# Global Constraints in Distributed Constraint Satisfaction (Extended Abstract)

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### **ABSTRACT**

Global constraints have been crucial for the success of centralized constraint programming. Here, we propose the inclusion of global constraints in distributed constraint satisfaction. We show how this inclusion can be done, considering different decompositions for global contraints. We provide experimental evidence of their benefits on several benchmarks solved with the ABT algorithm.

# **Categories and Subject Descriptors**

I.2.11 [Artificial Intelligence]: Distributed Artificial Intelligence

#### **General Terms**

Algorithms

## **Keywords**

Distributed constraint satisfaction, global constraints

## 1. INTRODUCTION

Global constraints have been crucial in the development of efficient constraint solvers [5]. They allow to capture global properties on an unbounded set of variables. In many cases, the exploitation of the semantic associated with each global constraint allows to codify propagators able to reach local consistency levels (typically generalized arc consistency, GAC) with polynomial complexity. This is a great advantage with respect to GAC propagators for generic non-binary constraints, which have complexity exponential in the constraint arity.

Often, it is implicitly assumed that distributed constraint reasoning precludes the use of global constraints. With the usual assumption that each agent contains a single variable (so agents and variables can be used interchangeably), an agent knows the constraint with each one of its neighbors, and nothing else [6]. These constraints are obviously binary. But this interpretation is too restrictive because there are distributed applications for which it is natural to use global constraints.

When adding global constraints in distributed reasoning we obtain several benefits. First, the expressivity of distributed constraint reasoning is enhanced since there are relations among several variables that cannot be expressed as a conjunction of binary relations (most global constraints are not binary decomposable). Second, the

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solving process can be done more efficiently. Local consistency can be more efficiently achieved when global constraints are involved [5]. Assuming a solving strategy maintaining some kind of local consistency, using global constraints improves its efficiency.

Accepting the interest of global constraints in distributed constraint reasoning, another question naturally follows: since some global constraints can be decomposed in simpler constraints, is it more efficient, to leave the global constraint as it was initially posted or to decompose it? If several decompositions are possible, which offers the best performance? We provide some answers to these questions, exploring two decompositions (binary [1] and nested for contractible constraints [4]) against the global constraint without decomposition, in two contexts: complete distributed search with / without unconditional GAC maintenance [3].

We assume that readers are familiar with constraint reasoning, specially with distributed constraint satisfaction problems (DisCSP) and the ABT algorithm [6].

# 2. ADDING GLOBAL CONSTRAINTS

A global constraint C is a class of constraints defined by a Boolean function  $f_C$  whose arity is not fixed. Constraints with different arities can be defined by the same Boolean function. For instance, all different  $(x_1, x_2, x_3)$  and all different  $(x_1, x_4, x_5, x_6)$  are two instances of the all different global constraint, where  $f_{all different}(T)$  returns true iff  $x_i \neq x_j, \forall x_i, x_j \in T$ . A global constraint C is contractible iff for any tuple t on  $x_{i_1}, \ldots, x_{i_{p+1}}$ , if t satisfies  $C(x_{i_1}, \ldots, x_{i_{p+1}})$  then the projection  $t[x_{i_1}, \ldots, x_{i_p}]$  of t on the first t variables satisfies t of binary decomposable without extra variables iff for any instance t of that the solutions of t are the solutions of t of t or a global constraint, while t of t or a global constraint, while t of variables t of that global constraint on the set of variables t.

We consider three different representations for a global constraint instance: direct, nested and binary. In the direct representation, C(T) is posted as a single constraint that allows all tuples on T satisfying C. Each agent in T includes C(T) in its constraint set. The nested representation is applicable to all contractible global constraints. The nested representation of C(T) with  $T=(x_{i_1},\ldots,x_{i_p})$  is the set of constraints  $\{C(x_{i_1},\ldots,x_{i_j})\mid j\in 2\ldots p\}$ . Each agent in T includes all constraints of the nested representation of C(T) that involve its variable in its constraints that are binary representation is applicable to all global constraints that are binary decomposable. The binary representation of C(T) is the set of constraints of its binary decomposition. Each agent in T includes all constraints of the binary decomposition of C(T) that involve its variable in its constraint set. The three representations for the

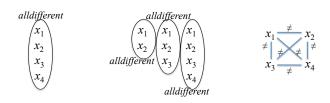


Figure 1: Representations for *alldifferent* $(x_1, x_2, x_3, x_4)$ : (left) direct, (center) nested, (right) binary.

 $\mathit{alldifferent}(x_1, x_2, x_3, x_4)$  global constraint appear in Figure 1.

Considering ABT as the solving algorithm, it is worth noting that ABT—originally proposed for binary constraints— can be easily generalized to handle constraints of any arity [2]. We assume that our ABT version contains such generalization.

In the direct representation, C(T) is posted as a single constraint. Each agent in T knows it. The lowest priority agent of T in the ABT order is in charge of evaluating it. Other agents in T put a link between themselves and that agent. In the nested representation, C(T),  $T=(x_{i_1},\ldots,x_{i_p})$ , is represented by the set of constraints  $\{C(x_{i_1},\ldots,x_{i_j})\mid j\in 2\ldots p\}$ . Thanks to the extra constraints that are posted, the checking of C(T) is not postponed to the last agent in T. In the binary representation, C(T) is represented by the set of constraints of its binary decomposition. These three representations of a global constraint instance are equivalent from the semantic point of view (they produce the same solutions). But they cause different ABT executions, so they can be seen as different models with dissimilar efficiency.

### 3. PROPAGATING GLOBAL CONSTRAINTS

Independently of the way a global constraint is included into ABT, this algorithm can be enhanced maintaining some form of local consistency during search. This was already investigated in [3], where limited/full forms of arc consistency (AC) were maintained during ABT execution for binary DisCSPs. While in [3] a limited form of AC causing unconditional deletions and full AC causing conditional deletions were considered, in this paper we only maintain the limited form of GAC that causes unconditional deletions (GAC because constraints may have arity higher than 2). Clearly, this limited GAC, that from now on we call UGAC, is less powerful than full GAC. Maintaining full GAC in the distributed context would cause a substantial load of extra messages which could overcome the benefits of domain pruning. We enforce UGAC on each considered global constraint by adapting the methods achieving GAC on them -developed in the centralized case- to this distributed setting, making them work inside each agent.

Before search, a suitable preprocess makes the problem GAC (before search any value deletion is unconditional, so GAC is equivalent to UGAC). During search, UGAC is enforced as follows: in ABT execution, if agent self receives a nogood message justifying the removal of its value v where the nogood has an empty left-hand side (see [6,3] for details), v can be unconditionally deleted from its domain. A deletion in the domain of  $x_{self}$  is propagated maintaining UGAC on the constraints connecting  $x_{self}$  with other variables, which may cause further deletions. Since the initial deletion is unconditional, deletions caused by the propagation are also unconditional.

To maintain UGAC during ABT search, some modifications are needed over the ABT algorithm: (1) the domain of variables constrained with *self* has to be represented in *self*; (2) only the agent

owner of a variable can modify its domain; if agent i deduces that a value could be deleted from the domain of  $x_j$ , it does nothing because that deduction will be done by agent j at some point; (3) there is a new message DEL to notify of value deletions:  $\mathrm{DEL}(self,k,v)$ —informing that self removes v from the domain of  $x_{self}$ —is sent from self to every agent k constrained with it; (4) a suitable preprocess makes all constraints GAC before ABT starts. These changes do not modify ABT correctness and completeness.

#### 4. EXPERIMENTS AND SUMMARY

We evaluated the impact of the addition of global constraints on random binary DisCSPs including instances of *alldifferent* and *atmost* global constraints. For ABT on DisCSPs with loose binary constraints, the most efficient representation is the binary one, followed by nested and finally direct. For ABT on DisCSPs with tight binary constraints, the most efficient representation is the direct one, followed by nested and finally binary. The same pattern appears considering both the number of exchanged messages and the number of non-concurrent constraint checks (NCCCs). For ABT-UGAC, enforcing UGAC propagation during search causes a drastic efficiency improvement from medium to high tightness, while the ranking of representations remains the same.

As summary, in this paper we propose the use of global constraints in distributed constraint reasoning, considering three different ways to represent global constraints. We evaluate the performance of ABT with or without UGAC maintenance on random DisCSPs containing some global constraints. We conclude that UGAC propagation of global constraints is never harmful in terms of messages, and in some cases it can significantly reduce the search space. Regarding the different representations of global constraints, the direct representation often is the less efficient one. For DisCSPs with loose binary constraints, the binary representation wins but for DisCSPs without solution, this representation degrades quickly generating too many nogood messages. The nested representation seems to offer a good compromise: it is never worse than direct, and in some cases it is better than binary. This is good news: there are many more constraints that are contractible than constraints that are binary decomposable.

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