

# Improving DPOP with Branch Consistency for Solving Distributed Constraint Optimization Problems

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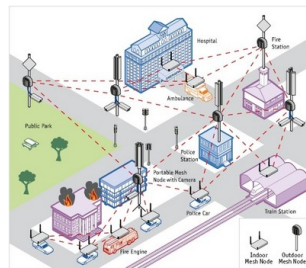
Sept. 9, 2014

# Talk Outline

- 1 Motivation and Background
- 2 Branch Consistency for Pseudo-Trees
- 3 Experiments and Results
- 4 Conclusions

# Distributed Optimization: Motivations

- Some problems cannot be realistically addressed in a centralized fashion.
- Agents cooperate to achieve a common objective.
- Simultaneously they can pursue private goals.
- Agents are constrained by limited communication capabilities.



Source: <http://kenanaonline.com/users/antennamaker>

# Distributed Optimization: Motivations

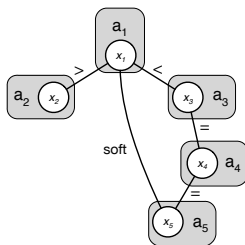
- Some problems cannot be realistically addressed in a centralized fashion.
- Agents cooperate to achieve a common objective.
- Simultaneously they can pursue private goals.
- Agents are constrained by limited communication capabilities.
- Solving time is important!



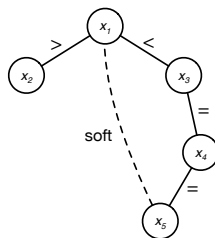
# Distributed Constrained Optimization (DCOP)

A *DCOP* is defined by a tuple  $\langle \mathcal{A}, \mathcal{X}, \mathcal{D}, \mathcal{F}, \alpha \rangle$ , where:

- $\mathcal{A}$  is a set of *agents*;
- $\mathcal{X}$  is a set of *variables*.
- $\mathcal{D}$  is a set of finite *domains*.
- $\mathcal{F}$  is a set of *utility functions*,  $f_i : \times_{x_j \in \text{scope}(f_i)} \mathcal{D}_j \mapsto \mathbb{N} \cup \{0, -\infty\}$ .
- $\alpha : \mathcal{X} \rightarrow \mathcal{A}$  maps each variable to one agent.



Constraint Graph



Pseudo-tree

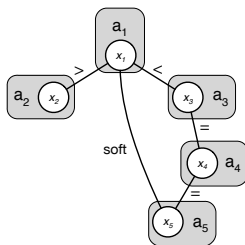
$x_1$	$x_5$	Utilities
0	0	20
0	1	8
0	2	10
0	3	3
...		
3	3	2

Soft Constraint Table

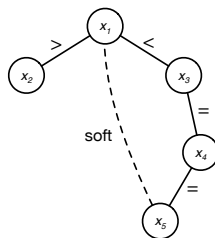
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Constraint Graph



Pseudo-tree

$x_1$	$x_3$	Utilities
0	0	$-\infty$
0	1	0
0	2	0
0	3	0
...		
3	3	$-\infty$

Hard Constraint Table

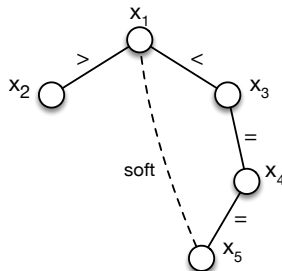
# Solving DCOP

- Find an utility maximal assignment for all the variables of the problem.
- Agents communicate exchanging messages.
- This is often the bottleneck!

# Solving DCOP (cont.)

## Distributed Pseudo-Tree Optimization Procedure (DPOP)

- 1 Pseudo-Tree Construction Phase.
- 2 UTIL propagation phase.
- 3 VALUE propagation phase.



UTIL Phase Computations of  $x_5$  ( $x_5$ ): UTIL Table

$x_1$	$x_4$	Utilities
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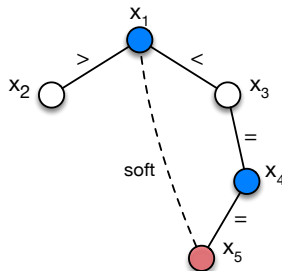
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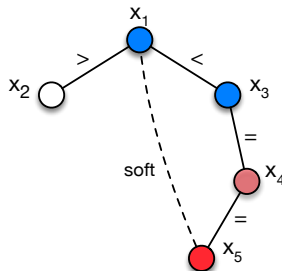
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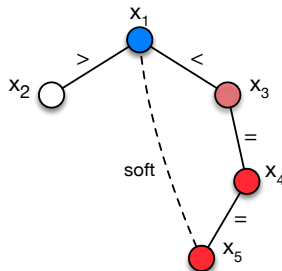
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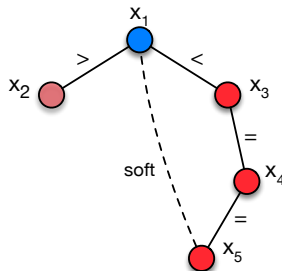
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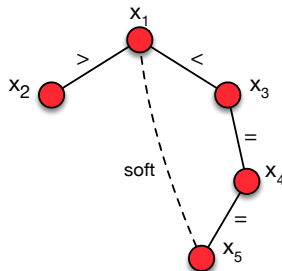
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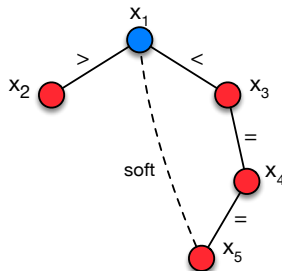
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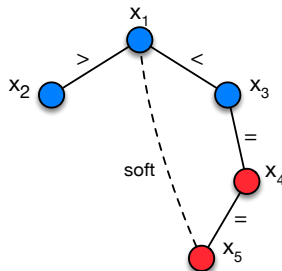
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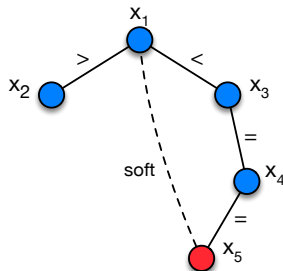
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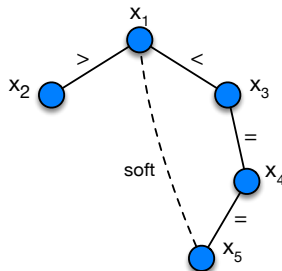
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# Distributed Optimization: Current Limitations

- 1 The DCOP community has not focused on exploiting hard constraints.<sup>1</sup>
- 2 Number of messages exchanged vs Message Size.

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<sup>1</sup>With few exceptions that do not take advantage of CP technologies.

# Distributed Optimization: Current Limitations

- ❶ The DCOP community has not focused on exploiting hard constraints.<sup>1</sup>
- ❷ Number of messages exchanged vs Message Size.
- ❸ Can we improve on these limitations and achieve better results?

---

<sup>1</sup>With few exceptions that do not take advantage of CP technologies.

# Exploiting Hard Constraints

- In the context of the **DPOP** algorithm.
- Assume that we are given a DCOP with hard constraints.
- First Trial: Integrating **Arc Consistency** to prune the UTIL table.

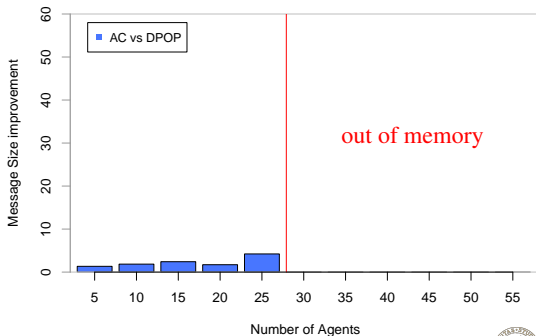
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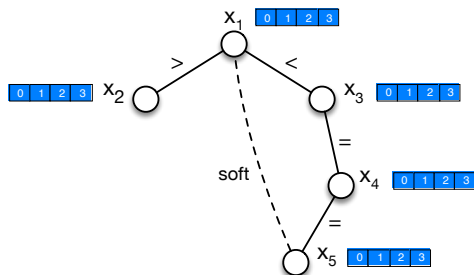
- Given a set of links.
- Assign them a frequency.
- The inference at the receivers:  $|x_i - x_j| > s$ .
- Use as few frequencies as possible.

Radio Frequency Assignment Problem



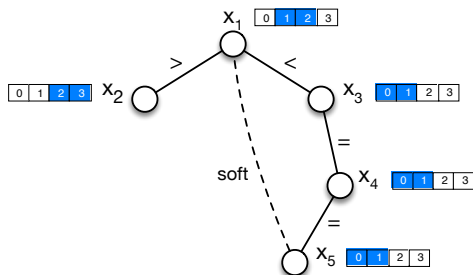
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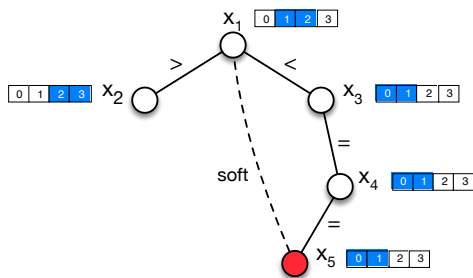
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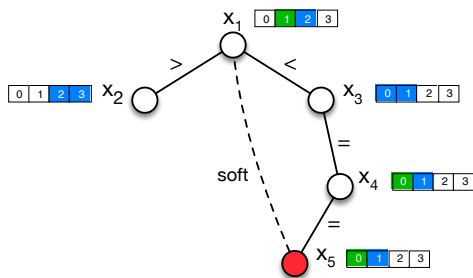
UTIL<sub>5</sub>

X5	X4	X1
0	0	1
0	0	2
0	1	1
0	1	2
1	0	1
1	0	2
1	1	1
1	1	2



# Exploiting Hard Constraints

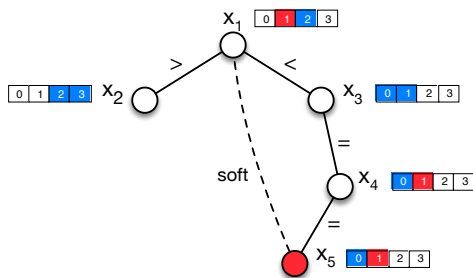
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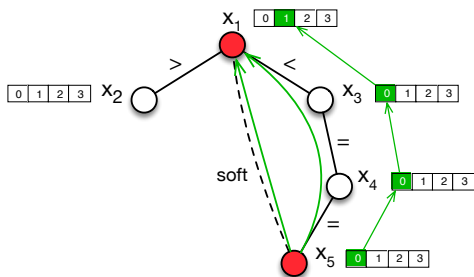


# Branch Consistency (BrC)

- Obs. 1: The domain store transmits a limited amount of information. It accounts for **no interaction among variables**.
- Obs. 2: A **Pseudo-Tree branch** contains the relevant information to build the UTIL tables.

# Branch Consistency (BrC) (cont.)

**Def. Branch Consistency** for pair of values:



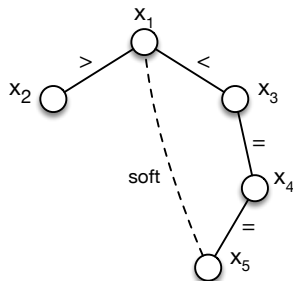
**Def.** A DCOP is **Branch Consistent (BrC)** iff for any pair of variables  $(x_i, x_j)$  with  $x_i$  and  $x_j$  in the same branch, and any  $(u, v) \in f_{ij}$ ,  $(u, v)$  is branch consistent.

**Def.** The **Value Reachability Matrix (VRM)**  $M_{ij}$  of  $x_i$  and  $x_j$ , with  $x_i$  ancestor of  $x_j$ , is a binary matrix of size  $|D_i| \times |D_j|$ , where  $M_{ij}[r, c] = 1$  iff  $(r, c)$  is BrC.

# BrC-DPOP

BrC-DPOP consists of 5 phases:

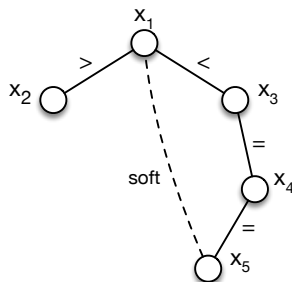
## 1. Pseudo-tree Generation Phase.



# BrC-DPOP

BrC-DPOP consists of 5 phases:

## 2. Path Construction Phase.

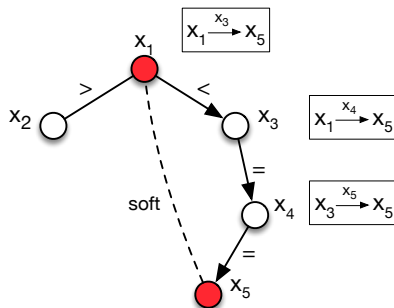


	$X_1 < X_2$	$X_1 > X_3$	$X_3 = X_4$	$X_4 = X_5$
	$X_2$ 0 1 2 3	$X_3$	$X_4$	$X_5$
$X_1$ 0	$\begin{bmatrix} 0 & 1 & 1 & 1 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$	$\begin{bmatrix} 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 1 & 1 & 0 \end{bmatrix}$	$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$	$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$
1				
2				
3				

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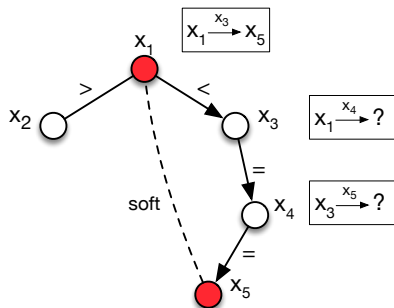


	$X1 < X2$	$X1 > X3$	$X3 = X4$	$X4 = X5$
$X2$	$\begin{bmatrix} 0 & 1 & 2 & 3 \end{bmatrix}$			
$X1$	$\begin{bmatrix} 0 & 1 & 1 & 1 \\ 1 & 0 & 0 & 1 \\ 2 & 0 & 0 & 1 \\ 3 & 0 & 0 & 0 \end{bmatrix}$	$\begin{bmatrix} 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 1 & 1 & 0 \end{bmatrix}$	$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$	$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$

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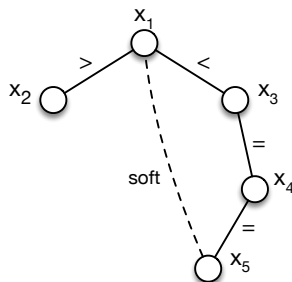
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$x_4$			$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$	
$x_5$				$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$



# BrC-DPOP

BrC-DPOP consists of 5 phases:

## 3. Arc Consistency Enforcement Phase.

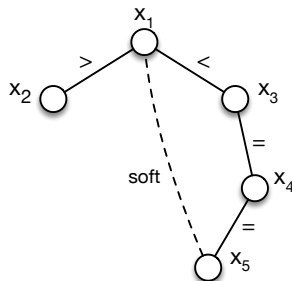


$X_1 < X_2$					$X_1 > X_3$				$X_3 = X_4$				$X_4 = X_5$			
$X_2$ 0 1 2 3					$X_3$				$X_4$				$X_5$			
$X_1$ 0	0	1	1	1	$X_1$ 0	0	0	0	$X_3$ 1	0	0	0	$X_4$ 1	0	0	0
1	0	0	1	1		1	0	0		0	1	0		0	1	0
2	0	0	0	1		1	1	0		0	0	1		0	0	1
3	0	0	0	0		1	1	1		0	0	0		0	0	1

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BrC-DPOP consists of 5 phases:

## 3. Arc Consistency Enforcement Phase.

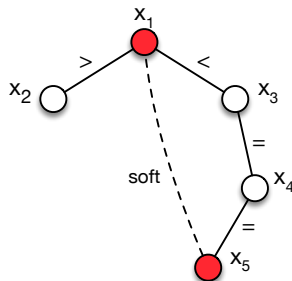


	$x_1 < x_2$	$x_1 > x_3$	$x_3 = x_4$	$x_4 = x_5$
$x_2$	0 1 2 3			
$x_1$	0 $\begin{bmatrix} 0 & 0 & 0 & 0 \end{bmatrix}$	$x_1$ $\begin{bmatrix} 0 & 0 & 0 & 0 \end{bmatrix}$	$x_3$ $\begin{bmatrix} 1 & 0 & 0 & 0 \end{bmatrix}$	$x_4$ $\begin{bmatrix} 1 & 0 & 0 & 0 \end{bmatrix}$
1	$\begin{bmatrix} 0 & 0 & 1 & 1 \end{bmatrix}$	$\begin{bmatrix} 1 & 0 & 0 & 0 \end{bmatrix}$	$\begin{bmatrix} 0 & 1 & 0 & 0 \end{bmatrix}$	$\begin{bmatrix} 0 & 1 & 0 & 0 \end{bmatrix}$
2	$\begin{bmatrix} 0 & 0 & 0 & 1 \end{bmatrix}$	$\begin{bmatrix} 1 & 1 & 0 & 0 \end{bmatrix}$	$\begin{bmatrix} 0 & 0 & 0 & 0 \end{bmatrix}$	$\begin{bmatrix} 0 & 0 & 0 & 0 \end{bmatrix}$
3	$\begin{bmatrix} 0 & 0 & 0 & 0 \end{bmatrix}$	$\begin{bmatrix} 0 & 0 & 0 & 0 \end{bmatrix}$	$\begin{bmatrix} 0 & 0 & 0 & 0 \end{bmatrix}$	$\begin{bmatrix} 0 & 0 & 0 & 0 \end{bmatrix}$

# BrC-DPOP

BrC-DPOP consists of 5 phases:

## 4. Branch Consistency Enforcement Phase.

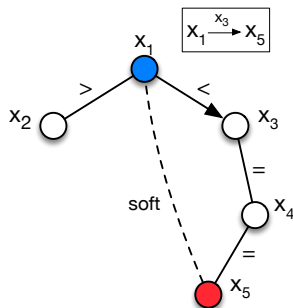


	$X_1 < X_2$	$X_1 > X_3$	$X_3 = X_4$	$X_4 = X_5$
$X_2$	0 1 2 3			
$X_1$	$\begin{bmatrix} 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 \\ 2 & 0 & 0 & 1 \\ 3 & 0 & 0 & 0 \end{bmatrix}$	$\begin{bmatrix} 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$	$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$	$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$

# BrC-DPOP

BrC-DPOP consists of 5 phases:

## 4. Branch Consistency Enforcement Phase.



$X_1 < X_2$					$X_1 > X_3$				$X_3 = X_4$				$X_4 = X_5$			
$X_2$					$X_3$				$X_4$				$X_5$			
$X_1$					$X_1$				$X_3$				$X_4$			
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0	0	1	1	1	0	0	0	0	1	0	0	0	1	0	0
2	0	0	0	1	1	1	0	0	0	0	0	0	0	0	0	0
3	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0

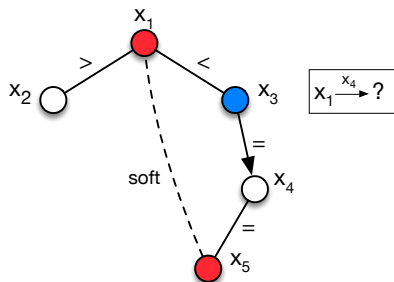
$$M_{11}$$

$$x_1 \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

# BrC-DPOP

BrC-DPOP consists of 5 phases:

## 4. Branch Consistency Enforcement Phase.



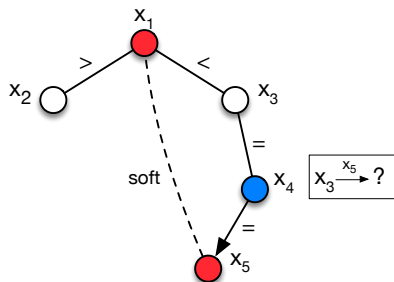
$X_1 < X_2$					$X_1 > X_3$				$X_3 = X_4$				$X_4 = X_5$						
$X_2$					$X_3$				$X_4$				$X_5$						
$X_1$	0	1	2	3	$X_1$	0	0	0	0	$X_3$	1	0	0	0	$X_4$	1	0	0	0
	0	0	0	0		0	0	0	0		1	0	0	0		1	0	0	0
1	0	0	1	1		1	0	0	0		0	1	0	0		0	1	0	0
2	0	0	0	1		1	1	0	0		0	0	0	0		0	0	0	0
3	0	0	0	0		0	0	0	0		0	0	0	0		0	0	0	0

$$\begin{array}{c} M_{31} \\ \begin{matrix} X_3 \\ x_1 \end{matrix} \begin{bmatrix} 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \end{array} \times \begin{array}{c} M_{11} \\ \begin{matrix} x_1 \\ x_1 \end{matrix} \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} = \begin{array}{c} M_{31} \\ \begin{matrix} X_3 \\ x_1 \end{matrix} \begin{bmatrix} 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

# BrC-DPOP

BrC-DPOP consists of 5 phases:

## 4. Branch Consistency Enforcement Phase.



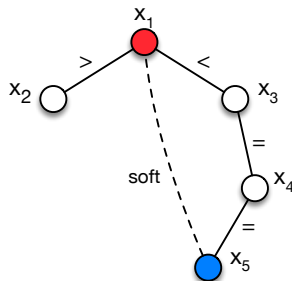
$X_1 < X_2$					$X_1 > X_3$				$X_3 = X_4$				$X_4 = X_5$			
		$X_2$	0	1	2	3			$X_3$			$X_4$			$X_5$	
$X_1$	0	0	0	0	0		$X_1$	0 <td>0</td> <td>0</td> <td>0</td> <th><math>X_3</math></th> <td>1</td> <td>0</td> <td>0</td> <td>0</td>	0	0	0	$X_3$	1	0	0	0
1	0	0	0	1	1		1	0	0	0		0	1	0	0	0
2	0	0	0	0	1		1	1	0	0		0	0	0	0	0
3	0	0	0	0	0		0	0	0	0		0	0	0	0	0

$$\begin{array}{c} M_{43} \\ x_4 \\ x_3 \end{array} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \times \begin{array}{c} M_{31} \\ x_3 \\ x_1 \end{array} \begin{bmatrix} 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} = \begin{array}{c} M_{41} \\ x_4 \\ x_1 \end{array} \begin{bmatrix} 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

# BrC-DPOP

BrC-DPOP consists of 5 phases:

## 4. Branch Consistency Enforcement Phase.



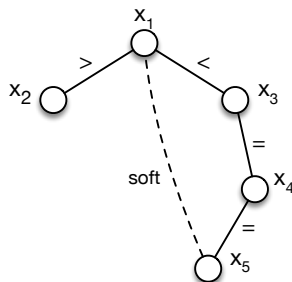
$x_1 < x_2$					$x_1 > x_3$				$x_3 = x_4$				$x_4 = x_5$			
$x_2$					$x_3$				$x_4$				$x_5$			
$x_1$					$x_1$				$x_3$				$x_4$			
0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0
1	0	0	1	1	1	0	0	0	0	1	0	0	0	1	0	0
2	0	0	0	1	1	1	0	0	0	0	0	0	0	0	0	0
3	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0

$$\begin{array}{c} M_{54} \\ x_5 \\ x_4 \end{array} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \times \begin{array}{c} M_{41} \\ x_4 \\ x_1 \end{array} \begin{bmatrix} 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} = \begin{array}{c} M_{51} \\ x_5 \\ x_1 \end{array} \begin{bmatrix} 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

# BrC-DPOP

BrC-DPOP consists of 5 phases:

## 5. UTIL and VALUE Phases.



UTIL<sub>5</sub>

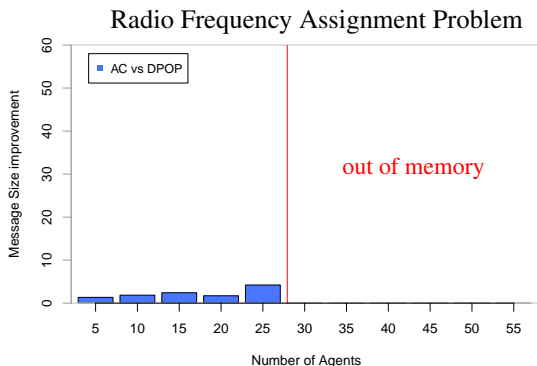
X5	X4	X1
0	0	1
0	0	2
1	1	2



# BrC-DPOP

BrC-DPOP consists of 5 phases:

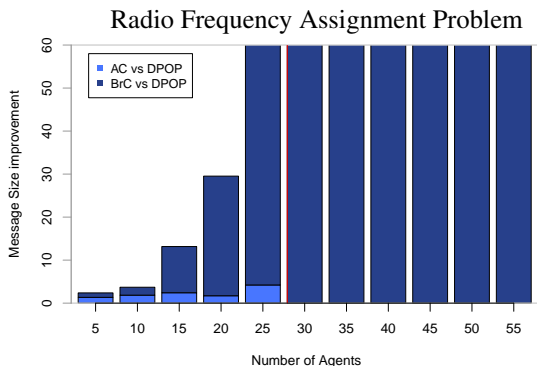
## 5. UTIL and VALUE Phases.



# BrC-DPOP

BrC-DPOP consists of 5 phases:

## 5. UTIL and VALUE Phases.



# BrC-DPOP: Theoretical Results

Let  $n = |\mathcal{A}|$ ,  $e = |\mathcal{F}|$  and  $d = \max_{x_i \in \mathcal{X}} |D_i|$ .

**Theorem 1.** The AC propagation phase requires  $O(nde)$  messages, each of size  $O(d)$ .

**Theorem 2.** The BrC propagation phase requires  $O(e)$  messages, each of size  $O(d^2)$ .

**Theorem 3.** The DCOP is arc consistent after the AC propagation phase.

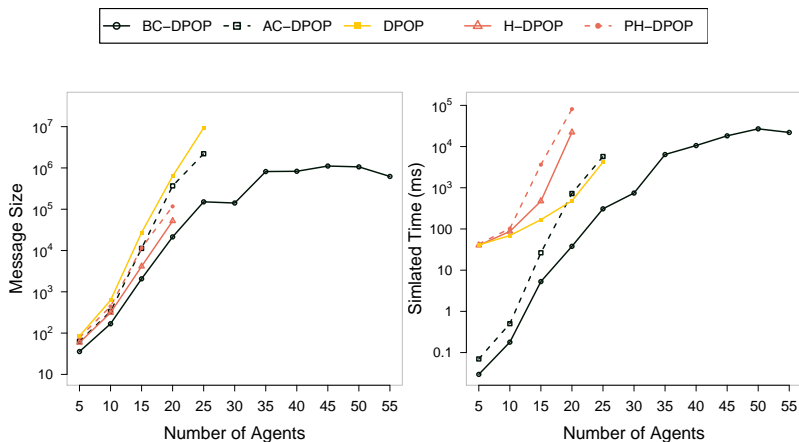
**Theorem 4.** The DCOP is branch consistent after the BrC propagation phase.

**Theorem 5.** BrC-DPOP is complete and correct.

# Experiments: Algorithms

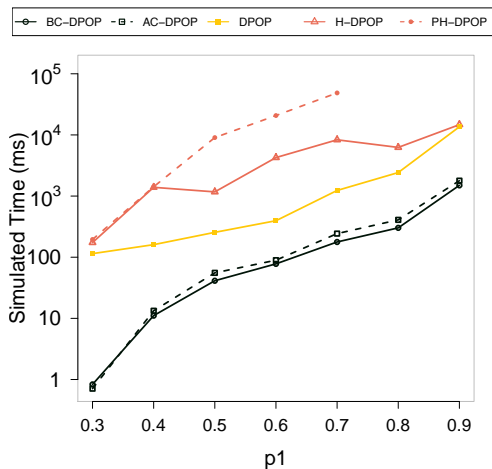
- DPOP
- H-DPOP
- P(rivacy-enhanced) H-DPOP
- AC-DPOP (only AC Propagation phase)
- BrC-DPOP

# Experiments: Radio Link Frequency Assignment



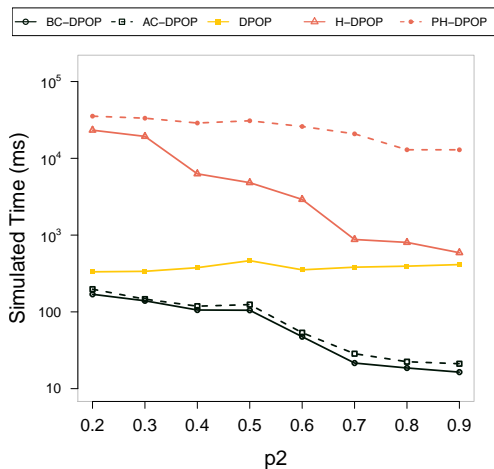
# Experiments: Random Graphs (Varying $p_1$ )

- $|\mathcal{A}| = 10$ .
- $|\mathcal{X}| = 10$ .
- $|D_i| = 8, \forall x_i \in \mathcal{X}$ .
- $p_2 = 0.6$
- We randomly injected hard constraints of type  $<$  or  $\neq$ .



# Experiments: Random Graphs (Varying $p_2$ )

- $|\mathcal{A}| = 10$ .
- $|\mathcal{X}| = 10$ .
- $|D_i| = 8, \forall x_i \in \mathcal{X}$ .
- $p_1 = 0.6$
- We randomly injected hard constraints of type  $<$  or  $\neq$ .



# Conclusions and Future Works

- Message size is one of the most important bottleneck in solving DCOPs.
- We have introduced the concept of *Branch Consistency* for Pseudo-Trees.
- We have introduced BrC-DPOP.
- Enhanced performances and scalability on random graphs and RLFA problems.
- We plan to extend this approach by:
  - Exploring propagation of soft constraints.
  - Handling high arity constraints.
- We are also interested in studying memory bounded solutions to scale up even to larger problems.



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Thank You!