set constraints in $(C)$

- $x : D = \llbracket [D] \rrbracket \dots \llbracket [D] \rrbracket \subseteq C$
- $y : E = \llbracket [E] \rrbracket \dots \llbracket [E] \rrbracket \subseteq C$
- $z : F = \llbracket [F] \rrbracket \dots \llbracket [F] \rrbracket \subseteq C$

constraint	propagator $p_{(C)}$
$x \subseteq y$	$x \subseteq [E] \wedge [D] \subseteq y$
$x \cap y = z$	$[D] \cap [E] \subseteq z \wedge z \subseteq [D] \cap [E]$ $[F] \subseteq x \wedge x \not\subseteq [E] \setminus [F]$ $[F] \subseteq y \wedge y \not\subseteq [D] \setminus [F]$

# Using views as propagation interface

## Access to data structures

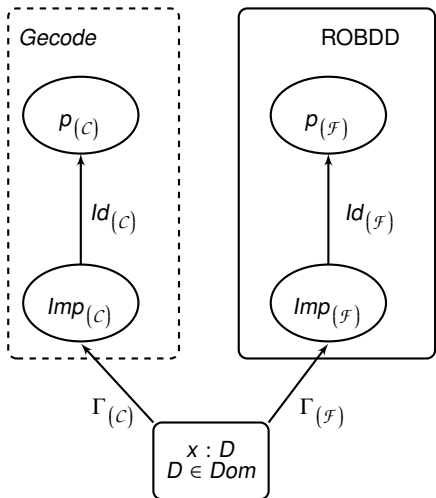
- propagators  $p_{(C)}$  on set variable  $x : E = \llbracket E \rrbracket .. \lceil E \rceil \llbracket C \rrbracket$ 
  - lookup interval bounds  $\llbracket E \rrbracket$  and  $\lceil E \rceil$
  - modify interval bounds  $\llbracket E \rrbracket$  and  $\lceil E \rceil$
- forwarded by propagation interface through variable view  
 $Id_{(C)} : (C) \rightarrow (C), Id_{(C)}(E) = E$

# Using views as propagation interface

## Domain information through iteration

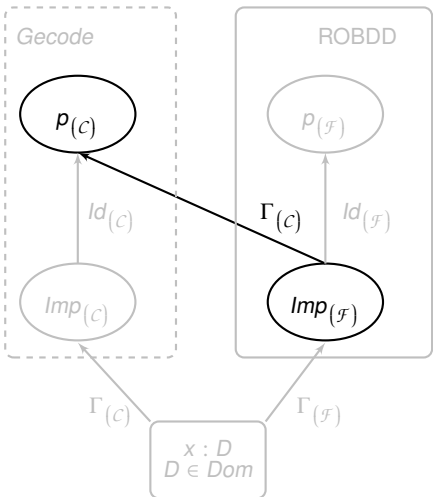
- introduced by Schulte and Tack [ST06]
- iterator `iter` provides functions:
  - `operator()()` test whether we can iterate further
  - `operator++()` increment to next value in set
- depending on structure:
  - `val()` value access
  - `min()` minimum of subinterval
  - `max()` maximum of subinterval

# Combining orthogonal concepts



- Orthogonal concepts
- Do they just coexist ?

# Combining orthogonal concepts



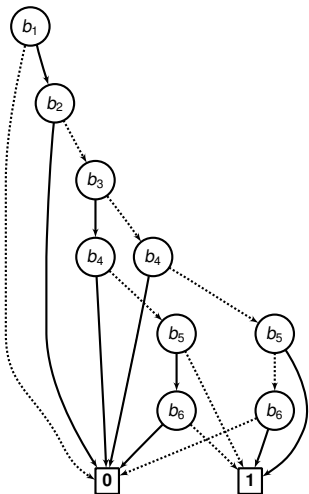
- Orthogonal concepts
- Do they just coexist ?
- No, we can connect them using **variable views**

# Crossing domain approximations

## From $(\mathcal{F})$ to $(\mathcal{C})$

- Simulate Cardinality set bounds interface for ROBDD representing  $D = \{\{1, 3\}, \{1, 5\}, \{1, 6\}, \{1, 3, 5\}, \{1, 3, 6\}\}$
- Apply view  $\Gamma_{(\mathcal{C})}$  on domain  $D \in (\mathcal{F})$ .
  - 1 extract set bounds
  - 2 extract cardinality bounds



Simulate  $(C)$  representation with  $(F)$  using  $\Gamma_{(C)}$ 

▶ Result

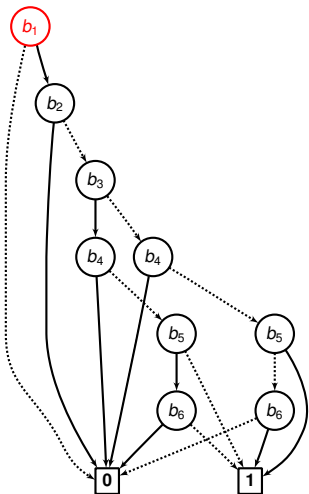
▶ Start

Stack

Inspected node

Flag

 $[[E]..[E]]_C$

Simulate  $(\mathcal{C})$  representation with  $(\mathcal{F})$  using  $\Gamma_{(\mathcal{C})}$ 

▶ Result

▶ Start

Stack

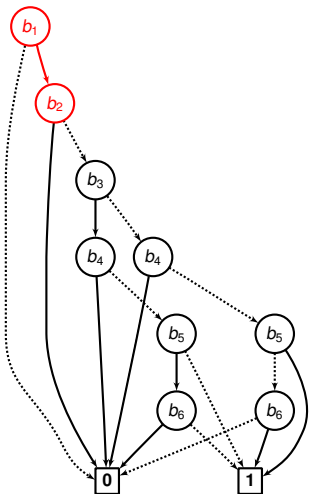
Inspected node

 $(b_1, b_2, \perp)$  Card (0,0)

Flag

INIT

 $[[E]..[E]]_{\mathcal{C}}$  $[\emptyset..[1, 2, 3, 4, 5, 6]]_{\mathcal{C}}$

Simulate  $(\mathcal{C})$  representation with  $(\mathcal{F})$  using  $\Gamma_{(\mathcal{C})}$ 

▶ Result

▶ Start

Stack

 $(b_2, \perp, b_3)$ 

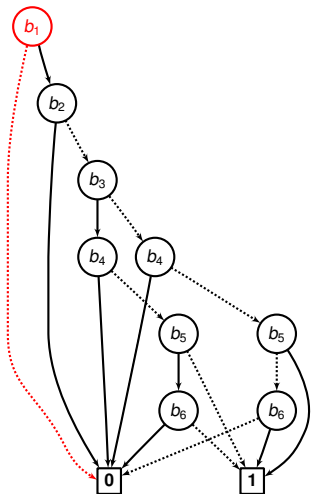
Inspected node

 $(b_1, b_2, \perp)$  **Card (0,0)**

Flag

**INIT** $[[E]..[E]]_{\mathcal{C}}$  $[\emptyset..[1, 2, 3, 4, 5, 6]]_{\mathcal{C}}$

# Simulate $(\mathcal{C})$ representation with $(\mathcal{F})$ using $\Gamma_{(\mathcal{C})}$



▶ Result

▶ Start

Stack

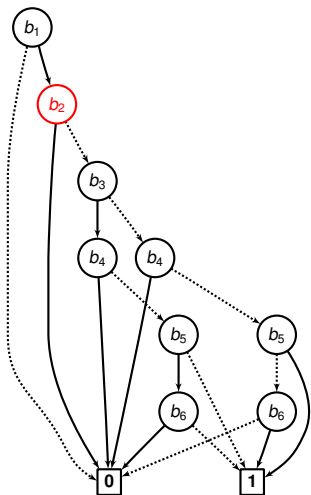
 $(b_2, \perp, b_3)$ 

Inspected node

 $(b_1, b_2, \perp)$  **Card (0,0)**

Flag

**FIX\_GLB** $[[E]..[E]]_{\mathcal{C}}$  $[\{1}..\{1, 2, 3, 4, 5, 6}\}]_{\mathcal{C}}$

Simulate  $(\mathcal{C})$  representation with  $(\mathcal{F})$  using  $\Gamma_{(\mathcal{C})}$ 

▶ Result

▶ Start

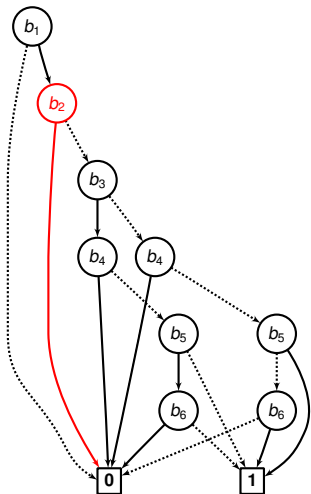
Stack

Inspected node

 $(b_2, \perp, b_3)$  **Card (1,1)**

Flag

**UNDET** $[[E]..[E]]_{\mathcal{C}}$  $[\{1}..\{1, 2, 3, 4, 5, 6}\}]_{\mathcal{C}}$

Simulate  $(\mathcal{C})$  representation with  $(\mathcal{F})$  using  $\Gamma_{(\mathcal{C})}$ 

Stack

Inspected node

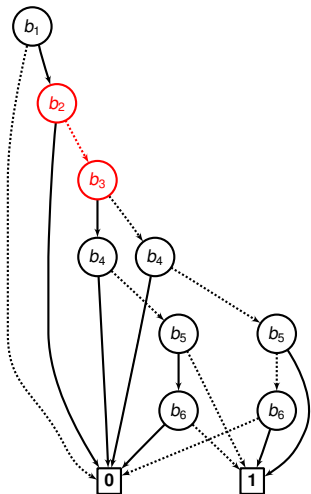
 $(b_2, \perp, b_3)$  **Card (1,1)**

Flag

**FIX\_NOT\_LUB** $[[E]..[E]]_{\mathcal{C}}$  $[\{1}..\{1, 3, 4, 5, 6}\}]_{\mathcal{C}}$ 

▶ Result

▶ Start

Simulate  $(\mathcal{C})$  representation with  $(\mathcal{F})$  using  $\Gamma_{(\mathcal{C})}$ 

▶ Result

▶ Start

Stack

 $(b_3, b_4, b_4)$ 

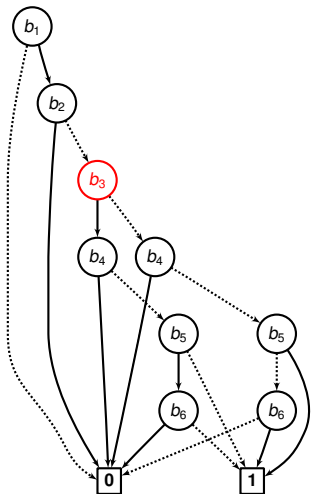
Inspected node

 $(b_2, \perp, b_3)$  **Card (1,1)**

Flag

**FIX\_NOT\_LUB** $[[E]..[E]]_{\mathcal{C}}$  $[\{1}..\{1, 3, 4, 5, 6}\}]_{\mathcal{C}}$

# Simulate $(C)$ representation with $(F)$ using $\Gamma_{(C)}$



▶ Result

▶ Start

Stack

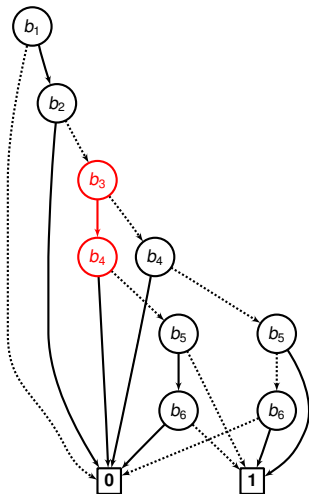
Inspected node

 $(b_3, b_4, b_4)$  **Card (1,1)**

Flag

**FIX\_UNKNOWN** $[[E]..[E]]_C$  $[\{1}..\{1, 3, 4, 5, 6}\}]_C$



Simulate  $(\mathcal{C})$  representation with  $(\mathcal{F})$  using  $\Gamma_{(\mathcal{C})}$ 

▶ Result

▶ Start

Stack

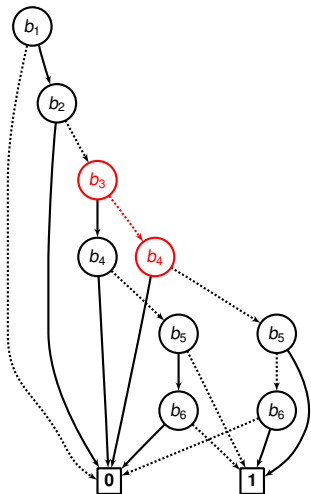
 $(b_4, \perp, b_5)$ 

Inspected node

 $(b_3, b_4, b_4)$  **Card (1,1)**

Flag

**FIX\_UNKNOWN** $[[E]..[E]]_{\mathcal{C}}$  $[\{1}..\{1, 3, 4, 5, 6}\}]_{\mathcal{C}}$

Simulate  $(\mathcal{C})$  representation with  $(\mathcal{F})$  using  $\Gamma_{(\mathcal{C})}$ 

▶ Result

▶ Start

Stack

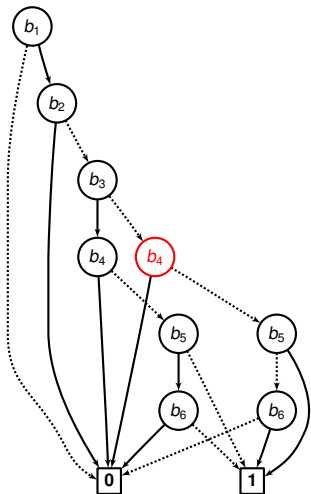
 $(b_4, \perp, b_5) (b_4, \perp, b_5)$ 

Inspected node

 $(b_3, b_4, b_4)$  **Card (1,1)**

Flag

**FIX\_UNKNOWN** $[[E]..[E]]_{\mathcal{C}}$  $[\{1}..\{1, 3, 4, 5, 6}\}]_{\mathcal{C}}$

Simulate  $(\mathcal{C})$  representation with  $(\mathcal{F})$  using  $\Gamma_{(\mathcal{C})}$ 

▶ Result

▶ Start

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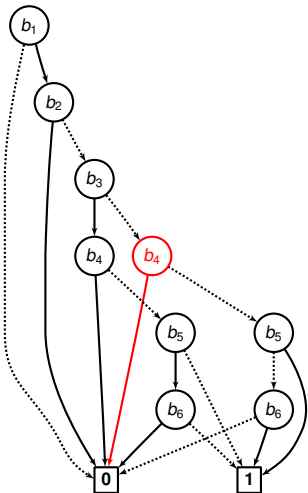
 $(b_4, \perp, b_5)$ 

Inspected node

 $(b_4, \perp, b_5)$  **Card (1,1)**

Flag

**UNDET** $[[E]..[E]]_{\mathcal{C}}$  $[\{1}..\{1, 3, 4, 5, 6}\}]_{\mathcal{C}}$

Simulate  $(\mathcal{C})$  representation with  $(\mathcal{F})$  using  $\Gamma_{(\mathcal{C})}$ 

▶ Result

▶ Start

Stack

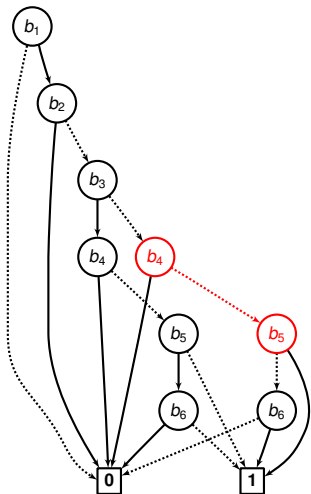
 $(b_4, \perp, b_5)$ 

Inspected node

 $(b_4, \perp, b_5)$  **Card (1,1)**

Flag

**FIX\_NOT\_LUB** $[[E]..[E]]_{\mathcal{C}}$  $[\{1}..\{1, 3, 4, 5, 6}\}]_{\mathcal{C}}$

Simulate  $(\mathcal{C})$  representation with  $(\mathcal{F})$  using  $\Gamma_{(\mathcal{C})}$ 

▶ Result

▶ Start

Stack

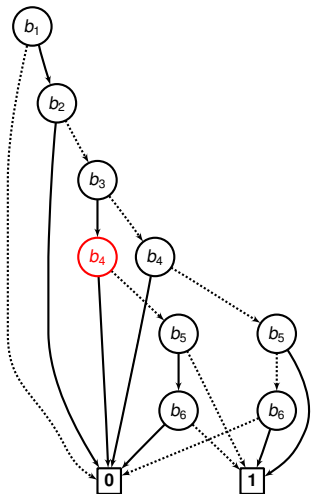
 $(b_5, \perp, b_6)$  $(b_4, \perp, b_5)$ 

Inspected node

 $(b_4, \perp, b_5)$  **Card (1,1)**

Flag

**FIX\_NOT\_LUB** $[[E] .. [E]]_{\mathcal{C}}$  $[\{1\} .. \{1, 3, 4, 5, 6\}]_{\mathcal{C}}$

Simulate  $(\mathcal{C})$  representation with  $(\mathcal{F})$  using  $\Gamma_{(\mathcal{C})}$ 

▶ Result

▶ Start

Stack

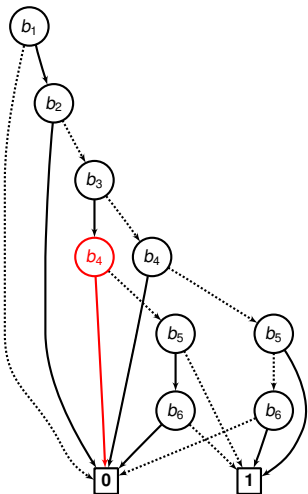
 $(b_5, \top, b_6)$ 

Inspected node

 $(b_4, \perp, b_5)$  **Card (2,2)**

Flag

**FIX\_NOT\_LUB** $[[E]..[E]]_{\mathcal{C}}$  $[\{1}..\{1, 3, 4, 5, 6}\}]_{\mathcal{C}}$

Simulate  $(\mathcal{C})$  representation with  $(\mathcal{F})$  using  $\Gamma_{(\mathcal{C})}$ 

▶ Result

▶ Start

Stack

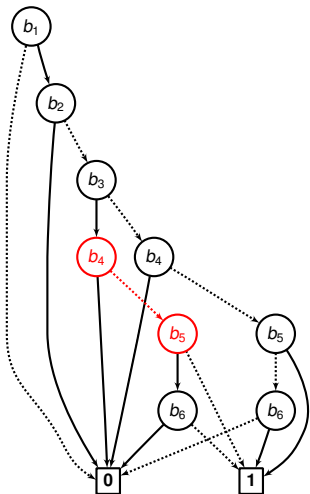
 $(b_5, \top, b_6)$ 

Inspected node

 $(b_4, \perp, b_5)$  **Card (2,2)**

Flag

**FIX\_NOT\_LUB** $[[E]..[E]]_{\mathcal{C}}$  $[\{1}..\{1, 3, 5, 6}]_{\mathcal{C}}$

Simulate  $(\mathcal{C})$  representation with  $(\mathcal{F})$  using  $\Gamma_{(\mathcal{C})}$ 

▶ Result

▶ Start

Stack

 $(b_5, \top, b_6)$   $(b_5, b_6, \top)$ 

Inspected node

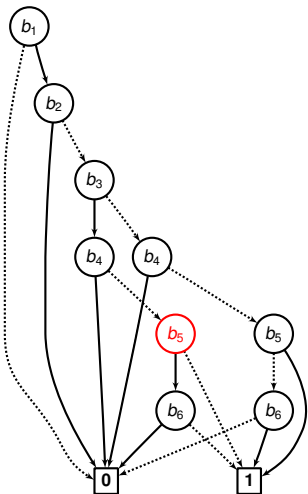
 $(b_4, \perp, b_5)$  **Card (2,2)**

Flag

**FIX\_NOT\_LUB** $[[E]..[E]]_{\mathcal{C}}$  $[\{1}..\{1, 3, 5, 6}]_{\mathcal{C}}$



# Simulate $(C)$ representation with $(F)$ using $\Gamma_{(C)}$



▶ Result   ▶ Start

Stack

$(b_5, \top, b_6)$

Inspected node

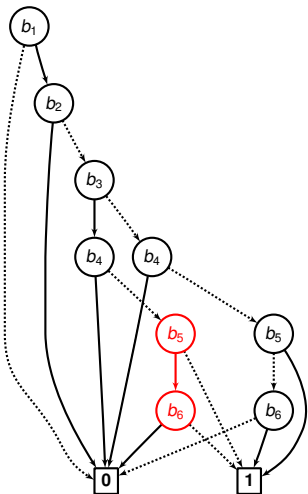
$(b_5, b_6, \top)$  **Card (2,2)**

Flag

**UNDET**

$[[E]..[E]]_C$

$[\{1}..\{1, 3, 5, 6}]_C$

Simulate  $(\mathcal{C})$  representation with  $(\mathcal{F})$  using  $\Gamma_{(\mathcal{C})}$ 

▶ Result

▶ Start

Stack

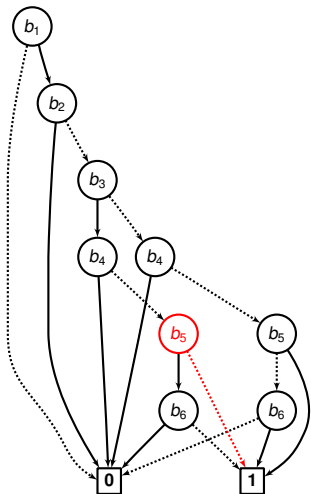
 $(b_5, \top, b_6)$  $(b_6, \perp, \top)$ 

Inspected node

 $(b_5, b_6, \top)$  **Card (2,2)**

Flag

**UNDET** $[[E]..[E]]_{\mathcal{C}}$  $[\{1\}..\{1, 3, 5, 6\}]_{\mathcal{C}}$

Simulate  $(\mathcal{C})$  representation with  $(\mathcal{F})$  using  $\Gamma_{(\mathcal{C})}$ 

▶ Result

▶ Start

Stack

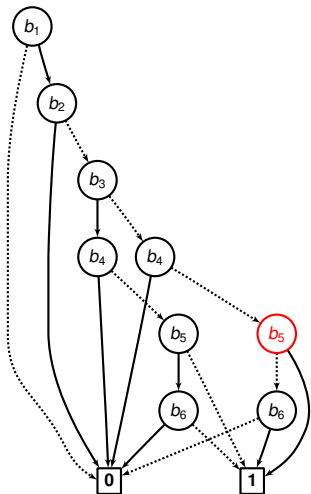
 $(b_5, \top, b_6)$  $(b_6, \perp, \top)$ 

Inspected node

 $(b_5, b_6, \top)$  **Card (2,2)**

Flag

**FIX\_NOT\_LUB** $[[E]..[E]]_{\mathcal{C}}$  $[\{1\}..\{1, 3, 5, 6\}]_{\mathcal{C}}$

Simulate  $(\mathcal{C})$  representation with  $(\mathcal{F})$  using  $\Gamma_{(\mathcal{C})}$ 

▶ Result

▶ Start

Stack

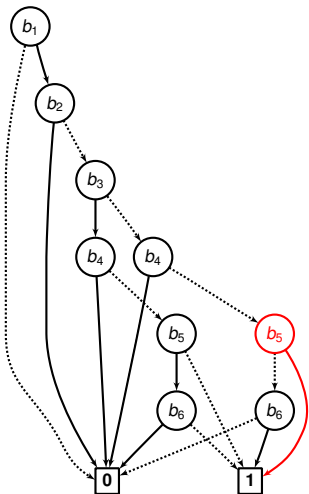
 $(b_6, \perp, \top)$ 

Inspected node

 $(b_5, \top, b_6)$  **Card (1,1)**

Flag

**FIX\_NOT\_LUB** $[[E]..[E]]_{\mathcal{C}}$  $[\{1\}..\{1, 3, 5, 6\}]_{\mathcal{C}}$

Simulate  $(\mathcal{C})$  representation with  $(\mathcal{F})$  using  $\Gamma_{(\mathcal{C})}$ 

▶ Result

▶ Start

Stack

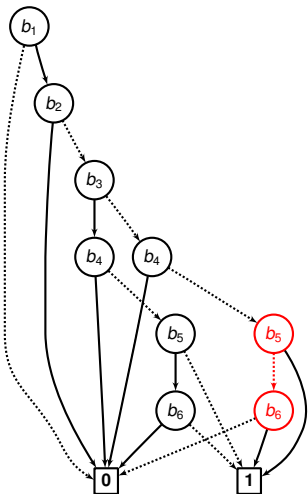
 $(b_6, \perp, \top)$ 

Inspected node

 $(b_5, \top, b_6)$  **Card (1,1)**

Flag

**UNDET** $[[E]..[E]]_{\mathcal{C}}$  $[\{1}..\{1, 3, 5, 6}\}]_{\mathcal{C}}$

Simulate  $(\mathcal{C})$  representation with  $(\mathcal{F})$  using  $\Gamma_{(\mathcal{C})}$ 

▶ Result

▶ Start

Stack

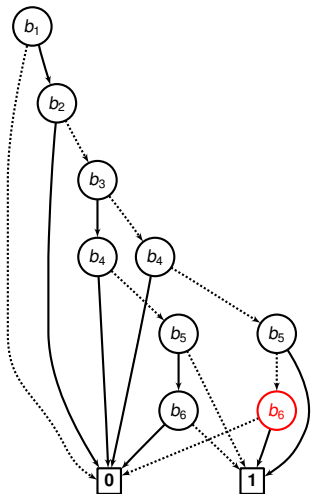
 $(b_6, \top, \perp) (b_6, \perp, \top)$ 

Inspected node

 $(b_5, \top, b_6)$  **Card (1,1)**

Flag

**UNDET** $[[E] .. [E]]_{\mathcal{C}}$  $[\{1\} .. \{1, 3, 5, 6\}]_{\mathcal{C}}$

Simulate  $(\mathcal{C})$  representation with  $(\mathcal{F})$  using  $\Gamma_{(\mathcal{C})}$ 

▶ Result

▶ Start

Stack

 $(b_6, \perp, \top)$ 

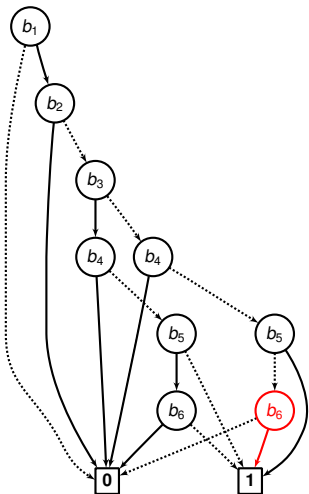
Inspected node

 $(b_6, \top, \perp)$  Card (1,1)

Flag

UNDET

 $[[E]..[E]]_{\mathcal{C}}$  $[\{1}..\{1, 3, 5, 6}\}_{\mathcal{C}}$

Simulate  $(\mathcal{C})$  representation with  $(\mathcal{F})$  using  $\Gamma_{(\mathcal{C})}$ 

▶ Result

▶ Start

Stack

 $(b_6, \perp, \top)$ 

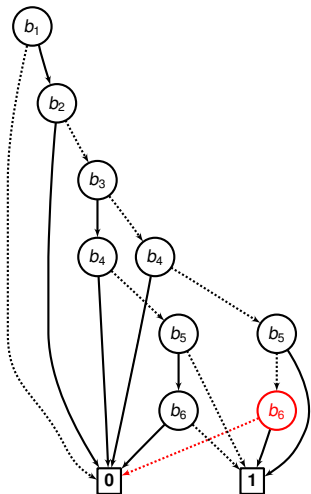
Inspected node

 $(b_6, \top, \perp)$  **Card (2,2)**

Flag

**FIX\_GLB** $[[E] .. [E]]_{\mathcal{C}}$  $[\{1\} .. \{1, 3, 5, 6\}]_{\mathcal{C}}$



Simulate  $(\mathcal{C})$  representation with  $(\mathcal{F})$  using  $\Gamma_{(\mathcal{C})}$ 

▶ Result

▶ Start

Stack

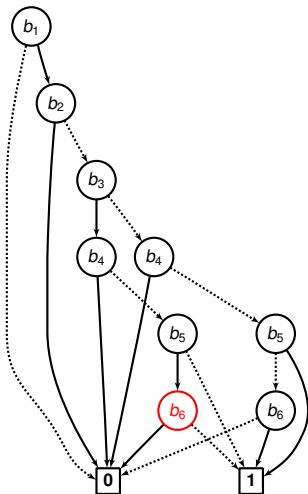
 $(b_6, \perp, \top)$ 

Inspected node

 $(b_6, \top, \perp)$  **Card (2,2)**

Flag

**FIX\_GLB** $[[E] .. [E]]_{\mathcal{C}}$  $[\{1\} .. \{1, 3, 5, 6\}]_{\mathcal{C}}$

Simulate  $(C)$  representation with  $(F)$  using  $\Gamma_{(C)}$ 

▶ Result

▶ Start

Stack

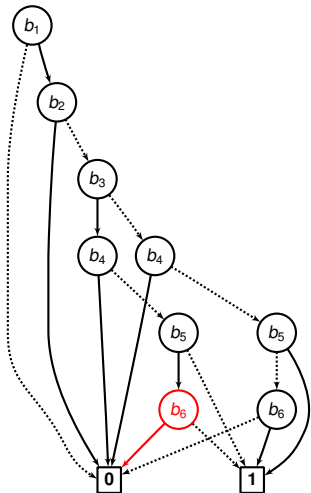
Inspected node

 $(b_6, \perp, \top)$  Card (3,3)

Flag

**FIX\_GLB** $[[E]..[E]]_C$  $[\{1}..\{1, 3, 5, 6}]_C$

# Simulate $(\mathcal{C})$ representation with $(\mathcal{F})$ using $\Gamma_{(\mathcal{C})}$



▶ Result

▶ Start

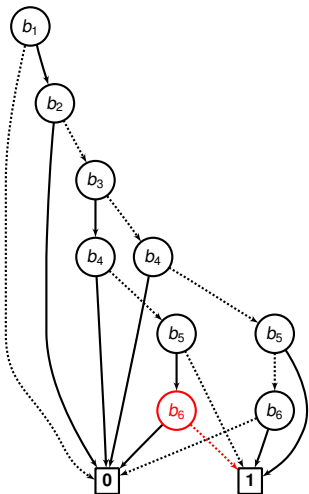
Stack

Inspected node

 $(b_6, \perp, \top)$  **Card (3,3)**

Flag

**UNDET**
 $[[E]..[E]]_{\mathcal{C}}$ 
 $[\{1}..\{1, 3, 5, 6}]_{\mathcal{C}}$

Simulate  $(\mathcal{C})$  representation with  $(\mathcal{F})$  using  $\Gamma_{(\mathcal{C})}$ 

▶ Result

▶ Start

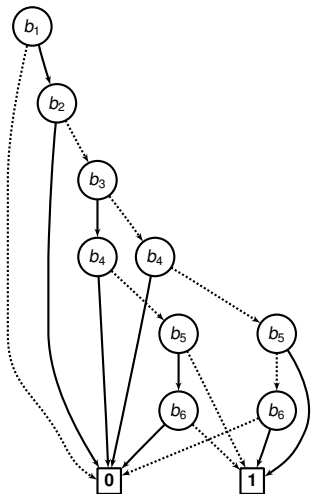
Stack

Inspected node

 $(b_6, \perp, \top)$  **Card (3,3)**

Flag

**UNDET** $[[E]..[E]]_{\mathcal{C}}$  $[\{1}..\{1, 3, 5, 6}]_{\mathcal{C}}$

Simulate  $(\mathcal{C})$  representation with  $(\mathcal{F})$  using  $\Gamma_{(\mathcal{C})}$ 

▶ Result

▶ Start

Stack

Inspected node

Flag

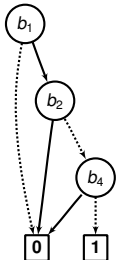
 $[[E]..[E]]_{\mathcal{C}}$  $[[\{1}..\{1, 3, 5, 6}]]_{\mathcal{C}}$

# Simulate $(C)$ representation with $(\mathcal{F})$ using $\Gamma_{(C)}$

## Resulting

- resulting ROBDD represents

- $E = \Gamma_{(C)}(D) = [\{1\}.. \{1, 3, 5, 6\}]_C$
- cardinality restrictions  $\text{card}(E, 2, 3)$



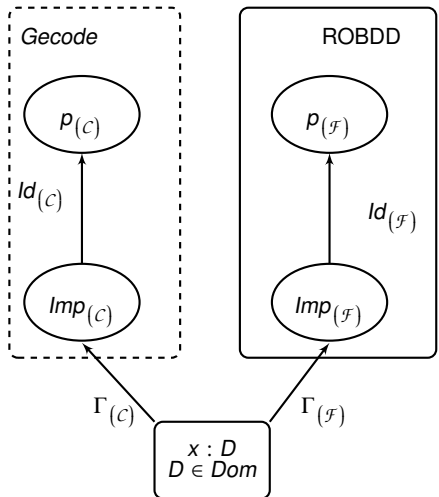
### Cardinality

$$c \in [2..3]$$

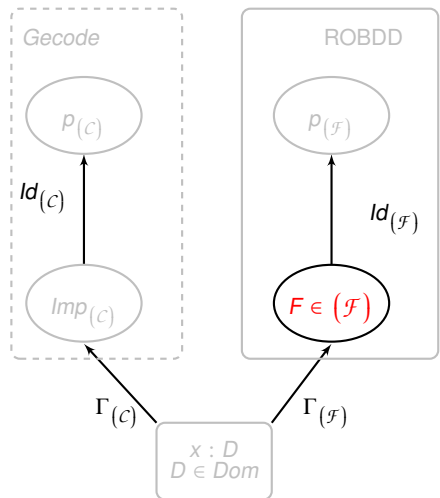
$$[[E].. [E]]_C$$

$$[\{1\}.. \{1, 3, 5, 6\}]_C$$

# Weaken propagation



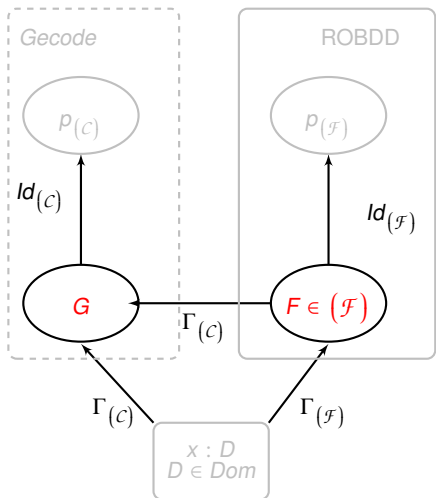
# Weaken propagation



- $F \in (\mathcal{F})$  is  $x$ -component of domain tuple  $\vec{F}$ ,  $\vec{F}.x = F$

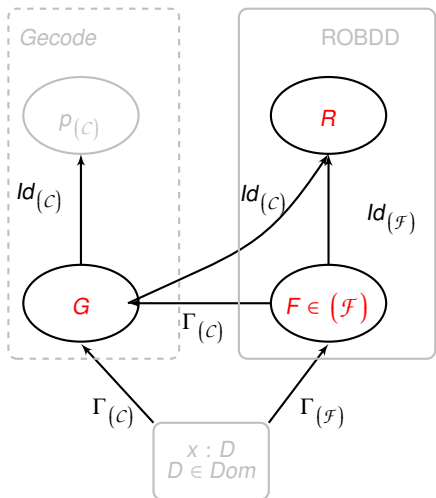


# Weaken propagation



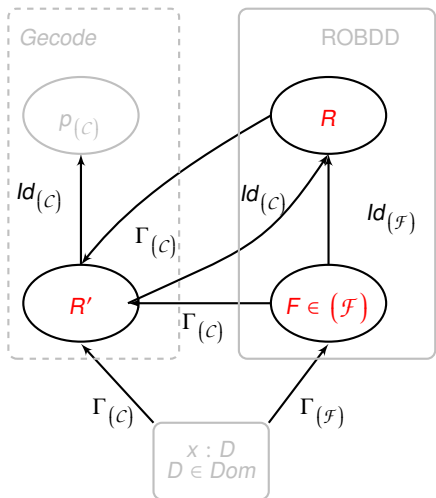
- 1  $F \in (\mathcal{F})$  is  $x$ -component of domain tuple  $\vec{F}$ ,  $\vec{F}.x = F$
- 2 Map  $F$  to the respective cardinality set bounds  $G = \Gamma_{(c)}(F)$

# Weaken propagation



- 1  $F \in (\mathcal{F})$  is  $x$ -component of domain tuple  $\vec{F}$ ,  $\vec{F}.x = F$
- 2 Map  $F$  to the respective cardinality set bounds  $G = \Gamma(c)(F)$
- 3 Since  $G \in (C) \subset (\mathcal{F})$  apply  $\rho_{(\mathcal{F})} : (\mathcal{F})^n \rightarrow (\mathcal{F})^n$
- 4 Propagation result  $R = \rho_{(\mathcal{F})}(\vec{G}).x$

# Weaken propagation



- 1  $F \in (\mathcal{F})$  is  $x$ -component of domain tuple  $\vec{F}$ ,  $\vec{F}.x = F$
- 2 Map  $F$  to the respective cardinality set bounds  $G = \Gamma(c)(F)$
- 3 Since  $G \in (\mathcal{C}) \subset (\mathcal{F})$  apply  $p_{(\mathcal{F})} : (\mathcal{F})^n \rightarrow (\mathcal{F})^n$
- 4 Propagation result  $R = p_{(\mathcal{F})}(\vec{G}).x$
- 5 Map result  $R$  again to  $R' = \Gamma(c)(R)$ .

# Using variable views to weaken propagation

## Changing consistency of propagation result

- Use a domain-consistent propagator

$$p_{(\mathcal{F})} : (\mathcal{F})^n \rightarrow (\mathcal{F})^n$$

- obtain bounds( $(\mathcal{C})$ )-consistent propagation  $\beta_{(\mathcal{C})} : (\mathcal{F})^n \rightarrow (\mathcal{C})^n$ ,  
 $\beta_{(\mathcal{C})}(\vec{F}) = \Gamma_{(\mathcal{C})}(p_{(\mathcal{F})}(\Gamma_{(\mathcal{C})}(\vec{F})))$

# Contributions

- 1 Implemented presented concepts in *Gecode*

# Contributions

## Gecode Constraint Library

- **g**eneric
- **c**onstraint
- **d**evelopment
- **e**nvironment



**Gecode** [The06a], a C++ library for constraint programming.  
Version 1.3.1 available from <http://www.gecode.org>

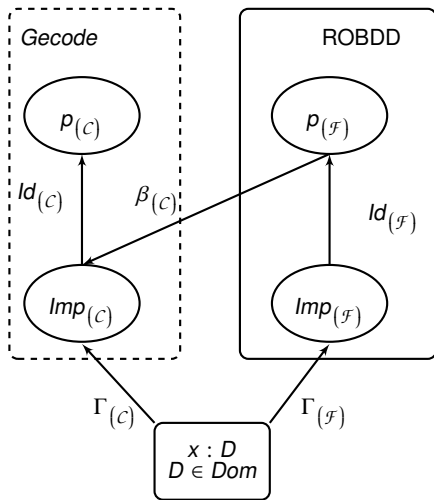
## Developers

- **Dr. Christian Schulte** (head, KTH, Sweden)
- **Guido Tack** (PS Lab, Saabrücken, Germany)

# Contributions

- 1 Implemented presented concepts in *Gecode*
  - 1 ROBDD set component
  - 2 Simulation  $(\mathcal{C})$  with  $(\mathcal{F})$  using view  $\Gamma_{(\mathcal{C})}$
  - 3 Propagation across approximations using  $\beta_{(\mathcal{C})}$

## Contributions

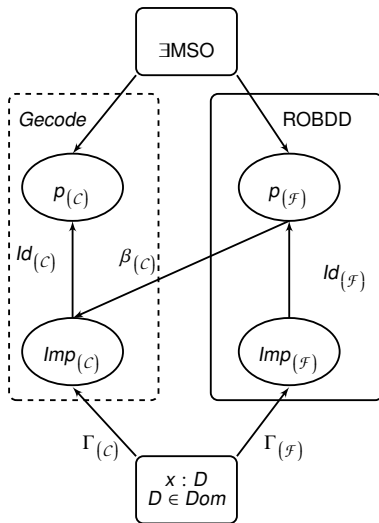




# Contributions

- 1 Implemented presented concepts in *Gecode*
  - 1 ROBDD set component
  - 2 Simulation  $(C)$  with  $(\mathcal{F})$  using view  $\Gamma_{(C)}$
  - 3 Propagation across approximations using  $\beta_{(C)}$
  - 4 Also implemented:
    - 1  $\beta_{(S)}$  for proper set bounds
    - 2  $\beta_{(\perp)}$  for lexicographic bounds
- 2 First framework to connect different implementations for set variables via variable views.
- 3 Prototype for generating set propagators from  $\exists$ MSO as uniform specification language [TSS06]

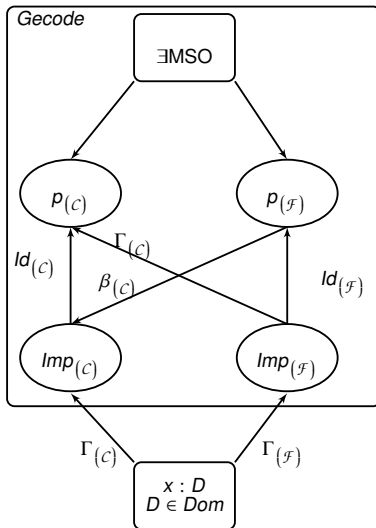
## Contributions



# Contributions

- 1 Implemented presented concepts in *Gecode*
  - 1 ROBDD set component
  - 2 Simulation  $(C)$  with  $(\mathcal{F})$  using view  $\Gamma_{(C)}$
  - 3 Propagation across approximations using  $\beta_{(C)}$
  - 4 Also implemented:
    - 1  $\beta_{(S)}$  for proper set bounds
    - 2  $\beta_{(L)}$  for lexicographic bounds
- 2 First framework to connect different implementations for set variables via variable views.
- 3 Prototype for generating set propagators from uniform specification language
- 4 Introduce different implementation for  $(C)$ 
  - compared different implementations for  $(C)$

## Contributions - Completing the picture



# Outlook and future work

- Generalize results to multisets by introducing
  - 1 approximations
  - 2 views
  - 3 constraints
- Comparison of used data structures with different data structures for example:
  - 1 Bit vectors

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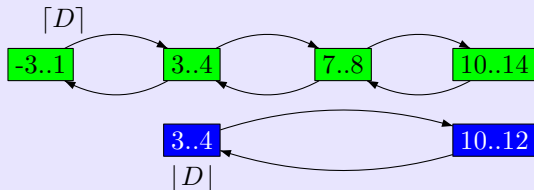
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# Sets in Gecode

## Representation of finite integer sets

- bounds representation of domain  $D$  by  $[[D] .. [D]]$
- each bound represented by a range list
- $D = [\{3, 4, 10, 11, 12\}.. \{-3, -2, -1, 0, 1, 3, 4, 7, 8, 10, 11, 12, 13, 14\}]$

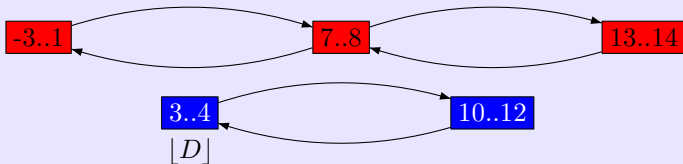


## Sets in Gecode

## Representation of finite integer sets

- remove  $\lfloor D \rfloor$  from  $\lceil D \rceil$
- obtain  $\Delta(D) = \lceil D \rceil \setminus \lfloor D \rfloor$

$$\Delta \lceil D \rceil \setminus \lfloor D \rfloor$$



# Sets in Gecode

## Minimal bounds representation [AB00]

- minimal bounds rep
- store disjoint union of  $\Delta(D)$  and  $\lfloor D \rfloor$  such that  $\Delta(D) \uplus \lfloor D \rfloor = \lceil D \rceil$

$\Delta \uplus \lfloor D \rfloor$



# Fragment of $\exists$ MSO

## Fragment

$$S ::= \exists x.\langle S \rangle \mid \langle F \rangle$$

$$F ::= \forall v.\langle B \rangle \mid \exists v.\langle B \rangle \mid \langle F \rangle \wedge \langle F \rangle$$

$$B ::= \langle B \rangle \wedge \langle B \rangle \mid \langle B \rangle \vee \langle B \rangle \mid \neg \langle B \rangle \mid v \in x \mid x \in \text{Var} \mid \perp$$

## Example

Express constraint in  $\exists$ MSO

- constraint  $c \equiv x \cap y = z$
- $\exists$ MSO-formula  $\phi_c = \forall v.v \in x \wedge y \in y \Leftrightarrow v \in z$