

# Associating Narrowing Functions to Constraints

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# Associating Narrowing Functions to Constraints

## Projection Function and its Enclosure

## Constraint Decomposition Method

- Primitive Constraints

- Inverse Functions

- Projection Function Enclosure with the Inverse Function

## Constraint Newton Method

- Interval Projections

- Properties of an Interval Projection

- Projection Function Enclosure with the Interval Projection

## Complementary Approaches

# Projection Function and its Enclosure

A set of narrowing functions is associated with a constraint by considering projections with respect to each variable in the scope

A projection function identifies from a box:

all the possible values of a particular variable for which there is a value combination belonging to the constraint relation

**Projection Function.** Let  $P=(X,D,C)$  be a CCSP. The projection function with respect to a constraint  $c=(s,\rho)\in C$  and a variable  $x_i\in s$ , denoted  $\pi_{x_i}^\rho$ , obtains a set of real values from a real box  $B$  and is defined by:

$$\pi_{x_i}^\rho(B) = \{ d[x_i] \mid d \in \rho \wedge d \in B \} = (\rho \cap B)[x_i]$$

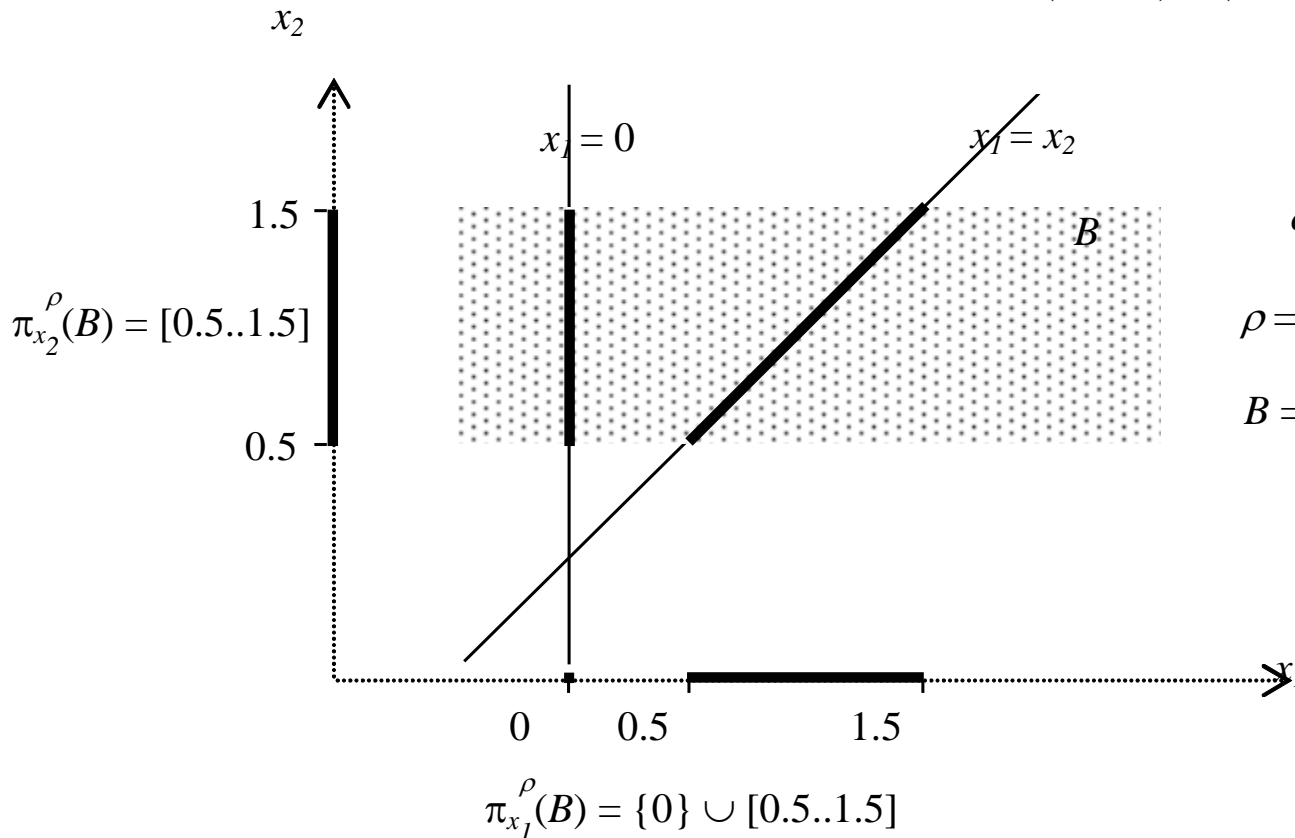
□

All value combinations within  $B$  with  $x_i$  values outside  $\pi_{x_i}^\rho(B)$  are outside the relation  $\rho$  and so they do not satisfy the constraint  $c$ .

# Projection Function and its Enclosure

$$P = (X, D, C) = (\langle x_1, x_2 \rangle, D_1 \times D_2, \{c\})$$

$$c \equiv x_1 \times (x_2 - x_1) = 0$$



$$c = (\langle x_1, x_2 \rangle, \rho)$$

$$\rho = \{ \langle x_1, x_2 \rangle \in D \mid x_1 \times (x_2 - x_1) = 0 \}$$

$$B = \langle [-0.5..2.5], [0.5..1.5] \rangle$$

All value combinations within  $B$  with  $x_i$  values outside  $\pi_{x_i}^\rho(B)$  are outside the relation  $\rho$  and so they do not satisfy the constraint  $c$ .

# Projection Function and its Enclosure

A box-narrowing function narrows the domain of one variable, from a box representing all the variables of the CCSP, eliminating some values that do not belong to a projection function

**Box-Narrowing Function.** Let  $P=(X,D,C)$  be a CCSP (with  $X=\langle x_1, \dots, x_i, \dots, x_n \rangle$ ). A box-narrowing function with respect to a constraint  $(s, \rho) \in C$  and a variable  $x_i \in s$  is a mapping, denoted  $\text{BNF}_{x_i}^\rho$ , that relates any  $F$ -box  $B=\langle I_{x_1}, \dots, I_{x_i}, \dots, I_{x_n} \rangle$  ( $B \subseteq D$ ) with the union of  $m$  ( $1 \leq m$ )  $F$ -boxes, defined by:

$$\text{BNF}_{x_i}^\rho(\langle I_{x_1}, \dots, I_{x_i}, \dots, I_{x_n} \rangle) = \langle I_{x_1}, \dots, I_1, \dots, I_{x_n} \rangle \cup \dots \cup \langle I_{x_1}, \dots, I_m, \dots, I_{x_n} \rangle$$

satisfying the property:

$$\pi_{x_i}^\rho(B[s]) \subseteq I_1 \cup \dots \cup I_m \subseteq I_{x_i}$$

□

Contractance follows from  $I_1 \cup \dots \cup I_m \subseteq I_{x_i}$  (the only changed domain is smaller than the original)

Correctness follows from  $\pi_{x_i}^\rho(B[s]) \subseteq I_1 \cup \dots \cup I_m$  (the eliminated combinations have  $x_i$  values outside the projection function)

# Constraint Decomposition Method

Decomposition of complex constraints into an equivalent set of primitive constraints whose projection functions can be easily computed by inverse functions

## Primitive Constraints

**Primitive Constraint.** Let  $e_c$  be a real expression with at most one basic operator and with no multiple occurrences of its variables. Let  $e_\theta$  be a real constant or a real variable not appearing in  $e_c$ . The constraint  $c$  is a primitive constraint iff it is expressed as:

$$e_c \diamond e_\theta \quad \text{with } \diamond \in \{\leq, =, \geq\}$$



A set of primitive constraints can be easily obtained from any non-primitive constraint:

A constraint may be decomposed by considering new variables and new equality constraints

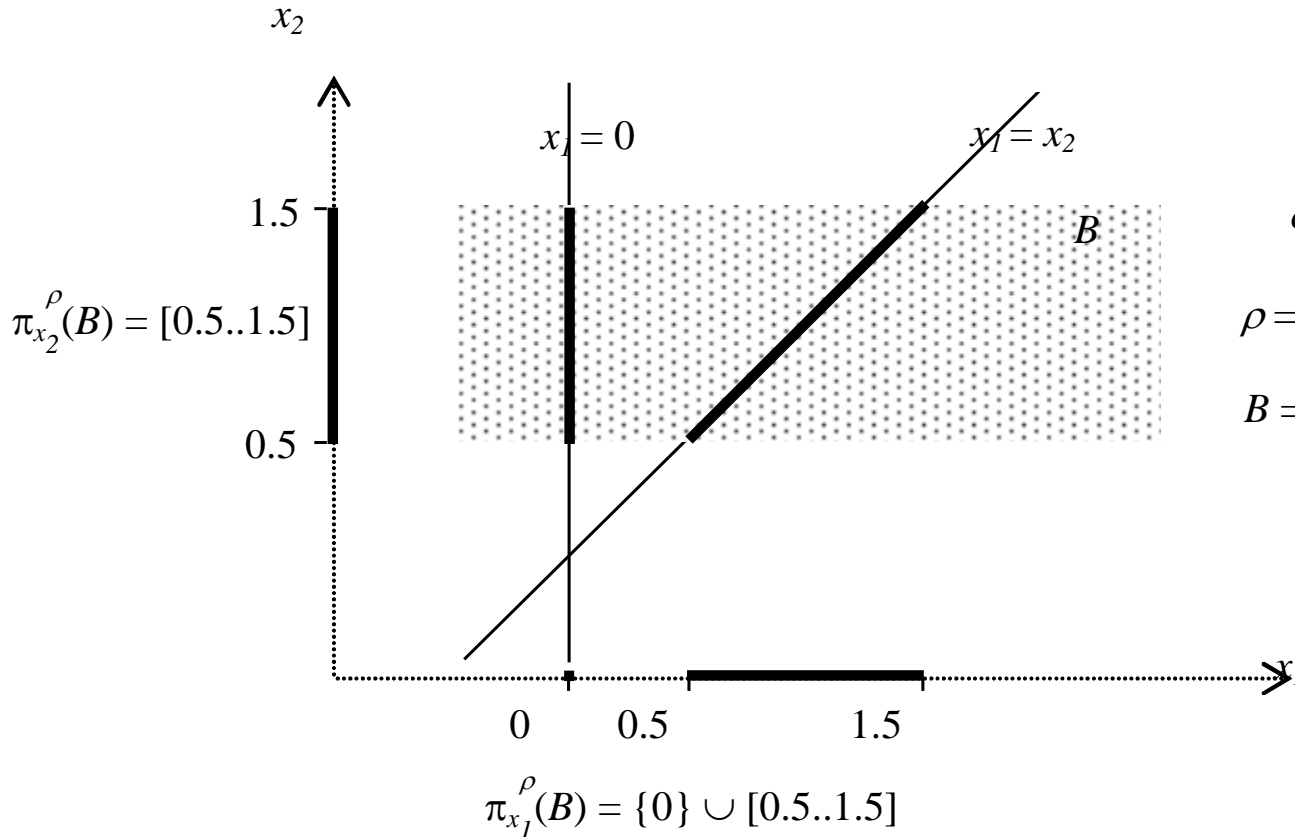
The whole set of primitives may be obtained by repeating this process until all constraints are primitive

# Constraint Decomposition Method

## Primitive Constraints

$$P = (X, D, C) = (\langle x_1, x_2 \rangle, D_1 \times D_2, \{c\})$$

$$c \equiv x_1 \times (x_2 - x_1) = 0$$



$$c = (\langle x_1, x_2 \rangle, \rho)$$

$$\rho = \{ \langle x_1, x_2 \rangle \in D \mid x_1 \times (x_2 - x_1) = 0 \}$$

$$B = \langle [-0.5, 2.5], [0.5, 1.5] \rangle$$

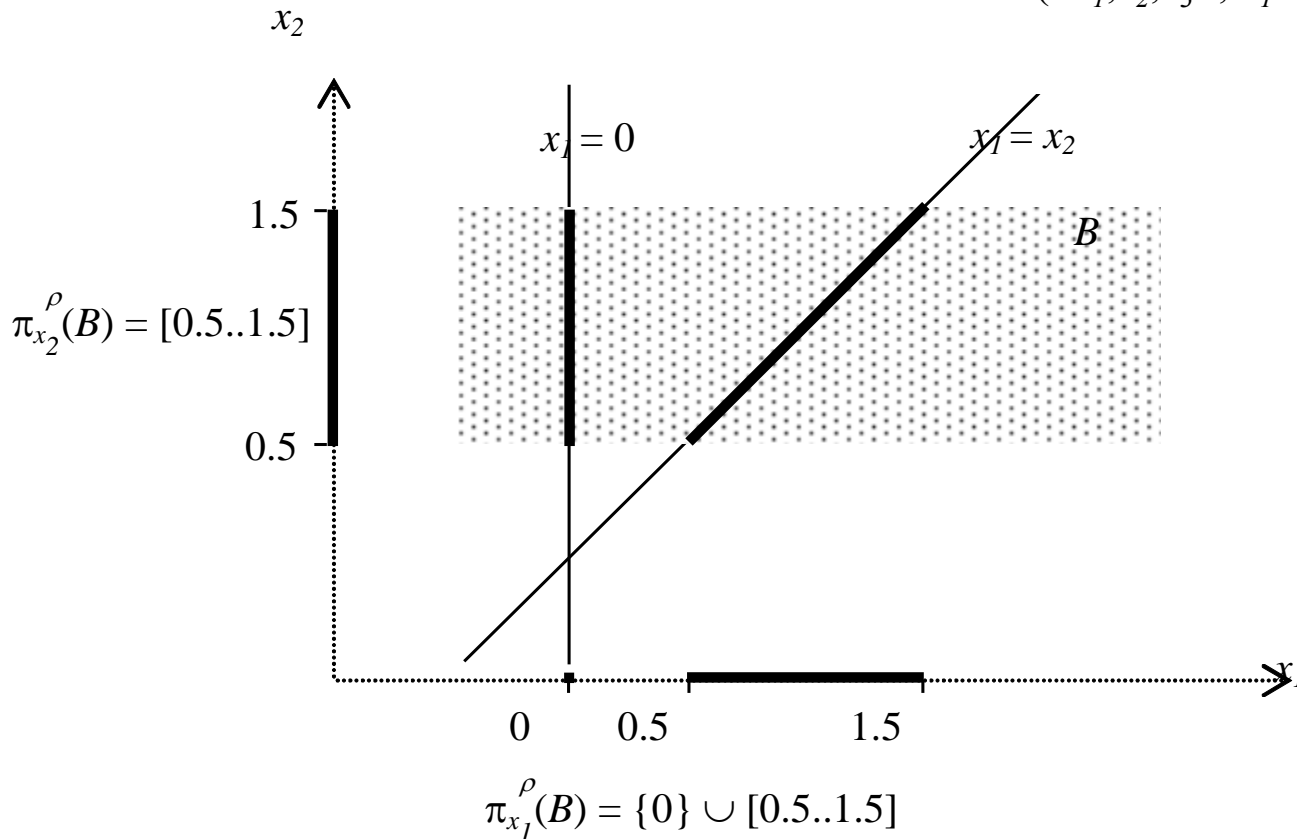
The constraint  $c$  is not primitive since it contains two basic arithmetic operators and the variable  $x_1$  occurs twice

# Constraint Decomposition Method

## Primitive Constraints

$$P' = (\langle x_1, x_2, x_3 \rangle, D_1 \times D_2 \times [-\infty..+\infty], \{c_1, c_2\})$$

$$c \begin{cases} c_1 \equiv x_1 \times x_3 = 0 \\ c_2 \equiv x_2 - x_1 = x_3 \end{cases}$$



a new variable  $x_3$  is introduced and  $c$  is replaced by  $c_1$  and  $c_2$   
 the domain of  $x_3$  is unbounded defining a new equivalent CCSP  $P'$



# Constraint Decomposition Method

## Inverse Functions

**Inverse Interval Expression.** Let  $c=(s,\rho)$  be a primitive constraint expressed in the form  $e_c \diamond e_0$  where  $e_c \equiv e_1$  or  $e_c \equiv \Phi(e_1, \dots, e_m)$  ( $\Phi$  is an  $m$ -ary basic operator and  $e_i$  a variable from  $s$  or a real constant). The inverse interval expression of  $c$  with respect to  $e_i$ , denoted  $\Psi e_i$ , is the natural interval expression of the expression obtained by solving algebraically, wrt  $e_i$ , the equality  $e_c = e_0$  if  $c$  is an equality or  $e_c = e_0 + k$  if  $c$  is an inequality (with  $k \leq 0$  for inequalities of the form  $e_c \leq e_0$  and  $k \geq 0$  for inequalities of the form  $e_c \geq e_0$ ).  $\square$

	$\Psi e_1$	$\Psi e_2$	$\Psi e_3$
$e_1 + e_2 \diamond e_3$	$(E_3 + K) - E_2$	$(E_3 + K) - E_1$	$(E_1 + E_2) - K$
$e_1 - e_2 \diamond e_3$	$(E_3 + K) + E_2$	$E_1 - (E_3 + K)$	$(E_1 - E_2) - K$
$e_1 \times e_2 \diamond e_3$	$(E_3 + K) / E_2$	$(E_3 + K) / E_1$	$(E_1 \times E_2) - K$
$e_1 / e_2 \diamond e_3$	$(E_3 + K) \times E_2$	$E_1 / (E_3 + K)$	$(E_1 / E_2) - K$
$e_1 \diamond e_2$	$(E_2 + K)$	$E_1 - K$	

$$\diamond \in \{\leq, =, \geq\}$$

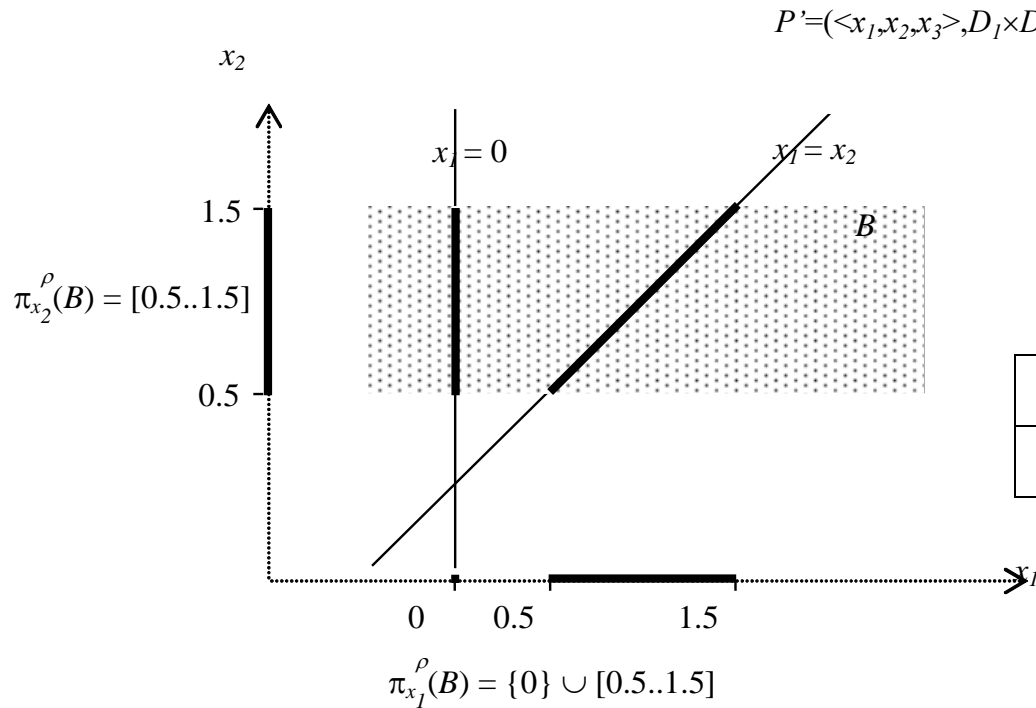
$e_i$  is a real variable or a real constant

$E_i$  is the natural interval extension of  $e_i$

$$K = \begin{cases} [-\infty..0] & \text{if } \diamond \equiv \leq \\ [0..0] & \text{if } \diamond \equiv = \\ [0..+\infty] & \text{if } \diamond \equiv \geq \end{cases}$$

# Constraint Decomposition Method

## Inverse Functions



	$\Psi e_1$	$\Psi e_2$	$\Psi e_3$
$x_1 \times x_3 = 0$	$0/X_3$	$0/X_1$	$X_1 \times X_3$
$x_2 - x_1 = x_3$	$X_3 + X_1$	$X_2 - X_3$	$X_2 - X_1$

The inverse interval expressions are associated with the primitive constraints of the decomposed CCSP  $P'$

# Constraint Decomposition Method

## Projection Function Enclosure with the Inverse Function

The inverse interval expression wrt a variable allows the definition of the projection function of the constraint wrt to that variable

**Projection Function based on the Inverse Interval Expression.** Let  $P=(X,D,C)$  be a CCSP. Let  $c=(s,\rho)\in C$  be an  $n$ -ary primitive constraint expressed in the form  $e_c\Diamond e_0$  where  $e_c\equiv e_1$  or  $e_c\equiv\Phi(e_1,\dots,e_m)$  (with  $\Phi$  an  $m$ -ary basic operator and  $e_i$  a variable from  $s$  or a real constant). Let  $\Psi_{x_i}$  be the inverse interval expression of  $c$  with respect to the variable  $x_i$  ( $e_i\equiv x_i$ ). The projection function  $\pi_{x_i}^\rho$  of the constraint  $c$  wrt variable  $x_i$  is the mapping:

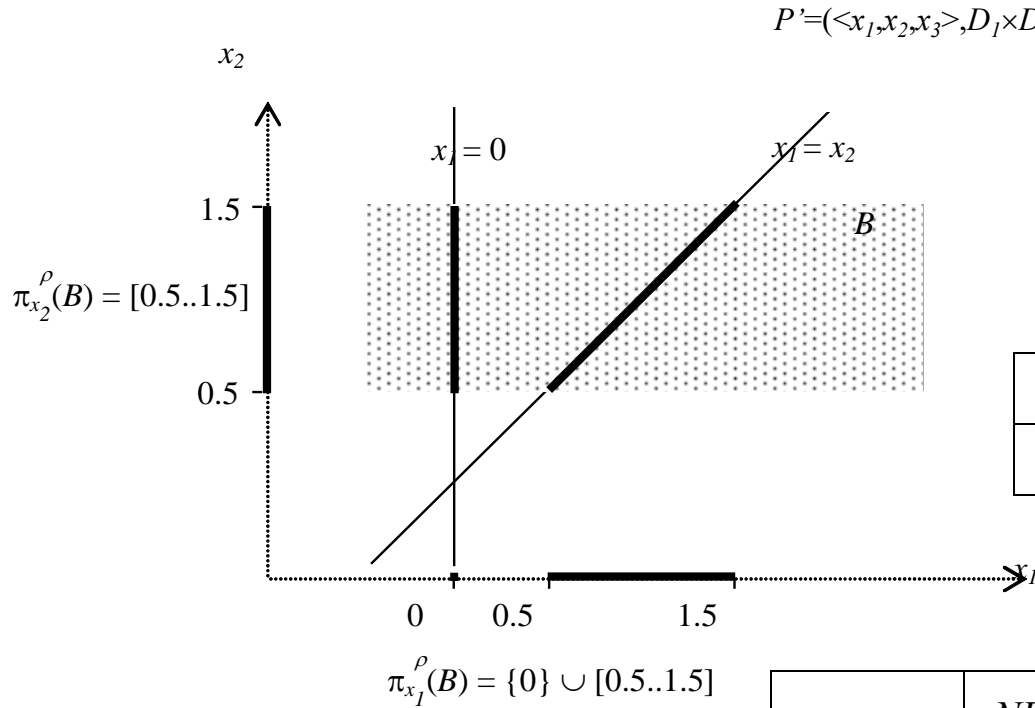
$$\pi_{x_i}^\rho(B) = \Psi_{x_i}(B) \cap B[x_i] \quad \text{where } B \text{ is an } n\text{-ary real box} \quad \square$$

$x_1 \times x_3 = 0$
$\pi_{x_1}^\rho(\langle I_1, I_3 \rangle) = (0/I_3) \cap I_1$
$\pi_{x_3}^\rho(\langle I_1, I_3 \rangle) = (0/I_1) \cap I_3$

$x_2 - x_1 = x_3$
$\pi_{x_1}^\rho(\langle I_1, I_2, I_3 \rangle) = (I_2 - I_3) \cap I_1$
$\pi_{x_2}^\rho(\langle I_1, I_2, I_3 \rangle) = (I_3 + I_1) \cap I_2$
$\pi_{x_3}^\rho(\langle I_1, I_2, I_3 \rangle) = (I_2 - I_1) \cap I_3$

# Constraint Decomposition Method

## Projection Function Enclosure with the Inverse Function



	$\Psi e_1$	$\Psi e_2$	$\Psi e_3$
$x_1 \times x_3 = 0$	$0/X_3$	$0/X_1$	$X_1 \times X_3$
$x_2 - x_1 = x_3$	$X_3 + X_1$	$X_2 - X_3$	$X_2 - X_1$

Box-narrowing functions are associated with the decomposed CCSP  $P'$

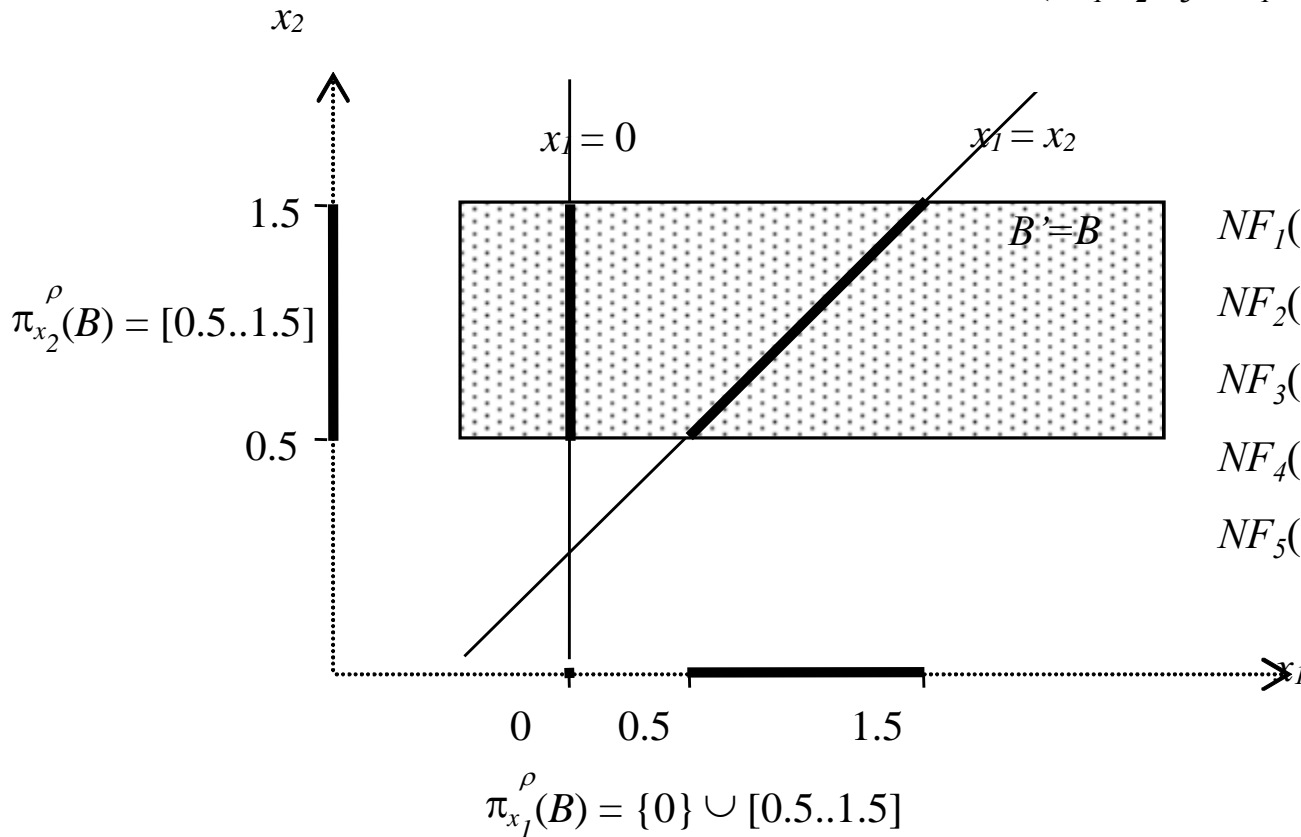
$x_1 \times x_3 = 0$	$NF_1$	$BNF_{x_1}^\rho(\langle I_1, I_2, I_3 \rangle) = \langle (0/I_3) \cap I_1, I_2, I_3 \rangle$
	$NF_2$	$BNF_{x_3}^\rho(\langle I_1, I_2, I_3 \rangle) = \langle I_1, I_2, (0/I_1) \cap I_3 \rangle$
$x_2 - x_1 = x_3$	$NF_3$	$BNF_{x_1}^\rho(\langle I_1, I_2, I_3 \rangle) = \langle (I_2 - I_3) \cap I_1, I_2, I_3 \rangle$
	$NF_4$	$BNF_{x_2}^\rho(\langle I_1, I_2, I_3 \rangle) = \langle I_1, (I_3 + I_1) \cap I_2, I_3 \rangle$
	$NF_5$	$BNF_{x_3}^\rho(\langle I_1, I_2, I_3 \rangle) = \langle I_1, I_2, (I_2 - I_1) \cap I_3 \rangle$

# Constraint Decomposition Method

## Example

$$P' = (\langle x_1, x_2, x_3 \rangle, D_1 \times D_2 \times [-\infty..+\infty], \{c_1, c_2\})$$

$$c \begin{cases} c_1 \equiv x_1 \times x_3 = 0 \\ c_2 \equiv x_2 - x_1 = x_3 \end{cases}$$



$$NF_1(\langle I_1, I_2, I_3 \rangle) = \langle (0/I_3) \cap I_1, I_2, I_3 \rangle$$

$$NF_2(\langle I_1, I_2, I_3 \rangle) = \langle I_1, I_2, (0/I_1) \cap I_3 \rangle$$

$$NF_3(\langle I_1, I_2, I_3 \rangle) = \langle (I_2 - I_3) \cap I_1, I_2, I_3 \rangle$$

$$NF_4(\langle I_1, I_2, I_3 \rangle) = \langle I_1, (I_3 + I_1) \cap I_2, I_3 \rangle$$

$$NF_5(\langle I_1, I_2, I_3 \rangle) = \langle I_1, I_2, (I_2 - I_1) \cap I_3 \rangle$$

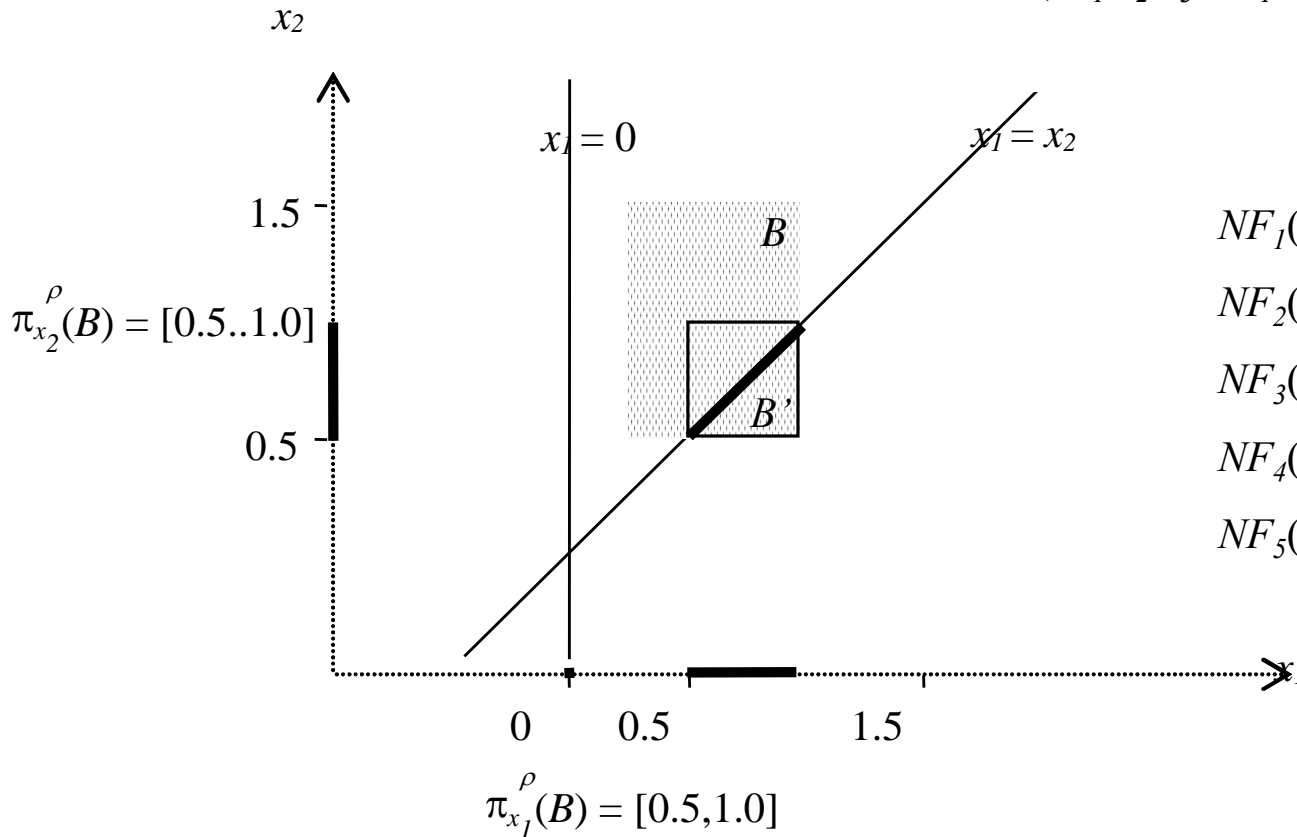
with  $B = \langle [-0.5, 2.5], [0.5, 1.5] \rangle$  no pruning would be obtained:  
 $B' = \langle [-0.5, 2.5], [0.5, 1.5] \rangle$  and  $x_3 = [-2.0, 2.0]$

# Constraint Decomposition Method

## Example

$$P' = (\langle x_1, x_2, x_3 \rangle, D_1 \times D_2 \times [-\infty..+\infty], \{c_1, c_2\})$$

$$c \begin{cases} c_1 \equiv x_1 \times x_3 = 0 \\ c_2 \equiv x_2 - x_1 = x_3 \end{cases}$$



$$NF_1(\langle I_1, I_2, I_3 \rangle) = \langle (0/I_3) \cap I_1, I_2, I_3 \rangle$$

$$NF_2(\langle I_1, I_2, I_3 \rangle) = \langle I_1, I_2, (0/I_1) \cap I_3 \rangle$$

$$NF_3(\langle I_1, I_2, I_3 \rangle) = \langle (I_2 - I_3) \cap I_1, I_2, I_3 \rangle$$

$$NF_4(\langle I_1, I_2, I_3 \rangle) = \langle I_1, (I_3 + I_1) \cap I_2, I_3 \rangle$$

$$NF_5(\langle I_1, I_2, I_3 \rangle) = \langle I_1, I_2, (I_2 - I_1) \cap I_3 \rangle$$

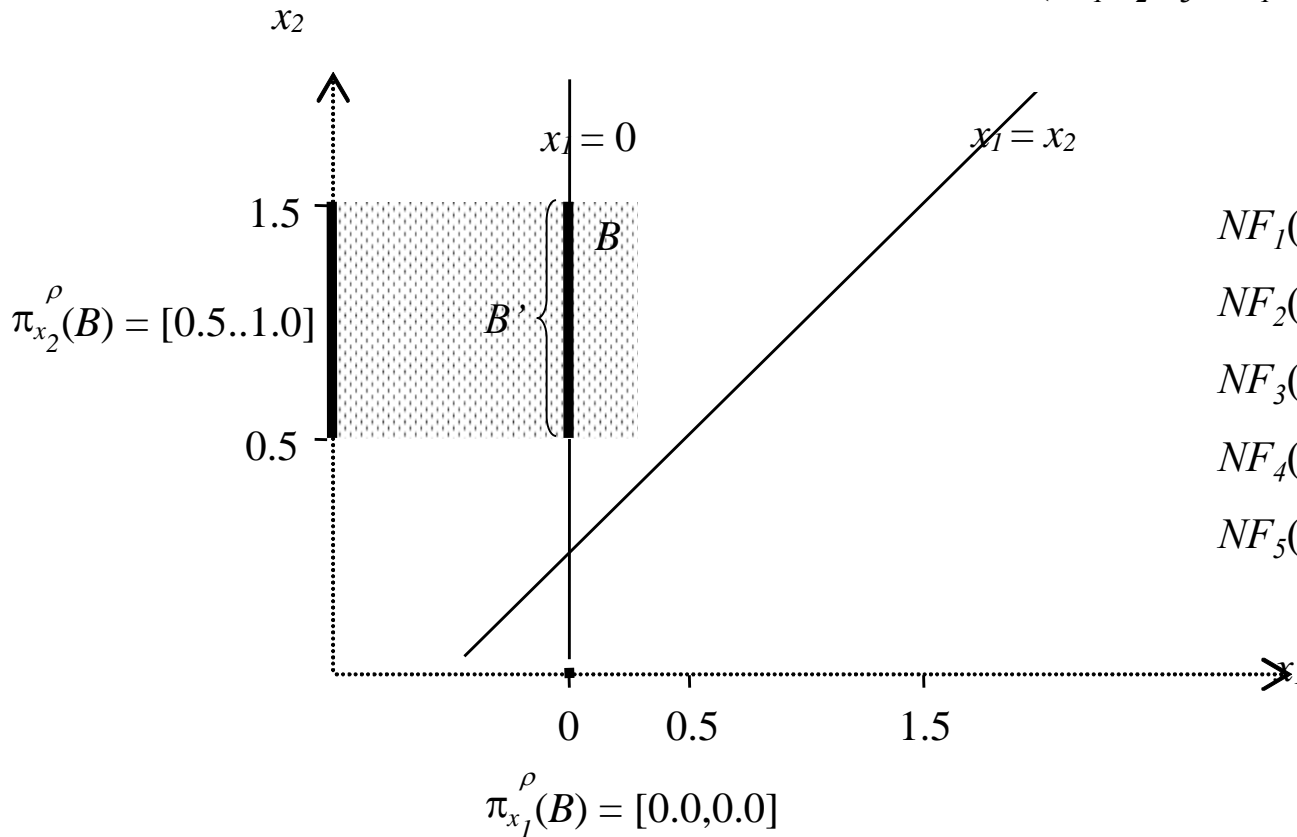
with  $B = \langle [0.25, 1.0], [0.5, 1.5] \rangle$  the best narrowing is obtained:  
 $B' = \langle [0.5, 1.0], [0.5, 1.0] \rangle$  and  $x_3 = [0.0, 0.0]$

# Constraint Decomposition Method

## Example

$$P' = (\langle x_1, x_2, x_3 \rangle, D_1 \times D_2 \times [-\infty..+\infty], \{c_1, c_2\})$$

$$c \begin{cases} c_1 \equiv x_1 \times x_3 = 0 \\ c_2 \equiv x_2 - x_1 = x_3 \end{cases}$$



$$NF_1(\langle I_1, I_2, I_3 \rangle) = \langle (0/I_3) \cap I_1, I_2, I_3 \rangle$$

$$NF_2(\langle I_1, I_2, I_3 \rangle) = \langle I_1, I_2, (0/I_1) \cap I_3 \rangle$$

$$NF_3(\langle I_1, I_2, I_3 \rangle) = \langle (I_2 - I_3) \cap I_1, I_2, I_3 \rangle$$

$$NF_4(\langle I_1, I_2, I_3 \rangle) = \langle I_1, (I_3 + I_1) \cap I_2, I_3 \rangle$$

$$NF_5(\langle I_1, I_2, I_3 \rangle) = \langle I_1, I_2, (I_2 - I_1) \cap I_3 \rangle$$

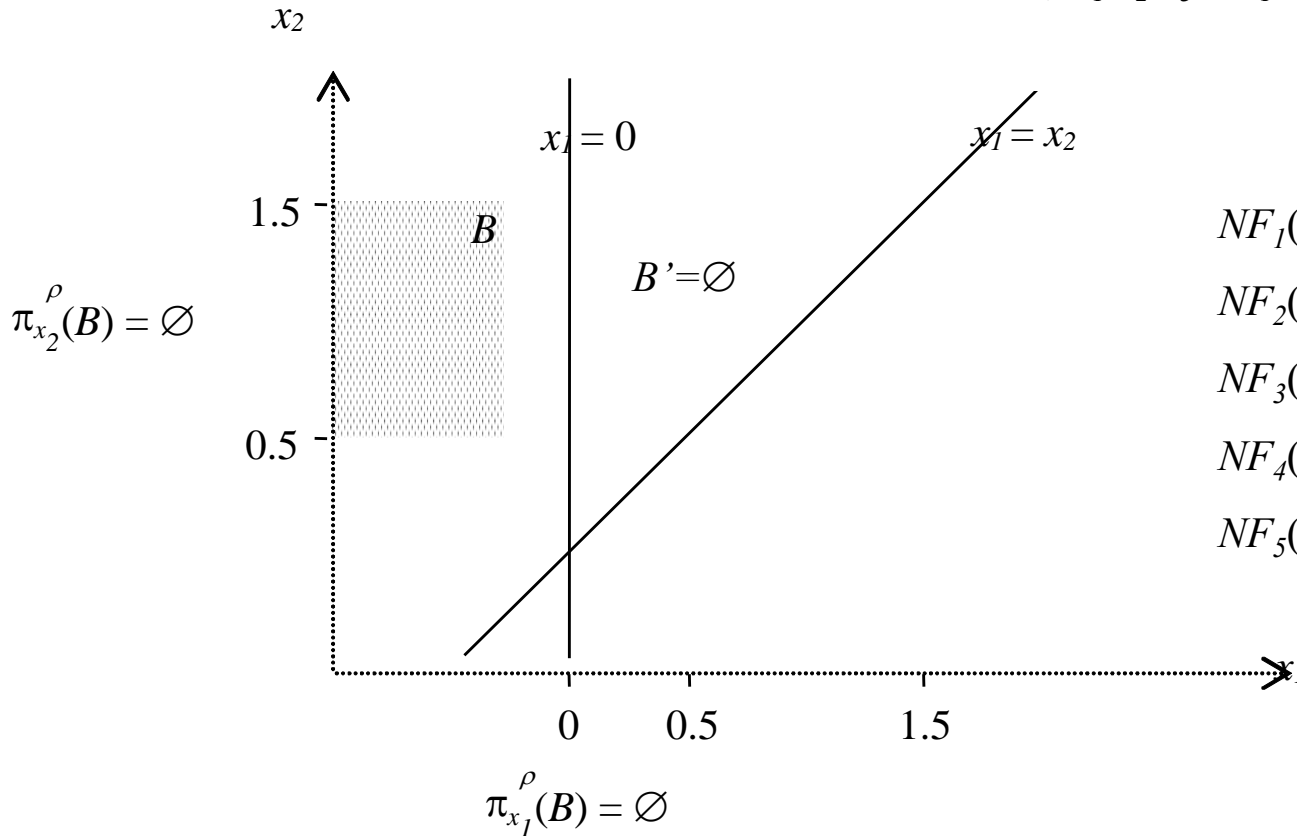
with  $B = \langle [-1.0, 0.25], [0.5, 1.5] \rangle$  the best narrowing is also obtained:  $B = \langle [0.0, 0.0], [0.5, 1.5] \rangle$  and  $x_3 = [0.5, 1.5]$

# Constraint Decomposition Method

## Example

$$P' = (\langle x_1, x_2, x_3 \rangle, D_1 \times D_2 \times [-\infty..+\infty], \{c_1, c_2\})$$

$$c \begin{cases} c_1 \equiv x_1 \times x_3 = 0 \\ c_2 \equiv x_2 - x_1 = x_3 \end{cases}$$



$$NF_1(\langle I_1, I_2, I_3 \rangle) = \langle (0/I_3) \cap I_1, I_2, I_3 \rangle$$

$$NF_2(\langle I_1, I_2, I_3 \rangle) = \langle I_1, I_2, (0/I_1) \cap I_3 \rangle$$

$$NF_3(\langle I_1, I_2, I_3 \rangle) = \langle (I_2 - I_3) \cap I_1, I_2, I_3 \rangle$$

$$NF_4(\langle I_1, I_2, I_3 \rangle) = \langle I_1, (I_3 + I_1) \cap I_2, I_3 \rangle$$

$$NF_5(\langle I_1, I_2, I_3 \rangle) = \langle I_1, I_2, (I_2 - I_1) \cap I_3 \rangle$$

with  $B = \langle [-1.0, -0.25], [0.5, 1.5] \rangle$  inconsistency is proved:

$$B' = \emptyset$$



# Constraint Newton Method

Complex constraints are handled without decomposition using a technique based on the interval Newton method for searching the zeros of univariate functions

## Interval Projections

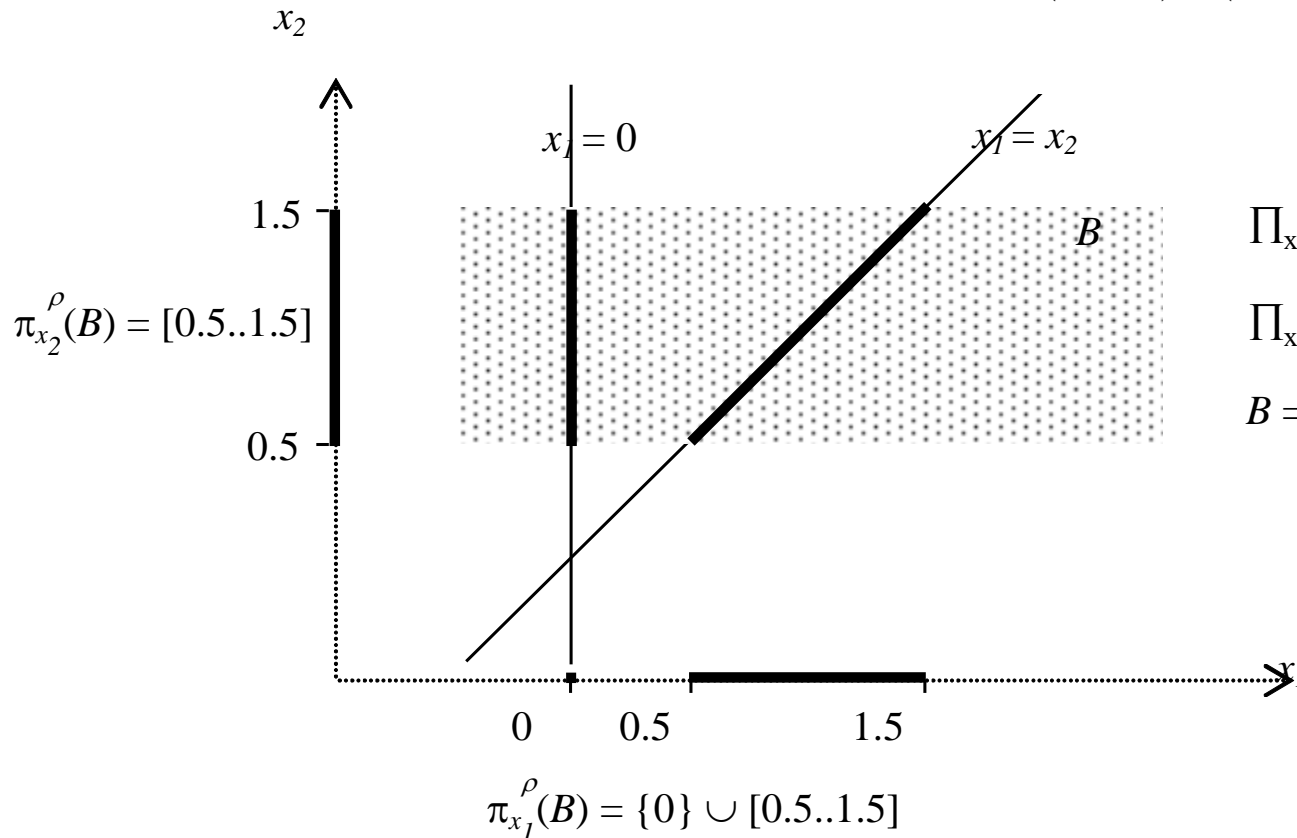
**Interval Projection.** Let  $P=(X,D,C)$  be a CCSP. Let  $c=(s,\rho)\in C$  be an  $n$ -ary constraint expressed in the form  $e_c \diamond 0$  (with  $\diamond \in \{\leq, =, \geq\}$  and  $e_c$  a real expression). Let  $B$  be an  $n$ -ary  $F$ -box. The interval projection of  $c$  wrt  $x_i \in s$  and  $B$  is the function, denoted  $\prod_{x_i}^{\rho B}$ , represented by the expression obtained by replacing in  $e_c$  each real variable  $x_j$  ( $x_j \neq x_i$ ) by the interval constant  $B[x_j]$ .  $\square$

# Constraint Newton Method

## Interval Projections

$$P = (X, D, C) = (\langle x_1, x_2 \rangle, D_1 \times D_2, \{c\})$$

$$c \equiv x_1 \times (x_2 - x_1) = 0$$



$$\Pi_{x_1}^{\rho B} \equiv x_1 \times ([0.5..1.5] - x_1)$$

$$\Pi_{x_2}^{\rho B} \equiv [-0.5..2.5] \times (x_2 - [-0.5..2.5])$$

$$B = \langle [-0.5..2.5], [0.5..1.5] \rangle$$

All value combinations within  $B$  with  $x_i$  values outside  $\pi_{x_i}^\rho(B)$  are outside the relation  $\rho$  and so they do not satisfy the constraint  $c$ .

# Constraint Newton Method

## Properties of an Interval Projection

From the properties of the interval projections, a strategy is devised for obtaining an enclosure of the projection function

**Properties of the Interval Projection.** Let  $P=(X,D,C)$  be a CCSP. Let  $c=(s,\rho)\in C$  be an  $n$ -ary constraint expressed in the form  $e_c \diamond 0$  (with  $\diamond \in \{\leq, =, \geq\}$  and  $e_c$  a real expression) and  $B$  an  $n$ -ary  $F$ -box. Let  $\prod_{x_i}^{\rho B}$  be the interval projection of  $c$  wrt variable  $x_i \in s$  and  $B$ . The following property is necessarily satisfied:

$$\forall r \in B[x_i] \ r \in \pi_{x_i}^{\rho}(B) \Rightarrow \exists v \in \prod_{x_i}^{\rho B}([r]): v \diamond 0$$

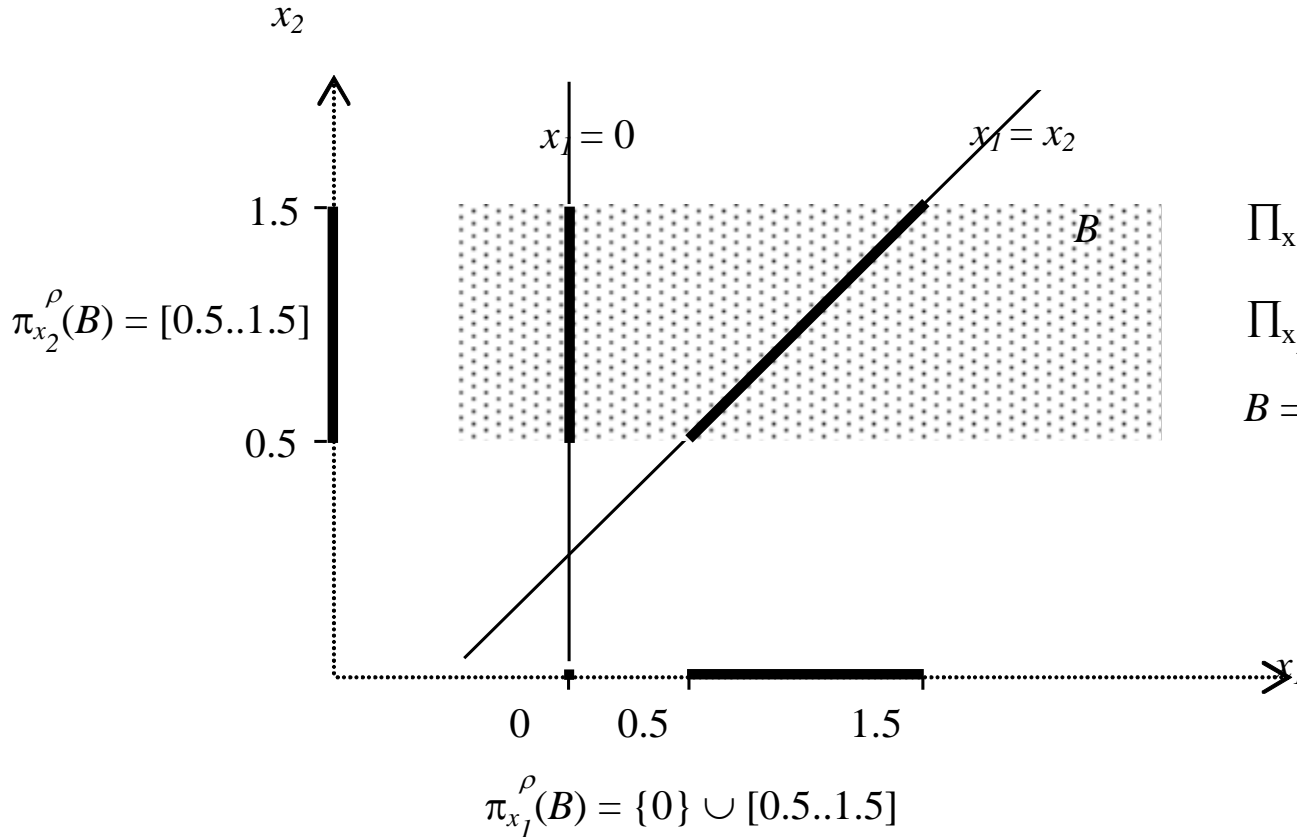
We will say that a real value  $r$  satisfies the interval projection condition if the right side of the implication is satisfied. □

# Constraint Newton Method

## Properties of an Interval Projection

$$P = (X, D, C) = (\langle x_1, x_2 \rangle, D_1 \times D_2, \{c\})$$

$$c \equiv x_1 \times (x_2 - x_1) = 0$$



$$\Pi_{x_1}^{\rho B} \equiv x_1 \times ([0.5..1.5] - x_1)$$

$$\Pi_{x_2}^{\rho B} \equiv [-0.5..2.5] \times (x_2 - [-0.5..2.5])$$

$$B = \langle [-0.5..2.5], [0.5..1.5] \rangle$$

$$\forall_{r \in [-0.5, 2.5]} r \in \pi_{x_1}^\rho(B) \Rightarrow \exists v \in [r] \times ([0.5..1.5] - [r]): v = 0$$

$$\forall_{r \in [0.5, 1.5]} r \in \pi_{x_2}^\rho(B) \Rightarrow \exists v \in [-0.5..2.5] \times ([r] - [-0.5..2.5]): v = 0$$

# Constraint Newton Method

## Projection Function Enclosure with the Interval Projection

The strategy used in the constraint Newton method is to search for the leftmost and the rightmost elements of the original variable domain satisfying the interval projection condition

**Projection Function Enclosure based on the Interval Projection.** Let  $P=(X,D,C)$  be a CCSP. Let  $c=(s,\rho)\in C$  be an  $n$ -ary constraint,  $B$  an  $n$ -ary  $F$ -box and  $x_i$  an element of  $s$ . Let  $a$  and  $b$  be respectively the leftmost and the rightmost elements of  $B[x_i]$  satisfying the interval projection condition. The following property necessarily holds:

$$\pi_{x_i}^{\rho}(B) \subseteq [a..b]$$

□

What is needed is a function, denoted *narrowBounds*, with the following property:

$$\pi_{x_i}^{\rho}(B) \subseteq [a..b] \subseteq \text{narrowBounds}(B[x_i])$$

# Constraint Newton Method

## Projection Function Enclosure with the Interval Projection

To obtain a new bound, the projection condition is firstly verified in the extreme of the original domain and only in case of failure the leftmost (rightmost) zero of the interval projection is searched

```
function narrowBounds(an F-interval [a..b])
  (1)  if a = b then if intervalProjCond([a]) then return [a] else return ∅; end if; end if;
  (2)  if not intervalProjCond([a..a+]) then a ← searchLeft([a+..b]);
  (3)  if a = ∅ then return ∅;
  (4)  if a = b then return [b];
  (5)  if not intervalProjCond([b-..b]) then b ← searchRight([a..b-]);
  (6)  return [a..b];
end function
```

In case of failure of an inequality condition, it assumes that the leftmost (rightmost) element satisfying the interval projection condition must be a zero of the interval projection

# Constraint Newton Method

## Projection Function Enclosure with the Interval Projection

The verification if the interval projection condition is satisfied in a canonical interval is straightforward

```
function intervalProjCond(a canonical  $F$ -interval  $I$ )  
  (1)  $[a..b] \leftarrow \prod_{x_i}^{\rho B}(I);$   
  (2) case  $\diamond$  of  
  (3)   “=”: return  $0 \in [a..b];$   
  (4)   “ $\leq$ ”: return  $a \leq 0;$   
  (5)   “ $\geq$ ”: return  $b \geq 0;$   
  (6) end case;  
end function
```

# Constraint Newton Method

## Projection Function Enclosure with the Interval Projection

The algorithm for searching for the leftmost zero of an interval projection uses a Newton Narrowing function ( $NN$ ) associated with the interval projection for reducing the search space

**function** *searchLeft*(an  $F$ -interval  $I$ )

- (1)  $Q \leftarrow \{I\};$
- (2) **while**  $Q \neq \emptyset$  **do**
- (3)     choose  $I_1 \in Q$  with the smallest left bound ( $\forall I \in Q \text{ left}(I_1) \leq \text{left}(I)$ );
- (4)      $Q \leftarrow Q \setminus \{I_1\};$
- (5)     **if**  $0 \in \prod_{x_i}^{\rho^A}(I_1)$  **then**
- (6)          $I_1 \leftarrow NN(I_1);$
- (7)         **if**  $I_1 \neq \emptyset$  **then**
- (8)              $I_0 \leftarrow \text{cleft}(I_1); I_1 \leftarrow [\text{right}(I_0).. \text{right}(I_1)];$
- (9)             **if**  $0 \in \prod_{x_i}^{\rho^B}(I_0)$  **then return**  $\text{left}(I_0);$
- (10)            **else**  $Q \leftarrow Q \cup \{[\text{left}(I_1).. \lfloor \text{center}(I_1) \rfloor], [\lfloor \text{center}(I_1) \rfloor.. \text{right}(I_1)]\};$  **end if;**
- (11)         **end if;**
- (12)     **end if;**
- (13) **end while;**
- (14) **return**  $\emptyset;$

**end function**

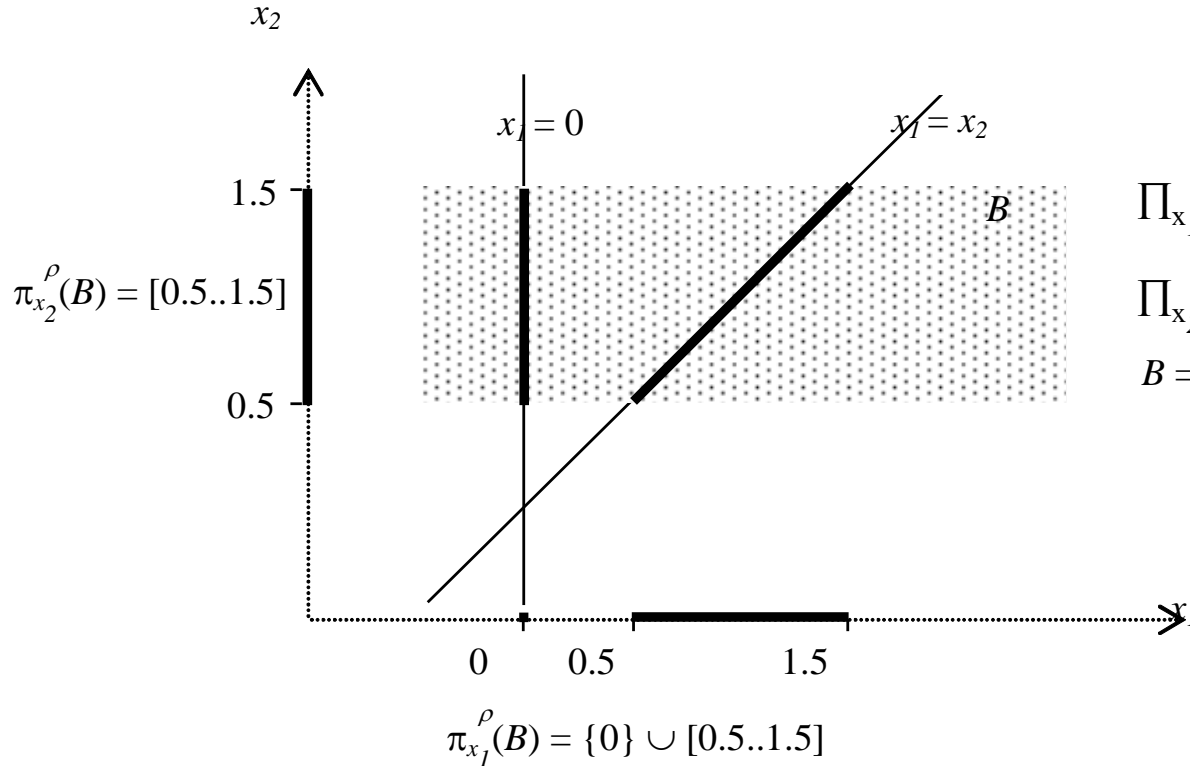


# Constraint Newton Method

## Example

$$P = (X, D, C) = (\langle x_1, x_2 \rangle, D_1 \times D_2, \{c\})$$

$$c \equiv x_1 \times (x_2 - x_1) = 0$$



$$\Pi_{x_1}^{\rho B} \equiv x_1 \times ([0.5..1.5] - x_1)$$

$$\Pi_{x_2}^{\rho B} \equiv [-0.5..2.5] \times (x_2 - [-0.5..2.5])$$

$$B = \langle [-0.5..2.5], [0.5..1.5] \rangle$$

Narrowing the domain of variable  $x_1$ : `narrowBounds([-0.5,2.5])`

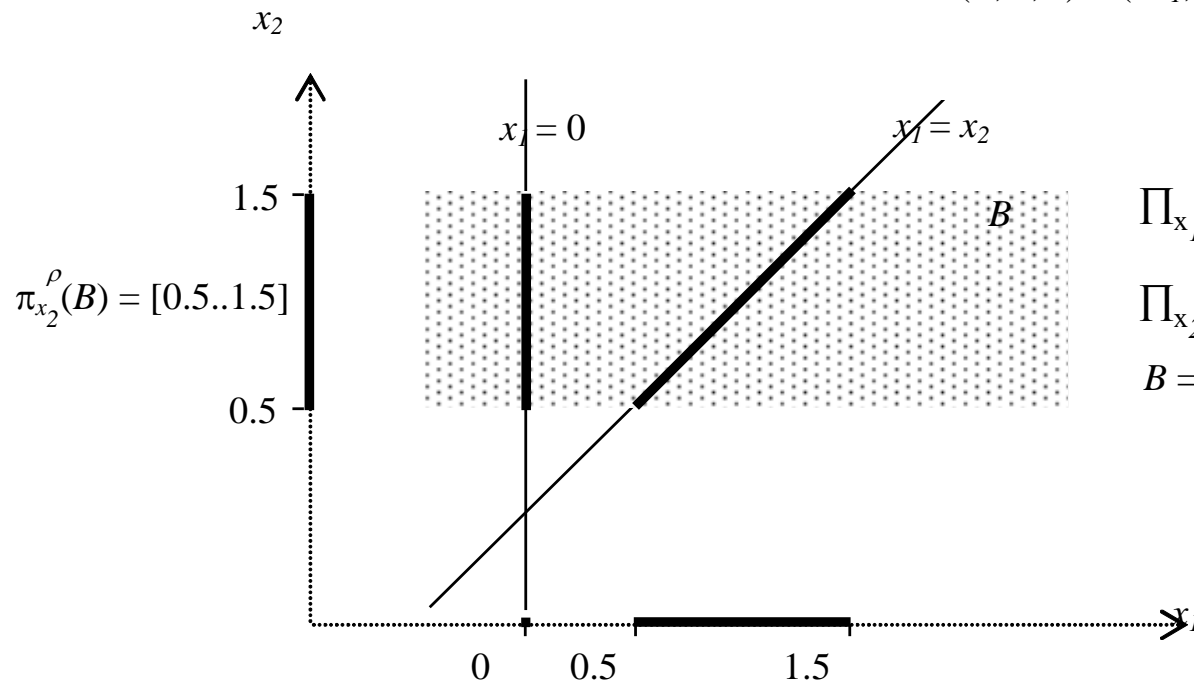
`intervalProjCond([-0.5,-0.499])`  $\rightarrow$  `False`  $0 \notin \Pi_{x_1}^{\rho B}([-0.5,-0.499]) = [-1,-0.499]$

# Constraint Newton Method

## Example

$$P = (X, D, C) = (\langle x_1, x_2 \rangle, D_1 \times D_2, \{c\})$$

$$c \equiv x_1 \times (x_2 - x_1) = 0$$



$$\Pi_{x_1}^{\rho B} \equiv x_1 \times ([0.5..1.5] - x_1)$$

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$$B = \langle [-0.5..2.5], [0.5..1.5] \rangle$$

$searchLeft([-0.499..2.5])$			
$Q = \{I_1, \dots, I_n\}$	$0 \in \Pi_{x_i}^{\rho B}(I_i)$	$NN(I_i)$	$0 \in \Pi_{x_i}^{\rho B}(I_i)$
$\{[-0.499..2.5]\}$	$0 \in [-5..4.998]$	$[-0.499..2.5]$	$0 \notin [-0.998.. -0.497]$
$\{[-0.498..1.001], [1.001..2.5]\}$	$0 \in [-0.995..2]$	$[-0.498..1.001]$	$0 \notin [-0.997.. -0.496]$
$\{[-0.497..0.252], [0.252..1.001], [1.001..2.5]\}$	$0 \in [-0.992..0.504]$	$[0..0.001]$	$0 \in [0..0.002]$

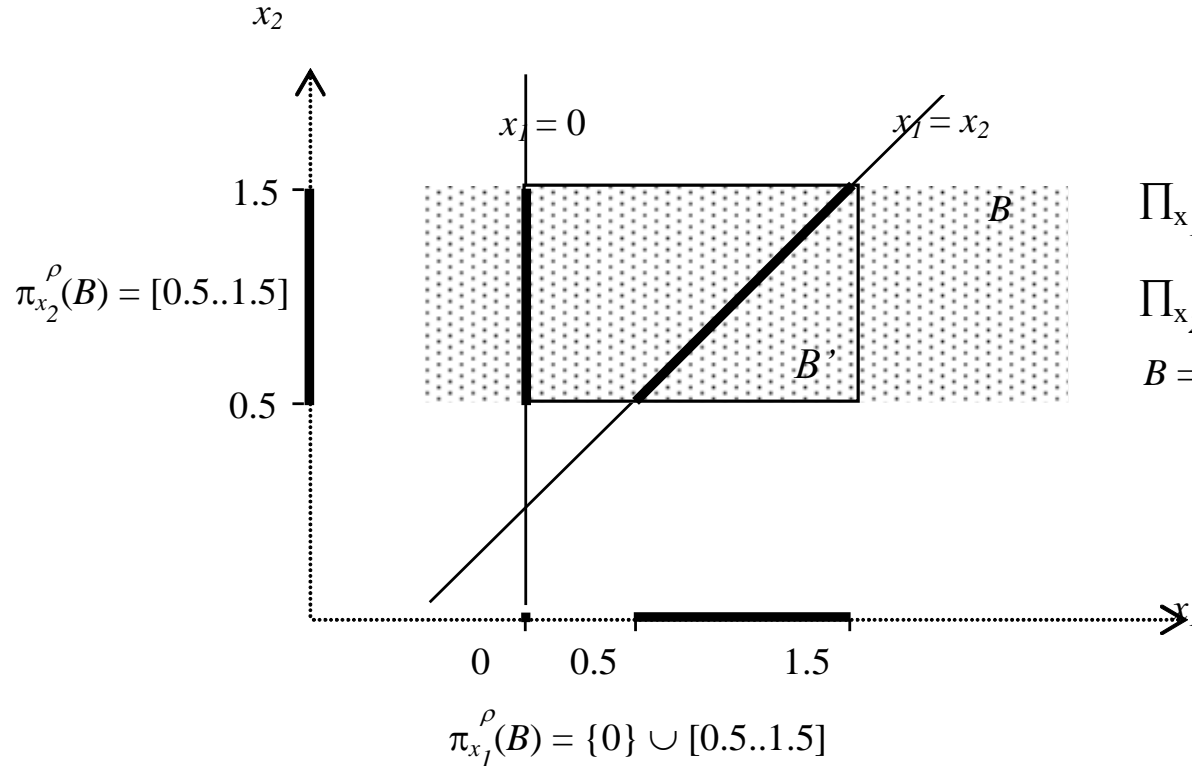
return 0

# Constraint Newton Method

## Example

$$P = (X, D, C) = (\langle x_1, x_2 \rangle, D_1 \times D_2, \{c\})$$

$$c \equiv x_1 \times (x_2 - x_1) = 0$$



$$\Pi_{x_1}^{\rho B} \equiv x_1 \times ([0.5..1.5] - x_1)$$

$$\Pi_{x_2}^{\rho B} \equiv [-0.5..2.5] \times (x_2 - [-0.5..2.5])$$

$$B = \langle [-0.5..2.5], [0.5..1.5] \rangle$$

Proceeding similarly for the upper bound of  $x_1$ , the best narrowing is obtained:

$$B' = \langle [0.0, 1.501], [0.5, 1.0] \rangle$$

# Complementary Approaches

A modification of the Newton method, is to use other interval extensions of the projection function associated with a constraint

A modification of the decomposition method, transforms the original set of constraints into an equivalent one where for each constraint (not necessarily primitive) the inverse interval expressions can be easily computed by interval arithmetic

Other modification is the introduction of a pre-processing phase preceding the definitions of the box-narrowing functions to obtain an equivalent CCSP (ex: introduction of redundant constraints)

# Complementary Approaches

Other variation is the development of an algorithm capable of implementing a narrowing function for any constraint without decomposing, with the same results as the decomposition method

A complementary approach take advantage of the way that a complex constraint is expressed: An algorithm that does not require decomposing complex constraints, makes it possible to combine both basic methods, and choose either one or the other, according to the form of the expression of the interval projection

Finally, some approaches consider narrowing functions capable of narrowing several variable domains simultaneously (ex: based on the multivariate interval Newton method)