

Robust Algorithms for CSPs

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Notation

Fix a finite set D

Constraint language Γ = finite set of relations on D

- CSP(Γ): given constraints $R_1(\mathbf{x}_1), \dots, R_q(\mathbf{x}_q)$ over a set of variables V , and such that each $R_i \in \Gamma$, decide if there is $\varphi : V \rightarrow D$ satisfying all constraints
- MAX CSP(Γ): optimisation, can repeat constraints
- Want: classify the complexity and approximability
- Many algorithmic and hardness results
- Many approaches: algebra, logic, graphs, analysis etc
- Many open questions

Algorithms for (Max) CSP: Good and Bad

- Exact polynomial-time algorithm for $\text{CSP}(\Gamma)$
 - Good: correctly identifies (un)satisfiable instances
 - Bad: 0.999-satisfiable vs. (< 0.001)-satisfiable?
- Approx polynomial-time algorithm for $\text{MAX CSP}(\Gamma)$
 - Good: treats all instances uniformly
 - Bad: satisfiable vs. 0.999-satisfiable?

Is there a good natural compromise between those two?

Robust Algorithms

- Call an algorithm for $\text{CSP}(\Gamma)$ **robust** [Zwick'98] if
 - it correctly identifies satisfiable instances, and
 - for a $(1 - \epsilon)$ -satisfiable instance, it outputs a $(1 - g(\epsilon))$ -satisfying assignment, where $g() \rightarrow 0$,
 - running time is polynomial in the size of instance, does not depend on ϵ
- Robust algorithms:
 - (almost) untouched for 12 years
 - meeting point of algebra and approximation

Previous Negative Results

3-LIN p : systems of linear 3-variable equations mod p

- Easy as a CSP
- Trivial $1/p$ -approximation algorithm for MAX CSP:
assign values to variables uniformly at random
- Best possible [Håstad'01]
- For any $\epsilon, \gamma > 0$, it is **NP**-hard to approximate within $1/p + \gamma$, even on $(1 - \epsilon)$ -satisfiable instances
- Hence no robust algorithm

Previous Positive Results

Restrict MAX CSP(Γ) to $(1 - \epsilon)$ -satisfiable instances,
algorithm finds, in expectation, $(1 - g(\epsilon))$ -sat assignment

- HORN- k -SAT $g(\epsilon) = \frac{\log k}{\log(1/\epsilon)}$ (LP-based) [Zwick'98]
- 2-SAT
 - $g(\epsilon) = 5 \cdot \epsilon^{1/3}$ (SDP-based) [Zwick'98]
 - $g(\epsilon) = O(\epsilon^{1/2})$ [Charikar, Makarychev²'09]
- UNIQUE GAMES- k
 - $g(\epsilon) = O(k^2 \epsilon^{1/5} \sqrt{\log(1/\epsilon)})$ [Khot'02]
 - $g(\epsilon) \approx 1 - O(k^{-\epsilon/2})$ [Charikar, Makarychev²'06]

The Issue with Equality

- Let $\text{CSP}(\Gamma)$ admit a robust algorithm with some $g(\epsilon)$
- Let $\Gamma_ =$ denote $\Gamma \cup \{=\}$.
- Assumption (A): $\text{CSP}(\Gamma_ =)$ admits a robust algorithm
 - need (A) to build general algebraic theory
 - current conjecture about RA does not rely on (A)
- Stronger A (SA): ditto + with $g' = \Omega(g)$
 - might need (SA) to study $g(\epsilon)$ quantitatively

Primitive Positive Definability

- Let Γ be a constraint language on D
- Let G be a sat instance of $\text{CSP}(\Gamma)$ with c constraints
- Let R be a projection of $\text{Sol}(G)$ on some coordinates
- Assume $\text{CSP}(\Gamma)$ admits RA with some $g(\epsilon)$
- Then $\text{CSP}(\Gamma \cup \{R\})$ admits RA with $c \cdot g(c\epsilon)$
- The same if G is an instance of $\text{CSP}(\Gamma_{=}) + (\text{A})/(\text{SA})$
 - logic/algebra lingo: R is **pp-definable** in Γ
- If Γ_1 and Γ_2 pp-define each other, treat them the same
- What controls pp-definability?

Polymorphisms by Example

Consider the relation $R = \{(0, 0, 0), (1, 0, 0), (0, 0, 1)\}$.

- The binary operation \min is a **polymorphism** of R .

$$\begin{array}{rcl} & \min & \min & \min \\ (& 1 & , & 0 & , & 0 &) & \in R \\ (& 0 & , & 0 & , & 1 &) & \in R \\ \hline (& 0 & , & 0 & , & 0 &) & \in R \end{array}$$

- The binary operation max is not.

Polymorphisms in Control

Let $\text{Pol}(\Gamma) = \{f \mid f \text{ is a polymorphism of each } R \in \Gamma\}$

Theorem 1 (Geiger'68; Bodnarchuk et al'69)

Γ_1 *pp-defines* Γ_2 *iff* $\text{Pol}(\Gamma_1) \subseteq \text{Pol}(\Gamma_2)$.

- Need to look at $\text{Pol}(\Gamma)$ rather than Γ
- “Nice” operations in $\text{Pol}(\Gamma) \Rightarrow$ “easy” $\text{CSP}(\Gamma)$
- Actually, algebra goes much deeper than that
- “Nice” = good identities, e.g.
 - $f(x, y) = f(y, x)$ min for HORN k -SAT
 - $m(x, x, y) = m(x, y, x) = m(y, x, x) = x$ 2-SAT
- Good identities \Rightarrow structure in relations/algebras

Important Technicalities

- Assume $f \in \text{Pol}(\Gamma)$ is unary non-bijective
- Apply f to any assignment, preserve sat constraints
- So, can ignore values in $D \setminus \text{Range}(f)$
- Take f with minimum range, remove $D \setminus \text{Range}(f)$
- What remains is the **core** of Γ (unique up to isomorph)
- If Γ_0 is the core of Γ then $\text{CSP}(\Gamma) = \text{CSP}(\Gamma_0)$
- If Γ is a core (of itself), and $\Gamma^c = \Gamma \cup \{\{a\} \mid a \in D\}$
 - Lemma: If Γ admits RA, then so does Γ^c

Bounded Width

For $l \leq k$, an (l, k) -strategy \mathcal{S} for I is a non-empty set of partial assignments with domain size $\leq l$ or k satisfying:

1. If $q \in \mathcal{S}$ and $p \subseteq q$ (i.e. q extends p) then $p \in \mathcal{S}$.
2. If $p \in \mathcal{S}$ and $|dom(p)| \leq l$ then, for each $u \in V$, there is $q \in \mathcal{S}$ with $|dom(q) = k|$ s.t. $p \subseteq q$ and $u \in dom(q)$.
3. If $p \in \mathcal{S}$ and C is a constraint in I whose variables are all contained in $dom(p)$ then p satisfies C .

Any satisfiable instance has an (l, k) -strategy for any l, k :
restrictions of any fixed satisfying assignment

$\text{CSP}(\Gamma)$ has width (l, k) if, $\forall I, (l, k)$ -strategy \Rightarrow satisfiable

Bounded Width Theorems

Bounded width = width (l, k) for some l, k

Wlog assume $\Gamma = \Gamma^c$.

Theorem 2 (Larose, Zádori'06; Valeriote'09)

If Γ pp-defines a constraint language whose core is 3LIN- p for some p then $\text{CSP}(\Gamma)$ does not have bounded width.

The above pp-definition can be chosen without equality.

Theorem 3 (Barto, Kozik'09; Bulatov'09)

In all other cases, $\text{CSP}(\Gamma)$ has width $(2, k)$ where $k = \max(3, \max ar)$.

The Guruswami-Zhou Conjecture

Guruswami and Zhou observed that

- the obstructions for bounded width are the same as the known obstructions for robust algorithms: $3\text{LIN-}p$
- For Boolean CSPs, $\text{BW}=\text{RA}$

Conjecture 1 (Guruswami,Zhou'11)

$\text{CSP}(\Gamma)$ admits a robust algorithm iff it has bounded width.

- The \Rightarrow direction holds by [Håstad] + previous slide.
- The \Leftarrow direction is the conjecture.

Width 1

- $\text{CSP}(\Gamma)$ has **width 1** if, $\forall I, (1, \text{maxar})\text{-strategy} \Rightarrow \text{sat}$
- HORN k -SAT has width 1
- 2-SAT and UG- k have bounded width, but not width 1

Theorem 4 (Kun et al'11; Dalmau,AK'11)

If $\text{CSP}(\Gamma)$ has width 1 then it admits a robust algorithm.

Reduction to (dual) HORN k' -SAT where $k' = \text{maxar} \cdot |D|$.

Define Γ' on domain $2^D \setminus \{\emptyset\}$ as follows [Feder, Vardi'98]

- $\Gamma' = \{R' \mid R \in \Gamma\}$
- $R' = \{(A_1, \dots, A_r) \mid \exists S \subseteq R \text{ s.t. } \forall i \ A_i = \text{pr}_i S\}$

Width 1 Cont'd

Facts about Γ' [Feder, Vardi'98]:

- Function $a \rightarrow \{a\}$ maps tuples in R to tuples in R'
- If $\text{CSP}(\Gamma)$ has width 1 then there exists $f : 2^D \rightarrow D$ (aka **set function**) mapping tuples in R' to tuples in R .
- Can transform an instance I of $\text{CSP}(\Gamma)$ to an instance I' of $\text{MAX CSP}(\Gamma')$ by replacing each R with R' .
- Can move between assignments for I and I' , preserving satisfied constraints
- Hence can replace Γ by Γ'

Width 1 Cont'd

- Obtain Γ'' by Booleanising Γ'
 - the domain of Γ'' is $\{0, 1\}$
 - $ar(R) = ar(R') = r \Rightarrow ar(R'') = r|D|$.
 - replace each $A \subseteq D$ by its characteristic vector
- $\text{MAX CSP}(\Gamma')$ and $\text{MAX CSP}(\Gamma'')$ are the same
- Trivially, \cup is a polymorphism of Γ'
- So, max is a polymorphism of Γ''
- Hence $\text{MAX CSP}(\Gamma'')$ is dual HORN k' -SAT where $k' = maxar \cdot |D|$
- Apply [Zwick'98], use set function f to go back to I

Beyond Width 1

Theorem 5 (Dalmau, AK'11) *If $d \in \text{Pol}(\Gamma)$ then $\text{CSP}(\Gamma)$ admits a robust algorithm with $g(\epsilon) = O(\epsilon^{1/10})$.*

The operation d (**dual discriminator**) is defined as

$$d(x, y, z) = \begin{cases} x & \text{if } x = y \\ z & \text{otherwise} \end{cases}$$

Can assume that relations in Γ are of three types [JCG'96]:

1. all unary relations (wlog of the form $D \setminus \{a\}$)
2. $(x = a \vee y = b)$ where $a, b \in D$ 2-SAT-like
3. $\{(a, \pi(a)) \mid a \in D\}$ where π is a permutation UG

SDP relaxation

Basic SDP for an instance I of $\text{CSP}(\Gamma)$ with vars V .

- Vector variables X_u^i , $u \in V$, $i \in D = \{1, \dots, k\}$.
(Rough intuition: $\|X_u^i\| \approx 1 \Leftrightarrow u$ is assigned i)
- SDP constraints

$$\|X_u^1\|^2 + \dots + \|X_u^k\|^2 = 1, \quad u \in V$$

$$X_u^i \cdot X_u^j = 0, \quad u \in V, \quad 1 \leq i \neq j \leq k$$

$$X_u^i \cdot X_v^j \geq 0, \quad u, v \in V, \quad 1 \leq i, j \leq k$$

$$\sum_{1 \leq i, j \leq k} X_u^i \cdot X_v^j = 1, \quad u, v \in V$$

Goal Function

- The goal function is $\sum_{C \in I} z_C$ where $z_C =$

$$\begin{cases} \sum_{1 \leq i \leq k} X_u^i \cdot X_v^{\pi(i)} & \text{if } C = ((u, v), \pi) \\ 1 - X_u^i \cdot X_u^i & \text{if } C = (u, D \setminus \{i\}) \\ 1 - \overline{X}_u^i \cdot \overline{X}_v^j & \text{if } C = (u = i \vee v = j) \end{cases}$$

where \overline{X}_u^i is a shorthand for $\sum_{1 \leq j \neq i \leq k} X_u^j$

2-Step Rounding

Combination of Zwick's 2-SAT and Khot's UG- k ideas.

- Set $S = \sum_{1 \leq i \leq k} X_u^i$ (independent of u , $\|S\| = 1$)
- Set $Y_u^i = 2X_u^i - S$, $u \in V, i \in D$ (again, $\|Y_u^i\| = 1$)
- Apply Zwick's rounding with appropriate ϵ to Y_u^i
 - rotate vectors Y_u^i towards the line given by S
 - take vector N_1 uniformly generated from the unit sphere, replace by $-N$ if necess to have $N_1 \cdot S > 0$
 - if there is a unique Y_u^i with $N_1 \cdot Y_u^i > 0$, set $u = i$
- Choose vector N_2 from distribution $N(0, 1)$, $N_2 \cdot S > 0$
- Set each undefined u to i such that $N_2 \cdot X_u^i$ is max

Mixed MAX CSP

In a **mixed** MAX CSP instance, constraints are of two sorts:

- **Hard** constraints that must be satisfied
- **Soft** constraints that can be falsified.

The case when all soft constraints are unary was studied:

- as MINCOSTCSP, complexity [Takhanov'10]
- as STRICT CSP, approximability [Kumar et al'11]

Robust Algorithms for MIXED CSP

- Can study robust algorithms for Mixed CSP
- Compute all fractions of the number of soft constraints
- Non-empty notion: MIXED UG- k admits RA
 - reduction to ordinary (soft) UG- k
 - eliminate all hard constraints
 - if $R_1(x, y)$ is hard, then, for each soft $R_2(y, z)$, replace $R_2(y, z)$ by $R_3(x, z)$ where $R_3 = R_1 \circ R_2$
 - eventually no soft constraint will touch y
 - now eliminate all hard constraints involving y

MIXED HORN k -SAT

Theorem 6 (Dalmau, AK'11)

If, for each fixed k' , MIXED HORN k' -SAT admits RA then the Guruswami-Zhou conjecture holds.

Proof Idea: Fix Γ such that $\text{CSP}(\Gamma)$ has bounded width.

Express existence of $(2, \text{maxar})$ -strategy for an instance I of $\text{CSP}(\Gamma)$ as an instance I' of MIXED HORN k' -SAT with $k' = \sum_{i=1}^{\text{maxar}} |D|^i$.

For each p with $|\text{dom}(p)| \leq \text{maxar}$, introduce a Boolean variable x_p stating “ p is **not** in the strategy”.

Since maxar is constant, polynomially many variables.

Clauses in I'

1. If $q \in \mathcal{S}$ and $p \subseteq q$ (i.e. q extends p) then $p \in \mathcal{S}$.
 - add hard $x_p \rightarrow x_q$
2. If $p \in \mathcal{S}$ and $|dom(p)| < k$ then for every $u \in V$ there exists some $q \in \mathcal{S}$ with $p \subseteq q$ and $u \in dom(q)$.
 - add hard $\bigvee x_q \rightarrow x_p$
where \bigvee is over all q with $u \in dom(q) \supseteq dom(p)$
3. If $p \in \mathcal{S}$ and C is a constraint in I whose scope is entirely contained in $dom(p)$ then p satisfies C
 - add soft x_p , for each C and p falsifying the above
4. non-empty strategy: add hard $\bigvee_p \bar{x}_p$

More about I'

- Since k is constant, I' has polynomially many constraints
- There are at most M different soft constraints in I' where $M = \sum_{i=1}^k |D|^i$
- Each soft constraint appears at most m times where m is the number of constraints in I
- Satisfying assignment φ' for $I' \Rightarrow$ strategy $\mathcal{S}_{\varphi'}$ for I consisting of all p with $x_p = 0$

Almost Satisfiable I'

Assume φ' fails to satisfy ϵ -fraction of soft constraints in I' .

- If a soft clause (x_p) is falsified then remove from I all constraints corresponding to it
- These form at most $(M \cdot \epsilon)$ -fraction of constraints in I
- Now $\mathcal{S}_{\varphi'}$ is a strategy for the remaining instance
- Bounded width algorithm finds a solution for it

Open Problems

- Settle the Guruswami-Zhou conjecture
- Study quantitative dependence of $g(\epsilon)$ on Γ
 - determined by $\text{Pol}(\Gamma)$, modulo (SA)
- Study Mixed CSPs
 - robust algorithm for MIXED HORN k -SAT ?
 - r -approx algorithm for MIXED HORN k -SAT ?
- Study CSPs with global constraints
 - MAX BISECTION = MAX CUT into equal parts
robust algorithm with $g(\epsilon) = O(\epsilon^{1/3} \log(1/\epsilon))$
(via expander decomposition) [Guruswami et al'11]