

# Submodular semilinear valued constraint satisfaction problems

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# VCSPs

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A **cost function** over  $D$  is any  $f: D^n \rightarrow \mathbb{Q}$ , for any  $n \in \mathbb{N}$ .

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The **valued constraint satisfaction problem** for  $\Gamma$ , **VCSP( $\Gamma$ )**, is a computational optimisation problem.

**INPUT:**

- a finite set  $V = \{x_1, \dots, x_n\}$  of variables, and
- an objective function  $\Phi(x_1, \dots, x_n) = \sum_{i=1}^k f_i(x_{i,1}, \dots, x_{i,q_i})$ , where  $1 \leq i \leq k$ ,  $x_{i,j} \in V$  and  $f_i$  is a cost function over  $D$ .

**GOAL:** find an **assignment** of labels (or **labeling**) to the variables that minimises  $\Phi$ .

## What we know

Let  $\Gamma$  be a valued constraint language over a finite domain  $D$ .

The computational complexity of  $\text{VCSP}(\Gamma)$  has attracted a lot of attention in the literature. All partial classifications were subsumed and generalised by:

**Theorem (Thapper and Živný, 2013)**

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Our goal is classify the computational complexity of VCSPs for semilinear languages.

# Semilinear VCSPs

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## Example

$$f: \mathbb{Q}^3 \rightarrow \mathbb{Q}$$

$$f(x, y, z) = \begin{cases} 5x + 7z & \text{if } x + y \leq 3 \\ -2 & \text{if } x + y > 3 \text{ and } \max(x, y) > z + 1 \\ \min(2y + 6, -1) & \text{otherwise} \end{cases}$$

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In a **semilinear VCSP** the underlying domain is  $\mathbb{Q}$  and the language is made up by semilinear cost functions.

# The Thapper&Živný's dichotomy

Let  $D$  be a finite set.

- Either  $\Gamma$  has a **symmetric fractional polymorphism** and  $\text{VCSP}(\Gamma)$  is in  $\text{P}$ ,
- or  $\text{VCSP}(\Gamma)$  is NP-hard.

## Fractional polymorphisms

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An  $m$ -ary **fractional operation**  $\omega$  on  $D$  is a probability distribution on  $\mathcal{O}_D^{(m)}$ .

The **support** of  $\omega$  is defined as  $Supp(\omega) = \{g \in \mathcal{O}_D^{(m)} \mid \omega(g) > 0\}$ .

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A  $m$ -ary fractional operation  $\omega$  on  $D$  with finite support is said to be a **fractional polymorphism** of a cost function  $f$  if, for any  $x_1, x_2, \dots, x_m \in D^n$ , we have

$$\sum_{g \in \text{Supp}(\omega)} \omega(g) f(g(x_1, x_2, \dots, x_m)) \leq \frac{1}{m} (f(x_1) + f(x_2) + \dots + f(x_m)),$$

where  $g$  is applied componentwise.

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where  $g$  is applied componentwise.

For a valued constraint language  $\Gamma$ ,  $fPol(\Gamma)$  denotes the set of fractional operations that are fractional polymorphisms of every cost function in  $\Gamma$ .

# Symmetric fractional polymorphisms

A ( $m$ -ary) fractional polymorphism is said to be **symmetric** if all operations  $g$  in its support is symmetric, i.e. for every permutation  $\pi \in \text{Sym}(1, \dots, m)$ , we have  $g(x_1, \dots, x_m) = g(x_{\pi(1)}, \dots, x_{\pi(m)})$ .



## Example: submodular semilinear languages

Let the domain  $D$  be totally ordered. We say that  $f: D^n \rightarrow \mathbb{Q}$  is **submodular** if for each  $x, y \in D^n$

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Submodularity is an important concept in discrete optimisation.

A cost function is submodular iff it has the (binary) fractional polymorphism  $\omega: \mathcal{O}_D^{(2)} \rightarrow [0, 1]$ ,

$$\omega(g) = \begin{cases} \frac{1}{2} & \text{if } g = \max \\ \frac{1}{2} & \text{if } g = \min \\ 0 & \text{if otherwise} \end{cases}$$

# Submodular semilinear VCSPs

$\Gamma$ : submodular language over a totally ordered domain  $D$ .

- If  $D$  is finite then the VCSP is in P (Cohen, Cooper, Jeavons, Krokhin).
- What is the computational complexity of VCSP( $\Gamma$ ) if  $\Gamma$  is a submodular semilinear language?

## Submodular semilinear functions

Examples of submodular semilinear cost functions:

- all unary cost functions are submodular;
- all linear cost functions are submodular;
- the cost function  $f: \mathbb{Q}^3 \rightarrow \mathbb{Q}, f(x_1, x_2, x_3) = \max(x_1, x_2, x_3)$  is submodular;
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### Lemma (Topkis, 1978)

*A binary function  $f: \mathbb{Q}^2 \rightarrow \mathbb{Q}$  is submodular if, and only if, for every  $\alpha_1 < \alpha_2$  and  $\beta_1 < \beta_2$  in  $\mathbb{Q}$*

$$f(\alpha_1, \beta_1) + f(\alpha_2, \beta_2) \leq f(\alpha_1, \beta_2) + f(\alpha_2, \beta_1).$$

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### Theorem (Topkis, 1978)

*$f: \mathbb{Q}^n \rightarrow \mathbb{Q}$  is submodular if and only if the (binary) projection to every plane parallel to one of the coordinate planes is submodular.*

## Submodular semilinear functions

A function  $f: \mathbb{Q}^n \rightarrow \mathbb{Q}$  is **separable** if  $f(x) = \sum_{i=1}^n f_i(x_i)$  for all  $x = (x_1, \dots, x_n)$ , with  $x_i \in \mathbb{Q}$  for  $i = 1, \dots, n$ .

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*If  $D_i$  is a chain (totally ordered set) for  $i = 1, \dots, n$ , then  $f$  is separable on  $\prod_{i=1}^n D_i$  if, and only if, both  $f$  and  $-f$  are submodular on  $\prod_{i=1}^n D_i$ .*



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## Proposition

*$f_i: \mathbb{Q} \rightarrow \mathbb{Q}$ ,  $i = 1, \dots, n$  finitely many unary semilinear cost functions. Then, find  $\inf_{x \in \mathbb{Q}} (f_1(x) + \dots + f_n(x))$  is in  $P$ .  
It follows that the VCSP for a language containing only separable semilinear cost functions is in  $P$ .*

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## Counterexample

$f: \mathbb{Q}^2 \rightarrow \mathbb{Q}, f(x_1, x_2) = \min(-x_1, -x_2 + 1)$ . It is minimum of non-increasing functions and it is not submodular. Take, for instance  $(2, 4), (5, -2) \in \mathbb{Q}^2$ .

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## Counterexample

$f: \mathbb{Q}^2 \rightarrow \mathbb{Q}, f(x_1, x_2) = \max(x_1, -x_2)$  is maximum of a non-decreasing function and a non-increasing function. It is not submodular: consider  $(-3, 2), (5, 1) \in \mathbb{Q}^2$ .

## The expressive power

Let  $\Gamma$  be a valued constraint language.

A  $k$ -ary cost function  $f$  is **expressible** over  $\Gamma$  if there exists an instance  $\mathcal{I}$  of  $VCSP(\Gamma)$  with objective function  $f_{\mathcal{I}}$  and with variables

$V = \{x_1, \dots, x_k, x_{k+1}, \dots, x_n\}$ , such that

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**Expressive power** of  $\Gamma$ : the set  $\langle \Gamma \rangle$  of all cost functions expressible over  $\Gamma$ .

**Remark:**  $\langle \Gamma \rangle$  is the closure of  $\Gamma$  under addition, non-negative scalar multiplication, minimisation over extra variables.

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**Proposition (Cohen, Cooper, Jeavons, 2006)**

$\Gamma$  valued constraint language over a finite domain. Then  $f\text{Pol}(\Gamma) = f\text{Pol}(\langle \Gamma \rangle)$ .

The proof works also for finite languages  $\Gamma$  over an infinite domain.

## Corollary

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$\Gamma$  is a **tame submodular semilinear language** if it satisfies the hypothesis of the corollary above.



## Algorithm

Consider the following objective function

$$\Phi(x, y, z) = f_1(x) + f_2(y) + \max(g_1(y), g_2(z)) + \min(h_1(x), h_2(z)).$$

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Where the **elementary unary functions** are:

$$f_1(x) = \begin{cases} 5x + 2 & x < 4 \\ 1 & x = 4 \\ 2x - 5 & x > 4 \end{cases}$$

$$f_2(y) = \begin{cases} -3y + 1 & y < -7 \\ -8 & y = -7 \\ y - 2 & y > -7 \end{cases}$$

$$g_1(y) = \begin{cases} 2y + 2 & y < 0 \\ 3 & y = 0 \\ y + 3 & y > 0 \end{cases}$$

$$g_2(z) = \begin{cases} z + 1 & z < 2 \\ 3 & z = 2 \\ 2z - 1 & 2 < z < 3 \\ 7 & z = 3 \\ 2z + 3 & z > 3 \end{cases}$$

$$h_1(x) = \begin{cases} x - 3 & x < -1 \\ 0 & x = -1 \\ x + 2 & x > -1 \end{cases}$$

$$h_2(z) = -z$$

# Algorithm

- Define  $B = \{-7, -1, 0, 2, 3, 4\}$  (special points).
- $(\mathbb{Q} \times E; \leq)$ , where  $E = \{-1, 0, 1\}$  and  $(a, b) \leq (c, d)$  iff  $a < c$  or  $a = c$  and  $b \leq d$ .

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$$\blacksquare \tilde{f}_1(x, \alpha) = \begin{cases} 5x + 2 & (x, \alpha) < (4, 0) \\ 1 & (x, \alpha) = (4, 0) \\ 2x - 5 & (x, \alpha) > (4, 0) \end{cases} \quad \tilde{f}_2(y, \alpha) = \begin{cases} -3y + 1 & (y, \alpha) < (-7, 0) \\ -8 & (y, \alpha) = (-7, 0) \\ y - 2 & (y, \alpha) > (-7, 0) \end{cases}$$

$$\tilde{g}_1(y, \alpha) = \begin{cases} 2y + 2 & (y, \alpha) < (0, 0) \\ 3 & (y, \alpha) = (0, 0) \\ y + 3 & (y, \alpha) > (0, 0) \end{cases} \quad \tilde{g}_2(z, \alpha) = \begin{cases} z + 1 & (z, \alpha) < (2, 0) \\ 3 & (z, \alpha) = (2, 0) \\ 2z - 1 & (2, 0) < (z, \alpha) < (3, 0) \\ 7 & (z, \alpha) = (3, 0) \\ 2z + 3 & (z, \alpha) > (3, 0) \end{cases}$$

$$\tilde{h}_1(x, \alpha) = \begin{cases} x - 3 & (x, \alpha) < (-1, 0) \\ 0 & (x, \alpha) = (-1, 0) \\ x + 2 & (x, \alpha) > (-1, 0) \end{cases} \quad \tilde{h}_2(z, \alpha) = -z$$

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- Every  $\tilde{f}$  is unary and inherits the monotonicity of  $f$ , therefore

$$\tilde{\Phi}((x, \alpha), (y, \beta), (z, \gamma)) = \tilde{f}_1(x, \alpha) + \tilde{f}_2(y, \beta) + \max(\tilde{g}_1(y, \beta), \tilde{g}_2(z, \gamma)) + \min(\tilde{h}_1(x, \alpha), \tilde{h}_2(z, \gamma))$$

is an instance of a VCSP for a new tame submodular semilinear language,  $\Gamma'$  over  $\mathbb{Q} \times E$ .

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- $\inf_{\mathbb{Q}} \Phi = \inf_{\mathbb{Q} \times E} \tilde{\Phi}$ .
- $D = \{(\alpha, 0), (\alpha, -1), (\alpha, 1) \mid \alpha \in B\}$  (finite).

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$$\tilde{\Phi}((x, \alpha), (y, \beta), (z, \gamma)) = \tilde{f}_1(x, \alpha) + \tilde{f}_2(y, \beta) + \max(\tilde{g}_1(y, \beta), \tilde{g}_2(z, \gamma)) + \min(\tilde{h}_1(x, \alpha), \tilde{h}_2(z, \gamma))$$

is an instance of a VCSP for a new tame submodular semilinear language,  $\Gamma'$  over  $\mathbb{Q} \times E$ .

- $\inf_{\mathbb{Q}} \Phi = \inf_{\mathbb{Q} \times E} \tilde{\Phi}$ .
- $D = \{(\alpha, 0), (\alpha, -1), (\alpha, 1) \mid \alpha \in B\}$  (finite).
- Cost functions in  $\Gamma'$  are still submodular over  $D \subset \mathbb{Q} \times E$ .



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## Fact

$\Gamma$  tame submodular semilinear language.

If  $\inf_{\mathbb{Q} \times E} \tilde{\Phi} = \inf_D \tilde{\Phi}$ , then there exists a polynomial-time reduction from a VCSP( $\Gamma$ ) to a VCSP for a submodular language over a finite domain.

In particular, VCSP( $\Gamma$ ) is in P.

## Next steps and open problems

- 1 Prove that  $\inf_{\mathbb{Q} \times E} \tilde{\Phi} = \inf_D \tilde{\Phi}$ .
- 2 Adapt the algorithm to the case in which all elementary unary cost function in the instance are linear (no special points).
- 3 Find a syntactic characterisation for all submodular semilinear functions.
- 4 Does  $fPol(\Gamma) = fPol(\Delta)$  implies  $\langle \Gamma \rangle = \langle \Delta \rangle$ ?

Thank you