

# Max CSP and lattices

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# Outline:

- The Max CSP problem
- Tools: Lattices, supermodularity, implementations, and cores
- Results and open questions:
  - $|D| = 3$
  - non-distributive lattices
  - single-predicate Max CSP
  - Max CSP with all constants ( $x = 0, x = 1, \dots$ )

# Basics

$D$  – a finite set with  $|D| > 1$  (the *domain*)

$R_D^{(m)} = \{f \mid f : D^m \rightarrow \{0, 1\}\}$  – the set of all  $m$ -ary *predicates*

$$R_D = \bigcup_{m=1}^{\infty} R_D^{(m)}$$

**Definition:** A *constraint* over a set of variables  $V = \{x_1, x_2, \dots, x_n\}$  is an expression of the form  $f(\mathbf{x})$  where

$f \in R_D^{(m)}$  is the *constraint predicate*; and

$\mathbf{x} = (x_{i_1}, \dots, x_{i_m})$  is the *constraint scope*.

The constraint  $f(\mathbf{x})$  is said to be *satisfied* on a tuple  $\mathbf{a} = (a_{i_1}, \dots, a_{i_m}) \in D^m$  if  $f(\mathbf{a}) = 1$ .

# Max CSP

A collection  $C = \{f_1(\mathbf{x}_1), \dots, f_m(\mathbf{x}_m)\}$  of constraints over  $V = \{x_1, \dots, x_n\}$ ;

each constraint  $f_i(\mathbf{x}_i)$  has a weight  $\alpha_i \in \mathbb{N}$ .

Find an *assignment*  $\phi : V \rightarrow D$  that maximizes the total weight of satisfied constraints; in other words, maximize the function  $f : D^n \rightarrow \mathbb{N}$ , defined by

$$f(x_1, \dots, x_n) = \sum_{i=1}^m \alpha_i \cdot f_i(\mathbf{x}_i).$$

## Parameterization of Max CSP

For a finite set of predicates  $\Gamma \subseteq R_D$ , Max CSP( $\Gamma$ ) is the set of Max CSP instances where all constraint predicates belong to  $\Gamma$ .

We say that  $\Gamma$  is a *constraint language*.

## Example

In the Max  $k$ -Cut problem, one is given an edge-weighted graph, and the goal is to divide it into  $k$  parts so as to maximize the total weight of edges between different parts.

Let  $neq_k$  be the disequality predicate on  $\{0, \dots, k - 1\}$ , that is,  $neq_k(x, y) = 1$  iff  $x \neq y$ . Then Max  $k$ -Cut  $\equiv$  Max CSP( $\{neq_k\}$ ).

# Approximation

**PO** is the class of optimization problems that can be solved to optimality in polynomial time.

**APX** is the class of optimization problems that can be optimized (in polynomial time) within some constant  $c > 1$ :

$$\frac{OPT(I)}{c} \leq m(A(I)) \leq c \cdot OPT(I)$$

Max CSP( $\Gamma$ ) can be approximated within  $|D|^a$  where  $a$  is the maximum arity of predicates in  $\Gamma$ .

A problem  $S$  is **APX**-complete if every problem in **APX** can be  $AP$ -reduced to  $S$ .

If  $S$  is **APX**-complete and  $\mathbf{P} \neq \mathbf{NP}$ , then

- there exists a constant  $c > 0$  such that  $S$  is not  $c$ -approximable;
- $S$  does not admit a polynomial-time approximation scheme;
- it is **NP**-hard to solve  $S$  exactly.

Max CSP( $\{neq_k\}$ ) is **APX**-complete.



## Classification when $|D| = 2$

[Creignou] [Khanna, Sudan, Williamson]

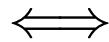
Let  $\Gamma$  be a constraint language over  $\{0, 1\}$ .  $\text{Max CSP}(\Gamma) \in \mathbf{PO}$  if and only if

- $\Gamma$  is 0-valid; or
- $\Gamma$  is 1-valid; or
- $\Gamma$  is 2-monotone.

Otherwise,  $\Gamma$  is **APX**-complete.

A predicate  $f : \{0, 1\}^n \rightarrow \{0, 1\}$  is 2-monotone if  $f$  can be expressed as follows:

$$f(x_1, \dots, x_n) = 1$$



$$(x_{i_1} \wedge \dots \wedge x_{i_s}) \vee (\neg x_{j_1} \wedge \dots \wedge \neg x_{j_t})$$

Both disjuncts are not required to contain literals.

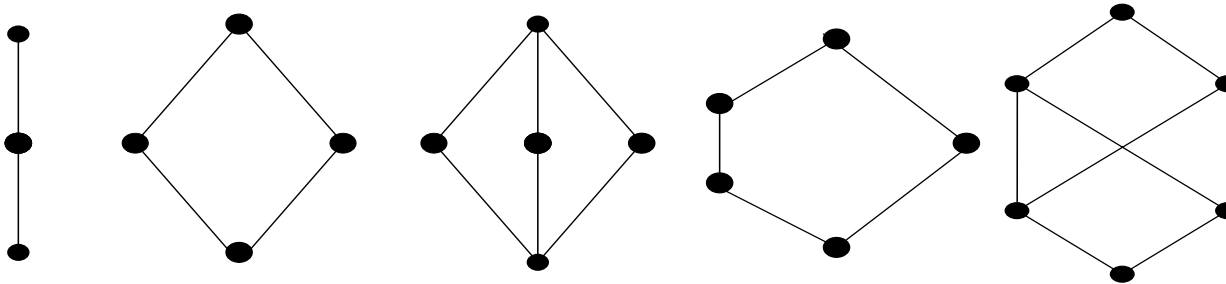
# Tools

- Lattices and supermodularity
- Strict implementations
- Cores

# Lattices

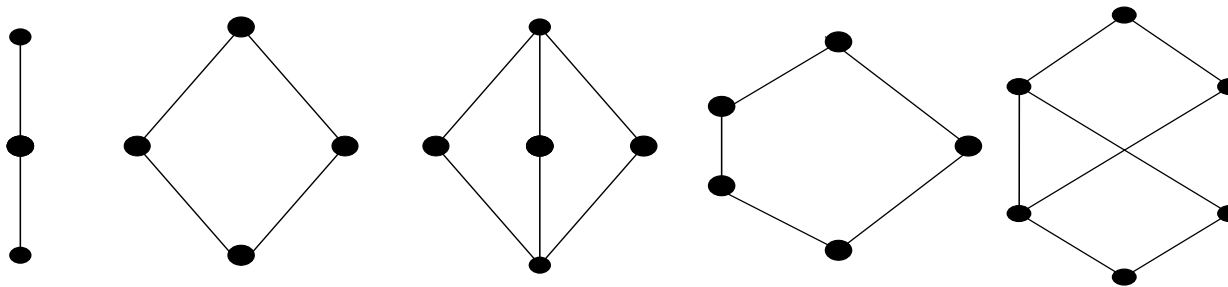
A *lattice*  $\mathcal{L}$  is a partial order in which any  $a, b \in \mathcal{L}$  have

- a least common upper bound (join)  $a \sqcup b$ , and
- a greatest common lower bound (meet)  $a \sqcap b$



A *chain* is a totally ordered lattice.

A lattice is called *distributive* iff it can be represented by subsets of a set, with lattice operations interpreted as union and intersection.

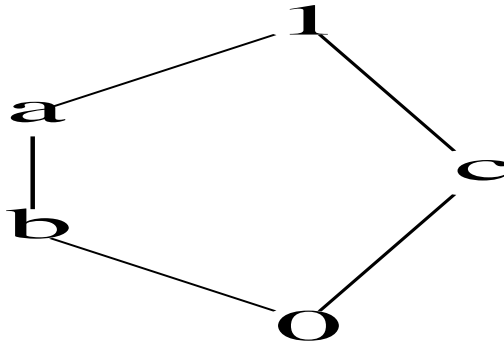


# Supermodular functions/predicates

Let  $\mathcal{L}$  be a lattice order on  $D$ . We say that an  $n$ -ary function  $f : D^n \rightarrow \mathbb{R}$  is *supermodular on  $\mathcal{L}$*  if

$$f(\mathbf{x}) + f(\mathbf{y}) \leq f(\mathbf{x} \sqcap \mathbf{y}) + f(\mathbf{x} \sqcup \mathbf{y}) \text{ for all } \mathbf{x}, \mathbf{y} \in D^n,$$

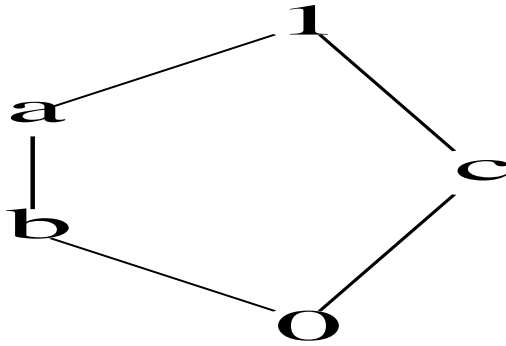
where  $\sqcup$  and  $\sqcap$  act point-wise.



$$f(a) = f(b) = f(c) = 1$$

$$f(0) = f(1) = 0$$

$$f(a) + f(c) = 2 \not\leq f(a \sqcap c) + f(a \sqcup b) = f(0) + f(1) = 0$$

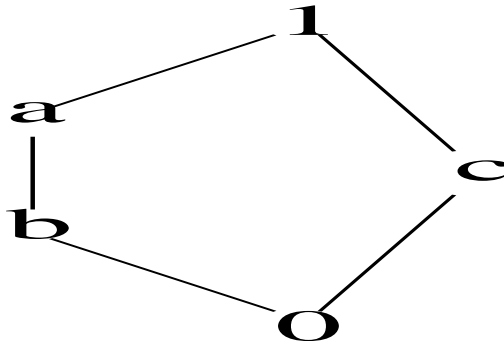


$$f(a) = f(b) = 1$$

$$f(c) = f(0) = f(1) = 0$$

$$f(a) + f(c) = 1 \not\leq f(a \sqcap c) + f(a \sqcup b) = f(0) + f(1) = 0$$





$$f(a) = f(b) = f(0) = f(1) = 1$$

$$f(c) = 0$$

$$x, y \in \{a, b, 0, 1\}$$

$$f(x) + f(c) = 1 \leq f(x \sqcap c) + f(x \sqcup c) = f(0) + f(1) = 2$$

$$f(x) + f(y) = 2 \leq f(x \sqcap y) + f(x \sqcup y) = f(x) + f(y) = 2$$

$$f(c) + f(c) = 0 \leq f(c \sqcap c) + f(c \sqcup c) = f(c) + f(c) = 0$$

## More examples

Every 2-monotone predicate is supermodular on  $0 \rightarrow 1$ .

Every unary predicate is supermodular on every chain.

## Max CSP and supermodularity

**Fact.** If  $f_1$  and  $f_2$  are supermodular predicates on  $\mathcal{L}$ , then  $\alpha \cdot f_1 + \beta \cdot f_2$ ,  $\alpha, \beta \geq 0$ , is supermodular on  $\mathcal{L}$ .

**Theorem.** [Schrijver]

Let  $\mathcal{L}$  be a distributive lattice order on a finite set  $D$ . A function  $f : D^n \rightarrow \mathbb{R}$  that is supermodular on  $\mathcal{L}$  can be maximized in polynomial time, if  $f$  and  $\mathcal{L}$  satisfy some mild restrictions.

Let  $\alpha_1, \dots, \alpha_m \geq 0$ . If predicates  $f_1, \dots, f_m$  are supermodular on  $\mathcal{L}$ , then so is

$$f(x_1, \dots, x_n) = \sum_{i=1}^m \alpha_i \cdot f_i(\mathbf{x}_i).$$

# Max CSP

A collection  $C = \{f_1(\mathbf{x}_1), \dots, f_m(\mathbf{x}_m)\}$  of constraints over  $V = \{x_1, \dots, x_n\}$ ;

each constraint  $f_i(\mathbf{x}_i)$  has a weight  $\alpha_i \in \mathbb{N}$ .

Find an *assignment*  $\phi : V \rightarrow D$  that maximizes the total weight of satisfied constraints; in other words, maximize the function  $f : D^n \rightarrow \mathbb{N}$ , defined by

$$f(x_1, \dots, x_n) = \sum_{i=1}^m \alpha_i \cdot f_i(\mathbf{x}_i).$$

**Theorem.** [Cohen, Cooper, Jeavons, Krokhin]

If  $\mathcal{L}$  is a distributive lattice and  $\Gamma$  consists of supermodular predicates on  $\mathcal{L}$ , then Max CSP( $\Gamma$ ) is in **PO**.

## Strict implementations

**Definition.** Let  $Y = \{y_1, \dots, y_m\}$  and  $Z = \{z_1, \dots, z_n\}$  be two disjoint sets of variables. Let  $g_1(\mathbf{y}_1), \dots, g_s(\mathbf{y}_s)$ ,  $s > 0$ , be constraints over  $Y \cup Z$ . If  $g(y_1, \dots, y_m)$  is a predicate such that the equality

$$g(y_1, \dots, y_m) = \max_Z \sum_{i=1}^s g_i(\mathbf{y}_i) - \alpha$$

is satisfied for all  $y_1, \dots, y_m$ , and some fixed  $\alpha > 0$ , then  $g$  is said to be *strictly implemented* from  $\{g_1, \dots, g_s\}$ .

**Lemma.** If a predicate  $g$  can be strictly implemented from  $\Gamma$  and  $\text{Max CSP}(\Gamma \cup \{g\})$  is **APX**-complete then so is  $\text{Max CSP}(\Gamma)$

How to strictly implement  $eq_2$  with  $neq_2$ :

$$eq_2(x, y) = \max_z (neq_2(x, z) + neq_2(y, z)) - 1$$

If  $x = y = 1$ , then let  $z = 0$ . Result: 1

If  $x = y = 0$ , then let  $z = 1$ . Result: 1

If  $x \neq y$ , then let  $z = 0$  (or  $z = 1$ ). Result: 0



# Cores

**Definition.** An *endomorphism* of  $\Gamma$  is a unary operation  $\gamma$  on  $D$  such that

$$f(a_1, \dots, a_m) = 1 \Rightarrow f(\gamma(a_1), \dots, \gamma(a_m)) = 1$$

for all  $f \in \Gamma$  and all  $(a_1, \dots, a_m) \in D^m$ . We will say that  $\Gamma$  is a *core* if every endomorphism of  $\Gamma$  is injective (i.e. a permutation).

**Intuition.** If  $\Gamma$  is *not* a core then  $\text{Max CSP}(\Gamma)$  reduces to a similar problem over a smaller domain obtained by removing elements *not in image*( $\gamma$ ).

**Fact.** For  $|D| = 2$ ,  $\Gamma$  is not a core iff there is  $a \in D$  such that  $f(a, \dots, a) = 1$  for all  $f \in \Gamma$ . In this case Max CSP( $\Gamma$ ) is trivial.

## Classification when $|D| = 2$ (version 2)

Let  $\Gamma$  be a constraint language over  $\{0, 1\}$  and assume (without loss of generality) that  $\Gamma$  is a core. Then,  $\text{Max CSP}(\Gamma) \in \mathbf{PO}$  if and only if  $\Gamma$  is supermodular on  $0 \rightarrow 1$ . Otherwise,  $\Gamma$  is **APX**-complete.

# Results and open questions

- $|D| = 3$
- Non-distributive lattices
- Single-predicate Max CSP
- Constraint languages that contain all constants

## Classification when $|D| = 3$

[Jonsson, Klasson, Krokhin]

Let  $\Gamma$  be a constraint language over  $\{0, 1, 2\}$  and assume (without loss of generality) that  $\Gamma$  is a core. Then,  $\text{Max CSP}(\Gamma) \in \mathbf{PO}$  if and only if  $\Gamma$  is supermodular on some chain over  $\{0, 1, 2\}$ . Otherwise,  $\Gamma$  is **APX**-complete.

The proof has many similarities with the proof for constraint languages with all constants.

pen question:

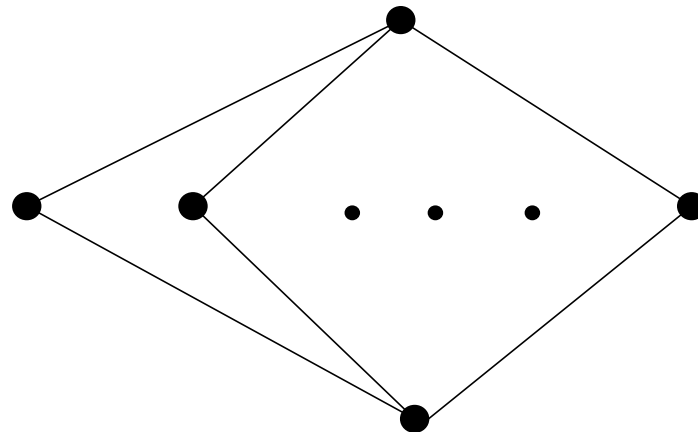
What is the complexity/approximability of Max CSP( $\Gamma$ ) when  $|D| > 3$ ?

# Hypothesis

Classification when  $|D| = k > 3$

Let  $\Gamma$  be a constraint language over  $\{0, \dots, k-1\}$  and assume that  $\Gamma$  is a core. Then,  $\text{Max CSP}(\Gamma) \in \mathbf{PO}$  if and only if  $\Gamma$  is supermodular on some distributive lattice over  $\{0, \dots, k-1\}$ . Otherwise,  $\Gamma$  is **APX**-complete.

There exist constraint languages  $\Gamma$  that are supermodular on



but not on any distributive lattice [Krokhin, Larose].



**Theorem.** [Krokhin, Larose]

If  $\Gamma$  consists of predicates that are supermodular on the  $k$ -diamond, then Max CSP( $\Gamma$ ) is in **PO**.

The algorithm runs in  $O(n^3)$  and it is inspired by algorithms for the Min Cut/Max Flow problem.

If  $\mathbf{V}, \mathbf{W}$  are classes of lattices, then  $\mathbf{V} \circ \mathbf{W}$  consists of all lattices  $\mathcal{L}$  such that there is a congruence  $\theta$  on  $\mathcal{L}$  with the following properties:

- the congruence lattice  $\mathcal{L}/\theta \in \mathbf{W}$ ; and
- every  $\theta$ -class is a lattice in  $\mathbf{V}$

**Theorem.** [Krokhin, Larose]

Suppose that  $\mathbf{V}, \mathbf{W}$  are finite classes of finite lattices. If supermodular optimization over  $\mathbf{V}$  and  $\mathbf{W}$  is in  $\mathbf{PO}$ , then supermodular optimization over  $\mathbf{V} \circ \mathbf{W}$  is in  $\mathbf{PO}$ , too.

**Corollary.**

If  $\Gamma$  consists of predicates that are supermodular on the pentagon, then  $\text{Max CSP}(\Gamma)$  is in  $\mathbf{PO}$ .

Let  $\Gamma$  be a core.

open question:

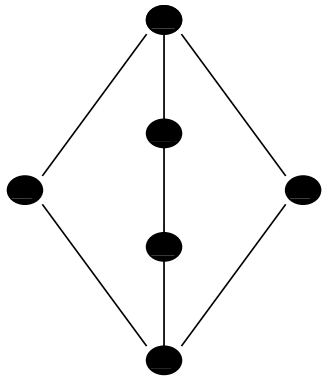
Is  $\text{Max CSP}(\Gamma) \in \mathbf{PO}$  whenever  $\Gamma$  is supermodular on *some* lattice?

open question:

Is  $\text{Max CSP}(\Gamma)$  **APX**-complete whenever  $\Gamma$  is not supermodular on *any* lattice?

pen question:

Assume that  $\Gamma$  is supermodular on



Is Max CSP( $\Gamma$ ) in **PO**?



# Complexity of single-predicate Max CSP

[Jonsson, Krokhin]

Let  $f : D^n \rightarrow \{0, 1\}$  be a predicate such that  $n > 1$ .

Max CSP( $\{f\}$ ) is in **PO** if and only if there exists a  $d \in D$  such that  $f(d, \dots, d) = 1$ . Otherwise, Max CSP( $\{f\}$ ) is **NP**-complete.

This is proved by two induction proofs. In the first part, it is assumed that  $f$  is binary and the induction is over  $|D|$ ; cores play an important rôle in the proof. In the second part, the induction is over the arity of  $f$ ; the main idea is to construct strict implementations that reduce the arity of predicates.

pen question:

Is  $\text{Max CSP}(\{f\})$  **APX**-complete whenever  $\text{Max CSP}(\{f\})$  is **NP**-complete?

# Constraint languages containing all constants

Given a finite set  $D'$ , we define the predicate  $u_{D'}$  such that

$$u_{D'}(x) = 1 \iff x \in D'.$$

Let  $\Gamma$  be a constraint language over domain  $D = \{0, \dots, d-1\}$ .  
 $\Gamma$  contains all constants if  $\{u_{\{0\}}, \dots, u_{\{d-1\}}\} \subseteq \Gamma$ .

Note:  $\Gamma$  is a core (the identity is the only endomorphism).

**Theorem.** [Deineko, Jonsson, Klasson, Krokhin]

Let  $\Gamma$  be a constraint language that contains all constants. Then,  $\text{Max CSP}(\Gamma) \in \mathbf{PO}$  if and only if  $\Gamma$  is supermodular on some chain. Otherwise,  $\text{Max CSP}(\Gamma)$  is **APX**-complete.



Every chain is a distributive lattice so we only need to prove the hardness part: Consequently, we assume that  $\Gamma$  is not supermodular on any chain over  $D$ .

**Step 1.** For every  $D' \subseteq D$ , the predicate  $u_{D'}$  can be strictly implemented by  $\Gamma$ . Henceforth, we assume that all unary predicates are in  $\Gamma$ .

**Step 2.**  $\Gamma$  contains all unary predicates. Then,  $\Gamma$  can strictly implement a constraint language  $\Gamma'$  such that  $\Gamma$  is not supermodular on any chain and every predicate in  $\Gamma'$  is at most binary. [Burkard, Klinz, Rudolf]

**Step 3.** If  $\Gamma$  is not supermodular on any chain, then there exists  $D' \subseteq D$  such that

- $|D'| \leq 4$ ; and
- $\Gamma|_{D'}$  is not supermodular on any chain.

The proof is inspired by how the COM-algorithm works [Deineko, Rudolf, Woeginger].

Is there an *AP*-reduction from  $\text{Max CSP}(\Gamma|_{D'})$  to  $\text{Max CSP}(\Gamma)$ ?

However:

Max CSP( $\Gamma|_{D'}$ )- $B$  *AP*-reduces to Max CSP( $\Gamma$ )- $B$ .

Strict implementations increase the degrees of variables, but not too much.

**Step 4.** If  $\Gamma$  is not supermodular on any chain, then there exists a subset  $\Gamma' \subseteq \Gamma$  such that

- $|\Gamma'| \leq 3$ ; and
- $\Gamma'$  is not supermodular on any chain.

By steps 1-4, we now have a constraint language  $\Gamma'$  satisfying the following properties:

- $\Gamma' = \{f_1, f_2, f_3\}$  where  $f_i : \{0, 1, 2, 3\}^2 \rightarrow \{0, 1\}$ ;
- $\Gamma'$  is not supermodular on any chain;
- $\text{Max CSP}(\Gamma')\text{-}B$  *AP*-reduces to  $\text{Max CSP}(\Gamma)\text{-}B$ .

By a computer-generated enumeration of strict implementations, it turns out that some predicate  $\neq_E$  with  $|E| = 2$  can be strictly implemented by every possible  $\Gamma'$ .

It is known that Max CSP( $\neq_E$ )-3 is **APX**-complete [Alimonti, Kann] which concludes the proof.

## pen question

Is there an elegant way of proving the previous result without using computer-assisted case analyses?



## References: Max CSP and lattices.

- [1] Nadia Creignou. A dichotomy theorem for maximum generalized satisfiability. *Journal of Computer and System Science* 51(3):511-522, 1995.
- [2] Sanjeev Khanna, Madhu Sudan, Luca Trevisan, and David Williamson. The approximability of constraint satisfaction problems. *SIAM Journal on Computing* 30(6):1863-1920, 2000.
- [3] Alexander Schrijver. A combinatorial algorithm minimizing submodular functions in polynomial time. *Journal of Combinatorial Theory Ser.B* 80:346-355,2000.
- [4] David Cohen, Martin Cooper, Peter Jeavons, and Andrei Krokhin. Supermodular functions and the complexity of Max CSP. *Discrete Applied Mathematics* 149(1-3):53-72, 2005.
- [5] Andrei Krokhin and Benoit Larose. Maximum constraint satisfaction on diamonds. Proc. CP-2005, 388-402, 2005.
- [6] Vladimir Deineko, Peter Jonsson, Mikael Klasson, and Andrei Krokhin. Supermodularity on chains and complexity of maximum constraint satisfaction problems. In *Proceedings of the European Conference on Combinatorics, Graph Theory and Applications (EuroCOMB-2005)*, pp. 51-56, 2005. Longer version available from [www.arxiv.org/ps/cs.CC/0602075](http://www.arxiv.org/ps/cs.CC/0602075).
- [7] Peter Jonsson, Mikael Klasson, and Andrei Krokhin. The approximability of three-valued MAX CSP. *SIAM Journal on Computing* 35(3):1329-1349, 2006. Draft version available from [www.arxiv.org/ps/cs.CC/0412042](http://www.arxiv.org/ps/cs.CC/0412042).