

Locally consistent constraint satisfaction problems

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CONSTRAINT SATISFACTION PROBLEMS

- constraint satisfaction problem (CSP)
universe U
constraint types $C_1 \subseteq U^{k_1}$, $C_2 \subseteq U^{k_2}$, \dots
- instance
set X of variables
constraints: k_i -tuples of variables for the i -th type of constraint
- goal
assignment $\varphi : X \rightarrow U$ such that each constraint is satisfied
- CSP includes satisfiability, graph 3-coloring, etc.
in particular, the problem is NP-complete in general
- graph 3-coloring formulated as CSP:
 $U = \{\color{red}\bullet, \color{green}\bullet, \color{blue}\bullet\}$, $X = V(G)$ and constraints correspond to edges

THE PROBLEM AND OUR RESULTS

- in practice, it is easy to identify small sets of inconsistent constraints
3-coloring — check all subgraphs on at most 6 vertices
- CSP instance is k -consistent if any k constraints can be satisfied
- What can be said about k -consistent instances?
 - bad news: definitely, not all constrained can be satisfied
 - good news: sometimes a substantial fraction of them can be satisfied
- we focus on the case when the constraints are Boolean predicates
design an asymptotically optimal approximation algorithm
analyze asymptotical behaviour as k tends to infinity
analyze completely predicates of small arity

NOTATION

- Π is a (finite) set of Boolean predicates
 $\Pi_{\text{SAT}} = \{(x_1), (x_1 \vee x_2), (x_1 \vee x_2 \vee x_3), \dots\}$
 $\Pi_{2\text{-SAT}} = \{(x_1), (x_1 \vee x_2)\}$
- Σ is a finite set of predicates from Π
 $\Sigma = \{(x), (\neg x \vee y), (\neg y \vee z), (\neg z)\}$ for $\Pi = \Pi_{2\text{-SAT}}$
- $\rho(\Sigma)$ denotes the fraction of predicates which can be satisfied
 $\rho(\Sigma) = 3/4$
- $\rho_k(\Pi) = \inf \rho(\Sigma)$ where Σ is a k -consistent system of predicates from Π
 $\rho_1(\Pi_{\text{SAT}}) = \rho_1(\Pi_{2\text{-SAT}}) = 1/2$
- $\rho_\infty(\Pi) = \lim_{k \rightarrow \infty} \rho_k(\Pi)$
- weighted variants $\rho_k^w(\Pi)$ and $\rho_\infty^w(\Pi)$

SAT PROBLEMS

- $\rho_2^w(\Pi_{\text{SAT}}) = \rho_2^w(\Pi_{2\text{-SAT}}) = \frac{\sqrt{5}-1}{2} \approx 0.6180$ [Lieberherr, Specker 1981]
- $\rho_3^w(\Pi_{\text{SAT}}) = \rho_3^w(\Pi_{2\text{-SAT}}) = 2/3$ [Lieberherr, Specker 1982]
a simple proof found by Yannakakis in 1992
- $\rho_4^w(\Pi_{\text{SAT}}) \approx 0.6992$ but $\rho_4^w(\Pi_{2\text{-SAT}}) > 0.6992$ [K. 2003]
- $\rho_\infty^w(\Pi_{\text{SAT}}) \leq 3/4$ [Huang, Lieberherr 1985]
- $\rho_\infty^w(\Pi_{\text{SAT}}) = \rho_\infty^w(\Pi_{2\text{-SAT}}) = 3/4$ [Trevisan 1997]
- Trevisan also addressed the case when Π_k is a set of all k -ary Boolean predicates and showed: $\rho_\infty^w(\Pi_k) = \rho_\infty(\Pi_k) = 2^{1-k}$

1-EXTENDABLE PREDICATES

- a Boolean predicate P is 1-extendable if the following holds:
if one of its arguments is fixed, the remaining ones can be chosen in such a way that the predicate is satisfied.
example: $P_1(x, y, z) = (x \vee y \vee z)$ vs. $P_2(x, y, z) = x \wedge (y \Leftrightarrow z)$
- $\sigma(P)$ is the fraction of satisfying assignments of P
example: $\sigma(P_1) = 7/8$
- $\rho_\infty^w(\{P\}) = \rho_\infty(\{P\}) = \sigma(P)$ if the arity of P is at least two
- upper bound: take a random assignment
- lower bound:
take a random hypergraph of high girth and associate predicates with its edges
random \implies you cannot do better than in random assignment
high girth \implies you do not have small sets of inconsistent predicates

GENERAL CASE

- a finite set Π , want to determine $\rho_\infty^w(\Pi)$ and $\rho_\infty(\Pi)$
- a restriction P' of a predicate P is obtained by fixing some of the arguments restriction can be described by a vector $\tau \in \{0, 1, \star\}^r$
 $P'(x, y) = P(x, y, 0) = x \wedge y$ where $P(x, y, z) = x \wedge (y \vee z)$ and $\tau = \star \star 0$
- $\pi_{P, \tau}(p) : \langle 0, 1 \rangle \rightarrow \langle 0, 1 \rangle$ is the probability that P is satisfied if x_i is true with the probability $1 - p$, p and $1/2$, if τ_i is 0, 1 and \star , respectively.
 $\pi_{P, \tau}(p)$ is a polynomial in p
- $\pi(\Pi)$ is the set of all $\pi_{P, \tau}$ for $P \in \Pi$ where the restriction of P is 1-extendable
- $\overline{\pi(\Pi)}$ is the set convex hull of $\pi(\Pi)$

$$\rho_\infty^w(\Pi) = \min_{f \in \overline{\pi(\Pi)}} \max_{p \in [0, 1]} f(p)$$

The same holds for $\rho_\infty(\Pi)$ if Π does not contain a predicate of arity one.

EXAMPLE 1

- $\Pi = \{P_1, P_2\}$ where $P_1(x_1) = (x_1)$ and $P_2(x_1, x_2) = (x_1 \vee x_2)$
- a single restriction of P_1 is 1-extendable
- four restriction of P_2 are 1-extendable: 11, 1 \star , \star 1 and $\star\star$
- $\pi_{P_1,1}(p) = p$, $\pi_{P_2,11}(p) = 2p - p^2$, $\pi_{P_2,1\star}(p) = \pi_{P_2,\star 1}(p) = (p + 1)/2$,
 $\pi_{P_2,\star\star}(p) = 3/4$.
- $\pi(\Pi) = \{p, 2p - p^2, (p + 1)/2, 3/4\}$
- $\rho_\infty^w(\Pi) = \min_{f \in \overline{\pi(\Pi)}} \max_{p \in [0,1]} f(p)$
each function in $\pi(\Pi)$ is at least $3/4$ for $p = 3/4$
 $\rho_\infty^w(\Pi) = 3/4$ [Trevisan 1997]

EXAMPLE 2

- $\Pi = \{P\}$ where $P(x_1, x_2, x_3, x_4, x_5) = (x_1 \wedge (x_2 \vee x_3 \vee x_4 \vee x_5))$
- several 1-extendable restrictions, but each is symmetric to one of the following:
1 * * * *, 10 * * *, 11 * * *, 100 * *, 110 * *, 111 * *, 1100*, 1100*, 1110*, 1111*,
11000, 11100, 11110 and 11111.

- $\rho_\infty^w(\Pi) = \min_{f \in \overline{\pi(\Pi)}} \max_{p \in [0,1]} f(p)$

$$\pi_{P,100**}(p) = p(1 - p^2/4)$$

$$\max \pi_{P,100**}(p) = 3/4$$

each function in $\pi(\Pi)$ is at least $3/4$ for $p = 3/4$

$$\rho_\infty^w(\Pi) = \rho_\infty(\Pi) = 3/4$$

SKETCH OF THE UPPER BOUND PROOF

- fix Π and $\varepsilon > 0$
- classify the variables according to the number of constraints forcing their values in the algorithm, one rather uses the depth of “forcing”
 $X_1, \dots, X_{2R/\varepsilon}$ and Y
- this also partitions constraints to $2R/\varepsilon$ levels
- choose i such that the fraction of constraints in the i -th level is small
it is at most $\varepsilon/2$
- fix variables from X_1, \dots, X_i to their forced values (0/1)
keep remaining variables free (\star)
- this leads to a polynomial $f(p)$ from $\overline{\pi(\Pi)}$
- take a random assignment for p that maximizes $f(p)$

ALGORITHMIC ASPECTS

- depth of “forcing” can be easily computed in time $O(1/\varepsilon)$
inconsistence found \implies the input is not sufficiently locally consistent
- the maximum of the polynomial $f(p)$ can be found with precision $\varepsilon/2$
in time $O(1/\varepsilon)$ (bounded derivatives)
- derandomization by the method of conditional expectations
- robust algorithm linear in the size of input and $1/\varepsilon$

GRAPH HOMOMORPHISMS

- a single binary constraint $\Gamma \subseteq U^2$ for a large universe U
- the target oriented graph H
 $V(H) = U$ and $wv \in E(H)$ if $(u, v) \in \Gamma$
- instance can be viewed as graph G with $V(G) = X$ and arcs corresponding to input constraints
- k -consistent: every $G' \subseteq G$ of size at most k can be mapped to H ($G \xrightarrow{k} H$)
- find a largest subgraph $G_0 \subseteq G$ that can be mapped to H
- the corresponding ratio $\rho_k(H) = \min_{G \xrightarrow{k} H} \frac{\|G_0\|}{\|G\|} \max_{G_0 \subseteq G, G_0 \rightarrow H}$
- if H contains an oriented cycle, then $\rho_k(H)$ is the density of the densest subgraph of H
if H contains a loop, then $\rho_k(H) = 1$

OPEN PROBLEMS

- determine $\rho_k(H)$ for acyclic oriented graph H
- determine $\rho_k(H)$ for oriented trees H
- determine $\rho_k(H)$ for consistently oriented paths H

GENERAL SETTING

- higher arity leads to hypergraph homomorphisms
- more constraints leads to edge-colored (hyper)graphs