# The complexity of conservative valued CSPs (Extended Abstract)

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## Abstract

We study the complexity of valued constraint satisfaction problems (VCSP). A problem from VCSP is characterised by a *constraint language*, a fixed set of cost functions over a finite domain. An instance of the problem is specified by a sum of cost functions from the language and the goal is to minimise the sum. Under the unique games conjecture, the approximability of finite-valued VCSPs is well-understood, see Raghavendra [FOCS'08]. However, there is no characterisation of finite-valued VCSPs, let alone general-valued VCSPs, that can be solved exactly in polynomial time, thus giving insights from a combinatorial optimisation perspective.

We consider the case of languages containing all possible unary cost functions. In the case of languages consisting of only  $\{0, \infty\}$ -valued cost functions (i.e. relations), such languages have been called *conservative* and studied by Bulatov [LICS'03] and recently by Barto [LICS'11]. Since we study valued languages, we call a language *conservative* if it contains all finite-valued unary cost functions. The computational complexity of conservative valued languages has been studied by Cohen *et al.* [AIJ'06] for languages over Boolean domains, by Deineko *et al.* [JACM'08] for  $\{0, 1\}$ -valued languages (a.k.a Max-CSP), and by Takhanov [STACS'10] for  $\{0, \infty\}$ -valued languages containing all finitevalued unary cost functions (a.k.a. Min-Cost-Hom).

We prove a Schaefer-like dichotomy theorem for conservative valued languages: if all cost functions in the language satisfy a certain condition (specified by a complementary combination of *STP and MJN multimorphisms*), then any instance can be solved in polynomial time (via a new algorithm developed in this paper), otherwise the language is NP-hard. This is the *first* complete complexity classification of *general-valued constraint languages* over non-Boolean domains. It is a common phenomenon that complexity classifications of problems over non-Boolean domains is significantly harder than the Boolean case. The polynomial-time algorithm we present for the tractable cases is a generalisation of the submodular minimisation problem and a result of Cohen *et al.* [TCS'08].

Our results generalise previous results by Takhanov [STACS'10] and (a subset of results) by Cohen *et al.* [AIJ'06] and Deineko *et al.* [JACM'08]. Moreover, our results do not rely on any computer-assisted search as in Deineko *et al.* [JACM'08], and provide a powerful tool for proving hardness of finite-valued and general-valued languages.

## 1 Introduction

The constraint satisfaction problem is a central generic problem in computer science. It provides a common framework for many theoretical problems as well as for many real-life applications, see [27] for a nice survey. An instance of the *constraint satisfaction problem* (CSP) consists of a collection of variables which must be assigned values subject to specified constraints. CSP is equivalent to the problem of evaluating conjunctive queries on databases [33], and to the homomorphism problem for relational structures [23].

An important line of research on the CSP is to identify all tractable cases; that is, cases that are recognisable and solvable in polynomial time. Most of this work has been focused on one of the two general approaches: either identifying structural properties of the way constraints interact which ensure tractability no matter what forms of constraints are imposed [21], or else identifying forms of constraints which are sufficiently restrictive to ensure tractability no matter how they are combined [10, 23].

The first approach has been used to characterise all tractable cases of bounded-arity CSPs: the *only* class of structures which ensures tractability (subject to a certain complexity theory assumption, namely FPT  $\neq W[1]$ ) are structures of bounded tree-width modulo homomorphic equivalence [19, 25, 26, 36]; and recently also for unbounded-arity CSPs [37]. The second approach has led to identifying certain algebraic properties known as polymorphisms [29] which are necessary for a set of constraint types to ensure tractability. A set of constraint types which ensures tractability is called a *tractable constraint language*.

Schaefer in his seminal work [41] gave a complete complexity classification of Boolean constraint languages. The algebraic approach based on polymorphisms [30] has been so far the most successful tool in generalising Schaefer's result to languages over a 3-element domain [9], languages with all unary relations [11, 4], languages comprised of a single binary relation without sources and sinks [3] (see also [6]), and languages comprised of a single binary relation that is a special triad [2]. The algebraic approach has also been essential in characterising the power of local consistency [5] and the "few subpowers property" [7], the two

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main tools known for solving tractable CSPs. A major open question in this line of research is the *Dichotomy Conjecture* of Feder and Vardi, which states that every constraint language is either tractable or NP-hard [23]. We remark that there are other approaches to the dichotomy conjecture; see, for instance, [27] for a nice survey of Hell and Nešetřil, and [34] for a connection between the Dichotomy Conjecture and probabilistically checkable proofs.

Since in practice many constraint satisfaction problems are over-constrained, and hence have no solution, or are under-constrained, and hence have many solutions, soft constraint satisfaction problems have been studied [20]. In an instance of the soft CSP, every constraint is associated with a cost function (rather than a relation as in the CSP) which represents preferences among different partial assignments, and the goal is to find the best assignment. Several very general soft CSP frameworks have been proposed in the literature [42, 8]. In this paper we focus on one of the very general frameworks, the valued constraint satisfaction problem (VCSP) [42]. Throughout the paper, we use the term constrain language (or just language) for a set of cost functions over a finite domain. If all cost functions from a given language  $\Gamma$  are  $\{0, \infty\}$ -valued (i.e. relations), we call  $\Gamma$  a *crisp* language. (If necessary, to stress the fact that  $\Gamma$  is a language, but not a crisp language, we call  $\Gamma$  a general-valued language.)

Similarly to the CSP, an important line of research on the VCSP is to identify tractable cases which are recognisable in polynomial time. Is is well known that structural reasons for tractability generalise to the VCSP [20]. In the case of language restrictions, only a few conditions are known to guarantee tractability of a given language [14, 13].

Related work The problem of characterising the complexity of different languages has received significant attention in the literature. For some classes researchers have established a Schaefer-like dichotomy theorem of the following form: if language  $\Gamma$  admits certain *polymorphisms* or *multimorphisms* then it is tractable, otherwise it is NPhard. Some of these classes are as follows: Boolean languages, i.e. languages with a 2-element domain (Cohen et al. [14]); crisp languages including all unary relations (Bulatov [11] and recently Barto [4]); crisp languages with a 3-element domain (Bulatov [9]);  $\{0, 1\}$ -valued languages including all unary cost functions (Deineko et al. [22]); crisp languages including additionally all finite-valued unary cost functions (Takhanov [43]); crisp languages including additionally a certain subset of finite-valued unary cost functions (Takhanov [44]).

Our proof exploits the results of Takhanov [43], who showed the existence of a majority polymorphism as a necessary condition for tractability of crisp languages including additionally all finite-valued unary cost functions. Other related work includes the work of Creignou *et al.* who studied various generalisations of the CSP to optimisation problems over Boolean domains [17], see also [18, 32]. Raghavendra [39] and Raghavendra and Steurer [40] have shown how to optimally approximate any finite-valued VCSP.

**Contributions** This paper focuses on valued languages containing all finite-valued unary cost functions; we call such languages *conservative*. Our main result is a dichotomy theorem for all conservative languages: if a conservative language  $\Gamma$  admits a complementary combination of *STP* (*symmetric tournament pair*) and *MJN* (*majority-majority-minority*) multimorphisms, then it is tractable, otherwise  $\Gamma$  is NP-hard. This is the first complete complexity classification of general-valued languages over non-Boolean domains, generalising previously obtained results in [14, 22, 43] as follows:

- Cohen *et al.* proved a dichotomy for arbitrary Boolean languages (|D| = 2). We generalise it to arbitrary domains ( $|D| \ge 2$ ), although only for conservative languages.
- Deineko *et al.* [22] and Takhanov [43] proved a dichotomy for the following languages, respectively:
  - {0,1}-valued languages containing additionally all unary cost functions;
  - $\{0,\infty\}$ -valued languages containing additionally all unary cost functions.

In both of these case the languages are conservative, so these classifications are special cases of our result. Note, however, that Deineko *et al.* additionally give a dichotomy with respect to approximability (PO vs. APX-hard), even when the number of occurrences of variables in instances is bounded; this part of [22] does not follow from our classification.

Moreover, our results provide a new powerful tool and do not rely on a computer-assisted search as in [22]. Building on techniques from this paper, Jonsson *et al.* [31] have recently shown that the same approach can be also used for certain non-conservative languages.

Since the complexity of Boolean conservative languages is known, we start, similarly to Bulatov and Takhanov [11, 43], by exploring the interactions between different 2element subdomains. Given a conservative language  $\Gamma$ , we will investigate properties of a certain graph  $G_{\Gamma}$  associated with the language and cost functions expressible over  $\Gamma$ . We link the complexity of  $\Gamma$  to certain properties of the graph  $G_{\Gamma}$ .

First, we show that if  $G_{\Gamma}$  does not satisfy certain properties, then  $\Gamma$  is intractable. Second, using  $G_{\Gamma}$ , we construct a (partial) *STP multimorphism* and a (partial) *MJN multimorphism*. Finally, we show that any language which admits a complementary combination of *STP and MJN multimorphisms* is tractable, thus generalising a tractable class of Cohen *et al.* [13], which in turn is a generalisation of the submodular minimisation problem. Thus we obtain a dichotomy theorem. The general-valued case is much more involved than the finite-valued case, and requires different techniques compared to previous results.

Given a finite language  $\Gamma$ , the graph  $G_{\Gamma}$  is finite as well, but depends on the expressive power of  $\Gamma$  (see Section 2 for precise definitions), which is infinite. In order to test whether  $\Gamma$  is tractable, we do not need to construct the graph  $G_{\Gamma}$  as it follows from our result that we just need to test for the existence of a complementary combination of two multimorphisms, which can be established in polynomial time.

Our results are formulated using the terminology of valued constraint satisfaction problems, but they apply to various other optimisation frameworks that are equivalent to valued constraint satisfaction problems such as Gibbs energy minimisation, Markov Random Fields, Min-Sum problems, and other models [35, 46].

**Organisation of the paper** The rest of the paper is organised as follows. In Section 2, we define valued constraint satisfaction problems (VCSPs), conservative languages, multimorphisms and other necessary definitions needed throughout the paper. We state our results in Section 3. Finally, Section 4 gives an overview of the omitted proofs, which will be given in full detail in the full version of this paper.

## 2 Background and notation

We denote by  $\mathbb{Q}_+$  the set of all non-negative rational numbers. We define  $\overline{\mathbb{Q}}_+ = \mathbb{Q}_+ \cup \{\infty\}$  with the standard addition operation extended so that for all  $a \in \mathbb{Q}_+$ ,  $a + \infty = \infty$ . Members of  $\overline{\mathbb{Q}}_+$  are called *costs*. Throughout the paper, we denote by D any fixed finite set, called a *domain*. Elements of D are called *domain values* or *labels*.

A function f from  $D^m$  to  $\overline{\mathbb{Q}}_+$  will be called a *cost* function on D of arity m. If the range of f lies entirely within  $\mathbb{Q}_+$ , then f is called a *finite-valued* cost function. If the range of f is  $\{0, \infty\}$ , then f is called a *crisp* cost function. If the range of a cost function f includes non-zero finite costs and infinity, we emphasise this fact by calling f a general-valued cost function. Let  $f: D^m \to \overline{\mathbb{Q}}_+$  be an m-ary cost function f. We denote dom  $f = \{x \in D^m \mid f(x) < \infty\}$  to be the effective domain of f. The argument of f is called an assignment or a labelling. Functions f of arity m = 2 are called *binary*.

A language is a set of cost functions with the same domain D. Language  $\Gamma$  is called finite-valued (crisp, generalvalued respectively) if all cost functions in  $\Gamma$  are finitevalued (crisp, general-valued respectively). A language  $\Gamma$ is *Boolean* if |D| = 2. DEFINITION 2.1. An instance  $\mathcal{I}$  of the valued constraint satisfaction problem (VCSP) is a function  $D^V \to \overline{\mathbb{Q}}_+$  given by

$$Cost_{\mathcal{I}}(\boldsymbol{x}) = \sum_{t \in T} f_t \left( x_{i(t,1)}, \dots, x_{i(t,m_t)} \right)$$

It is specified by a finite set of nodes V, finite set of terms (also known as constraints) T, cost functions  $f_t : D^{m_t} \rightarrow \overline{\mathbb{Q}}_+$  or arity  $m_t$  and indices  $i(t,k) \in V$  for  $t \in T$ ,  $k = 1, \ldots, m_t$ . A solution to  $\mathcal{I}$  is an assignment  $x \in D^V$  with the minimum cost.

We denote by VCSP( $\Gamma$ ) the class of all VCSP instances whose terms  $f_t$  belong to  $\Gamma$ . A finite language  $\Gamma$  is called *tractable* if VCSP( $\Gamma$ ) can be solved in polynomial time, and *intractable* if VCSP( $\Gamma$ ) is NP-hard. An infinite language  $\Gamma$  is tractable if every finite subset  $\Gamma' \subseteq \Gamma$  is tractable, and intractable if there is a finite subset  $\Gamma' \subseteq \Gamma$  that is intractable.

The idea behind conservative languages is to contain all possible unary cost functions: Bulatov has called a crisp language  $\Gamma$  conservative if  $\Gamma$  contains all unary relations [11]. We are interested in valued languages containing all possible unary cost functions and hence define conservative languages as follows:

DEFINITION 2.2. Language  $\Gamma$  is called conservative if  $\Gamma$  contains all  $\{0,1\}$ -valued unary cost functions  $u: D \rightarrow \{0,1\}$ .

Such languages have been studied by Deineko *et al.* [22] and Takhanov [43]. Note, we could have defined  $\Gamma$  to be conservative if it contains all possible general-valued unary cost functions  $u: D \to \overline{\mathbb{Q}}_+$ . However, the weaker definition 2.2 will be sufficient for our purposes: as shown in the full version of this paper, adding all possible unary cost functions  $u: D \to \overline{\mathbb{Q}}_+$  to a conservative language  $\Gamma$  does not change the complexity of  $\Gamma$ .

DEFINITION 2.3. A mapping  $F: D^k \to D, k \ge 1$  is called a polymorphism of a cost function  $f: D^m \to \overline{\mathbb{Q}}_+$  if

$$F(\boldsymbol{x}_1,\ldots,\boldsymbol{x}_k) \in \operatorname{dom} f \qquad \forall \boldsymbol{x}_1,\ldots,\boldsymbol{x}_k \in \operatorname{dom} f$$

where F is applied component-wise. F is a polymorphism of a language  $\Gamma$  if F is a polymorphism of every cost function in  $\Gamma$ .

Multimorphisms [14] are generalisations of polymorphisms. To make the paper easier to read, we only define binary and ternary multimorphisms as we will not need multimorphisms of higher arities.

DEFINITION 2.4. Let  $\langle \Box, \sqcup \rangle$  be a pair of operations, where  $\Box, \sqcup : D \times D \rightarrow D$ , and let  $\langle F_1, F_2, F_3 \rangle$  be a triple of operations, where  $F_i : D \times D \times D \rightarrow D$ ,  $1 \le i \le 3$ .

• Pair  $\langle \Box, \sqcup \rangle$  is called a (binary) multimorphism of cost function  $f: D^m \to \overline{\mathbb{Q}}_+$  if

$$(2.1) f(\boldsymbol{x} \sqcap \boldsymbol{y}) + f(\boldsymbol{x} \sqcup \boldsymbol{y}) \leq f(\boldsymbol{x}) + f(\boldsymbol{y}) \quad \forall \boldsymbol{x}, \boldsymbol{y} \in \text{dom} f$$

where operations  $\Box, \sqcup$  are applied component-wise.  $\langle \Box, \sqcup \rangle$  is a multimorphism of language  $\Gamma$  if  $\langle \Box, \sqcup \rangle$  is a multimorphism of every f from  $\Gamma$ .

• Triple  $\langle F_1, F_2, F_3 \rangle$  is called a (ternary) multimorphism of cost function  $f: D^m \to \overline{\mathbb{Q}}_+$  if

(2.2) 
$$f(F_1(\boldsymbol{x}, \boldsymbol{y}, \boldsymbol{z})) + f(F_2(\boldsymbol{x}, \boldsymbol{y}, \boldsymbol{z})) + f(F_3(\boldsymbol{x}, \boldsymbol{y}, \boldsymbol{z}))$$

$$\leq f(\boldsymbol{x}) + f(\boldsymbol{y}) + f(\boldsymbol{z}) \qquad \forall \boldsymbol{x}, \boldsymbol{y}, \boldsymbol{z} \in \text{dom} f$$

where operations  $F_1, F_2, F_3$  are applied componentwise.  $\langle F_1, F_2, F_3 \rangle$  is a multimorphism of language  $\Gamma$  if  $\langle F_1, F_2, F_3 \rangle$  is a multimorphism of every f from  $\Gamma$ .

- Operation  $F : D^k \to D$  is called conservative if  $F(x_1, \ldots, k_k) \in \{x_1, \ldots, x_k\}$  for all  $x_1, \ldots, x_k \in D$ .
- Pair  $\langle \Box, \sqcup \rangle$  is called conservative if  $\{\{a \Box b, a \sqcup b\}\} = \{\{a, b\}\}\$  for all  $a, b \in D$ , where  $\{\{\ldots\}\}\$  denotes a multiset, *i.e.* in the case of repetitions elements' multiplicities are taken into account. Similarly, triple  $\langle F_1, F_2, F_3 \rangle$  is called conservative if  $\{\{F_1(a, b, c), F_2(a, b, c), F_3(a, b, c)\}\} = \{\{a, b, c\}\}\$  for all  $a, b, c \in D$ . In other words, applying  $\langle F_1, F_2, F_3 \rangle$  to (a, b, c) should give a permutation of (a, b, c).
- *Pair*  $\langle \Box, \sqcup \rangle$  *is called a* symmetric tournament pair (STP) *if it is conservative and both operations*  $\Box, \sqcup$  *are commutative, i.e.*  $a \Box b = b \Box a$  *and*  $a \sqcup b = b \sqcup a$  *for all*  $a, b \in D$ .
- An operation  $Mj: D^3 \to D$  is called a majority operation if for every tuple  $(a, b, c) \in D^3$  with  $|\{a, b, c\}| = 2$ operation Mj returns the unique majority element among a, b, c (that occurs twice). An operation  $Mn: D^3 \to D$ is called a minority operation if for every tuple  $(a, b, c) \in$  $D^3$  with  $|\{a, b, c\}| = 2$  operation Mn returns the unique minority element among a, b, c (that occurs once).
- *Triple* (Mj<sub>1</sub>, Mj<sub>2</sub>, Mn<sub>3</sub>) *is called an* MJN *if it is conservative*, Mj<sub>1</sub>, Mj<sub>2</sub> *are (possibly different) majority operations, and* Mn<sub>3</sub> *is a minority operation.*

We say that  $\langle \Box, \sqcup \rangle$  is a multimorphism of language  $\Gamma$ , or  $\Gamma$  admits  $\langle \Box, \sqcup \rangle$ , if all cost functions  $f \in \Gamma$  satisfy (2.1). Using a polynomial-time algorithm for minimising submodular functions, Cohen *et al.* have obtained the following result:

THEOREM 2.1. ([13]) If a language  $\Gamma$  admits an STP, then  $\Gamma$  is tractable.

The existence of an MJN multimorphism also leads to tractability. This was shown for a specific choice of an MJN by Cohen *et al.* [14].

Our tractability result, presented in the next section, will include both above-mentioned tractable classes as special cases.

**Expressibility** Finally, we define the important notion of expressibility, which captures the idea of introducing auxiliary variables in a VCSP instance and the possibility of minimising over these auxiliary variables. (For crisp languages, this is equivalent to *implementation* [18].)

DEFINITION 2.5. A cost function  $f : D^m \to \overline{\mathbb{Q}}_+$  is expressible over a language  $\Gamma$  if there exists an instance  $\mathcal{I} \in VCSP(\Gamma)$  with the set of nodes  $V = \{1, \ldots, m, m + 1, \ldots, m + k\}$  where  $k \ge 0$  such that

$$f(\boldsymbol{x}) = \min_{\boldsymbol{y} \in D^k} Cost_{\mathcal{I}}(\boldsymbol{x}, \boldsymbol{y}) \qquad \forall \boldsymbol{x} \in D^m$$

We define  $\Gamma^*$  to be the expressive power of  $\Gamma$ ; that is, the set of all cost functions f such that f is expressible over  $\Gamma$ .

The importance of expressibility is in the following result:

THEOREM 2.2. ([14]) For any language  $\Gamma$ ,  $\Gamma$  is tractable iff  $\Gamma^*$  is tractable.

It is easy to observe and well known that any polymorphism (multimorphism) of  $\Gamma$  is also a polymorphism (multimorphism) of  $\Gamma^*$  [14].

#### **3** Our results

In this section, we relate the complexity of a conservative language  $\Gamma$  to properties of a certain graph  $G_{\Gamma}$  associated with  $\Gamma$ .

Given a conservative language  $\Gamma$ , let  $G_{\Gamma} = (P, E)$  be the graph with the set of nodes  $P = \{(a, b) | a, b \in D, a \neq b\}$ and the set of edges E defined as follows: there is an edge between  $(a, b) \in P$  and  $(a', b') \in P$  iff there exists binary cost function  $f \in \Gamma^*$  such that

(3.3) 
$$f(a,a') + f(b,b') > f(a,b') + f(b,a') ,$$
$$(a,b'), (b,a') \in \text{dom} f$$

Note that  $G_{\Gamma}$  may have self-loops. For node  $p \in P$ we denote the self-loop by  $\{p, p\}$ . We say that edge  $\{(a, b), (a', b')\} \in E$  is *soft* if there exists binary  $f \in \Gamma^*$  satisfying (3.3) such that at least one of the assignments (a, a'), (b, b') is in dom f. Edges in E that are not soft are called *hard*. For node  $p = (a, b) \in P$  we denote  $\bar{p} = (b, a) \in P$ . Note, a somewhat similar graph (but not the same) was used by Takhanov [43] for languages  $\Gamma$  containing crisp functions and finite unary cost functions.<sup>1</sup>

We denote  $M \subseteq P$  to be the set of vertices  $(a, b) \in P$ without self-loops, and  $\overline{M} = P - M$  to be the complement of M. It follows from the definition that set M is symmetric, i.e.  $(a,b) \in M$  iff  $(b,a) \in M$ . We will write  $\{a,b\} \in M$ to indicate that  $(a,b) \in M$ ; this is consistent due to the symmetry of M. Similarly, we will write  $\{a,b\} \in \overline{M}$  if  $(a,b) \in \overline{M}$ , and  $\{a,b\} \in P$  if  $(a,b) \in P$ , i.e.  $a,b \in D$  and  $a \neq b$ .

**DEFINITION 3.1.** Let  $(\Box, \sqcup)$  and  $(Mj_1, Mj_2, Mn_3)$  be binary and ternary operations respectively.

- Pair  $\langle \Box, \sqcup \rangle$  is an STP on M if  $\langle \Box, \sqcup \rangle$  is conservative on  $P \cup \{\{a\} \mid a \in D\}$  and commutative on M.
- Triple  $(Mj_1, Mj_2, Mn_3)$  is an MJN on  $\overline{M}$  if it is conservative and for each triple  $(a, b, c) \in D^3$  with  $\{a, b, c\} = \{x, y\} \in \overline{M}$  operations  $Mj_1(a, b, c), Mj_2(a, b, c)$  return the unique majority element among a, b, c (that occurs twice) and  $Mn_3(a, b, c)$  returns the remaining minority element.

Our main results are given by the following three theorems.

THEOREM 3.1. Let  $\Gamma$  be a conservative language.

- (a) If  $G_{\Gamma}$  has a soft self-loop then  $\Gamma$  is NP-hard.
- (b) If G<sub>Γ</sub> does not have soft self-loops then Γ admits a pair ⟨⊔, ⊓⟩ which is an STP on M and satisfies additionally a ⊓ b = a, a ⊔ b = b for {a, b} ∈ M.

THEOREM 3.2. Let  $\Gamma$  be a conservative language. If  $\Gamma$  does not admit an MJN on  $\overline{M}$  then it is NP-hard.

THEOREM 3.3. Suppose language  $\Gamma$  admits an STP on Mand an MJN on  $\overline{M}$ , for some choice of symmetric  $M \subseteq P$ . Then  $\Gamma$  is tractable.

Theorems 3.1-3.3 give the dichotomy result for conservative languages:

COROLLARY 3.1. If a conservative language  $\Gamma$  admits an STP on M and an MJN on  $\overline{M}$  for some symmetric  $M \subseteq P$  then  $\Gamma$  is tractable. Otherwise  $\Gamma$  is NP-hard.

*Proof.* The first part follows from Theorem 3.3; let us show the second part. Suppose that the precondition of the corollary does not hold, then one of the following cases must be true (we assume below that M is the set of nodes without self-loops in  $G_{\Gamma}$ ):

- $G_{\Gamma}$  has a soft self-loop. Then  $\Gamma$  is NP-hard by Theorem 3.1(a).
- $G_{\Gamma}$  does not have soft self-loops and  $\Gamma$  does not admit an STP on M. This is a contradiction by Theorem 3.1(b).
- $G_{\Gamma}$  does not have soft self-loops and  $\Gamma$  does not admit an MJN on  $\overline{M}$ . Then  $\Gamma$  is NP-hard by Theorem 3.2.

In the finite-valued case, we get a simpler tractability criterion:

COROLLARY 3.2. If a conservative finite-valued language  $\Gamma$  admits an STP then  $\Gamma$  is tractable. Otherwise  $\Gamma$  is NP-hard.<sup>2</sup>

**Proof.** Consider the graph  $G_{\Gamma}$  associated with  $\Gamma$ . If  $G_{\Gamma}$  contains a soft self-loop, then, by Theorem 3.1(a),  $\Gamma$  is NP-hard. Suppose that  $G_{\Gamma}$  does not contain soft self-loops. As  $\Gamma$  is finite-valued,  $G_{\Gamma}$  cannot have hard self-loops. Therefore,  $\overline{M}$  is empty and M = P. By Theorem 3.1(b),  $\Gamma$  admits an STP. The tractability then follows from Theorem 3.3.

#### 4 Overview of the proofs

In this section we sketch proofs of our results. Complete proofs of Theorems 3.1-3.3 will be given in the full version of this paper.

First, we show that we can strengthen the definition of conservative languages without loss of generality. Let  $\overline{\Gamma}$  be the language obtained from  $\Gamma$  by adding all possible general-valued unary cost functions  $u : D \to \overline{\mathbb{Q}}_+$ . Note,  $\overline{\Gamma}$  may be different from  $\Gamma$  since  $\Gamma$  is only guaranteed to have all possible  $\{0, 1\}$ -valued unary cost functions.

**PROPOSITION 4.1.** (a) Graphs  $G_{\Gamma}$  and  $G_{\overline{\Gamma}}$  are the same: if  $\{p,q\}$  is a soft (hard) edge in  $G_{\Gamma}$  then it is also a soft (hard) edge in  $G_{\overline{\Gamma}}$ , and vice versa. (b) If  $\overline{\Gamma}$  is NP-hard then so is  $\Gamma$ .

This shows that it suffices to prove Theorems 3.1 and 3.2 for language  $\overline{\Gamma}$ . Indeed, consider Theorem 3.1 for a conservative language G. If  $G_{\Gamma}$  has a soft self-loop then by Proposition 4.1(a) so does  $G_{\overline{\Gamma}}$ . Theorem 3.1(a) for language  $\overline{\Gamma}$ would imply that  $\overline{\Gamma}$  is NP-hard, and therefore  $\Gamma$  is also NPhard by Proposition 4.1(b). If  $G_{\Gamma}$  does not have soft selfloops then neither does  $G_{\overline{\Gamma}}$ . Theorem 3.1(b) for language  $\overline{\Gamma}$ would imply that  $\overline{\Gamma}$  admits the appropriate multimorphism

<sup>&</sup>lt;sup>1</sup>Roughly speaking, the graph structure in [43] was defined via a "min" polymorphism rather than a  $\langle \min, \max \rangle$  multimorphism, so the property  $\{p, q\} \in E \Rightarrow \{\bar{p}, \bar{q}\} \in E$  (that we prove for our graph in the next section) might not hold in Takhanov's case. Also, in [43] edges were not classified as being soft or hard.

 $<sup>^{2}</sup>$ It can be shown that if a finite-valued language admits an STP multimorphism, it also admits a submodularity multimorphism. This result is implicitly contained in [13]. Namely, after reducing the domains as in [13, Theorem 8.3], the STP might contain cycles. [13, Lemma 7.15] tells us that on cycles we have, in the finite-valued case, only unary cost functions. It follows that the cost functions admitting the STP must be submodular w.r.t. some total order [16].

This simplifies the tractability criterion in the finite-valued case (though we do not exploit this fact anywhere in the paper).

 $\langle \sqcup, \sqcap \rangle$ . Since  $\Gamma \subseteq \overline{\Gamma}$ ,  $\langle \sqcup, \sqcap \rangle$  is also a multimorphism of  $\Gamma$ . Lemma 4.1(a) so is  $\overline{p}$ ). Clearly, property (i) still holds. Property (ii) holds for Theorem 3.2.

We can therefore make the following assumption without loss of generality when proving Theorems 3.1 and 3.2:

Assumption 1.  $\Gamma$  contains all general-valued unary cost functions  $u: D \to \overline{\mathbb{Q}}_+$ .

**4.1** Sketch of the proof of Theorem 3.1 Using unary cost functions, Theorem 3.1(a) can be proved via a reduction from the Max-SAT problem with XOR clauses and the Independent Set problem. We focus on Theorem 3.1(b). Graph  $G_{\Gamma}$  and its properties play a crucial role in the proofs.

In the lemma below, a *path* of length k is a sequence of edges  $\{p_0, p_1\}, \{p_1, p_2\}, \ldots, \{p_{k-1}, p_k\}$ , where  $\{p_{i-1}, p_i\} \in E$ . Note that we allow edge repetitions. A path is *even* iff its length is even. A path is a *cycle* if  $p_0 = p_k$ . If  $X \subseteq P$  then (X, E[X]) denotes the subgraph of (P, E) induced by X.

LEMMA 4.1. Graph  $G_{\Gamma} = (P, E)$  satisfies the following properties:

- (a)  $\{p,q\} \in E$  implies  $\{\bar{p},\bar{q}\} \in E$  and vice versa. The two edges are either both soft or both hard.
- (b) Suppose that  $\{p,q\} \in E$  and  $\{q,r\} \in E$ , then  $\{p,\bar{r}\} \in E$ . If at least one of the first two edges is soft then the third edge is also soft.
- (c) For each  $p \in P$ , nodes p and  $\overline{p}$  are either both in M or both in  $\overline{M}$ .
- (d) There are no edges from M to  $\overline{M}$ .
- (e) Graph (M, E[M]) does not have odd cycles.
- (f) If node p is not isolated (i.e. it has at least one incident edge  $\{p,q\} \in E$ ) then  $\{p,\bar{p}\} \in E$ .
- (g) Nodes  $p \in \overline{M}$  do not have incident soft edges.

We now construct a pair of operations  $\langle \Box, \sqcup \rangle$  for  $\Gamma$  that behaves as an STP on M and as a multi-projection (returning its two arguments in the same order) on  $\overline{M}$ .

LEMMA 4.2. There exists an assignment  $\sigma$  :  $M \rightarrow \{-1, +1\}$  such that (i)  $\sigma(p) = -\sigma(q)$  for all edges  $\{p, q\} \in E$ , and (ii)  $\sigma(p) = -\sigma(\bar{p})$  for all  $p \in M$ .

*Proof.* By Lemma 4.1(e) graph (M, E[M]) does not have odd cycles. Therefore, by Harary's Theorem, graph (M, E[M]) is bipartite and there exists an assignment  $\sigma$ :  $M \rightarrow \{-1, +1\}$  that satisfies property (i). Let us modify this assignment as follows: for each isolated node  $p \in M$ (i.e. node without incident edges) set  $\sigma(p)$ ,  $\sigma(\bar{p})$  so that  $\sigma(p) = -\sigma(\bar{p}) \in \{-1, +1\}$ . (Note, if p is isolated then by Lemma 4.1(a) so is  $\bar{p}$ ). Clearly, property (i) still holds. Property (ii) holds for each node  $p \in M$  as well: if p is isolated then (ii) holds by construction, otherwise by Lemma 4.1(f) there exists edge  $\{p, \bar{p}\} \in E$ , and so (ii) follows from property (i).

Given assignment  $\sigma$  constructed in Lemma 4.2, we now define operations  $\sqcap, \sqcup : D^2 \to D$  as follows:

- $a \sqcap a = a \sqcup a = a$  for  $a \in D$ .
- If  $(a,b) \in M$  then  $a \sqcap b$  and  $a \sqcup b$  are the unique elements of D satisfying  $\{a \sqcap b, a \sqcup b\} = \{a,b\}$  and  $\sigma(a \sqcap b, a \sqcup b) = +1$ .
- If  $(a, b) \in \overline{M}$  then  $a \sqcap b = a$  and  $a \sqcup b = b$ .

LEMMA 4.3. For any binary cost function  $f \in \Gamma^*$  and any  $x, y \in \text{dom } f$  there holds

(4.4) 
$$f(\boldsymbol{x} \sqcap \boldsymbol{y}) + f(\boldsymbol{x} \sqcup \boldsymbol{y}) \le f(\boldsymbol{x}) + f(\boldsymbol{y})$$

In order to proceed, we introduce the following notation. Given a cost function f of arity m, we denote by V the set of variables corresponding to the arguments of f, with |V| = m. For two assignments  $x, y \in D^m$  we denote  $\Delta(x, y) = \{i \in V \mid x_i \neq y_i\}$  to be the set of variables on which x and y differ.

Having Lemma 4.3 as a base case and using induction on |V|, we can prove the following

LEMMA 4.4. Condition (4.4) holds for any cost function  $f \in \Gamma^*$  and assignments  $x, y \in \text{dom } f$  with  $|\Delta(x, y)| \leq 2$ .

Having Lemma 4.4 as a base case and using induction on  $|\Delta(x, y)|$ , we can prove the following

LEMMA 4.5. Condition (4.4) holds for any cost function  $f \in \Gamma^*$  and any  $x, y \in \text{dom } f$ .

The previous lemma finishes the proof of Theorem 3.1(b).

We remark that the idea behind the previous three lemmas comes from the proof that a k-ary finite-valued cost function f is submodular iff every binary projection of f is submodular [45]. However, this is known to not hold for general-valued cost functions (i.e. cost functions taking on both finite and infinite costs) [12], and hence our proofs are more elaborate.

**4.2** Sketch of the proof of Theorem **3.2** First, we introduce the following simplifying assumptions:

**Assumption 2.**  $\Gamma$  admits a majority polymorphism.

**Assumption 3.**  $G_{\Gamma}$  does not have soft self-loops.

(If these assumptions do not hold then it can be shown that  $\Gamma$  is NP-hard, using results of Takhanov [43] and Theorem 3.1(a)). Our goal is then to construct an MJN multi-morphism on  $\overline{M}$  under assumptions 1-3.

In order to do this, we introduce a function  $\mu$  which maps every set  $\{a, b, c\} \subseteq D$  with  $|\{a, b, c\}| = 3$  to a subset of  $\{a, b, c\}$ . This subset is defined as follows:  $c \in$  $\mu(\{a, b, c\})$  iff there exists binary function  $f \in \Gamma^*$  and a pair  $(a', b') \in \overline{M}$  such that

dom 
$$f = \{(a, a'), (b, a'), (c, b')\}$$

LEMMA 4.6. Set  $\mu(\{a, b, c\})$  contains at most one label. Furthermore, if  $\mu(\{a, b, c\}) = \{c\}$  then  $(a, c) \in \overline{M}$  and  $(b, c) \in \overline{M}$ .

For convenience, we define  $\mu(\{a, b, c\}) = \emptyset$  if  $|\{a, b, c\}| \leq 2$ . We are now ready to construct operation  $MJN = \langle Mj_1, Mj_2, Mn_3 \rangle$ . Given a tuple  $(a, b, c) \in D^3$ , we define

$$\begin{split} & \text{MJN}(a, b, c) &= \\ & (4.5a) \\ & (4.5b) \\ & (4.5c) \\ & (4.5c) \\ & (4.5d) \\ \end{split} \begin{cases} (x, x, y) & \text{if } \{\{a, b, c\}\} = \{\{x, x, y\}\}, \{x, y\} \in \overline{M} \\ & (b \sqcap c, b \sqcup c, a) & \text{if } \mu(\{a, b, c\}) = \{a\} \\ & (a \sqcap c, a \sqcup c, b) & \text{if } \mu(\{a, b, c\}) = \{b\} \\ & (a \sqcap b, a \sqcup b, c) & \text{in any other case} \\ \end{split}$$

where {{...}} denotes a *multiset*, i.e. elements' multiplicities are taken into account.

To prove Theorem 3.2, we need to prove for every  $f \in \Gamma^*$  and for every  $x, y, z \in \text{dom } f$ ,

$$\begin{array}{ll} (4.6) & f(\operatorname{Mj}_1(\boldsymbol{x},\boldsymbol{y},\boldsymbol{z})) + f(\operatorname{Mj}_2(\boldsymbol{x},\boldsymbol{y},\boldsymbol{z})) + f(\operatorname{Mn}_3(\boldsymbol{x},\boldsymbol{y},\boldsymbol{z})) \\ \\ & \leq & f(\boldsymbol{x}) + f(\boldsymbol{y}) + f(\boldsymbol{z}) \end{array}$$

We say that an instance  $(f, \boldsymbol{x}, \boldsymbol{y}, \boldsymbol{z})$  is valid if  $f \in \Gamma^*$  and  $\boldsymbol{x}, \boldsymbol{y}, \boldsymbol{z} \in \text{dom } f$ . It is satisfiable if (4.6) holds, and unsatisfiable otherwise. For a triple  $\boldsymbol{x}, \boldsymbol{y}, \boldsymbol{z} \in D^V$ denote  $\delta(\boldsymbol{x}, \boldsymbol{y}, \boldsymbol{z}) = \sum_{i \in V} |\{x_i, y_i, z_i\}|, \ \Delta(\boldsymbol{x}, \boldsymbol{y}, \boldsymbol{z}) = \{i \in V \mid x_i \neq y_i\} \text{ and } \Delta^M(\boldsymbol{x}, \boldsymbol{y}, \boldsymbol{z}) = \{i \in \Delta(\boldsymbol{x}, \boldsymbol{y}, \boldsymbol{z}) \mid \{x_i, y_i, z_i\} = \{a, b\} \in M\}.$ 

Suppose that an unsatisfiable instance exists. Let (f, x, y, z) be a lowest unsatisfiable instance with respect to the partial order  $\leq$  defined as the lexicographical order with components

(4.7) 
$$( \begin{array}{c} \delta(\boldsymbol{x}, \boldsymbol{y}, \boldsymbol{z}), \\ |\Delta(\boldsymbol{x}, \boldsymbol{y}, \boldsymbol{z})|, \\ |\Delta^{M}(\boldsymbol{x}, \boldsymbol{y}, \boldsymbol{z})|, \\ |\{i \in V \mid \mu(\{x_{i}, y_{i}, z_{i}\}) = \{x_{i}\}\}| \end{array} )$$

(the first component is more significant). We denote  $\delta_{\min} = \delta(\boldsymbol{x}, \boldsymbol{y}, \boldsymbol{z})$ . Thus, we can assume that all valid instances  $(f, \boldsymbol{x}', \boldsymbol{y}', \boldsymbol{z}')$  with  $(\boldsymbol{x}', \boldsymbol{y}', \boldsymbol{z}') \prec (\boldsymbol{x}, \boldsymbol{y}, \boldsymbol{z})$  (and in particular with  $\delta(\boldsymbol{x}', \boldsymbol{y}', \boldsymbol{z}') < \delta_{\min}$ ) are satisfiable, while the instance  $(f, \boldsymbol{x}, \boldsymbol{y}, \boldsymbol{z})$  is unsatisfiable.

By analysing this minimal counter-example, we show that none of the cases (4.5a)-(4.5c) are possible. The arguments that we use are more complicated versions of the induction arguments used for proving theorem 3.1; additionally, for one of the cases we exploit the fact that f admits a majority polymorphism.

We obtain that MJN(x, y, z) is defined via (4.5d) for all nodes. But then inequality (4.6) follows from the fact that  $\langle \Box, \sqcup \rangle$  is a multimorphism of f, contradicting to the choice of (f, x, y, z). This proves Theorem 3.2.

**4.3** Sketch of the proof of Theorem 3.3 In this section we present an algorithm for minimising instances from VCSP( $\Gamma$ ). The idea for the algorithm and some of the proof techniques have been influenced by the techniques used by Takhanov [43] for proving the absence of *arithmetical deadlocks* in certain instances. However, the algorithm itself is very different from Takhanov's approach. (The latter does not rely on submodular minimization algorithms; instead, it performs a reduction to an optimization problem in a perfect graph).

Let  $f: \mathcal{D} \to \overline{\mathbb{Q}}_+$  be the function to be minimised, V be the set of its variables (which we will also call nodes), and  $D_i$  be the domain of variable  $i \in V$  with  $\mathcal{D} = \times_{i \in V} D_i$ . In the beginning all domains are the same  $(D_i = D)$ , but as the algorithm progresses we will allow  $D_i$  to become different for different  $i \in V$ . As a consequence, operations  $\Box, \sqcup$  may act differently on different components of vectors  $\boldsymbol{x}, \boldsymbol{y} \in \mathcal{D}$ We denote  $\Box_i, \sqcup_i : D_i \times D_i \to D_i$  to be the *i*-th operations of  $\langle \Box, \sqcup \rangle$ . Similarly, we denote by  $\mathtt{Mj}_{1i}, \mathtt{Mj}_{2i}, \mathtt{Mn}_{3i} : D_i \times D_i \times D_i \to D_i$  to be the *i*-th operations of  $\langle \mathtt{Mj}_1, \mathtt{Mj}_2, \mathtt{Mn}_3 \rangle$ .

We denote by P the collection of sets  $P = (P_i)_{i \in V}$ where  $P_i = \{\{a, b\} \mid a, b \in D_i, a \neq b\}$ . We denote by M a collection of subsets  $M = (M_i)_{i \in V}$ ,  $M_i \subseteq P_i$ , and  $\overline{M} = (\overline{M}_i)_{i \in V}$ ,  $\overline{M}_i = P_i - M_i$ . We now extend Definition 3.1 as follows.

**DEFINITION 4.1.** Let  $\langle \Box, \sqcup \rangle$  and  $\langle Mj_1, Mj_2, Mn_3 \rangle$  be collections of binary and ternary operations respectively.

- Pair  $\langle \Box, \sqcup \rangle$  is an STP on M if for all  $i \in V$  pair  $\langle \Box_i, \sqcup_i \rangle$  is an STP on  $M_i$ .
- Triple  $\langle Mj_1, Mj_2, Mn_3 \rangle$  is an MJN on  $\overline{M}$  if for all  $i \in V$ triple  $\langle Mj_{1i}, Mj_{2i}, Mn_{3i} \rangle$  is an MJN on  $\overline{M}_i$ .

We will assume without loss of generality that  $\langle \Box_i, \sqcup_i \rangle$ is non-commutative on any  $\{a, b\} \in \overline{M}_i$  (if not, we can simply add such  $\{a, b\}$  to  $M_i$ ).

We are now ready to present the algorithm; it will consist of three stages.

**Stage 1: Decomposition into binary relations** It can be shown that the instance admits a majority polymorphism, and hence every cost function f can be decomposed [1] into unary relations  $\rho_i \subseteq D_i$ ,  $i \in D_i$  and binary relations  $\rho_{ij} \subseteq D_i \times D_j$ ,  $i, j \in V$ ,  $i \neq j$  such that

$$x \in \operatorname{dom} f \quad \Leftrightarrow \quad$$

$$[x_i \in \rho_i \ \forall i \in V]$$
 and  $[(x_i, x_j) \in \rho_{ij} \ \forall i, j \in V, i \neq j]$ 

We will always assume that binary relations are symmetric, i.e.  $(x, y) \in \rho_{ij} \Leftrightarrow (y, x) \in \rho_{ji}$ . We use the following notation for relations:

• If  $\rho_{ij} \in D_i \times D_j$ ,  $X \subseteq D_i$  and  $Y \subseteq D_j$  then

$$\rho_{ij}(X, \cdot) = \{ y \mid \exists x \in X \text{ s.t. } (x, y) \in \rho_{ij} \}$$
  
$$\rho_{ij}(\cdot, Y) = \{ x \mid \exists y \in Y \text{ s.t. } (x, y) \in \rho_{ij} \}$$

If  $X = \{x\}$  and  $Y = \{y\}$  then these two sets will be denoted as  $\rho_{ij}(x, \cdot)$  and  $\rho_{ij}(\cdot, y)$  respectively.

In the first stage we establish *strong 3-consistency* using the standard constraint-processing techniques [15] so that the resulting relations satisfy *arc-consistency*:

$$\{x \mid (\exists y)(x,y) \in \rho_{ij}\} = \rho_i \qquad \forall \text{ distinct } i, j \in V$$

and path-consistency:

$$\rho_{ik}(x,\cdot) \cap \rho_{jk}(y,\cdot) \neq \emptyset \quad \forall \text{ distinct } i, j, k \in V, (x,y) \in \rho_{ij}$$

It is known that in the presence of a majority polymorphism strong 3-consistency is equivalent to global consistency [28]; that is dom f is empty iff all  $\rho_i$  and  $\rho_{ij}$  are empty. Using this fact, it is not difficult to show that the strong 3-consistency relations  $\rho_i$ ,  $\rho_{ij}$  are uniquely determined by f via

$$\rho_i = \{x_i \mid \boldsymbol{x} \in \operatorname{dom} f\} \qquad \rho_{ij} = \{(x_i, x_j) \mid \boldsymbol{x} \in \operatorname{dom} f\}$$

The second equation implies that any polymorphism of f is also a polymorphism of  $\rho_{ij}$ .

From now on we will assume that  $D_i = \rho_i$  for all  $i \in V$ . This can be achieved by reducing sets  $D_i$  if necessary. We will also assume that all sets  $D_i$  are non-empty.

Stage 2: Modifying M and  $\langle \Box, \sqcup \rangle$  The second stage of the algorithm works by iteratively growing sets  $M_i$  and simultaneously modifying operations  $\langle \Box_i, \sqcup_i \rangle$  so that (i)  $\langle \Box_i, \sqcup_i \rangle$  is still a conservative pair which is commutative on  $M_i$  and non-commutative on  $\overline{M}_i$ , and (ii)  $\langle \Box, \sqcup \rangle$  is a multimorphism of f. It stops when we get  $M_i = P_i$  for all  $i \in V$ .

We now describe one iteration. First, we identify subset  $U \subseteq V$  and subsets  $A_i, B_i \subseteq D_i$  for each  $i \in U$  using the following algorithm:

- 1: pick node  $k \in V$  and pair  $\{a, b\} \in \overline{M}_k$ . (If they do not exist, terminate and go to Stage 3.)
- 2: set  $U = \{k\}, A_k = \{a\}, B_k = \{b\}$

3: while there exists  $i \in V - U$  such that  $\rho_{ki}(A_k, \cdot) \cap \rho_{ki}(B_k, \cdot) = \emptyset$  do

4: add *i* to *U*, set  $A_i = \rho_{ki}(A_k, \cdot)$ ,  $B_i = \rho_{ki}(B_k, \cdot)$ *I* compute "closure" of sets  $A_i$  for  $i \in U$ 

5: while there exists  $a \in D_k - A_k$  s.t.  $a \in \rho_{ki}(\cdot, A_i)$  for some  $i \in U - \{k\}$  do

- add a to A<sub>k</sub> set A<sub>j</sub> = ρ<sub>kj</sub>(A<sub>k</sub>, ·) for all j ∈ U − {k}
  end while // compute "closure" of sets B<sub>i</sub> for i ∈ U
- 8: while there exists  $b \in D_k B_k$  s.t.  $b \in \rho_{ki}(\cdot, B_i)$  for some  $i \in U - \{k\}$  do

9: add b to 
$$B_k$$
  
set  $B_j = \rho_{kj}(B_k, \cdot)$  for all  $j \in U - \{k\}$   
10: **end while**  
// done

11: end while

12: **return** set  $U \subseteq V$  and sets  $A_i, B_i \subseteq D_i$  for  $i \in U$ 

LEMMA 4.7. Sets U and  $A_i, B_i$  for  $i \in U$  produced by the algorithm have the following properties:

- (a) Sets  $A_i$  and  $B_i$  for  $i \in U$  are disjoint.
- (b)  $\{a,b\} \in \overline{M}_i$  for all  $i \in U$ ,  $a \in A_i$ ,  $b \in B_i$ .
- (c)  $\rho_{ki}(A_k, \cdot) = A_i$ ,  $\rho_{ki}(B_k, \cdot) = B_i$ ,  $\rho_{ki}(\cdot, A_i) = A_k$ ,  $\rho_{ki}(\cdot, B_i) = B_k$  for all  $i \in U - \{k\}$  where k is the node chosen in line 1.
- (d) Suppose that  $i \in U$  and  $j \in \overline{U} \equiv V U$ . If  $(c, x) \in \rho_{ij}$ where  $c \in A_i \cup B_i$  and  $x \in D_j$  then  $(d, x) \in \rho_{ij}$  for all  $d \in A_i \cup B_i$ .

To complete the iteration, we modify sets  $M_i$  and operations  $\Box_i, \sqcup_i$  for each  $i \in U$  as follows:

- add all pairs  $\{a, b\}$  to  $M_i$  where  $a \in A_i, b \in B_i$ .
- redefine  $a \sqcap_i b = b \sqcap_i a = a$ ,  $a \sqcup_i b = b \sqcup_i a = b$  for all  $a \in A_i, b \in B_i$

**LEMMA 4.8.** The new pair of operations  $\langle \Box, \sqcup \rangle$  is a multimorphism of f.

The two lemmas imply that all steps are well-defined, and upon termination the algorithm produces a pair  $\langle \Box, \sqcup \rangle$  which is an STP multimorphism of f.

**Stage 3: Reduction to a submodular minimisation problem** At this stage we have an STP multimorphism. Hence, the instance can be solved by Theorem 2.1.

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