

Constraints, Graphs, Algebra, Logic, and complexity

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Constraint Satisfaction Problem (CSP)

Input: (V, D, C) :

- A finite set V of *variables*
- A finite set D of *values*
- A finite set C of *constraints* restricting the values that tuples of variables can take.

Constraint: (t, R)

- t : a tuple of variables over V
- R : a relation of arity $|t|$

Solution: $h : V \rightarrow D$

- $h(t) \in R$: for all $(t, R) \in C$

Question: Does (V, D, C) have a solution? I.e., is there an assignment of values to the variables such that all constraints are satisfied?

Constraint Satisfaction

Applications:

- belief maintenance
- machine vision
- natural language processing
- planning and scheduling
- temporal reasoning
- type reconstruction
- bioinformatics
- ...

3-Colorability

3-COLOR: Given an undirected graph $A = (V, E)$, is it 3-colorable?

- The variables are the nodes in V .
- The values are the elements in $\{\mathbf{R}, \mathbf{G}, \mathbf{B}\}$.
- The constraints are $\{(\langle u, v \rangle, \rho) : (u, v) \in E\}$, where $\rho = \{(R, G), (R, B), (G, R), (G, B), (B, R), (B, G)\}$.

Introduction to Database Theory

Basic Concepts:

- *Relation Scheme*: a set of attributes
- *Tuple*: mapping from relation scheme to data values
- *Tuple Projection*: if t is a tuple on P , and $Q \subseteq P$, then $t[Q]$ is the restriction of t to Q .
- *Relation*: a set of tuples over a relation scheme
- *Relational Projection*: if R is a relation on P , and $Q \subseteq P$, then $R[Q]$ is the relation $\{t[Q] : t \in R\}$.
- *Join*: Let R_i be a relation over relation scheme S_i . Then $\bowtie_i R_i$ is a relation over the relation scheme $\cup_i S_i$ defined by $\bowtie_i R_i = \{t : t[S_i] \in R_i\}$.

Database Perspective of CSP

Given: $(V, D, \{C_1, \dots, C_m\})$, where $C_i = (t_i, R_i)$.

Assume (wlog): Each t_i consists of distinct elements.

Database Perspective:

- V : attributes
- D : values
- (t_i, R_i) : relation R_i over relation scheme t_i

Fact: (Bibel, Gyssens, Jeavons, Cohen)

$(V, D, \{C_1, \dots, C_m\})$ has a solution iff $\bowtie_1^m R_i$ is nonempty.

Homomorphisms

Homomorphism: Let $\mathbf{A} = (A, R_1^{\mathbf{A}}, \dots, R_m^{\mathbf{A}})$ and $\mathbf{B} = (B, R_1^{\mathbf{B}}, \dots, R_m^{\mathbf{B}})$ be two relational structures.

$h : A \rightarrow B$ is a *homomorphism* from \mathbf{A} to \mathbf{B} if for every $i \leq m$ and every tuple $(a_1, \dots, a_n) \in A^n$,

$$R_i^{\mathbf{A}}(a_1, \dots, a_n) \implies R_i^{\mathbf{B}}(h(a_1), \dots, h(a_n)).$$

The Homomorphism Problem: Given relational structures \mathbf{A} and \mathbf{B} , is there a homomorphism $h : \mathbf{A} \rightarrow \mathbf{B}$?

Example: An undirected graph $\mathbf{A} = (V, E)$ is 3-colorable



there is a homomorphism $h : \mathbf{A} \rightarrow K_3$, where K_3 is the *3-clique*.

Homomorphism Problems

Examples:

- k -Clique: $K_k \xrightarrow{h} (V, E)?$
- Hamiltonian Cycle: $(V, C_{|V|}, \neq) \xrightarrow{h} (V, E, \neq)?$
- Subgraph Isomorphism: $(V, E, \overline{E}) \xrightarrow{h} (V', E', \overline{E'})?$
- s - t Connectivity: $(V, E, \{\langle s, t \rangle\}) \xrightarrow{h} (\{0, 1\}, =, \neq)?$

Fact: (Levin, 1973)

The homomorphism problem is NP-complete.

CSP vs. Homomorphisms

From CSP to Homomorphism:

Given: $(V, D, \{C_1, \dots, C_m\})$, where $C_i = (t_i, R_i)$.

Define **A**, **B**:

- $\mathbf{A} = (V, \{t_1\}, \dots, \{t_m\})$
- $\mathbf{B} = (D, R_1, \dots, R_m)$

Fact: (V, D, C) has a solution iff there is homomorphism from **A** to **B**.

CSP vs. Homomorphisms

From Homomorphism to CSP:

Given: $\mathbf{A} = (A, R_1^A, \dots, R_m^A)$, $\mathbf{B} = (B, R_1^B, \dots, R_m^B)$.

Define (V, D, C) :

- $V = A$: elements of \mathbf{A} are variables.
- $D = B$: elements of \mathbf{B} are values.
- $C = \{(t, R_i^B) : t \in R_i^A\}$: constraints derived from \mathbf{A}, \mathbf{B} .

Fact: There is homomorphism from \mathbf{A} to \mathbf{B} iff (V, D, C) has a solution.

Conclusion: CSP=Homomorphism Problem

- Feder&V., 1993
- Garey&Johnson, 1979: Homomorphism in, CSP not.

Uniform CSP vs. Non-Uniform CSP

Uniform CSP:

$$\{(\mathbf{A}, \mathbf{B}) : \exists \text{ homomorphism } h : \mathbf{A} \rightarrow \mathbf{B}\}$$

Complexity of Uniform CSP: NP-complete

Non-uniform CSP: Fix a structure \mathbf{B}

$$\text{CSP}(\mathbf{B}) = \{\mathbf{A} : \exists \text{ homomorphism } h : \mathbf{A} \rightarrow \mathbf{B}\}$$

Complexity of Non-Uniform CSP: Depends on \mathbf{B}

- $\text{CSP}(K_2)$ is in PTIME (2-COLORABILITY)
- $\text{CSP}(K_3)$ is NP-complete (3-COLORABILITY)

Complexity of Non-Uniform CSP

Research Program:

Identify the tractable cases of non-uniform CSP

Dichotomy Conjecture: (Feder&V., 1993)

For every structure \mathbf{B} ,

- either $\text{CSP}(\mathbf{B})$ is in PTIME
- or $\text{CSP}(\mathbf{B})$ is NP-complete.

Recall: $P \neq NP \Rightarrow NP - NPC - P \neq \emptyset$ (Ladner, 1975)

Intuition: CSP is not expressive enough to diagonalize over PTIME.

“Evidence” for the Conjecture

“Evidence 1”: (Hell&Nešetřil, 1990)

Let \mathbf{B} be an *undirected* graph.

- \mathbf{B} bipartite \implies $\text{CSP}(\mathbf{B})$ is in PTIME
- \mathbf{B} non-bipartite \implies $\text{CSP}(\mathbf{B})$ is NP-complete

Intuition: Every undirected graph homomorphism problem is equivalent either to 2-COLOR or 3-COLOR.

More “Evidence”: Boolean CSP

$$B = \{0, 1\}$$

E.g.: 2-SAT

B:

$$x \vee y: \begin{array}{|c|c|} \hline 0 & 1 \\ \hline 1 & 0 \\ \hline 1 & 1 \\ \hline \end{array} \quad \neg x \vee y: \begin{array}{|c|c|} \hline 0 & 0 \\ \hline 0 & 1 \\ \hline 1 & 1 \\ \hline \end{array} \quad \neg x \vee \neg y: \begin{array}{|c|c|} \hline 0 & 0 \\ \hline 0 & 1 \\ \hline 1 & 0 \\ \hline \end{array}$$

Dichotomy Theorem: (Schaefer, 1978)

Let **B** have a *Boolean* domain, then

- either **B** is trivial, Horn, anti-Horn, disjunctive, or affine, and $\text{CSP}(\mathbf{B})$ is in PTIME,
- otherwise $\text{CSP}(\mathbf{B})$ is NP-complete.

Dichotomy and Classification

Question: How far from CSP we need go to get a provable dichotomy?

Feder&V., 1993: It suffices to consider directed graphs to settle the Dichotomy Conjecture!

Classification Question:

For a given structure \mathbf{B} ,

- when is $\text{CSP}(\mathbf{B})$ in PTIME?
- when is $\text{CSP}(\mathbf{B})$ NP-complete?

Recent Progress on the Dichotomy Conjecture

Theorem: [Bulatov, 2002]

The Dichotomy Conjecture holds when $|B| = 3$.

Definition: A relational structure $\mathbf{B} = (B, R_1^{\mathbf{B}}, \dots, R_m^{\mathbf{B}})$ is *conservative* if it contains all possible monadic relations over the domain of the structure.

Intuition: All possible constraints over individual variables are available.

Theorem: [Bulatov, 2003]

The Dichotomy Conjecture holds when B is conservative.

Sources of Tractability

Empirical Observation: Feder&V., 1993

All known tractable CS problems can be explained as

- *combinatorial* (Datalog)
- *algebraic* (group-theoretic)

Classification Conjecture: (Feder&V., 1993)

Two explanations for tractability of $\text{CSP}(\mathbf{B})$

- Datalog
- group-theoretic

Bulatov, 2002 showed that the group-theoretic explanation is too weak – more general algebraic techniques required.

Datalog and Non-Uniform CSP

Example: NON 2-COLORABILITY

$$O(X, Y) : - E(X, Y)$$

$$O(X, Y) : - O(X, Z), E(Z, W), E(W, Y)$$

$$Q : - O(X, X)$$

Recall: Datalog \subseteq PTIME

Define: $\overline{\text{CSP}(\mathbf{B})} = \{\mathbf{A} : \mathbf{A} \notin \text{CSP}(\mathbf{B})\}$.

Datalog vs. Non-Uniform CSP: Explanation for many tractability results

- $\overline{\text{CSP}(\mathbf{B})}$ is expressible in Datalog

Note: $\overline{\text{CSP}(\mathbf{B})}$ is positively monotone.

k -Datalog

Definition:

- k -Datalog: Datalog with at most k variables per rule (NON 2-COLORABILITY is in 4-Datalog)
- $\exists\text{IL}^k$: k -variable existential positive infinitary logic
 - variables: x_1, \dots, x_k
 - no universal quantifiers
 - no negations
 - infinitary conjunctions and disjunctions

Facts: Fix $k \geq 1$

- k -Datalog $\subset \exists\text{IL}^k$
- $\exists\text{IL}^k$ can be characterized in terms of *existential k -pebble games* between the *Spoiler* and the *Duplicator*.
- There is a PTIME algorithm to decide whether the *Spoiler* or the *Duplicator* wins the existential k -pebble game.

Existential k -Pebble Games

A, B: structures

- **Spoiler:** *places on or removes* a pebble from an element of A.
- **Duplicator:** tries to duplicate move on B.

A: a_1, a_2, \dots, a_l $l \leq k$

B: b_1, b_2, \dots, b_l

- *Spoiler wins:* $h(a_i) = b_i, 1 \leq i \leq l$ is not a homomorphism.
- *Duplicator wins:* otherwise.

Fact: (Kolaitis&V., 1995)

B satisfies the same $\exists \text{IL}^k$ sentences as A iff the Duplicator wins the existential k -pebble game on A, B.

k -Datalog and CSP

Theorem: (Kolaitis&V., 1998): TFAE for $k \geq 1$ and a structure \mathbf{B} :

- $\overline{\text{CSP}(\mathbf{B})}$ is expressible in k -Datalog
- $\overline{\text{CSP}(\mathbf{B})}$ is expressible in $\exists\text{IL}^k$
- $\text{CSP}(\mathbf{B}) = \{\mathbf{A} : \text{Duplicator wins the existential } k\text{-pebble game on } \mathbf{A} \text{ and } \mathbf{B}\}.$

Intuition: $\overline{\text{CSP}(\mathbf{B})} \in k\text{-Datalog}$ implies that existence of homomorphism is equivalent to the Duplicator winning the existential k -pebble game.

k -Datalog and CSP

Proposition: (Kolaitis&V., 1998)

For a fixed structure \mathbf{B} , there is a k -Datalog program $\rho_{\mathbf{B}}^k$ such that $\rho_{\mathbf{B}}^k(\mathbf{A})$ is nonempty iff the Spoiler wins the existential k -pebble game on \mathbf{A}, \mathbf{B} .

$\rho_{\mathbf{B}}^k$:

- If $\rho_{\mathbf{B}}^k(\mathbf{A})$ is nonempty, then $\mathbf{A} \notin \text{CSP}(\mathbf{B})$.
- If $\overline{\text{CSP}(\mathbf{B})}$ is definable in k -Datalog, then it is definable by $\rho_{\mathbf{B}}^k$.
- *Open question:* Decide for a given \mathbf{B} whether $\overline{\text{CSP}(\mathbf{B})}$ is definable by $\rho_{\mathbf{B}}^k$.

Classification Questions

For a given structure \mathbf{B} :

- Is $\overline{\text{CSP}(\mathbf{B})}$ in k -Datalog, for a fixed $k > 0$?
- Is $\overline{\text{CSP}(\mathbf{B})}$ in k -Datalog, for some $k > 0$?

Group Theory

Example: Affine satisfiability - linear equations mod 2

$$x_1 - x_2 + x_3 = 1$$

$$x_1 + x_2 - x_3 = 1$$

Definition: $CSP(\mathbf{B}) \in \text{Subgroup}$ if there is a finite group G such that each k -ary relation in \mathbf{B} is a coset of G^k .

Theorem: Feder&V., 1993

$CSP(\mathbf{B}) \in \text{Subgroup}$ implies $CSP(\mathbf{B}) \in PTIME$.

Jeavons et al.: extensions of the algebraic framework.

The Product Operation

Definition: Let $G_1 = (V_1, E_1)$ and $G_2 = (V_2, E_2)$ be two graphs. The *product* of these graphs is the graph $G_1 \times G_2 = (V_1 \times V_2, E_1 \times E_2)$, where $(\langle u, u' \rangle, \langle v, v' \rangle) \in E_1 \times E_2$ iff $(u, v) \in E_1$ and $(u', v') \in E_2$.

Note: This definition can be extended to pairs of relational structures.

Polymorphisms

Definition: Let $\mathbf{B} = (B, R_1^{\mathbf{B}}, \dots, R_m^{\mathbf{B}})$ be a relational structure. A k -ary *polymorphism* is a homomorphism $f : \mathbf{B}^k \rightarrow \mathbf{B}$ (*closure condition*).

$Poly(\mathbf{B})$: set of polymorphisms of \mathbf{B}

Theorem: [Bulatov&Krokhin&Jeavons, 2000]

$Poly(\mathbf{B}_1) = Poly(\mathbf{B}_2) \Rightarrow CSP(\mathbf{B}_1) \equiv_p CSP(\mathbf{B}_2)$.

Conclusion: $Poly(\mathbf{B})$ characterizes the complexity of $CSP(\mathbf{B})$.

The Algebraic Approach to CSP: Study $Poly(\mathbf{B})$.

Definition: A *Maltsev operation* is a ternary function f such that $f(a, a, b) = f(b, a, a) = b$ for all a, b in its domain.

Theorem [Bulatov, 2002]

If $Poly(\mathbf{B})$ contains a Maltsev operation, then $CSP(\mathbf{B})$ is in PTIME.

Back to Datalog

Definition: A k -ary *near-unanimity operation* is a k -ary function f such that $f(x_1, x_2, \dots, x_k) = a$ whenever at least $k - 1$ of the x_i 's equal a .

Example: Majority is a near-unanimity operation.

Theorem: [Feder&V., 1993]

If $\text{Poly}(\mathbf{B})$ contains a near-unanimity function, then $\overline{\text{CSP}}(\mathbf{B})$ is definable in Datalog.

More on Datalog

Definition: A k -ary *weak near-unanimity operation* is a k -ary function f such that $(a, a, \dots, a) = a$, and $f(b, a, \dots, a) = f(a, b, a, \dots, a) = \dots = f(a, a, \dots, b)$, for all a, b in the domain.

Definition: A structure B is a *core* if every homomorphism $h : \mathbf{B} \rightarrow \mathbf{B}$ is an isomorphism.

WLOG: Restrict attention to cores

Theorem: [Barto&Kozik, 2009]

$\overline{CSP}(\mathbf{B})$ is definable in Datalog iff $Poly(B)$ contains weak near-unanimity operations for all sufficiently large arities. This condition can be checked in exponential time.

Uniform Tractability

General Problem: $\text{CSP}(\mathcal{C}, \mathcal{D})$, where \mathcal{C}, \mathcal{D} are classes of structures

- is there a homomorphism from \mathbf{A} to \mathbf{B} , where $\mathbf{A} \in \mathcal{C}$ and $\mathbf{B} \in \mathcal{D}$.

Question: When is $\text{CSP}(\mathcal{C}, \mathcal{D})$ tractable?

- Non-uniform case: $\text{CSP}(\text{All}, \mathbf{B})$ for a fixed structure \mathbf{B} .

Another important case: When is $\text{CSP}(\mathcal{C}, \text{All})$ tractable?

Bounded Treewidth

Definition: A *tree decomposition* of a structure $\mathbf{A} = (A, R_1, \dots, R_m)$ is a labeled tree T such that

- Each label is a non-empty subset of A ;
- For every R_i and every $(a_1, \dots, a_n) \in R_i$, there is a node whose label contains $\{a_1, \dots, a_n\}$.
- For every $a \in A$, the nodes whose label contain a form a subtree.

The *treewidth* $\text{tw}(\mathbf{A})$ of \mathbf{A} is defined by

$$\text{tw}(\mathbf{A}) = \min_T \{ \max \{ \text{label size in } T \} \} - 1$$

Note: Generalizes the *treewidth* of a *graph*.

Tree Decomposition

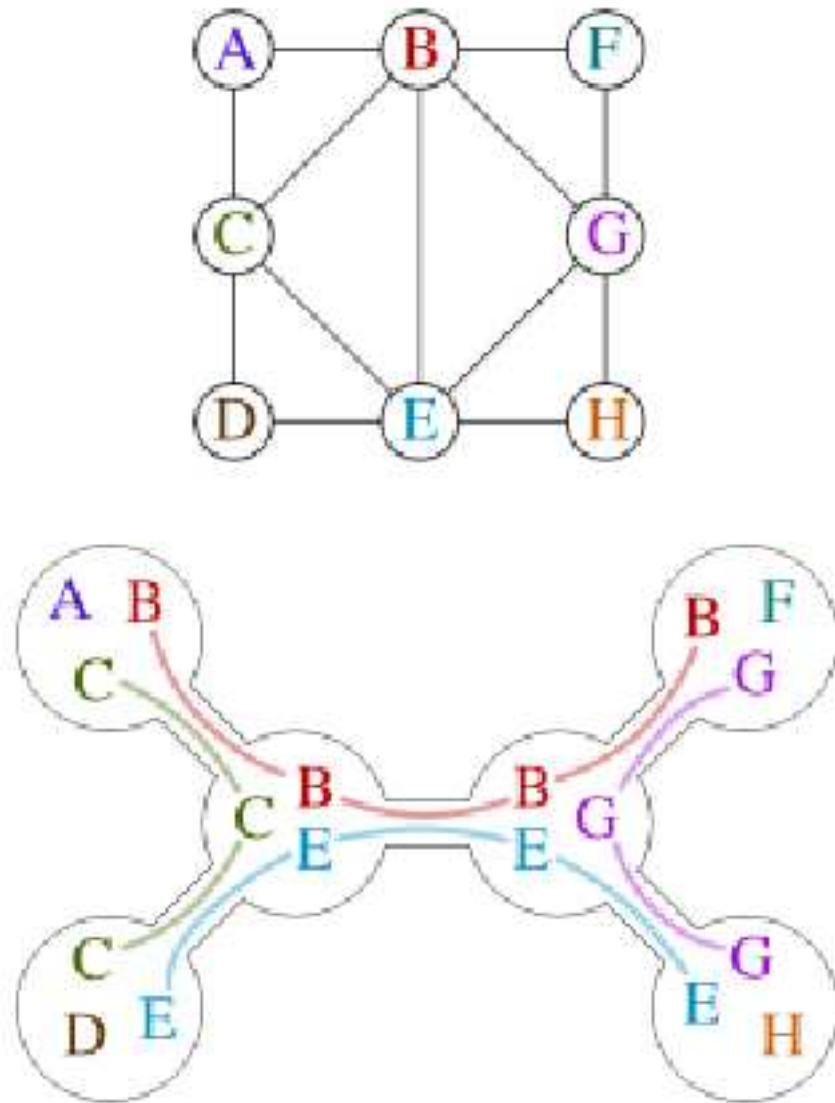


Figure 1: Treewidth 2

Bounded Treewidth and CSP

$$\mathcal{T}_k = \{\mathbf{A} : \text{tw}(\mathbf{A}) \leq k\}$$

Theorem: (Freuder, 1990)

$\text{CSP}(\mathcal{T}_k, \text{All})$ is in PTIME.

Note:

- Complexity is exponential in k .
- Determining treewidth of \mathbf{B} is NP-hard.
- Checking if treewidth is k is in linear time.

Complexity of Query Evaluation

Expression Complexity: Fix \mathbf{B}

$\{Q : Q(\mathbf{B}) \text{ is nonempty}\}$

Data Complexity: Fix Q

$\{\mathbf{B} : Q(\mathbf{B}) \text{ is nonempty}\}$

Exponential Gap: (V., 1982)

- Data complexity of FO: *LOGSPACE*
- Expression complexity of FO: *PSPACE-complete*

Mystery: practical query evaluation

Variable-Confined Queries

Definition: FO^k is first-order logic with at most k variables.

In Practice: (V., 1995)

- Queries often can be rewritten to use a small number of variables.
- Variable-confined queries have lower expression complexity.
- E.g.: expression complexity of FO^k is *PTIME-complete*

CSP and Database Queries

Theorem: Chandra&Merlin, 1977

Given \mathbf{A} , we can construct in polynomial time an existential, positive, conjunctive first-order query $Q_{\mathbf{A}}$ such that $h : \mathbf{A} \rightarrow \mathbf{B}$ iff $Q_{\mathbf{A}}(\mathbf{B})$ is nonempty.

Definition: The *core* of a structure is its (unique) minimal homomorphic substructure. Let \mathcal{C}_k consists of structures with cores of treewidth at most k .

Lemma: Chandra&Merlin, 1977

$Q_{\mathbf{A}}$ is logically equivalent to $Q_{\text{core}(\mathbf{A})}$

Theorem: [Kolaitis&V., 1998]

$\text{core}(\mathbf{A})$ has treewidth k iff $Q_{\mathbf{A}}$ is expressible in existential, positive FO with $k + 1$ variables.

Corollary [Dalmau&Kolaitis&V., 2002]

$\text{CSP}(\mathcal{C}_k, \text{All})$ is tractable; can be solved using k -Datalog.

Lower Bounds

Theorem: [Grohe, 2005]

Assume $FPT \neq W[1]$. Then $\text{CSP}((\mathcal{A}), \text{All})$ is tractable only if $\mathcal{A} \subseteq \mathcal{C}_k$.

Theorem: [Atserias&Bulatov&Dalmau, 2007]

$\text{CSP}((\mathcal{A}), \text{All})$ is solvable by k -Datalog only if $\mathcal{A} \subseteq \mathcal{C}_k$.

In Conclusion

CSP: a paradigmatic problem with connection to

- Graph theory,
- Algebra, and
- Logic,

with several outstanding open questions of theoretical and practical importance.